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Introduction

0.1 About this book

This first volume is a one semester course in basic analysis. Together with the second volume it is a year-long course. It started its life as my lecture notes for teaching Math 444 at the University of Illinois at Urbana-Champaign (UIUC) in Fall semester 2009. Later I added the metric space chapter to teach Math 521 at University of Wisconsin–Madison (UW). Volume II was added to teach Math 4143/4153 at Oklahoma State University (OSU). A prerequisite for these courses is usually a basic proof course, using for example [H], [F], or [DW].

It should be possible to use the book for both a basic course for students who do not necessarily wish to go to graduate school (such as UIUC 444), but also as a more advanced one-semester course that also covers topics such as metric spaces (such as UW 521). Here are my suggestions for what to cover in a semester course. For a slower course such as UIUC 444:

§0.3, §1.1–§1.4, §2.1–§2.5, §3.1–§3.4, §4.1–§4.2, §5.1–§5.3, §6.1–§6.3

For a more rigorous course covering metric spaces that runs quite a bit faster (such as UW 521):

§0.3, §1.1–§1.4, §2.1–§2.5, §3.1–§3.4, §4.1–§4.2, §5.1–§5.3, §6.1–§6.2, §7.1–§7.6

It should also be possible to run a faster course without metric spaces covering all sections of chapters 0 through 6. The approximate number of lectures given in the section notes through chapter 6 are a very rough estimate and were designed for the slower course. The first few chapters of the book can be used in an introductory proofs course as is done, for example, at Iowa State University Math 201, where this book is used in conjunction with Hammack’s Book of Proof [H].

With volume II one can run a year-long course that also covers multivariable topics. It may make sense in this case to cover most of the first volume in the first semester while leaving metric spaces for the beginning of the second semester.

The book normally used for the class at UIUC is Bartle and Sherbert, *Introduction to Real Analysis* third edition [BS]. The structure of the beginning of the book somewhat follows the standard syllabus of UIUC Math 444 and therefore has some similarities with [BS]. A major difference is that we define the Riemann integral using Darboux sums and not tagged partitions. The Darboux approach is far more appropriate for a course of this level.

Our approach allows us to fit a course such as UIUC 444 within a semester and still spend some time on the interchange of limits and end with Picard’s theorem on the existence and uniqueness of solutions of ordinary differential equations. This theorem is a wonderful example that uses many results proved in the book. For more advanced students, material may be covered faster so that we arrive at metric spaces and prove Picard’s theorem using the fixed point theorem as is usual.
Other excellent books exist. My favorite is Rudin’s excellent *Principles of Mathematical Analysis* [R2] or, as it is commonly and lovingly called, *baby Rudin* (to distinguish it from his other great analysis textbook, *big Rudin*). I took a lot of inspiration and ideas from Rudin. However, Rudin is a bit more advanced and ambitious than this present course. For those that wish to continue mathematics, Rudin is a fine investment. An inexpensive and somewhat simpler alternative to Rudin is Rosenlicht’s *Introduction to Analysis* [R1]. There is also the freely downloadable *Introduction to Real Analysis* by William Trench [T].

A note about the style of some of the proofs: Many proofs traditionally done by contradiction, I prefer to do by a direct proof or by contrapositive. While the book does include proofs by contradiction, I only do so when the contrapositive statement seemed too awkward, or when contradiction follows rather quickly. In my opinion, contradiction is more likely to get beginning students into trouble, as we are talking about objects that do not exist.

I try to avoid unnecessary formalism where it is unhelpful. Furthermore, the proofs and the language get slightly less formal as we progress through the book, as more and more details are left out to avoid clutter.

As a general rule, I use := instead of = to define an object rather than to simply show equality. I use this symbol rather more liberally than is usual for emphasis. I use it even when the context is “local,” that is, I may simply define a function $f(x) := x^2$ for a single exercise or example.

Finally, I would like to acknowledge Jana Maříková, Glen Pugh, Paul Vojta, Frank Beatrous, Sönmez Şahutoğlu, Jim Brandt, Kenji Kozai, Arthur Busch, Anton Petrunin, Mark Meilstrup, Harold P. Boas, Atilla Yılmaz, Thomas Mahoney, Scott Armstrong, and Paul Sacks, Matthias Weber, Manuele Santoprete, Robert Niemeyer, Amanullah Nabavi, for teaching with the book and giving me lots of useful feedback. Frank Beatrous wrote the University of Pittsburgh version extensions, which served as inspiration for many more recent additions. I would also like to thank Dan Stoneham, Jeremy Sutter, Eliya Gwetta, Daniel Pimentel-Alarcón, Steve Hoerning, Yi Zhang, Nicole Caviris, Kristopher Lee, Baoyue Bi, Hannah Lund, Trevor Mannella, Mitchel Meyer, Gregory Beauregard, Chase Meadors, Andreas Giannopoulos, Nick Nelsen, Ru Wang, Trevor Fancher, Brandon Tague, Wang KP, an anonymous reader or two, and in general all the students in my classes for suggestions and finding errors and typos.
0.2 About analysis

Analysis is the branch of mathematics that deals with inequalities and limits. The present course deals with the most basic concepts in analysis. The goal of the course is to acquaint the reader with rigorous proofs in analysis and also to set a firm foundation for calculus of one variable (and several variables if volume II is also considered).

Calculus has prepared you, the student, for using mathematics without telling you why what you learned is true. To use, or teach, mathematics effectively, you cannot simply know what is true, you must know why it is true. This course shows you why calculus is true. It is here to give you a good understanding of the concept of a limit, the derivative, and the integral.

Let us use an analogy. An auto mechanic that has learned to change the oil, fix broken headlights, and charge the battery, will only be able to do those simple tasks. He will be unable to work independently to diagnose and fix problems. A high school teacher that does not understand the definition of the Riemann integral or the derivative may not be able to properly answer all the students’ questions. To this day I remember several nonsensical statements I heard from my calculus teacher in high school, who simply did not understand the concept of the limit, though he could “do” the problems in the textbook.

We start with a discussion of the real number system, most importantly its completeness property, which is the basis for all that comes after. We then discuss the simplest form of a limit, the limit of a sequence. Afterwards, we study functions of one variable, continuity, and the derivative. Next, we define the Riemann integral and prove the fundamental theorem of calculus. We discuss sequences of functions and the interchange of limits. Finally, we give an introduction to metric spaces.

Let us give the most important difference between analysis and algebra. In algebra, we prove equalities directly; we prove that an object, a number perhaps, is equal to another object. In analysis, we usually prove inequalities, and we prove those inequalities by estimating. To illustrate the point, consider the following statement.

Let $x$ be a real number. If $x < \varepsilon$ is true for all real numbers $\varepsilon > 0$, then $x \leq 0$.

This statement is the general idea of what we do in analysis. Suppose next we really wish to prove the equality $x = 0$. In analysis, we prove two inequalities: $x \leq 0$ and $x \geq 0$. To prove the inequality $x \leq 0$, we prove $x < \varepsilon$ for all positive $\varepsilon$. To prove the inequality $x \geq 0$, we prove $x > -\varepsilon$ for all positive $\varepsilon$.

The term real analysis is a little bit of a misnomer. I prefer to use simply analysis. The other type of analysis, complex analysis, really builds up on the present material, rather than being distinct. Furthermore, a more advanced course on real analysis would talk about complex numbers often. I suspect the nomenclature is historical baggage.

Let us get on with the show...
0.3 Basic set theory

Note: 1–3 lectures (some material can be skipped, covered lightly, or left as reading)

Before we start talking about analysis, we need to fix some language. Modern* analysis uses the language of sets, and therefore that is where we start. We talk about sets in a rather informal way, using the so-called “naive set theory.” Do not worry, that is what majority of mathematicians use, and it is hard to get into trouble. The reader has hopefully seen the very basics of set theory and proof writing before, and this section should be a quick refresher.

0.3.1 Sets

Definition 0.3.1. A set is a collection of objects called elements or members. A set with no objects is called the empty set and is denoted by ∅ (or sometimes by {}).

Think of a set as a club with a certain membership. For example, the students who play chess are members of the chess club. The same student can be a member of many different clubs. However, do not take the analogy too far. A set is only defined by the members that form the set; two sets that have the same members are the same set.

Most of the time we will consider sets of numbers. For example, the set

\[ S := \{0, 1, 2\} \]

is the set containing the three elements 0, 1, and 2. By “:=”, we mean we are defining what S is, rather than just showing equality. We write

\[ 1 \in S \]

to denote that the number 1 belongs to the set S. That is, 1 is a member of S. At times we want to say that two elements are in a set S, so we write “1, 2 ∈ S” as a shorthand for “1 ∈ S and 2 ∈ S.”

Similarly, we write

\[ 7 \notin S \]

to denote that the number 7 is not in S. That is, 7 is not a member of S.

The elements of all sets under consideration come from some set we call the universe. For simplicity, we often consider the universe to be the set that contains only the elements we are interested in. The universe is generally understood from context and is not explicitly mentioned. In this course, our universe will most often be the set of real numbers.

While the elements of a set are often numbers, other objects, such as other sets, can be elements of a set. A set may also contain some of the same elements as another set. For example,

\[ T := \{0, 2\} \]

contains the numbers 0 and 2. In this case all elements of T also belong to S. We write \( T \subset S \). See Figure 1 for a diagram.

*The term “modern” refers to late 19th century up to the present.
Definition 0.3.2.

(i) A set \( A \) is a subset of a set \( B \) if \( x \in A \) implies \( x \in B \), and we write \( A \subset B \). That is, all members of \( A \) are also members of \( B \). At times we write \( B \supset A \) to mean the same thing.

(ii) Two sets \( A \) and \( B \) are equal if \( A \subset B \) and \( B \subset A \). We write \( A = B \). That is, \( A \) and \( B \) contain exactly the same elements. If it is not true that \( A \) and \( B \) are equal, then we write \( A \neq B \).

(iii) A set \( A \) is a proper subset of \( B \) if \( A \subset B \) and \( A \neq B \). We write \( A \varsubsetneq B \).

For the example \( S \) and \( T \) defined above, \( T \subset S \), but \( T \neq S \). So \( T \) is a proper subset of \( S \). If \( A = B \), then \( A \) and \( B \) are simply two names for the same exact set.

To define sets, one often uses the set building notation,

\[ \{ x \in A : P(x) \} \]

This notation refers to a subset of the set \( A \) containing all elements of \( A \) that satisfy the property \( P(x) \). Using \( S = \{0,1,2\} \) as above, \( \{x \in S : x \neq 2\} \) is the set \( \{0,1\} \). The notation is sometimes abbreviated as \( \{x : P(x)\} \), that is, \( A \) is not mentioned when understood from context. Furthermore, \( x \in A \) is sometimes replaced with a formula to make the notation easier to read.

Example 0.3.3: The following are sets including the standard notations.

(i) The set of natural numbers, \( \mathbb{N} := \{1,2,3,\ldots\} \).

(ii) The set of integers, \( \mathbb{Z} := \{0,-1,1,-2,2,\ldots\} \).

(iii) The set of rational numbers, \( \mathbb{Q} := \{ \frac{m}{n} : m,n \in \mathbb{Z} \text{ and } n \neq 0 \} \).

(iv) The set of even natural numbers, \( \{2m : m \in \mathbb{N}\} \).

(v) The set of real numbers, \( \mathbb{R} \).

Note that \( \mathbb{N} \subset \mathbb{Z} \subset \mathbb{Q} \subset \mathbb{R} \).

We create new sets out of old ones by applying some natural operations.

Definition 0.3.4.

(i) A union of two sets \( A \) and \( B \) is defined as

\[ A \cup B := \{ x : x \in A \text{ or } x \in B \} \]

(ii) An intersection of two sets \( A \) and \( B \) is defined as

\[ A \cap B := \{ x : x \in A \text{ and } x \in B \} \]
(iii) A complement of $B$ relative to $A$ (or set-theoretic difference of $A$ and $B$) is defined as

$$A \setminus B := \{x : x \in A \text{ and } x \notin B\}.$$

(iv) We say complement of $B$ and write $B^c$ instead of $A \setminus B$ if the set $A$ is either the entire universe or if it is the obvious set containing $B$, and is understood from context.

(v) We say sets $A$ and $B$ are disjoint if $A \cap B = \emptyset$.

The notation $B^c$ may be a little vague at this point. If the set $B$ is a subset of the real numbers $\mathbb{R}$, then $B^c$ means $\mathbb{R} \setminus B$. If $B$ is naturally a subset of the natural numbers, then $B^c$ is $\mathbb{N} \setminus B$. If ambiguity can arise, we use the set difference notation $A \setminus B$.

\begin{figure}
\centering
\includegraphics[width=\textwidth]{venn_diagrams.png}
\caption{Venn diagrams of set operations, the result of the operation is shaded.}
\end{figure}

We illustrate the operations on the \textit{Venn diagrams} in \textbf{Figure 2}. Let us now establish one of most basic theorems about sets and logic.

\textbf{Theorem 0.3.5 (DeMorgan).} Let $A, B, C$ be sets. Then

$$(B \cup C)^c = B^c \cap C^c,$$

$$(B \cap C)^c = B^c \cup C^c,$$

or, more generally,

$$A \setminus (B \cup C) = (A \setminus B) \cap (A \setminus C),$$

$$A \setminus (B \cap C) = (A \setminus B) \cup (A \setminus C).$$
0.3. BASIC SET THEORY

Proof. The first statement is proved by the second statement if we assume the set \( A \) is our “universe.”

Let us prove \( A \setminus (B \cup C) = (A \setminus B) \cap (A \setminus C) \). Remember the definition of equality of sets. First, we must show that if \( x \in A \setminus (B \cup C) \), then \( x \in (A \setminus B) \cap (A \setminus C) \). Second, we must also show that if \( x \in (A \setminus B) \cap (A \setminus C) \), then \( x \in A \setminus (B \cup C) \).

So let us assume \( x \in A \setminus (B \cup C) \). Then \( x \) is in \( A \), but not in \( B \) nor \( C \). Hence \( x \) is in \( A \) and not in \( B \), that is, \( x \in A \setminus B \). Similarly \( x \in A \setminus C \). Thus \( x \in (A \setminus B) \cap (A \setminus C) \).

On the other hand suppose \( x \in (A \setminus B) \cap (A \setminus C) \). In particular, \( x \in (A \setminus B) \), so \( x \in A \) and \( x \notin B \). Also as \( x \in (A \setminus C) \), then \( x \notin C \). Hence \( x \in A \setminus (B \cup C) \).

The proof of the other equality is left as an exercise. \( \square \)

The result above we called a Theorem, while most results we call a Proposition, and a few we call a Lemma (a result leading to another result) or Corollary (a quick consequence of the preceding result). Do not read too much into the naming. Some of it is traditional, some of it is stylistic choice.

It is not necessarily true that a Theorem is always “more important” than a Proposition or a Lemma.

We will also need to intersect or union several sets at once. If there are only finitely many, then we simply apply the union or intersection operation several times. However, suppose we have an infinite collection of sets (a set of sets) \( \{A_1, A_2, A_3, \ldots \} \). We define

\[
\bigcup_{n=1}^{\infty} A_n := \{ x : x \in A_n \text{ for some } n \in \mathbb{N} \},
\]

\[
\bigcap_{n=1}^{\infty} A_n := \{ x : x \in A_n \text{ for all } n \in \mathbb{N} \}.
\]

We can also have sets indexed by two natural numbers. For example, we can have the set of sets \( \{A_{1,1}, A_{1,2}, A_{2,1}, A_{2,2}, A_{3,1}, \ldots \} \). Then we write

\[
\bigcup_{n=1}^{\infty} \bigcup_{m=1}^{\infty} A_{n,m} = \bigcup_{n=1}^{\infty} \left( \bigcup_{m=1}^{\infty} A_{n,m} \right).
\]

And similarly with intersections.

It is not hard to see that we can take the unions in any order. However, switching the order of unions and intersections is not generally permitted without proof. For instance,

\[
\bigcup_{n=1}^{\infty} \bigcap_{m=1}^{\infty} \{ k \in \mathbb{N} : mk < n \} = \bigcap_{n=1}^{\infty} \emptyset = \emptyset.
\]

However,

\[
\bigcap_{m=1}^{\infty} \bigcup_{n=1}^{\infty} \{ k \in \mathbb{N} : mk < n \} = \bigcap_{m=1}^{\infty} \mathbb{N} = \mathbb{N}.
\]

Sometimes, the index set is not the natural numbers. In such a case we require a more general notation. Suppose \( I \) is some set and for each \( \lambda \in I \), there is a set \( A_{\lambda} \). Then we define

\[
\bigcup_{\lambda \in I} A_{\lambda} := \{ x : x \in A_{\lambda} \text{ for some } \lambda \in I \}, \quad \bigcap_{\lambda \in I} A_{\lambda} := \{ x : x \in A_{\lambda} \text{ for all } \lambda \in I \}.
\]
0.3.2 Induction

When a statement includes an arbitrary natural number, a common method of proof is the principle of induction. We start with the set of natural numbers \( \mathbb{N} = \{1, 2, 3, \ldots\} \), and we give them their natural ordering, that is, \( 1 < 2 < 3 < 4 < \cdots \). By \( S \subset \mathbb{N} \) having a least element, we mean that there exists an \( x \in S \), such that for every \( y \in S \), we have \( x \leq y \).

The natural numbers \( \mathbb{N} \) ordered in the natural way possess the so-called well ordering property. We take this property as an axiom; we simply assume it is true.

Well ordering property of \( \mathbb{N} \). Every nonempty subset of \( \mathbb{N} \) has a least (smallest) element.

The principle of induction is the following theorem, which is in a sense* equivalent to the well ordering property of the natural numbers.

**Theorem 0.3.6** (Principle of induction). Let \( P(n) \) be a statement depending on a natural number \( n \). Suppose that

(i) **(basis statement)** \( P(1) \) is true.

(ii) **(induction step)** If \( P(n) \) is true, then \( P(n + 1) \) is true.

Then \( P(n) \) is true for all \( n \in \mathbb{N} \).

**Proof.** Let \( S \) be the set of natural numbers \( m \) for which \( P(m) \) is not true. Suppose for contradiction that \( S \) is nonempty. Then \( S \) has a least element by the well ordering property. Call \( m \in S \) the least element of \( S \). We know \( 1 \notin S \) by hypothesis. So \( m > 1 \), and \( m - 1 \) is a natural number as well. Since \( m \) is the least element of \( S \), we know that \( P(m - 1) \) is true. But the induction step says that \( P(m - 1 + 1) = P(m) \) is true, contradicting the statement that \( m \in S \). Therefore, \( S \) is empty and \( P(n) \) is true for all \( n \in \mathbb{N} \). \( \square \)

Sometimes it is convenient to start at a different number than 1, all that changes is the labeling. The assumption that \( P(n) \) is true in “if \( P(n) \) is true, then \( P(n + 1) \) is true” is usually called the induction hypothesis.

**Example 0.3.7:** Let us prove that for all \( n \in \mathbb{N} \),

\[
2^{n-1} \leq n! \quad \text{(recall } n! = 1 \cdot 2 \cdot 3 \cdots n). \]

We let \( P(n) \) be the statement that \( 2^{n-1} \leq n! \) is true. By plugging in \( n = 1 \), we see that \( P(1) \) is true.

Suppose \( P(n) \) is true. That is, suppose \( 2^{n-1} \leq n! \) holds. Multiply both sides by 2 to obtain

\[
2^n \leq 2(n!).
\]

As \( 2 \leq (n + 1) \) when \( n \in \mathbb{N} \), we have \( 2(n!) \leq (n + 1)(n!) = (n + 1)! \). That is,

\[
2^n \leq 2(n!) \leq (n + 1)!,
\]

and hence \( P(n + 1) \) is true. By the principle of induction, \( P(n) \) is true for all \( n \in \mathbb{N} \). In other words, \( 2^{n-1} \leq n! \) is true for all \( n \in \mathbb{N} \).

*To be completely rigorous, this equivalence is only true if we also assume as an axiom that \( n - 1 \) exists for all natural numbers bigger than 1, which we do. In this book, we are assuming all the usual arithmetic holds.
Example 0.3.8: We claim that for all \( c \neq 1 \),
\[
1 + c + c^2 + \cdots + c^n = \frac{1 - c^{n+1}}{1 - c}.
\]

Proof: It is easy to check that the equation holds with \( n = 1 \). Suppose it is true for \( n \). Then
\[
1 + c + c^2 + \cdots + c^n + c^{n+1} = (1 + c + c^2 + \cdots + c^n) + c^{n+1}
\]
\[
= \frac{1 - c^{n+1}}{1 - c} + c^{n+1}
\]
\[
= \frac{1 - c^{n+1} + (1 - c)c^{n+1}}{1 - c}
\]
\[
= \frac{1 - c^{n+2}}{1 - c}.
\]

Sometimes, it is easier to use in the inductive step that \( P(k) \) is true for all \( k = 1, 2, \ldots, n \), not just for \( k = n \). This principle is called strong induction and is equivalent to the normal induction above. The proof of that equivalence is left as an exercise.

Theorem 0.3.9 (Principle of strong induction). Let \( P(n) \) be a statement depending on a natural number \( n \). Suppose that

(i) (basis statement) \( P(1) \) is true.

(ii) (induction step) If \( P(k) \) is true for all \( k = 1, 2, \ldots, n \), then \( P(n+1) \) is true.

Then \( P(n) \) is true for all \( n \in \mathbb{N} \).

0.3.3 Functions

Informally, a set-theoretic function \( f \) taking a set \( A \) to a set \( B \) is a mapping that to each \( x \in A \) assigns a unique \( y \in B \). We write \( f : A \rightarrow B \). An example function \( f : S \rightarrow T \) taking \( S := \{0, 1, 2\} \) to \( T := \{0, 2\} \) can be defined by assigning \( f(0) := 2 \), \( f(1) := 2 \), and \( f(2) := 0 \). That is, a function \( f : A \rightarrow B \) is a black box, into which we stick an element of \( A \) and the function spits out an element of \( B \). Sometimes \( f \) is called a mapping or a map, and we say \( f \) maps \( A \) to \( B \).

Often, functions are defined by some sort of formula; however, you should really think of a function as just a very big table of values. The subtle issue here is that a single function can have several formulas, all giving the same function. Also, for many functions, there is no formula that expresses its values.

To define a function rigorously, first let us define the Cartesian product.

Definition 0.3.10. Let \( A \) and \( B \) be sets. The Cartesian product is the set of tuples defined as
\[
A \times B := \{(x, y) : x \in A, y \in B\}.
\]

For instance, \( \{a, b\} \times \{c, d\} = \{(a, c), (a, d), (b, c), (b, d)\} \). A more complicated example is the set \( [0, 1] \times [0, 1] \): a subset of the plane bounded by a square with vertices \((0, 0), (0, 1), (1, 0), \) and \((1, 1)\). When \( A \) and \( B \) are the same set we sometimes use a superscript 2 to denote such a product. For example, \([0, 1]^2 = [0, 1] \times [0, 1]\) or \( \mathbb{R}^2 = \mathbb{R} \times \mathbb{R} \) (the Cartesian plane).
Definition 0.3.11. A function $f: A \to B$ is a subset $f$ of $A \times B$ such that for each $x \in A$, there exists a unique $y \in B$ for which $(x,y) \in f$. We write $f(x) = y$. Sometimes the set $f$ is called the graph of the function rather than the function itself.

The set $A$ is called the domain of $f$ (and sometimes confusingly denoted $D(f)$). The set

$$R(f) := \{ y \in B : \text{there exists an } x \in A \text{ such that } f(x) = y \}$$

is called the range of $f$. The set $B$ is called the codomain of $f$.

It is possible that the range $R(f)$ is a proper subset of the codomain $B$, while the domain of $f$ is always equal to $A$. We generally assume that the domain of $f$ is nonempty.

Example 0.3.12: From calculus, you are most familiar with functions taking real numbers to real numbers. However, you saw some other types of functions as well. The derivative is a function mapping the set of differentiable functions to the set of all functions. Another example is the Laplace transform, which also takes functions to functions. Yet another example is the function that takes a continuous function $g$ defined on the interval $[0, 1]$ and returns the number $\int_0^1 g(x) \, dx$.

Definition 0.3.13. Consider a function $f: A \to B$. Define the image (or direct image) of a subset $C \subset A$ as

$$f(C) := \{ f(x) \in B : x \in C \}.$$ 

Define the inverse image of a subset $D \subset B$ as

$$f^{-1}(D) := \{ x \in A : f(x) \in D \}.$$ 

In particular, $R(f) = f(A)$, the range is the direct image of the domain $A$.

---

Example 0.3.14: Define the function $f: \mathbb{R} \to \mathbb{R}$ by $f(x) := \sin(\pi x)$. Then $f([0, 1/2]) = [0, 1]$, $f^{-1}(\{0\}) = \mathbb{Z}$, etc.

Proposition 0.3.15. Consider $f: A \to B$. Let $C, D$ be subsets of $B$. Then

$$f^{-1}(C \cup D) = f^{-1}(C) \cup f^{-1}(D),$$

$$f^{-1}(C \cap D) = f^{-1}(C) \cap f^{-1}(D),$$

$$f^{-1}(C^c) = (f^{-1}(C))^c.$$
Read the last line of the proposition as \( f^{-1}(B \setminus C) = A \setminus f^{-1}(C) \).

**Proof.** We start with the union. If \( x \in f^{-1}(C \cup D) \), then \( x \) is taken to \( C \) or \( D \), that is, \( f(x) \in C \) or \( f(x) \in D \). Thus \( f^{-1}(C \cup D) \subset f^{-1}(C) \cup f^{-1}(D) \). Conversely if \( x \in f^{-1}(C) \), then \( x \in f^{-1}(C \cup D) \).

Similarly for \( x \in f^{-1}(D) \). Hence \( f^{-1}(C \cup D) \supset f^{-1}(C) \cup f^{-1}(D) \), and we have equality.

The rest of the proof is left as an exercise. \( \square \)

The proposition does not hold for direct images. We do have the following weaker result.

**Proposition 0.3.16.** Consider \( f : A \to B \). Let \( C, D \) be subsets of \( A \). Then

\[
\begin{align*}
    f(C \cup D) &= f(C) \cup f(D), \\
    f(C \cap D) &\subset f(C) \cap f(D).
\end{align*}
\]

The proof is left as an exercise.

**Definition 0.3.17.** Let \( f : A \to B \) be a function. The function \( f \) is said to be *injective* or *one-to-one* if \( f(x_1) = f(x_2) \) implies \( x_1 = x_2 \). In other words, \( f \) is injective if for all \( y \in B \), the set \( f^{-1}\{y\} \) is empty or consists of a single element. We call such an \( f \) an *injection*.

If \( f(A) = B \), then we say \( f \) is *surjective* or *onto*. In other words, \( f \) is surjective if the range and the codomain of \( f \) are equal. We call such an \( f \) a *surjection*.

If \( f \) is both an surjective and injective, then we say \( f \) is *bijective* or that \( f \) is a *bijection*.

When \( f : A \to B \) is a bijection, then the inverse image of a single element, \( f^{-1}\{y\} \), is always a unique element of \( A \). We then consider \( f^{-1} \) as a function \( f^{-1} : B \to A \) and we write simply \( f^{-1}(y) \).

In this case, we call \( f^{-1} \) the *inverse function* of \( f \). For instance, for the bijection \( f : \mathbb{R} \to \mathbb{R} \) defined by \( f(x) := x^3 \), we have \( f^{-1}(x) = \sqrt[3]{x} \).

**Definition 0.3.18.** Consider \( f : A \to B \) and \( g : B \to C \). The *composition* of the functions \( f \) and \( g \) is the function \( g \circ f : A \to C \) defined as

\[
(g \circ f)(x) := g(f(x)).
\]

For example, if \( f : \mathbb{R} \to \mathbb{R} \) is \( f(x) := x^3 \) and \( g : \mathbb{R} \to \mathbb{R} \) is \( g(y) = \sin(y) \), then \( (g \circ f)(x) = \sin(x^3) \).

### 0.3.4 Relations and equivalence classes

We often compare two objects in some way. We say \( 1 < 2 \) for natural numbers, or \( 1/2 = 2/4 \) for rational numbers, or \( \{a, c\} \subset \{a, b, c\} \) for sets. The ‘\(<\)’, ‘\(=\)’, and ‘\(<\)’ are examples of relations.

**Definition 0.3.19.** Given a set \( A \), a *binary relation* on \( A \) is a subset \( \mathcal{R} \subset A \times A \), which are those pairs where the relation is said to hold. Instead of \( (a, b) \in \mathcal{R} \), we write \( a \mathcal{R} b \).

**Example 0.3.20:** Take \( A := \{1, 2, 3\} \).

Consider the relation ‘<’. The corresponding set of pairs is \( \{(1, 2), (1, 3), (2, 3)\} \). So \( 1 < 2 \) holds as \( (1, 2) \) is in the corresponding set of pairs, but \( 3 < 1 \) does not hold as \( (3, 1) \) is not in the set.

Similarly, the relation ‘\(=\)’ is defined by the set of pairs \( \{(1, 1), (2, 2), (3, 3)\} \).

Any subset of \( A \times A \) is a relation. Let us define the relation \( \dagger \) via \( \{(1, 2), (2, 1), (2, 3), (3, 1)\} \), then \( 1 \dagger 2 \) and \( 3 \dagger 1 \) are true, but \( 1 \dagger 3 \) is not.
Definition 0.3.21. Let $R$ be a relation on a set $A$. Then $R$ is said to be

(i) Reflexive if $a R a$ for all $a \in A$.

(ii) Symmetric if $a R b$ implies $b R a$.

(iii) Transitive if $a R b$ and $b R c$ implies $a R c$.

If $R$ is reflexive, symmetric, and transitive, then it is said to be an equivalence relation.

Example 0.3.22: Let $A := \{1, 2, 3\}$ as above. The relation ‘$<$’ is transitive, but neither reflexive nor symmetric. The relation ‘$\leq$’ defined by $\{(1, 1), (1, 2), (1, 3), (2, 2), (2, 3), (3, 3)\}$ is reflexive and transitive, but not symmetric. Finally, a relation ‘$\star$’ defined by $\{(1, 1), (1, 2), (2, 1), (2, 2), (3, 3)\}$ is an equivalence relation.

Equivalence relations are useful in that they divide a set into sets of “equivalent” elements.

Definition 0.3.23. Let $A$ be a set and $R$ an equivalence relation. An equivalence class of $a \in A$, often denoted by $[a]$, is the set $\{x \in A : a R x\}$.

For example, given the relation ‘$\star$’ above, there are two equivalence classes, $[1] = [2] = \{1, 2\}$ and $[3] = \{3\}$.

Reflexivity guarantees that $a \in [a]$. Symmetry guarantees that if $b \in [a]$, then $a \in [b]$. Finally, transitivity guarantees that if $a \in [b]$ and $b \in [c]$, then $a \in [c]$. In particular, we have the following proposition, whose proof is an exercise.

Proposition 0.3.24. If $R$ is an equivalence relation on a set $A$, then every $a \in A$ is in exactly one equivalence class. In particular, $a R b$ if and only $[a] = [b]$.

Example 0.3.25: The set of rational numbers can be defined as equivalence classes of a pair of an integer and a natural number, that is elements of $\mathbb{Z} \times \mathbb{N}$. The relation is defined by $(a, b) \sim (c, d)$ whenever $ad = bc$. It is left as an exercise to prove that ‘$\sim$’ is an equivalence relation. Usually the equivalence class $[(a, b)]$ is written as $a/b$.

0.3.5 Cardinality

A subtle issue in set theory and one generating a considerable amount of confusion among students is that of cardinality, or “size” of sets. The concept of cardinality is important in modern mathematics in general and in analysis in particular. In this section, we will see the first really unexpected theorem.

Definition 0.3.26. Let $A$ and $B$ be sets. We say $A$ and $B$ have the same cardinality when there exists a bijection $f : A \to B$. We denote by $|A|$ the equivalence class of all sets with the same cardinality as $A$ and we simply call $|A|$ the cardinality of $A$.

For example, $\{1, 2, 3\}$ has the same cardinality as $\{a, b, c\}$ by defining a bijection $f(1) := a$, $f(2) := b$, $f(3) := c$. Clearly the bijection is not unique.

The existence of a bijection really is an equivalence relation. The identity, $f(x) := x$, is a bijection showing reflexivity. If $f$ is a bijection, then so is $f^{-1}$ showing symmetricity. If $f : A \to B$ and $g : B \to C$ are bijections, then $g \circ f$ is a bijection of $A$ and $C$ showing transitivity. A set $A$ has the same cardinality as the empty set if and only if $A$ itself is the empty set: If $B$ is nonempty, then no function $f : B \to \emptyset$ can exist. In particular, there is no bijection of $B$ and $\emptyset$. 
Definition 0.3.27. Suppose \( A \) has the same cardinality as \( \{1, 2, 3, \ldots, n\} \) for some \( n \in \mathbb{N} \). We then write \( |A| := n \). If \( A \) is empty, we write \( |A| := 0 \). In either case, we say that \( A \) is finite.

We say \( A \) is infinite or “of infinite cardinality” if \( A \) is not finite.

That the notation \( |A| = n \) is justified we leave as an exercise. That is, for each nonempty finite set \( A \), there exists a unique natural number \( n \) such that there exists a bijection from \( A \) to \( \{1, 2, 3, \ldots, n\} \).

We can order sets by size.

Definition 0.3.28. We write

\[
|A| \leq |B|
\]

if there exists an injection from \( A \) to \( B \). We write \( |A| = |B| \) if \( A \) and \( B \) have the same cardinality. We write \( |A| < |B| \) if \( |A| \leq |B| \), but \( A \) and \( B \) do not have the same cardinality.

We state without proof that \( A \) and \( B \) have the same cardinality if and only if \( |A| \leq |B| \) and \( |B| \leq |A| \). This is the so-called Cantor–Bernstein–Schröder theorem. Furthermore, if \( A \) and \( B \) are any two sets, we can always write \( |A| \leq |B| \) or \( |B| \leq |A| \). The issues surrounding this last statement are very subtle. As we do not require either of these two statements, we omit proofs.

The truly interesting cases of cardinality are infinite sets. We will distinguish two types of infinite cardinality.

Definition 0.3.29. If \( |A| = |\mathbb{N}| \), then \( A \) is said to be countably infinite. If \( A \) is finite or countably infinite, then we say \( A \) is countable. If \( A \) is not countable, then \( A \) is said to be uncountable.

The cardinality of \( \mathbb{N} \) is usually denoted as \( \aleph_0 \) (read as aleph-naught).

Example 0.3.30: The set of even natural numbers has the same cardinality as \( \mathbb{N} \). Proof: Let \( E \subset \mathbb{N} \) be the set of even natural numbers. Given \( k \in E \), write \( k = 2n \) for some \( n \in \mathbb{N} \). Then \( f(n) := 2n \) defines a bijection \( f : \mathbb{N} \to E \).

In fact, let us mention without proof the following characterization of infinite sets: A set is infinite if and only if it is in one-to-one correspondence with a proper subset of itself.

Example 0.3.31: \( \mathbb{N} \times \mathbb{N} \) is a countably infinite set. Proof: Arrange the elements of \( \mathbb{N} \times \mathbb{N} \) as follows \((1, 1), (1, 2), (2, 1), (1, 3), (2, 2), (3, 1), \ldots \). That is, always write down first all the elements whose two entries sum to \( k \), then write down all the elements whose entries sum to \( k + 1 \) and so on. Define a bijection with \( \mathbb{N} \) by letting \( 1 \) go to \((1, 1)\), \( 2 \) go to \((1, 2)\), and so on. See Figure 4.

Example 0.3.32: The set of rational numbers is countable. Proof: (informal) For positive rational numbers follow the same procedure as in the previous example, writing \( 1/1, 1/2, 2/1, \) etc. However, leave out fractions (such as \( 2/2 \)) that have already appeared. The list would continue: \( 1/3, 3/1, 1/4, 2/3, \) etc. For all rational numbers, include 0 and the negative numbers: \( 0, 1/1, -1/1, 1/2, -1/2, \) etc.

For completeness, we mention the following statements from the exercises. If \( A \subset B \) and \( B \) is countably infinite, then \( A \) is countable. The contrapositive of the statement is that if \( A \) is uncountable, then \( B \) is uncountable. As a consequence, if \( |A| < |\mathbb{N}| \), then \( A \) is finite. Similarly, if \( B \) is finite and \( A \subset B \), then \( A \) is finite.

*For the fans of the TV show *Futurama*, there is a movie theater in one episode called an \( \aleph_0 \)-plex.
We give the first truly striking result. First, we need a notation for the set of all subsets of a set.

**Definition 0.3.33.** The \( \text{power set} \) of a set \( A \), denoted by \( P(A) \), is the set of all subsets of \( A \).

For example, if \( A := \{1, 2\} \), then \( P(A) = \{\emptyset, \{1\}, \{2\}, \{1, 2\}\} \). In particular, \( |A| = 2 \) and \( |P(A)| = 4 = 2^2 \). In general, for a finite set \( A \) of cardinality \( n \), the cardinality of \( P(A) \) is \( 2^n \). This fact is left as an exercise. Hence, for a finite set \( A \), the cardinality of \( P(A) \) is strictly larger than the cardinality of \( A \). What is an unexpected and striking fact is that this statement is still true for infinite sets.

**Theorem 0.3.34 (Cantor\(^*\)).** \( |A| < |P(A)| \). In particular, there exists no surjection from \( A \) onto \( P(A) \).

**Proof.** There exists an injection \( f : A \to P(A) \). For \( x \in A \), define \( f(x) := \{x\} \). Thus, \( |A| \leq |P(A)| \).

To finish the proof, we must show that no function \( g : A \to P(A) \) is a surjection. Suppose \( g : A \to P(A) \) is a function. So for \( x \in A \), \( g(x) \) is a subset of \( A \). Define the set

\[
B := \{x \in A : x \notin g(x)\}.
\]

We claim that \( B \) is not in the range of \( g \) and hence \( g \) is not a surjection. Suppose for contradiction that there exists an \( x_0 \) such that \( g(x_0) = B \). Either \( x_0 \in B \) or \( x_0 \notin B \). If \( x_0 \in B \), then \( x_0 \notin g(x_0) = B \), which is a contradiction. If \( x_0 \notin B \), then \( x_0 \in g(x_0) = B \), which is again a contradiction. Thus such an \( x_0 \) does not exist. Therefore, \( B \) is not in the range of \( g \), and \( g \) is not a surjection. As \( g \) was an arbitrary function, no surjection exists.

One particular consequence of this theorem is that there do exist uncountable sets, as \( P(\mathbb{N}) \) must be uncountable. A related fact is that the set of real numbers (which we study in the next chapter) is uncountable. The existence of uncountable sets may seem unintuitive, and the theorem caused quite a controversy at the time it was announced. The theorem not only says that uncountable sets exist, but that there in fact exist progressively larger and larger infinite sets \( \mathbb{N}, P(\mathbb{N}), P(P(\mathbb{N})), P(P(P(\mathbb{N}))), \) etc.

\(^*\)Named after the German mathematician Georg Ferdinand Ludwig Philipp Cantor (1845–1918).
0.3. BASIC SET THEORY

0.3.6 Exercises

Exercise 0.3.1: Show \(A \setminus (B \cap C) = (A \setminus B) \cup (A \setminus C)\).

Exercise 0.3.2: Prove that the principle of strong induction is equivalent to the standard induction.

Exercise 0.3.3: Finish the proof of Proposition 0.3.15.

Exercise 0.3.4:

a) Prove Proposition 0.3.16.

b) Find an example for which equality of sets in \(f(C \cap D) \subset f(C) \cap f(D)\) fails. That is, find an \(f\), \(A\), \(B\), \(C\), and \(D\) such that \(f(C \cap D)\) is a proper subset of \(f(C) \cap f(D)\).

Exercise 0.3.5 (Tricky): Prove that if \(A\) is nonempty and finite, then there exists a unique \(n \in \mathbb{N}\) such that there exists a bijection between \(A\) and \(\{1, 2, 3, \ldots, n\}\). In other words, the notation \(|A| := n\) is justified. Hint: Show that if \(n > m\), then there is no injection from \(\{1, 2, 3, \ldots, n\}\) to \(\{1, 2, 3, \ldots, m\}\).

Exercise 0.3.6: Prove:

a) \(A \cap (B \cup C) = (A \cap B) \cup (A \cap C)\).

b) \(A \cup (B \cap C) = (A \cup B) \cap (A \cup C)\).

Exercise 0.3.7: Let \(A \Delta B\) denote the symmetric difference, that is, the set of all elements that belong to either \(A\) or \(B\), but not to both \(A\) and \(B\).

a) Draw a Venn diagram for \(A \Delta B\).

b) Show \(A \Delta B = (A \setminus B) \cup (B \setminus A)\).

c) Show \(A \Delta B = (A \cup B) \setminus (A \cap B)\).

Exercise 0.3.8: For each \(n \in \mathbb{N}\), let \(A_n := \{(n + 1)k : k \in \mathbb{N}\}\).

a) Find \(A_1 \cap A_2\).

b) Find \(\bigcup_{n=1}^{\infty} A_n\).

c) Find \(\bigcap_{n=1}^{\infty} A_n\).

Exercise 0.3.9: Determine \(\mathcal{P}(S)\) (the power set) for each of the following:

a) \(S = \emptyset\),

b) \(S = \{1\}\),

c) \(S = \{1, 2\}\),

d) \(S = \{1, 2, 3, 4\}\).

Exercise 0.3.10: Let \(f : A \rightarrow B\) and \(g : B \rightarrow C\) be functions.

a) Prove that if \(g \circ f\) is injective, then \(f\) is injective.

b) Prove that if \(g \circ f\) is surjective, then \(g\) is surjective.

c) Find an explicit example where \(g \circ f\) is bijective, but neither \(f\) nor \(g\) is bijective.

Exercise 0.3.11: Prove by induction that \(n < 2^n\) for all \(n \in \mathbb{N}\).

Exercise 0.3.12: Show that for a finite set \(A\) of cardinality \(n\), the cardinality of \(\mathcal{P}(A)\) is \(2^n\).
**Exercise 0.3.13:** Prove \(\frac{1}{1^2} + \frac{1}{2^3} + \cdots + \frac{1}{n(n+1)} = \frac{n}{n+1}\) for all \(n \in \mathbb{N}\).

**Exercise 0.3.14:** Prove \(1^3 + 2^3 + \cdots + n^3 = \left(\frac{n(n+1)}{2}\right)^2\) for all \(n \in \mathbb{N}\).

**Exercise 0.3.15:** Prove that \(n^3 + 5n\) is divisible by 6 for all \(n \in \mathbb{N}\).

**Exercise 0.3.16:** Find the smallest \(n \in \mathbb{N}\) such that \(2(n + 5)^2 < n^3\) and call it \(n_0\). Show that \(2(n + 5)^2 < n^3\) for all \(n \geq n_0\).

**Exercise 0.3.17:** Find all \(n \in \mathbb{N}\) such that \(n^2 < 2^n\).

**Exercise 0.3.18:** Prove the well ordering property of \(\mathbb{N}\) using the principle of induction.

**Exercise 0.3.19:** Give an example of a countably infinite collection of finite sets \(A_1, A_2, \ldots\), whose union is not a finite set.

**Exercise 0.3.20:** Give an example of a countably infinite collection of infinite sets \(A_1, A_2, \ldots\), with \(A_j \cap A_k\) being infinite for all \(j\) and \(k\), such that \(\bigcap_{j=1}^{\infty} A_j\) is nonempty and finite.

**Exercise 0.3.21:** Suppose \(A \subset B\) and \(B\) is finite. Prove that \(A\) is finite. That is, if \(A\) is nonempty, construct a bijection of \(A\) to \(\{1, 2, \ldots, n\}\).

**Exercise 0.3.22:** Prove Proposition 0.3.24. That is, prove that if \(\mathcal{R}\) is an equivalence relation on a set \(A\), then every \(a \in A\) is in exactly one equivalence class. Then prove that \(a \mathcal{R} b\) if and only if \([a] = [b]\).

**Exercise 0.3.23:** Prove that the relation ‘\(\sim\)’ in Example 0.3.25 is an equivalence relation.

**Exercise 0.3.24:**

a) Suppose \(A \subset B\) and \(B\) is countably infinite. By constructing a bijection, show that \(A\) is countable (that is, \(A\) is empty, finite, or countably infinite).

b) Use part a) to show that if \(|A| < |\mathbb{N}|\), then \(A\) is finite.

**Exercise 0.3.25** (Challenging): Suppose \(|\mathbb{N}| \leq |S|\), or in other words, \(S\) contains a countably infinite subset. Show that there exists a countably infinite subset \(A \subset S\) and a bijection between \(S \setminus A\) and \(S\).
Chapter 1

Real Numbers

1.1 Basic properties

Note: 1.5 lectures

The main object we work with in analysis is the set of real numbers. As this set is so fundamental, often much time is spent on formally constructing the set of real numbers. However, we take an easier approach here and just assume that a set with the correct properties exists. We start with the definitions of those properties.

Definition 1.1.1. An ordered set is a set $S$ together with a relation $<$ such that

(i) (trichotomy) For all $x, y \in S$, exactly one of $x < y$, $x = y$, or $y < x$ holds.

(ii) (transitivity) If $x, y, z \in S$ are such that $x < y$ and $y < z$, then $x < z$.

We write $x \leq y$ if $x < y$ or $x = y$. We define $>$ and $\geq$ in the obvious way.

The set of rational numbers $\mathbb{Q}$ is an ordered set by letting $x < y$ if and only if $y - x$ is a positive rational number, that is if $y - x = p/q$ where $p, q \in \mathbb{N}$. Similarly, $\mathbb{N}$ and $\mathbb{Z}$ are also ordered sets.

There are other ordered sets than sets of numbers. For example, the set of countries can be ordered by landmass, so India $>$ Lichtenstein. A typical ordered set that you have used since primary school is the dictionary. It is the ordered set of words where the order is the so-called lexicographic ordering. Such ordered sets often appear, for example, in computer science. In this book we will mostly be interested in ordered sets of numbers.

Definition 1.1.2. Let $E \subset S$, where $S$ is an ordered set.

(i) If there exists a $b \in S$ such that $x \leq b$ for all $x \in E$, then we say $E$ is bounded above and $b$ is an upper bound of $E$.

(ii) If there exists a $b \in S$ such that $x \geq b$ for all $x \in E$, then we say $E$ is bounded below and $b$ is a lower bound of $E$.

(iii) If there exists an upper bound $b_0$ of $E$ such that whenever $b$ is an upper bound for $E$ we have $b_0 \leq b$, then $b_0$ is called the least upper bound or the supremum of $E$. See Figure 1.1. We write $\sup E := b_0$. 
(iv) Similarly, if there exists a lower bound $b_0$ of $E$ such that whenever $b$ is a lower bound for $E$ we have $b_0 \geq b$, then $b_0$ is called the greatest lower bound or the infimum of $E$. We write $$\inf E := b_0.$$ When a set $E$ is both bounded above and bounded below, we say simply that $E$ is bounded.

The notation $\sup E$ and $\inf E$ is justified as the supremum (or infimum) is unique (if it exists): If $b$ and $b'$ are suprema of $E$, then $b \leq b'$ and $b' \leq b$, because both $b$ and $b'$ are the least upper bounds, so $b = b'$.

![Figure 1.1: A set $E$ bounded above and the least upper bound of $E$.](image)

A simple example: Let $S := \{a, b, c, d, e \}$ be ordered as $a < b < c < d < e$, and let $E := \{a, c\}$. Then $c$, $d$, and $e$ are upper bounds of $E$, and $c$ is the least upper bound or supremum of $E$.

A supremum or infimum for $E$ (even if it exists) need not be in $E$. The set $E := \{x \in \mathbb{Q} : x < 1\}$ has a least upper bound of 1, but 1 is not in the set $E$ itself. The set $G := \{x \in \mathbb{Q} : x \leq 1\}$ also has an upper bound of 1, and in this case $1 \in G$. The set $P := \{x \in \mathbb{Q} : x \geq 0\}$ has no upper bound (why?) and therefore it cannot have a least upper bound. The set $P$ does have a greatest lower bound: 0.

**Definition 1.1.3.** An ordered set $S$ has the least-upper-bound property if every nonempty subset $E \subset S$ that is bounded above has a least upper bound, that is sup $E$ exists in $S$.

The least-upper-bound property is sometimes called the completeness property or the Dedekind completeness property*. As we will note in the next section, the real numbers have this property.

**Example 1.1.4:** The set $\mathbb{Q}$ of rational numbers does not have the least-upper-bound property. The subset $\{x \in \mathbb{Q} : x^2 < 2\}$ does not have a supremum in $\mathbb{Q}$. We will see later (Example 1.2.3) that the supremum is $\sqrt{2}$, which is not rational\(^1\). Suppose $x \in \mathbb{Q}$ such that $x^2 = 2$. Write $x = m/n$ in lowest terms. So $(m/n)^2 = 2$ or $m^2 = 2n^2$. Hence, $m^2$ is divisible by 2, and so $m$ is divisible by 2. Write $m = 2k$ and so $(2k)^2 = 2n^2$. Divide by 2 and note that $2k^2 = n^2$, and hence $n$ is divisible by 2. But that is a contradiction as $m/n$ is in lowest terms.

That $\mathbb{Q}$ does not have the least-upper-bound property is one of the most important reasons why we work with $\mathbb{R}$ in analysis. The set $\mathbb{Q}$ is just fine for algebraists. But us analysts require the least-upper-bound property to do any work. We also require our real numbers to have many algebraic properties. In particular, we require that they are a field.

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*Named after the German mathematician Julius Wilhelm Richard Dedekind (1831–1916).

\(^1\)This is true for all other roots of 2, and interestingly, the fact that $\sqrt{2}$ is never rational for $k > 1$ implies no piano can ever be perfectly tuned in all keys. See for example: [https://youtu.be/1Hqm0dYKUx4](https://youtu.be/1Hqm0dYKUx4).
1.1. BASIC PROPERTIES

Definition 1.1.5. A set $F$ is called a field if it has two operations defined on it, addition $x+y$ and multiplication $xy$, and if it satisfies the following axioms:

(A1) If $x \in F$ and $y \in F$, then $x + y \in F$.

(A2) (commutativity of addition) $x + y = y + x$ for all $x, y \in F$.

(A3) (associativity of addition) $(x + y) + z = x + (y + z)$ for all $x, y, z \in F$.

(A4) There exists an element $0 \in F$ such that $0 + x = x$ for all $x \in F$.

(A5) For every element $x \in F$, there exists an element $-x \in F$ such that $x + (-x) = 0$.

(M1) If $x \in F$ and $y \in F$, then $xy \in F$.

(M2) (commutativity of multiplication) $xy = yx$ for all $x, y \in F$.

(M3) (associativity of multiplication) $(xy)z = x(yz)$ for all $x, y, z \in F$.

(M4) There exists an element $1 \in F$ (and $1 \neq 0$) such that $1x = x$ for all $x \in F$.

(M5) For every $x \in F$ such that $x \neq 0$ there exists an element $\frac{1}{x} \in F$ such that $x\left(\frac{1}{x}\right) = 1$.

(D) (distributive law) $x(y + z) = xy + xz$ for all $x, y, z \in F$.

Example 1.1.6: The set $\mathbb{Q}$ of rational numbers is a field. On the other hand $\mathbb{Z}$ is not a field, as it does not contain multiplicative inverses. For example, there is no $x \in \mathbb{Z}$ such that $2x = 1$, so (M5) is not satisfied. You can check that (M5) is the only property that fails*.

We will assume the basic facts about fields that are easily proved from the axioms. For example, $0x = 0$ is easily proved by noting that $xx = (0 + x)x = 0x + xx$, using (A4), (D), and (M2). Then using (A5) on $xx$, along with (A2), (A3), and (A4), we obtain $0 = 0x$.

Definition 1.1.7. A field $F$ is said to be an ordered field if $F$ is also an ordered set such that

(i) For $x, y, z \in F$, $x < y$ implies $x + z < y + z$.

(ii) For $x, y \in F$, $x > 0$ and $y > 0$ implies $xy > 0$.

If $x > 0$, we say $x$ is positive. If $x < 0$, we say $x$ is negative. We also say $x$ is nonnegative if $x \geq 0$, and $x$ is nonpositive if $x \leq 0$.

It can be checked that the rational numbers $\mathbb{Q}$ with the standard ordering is an ordered field.

Proposition 1.1.8. Let $F$ be an ordered field and $x, y, z, w \in F$. Then

(i) If $x > 0$, then $-x < 0$ (and vice versa).

(ii) If $x > 0$ and $y < z$, then $xy < xz$.

(iii) If $x < 0$ and $y < z$, then $xy > xz$.

(iv) If $x \neq 0$, then $x^2 > 0$.

(v) If $0 < x < y$, then $0 < \frac{1}{y} < \frac{1}{x}$.

(vi) If $0 < x < y$, then $x^2 < y^2$.

(vii) If $x \leq y$ and $z \leq w$, then $x + z \leq y + w$.

*An algebraist would say that $\mathbb{Z}$ is an ordered ring, or perhaps more precisely a commutative ordered ring.
Note that (iv) implies in particular that $1 > 0$.

**Proof.** Let us prove (i). The inequality $x > 0$ implies by item (i) of definition of ordered fields that $x + (-x) > 0 + (-x)$. Apply the algebraic properties of fields to obtain $0 > -x$. The “vice versa” follows by similar calculation.

For (ii), notice that $y < z$ implies $0 < z - y$ by item (i) of the definition of ordered fields. Apply item (ii) of the definition of ordered fields to obtain $0 < x(z - y)$. By algebraic properties, $0 < xz - xy$. Again by item (i) of the definition, $xy < xz$.

Part (iii) is left as an exercise.

To prove part (iv) first suppose $x > 0$. By item (ii) of the definition of ordered fields, $x^2 > 0$ (use $y = x$). If $x < 0$, we use part (iii) of this proposition, where we plug in $y = x$ and $z = 0$.

To prove part (v), notice that $1/x$ cannot be equal to zero (why?). Suppose $1/x < 0$, then $-1/x > 0$ by (i). Apply part (ii) (as $x > 0$) to obtain $x(-1/x) > 0x$ or $-1 > 0$, which contradicts $1 > 0$ by using part (i) again. Hence $1/x > 0$. Similarly, $1/y > 0$. Thus $(1/x)(1/y) > 0$ by definition of ordered field and by part (ii)

$$\frac{1}{x}\frac{1}{y}x < (\frac{1}{x})(\frac{1}{y})y.$$  

By algebraic properties we get $1/y < 1/x$.

Parts (vi) and (vii) are left as exercises.

The product of two positive numbers (elements of an ordered field) is positive. However, it is not true that if the product is positive, then each of the two factors must be positive. For instance, $(-1)(-1) = 1 > 0$.

**Proposition 1.1.9.** Let $x, y \in F$, where $F$ is an ordered field. If $xy > 0$, then either both $x$ and $y$ are positive, or both are negative.

**Proof.** We show the contrapositive: If either one of $x$ or $y$ is zero, or if $x$ and $y$ have opposite signs, then $xy$ is not positive. If either $x$ or $y$ is zero, then $xy$ is zero and hence not positive. Hence assume that $x$ and $y$ are nonzero and have opposite signs. Without loss of generality suppose $x > 0$ and $y < 0$. Multiply $y < 0$ by $x$ to get $xy < 0x = 0$.

**Example 1.1.10:** The reader may also know about the complex numbers, usually denoted by $\mathbb{C}$. That is, $\mathbb{C}$ is the set of numbers of the form $x + iy$, where $x$ and $y$ are real numbers, and $i$ is the imaginary number, a number such that $i^2 = -1$. The reader may remember from algebra that $\mathbb{C}$ is also a field; however, it is not an ordered field. While one can make $\mathbb{C}$ into an ordered set in some way, it is not possible to put an order on $\mathbb{C}$ that would make it an ordered field: In every ordered field, $-1 < 0$ and $x^2 > 0$ for all nonzero $x$, but in $\mathbb{C}$, $i^2 = -1$.

Finally, an ordered field that has the least-upper-bound property has the corresponding property for greatest lower bounds.

**Proposition 1.1.11.** Let $F$ be an ordered field with the least-upper-bound property. Let $A \subset F$ be a nonempty set that is bounded below. Then $\inf A$ exists.

**Proof.** Let $B := \{-x : x \in A\}$. Let $b \in F$ be a lower bound for $A$: If $x \in A$, then $x \geq b$. In other words, $-x \leq -b$. So $-b$ is an upper bound for $B$. Since $F$ has the least-upper-bound property, $c := \sup B$ exists, and $c \leq -b$. As $y \leq c$ for all $y \in B$, then $-c \leq x$ for all $x \in A$. So $-c$ is a lower bound for $A$. As $-c \geq b$, $-c$ is the greatest lower bound of $A$. 

\[ \square \]
1.1. Exercises

Exercise 1.1.1: Prove part (iii) of Proposition 1.1.8. That is, let $F$ be an ordered field and $x, y, z \in F$. Prove if $x < 0$ and $y < z$, then $xy > xz$.

Exercise 1.1.2: Let $S$ be an ordered set. Let $A \subset S$ be a nonempty finite subset. Then $A$ is bounded. Furthermore, $\inf A$ exists and is in $A$ and $\sup A$ exists and is in $A$. Hint: Use induction.

Exercise 1.1.3: Prove part (vi) of Proposition 1.1.8. That is, let $x, y \in F$, where $F$ is an ordered field, such that $0 < x < y$. Show that $x^2 < y^2$.

Exercise 1.1.4: Let $S$ be an ordered set. Let $B \subset S$ be bounded (above and below). Let $A \subset B$ be a nonempty subset. Suppose all the $\inf$s and $\sup$s exist. Show that $\inf B \leq \inf A \leq \sup A \leq \sup B$.

Exercise 1.1.5: Let $S$ be an ordered set. Let $A \subset S$ and suppose $b$ is an upper bound for $A$. Suppose $b \in A$. Show that $b = \sup A$.

Exercise 1.1.6: Let $D$ be the ordered set of all possible words (not just English words, all strings of letters) using the Latin alphabet using only lower case letters. The order is the lexicographic order as in a dictionary (e.g. $aa < aaa < dog < door$). Let $A$ be the subset of $D$ containing the words whose first letter is ‘a’ (e.g. $a \in A$, $abcd \in A$). Show that $A$ has a supremum and find what it is.

Exercise 1.1.7: Find a (nonstandard) ordering of the set of natural numbers $\mathbb{N}$ such that there exists a nonempty proper subset $A \subseteq \mathbb{N}$ and such that $\sup A$ exists in $\mathbb{N}$, but $\sup A \notin A$. To keep things straight it might be a good idea to use a different notation for the nonstandard ordering such as $n \prec m$.

Exercise 1.1.8: Let $F := \{0, 1, 2\}$.

a) Prove that there is exactly one way to define addition and multiplication so that $F$ is a field if 0 and 1 have their usual meaning of (A4) and (M4).

b) Show that $F$ cannot be an ordered field.

Exercise 1.1.9: Let $S$ be an ordered set and $A$ is a nonempty subset such that $\sup A$ exists. Suppose there is a $B \subset A$ such that whenever $x \in A$ there is a $y \in B$ such that $x \leq y$. Show that $\sup B$ exists and $\sup B = \sup A$.

Exercise 1.1.10: Let $D$ be the ordered set of all possible words (not just English words, all strings of letters of arbitrary length) using the Latin alphabet using only lower case letters. The order is the lexicographic order as in a dictionary (e.g. $aa < aaa < dog < door$). Let $A$ be the subset of $D$ containing the words whose first letter is ‘a’ (e.g. $a \in A$, $abcd \in A$). Show that $A$ has a supremum and find what it is.

Exercise 1.1.11: Let $F$ be an ordered field and $x, y, z, w \in F$.

a) Prove part (vii) of Proposition 1.1.8. That is, if $x \leq y$ and $z \leq w$, then $x + z \leq y + w$.

b) Prove that if $x < y$ and $z \leq w$, then $x + z < y + w$.

Exercise 1.1.12: Prove that any ordered field must contain a countably infinite set.

Exercise 1.1.13: Let $\mathbb{N}_\infty := \mathbb{N} \cup \{\infty\}$, where elements of $\mathbb{N}$ are ordered in the usual way amongst themselves, and $k < \infty$ for every $k \in \mathbb{N}$. Show $\mathbb{N}_\infty$ is an ordered set and that every subset $E \subset \mathbb{N}_\infty$ has a supremum in $\mathbb{N}_\infty$ (make sure to also handle the case of an empty set).

Exercise 1.1.14: Let $S := \{a_k : k \in \mathbb{N}\} \cup \{b_k : k \in \mathbb{N}\}$, ordered such that $a_k < b_j$ for every $k$ and $j$, $a_k < a_m$ whenever $k < m$, and $b_k > b_m$ whenever $k < m$.

a) Show that $S$ is an ordered set.

b) Show that every subset of $S$ is bounded (both above and below).

c) Find a bounded subset of $S$ that has no least upper bound.
1.2 The set of real numbers

Note: 2 lectures, the extended real numbers are optional

1.2.1 The set of real numbers

We finally get to the real number system. To simplify matters, instead of constructing the real number set from the rational numbers, we simply state their existence as a theorem without proof. Notice that \( \mathbb{Q} \) is an ordered field.

Theorem 1.2.1. There exists a unique* ordered field \( \mathbb{R} \) with the least-upper-bound property such that \( \mathbb{Q} \subset \mathbb{R} \).

Note that also \( \mathbb{N} \subset \mathbb{Q} \). We saw that 1 > 0. By induction (exercise) we can prove that \( n > 0 \) for all \( n \in \mathbb{N} \). Similarly, we verify simple statements about rational numbers. For example, we proved that if \( n > 0 \), then \( 1/n > 0 \). Then \( m < k \) implies \( m/n < k/n \).

Let us prove one of the most basic but useful results about the real numbers. The following proposition is essentially how an analyst proves an inequality.

Proposition 1.2.2. If \( x \in \mathbb{R} \) is such that \( x \leq \varepsilon \) for all \( \varepsilon \in \mathbb{R} \) where \( \varepsilon > 0 \), then \( x \leq 0 \).

Proof. If \( x > 0 \), then \( 0 < x/2 < x \) (why?). Taking \( \varepsilon = x/2 \) obtains a contradiction. Thus \( x \leq 0 \).

Another useful version of this idea is the following equivalent statement for nonnegative numbers: If \( x \geq 0 \) is such that \( x \leq \varepsilon \) for all \( \varepsilon > 0 \), then \( x = 0 \). And to prove that \( x \geq 0 \) in the first place, an analyst might prove that all \( x \geq -\varepsilon \) for all \( \varepsilon > 0 \). From now on, when we say \( x \geq 0 \) or \( \varepsilon > 0 \), we automatically mean that \( x, \varepsilon \in \mathbb{R} \).

A related simple fact is that any time we have two real numbers \( a < b \), then there is another real number \( c \) such that \( a < c < b \). Take, for example, \( c = a + b/2 \) (why?). In fact, there are infinitely many real numbers between \( a \) and \( b \). We will use this fact in the next example.

The most useful property of \( \mathbb{R} \) for analysts is not just that it is an ordered field, but that it has the least-upper-bound property. Essentially, we want \( \mathbb{Q} \), but we also want to take suprema (and infima) willy-nilly. So what we do is take \( \mathbb{Q} \) and throw in enough numbers to obtain \( \mathbb{R} \).

We mentioned already that \( \mathbb{R} \) contains elements that are not in \( \mathbb{Q} \) because of the least-upper-bound property. Let us prove it. We saw there is no rational square root of two. The set \( \{ x \in \mathbb{Q} : x^2 < 2 \} \) implies the existence of the real number \( \sqrt{2} \), although this fact requires a bit of work. See also Exercise 1.2.14.

Example 1.2.3: Claim: There exists a unique positive \( r \in \mathbb{R} \) such that \( r^2 = 2 \). We denote \( r \) by \( \sqrt{2} \).

Proof. Take the set \( A := \{ x \in \mathbb{R} : x^2 < 2 \} \). We first show that \( A \) is bounded above and nonempty. The equation \( x \geq 2 \) implies \( x^2 \geq 4 \) (see Exercise 1.1.3), so if \( x^2 < 2 \), then \( x < 2 \), and \( A \) is bounded above. As 1 \( \in A \), the set \( A \) is nonempty. We can therefore find the supremum.

Let \( r := \sup A \). We will show that \( r^2 = 2 \) by showing that \( r^2 \geq 2 \) and \( r^2 \leq 2 \). This is the way analysts show equality, by showing two inequalities. We already know that \( r \geq 1 > 0 \).

*Uniqueness is up to isomorphism, but we wish to avoid excessive use of algebra. For us, it is simply enough to assume that a set of real numbers exists. See Rudin [R2] for the construction and more details.
In the following, it may seem we are pulling certain expressions out of a hat. When writing a proof such as this we would, of course, come up with the expressions only after playing around with what we wish to prove. The order in which we write the proof is not necessarily the order in which we come up with the proof.

Let us first show that \( r^2 \geq 2 \). Take a positive number \( s \) such that \( s^2 < 2 \). We wish to find an \( h > 0 \) such that \((s + h)^2 < 2\). As \( 2 - s^2 > 0 \), we have \( \frac{2-s^2}{2s+1} > 0 \). Choose an \( h \in \mathbb{R} \) such that \( 0 < h < \frac{2-s^2}{2s+1} \). Furthermore, assume \( h < 1 \). Estimate,

\[
(s + h)^2 - s^2 = h(2s + h) < h(2s + 1) \quad \text{(since } h < 1) \\
< 2 - s^2 \quad \text{(since } h < \frac{2-s^2}{2s+1}).
\]

Therefore, \((s + h)^2 < 2\). Hence \( s + h \in A \), but as \( h > 0 \), we have \( s + h > s \). So \( s < r = \sup A \). As \( s \) was an arbitrary positive number such that \( s^2 < 2 \), it follows that \( r^2 \geq 2 \).

Now take a positive number \( s \) such that \( s^2 > 2 \). We wish to find an \( h > 0 \) such that \((s - h)^2 > 2\) and \( s - h \) is still positive. As \( s^2 - 2 > 0 \), we have \( \frac{s^2-2}{2s} > 0 \). Let \( h := \frac{s^2-2}{2s} \), and check \( s - h = s - \frac{s^2-2}{2s} = \frac{s}{2} + \frac{1}{s} > 0 \). Estimate,

\[
s^2 - (s - h)^2 = 2sh - h^2 < 2sh \quad \text{(since } h^2 > 0 \text{ as } h \neq 0) \\
= s^2 - 2 \quad \text{(since } h = \frac{s^2-2}{2s}).
\]

By subtracting \( s^2 \) from both sides and multiplying by \(-1\), we find \((s - h)^2 > 2\). Therefore, \( s - h \notin A \).

Moreover, if \( x \geq s - h \), then \( x^2 \geq (s - h)^2 > 2 \) (as \( x > 0 \) and \( s - h > 0 \)) and so \( x \notin A \). Thus, \( s - h \) is an upper bound for \( A \). However, \( s - h < s \), or in other words, \( s > r = \sup A \). Hence, \( r^2 \leq 2 \).

Together, \( r^2 \geq 2 \) and \( r^2 \leq 2 \) imply \( r^2 = 2 \). The existence part is finished. We still need to handle uniqueness. Suppose \( s \in \mathbb{R} \) such that \( s^2 = 2 \) and \( s > 0 \). Thus \( s^2 = r^2 \). However, if \( 0 < s < r \), then \( s^2 < r^2 \). Similarly, \( 0 < r < s \) implies \( r^2 < s^2 \). Hence \( s = r \). \( \Box \)

The number \( \sqrt{2} \notin \mathbb{Q} \). The set \( \mathbb{R} \setminus \mathbb{Q} \) is called the set of irrational numbers. We just saw that \( \mathbb{R} \setminus \mathbb{Q} \) is nonempty. Not only is it nonempty, we will see later that it is very large indeed.

Using the same technique as above, we can show that a positive real number \( x^{1/n} \) exists for all \( n \in \mathbb{N} \) and all \( x > 0 \). That is, for each \( x > 0 \), there exists a unique positive real number \( r \) such that \( r^n = x \). The proof is left as an exercise.

### 1.2.2 Archimedean property

As we have seen, there are plenty of real numbers in any interval. But there are also infinitely many rational numbers in any interval. The following is one of the fundamental facts about the real numbers. The two parts of the next theorem are actually equivalent, even though it may not seem like that at first sight.
Theorem 1.2.4.

(i) (Archimedean property) If \( x, y \in \mathbb{R} \) and \( x > 0 \), then there exists an \( n \in \mathbb{N} \) such that
\[
x x > y.
\]

(ii) (\( \mathbb{Q} \) is dense in \( \mathbb{R} \)) If \( x, y \in \mathbb{R} \) and \( x < y \), then there exists an \( r \in \mathbb{Q} \) such that \( x < r < y \).

Proof. Let us prove (i). Divide through by \( x \). Then (i) says that for every real number \( t := y/x \), we can find \( n \in \mathbb{N} \) such that \( n > t \). In other words, (i) says that \( \mathbb{N} \subset \mathbb{R} \) is not bounded above. Suppose for contradiction that \( \mathbb{N} \) is bounded above. Let \( b := \sup \mathbb{N} \). The number \( b - 1 \) cannot possibly be an upper bound for \( \mathbb{N} \) as it is strictly less than \( b \) (the least upper bound). Thus there exists an \( m \in \mathbb{N} \) such that \( m > b - 1 \). Add one to obtain \( m + 1 > b \), contradicting \( b \) being an upper bound.

![Figure 1.2: Idea of the proof of the density of \( \mathbb{Q} \): Find \( n \) such that \( y - x > 1/n \), then take the least \( m \) such that \( m/n > x \).](image)

Let us tackle (ii). See Figure 1.2 for a picture of the idea behind the proof. First assume \( x \geq 0 \). Note that \( y - x > 0 \). By (i), there exists an \( n \in \mathbb{N} \) such that
\[
n(y - x) > 1 \quad \text{or} \quad y - x > 1/n.
\]

Again by (i) the set \( A := \{ k \in \mathbb{N} : k > nx \} \) is nonempty. By the well ordering property of \( \mathbb{N} \), \( A \) has a least element \( m \), and as \( m \in A \), then \( m > nx \). Divide through by \( n \) to get \( x < m/n \). As \( m \) is the least element of \( A \), \( m - 1 \notin A \). If \( m > 1 \), then \( m - 1 \in \mathbb{N} \), but \( m - 1 \notin A \) and so \( m - 1 \leq nx \). If \( m = 1 \), then \( m - 1 = 0 \), and \( m - 1 \leq nx \) still holds as \( x \geq 0 \). In other words,
\[
m - 1 \leq nx \quad \text{or} \quad m \leq nx + 1.
\]

On the other hand, from \( n(y - x) > 1 \) we obtain \( ny > 1 + nx \). Hence \( ny > 1 + nx \geq m \), and therefore \( y > m/n \). Putting everything together we obtain \( x < m/n < y \). So take \( r = m/n \).

Now assume \( x < 0 \). If \( y > 0 \), then just take \( r = 0 \). If \( y \leq 0 \), then \( 0 \leq -y < -x \), and we find a rational \( q \) such that \(-y < q < -x\). Then take \( r = -q \).

Let us state and prove a simple but useful corollary of the Archimedean property.

Corollary 1.2.5. \( \inf \{1/n : n \in \mathbb{N} \} = 0 \).

Proof. Let \( A := \{1/n : n \in \mathbb{N} \} \). Obviously \( A \) is not empty. Furthermore, \( 1/n > 0 \) for all \( n \in \mathbb{N} \), and so 0 is a lower bound, and \( b := \inf A \) exists. As 0 is a lower bound, then \( b \geq 0 \). Take an arbitrary \( a > 0 \). By the Archimedean property there exists an \( n \) such that \( na > 1 \), or in other words \( a > 1/n \in A \). Therefore, \( a \) cannot be a lower bound for \( A \). Hence \( b = 0 \).

*Named after the Ancient Greek mathematician Archimedes of Syracuse (c. 287 BC – c. 212 BC). This property is Axiom V from Archimedes’ “On the Sphere and Cylinder” 225 BC.
1.2. THE SET OF REAL NUMBERS

1.2.3 Using supremum and infimum

Suprema and infima are compatible with algebraic operations. For a set $A \subset \mathbb{R}$ and $x \in \mathbb{R}$ define

$$x + A := \{x + y : y \in A\},$$
$$xA := \{xy : y \in A\}.$$

For example, if $A = \{1, 2, 3\}$, then $5 + A = \{6, 7, 8\}$ and $3A = \{3, 6, 9\}$.

**Proposition 1.2.6.** Let $A \subset \mathbb{R}$ be nonempty.

(i) If $x \in \mathbb{R}$ and $A$ is bounded above, then $\sup(x + A) = x + \sup A$.

(ii) If $x \in \mathbb{R}$ and $A$ is bounded below, then $\inf(x + A) = x + \inf A$.

(iii) If $x > 0$ and $A$ is bounded above, then $\sup(xA) = x\sup A$.

(iv) If $x > 0$ and $A$ is bounded below, then $\inf(xA) = x\inf A$.

(v) If $x < 0$ and $A$ is bounded below, then $\sup(xA) = x\inf A$.

(vi) If $x < 0$ and $A$ is bounded above, then $\inf(xA) = x\sup A$.

Do note that multiplying a set by a negative number switches supremum for an infimum and vice versa. Also, as the proposition implies that supremum (resp. infimum) of $x + A$ or $xA$ exists, it also implies that $x + A$ or $xA$ is nonempty and bounded above (resp. below).

**Proof.** Let us only prove the first statement. The rest are left as exercises.

Suppose $b$ is an upper bound for $A$. That is, $y \leq b$ for all $y \in A$. Then $x + y \leq x + b$ for all $y \in A$, and so $x + b$ is an upper bound for $x + A$. In particular, if $b = \sup A$, then

$$\sup(x + A) \leq x + b = x + \sup A.$$ 

The opposite inequality is similar. If $b$ is an upper bound for $x + A$, then $x + y \leq b$ for all $y \in A$ and so $y \leq b - x$ for all $y \in A$. So $b - x$ is an upper bound for $A$. If $b = \sup(x + A)$, then

$$\sup A \leq b - x = \sup(x + A) - x.$$ 

The result follows. 

Sometimes we need to apply supremum or infimum twice. Here is an example.

**Proposition 1.2.7.** Let $A, B \subset \mathbb{R}$ be nonempty sets such that $x \leq y$ whenever $x \in A$ and $y \in B$. Then $A$ is bounded above, $B$ is bounded below, and $\sup A \leq \inf B$.

**Proof.** Any $x \in A$ is a lower bound for $B$. Therefore $x \leq \inf B$ for all $x \in A$, so $\inf B$ is an upper bound for $A$. Hence, $\sup A \leq \inf B$. 

We must be careful about strict inequalities and taking suprema and infima. Note that $x < y$ whenever $x \in A$ and $y \in B$ still only implies $\sup A \leq \inf B$, and not a strict inequality. This is an important subtle point that comes up often. For example, take $A := \{0\}$ and take $B := \{1/n : n \in \mathbb{N}\}$. Then $0 < 1/n$ for all $n \in \mathbb{N}$. However, $\sup A = 0$ and $\inf B = 0$.

The proof of the following often used elementary fact is left to the reader. A similar statement holds for infima.
Proposition 1.2.8. If $S \subset \mathbb{R}$ is nonempty and bounded above, then for every $\varepsilon > 0$ there exists an $x \in S$ such that $(\sup S) - \varepsilon < x \leq \sup S$.

To make using suprema and infima even easier, we may want to write $\sup A$ and $\inf A$ without worrying about $A$ being bounded and nonempty. We make the following natural definitions.

Definition 1.2.9. Let $A \subset \mathbb{R}$ be a set.

(i) If $A$ is empty, then $\sup A := -\infty$.

(ii) If $A$ is not bounded above, then $\sup A := \infty$.

(iii) If $A$ is empty, then $\inf A := \infty$.

(iv) If $A$ is not bounded below, then $\inf A := -\infty$.

For convenience, $\infty$ and $-\infty$ are sometimes treated as if they were numbers, except we do not allow arbitrary arithmetic with them. We make $\mathbb{R}^* := \mathbb{R} \cup \{-\infty, \infty\}$ into an ordered set by letting

$-\infty < \infty$ and $-\infty < x$ and $x < \infty$ for all $x \in \mathbb{R}$.

The set $\mathbb{R}^*$ is called the set of extended real numbers. It is possible to define some arithmetic on $\mathbb{R}^*$. Most operations are extended in an obvious way, but we must leave $\infty - \infty$, $0 \cdot (\pm \infty)$, and $\frac{\pm \infty}{\pm \infty}$ undefined. We refrain from using this arithmetic, it leads to easy mistakes as $\mathbb{R}^*$ is not a field. Now we can take suprema and infima without fear of emptiness or unboundedness. In this book, we mostly avoid using $\mathbb{R}^*$ outside of exercises, and leave such generalizations to the interested reader.

1.2.4 Maxima and minima

By Exercise 1.1.2, a finite set of numbers always has a supremum or an infimum that is contained in the set itself. In this case we usually do not use the words supremum or infimum.

When a set $A$ of real numbers is bounded above, such that $\sup A \in A$, then we can use the word maximum and the notation $\text{max} A$ to denote the supremum. Similarly for infimum: When a set $A$ is bounded below and $\inf A \in A$, then we can use the word minimum and the notation $\text{min} A$. For example,

$$\text{max}\{1, 2.4, \pi, 100\} = 100,$$

$$\text{min}\{1, 2.4, \pi, 100\} = 1.$$ 

While writing sup and inf may be technically correct in this situation, max and min are generally used to emphasize that the supremum or infimum is in the set itself.

1.2.5 Exercises

Exercise 1.2.1: Prove that if $t > 0 \ (t \in \mathbb{R})$, then there exists an $n \in \mathbb{N}$ such that $\frac{1}{n^2} < t$.

Exercise 1.2.2: Prove that if $t \geq 0 \ (t \in \mathbb{R})$, then there exists an $n \in \mathbb{N}$ such that $n - 1 \leq t < n$.

Exercise 1.2.3: Finish the proof of Proposition 1.2.6.
1.2. THE SET OF REAL NUMBERS

Exercise 1.2.4: Let $x, y \in \mathbb{R}$. Suppose $x^2 + y^2 = 0$. Prove that $x = 0$ and $y = 0$.

Exercise 1.2.5: Show that $\sqrt{3}$ is irrational.

Exercise 1.2.6: Let $n \in \mathbb{N}$. Show that either $\sqrt{n}$ is either an integer or it is irrational.

Exercise 1.2.7: Prove the arithmetic-geometric mean inequality. That is, for two positive real numbers $x, y$, we have

$$\sqrt{xy} \leq \frac{x + y}{2}.$$  

Furthermore, equality occurs if and only if $x = y$.

Exercise 1.2.8: Show that for every pair of real numbers $x$ and $y$ such that $x < y$, there exists an irrational number $s$ such that $x < s < y$. Hint: Apply the density of $\mathbb{Q}$ to $\sqrt{x}/\sqrt{2}$ and $\sqrt{y}/\sqrt{2}$.

Exercise 1.2.9: Let $A$ and $B$ be two nonempty bounded sets of real numbers. Let $C := \{a + b : a \in A, b \in B\}$. Show that $C$ is a bounded set and that

$$\sup C = \sup A + \sup B \quad \text{and} \quad \inf C = \inf A + \inf B.$$ 

Exercise 1.2.10: Let $A$ and $B$ be two nonempty bounded sets of nonnegative real numbers. Define the set $C := \{ab : a \in A, b \in B\}$. Show that $C$ is a bounded set and that

$$\sup C = (\sup A)(\sup B) \quad \text{and} \quad \inf C = (\inf A)(\inf B).$$

Exercise 1.2.11 (Hard): Given $x > 0$ and $n \in \mathbb{N}$, show that there exists a unique positive real number $r$ such that $x = r^n$. Usually $r$ is denoted by $x^{1/n}$.

Exercise 1.2.12 (Easy): Prove Proposition 1.2.8.

Exercise 1.2.13: Prove the so-called Bernoulli’s inequality*: If $1 + x > 0$, then for all $n \in \mathbb{N}$, we have $(1 + x)^n \geq 1 + nx$.

Exercise 1.2.14: Prove $\sup \{x \in \mathbb{Q} : x^2 < 2\} = \sup \{x \in \mathbb{R} : x^2 < 2\}$.

Exercise 1.2.15:

a) Prove that given $y \in \mathbb{R}$, we have $\sup \{x \in \mathbb{Q} : x < y\} = y$.

b) Let $A \subset \mathbb{Q}$ be a set that is bounded above such that whenever $x \in A$ and $t \in \mathbb{Q}$ with $t < x$, then $t \in A$. Further suppose $\sup A \notin A$. Show that there exists a $y \in \mathbb{R}$ such that $A = \{x \in \mathbb{Q} : x < y\}$. A set such as $A$ is called a Dedekind cut.

c) Show that there is a bijection between $\mathbb{R}$ and Dedekind cuts.

Note: Dedekind used sets as in part b) in his construction of the real numbers.

Exercise 1.2.16: Prove that if $A \subset \mathbb{Z}$ is a nonempty subset bounded below, then there exists a least element in $A$. Now describe why this statement would simplify the proof of Theorem 1.2.4 part (ii) so that you do not have to assume $x \geq 0$.

*Named after the Swiss mathematician Jacob Bernoulli (1655–1705).
**Exercise 1.2.17:** Let us suppose we know $x^{1/n}$ exists for every $x > 0$ and every $n \in \mathbb{N}$ (see Exercise 1.2.11 above). For integers $p$ and $q > 0$ where $p/q$ is in lowest terms, define $x^{p/q} := (x^{1/q})^p$.

a) Show that the power is well-defined even if the fraction is not in lowest terms: If $p/q = m/k$ where $m$ and $k > 0$ are integers, then $(x^{1/q})^p = (x^{1/m})^k$.

b) Let $x$ and $y$ be two positive numbers and $r$ a rational number. Assuming $r > 0$, show $x < y$ if and only if $x^r < y^r$. Then suppose $r < 0$ and show: $x < y$ if and only if $x^r > y^r$.

c) Suppose $x > 1$ and $r, s$ are rational where $r < s$. Show $x^r < x^s$. If $0 < x < 1$ and $r < s$, show that $x^r > x^s$.

Hint: Write $r$ and $s$ with the same denominator.

d) (Challenging) For an irrational $z \in \mathbb{R} \setminus \mathbb{Q}$ and $x > 1$ define $x^z := \sup\{x^r : r \leq z, r \in \mathbb{Q}\}$, for $x = 1$ define $1^z = 1$, and for $0 < x < 1$ define $x^z := \inf\{x^r : r \leq z, r \in \mathbb{Q}\}$. Prove the two assertions of part b) for all real $z$.

*In §5.4 we will define exponential and the logarithm and define $x^z := \exp(z \ln x)$. We will then have sufficient machinery to make proofs of these assertions far easier. At this point, however, we do not yet have these tools.*
1.3 Absolute value and bounded functions

Note: 0.5–1 lecture

A concept we will encounter over and over is the concept of absolute value. You want to think of the absolute value as the “size” of a real number. Let us give a formal definition.

\[ |x| := \begin{cases} 
  x & \text{if } x \geq 0, \\
  -x & \text{if } x < 0. 
\end{cases} \]

Let us give the main features of the absolute value as a proposition.

**Proposition 1.3.1.**

(i) \(|x| \geq 0, \text{ moreover, } |x| = 0 \text{ if and only if } x = 0.\)

(ii) \(|-x| = |x| \text{ for all } x \in \mathbb{R}.\)

(iii) \(|xy| = |x||y| \text{ for all } x, y \in \mathbb{R}.\)

(iv) \(|x|^2 = x^2 \text{ for all } x \in \mathbb{R}.\)

(v) \(|x| \leq y \text{ if and only if } -y \leq x \leq y.\)

(vi) \(-|x| \leq x \leq |x| \text{ for all } x \in \mathbb{R}.\)

**Proof.** (i): First suppose \(x \geq 0.\) Then \(|x| = x \geq 0.\) Also, \(|x| = x = 0\) if and only if \(x = 0.\) On the other hand, if \(x < 0,\) then \(|x| = -x > 0,\) and \(|x|\) is never zero.

(ii): If \(x > 0,\) then \((-x) < 0\) and so \(|-x| = -(-x) = x = |x|.\) Similarly when \(x < 0,\) or \(x = 0.\)

(iii): If \(x\) or \(y\) is zero, then the result is immediate. When \(x\) and \(y\) are both positive, then \(|x||y| = xy.\) \(xy\) is also positive and hence \(xy = |xy|.\) If \(x\) and \(y\) are both negative, then \(xy\) is still positive and \(xy = |xy|,\) and \(|x||y| = (-x)(-y) = xy.\) Next assume \(x > 0\) and \(y < 0.\) Then \(|x||y| = x(-y) = -(xy).\) Now \(xy\) is negative and hence \(|xy| = -(xy).\) Similarly if \(x < 0\) and \(y > 0.\)

(iv): Immediate if \(x \geq 0.\) If \(x < 0,\) then \(|x|^2 = (-x)^2 = x^2.\)

(v): Suppose \(|x| \leq y.\) If \(x \geq 0,\) then \(x \leq y.\) It follows that \(y \geq 0,\) leading to \(-y \leq 0 \leq x.\) So \(-y \leq x \leq y\) holds. If \(x < 0,\) then \(|x| \leq y\) means \(-x \leq y.\) Negating both sides we get \(x \geq -y.\) Again \(y \geq 0\) and so \(y \geq 0 > x.\) Hence, \(-y \leq x \leq y.\)

On the other hand, suppose \(-y \leq x \leq y\) is true. If \(x \geq 0,\) then \(x \leq y\) is equivalent to \(|x| \leq y.\) If \(x < 0,\) then \(-y \leq x\) implies \((-x) \leq y,\) which is equivalent to \(|x| \leq y.\)

(vi): Apply (v) with \(y = |x|.\) \(\square\)

A property used frequently enough to give it a name is the so-called triangle inequality.

**Proposition 1.3.2** (Triangle Inequality). \(|x + y| \leq |x| + |y| \text{ for all } x, y \in \mathbb{R}.\)

**Proof.** Proposition 1.3.1 gives \(-|x| \leq x \leq |x|\) and \(-|y| \leq y \leq |y|.\) Add these two inequalities to obtain

\[-(|x| + |y|) \leq x + y \leq |x| + |y|.\]

Apply Proposition 1.3.1 again to find \(|x + y| \leq |x| + |y|.\) \(\square\)
There are other often applied versions of the triangle inequality.

**Corollary 1.3.3.** Let \( x, y \in \mathbb{R} \).

(i) (reverse triangle inequality) \(|(|x| - |y|)| \leq |x - y|\).

(ii) \(|x - y| \leq |x| + |y|\).

**Proof.** Let us plug in \( x = a - b \) and \( y = b \) into the standard triangle inequality to obtain

\[ |a| = |a - b + b| \leq |a - b| + |b|, \]

or \(|a| - |b| \leq |a - b|\). Switching the roles of \( a \) and \( b \) we find \(|b| - |a| \leq |b - a| = |a - b|\). Applying Proposition 1.3.1, we obtain the reverse triangle inequality.

The second version of the triangle inequality is obtained from the standard one by just replacing \( y \) with \(-y\), and noting \(|-y| = |y|\).

**Corollary 1.3.4.** Let \( x_1, x_2, \ldots, x_n \in \mathbb{R} \). Then

\[ |x_1 + x_2 + \cdots + x_n| \leq |x_1| + |x_2| + \cdots + |x_n|. \]

**Proof.** We proceed by induction. The conclusion holds trivially for \( n = 1 \), and for \( n = 2 \) it is the standard triangle inequality. Suppose the corollary holds for \( n \). Take \( n + 1 \) numbers \( x_1, x_2, \ldots, x_{n+1} \) and first use the standard triangle inequality, then the induction hypothesis

\[ |x_1 + x_2 + \cdots + x_n + x_{n+1}| \leq |x_1 + x_2 + \cdots + x_n| + |x_{n+1}| \leq |x_1| + |x_2| + \cdots + |x_n| + |x_{n+1}|. \]

Let us see an example of the use of the triangle inequality.

**Example 1.3.5:** Find a number \( M \) such that \(|x^2 - 9x + 1| \leq M \) for all \(-1 \leq x \leq 5\).

Using the triangle inequality, write

\[ |x^2 - 9x + 1| \leq |x^2| + |9x| + |1| = |x|^2 + 9|x| + 1. \]

The expression \(|x|^2 + 9|x| + 1\) is largest when \(|x|\) is largest (why?). In the interval provided, \(|x|\) is largest when \( x = 5 \) and so \(|x| = 5\). One possibility for \( M \) is

\[ M = 5^2 + 9(5) + 1 = 71. \]

There are, of course, other \( M \) that work. The bound of 71 is much higher than it need be, but we didn’t ask for the best possible \( M \), just one that works.

The last example leads us to the concept of bounded functions.

**Definition 1.3.6.** Suppose \( f : D \to \mathbb{R} \) is a function. We say \( f \) is bounded if there exists a number \( M \) such that \(|f(x)| \leq M \) for all \( x \in D \).
1.3. ABSOLUTE VALUE AND BOUNDED FUNCTIONS

In the example, we proved \( x^2 - 9x + 1 \) is bounded when considered as a function on \( D = \{ x : -1 \leq x \leq 5 \} \). On the other hand, if we consider the same polynomial as a function on the whole real line \( \mathbb{R} \), then it is not bounded.

For a function \( f : D \to \mathbb{R} \), we write (see Figure 1.3 for an example)

\[
\sup_{x \in D} f(x) := \sup f(D), \\
\inf_{x \in D} f(x) := \inf f(D).
\]

We also sometimes replace the “\( x \in D \)” with an expression. For example if, as before, \( f(x) = x^2 - 9x + 1 \), for \( -1 \leq x \leq 5 \), a little bit of calculus shows

\[
\sup_{x \in D} f(x) = \sup_{-1 \leq x \leq 5} (x^2 - 9x + 1) = 11, \\
\inf_{x \in D} f(x) = \inf_{-1 \leq x \leq 5} (x^2 - 9x + 1) = -77/4.
\]

**Proposition 1.3.7.** If \( f : D \to \mathbb{R} \) and \( g : D \to \mathbb{R} \) (\( D \) nonempty) are bounded* functions and

\[
f(x) \leq g(x) \quad \text{for all } x \in D,
\]

then

\[
\sup_{x \in D} f(x) \leq \sup_{x \in D} g(x) \quad \text{and} \quad \inf_{x \in D} f(x) \leq \inf_{x \in D} g(x).
\] (1.1)

Be careful with the variables. The \( x \) on the left side of the inequality in (1.1) is different from the \( x \) on the right. You should really think of, say, the first inequality as

\[
\sup_{x \in D} f(x) \leq \sup_{y \in D} g(y).
\]

Let us prove this inequality. If \( b \) is an upper bound for \( g(D) \), then \( f(x) \leq g(x) \leq b \) for all \( x \in D \), and hence \( b \) is also an upper bound for \( f(D) \), or \( f(x) \leq b \) for all \( x \in D \). Take the least upper bound of \( g(D) \) to get that for all \( x \in D \)

\[
f(x) \leq \sup_{y \in D} g(y).
\]

*The boundedness hypothesis is for simplicity, it can be dropped if we allow for the extended real numbers.
Therefore, \( \sup_{y \in D} g(y) \) is an upper bound for \( f(D) \) and thus greater than or equal to the least upper bound of \( f(D) \).

\[
\sup_{x \in D} f(x) \leq \sup_{y \in D} g(y).
\]

The second inequality (the statement about the inf) is left as an exercise (Exercise 1.3.4).

A common mistake is to conclude

\[
\sup_{x \in D} f(x) \leq \inf_{y \in D} g(y).
\]  \hspace{1cm} (1.2)

The inequality (1.2) is not true given the hypothesis of the proposition above. For this stronger inequality we need the stronger hypothesis

\[ f(x) \leq g(y) \quad \text{for all } x \in D \text{ and } y \in D. \]

The proof as well as a counterexample is left as an exercise (Exercise 1.3.5).

1.3.1 Exercises

**Exercise 1.3.1:** Show that \( |x - y| < \varepsilon \) if and only if \( x - \varepsilon < y < x + \varepsilon \).

**Exercise 1.3.2:** Show:

a) \( \max\{x, y\} = \frac{x + y + |x - y|}{2} \)

b) \( \min\{x, y\} = \frac{x + y - |x - y|}{2} \)

**Exercise 1.3.3:** Find a number \( M \) such that \( |x^3 - x^2 + 8x| \leq M \) for all \(-2 \leq x \leq 10\).

**Exercise 1.3.4:** Finish the proof of Proposition 1.3.7. That is, prove that given a set \( D \), and two bounded functions \( f : D \rightarrow \mathbb{R} \) and \( g : D \rightarrow \mathbb{R} \) such that \( f(x) \leq g(x) \) for all \( x \in D \), then

\[
\inf_{x \in D} f(x) \leq \inf_{x \in D} g(x).
\]

**Exercise 1.3.5:** Let \( f : D \rightarrow \mathbb{R} \) and \( g : D \rightarrow \mathbb{R} \) be functions (\( D \) nonempty).

a) Suppose \( f(x) \leq g(y) \) for all \( x \in D \) and \( y \in D \). Show that

\[
\sup_{x \in D} f(x) \leq \inf_{x \in D} g(x).
\]

b) Find a specific \( D, f, \) and \( g \), such that \( f(x) \leq g(x) \) for all \( x \in D \), but

\[
\sup_{x \in D} f(x) > \inf_{x \in D} g(x).
\]

**Exercise 1.3.6:** Prove Proposition 1.3.7 without the assumption that the functions are bounded. Hint: You need to use the extended real numbers.

**Exercise 1.3.7:** Let \( D \) be a nonempty set. Suppose \( f : D \rightarrow \mathbb{R} \) and \( g : D \rightarrow \mathbb{R} \) are bounded functions.

a) Show

\[
\sup_{x \in D} (f(x) + g(x)) \leq \sup_{x \in D} f(x) + \sup_{x \in D} g(x) \quad \text{and} \quad \inf_{x \in D} (f(x) + g(x)) \geq \inf_{x \in D} f(x) + \inf_{x \in D} g(x).
\]

b) Find examples where we obtain strict inequalities.
Exercise 1.3.8: Suppose \( f : D \to \mathbb{R} \) and \( g : D \to \mathbb{R} \) are bounded functions and \( \alpha \in \mathbb{R} \).

a) Show that \( \alpha f : D \to \mathbb{R} \) defined by \((\alpha f)(x) := \alpha f(x)\) is a bounded function.

b) Show that \( f + g : D \to \mathbb{R} \) defined by \((f + g)(x) := f(x) + g(x)\) is a bounded function.

Exercise 1.3.9: Let \( f : D \to \mathbb{R} \) and \( g : D \to \mathbb{R} \) be functions, \( \alpha \in \mathbb{R} \), and recall what \( f + g \) and \( \alpha f \) means from the previous exercise.

a) Prove that if \( f + g \) and \( g \) are bounded, then \( f \) is bounded.

b) Find an example where \( f \) and \( g \) are both unbounded, but \( f + g \) is bounded.

c) Prove that if \( f \) is bounded but \( g \) is unbounded, then \( f + g \) is unbounded.

d) Find an example where \( f \) is unbounded but \( \alpha f \) is bounded.
1.4 Intervals and the size of \( \mathbb{R} \)

Note: 0.5–1 lecture (proof of uncountability of \( \mathbb{R} \) can be optional)

You surely saw the notation for intervals before, but let us give a formal definition here. For \( a, b \in \mathbb{R} \) such that \( a < b \) we define

\[
\begin{align*}
[a, b] &:= \{ x \in \mathbb{R} : a \leq x \leq b \}, \\
(a, b) &:= \{ x \in \mathbb{R} : a < x < b \}, \\
(a, b] &:= \{ x \in \mathbb{R} : a < x \leq b \}, \\
[a, b) &:= \{ x \in \mathbb{R} : a \leq x < b \}.
\end{align*}
\]

The interval \([a, b]\) is called a closed interval and \((a, b)\) is called an open interval. The intervals of the form \((a, b]\) and \([a, b)\) are called half-open intervals.

The intervals above were all bounded intervals, since both \(a\) and \(b\) were real numbers. We define unbounded intervals,

\[
\begin{align*}
[a, \infty) &:= \{ x \in \mathbb{R} : a \leq x \}, \\
(a, \infty) &:= \{ x \in \mathbb{R} : a < x \}, \\
(-\infty, b] &:= \{ x \in \mathbb{R} : x \leq b \}, \\
(-\infty, b) &:= \{ x \in \mathbb{R} : x < b \}.
\end{align*}
\]

For completeness, we define \((-\infty, \infty) := \mathbb{R}\). The intervals \([a, \infty), (-\infty, b]\), and \(\mathbb{R}\) are sometimes called unbounded closed intervals, and \((a, \infty), (-\infty, b), \) and \(\mathbb{R}\) are sometimes called unbounded open intervals.

The proof of the following proposition is left as an exercise. In short, an interval is a set with at least two points that contains all points between any two points.*

**Proposition 1.4.1.** A set \( I \subset \mathbb{R} \) is an interval if and only if \( I \) contains at least 2 points and for all \(a, c \in I\) and \( b \in \mathbb{R} \) such that \(a < b < c\), we have \(b \in I\).

We have already seen that every open interval \((a, b)\) (where \(a < b\) of course) must be nonempty. For example, it contains the number \( \frac{a+b}{2} \). An unexpected fact is that from a set-theoretic perspective, all intervals have the same “size,” that is, they all have the same cardinality. For example the map \( f(x) := 2x \) takes the interval \([0, 1]\) bijectively to the interval \([0, 2]\).

Maybe more interestingly, the function \( f(x) := \tan(x) \) is a bijective map from \((-\pi/2, \pi/2)\) to \(\mathbb{R}\). Hence the bounded interval \((-\pi/2, \pi/2)\) has the same cardinality as \(\mathbb{R}\). It is not completely straightforward to construct a bijective map from \([0, 1]\) to \((0, 1)\), but it is possible.

And do not worry, there does exist a way to measure the “size” of subsets of real numbers that “sees” the difference between \([0, 1]\) and \([0, 2]\). However, its proper definition requires much more machinery than we have right now.

Let us say more about the cardinality of intervals and hence about the cardinality of \(\mathbb{R}\). We have seen that there exist irrational numbers, that is \(\mathbb{R} \setminus \mathbb{Q}\) is nonempty. The question is: How

\*Sometimes single point sets and the empty set are also called intervals, but in this book, intervals have at least 2 points. That is, we only defined the bounded intervals if \(a < b\).
many irrational numbers are there? It turns out there are a lot more irrational numbers than rational numbers. We have seen that \( \mathbb{Q} \) is countable, and we will show that \( \mathbb{R} \) is uncountable. In fact, the cardinality of \( \mathbb{R} \) is the same as the cardinality of \( \mathcal{P}(\mathbb{N}) \), although we will not prove this claim here.

**Theorem 1.4.2** (Cantor). \( \mathbb{R} \) is uncountable.

We give a version of Cantor’s original proof from 1874 as this proof requires the least setup. Normally this proof is stated as a contradiction, but a proof by contrapositive is easier to understand.

**Proof.** Let \( X \subset \mathbb{R} \) be a countably infinite subset such that for every pair of real numbers \( a < b \), there is an \( x \in X \) such that \( a < x < b \). Were \( \mathbb{R} \) countable, we could take \( X = \mathbb{R} \). We will show that \( X \) is necessarily a proper subset, and so \( \mathbb{R} \) cannot equal \( \mathbb{R} \), and \( \mathbb{R} \) must be uncountable.

As \( X \) is countably infinite, there is a bijection from \( \mathbb{N} \) to \( X \). We write \( X \) as a sequence of real numbers \( x_1, x_2, x_3, \ldots, \) such that each number in \( X \) is given by \( x_n \) for some \( n \in \mathbb{N} \).

We inductively construct two sequences of real numbers \( a_1, a_2, a_3, \ldots \) and \( b_1, b_2, b_3, \ldots \). Let \( a_1 := x_1 \) and \( b_1 := x_1 + 1 \). Note that \( a_1 < b_1 \) and \( x_1 \notin (a_1, b_1) \). For some \( k > 1 \), suppose \( a_j \) and \( b_j \) have been defined for \( j = 1, 2, \ldots, k-1 \), suppose the open interval \( (a_j, b_j) \) does not contain \( x_k \) for \( \ell = 1, 2, \ldots, j \), and suppose \( a_1 < a_2 < \cdots < a_{k-1} < b_{k-1} < \cdots < b_2 < b_1 \).

(i) Define \( a_k := x_n \), where \( n \) is the smallest \( n \in \mathbb{N} \) such that \( x_n \in (a_{k-1}, b_{k-1}) \). Such an \( x_n \) exists by our assumption on \( X \), and \( n \geq k \) by the assumption on \( (a_{k-1}, b_{k-1}) \).

(ii) Next, define \( b_k \) to be some real number in \( (a_k, b_{k-1}) \).

Notice that \( a_{k-1} < a_k < b_k < b_{k-1} \). Also notice that \( (a_k, b_k) \) does not contain \( x_k \) and hence does not contain \( x_j \) for \( j = 1, 2, \ldots, k \). The two sequences are now defined.

Claim: \( a_n < b_m \) for all \( n \) and \( m \) in \( \mathbb{N} \). Proof: Let us first assume \( n < m \). Then \( a_n < a_{n+1} < \cdots < a_{m-1} < a_m < b_m \). Similarly for \( n > m \). The claim follows.

Let \( A := \{a_n : n \in \mathbb{N}\} \) and \( B := \{b_n : n \in \mathbb{N}\} \). By Proposition 1.2.7 and the claim above,

\[
\sup A \leq \inf B.
\]

Define \( y := \sup A \). The number \( y \) cannot be a member of \( A \): If \( y = a_n \) for some \( n \), then \( y < a_{n+1} \), which is impossible. Similarly, \( y \) cannot be a member of \( B \). Therefore, \( a_n < y \) for all \( n \in \mathbb{N} \) and \( y < b_n \) for all \( n \in \mathbb{N} \). In other words, for every \( n \in \mathbb{N} \), we have \( y \in (a_n, b_n) \). By the construction of the sequence, \( x_n \notin (a_n, b_n) \), and so \( y \neq x_n \). As this was true for all \( n \in \mathbb{N} \), we have that \( y \notin X \).

We have constructed a real number \( y \) that is not in \( X \), and thus \( X \) is a proper subset of \( \mathbb{R} \). The sequence \( x_1, x_2, \ldots \) cannot contain all elements of \( \mathbb{R} \) and thus \( \mathbb{R} \) is uncountable. \( \square \)

### 1.4.1 Exercises

**Exercise 1.4.1:** For \( a < b \), construct an explicit bijection from \( \{a, b\} \) to \( (0, 1) \).

**Exercise 1.4.2:** Suppose \( f : [0, 1] \to (0, 1) \) is a bijection. Using \( f \), construct a bijection from \( [-1, 1] \) to \( \mathbb{R} \).

**Exercise 1.4.3:** Prove Proposition 1.4.1. That is, suppose \( I \subset \mathbb{R} \) is a subset with at least 2 elements such that if \( a < b < c \) and \( a, c \in I \), then \( b \in I \). Prove that \( I \) is one of the nine types of intervals explicitly given in this section. Furthermore, prove that the intervals given in this section all satisfy this property.
Exercise 1.4.4 (Hard): Construct an explicit bijection from $(0, 1]$ to $(0, 1)$. Hint: One approach is as follows: First map $(1/2, 1]$ to $(0, 1/2]$, then map $(1/4, 1/2]$ to $(1/2, 3/4]$, etc. Write down the map explicitly, that is, write down an algorithm that tells you exactly what number goes where. Then prove that the map is a bijection.

Exercise 1.4.5 (Hard): Construct an explicit bijection from $[0, 1]$ to $(0, 1)$.

Exercise 1.4.6:

a) Show that every closed interval $[a, b]$ is the intersection of countably many open intervals.

b) Show that every open interval $(a, b)$ is a countable union of closed intervals.

c) Show that an intersection of a possibly infinite family of bounded closed intervals, $\bigcap_{\lambda \in I} [a_\lambda, b_\lambda]$, is either empty, a single point, or a bounded closed interval.

Exercise 1.4.7: Suppose $S$ is a set of disjoint open intervals in $\mathbb{R}$. That is, if $(a, b) \in S$ and $(c, d) \in S$, then either $(a, b) = (c, d)$ or $(a, b) \cap (c, d) = \emptyset$. Prove $S$ is a countable set.

Exercise 1.4.8: Prove that the cardinality of $[0, 1]$ is the same as the cardinality of $(0, 1)$ by showing that $|[0, 1]| \leq |(0, 1)|$ and $|(0, 1)| \leq |[0, 1]|$. See Definition 0.3.28. This proof requires the Cantor–Bernstein–Schröder theorem we stated without proof. Note that this proof does not give you an explicit bijection.

Exercise 1.4.9 (Challenging): A number $x$ is algebraic if $x$ is a root of a polynomial with integer coefficients, in other words, $a_n x^n + a_{n-1} x^{n-1} + \cdots + a_1 x + a_0 = 0$ where all $a_n \in \mathbb{Z}$.

a) Show that there are only countably many algebraic numbers.

b) Show that there exist non-algebraic (transcendental) numbers (follow in the footsteps of Cantor, use the uncountability of $\mathbb{R}$).

Hint: Feel free to use the fact that a polynomial of degree $n$ has at most $n$ real roots.

Exercise 1.4.10 (Challenging): Let $F$ be the set of all functions $f: \mathbb{R} \to \mathbb{R}$. Prove $|\mathbb{R}| < |F|$ using Cantor’s Theorem 0.3.34.*

*Interestingly, if $C$ is the set of continuous functions, then $|\mathbb{R}| = |C|$. 
1.5 Decimal representation of the reals

Note: 1 lecture (optional)

We often think of real numbers as their decimal representation. For a positive integer $n$, we find the digits $d_K, d_{K-1}, \ldots, d_2, d_1, d_0$ for some $K$, where each $d_j$ is an integer between 0 and 9, then

$$n = d_K 10^K + d_{K-1} 10^{K-1} + \cdots + d_2 10^2 + d_1 10 + d_0.$$  

We often assume $d_K \neq 0$. To represent $n$ we write the sequence of digits: $n = d_K d_{K-1} \cdots d_2 d_1 d_0$.

By a (decimal) digit, we mean an integer between 0 and 9. Similarly, we represent some rational numbers. That is, for certain numbers $x$, we can find negative integer $-M$, a positive integer $K$, and digits $d_K, d_{K-1}, \ldots, d_1, d_0, d_{-1}, \ldots, d_{-M}$, such that

$$x = d_K 10^K + d_{K-1} 10^{K-1} + \cdots + d_2 10^2 + d_1 10 + d_0 + d_{-1} 10^{-1} + d_{-2} 10^{-2} + \cdots + d_{-M} 10^{-M}.$$  

We write $x = d_K d_{K-1} \cdots d_1 d_0 \cdot d_{-1} d_{-2} \cdots d_{-M}$.

Not every real number has such a representation, even the simple rational number $1/3$ does not. The irrational number $\sqrt{2}$ does not have such a representation either. To get a representation for all real numbers, we must allow infinitely many digits.

Let us consider only real numbers in the interval $(0, 1]$. If we find a representation for these, adding integers to them obtains a representation for all real numbers. Take an infinite sequence of decimal digits:

$$0.d_1 d_2 d_3 \ldots.$$  

That is, we have a digit $d_j$ for every $j \in \mathbb{N}$. We renumbered the digits to avoid the negative signs. We call the number

$$D_n := \frac{d_1}{10} + \frac{d_2}{10^2} + \frac{d_3}{10^3} + \cdots + \frac{d_n}{10^n},$$

the truncation of $x$ to $n$ decimal digits. We say this sequence of digits represents a real number $x$ if

$$x = \sup_{n \in \mathbb{N}} \left( \frac{d_1}{10} + \frac{d_2}{10^2} + \frac{d_3}{10^3} + \cdots + \frac{d_n}{10^n} \right) = \sup_{n \in \mathbb{N}} D_n.$$  

Proposition 1.5.1.

(i) Every infinite sequence of digits $0.d_1 d_2 d_3 \ldots$ represents a unique real number $x \in [0, 1]$, and

$$D_n \leq x \leq D_n + \frac{1}{10^n} \quad \text{for all } n \in \mathbb{N}.$$  

(ii) For every $x \in (0, 1]$ there exists an infinite sequence of digits $0.d_1 d_2 d_3 \ldots$ that represents $x$. There exists a unique representation such that

$$D_n < x \leq D_n + \frac{1}{10^n} \quad \text{for all } n \in \mathbb{N}.$$
Proof. We start with the first item. Take an arbitrary infinite sequence of digits $0.d_1 d_2 d_3 \ldots$. Use the geometric sum formula to write
\[
D_n = \frac{d_1}{10} + \frac{d_2}{10^2} + \frac{d_3}{10^3} + \cdots + \frac{d_n}{10^n} \leq \frac{9}{10} + \frac{9}{10^2} + \frac{9}{10^3} + \cdots + \frac{9}{10^n} = \frac{9}{10} \left(1 + \frac{1}{10} + \frac{1}{10^2} + \cdots + \frac{1}{10}^{n-1}\right) = \frac{9}{10} \left(1 - \frac{1}{10^n}\right) = 1 - \frac{1}{10^n} < 1.
\]
In particular, $D_n < 1$ for all $n$. A sum of nonnegative numbers is nonnegative so $D_n \geq 0$, and hence
\[
0 \leq \sup_{n \in \mathbb{N}} D_n \leq 1.
\]
Therefore, $0.d_1 d_2 d_3 \ldots$ represents a unique number $x := \sup_{n \in \mathbb{N}} D_n \in [0, 1]$. As $x$ is a supremum, then $D_n \leq x$. Take $m \in \mathbb{N}$. If $m < n$, then $D_m - D_n \leq 0$. If $m > n$, then computing as above
\[
D_m - D_n = \frac{d_{n+1}}{10^{n+1}} + \frac{d_{n+2}}{10^{n+2}} + \frac{d_{n+3}}{10^{n+3}} + \cdots + \frac{d_m}{10^m} \leq \frac{1}{10^n} \left(1 - \frac{1}{10}^{m-n}\right) < \frac{1}{10^n}.
\]
Take the supremum over $m$ to find
\[
x - D_n \leq \frac{1}{10^n}.
\]
We move on to the second item. Take any $x \in (0, 1]$. First let us tackle the existence. For convenience, let $D_0 := 0$. Then, $D_0 < x \leq D_0 + 10^{-0}$. Suppose we defined the digits $d_1, d_2, \ldots, d_n$, and that $D_k < x \leq D_k + 10^{-k}$, for $k = 0, 1, 2, \ldots, n$. We need to define $d_{n+1}$.

By the Archimedean property of the real numbers, find an integer $j$ such that $x - D_n \leq j10^{-(n+1)}$. Take the least such $j$ and obtain
\[
(j - 1)10^{-(n+1)} < x - D_n \leq j10^{-(n+1)}.
\]
Let $d_{n+1} := j - 1$. As $D_n < x$, then $d_{n+1} = j - 1 \geq 0$. On the other hand, since $x - D_n \leq 10^{-n}$, we have that $j$ is at most 10, and therefore $d_{n+1} \leq 9$. So $d_{n+1}$ is a decimal digit. Since $D_{n+1} = D_n + d_{n+1}10^{-(n+1)}$ add $D_n$ to the inequality (1.3) above:
\[
D_{n+1} = D_n + (j - 1)10^{-(n+1)} < x \leq D_n + j10^{-(n+1)} = D_n + (j - 1)10^{-(n+1)} + 10^{-(n+1)} = D_{n+1} + 10^{-(n+1)}.
\]
And so $D_{n+1} \leq x - D_{n+1} + 10^{-(n+1)}$ holds. We inductively defined an infinite sequence of digits $0.d_1 d_2 d_3 \ldots$.

Consider $D_n < x \leq D_n + 10^{-n}$. As $D_n < x$ for all $n$, then $\sup\{D_n : n \in \mathbb{N}\} \leq x$. The second inequality for $D_n$ implies
\[
x - \sup\{D_m : m \in \mathbb{N}\} \leq x - D_n \leq 10^{-n}.
\]
As the inequality holds for all $n$ and $10^{-n}$ can be made arbitrarily small (see Exercise 1.5.8), we have $x \leq \sup\{D_m : m \in \mathbb{N}\}$. Therefore, $\sup\{D_m : m \in \mathbb{N}\} = x$. 


What is left to show is the uniqueness. Suppose \(0.e_1 e_2 e_3 \ldots\) is another representation of \(x\). Let \(E_n\) be the \(n\)-digit truncation of \(0.e_1 e_2 e_3 \ldots\), and suppose \(E_n < x \leq E_n + 10^{-n}\) for all \(n \in \mathbb{N}\). Suppose for some \(K \in \mathbb{N}\), \(e_n = d_n\) for all \(n < K\), so \(D_{K-1} = E_{K-1}\). Then

\[
E_K = D_{K-1} + e_K 10^{-K} < x \leq E_K + 10^{-K} = D_{K-1} + e_K 10^{-K} + 10^{-K}.
\]

Subtracting \(D_{K-1}\) and multiplying by \(10^K\) we get

\[
e_K < (x - D_{K-1}) 10^K \leq e_K + 1.
\]

Similarly,

\[
d_K < (x - D_{K-1}) 10^K \leq d_K + 1.
\]

Hence, both \(e_K\) and \(d_K\) are the largest integer \(j\) such that \(j < (x - D_{K-1}) 10^K\), and therefore \(e_K = d_K\). That is, the representation is unique. \(\square\)

The representation is not unique if we do not require \(D_n < x\) for all \(n\). For example, for the number \(1/2\), the method in the proof obtains the representation

\[
0.49999 \ldots.
\]

However, \(1/2\) also has the representation \(0.50000 \ldots\).

The only numbers that have nonunique representations are ones that end either in an infinite sequence of 0s or 9s, because the only representation for which \(D_n = x\) is one where all digits past the \(n\)th digit are zero. In this case there are exactly two representations of \(x\) (see the exercises).

Let us give another proof of the uncountability of the reals using decimal representations. This is Cantor’s second proof, and is probably better known. This proof may seem shorter, but it is because we already did the hard part above and we are left with a slick trick to prove that \(\mathbb{R}\) is uncountable. This trick is called Cantor diagonalization and finds use in other proofs as well.

**Theorem 1.5.2** (Cantor). The set \((0, 1]\) is uncountable.

**Proof.** Let \(X := \{x_1, x_2, x_3, \ldots\}\) be any countable subset of real numbers in \((0, 1]\). We will construct a real number not in \(X\). Let

\[
x_n = 0.d_1^n d_2^n d_3^n \ldots
\]

be the unique representation from the proposition, that is, \(d_j^n\) is the \(j\)th digit of the \(n\)th number. Let

\[
e_n := \begin{cases} 1 & \text{if } d_j^n \neq 1, \\ 2 & \text{if } d_j^n = 1. \end{cases}
\]

Let \(E_n\) be the \(n\)-digit truncation of \(y = 0.e_1 e_2 e_3 \ldots\). Because all the digits are nonzero we get \(E_n < E_{n+1} \leq y\). Therefore

\[
E_n < y \leq E_n + 10^{-n}
\]

for all \(n\), and the representation is the unique one for \(y\) from the proposition. For every \(n\), the \(n\)th digit of \(y\) is different from the \(n\)th digit of \(x_n\), so \(y \neq x_n\). Therefore \(y \notin X\), and as \(X\) was an arbitrary countable subset, \((0, 1]\) must be uncountable. See Figure 1.4 for an example. \(\square\)
Using decimal digits we can also find lots of numbers that are not rational. The following proposition is true for every rational number, but we give it only for $x \in (0, 1]$ for simplicity.

**Proposition 1.5.3.** If $x \in (0, 1]$ is a rational number and $x = 0.d_1d_2d_3\ldots$, then the decimal digits eventually start repeating. That is, there are positive integers $N$ and $P$, such that for all $n \geq N$, $d_n = d_{n+P}$.

**Proof.** Suppose $x = p/q$ for positive integers $p$ and $q$. Suppose also that $x$ is a number with a unique representation, as otherwise we have seen above that both its representations are repeating, see also Exercise 1.5.3. This also means that $x \neq 1$ so $p < q$.

To compute the first digit we take $10p$ and divide by $q$. Let $d_1$ be the quotient, and the remainder $r_1$ is some integer between 0 and $q - 1$. That is, $d_1$ is the largest integer such that $d_1q \leq 10p$ and then $r_1 = 10p - d_1q$. As $p < q$, then $d_1 < 10$, so $d_1$ is a digit. Furthermore,

$$
\frac{d_1}{10} \leq \frac{p}{q} = \frac{d_1}{10} + \frac{r_1}{10q} \leq \frac{d_1}{10} + \frac{1}{10}.
$$

The first inequality must be strict since $x$ has a unique representation. That is, $d_1$ really is the first digit. What is left is $r_1/(10q)$. This is the same as computing the first digit of $r_1/q$. To compute $d_2$ divide $10r_1$ by $q$, and so on. After computing $n - 1$ digits, we have $r_{n-1}/q = D_{n-1} + r_{n-1}/(10^nq)$. To get the $n$th digit, divide $10r_{n-1}$ by $q$ to get quotient $d_n$, remainder $r_n$, and the inequalities

$$
\frac{d_n}{10} \leq \frac{r_{n-1}}{q} = \frac{d_n}{10} + \frac{r_n}{10q} \leq \frac{d_n}{10} + \frac{1}{10}.
$$

Dividing by $10^{n-1}$ and adding $D_{n-1}$ we find

$$
D_n \leq D_{n-1} + \frac{r_{n-1}}{10^nq} = \frac{p}{q} \leq D_n + \frac{1}{10^n}.
$$

By uniqueness we really have the $n$th digit $d_n$ from the construction.

The new digit depends only the remainder from the previous step. There are at most $q$ possible remainders and hence at some step the process must start repeating itself, and $P$ is at most $q$. 

The converse of the proposition is also true and is left as an exercise.
Example 1.5.4: The number

\[ x = 0.101001000100001000001 \ldots , \]

is irrational. That is, the digits are \( n \) zeros, then a one, then \( n + 1 \) zeros, then a one, and so on and so forth. The fact that \( x \) is irrational follows from the proposition; the digits never start repeating. For every \( P \), if we go far enough, we find a 1 followed by at least \( P + 1 \) zeros.

1.5.1 Exercises

Exercise 1.5.1 (Easy): What is the decimal representation of 1 guaranteed by Proposition 1.5.1? Make sure to show that it does satisfy the condition.

Exercise 1.5.2: Prove the converse of Proposition 1.5.3, that is, if the digits in the decimal representation of \( x \) are eventually repeating, then \( x \) must be rational.

Exercise 1.5.3: Show that real numbers \( x \in (0, 1) \) with nonunique decimal representation are exactly the rational numbers that can be written as \( \frac{m}{10^n} \) for some integers \( m \) and \( n \). In this case show that there exist exactly two representations of \( x \).

Exercise 1.5.4: Let \( b \geq 2 \) be an integer. Define a representation of a real number in \([0, 1]\) in terms of base \( b \) rather than base 10 and prove Proposition 1.5.1 for base \( b \).

Exercise 1.5.5: Using the previous exercise with \( b = 2 \) (binary), show that cardinality of \( \mathbb{R} \) is the same as the cardinality of \( \mathcal{P}(\mathbb{N}) \), obtaining yet another (though related) proof that \( \mathbb{R} \) is uncountable. Hint: Construct two injections, one from \([0, 1]\) to \( \mathcal{P}(\mathbb{N}) \) and one from \( \mathcal{P}(\mathbb{N}) \) to \([0, 1]\). Hint 2: Given a set \( A \subset \mathbb{N} \), let the nth binary digit of \( x \) be 1 if \( n \in A \).

Exercise 1.5.6 (Challenging): Explicitly construct an injection from \([0, 1] \times [0, 1]\) to \([0, 1]\) (think about why this is so surprising*). Then describe the set of numbers in \([0, 1]\) not in the image of your injection (unless, of course, you managed to construct a bijection). Hint: Consider even and odd digits of the decimal expansion.

Exercise 1.5.7: Prove that if \( x = \frac{n}{q} \in (0, 1) \) is a rational number, \( q > 1 \), then the period \( P \) of repeating digits in the decimal representation of \( x \) is in fact less than or equal to \( q - 1 \).

Exercise 1.5.8: Prove that if \( b \in \mathbb{N} \) and \( b \geq 2 \), then for every \( \varepsilon > 0 \), there is an \( n \in \mathbb{N} \) such that \( b^{-n} < \varepsilon \). Hint: One possibility is to first prove that \( b^n > n \) for all \( n \in \mathbb{N} \) by induction.

Exercise 1.5.9: Explicitly construct an injection \( f : \mathbb{R} \to \mathbb{R} \setminus \mathbb{Q} \) using Proposition 1.5.3.

*With quite a bit more work (or by applying the Cantor–Bernstein–Schröder theorem) one can prove that there is a bijection. When he proved this result, Cantor apparently wrote “I see it but I don’t believe it.”
Chapter 2
Sequences and Series

2.1 Sequences and limits

Note: 2.5 lectures

Analysis is essentially about taking limits. The most basic type of a limit is a limit of a sequence of real numbers. We have already seen sequences used informally. Let us give the formal definition.

Definition 2.1.1. A sequence (of real numbers) is a function \( x : \mathbb{N} \rightarrow \mathbb{R} \). Instead of \( x(n) \), we usually denote the \( n \)th element in the sequence by \( x_n \). We use the notation \( \{x_n\}_{n=1}^{\infty} \), or more precisely \( \{x_n\}_{n=1}^{\infty} \), to denote a sequence.

A sequence \( \{x_n\} \) is bounded if there exists a \( B \in \mathbb{R} \) such that \( |x_n| \leq B \) for all \( n \in \mathbb{N} \).

In other words, the sequence \( \{x_n\} \) is bounded whenever the set \( \{x_n : n \in \mathbb{N}\} \) is bounded, or equivalently when it is bounded as a function.

When we need to give a concrete sequence we often give each term as a formula in terms of \( n \). For example, \( \{1/n\}_{n=1}^{\infty} \), or simply \( \{1/n\} \), stands for the sequence \( 1, 1/2, 1/3, 1/4, 1/5, \ldots \). The sequence \( \{1/n\} \) is a bounded sequence (\( B = 1 \) suffices). On the other hand the sequence \( \{n\} \) stands for \( 1, 2, 3, 4, \ldots \), and this sequence is not bounded (why?).

While the notation for a sequence is similar\(^*\) to that of a set, the notions are distinct. For example, the sequence \( \{(-1)^n\} \) is the sequence \(-1, 1, -1, 1, -1, 1, \ldots \), whereas the set of values, the range of the sequence, is just the set \( \{-1, 1\} \). We can write this set as \( \{(-1)^n : n \in \mathbb{N}\} \). When ambiguity can arise, we use the words sequence or set to distinguish the two concepts.

Another example of a sequence is the so-called constant sequence. That is a sequence \( \{c\} = c, c, c, c, \ldots \) consisting of a single constant \( c \in \mathbb{R} \) repeating indefinitely.

We now get to the idea of a limit of a sequence. We will see in Proposition 2.1.6 that the notation below is well-defined. That is, if a limit exists, then it is unique. So it makes sense to talk about the limit of a sequence.

\(^*\)[BS] use the notation \( (x_n) \) to denote a sequence instead of \( \{x_n\} \), which is what [R2] uses. Both are common.
**Definition 2.1.2.** A sequence \( \{x_n\} \) is said to converge to a number \( x \in \mathbb{R} \) if for every \( \varepsilon > 0 \), there exists an \( M \in \mathbb{N} \) such that \( |x_n - x| < \varepsilon \) for all \( n \geq M \). The number \( x \) is said to be the limit of \( \{x_n\} \). We write
\[
\lim_{n \to \infty} x_n := x.
\]

A sequence that converges is said to be convergent. Otherwise, we say the sequence diverges or that it is divergent.

It is good to know intuitively what a limit means. It means that eventually every number in the sequence is close to the number \( x \). More precisely, we can get arbitrarily close to the limit, provided we go far enough in the sequence. It does not mean we ever reach the limit. It is possible, and quite common, that there is no \( x_n \) in the sequence that equals the limit \( x \). We illustrate the concept in Figure 2.1. In the figure we first think of the sequence as a graph, as it is a function of \( \mathbb{N} \). Secondly we also plot it as a sequence of labeled points on the real line.

![Illustration of convergence](image)

**Figure 2.1:** Illustration of convergence. On top, we show the first ten points of the sequence as a graph with \( M \) and the interval around the limit \( x \) marked. On bottom, the points of the same sequence are marked on the number line.

When we write \( \lim x_n = x \) for some real number \( x \), we are saying two things: first, that \( \{x_n\} \) is convergent, and second, that the limit is \( x \).

The definition above is one of the most important definitions in analysis, and it is necessary to understand it perfectly. The key point in the definition is that given any \( \varepsilon > 0 \), we can find an \( M \). The \( M \) can depend on \( \varepsilon \), so we only pick an \( M \) once we know \( \varepsilon \). Let us illustrate convergence on a few examples.

**Example 2.1.3:** The constant sequence \( 1, 1, 1, 1, \ldots \) is convergent and the limit is 1. For every \( \varepsilon > 0 \), we pick \( M = 1 \).
Example 2.1.4: Claim: The sequence \( \{1/n\} \) is convergent and
\[
\lim_{n \to \infty} \frac{1}{n} = 0.
\]
Proof: Given an \( \varepsilon > 0 \), we find an \( M \in \mathbb{N} \) such that \( 0 < 1/M < \varepsilon \) (Archimedean property at work). Then for all \( n \geq M \),
\[
|x_n - 0| = \left| \frac{1}{n} \right| = \frac{1}{n} \leq \frac{1}{M} < \varepsilon.
\]

Example 2.1.5: The sequence \( \{(−1)^n\} \) is divergent. Proof: If there were a limit \( x \), then for \( \varepsilon = \frac{1}{2} \) we expect an \( M \) that satisfies the definition. Suppose such an \( M \) exists. Then for an even \( n \geq M \) we compute
\[
\frac{1}{2} > |x_n - x| = |1 - x| \quad \text{and} \quad \frac{1}{2} > |x_{n+1} - x| = |-1 - x|.
\]
But
\[
2 = |1 - x - (-1 - x)| \leq |1 - x| + |-1 - x| < \frac{1}{2} + \frac{1}{2} = 1,
\]
and that is a contradiction.

Proposition 2.1.6. A convergent sequence has a unique limit.

The proof of this proposition exhibits a useful technique in analysis. Many proofs follow the same general scheme. We want to show a certain quantity is zero. We write the quantity using the triangle inequality as two quantities, and we estimate each one by arbitrarily small numbers.

Proof. Suppose the sequence \( \{x_n\} \) has limits \( x \) and \( y \). Take an arbitrary \( \varepsilon > 0 \). From the definition find an \( M_1 \) such that for all \( n \geq M_1 \), \( |x_n - x| < \varepsilon/2 \). Similarly, find an \( M_2 \) such that for all \( n \geq M_2 \), we have \( |x_n - y| < \varepsilon/2 \). Now take an \( n \) such that \( n \geq M_1 \) and also \( n \geq M_2 \), and estimate
\[
|y - x| = |x_n - x - (x_n - y)|
\leq |x_n - x| + |x_n - y|
< \varepsilon/2 + \varepsilon/2 = \varepsilon.
\]
As \( |y - x| < \varepsilon \) for all \( \varepsilon > 0 \), then \( |y - x| = 0 \) and \( y = x \). Hence the limit (if it exists) is unique.

Proposition 2.1.7. A convergent sequence \( \{x_n\} \) is bounded.

Proof. Suppose \( \{x_n\} \) converges to \( x \). Thus there exists an \( M \in \mathbb{N} \) such that for all \( n \geq M \), we have \( |x_n - x| < 1 \). Let \( B_1 := |x| + 1 \) and note that for \( n \geq M \),
\[
|x_n| = |x_n - x + x|
\leq |x_n - x| + |x|
< 1 + |x| = B_1.
\]
The set \( \{|x_1|, |x_2|, \ldots, |x_{M-1}|\} \) is a finite set and hence let
\[
B_2 := \max\{|x_1|, |x_2|, \ldots, |x_{M-1}|\}.
\]
Let \( B := \max\{B_1, B_2\} \). Then for all \( n \in \mathbb{N} \),
\[
|x_n| \leq B.
\]
The sequence \((-1)^n\) shows that the converse does not hold. A bounded sequence is not necessarily convergent.

**Example 2.1.8:** Let us show the sequence \(\left\{\frac{n^2+1}{n^2+n}\right\}\) converges and

\[
\lim_{n \to \infty} \frac{n^2+1}{n^2+n} = 1.
\]

Given \(\varepsilon > 0\), find \(M \in \mathbb{N}\) such that \(\frac{1}{M} < \varepsilon\). Then for all \(n \geq M\),

\[
\left|\frac{n^2+1}{n^2+n} - 1\right| = \left|\frac{n^2+1-(n^2+n)}{n^2+n}\right| = \left|\frac{1-n}{n^2+n}\right| \leq \frac{n}{n^2+n} = \frac{1}{n+1} \leq \frac{1}{n} \leq \frac{1}{M} < \varepsilon.
\]

Therefore, \(\lim n^2+1 = 1\). This example shows that sometimes to get what you want, you must throw away some information to get a simpler estimate.

### 2.1.1 Monotone sequences

The simplest type of a sequence is a monotone sequence. Checking that a monotone sequence converges is as easy as checking that it is bounded. It is also easy to find the limit for a convergent monotone sequence, provided we can find the supremum or infimum of a countable set of numbers.

**Definition 2.1.9.** A sequence \(\{x_n\}\) is **monotone increasing** if \(x_n \leq x_{n+1}\) for all \(n \in \mathbb{N}\). A sequence \(\{x_n\}\) is **monotone decreasing** if \(x_n \geq x_{n+1}\) for all \(n \in \mathbb{N}\). If a sequence is either monotone increasing or monotone decreasing, we can simply say the sequence is **monotone**.

For example, \(\{n\}\) is monotone increasing, \(\{1/n\}\) is monotone decreasing, the constant sequence \(\{1\}\) is both monotone increasing and monotone decreasing, and \(\{(-1)^n\}\) is not monotone. First few terms of a sample monotone increasing sequence are shown in Figure 2.2.

**Proposition 2.1.10.** A monotone sequence \(\{x_n\}\) is bounded if and only if it is convergent. Furthermore, if \(\{x_n\}\) is monotone increasing and bounded, then

\[
\lim_{n \to \infty} x_n = \sup\{x_n : n \in \mathbb{N}\}.
\]

If \(\{x_n\}\) is monotone decreasing and bounded, then

\[
\lim_{n \to \infty} x_n = \inf\{x_n : n \in \mathbb{N}\}.
\]

*Some authors use the word monotonic.*
Proof. Consider a monotone increasing sequence \( \{x_n\} \). Suppose the sequence is bounded, that is, the set \( \{x_n : n \in \mathbb{N}\} \) is bounded. Let

\[
x := \sup\{x_n : n \in \mathbb{N}\}.
\]

Let \( \varepsilon > 0 \) be arbitrary. As \( x \) is the supremum, then there must be at least one \( M \in \mathbb{N} \) such that \( x_M > x - \varepsilon \) (because \( x \) is the supremum). As \( \{x_n\} \) is monotone increasing, then it is easy to see (by induction) that \( x_n \geq x_M \) for all \( n \geq M \). Hence for all \( n \geq M \),

\[
|x_n - x| = x - x_n \leq x - x_M < \varepsilon.
\]

Therefore, the sequence converges to \( x \), so a bounded monotone increasing sequence converges. For the other direction, we have already proved that a convergent sequence is bounded.

The proof for monotone decreasing sequences is left as an exercise. \( \square \)

**Example 2.1.11:** Take the sequence \( \left\{ \frac{1}{\sqrt{n}} \right\} \).

The sequence is bounded below as \( \frac{1}{\sqrt{n}} > 0 \) for all \( n \in \mathbb{N} \). Let us show that it is monotone decreasing. We start with \( \sqrt{n+1} \geq \sqrt{n} \) (why is that true?). From this inequality we obtain

\[
\frac{1}{\sqrt{n+1}} \leq \frac{1}{\sqrt{n}}.
\]

So the sequence is monotone decreasing and bounded below (hence bounded). **Proposition 2.1.10** says that that the sequence is convergent and

\[
\lim_{n \to \infty} \frac{1}{\sqrt{n}} = \inf\left\{ \frac{1}{\sqrt{n}} : n \in \mathbb{N} \right\}.
\]

We already know that the infimum is greater than or equal to 0, as 0 is a lower bound. Take a number \( b \geq 0 \) such that \( b \leq \frac{1}{\sqrt{n}} \) for all \( n \). We square both sides to obtain

\[
b^2 \leq \frac{1}{n} \quad \text{for all} \ n \in \mathbb{N}.
\]
We have seen before that this implies that $b^2 \leq 0$ (a consequence of the Archimedean property). As $b^2 \geq 0$ as well, we have $b^2 = 0$ and so $b = 0$. Hence, $b = 0$ is the greatest lower bound, and $\lim_{n \to \infty} \frac{1}{\sqrt{n}} = 0$.

**Example 2.1.12:** A word of caution: We must show that a monotone sequence is bounded in order to use Proposition 2.1.10 to conclude a sequence converges. The sequence $\{1 + 1/2 + \cdots + 1/n\}$ is a monotone increasing sequence that grows very slowly. We will see, once we get to series, that this sequence has no upper bound and so does not converge. It is not at all obvious that this sequence has no upper bound.

A common example of where monotone sequences arise is the following proposition. The proof is left as an exercise.

**Proposition 2.1.13.** Let $S \subset \mathbb{R}$ be a nonempty bounded set. Then there exist monotone sequences $\{x_n\}$ and $\{y_n\}$ such that $x_n, y_n \in S$ and

$$\sup S = \lim_{n \to \infty} x_n \quad \text{and} \quad \inf S = \lim_{n \to \infty} y_n.$$ 

**2.1.2 Tail of a sequence**

**Definition 2.1.14.** For a sequence $\{x_n\}$, the $K$-tail (where $K \in \mathbb{N}$), or just the tail, of $\{x_n\}$ is the sequence starting at $K + 1$, usually written as $\{x_{n+K}\}_{n=1}^{\infty}$ or $\{x_n\}_{n=K+1}^{\infty}$.

For example, the 4-tail of $\{1/n\}$ is $1/5, 1/6, 1/7, 1/8, \ldots$. The 0-tail of a sequence is the sequence itself. The convergence and the limit of a sequence only depends on its tail.

**Proposition 2.1.15.** Let $\{x_n\}_{n=1}^{\infty}$ be a sequence. Then the following statements are equivalent:

(i) The sequence $\{x_n\}_{n=1}^{\infty}$ converges.

(ii) The K-tail $\{x_{n+K}\}_{n=1}^{\infty}$ converges for all $K \in \mathbb{N}$.

(iii) The K-tail $\{x_{n+K}\}_{n=1}^{\infty}$ converges for some $K \in \mathbb{N}$.

Furthermore, if any (and hence all) of the limits exist, then for all $K \in \mathbb{N}$

$$\lim_{n \to \infty} x_n = \lim_{n \to \infty} x_{n+K}.$$ 

**Proof.** It is clear that (ii) implies (iii). We will therefore show first that (i) implies (ii), and then we will show that (iii) implies (i). That is,

```
    (i) to prove
  (ii) (iii)
```

In the process we will also show that the limits are equal.

We start with (i) implies (ii). Suppose $\{x_n\}$ converges to some $x \in \mathbb{R}$. Let $K \in \mathbb{N}$ be arbitrary, and define $y_n := x_{n+K}$. We wish to show that $\{y_n\}$ converges to $x$. Given an $\varepsilon > 0$, there exists an
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\( M \in \mathbb{N} \) such that \(|x - x_n| < \varepsilon\) for all \( n \geq M \). Note that \( n \geq M \) implies \( n + K \geq M \). Therefore, for all \( n \geq M \), we have

\[ |x - y_n| = |x - x_{n+K}| < \varepsilon. \]

Consequently, \( \{y_n\} \) converges to \( x \).

Let us move to (iii) implies (i). Let \( K \in \mathbb{N} \) be given, define \( y_n := x_{n+K} \), and suppose that \( \{y_n\} \) converges to \( x \in \mathbb{R} \). That is, given an \( \varepsilon > 0 \), there exists an \( M' \in \mathbb{N} \) such that \(|x - y_n| < \varepsilon\) for all \( n \geq M' \). Let \( M := M' + K \). Then \( n \geq M \) implies \( n - K \geq M' \). Thus, whenever \( n \geq M \), we have

\[ |x - x_n| = |x - y_{n-K}| < \varepsilon. \]

Therefore, \( \{x_n\} \) converges to \( x \). \(\square\)

At the end of the day, the limit does not care about how the sequence begins, it only cares about the tail of the sequence. The beginning of the sequence may be arbitrary.

For example, the sequence defined by \( x_n := \frac{n}{n^2 + 16} \) is decreasing if we start at \( n = 4 \) (it is increasing before). That is: \( \{x_n\} = \frac{1}{17}, \frac{1}{10}, \frac{3}{25}, \frac{1}{8}, \frac{5}{41}, \frac{3}{26}, \frac{7}{65}, \frac{1}{10}, \frac{9}{97}, \frac{5}{58}, \ldots \) and

\[ 1/17 < 1/10 < 3/25 < 1/8 < 5/41 > 3/26 > 7/65 > 1/10 > 9/97 > 5/58 > \ldots. \]

If we throw away the first 3 terms and look at the 3-tail, it is decreasing. The proof is left as an exercise. Since the 3-tail is monotone and bounded below by zero, it is convergent, and therefore the sequence is convergent.

### 2.1.3 Subsequences

It is useful to sometimes consider only some terms of a sequence. A subsequence of \( \{x_n\} \) is a sequence that contains only some of the numbers from \( \{x_n\} \) in the same order.

**Definition 2.1.16.** Let \( \{x_n\} \) be a sequence. Let \( \{n_i\} \) be a strictly increasing sequence of natural numbers, that is, \( n_i < n_{i+1} \) for all \( i \) (in other words \( n_1 < n_2 < n_3 < \cdots \)). The sequence

\[ \{x_{n_i}\}_{i=1}^{\infty} \]

is called a *subsequence* of \( \{x_n\} \).

So the subsequence is the sequence \( x_{n_1}, x_{n_2}, x_{n_3}, \ldots \). Consider the sequence \( \{1/n\} \). The sequence \( \{1/3n\} \) is a subsequence. To see how these two sequences fit in the definition, take \( n_i := 3i \). The numbers in the subsequence must come from the original sequence. So \( 1, 0, 1/3, 0, 1/5, \ldots \) is not a subsequence of \( \{1/n\} \). Similarly, order must be preserved. So the sequence \( 1, 1/3, 1/2, 1/5, \ldots \) is not a subsequence of \( \{1/n\} \).

A tail of a sequence is one special type of a subsequence. For an arbitrary subsequence, we have the following proposition about convergence.

**Proposition 2.1.17.** If \( \{x_n\} \) is a convergent sequence, then every subsequence \( \{x_{n_i}\} \) is also convergent, and

\[ \lim_{n \to \infty} x_n = \lim_{i \to \infty} x_{n_i}. \]
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Proof. Suppose \( \lim_{n \to \infty} x_n = x \). So for every \( \varepsilon > 0 \) there is an \( M \in \mathbb{N} \) such that for all \( n \geq M \),

\[
|x_n - x| < \varepsilon.
\]

It is not hard to prove (do it!) by induction that \( n_i \geq i \). Hence \( i \geq M \) implies \( n_i \geq M \). Thus, for all \( i \geq M \),

\[
|x_{n_i} - x| < \varepsilon,
\]

and we are done. \( \square \)

Example 2.1.18: Existence of a convergent subsequence does not imply convergence of the sequence itself. Take the sequence 0, 1, 0, 1, 0, 1, \ldots. That is, \( x_n = 0 \) if \( n \) is odd, and \( x_n = 1 \) if \( n \) is even. The sequence \( \{x_n\} \) is divergent; however, the subsequence \( \{x_{2n}\} \) converges to 1 and the subsequence \( \{x_{2n+1}\} \) converges to 0. Compare Proposition 2.3.7.

2.1.4 Exercises

In the following exercises, feel free to use what you know from calculus to find the limit, if it exists. But you must prove that you found the correct limit, or prove that the series is divergent.

Exercise 2.1.1: Is the sequence \( \{3n\} \) bounded? Prove or disprove.

Exercise 2.1.2: Is the sequence \( \{n\} \) convergent? If so, what is the limit?

Exercise 2.1.3: Is the sequence \( \left\{ \frac{(-1)^n}{2n} \right\} \) convergent? If so, what is the limit?

Exercise 2.1.4: Is the sequence \( \{2^{-n}\} \) convergent? If so, what is the limit?

Exercise 2.1.5: Is the sequence \( \left\{ \frac{n}{n+1} \right\} \) convergent? If so, what is the limit?

Exercise 2.1.6: Is the sequence \( \left\{ \frac{n}{n^2+1} \right\} \) convergent? If so, what is the limit?

Exercise 2.1.7: Let \( \{x_n\} \) be a sequence.

a) Show that \( \lim x_n = 0 \) (that is, the limit exists and is zero) if and only if \( \lim |x_n| = 0 \).

b) Find an example such that \( \{|x_n|\} \) converges and \( \{x_n\} \) diverges.

Exercise 2.1.8: Is the sequence \( \left\{ \frac{2^n}{n!} \right\} \) convergent? If so, what is the limit?

Exercise 2.1.9: Show that the sequence \( \left\{ \frac{1}{\sqrt{n}} \right\} \) is monotone and bounded. Then use Proposition 2.1.10 to find the limit.

Exercise 2.1.10: Show that the sequence \( \left\{ \frac{n+1}{n} \right\} \) is monotone and bounded. Then use Proposition 2.1.10 to find the limit.

Exercise 2.1.11: Finish the proof of Proposition 2.1.10 for monotone decreasing sequences.
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Exercise 2.1.13: Let \( \{x_n\} \) be a convergent monotone sequence. Suppose there exists a \( k \in \mathbb{N} \) such that
\[
\lim_{n \to \infty} x_n = x_k.
\]
Show that \( x_n = x_k \) for all \( n \geq k \).

Exercise 2.1.14: Find a convergent subsequence of the sequence \( \{(-1)^n\} \).

Exercise 2.1.15: Let \( \{x_n\} \) be a sequence defined by
\[
x_n := \begin{cases} n & \text{if } n \text{ is odd}, \\ \frac{1}{n} & \text{if } n \text{ is even}. \end{cases}
\]
a) Is the sequence bounded? (prove or disprove)
b) Is there a convergent subsequence? If so, find it.

Exercise 2.1.16: Let \( \{x_n\} \) be a sequence. Suppose there are two convergent subsequences \( \{x_{n_i}\} \) and \( \{x_{m_i}\} \). Suppose
\[
\lim_{i \to \infty} x_{n_i} = a \quad \text{and} \quad \lim_{i \to \infty} x_{m_i} = b,
\]
where \( a \neq b \). Prove that \( \{x_n\} \) is not convergent, without using Proposition 2.1.17.

Exercise 2.1.17 (Tricky): Find a sequence \( \{x_n\} \) such that for every \( y \in \mathbb{R} \), there exists a subsequence \( \{x_{n_i}\} \) converging to \( y \).

Exercise 2.1.18 (Easy): Let \( \{x_n\} \) be a sequence and \( x \in \mathbb{R} \). Suppose for every \( \varepsilon > 0 \), there is an \( M \) such that
\[
|x_n - x| \leq \varepsilon \quad \text{for all } n \geq M.
\]
Show that \( \lim x_n = x \).

Exercise 2.1.19 (Easy): Let \( \{x_n\} \) be a sequence and \( x \in \mathbb{R} \) such that there exists a \( k \in \mathbb{N} \) such that for all \( n \geq k \), \( x_n = x \). Prove that \( \{x_n\} \) converges to \( x \).

Exercise 2.1.20: Let \( \{x_n\} \) be a sequence and define a sequence \( \{y_n\} \) by \( y_{2k} := x_k \) and \( y_{2k-1} := x_k \) for all \( k \in \mathbb{N} \). Prove that \( \{x_n\} \) converges if and only if \( \{y_n\} \) converges. Furthermore, prove that if they converge, then \( \lim x_n = \lim y_n \).

Exercise 2.1.21: Show that the 3-tail of the sequence defined by \( x_n := \frac{n}{n^2 + 16} \) is monotone decreasing. Hint: Suppose \( n \geq m \geq 4 \) and consider the numerator of the expression \( x_n - x_m \).

Exercise 2.1.22: Suppose that \( \{x_n\} \) is a sequence such that the subsequences \( \{x_{2n}\}, \{x_{2n-1}\}, \text{and } \{x_{3n}\} \) all converge. Show that \( \{x_n\} \) is convergent.

Exercise 2.1.23: Suppose that \( \{x_n\} \) is a monotone increasing sequence that has a convergent subsequence. Show that \( \{x_n\} \) is convergent. Note: So Proposition 2.1.17 is an “if and only if” for monotone sequences.
2.2 Facts about limits of sequences

Note: 2–2.5 lectures, recursively defined sequences can safely be skipped

In this section we go over some basic results about the limits of sequences. We start by looking at how sequences interact with inequalities.

2.2.1 Limits and inequalities

A basic lemma about limits and inequalities is the so-called squeeze lemma. It allows us to show convergence of sequences in difficult cases if we find two other simpler convergent sequences that “squeeze” the original sequence.

**Lemma 2.2.1** (Squeeze lemma). Let \( \{a_n\} \), \( \{b_n\} \), and \( \{x_n\} \) be sequences such that

\[
a_n \leq x_n \leq b_n \quad \text{for all } n \in \mathbb{N}.
\]

Suppose \( \{a_n\} \) and \( \{b_n\} \) converge and

\[
\lim_{n \to \infty} a_n = \lim_{n \to \infty} b_n.
\]

Then \( \{x_n\} \) converges and

\[
\lim_{n \to \infty} x_n = \lim_{n \to \infty} a_n = \lim_{n \to \infty} b_n.
\]

**Proof.** Let \( x := \lim a_n = \lim b_n \). Let \( \varepsilon > 0 \) be given. Find an \( M_1 \) such that for all \( n \geq M_1 \), we have that \( |a_n - x| < \varepsilon \), and an \( M_2 \) such that for all \( n \geq M_2 \), we have \( |b_n - x| < \varepsilon \). Set \( M := \max\{M_1, M_2\} \).

Suppose \( n \geq M \). In particular, \( x - a_n < \varepsilon \), or \( x - \varepsilon < a_n \). Similarly, \( b_n < x + \varepsilon \). Putting everything together, we find

\[
x - \varepsilon < a_n \leq x_n \leq b_n < x + \varepsilon.
\]

In other words, \( -\varepsilon < x_n - x < \varepsilon \) or \( |x_n - x| < \varepsilon \). So \( \{x_n\} \) converges to \( x \). See Figure 2.3. \( \square \)

![Figure 2.3: Squeeze lemma proof in picture.](image)

**Example 2.2.2:** One application of the squeeze lemma is to compute limits of sequences using limits that we already know. For example, consider the sequence \( \{1/n\sqrt{n}\} \). Since \( \sqrt{n} \geq 1 \) for all \( n \in \mathbb{N} \), we have

\[
0 \leq \frac{1}{n\sqrt{n}} \leq \frac{1}{n}
\]

for all \( n \in \mathbb{N} \). We already know \( \lim 1/n = 0 \). Hence, using the constant sequence \( \{0\} \) and the sequence \( \{1/n\} \) in the squeeze lemma, we conclude

\[
\lim_{n \to \infty} \frac{1}{n\sqrt{n}} = 0.
\]
2.2. FACTS ABOUT LIMITS OF SEQUENCES

Limits, when they exist, preserve non-strict inequalities.

**Lemma 2.2.3.** Let \( \{x_n\} \) and \( \{y_n\} \) be convergent sequences and

\[
x_n \leq y_n,
\]

for all \( n \in \mathbb{N} \). Then

\[
\lim_{n \to \infty} x_n \leq \lim_{n \to \infty} y_n.
\]

**Proof.** Let \( x := \lim x_n \) and \( y := \lim y_n \). Let \( \varepsilon > 0 \) be given. Find an \( M_1 \) such that for all \( n \geq M_1 \), we have \( |x_n - x| < \varepsilon/2 \). Find an \( M_2 \) such that for all \( n \geq M_2 \), we have \( |y_n - y| < \varepsilon/2 \). In particular, for some \( n \geq \max \{M_1, M_2\} \), we have \( x - x_n < \varepsilon/2 \) and \( y_n - y < \varepsilon/2 \). We add these inequalities to obtain

\[
y_n - x_n + x - y < \varepsilon, \quad \text{or} \quad y_n - x_n < y - x + \varepsilon.
\]

Since \( x_n \leq y_n \), we have \( 0 \leq y_n - x_n \) and hence \( 0 < y - x + \varepsilon \). In other words,

\[
x - y < \varepsilon.
\]

Because \( \varepsilon > 0 \) was arbitrary, we obtain \( x - y \leq 0 \). Therefore, \( x \leq y \).

The next corollary follows by using constant sequences in Lemma 2.2.3. The proof is left as an exercise.

**Corollary 2.2.4.**

(i) If \( \{x_n\} \) is a convergent sequence such that \( x_n \geq 0 \), then

\[
\lim_{n \to \infty} x_n \geq 0.
\]

(ii) Let \( a, b \in \mathbb{R} \) and let \( \{x_n\} \) be a convergent sequence such that

\[
a \leq x_n \leq b,
\]

for all \( n \in \mathbb{N} \). Then

\[
a \leq \lim_{n \to \infty} x_n \leq b.
\]

In Lemma 2.2.3 and Corollary 2.2.4 we cannot simply replace all the non-strict inequalities with strict inequalities. For example, let \( x_n := -1/n \) and \( y_n := 1/n \). Then \( x_n < y_n, x_n < 0, \) and \( y_n > 0 \) for all \( n \). However, these inequalities are not preserved by the limit operation as \( \lim x_n = \lim y_n = 0 \). The moral of this example is that strict inequalities may become non-strict inequalities when limits are applied; if we know \( x_n < y_n \) for all \( n \), we may only conclude

\[
\lim_{n \to \infty} x_n \leq \lim_{n \to \infty} y_n,
\]

This issue is a common source of errors.
2.2.2 Continuity of algebraic operations

Limits interact nicely with algebraic operations.

**Proposition 2.2.5.** Let \( \{x_n\} \) and \( \{y_n\} \) be convergent sequences.

(i) The sequence \( \{z_n\} \), where \( z_n := x_n + y_n \), converges and
\[
\lim_{n \to \infty} (x_n + y_n) = \lim_{n \to \infty} z_n = \lim_{n \to \infty} x_n + \lim_{n \to \infty} y_n.
\]

(ii) The sequence \( \{z_n\} \), where \( z_n := x_n - y_n \), converges and
\[
\lim_{n \to \infty} (x_n - y_n) = \lim_{n \to \infty} z_n = \lim_{n \to \infty} x_n - \lim_{n \to \infty} y_n.
\]

(iii) The sequence \( \{z_n\} \), where \( z_n := x_n y_n \), converges and
\[
\lim_{n \to \infty} (x_n y_n) = \lim_{n \to \infty} z_n = \left( \lim_{n \to \infty} x_n \right) \left( \lim_{n \to \infty} y_n \right).
\]

(iv) If \( \lim y_n \neq 0 \) and \( y_n \neq 0 \) for all \( n \in \mathbb{N} \), then the sequence \( \{z_n\} \), where \( z_n := \frac{x_n}{y_n} \), converges and
\[
\lim_{n \to \infty} \frac{x_n}{y_n} = \lim_{n \to \infty} z_n = \frac{\lim_{n \to \infty} x_n}{\lim_{n \to \infty} y_n}.
\]

**Proof.** We start with (i). Suppose \( \{x_n\} \) and \( \{y_n\} \) are convergent sequences and write \( z_n := x_n + y_n \).

Let \( \varepsilon > 0 \) be given. Find an \( M_1 \) such that for all \( n \geq M_1 \), we have \( |x_n - x| < \varepsilon/2 \). Find an \( M_2 \) such that for all \( n \geq M_2 \), we have \( |y_n - y| < \varepsilon/2 \). Take \( M := \max\{M_1, M_2\} \). For all \( n \geq M \), we have
\[
|z_n - z| = |(x_n + y_n) - (x + y)|
= |x_n - x + y_n - y|
\leq |x_n - x| + |y_n - y|
< \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon.
\]

Therefore (i) is proved. Proof of (ii) is almost identical and is left as an exercise.

Let us tackle (iii). Suppose again that \( \{x_n\} \) and \( \{y_n\} \) are convergent sequences and write \( z_n := x_n y_n \).

Let \( \varepsilon > 0 \) be given. Let \( K := \max\{|x|, |y|, \varepsilon/3, 1\} \). Find an \( M_1 \) such that for all \( n \geq M_1 \), we have \( |x_n - x| < \frac{\varepsilon}{3K} \). Find an \( M_2 \) such that for all \( n \geq M_2 \), we have \( |y_n - y| < \frac{\varepsilon}{3K} \). Take \( M := \max\{M_1, M_2\} \).

For all \( n \geq M \), we have
\[
|z_n - z| = |(x_n y_n) - (xy)|
= |(x_n - x + x)(y_n - y + y) - xy|
= |(x_n - x)y + x(y_n - y) + (x_n - x)(y_n - y)|
\leq |(x_n - x)y| + |x(y_n - y)| + |(x_n - x)(y_n - y)|
= |x_n - x||y| + |x||y_n - y| + |x_n - x||y_n - y|
< \frac{\varepsilon}{3K} K + K \frac{\varepsilon}{3K} + \frac{\varepsilon}{3K} \frac{\varepsilon}{3K}
\quad (\text{now notice that } \frac{\varepsilon}{3K} \leq 1 \text{ and } K \geq 1)
\leq \frac{\varepsilon}{3} + \frac{\varepsilon}{3} + \frac{\varepsilon}{3} = \varepsilon.
\]
Finally, we examine (iv). Instead of proving (iv) directly, we prove the following simpler claim:

Claim: If \( \{y_n\} \) is a convergent sequence such that \( \lim y_n \neq 0 \) and \( y_n \neq 0 \) for all \( n \in \mathbb{N} \), then \( \{1/y_n\} \) converges and

\[
\lim_{n \to \infty} \frac{1}{y_n} = \frac{1}{\lim y_n}.
\]

Once the claim is proved, we take the sequence \( \{1/y_n\} \), multiply it by the sequence \( \{x_n\} \) and apply item (iii).

Proof of claim: Let \( \varepsilon > 0 \) be given. Let \( y := \lim y_n \). As \( |y| \neq 0 \), then \( \min\left\{|y|^2 \frac{\varepsilon}{2}, \frac{|y|}{2}\right\} > 0 \). Find an \( M \) such that for all \( n \geq M \), we have

\[
|y_n - y| < \min\left\{|y|^2 \frac{\varepsilon}{2}, \frac{|y|}{2}\right\}.
\]

For all \( n \geq M \), we have \( |y - y_n| < |y|/2 \), and so

\[
|y| = |y_n - y_n + y_n| \leq |y - y_n| + |y_n| < \frac{|y|}{2} + |y_n|.
\]

Subtracting \(|y|/2\) from both sides we obtain \(|y|/2 < |y_n|\), or in other words,

\[
\frac{1}{|y_n|} < \frac{2}{|y|}.
\]

We finish the proof of the claim:

\[
\left|\frac{1}{y_n} - \frac{1}{y}\right| = \left|\frac{y - y_n}{y y_n}\right| = \frac{|y - y_n|}{|y| |y_n|} \leq \frac{|y - y_n|}{|y|} \frac{2}{|y|} \leq \frac{|y|^2 \frac{\varepsilon}{2}}{|y|} = \varepsilon.
\]

And we are done.

By plugging in constant sequences, we get several easy corollaries. If \( c \in \mathbb{R} \) and \( \{x_n\} \) is a convergent sequence, then for example

\[
\lim_{n \to \infty} cx_n = c \left(\lim_{n \to \infty} x_n\right) \quad \text{and} \quad \lim_{n \to \infty} (c + x_n) = c + \lim_{n \to \infty} x_n.
\]

Similarly, we find such equalities for constant subtraction and division.

As we can take limits past multiplication we can show (exercise) that \( \lim x_n^k = (\lim x_n)^k \) for all \( k \in \mathbb{N} \). That is, we can take limits past powers. Let us see if we can do the same with roots.
Proposition 2.2.6. Let \( \{x_n\} \) be a convergent sequence such that \( x_n \geq 0 \). Then
\[
\lim_{n \to \infty} \sqrt{x_n} = \sqrt{\lim_{n \to \infty} x_n}.
\]

Of course, to even make this statement, we need to apply Corollary 2.2.4 to show that \( \lim x_n \geq 0 \), so that we can take the square root without worry.

**Proof.** Let \( \{x_n\} \) be a convergent sequence and let \( x := \lim x_n \). As we just mentioned, \( x \geq 0 \).

First suppose \( x = 0 \). Let \( \varepsilon > 0 \) be given. Then there is an \( M \) such that for all \( n \geq M \), we have \( x_n = |x_n| < \varepsilon^2 \), or in other words, \( \sqrt{x_n} < \varepsilon \). Hence,
\[
\left| \sqrt{x_n} - \sqrt{x} \right| = \sqrt{x_n} < \varepsilon.
\]

Now suppose \( x > 0 \) (and hence \( \sqrt{x} > 0 \)).
\[
\left| \sqrt{x_n} - \sqrt{x} \right| = \left| \frac{x_n - x}{\sqrt{x_n} + \sqrt{x}} \right| = \frac{1}{\sqrt{x_n} + \sqrt{x}} |x_n - x| \leq \frac{1}{\sqrt{x}} |x_n - x|.
\]

We leave the rest of the proof to the reader.

A similar proof works for the \( k \)th root. That is, we also obtain \( \lim x_n^{1/k} = (\lim x_n)^{1/k} \). We leave this to the reader as a challenging exercise.

We may also want to take the limit past the absolute value sign. The converse of this proposition is not true, see Exercise 2.1.7 part b).

Proposition 2.2.7. If \( \{x_n\} \) is a convergent sequence, then \( \{|x_n|\} \) is convergent and
\[
\lim_{n \to \infty} |x_n| = \left| \lim_{n \to \infty} x_n \right|.
\]

**Proof.** We simply note the reverse triangle inequality
\[
\left| |x_n| - |x| \right| \leq |x_n - x|.
\]
Hence if \( |x_n - x| \) can be made arbitrarily small, so can \( |x_n| - |x| \). Details are left to the reader.

Let us see an example putting the propositions above together. Since \( \lim 1/n = 0 \), then
\[
\lim_{n \to \infty} \sqrt{1 + 1/n - 100/n^2} = \sqrt{1 + (\lim 1/n) - 100(\lim 1/n)(\lim 1/n)} = 1.
\]
That is, the limit on the left-hand side exists because the right-hand side exists. You really should read the equality above from right to left.

On the other hand you must apply the propositions carefully. For example, by rewriting the expression with common denominator first we find
\[
\lim_{n \to \infty} \left( \frac{n^2}{n + 1} - n \right) = -1.
\]
However, \( \{n^2/(n+1)\} \) and \( \{n\} \) are not convergent, so \( (\lim n^2/(n+1)) - (\lim n) \) is nonsense.
2.2.3 Recursively defined sequences

Now that we know we can interchange limits and algebraic operations, we can compute the limits of many sequences. One such class are recursively defined sequences, that is, sequences where the next number in the sequence is computed using a formula from a fixed number of preceding elements in the sequence.

Example 2.2.8: Let \( \{x_n\} \) be defined by \( x_1 := 2 \) and

\[
x_{n+1} := x_n - \frac{x_n^2 - 2}{2x_n}.
\]

We must first find out if this sequence is well-defined; we must show we never divide by zero. Then we must find out if the sequence converges. Only then can we attempt to find the limit.

So let us prove \( x_n \) exists and \( x_n > 0 \) for all \( n \) (so the sequence is well-defined and bounded below). Let us show this by induction. We know that \( x_1 = 2 > 0 \). For the induction step, suppose \( x_n > 0 \). Then

\[
x_{n+1} = x_n - \frac{x_n^2 - 2}{2x_n} = \frac{2x_n^2 - x_n^2 + 2}{2x_n} = \frac{x_n^2 + 2}{2x_n}.
\]

It is always true that \( x_n^2 + 2 > 0 \), and as \( x_n > 0 \), then \( \frac{x_n^2 + 2}{2x_n} > 0 \) and hence \( x_{n+1} > 0 \).

Next let us show that the sequence is monotone decreasing. If we show that \( x_{n+1}^2 - 2 \geq 0 \) for all \( n \), then \( x_{n+1} \leq x_n \) for all \( n \). Obviously \( x_1^2 - 2 = 4 - 2 = 2 > 0 \). For an arbitrary \( n \), we have

\[
x_{n+1}^2 - 2 = \left( \frac{x_n^2 + 2}{2x_n} \right)^2 - 2 = \frac{4x_n^4 + 4x_n^2 + 4 - 8x_n^2}{4x_n^2} = \frac{x_n^4 - 4x_n^2 + 4}{4x_n^2} = \frac{(x_n^2 - 2)^2}{4x_n^2}.
\]

Since squares are nonnegative, \( x_{n+1}^2 - 2 \geq 0 \) for all \( n \). Therefore, \( \{x_n\} \) is monotone decreasing and bounded (\( x_n > 0 \) for all \( n \)), and so the limit exists. It remains to find the limit.

Write

\[
2x_n x_{n+1} = x_n^2 + 2.
\]

Since \( \{x_{n+1}\} \) is the 1-tail of \( \{x_n\} \), it converges to the same limit. Let us define \( x := \lim x_n \). Take the limit of both sides to obtain

\[
2x^2 = x^2 + 2,
\]

or \( x^2 = 2 \). As \( x_n > 0 \) for all \( n \) we get \( x \geq 0 \), and therefore \( x = \sqrt{2} \).

You may have seen the sequence above before. It is Newton’s method* for finding the square root of 2. This method comes up often in practice and converges very rapidly. We used the fact that \( x_1^2 - 2 > 0 \), although it was not strictly needed to show convergence by considering a tail of the sequence. The sequence converges as long as \( x_1 \neq 0 \), although with a negative \( x_1 \) we would arrive at \( x = -\sqrt{2} \). By replacing the 2 in the numerator we obtain the square root of any positive number. These statements are left as an exercise.

You should, however, be careful. Before taking any limits, you must make sure the sequence converges. Let us see an example.

---

*Named after the English physicist and mathematician Isaac Newton (1642–1726/7).
**Example 2.2.9:** Suppose \( x_1 := 1 \) and \( x_{n+1} := x_n^2 + x_n \). If we blindly assumed that the limit exists (call it \( x \)), then we would get the equation \( x = x^2 + x \), from which we might conclude \( x = 0 \). However, it is not hard to show that \( \{x_n\} \) is unbounded and therefore does not converge.

The thing to notice in this example is that the method still works, but it depends on the initial value \( x_1 \). If we set \( x_1 := 0 \), then the sequence converges and the limit really is 0. An entire branch of mathematics, called dynamics, deals precisely with these issues. See Exercise 2.2.14.

**2.2.4 Some convergence tests**

It is not always necessary to go back to the definition of convergence to prove that a sequence is convergent. We first give a simple convergence test. The main idea is that \( \{x_n\} \) converges to \( x \) if and only if \( \{|x_n - x|\} \) converges to zero.

**Proposition 2.2.10.** Let \( \{x_n\} \) be a sequence. Suppose there is an \( x \in \mathbb{R} \) and a convergent sequence \( \{a_n\} \) such that

\[
\lim_{n \to \infty} a_n = 0
\]

and

\[
|x_n - x| \leq a_n \quad \text{for all } n \in \mathbb{N}.
\]

Then \( \{x_n\} \) converges and \( \lim x_n = x \).

**Proof.** Let \( \varepsilon > 0 \) be given. Note that \( a_n \geq 0 \) for all \( n \). Find an \( M \in \mathbb{N} \) such that for all \( n \geq M \), we have \( a_n = |a_n - 0| < \varepsilon \). Then, for all \( n \geq M \), we have

\[
|x_n - x| \leq a_n < \varepsilon.
\]

As the proposition shows, to study when a sequence has a limit is the same as studying when another sequence goes to zero. In general, it may be hard to decide if a sequence converges, but for certain sequences there exist easy to apply tests that tell us if the sequence converges or not. Let us see one such test. First, let us compute the limit of a certain specific sequence.

**Proposition 2.2.11.** Let \( c > 0 \).

(i) If \( c < 1 \), then

\[
\lim_{n \to \infty} c^n = 0.
\]

(ii) If \( c > 1 \), then \( \{c^n\} \) is unbounded.

**Proof.** First consider \( c < 1 \). As \( c > 0 \), then \( c^n > 0 \) for all \( n \in \mathbb{N} \) by induction. As \( c < 1 \), then \( c^{n+1} < c^n \) for all \( n \). So \( \{c^n\} \) is a decreasing sequence that is bounded below. Hence, it is convergent. Let \( L := \lim c^n \). The 1-tail \( \{c^{n+1}\} \) also converges to \( L \). Taking the limit of both sides of \( c^{n+1} = c \cdot c^n \), we obtain \( L = cL \), or \( (1 - c)L = 0 \). It follows that \( L = 0 \) as \( c \neq 1 \).

Now consider \( c > 1 \). Let \( B > 0 \) be arbitrary. As \( 1/c < 1 \), then \( \{(1/c)^n\} \) converges to 0. Hence for some large enough \( n \), we get

\[
\frac{1}{c^n} = \left(\frac{1}{c}\right)^n < \frac{1}{B}.
\]

In other words, \( c^n > B \), and \( B \) is not an upper bound for \( \{c^n\} \). As \( B \) was arbitrary, \( \{c^n\} \) is unbounded. \( \square \)
In the proposition above, the ratio of the \((n+1)\)th term and the \(n\)th term is \(c\). We generalize this simple result to a larger class of sequences. The following lemma will come up again once we get to series.

**Lemma 2.2.12** (Ratio test for sequences). Let \(\{x_n\}\) be a sequence such that \(x_n \neq 0\) for all \(n\) and such that the limit

\[
L := \lim_{n \to \infty} \frac{|x_{n+1}|}{|x_n|}
\]

exists.

(i) If \(L < 1\), then \(\{x_n\}\) converges and \(\lim x_n = 0\).

(ii) If \(L > 1\), then \(\{x_n\}\) is unbounded (hence diverges).

If \(L\) exists, but \(L = 1\), the lemma says nothing. We cannot make any conclusion based on that information alone. For example, the sequence \(\{1/n\}\) converges to zero, but \(L = 1\). The constant sequence \(\{1\}\) converges to 1, not zero, and \(L = 1\) as well. Finally, the sequence \(\{n\}\) is unbounded, yet again \(L = 1\). The statement of the lemma may be strengthened somewhat, see exercises 2.2.13 and 2.3.15.

**Proof.** Suppose \(L < 1\). As \(\frac{|x_{n+1}|}{|x_n|} \geq 0\) for all \(n\), then \(L \geq 0\). Pick \(r\) such that \(L < r < 1\). We wish to compare the sequence \(\{x_n\}\) to the sequence \(\{r^n\}\). The idea is that while the ratio \(\frac{|x_{n+1}|}{|x_n|}\) is not going to be less than \(L\) eventually, it will eventually be less than \(r\), which is still less than 1. The intuitive idea of the proof is illustrated in Figure 2.4.

![Figure 2.4: Proof of ratio test in picture. The short lines represent the ratios \(\frac{|x_{n+1}|}{|x_n|}\) approaching \(L\).](image)

As \(r - L > 0\), there exists an \(M \in \mathbb{N}\) such that for all \(n \geq M\), we have

\[
\left| \frac{|x_{n+1}|}{|x_n|} - L \right| < r - L.
\]

Therefore, for \(n \geq M\),

\[
\frac{|x_{n+1}|}{|x_n|} - L < r - L \quad \text{or} \quad \frac{|x_{n+1}|}{|x_n|} < r.
\]

For \(n > M\) (that is for \(n \geq M + 1\)) write

\[
|x_n| = |x_M| \frac{|x_{M+1}|}{|x_M|} \frac{|x_{M+2}|}{|x_{M+1}|} \ldots \frac{|x_n|}{|x_{n-1}|} < |x_M| r r \ldots r = |x_M| r^{n-M} = (|x_M| r^{-M}) r^n.
\]

The sequence \(\{r^n\}\) converges to zero and hence \(|x_M| r^{-M} r^n\) converges to zero. By Proposition 2.2.10, the \(M\)-tail of \(\{x_n\}\) converges to zero and therefore \(\{x_n\}\) converges to zero.

Now suppose \(L > 1\). Pick \(r\) such that \(1 < r < L\). As \(L - r > 0\), there exists an \(M \in \mathbb{N}\) such that for all \(n \geq M\)

\[
\left| \frac{|x_{n+1}|}{|x_n|} - L \right| < L - r.
\]
Therefore, 
\[ \frac{|x_{n+1}|}{|x_n|} > r. \]
Again for \( n > M \), write
\[ |x_n| = |x_M| \frac{|x_{M+1}|}{|x_M|} \frac{|x_{M+2}|}{|x_{M+1}|} \cdots \frac{|x_n|}{|x_{n-1}|} > |x_M| r r \cdots r = |x_M| r^{n-M} = (|x_M| r^{-M}) r^n. \]
The sequence \( \{r^n\} \) is unbounded (since \( r > 1 \)), and so \( \{x_n\} \) cannot be bounded (if \( |x_n| \leq B \) for all \( n \), then \( r^n < \frac{B}{|x_M|} r^M \) for all \( n > M \), which is impossible). Consequently, \( \{x_n\} \) cannot converge. \( \square \)

**Example 2.2.13:** A simple application of the lemma above is to prove
\[ \lim_{n \to \infty} \frac{2^n}{n!} = 0. \]

Proof: Compute
\[ \frac{2^{n+1}/(n+1)!}{2^n/n!} = \frac{2^{n+1}}{2^n} \frac{n!}{(n+1)!} = \frac{2}{n+1}. \]
It is not hard to see that \( \left\{ \frac{2}{n+1} \right\} \) converges to zero. The conclusion follows by the lemma.

**Example 2.2.14:** A more complicated (and useful) application of the ratio test is to prove
\[ \lim_{n \to \infty} n^{1/n} = 1. \]

Proof: Let \( \varepsilon > 0 \) be given. Consider the sequence \( \left\{ \frac{n}{(1+\varepsilon)^n} \right\} \). Compute
\[ \frac{(n+1)/(1+\varepsilon)^{n+1}}{n/(1+\varepsilon)^n} = \frac{n+1}{n} \frac{1}{1+\varepsilon}. \]
The limit of \( \frac{n+1}{n} = 1 + \frac{1}{n} \) as \( n \to \infty \) is 1, and so
\[ \lim_{n \to \infty} \frac{(n+1)/(1+\varepsilon)^{n+1}}{n/(1+\varepsilon)^n} = \frac{1}{1+\varepsilon} < 1. \]
Therefore, \( \left\{ \frac{n}{(1+\varepsilon)^n} \right\} \) converges to 0. In particular, there exists an \( M \) such that for \( n \geq M \), we have \( \frac{n}{(1+\varepsilon)^n} < 1 \), or \( n < (1+\varepsilon)^n \), or \( n^{1/n} < 1 + \varepsilon \). As \( n \geq 1 \), then \( n^{1/n} \geq 1 \), and so \( 0 \leq n^{1/n} - 1 < \varepsilon \). Consequently, \( \lim n^{1/n} = 1 \).

### 2.2.5 Exercises

**Exercise 2.2.1:** Prove Corollary 2.2.4. Hint: Use constant sequences and Lemma 2.2.3.

**Exercise 2.2.2:** Prove part (ii) of Proposition 2.2.5.
Exercise 2.2.3: Prove that if \( \{x_n\} \) is a convergent sequence, \( k \in \mathbb{N} \), then
\[
\lim_{n \to \infty} x_n^k = \left( \lim_{n \to \infty} x_n \right)^k.
\]

Hint: Use induction.

Exercise 2.2.4: Suppose \( x_1 := \frac{1}{2} \) and \( x_{n+1} := x_n^2 \). Show that \( \{x_n\} \) converges and find \( \lim x_n \). Hint: You cannot divide by zero!

Exercise 2.2.5: Let \( x_n := \frac{n - \cos(n)}{n} \). Use the squeeze lemma to show that \( \{x_n\} \) converges and find the limit.

Exercise 2.2.6: Let \( x_n := \frac{1}{n^2} \) and \( y_n := \frac{1}{n} \). Define \( z_n := \frac{x_n}{y_n} \) and \( w_n := \frac{y_n}{x_n} \). Do \( \{z_n\} \) and \( \{w_n\} \) converge? What are the limits? Can you apply Proposition 2.2.5? Why or why not?

Exercise 2.2.7: True or false, prove or find a counterexample. If \( \{x_n\} \) is a sequence such that \( \{x_{2n}\} \) converges, then \( \{x_n\} \) converges.

Exercise 2.2.8: Show that
\[
\lim_{n \to \infty} \frac{n^2}{2^n} = 0.
\]

Exercise 2.2.9: Suppose \( \{x_n\} \) is a sequence and suppose for some \( x \in \mathbb{R} \), the limit
\[
L := \lim_{n \to \infty} \frac{|x_{n+1} - x|}{|x_n - x|}
\]
eexists and \( L < 1 \). Show that \( \{x_n\} \) converges to \( x \).

Exercise 2.2.10 (Challenging): Let \( \{x_n\} \) be a convergent sequence such that \( x_n \geq 0 \) and \( k \in \mathbb{N} \). Then
\[
\lim_{n \to \infty} x_n^{1/k} = \left( \lim_{n \to \infty} x_n \right)^{1/k}.
\]

Hint: Find an expression \( q \) such that \( \frac{x_n^{1/k} - 1}{x_n - 1} = \frac{1}{q} \).

Exercise 2.2.11: Let \( r > 0 \). Show that starting with an arbitrary \( x_1 \neq 0 \), the sequence defined by
\[
x_{n+1} := x_n - \frac{x_n^2 - r}{2x_n}
\]
converges to \( \sqrt{r} \) if \( x_1 > 0 \) and \( -\sqrt{r} \) if \( x_1 < 0 \).

Exercise 2.2.12:

a) Suppose \( \{a_n\} \) is a bounded sequence and \( \{b_n\} \) is a sequence converging to 0. Show that \( \{a_n b_n\} \) converges to 0.

b) Find an example where \( \{a_n\} \) is unbounded, \( \{b_n\} \) converges to 0, and \( \{a_n b_n\} \) is not convergent.

c) Find an example where \( \{a_n\} \) is bounded, \( \{b_n\} \) converges to some \( x \neq 0 \), and \( \{a_n b_n\} \) is not convergent.
**Exercise 2.2.13 (Easy):** Prove the following stronger version of Lemma 2.2.12, the ratio test. Suppose \( \{x_n\} \) is a sequence such that \( x_n \neq 0 \) for all \( n \).

a) Prove that if there exists an \( r < 1 \) and \( M \in \mathbb{N} \) such that for all \( n \geq M \), we have

\[
\left| \frac{x_{n+1}}{x_n} \right| \leq r,
\]

then \( \{x_n\} \) converges to 0.

b) Prove that if there exists an \( r > 1 \) and \( M \in \mathbb{N} \) such that for all \( n \geq M \), we have

\[
\left| \frac{x_{n+1}}{x_n} \right| \geq r,
\]

then \( \{x_n\} \) is unbounded.

**Exercise 2.2.14:** Suppose \( x_1 := c \) and \( x_{n+1} := x_n^2 + x_n \). Show that \( \{x_n\} \) converges if and only if \( -1 \leq c \leq 0 \), in which case it converges to 0.

**Exercise 2.2.15:** Prove \( \lim_{n \to \infty} (n^2 + 1)^{1/n} = 1 \).

**Exercise 2.2.16:** Prove that \( \{(n!)^{1/n}\} \) is unbounded. Hint: Show that \( \{\frac{C}{n!}\} \) converges to zero for all \( C > 0 \).
2.3 Limit superior, limit inferior, and Bolzano–Weierstrass

Note: 1–2 lectures, alternative proof of BW optional

In this section we study bounded sequences and their subsequences. In particular, we define the so-called limit superior and limit inferior of a bounded sequence and talk about limits of subsequences. Furthermore, we prove the Bolzano–Weierstrass theorem*, an indispensable tool in analysis, showing the existence of convergent subsequences.

We have seen that every convergent sequence is bounded, although there exist many bounded divergent sequences. For example, the sequence \( \{(-1)^n\} \) is bounded, but divergent. All is not lost, however, and we can still compute certain limits with a bounded divergent sequence.

2.3.1 Upper and lower limits

There are ways of creating monotone sequences out of any sequence, and in this fashion we get the so-called limit superior and limit inferior. These limits always exist for bounded sequences.

If a sequence \( \{x_n\} \) is bounded, then the set \( \{x_k : k \in \mathbb{N}\} \) is bounded. For every \( n \), the set \( \{x_k : k \geq n\} \) is also bounded (as it is a subset), so we take its supremum and infimum.

**Definition 2.3.1.** Let \( \{x_n\} \) be a bounded sequence. Define the sequences \( \{a_n\} \) and \( \{b_n\} \) by

\[
    a_n := \sup\{x_k : k \geq n\} \quad \text{and} \quad b_n := \inf\{x_k : k \geq n\}.
\]

Define, if the limits exist,

\[
    \limsup_{n \to \infty} x_n := \lim_{n \to \infty} a_n, \quad \liminf_{n \to \infty} x_n := \lim_{n \to \infty} b_n.
\]

For a bounded sequence, liminf and limsup always exist (see below). It is possible to define liminf and limsup for unbounded sequences if we allow \( \infty \) and \( -\infty \), and we do so later in this section. It is not hard to generalize the following results to include unbounded sequences; however, we first restrict our attention to bounded ones.

**Proposition 2.3.2.** Let \( \{x_n\} \) be a bounded sequence. Let \( a_n \) and \( b_n \) be as in the definition above.

(i) The sequence \( \{a_n\} \) is bounded monotone decreasing and \( \{b_n\} \) is bounded monotone increasing. In particular, \( \liminf_{n \to \infty} x_n \) and \( \limsup_{n \to \infty} x_n \) exist.

(ii) \( \limsup_{n \to \infty} x_n = \inf\{a_n : n \in \mathbb{N}\} \) and \( \liminf_{n \to \infty} x_n = \sup\{b_n : n \in \mathbb{N}\} \).

(iii) \( \liminf_{n \to \infty} x_n \leq \limsup_{n \to \infty} x_n \).

**Proof.** Let us see why \( \{a_n\} \) is a decreasing sequence. As \( a_n \) is the least upper bound for \( \{x_k : k \geq n\} \), it is also an upper bound for the subset \( \{x_k : k \geq (n+1)\} \). Therefore \( a_{n+1} \), the least upper bound for \( \{x_k : k \geq (n+1)\} \), has to be less than or equal to \( a_n \), the least upper bound for \( \{x_k : k \geq n\} \). That is, \( a_n \geq a_{n+1} \) for all \( n \). Similarly (an exercise), \( \{b_n\} \) is an increasing sequence. It is left as an exercise to show that if \( \{x_n\} \) is bounded, then \( \{a_n\} \) and \( \{b_n\} \) must be bounded.

The second item follows as the sequences \( \{a_n\} \) and \( \{b_n\} \) are monotone and bounded.

*Named after the Czech mathematician Bernhard Placidus Johann Nepomuk Bolzano (1781–1848), and the German mathematician Karl Theodor Wilhelm Weierstrass (1815–1897).
For the third item, note that \( b_n \leq a_n \), as the inf of a nonempty set is less than or equal to its sup. The sequences \( \{a_n\} \) and \( \{b_n\} \) converge to the limsup and the liminf respectively. Apply Lemma 2.2.3 to obtain

\[
\lim_{n \to \infty} b_n \leq \lim_{n \to \infty} a_n.
\]

\[\limsup_{n \to \infty} x_n \]

\[\liminf_{n \to \infty} x_n \]

Figure 2.5: First 50 terms of an example sequence. Terms \( x_n \) of the sequence are marked with dots (•),\( a_n \) are marked with circles (○), and \( b_n \) are marked with diamonds (⋄).

Example 2.3.3: Let \( \{x_n\} \) be defined by

\[
x_n := \begin{cases} 
\frac{n+1}{n} & \text{if } n \text{ is odd}, \\
0 & \text{if } n \text{ is even}.
\end{cases}
\]

Let us compute the lim inf and lim sup of this sequence. See also Figure 2.6. First the limit inferior:

\[
\liminf_{n \to \infty} x_n = \lim_{n \to \infty} \left( \inf \{x_k : k \geq n\} \right) = \lim_{n \to \infty} 0 = 0.
\]

For the limit superior, we write

\[
\limsup_{n \to \infty} x_n = \lim_{n \to \infty} \left( \sup \{x_k : k \geq n\} \right).
\]

It is not hard to see that

\[
\sup \{x_k : k \geq n\} = \begin{cases} 
\frac{n+1}{n} & \text{if } n \text{ is odd}, \\
\frac{n+2}{n+1} & \text{if } n \text{ is even}.
\end{cases}
\]

We leave it to the reader to show that the limit is 1. That is,

\[
\lim_{n \to \infty} x_n = 1.
\]

Do note that the sequence \( \{x_n\} \) is not a convergent sequence.
2.3. LIMIT SUPERIOR, LIMIT INFERIOR, AND BOLZANO–WEIERSTRASS

2.3. LIMIT SUPERIOR, LIMIT INFERIOR, AND BOLZANO–WEIERSTRASS

We associate certain subsequences with limsup and liminf. It is important to notice that \( \{a_n\} \) and \( \{b_n\} \) are not necessarily subsequences of \( \{x_n\} \), nor do they have to even consist of the same numbers. For example, for the sequence \( \{1/n\} \), \( b_n = 0 \) for all \( n \in \mathbb{N} \).

**Theorem 2.3.4.** If \( \{x_n\} \) is a bounded sequence, then there exists a subsequence \( \{x_{n_k}\} \) such that

\[
\lim_{k \to \infty} x_{n_k} = \limsup_{n \to \infty} x_n.
\]

Similarly, there exists a (perhaps different) subsequence \( \{x_{m_k}\} \) such that

\[
\lim_{k \to \infty} x_{m_k} = \liminf_{n \to \infty} x_n.
\]

**Proof.** Define \( a_n := \sup \{x_k : k \geq n\} \). Write \( x := \limsup x_n = \lim a_n \). We define the subsequence inductively. Let \( n_1 := 1 \) and suppose we have defined the subsequence until \( n_k \) for some \( k \). Pick some \( m > n_k \) such that

\[
a_{(n_k + 1)} - x_m < \frac{1}{k+1}.
\]

We can do this as \( a_{(n_k + 1)} \) is a supremum of the set \( \{x_n : n \geq n_k + 1\} \) and hence there are elements of the sequence arbitrarily close (or even possibly equal) to the supremum. Set \( n_{k+1} := m \). The subsequence \( \{x_{n_k}\} \) is defined. Next we need to prove that it converges and has the right limit.

For all \( k \geq 2 \), we have \( a_{(n_{k+1})} \geq a_{n_k} \) (why?) and \( a_{n_k} \geq x_{n_k} \). Therefore, for every \( k \geq 2 \),

\[
|a_{n_k} - x_{n_k}| = a_{n_k} - x_{n_k} \leq a_{(n_{k-1} + 1)} - x_{n_k} < \frac{1}{k+1}.
\]

Let us show that \( \{x_{n_k}\} \) converges to \( x \). Note that the subsequence need not be monotone. Let \( \varepsilon > 0 \) be given. As \( \{a_n\} \) converges to \( x \), the subsequence \( \{a_{n_k}\} \) converges to \( x \). Thus there exists an
$M_1 \in \mathbb{N}$ such that for all $k \geq M_1$, we have

$$|a_{n_k} - x| < \frac{\varepsilon}{2}.$$ 

Find an $M_2 \in \mathbb{N}$ such that

$$\frac{1}{M_2} \leq \frac{\varepsilon}{2}.$$ 

Take $M := \max\{M_1, M_2, 2\}$ and compute. For all $k \geq M$, we have

$$|x - x_{n_k}| = |a_{n_k} - x_{n_k} + x - a_{n_k}|$$

$$\leq |a_{n_k} - x_{n_k}| + |x - a_{n_k}|$$

$$\leq \frac{1}{k} + \varepsilon$$

$$\leq \frac{1}{M_2} + \frac{\varepsilon}{2} \leq \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon.$$ 

We leave the statement for $\liminf$ as an exercise. \qed

### 2.3.2 Using limit inferior and limit superior

The advantage of $\liminf$ and $\limsup$ is that we can always write them down for any (bounded) sequence. If we could somehow compute them, we could also compute the limit of the sequence if it exists, or show that the sequence diverges. Working with $\liminf$ and $\limsup$ is a little bit like working with limits, although there are subtle differences.

**Proposition 2.3.5.** Let $\{x_n\}$ be a bounded sequence. Then $\{x_n\}$ converges if and only if

$$\liminf_{n \to \infty} x_n = \limsup_{n \to \infty} x_n.$$ 

Furthermore, if $\{x_n\}$ converges, then

$$\lim_{n \to \infty} x_n = \liminf_{n \to \infty} x_n = \limsup_{n \to \infty} x_n.$$ 

**Proof.** Let $a_n$ and $b_n$ be as in Definition 2.3.1. In particular, for all $n \in \mathbb{N}$,

$$b_n \leq x_n \leq a_n.$$ 

If $\liminf x_n = \limsup x_n$, then we know that $\{a_n\}$ and $\{b_n\}$ both converge to the same limit. By the squeeze lemma (Lemma 2.2.1), $\{x_n\}$ converges and

$$\lim_{n \to \infty} b_n = \lim_{n \to \infty} x_n = \lim_{n \to \infty} a_n.$$ 

Now suppose $\{x_n\}$ converges to $x$. By Theorem 2.3.4, there exists a subsequence $\{x_{n_k}\}$ that converges to $\limsup x_n$. As $\{x_n\}$ converges to $x$, every subsequence converges to $x$ and therefore $\limsup x_n = \lim x_{n_k} = x$. Similarly, $\liminf x_n = x$. \qed
Limit superior and limit inferior behave nicely with subsequences.

**Proposition 2.3.6.** Suppose \( \{x_n\} \) is a bounded sequence and \( \{x_{n_k}\} \) is a subsequence. Then
\[
\liminf_{n \to \infty} x_n \leq \liminf_{k \to \infty} x_{n_k} \leq \limsup_{k \to \infty} x_{n_k} \leq \limsup_{n \to \infty} x_n.
\]

**Proof.** The middle inequality has been proved already. We will prove the third inequality, and leave the first inequality as an exercise.

We want to prove that \( \limsup_{k \to \infty} x_{n_k} \leq \limsup_{n \to \infty} x_n \). Define \( a_j := \sup \{x_k : k \geq j\} \) as usual. Also define \( c_j := \sup \{x_{n_k} : k \geq j\} \). It is not true that \( \{c_j\} \) is necessarily a subsequence of \( \{a_j\} \). However, as \( n_k \geq k \) for all \( k \), we have that \( \{x_{n_k} : k \geq j\} \subset \{x_k : k \geq j\} \). A supremum of a subset is less than or equal to the supremum of the set and therefore
\[
c_j \leq a_j \quad \text{for all} \ j.
\]

**Lemma 2.2.3** gives
\[
\lim_{j \to \infty} c_j \leq \lim_{j \to \infty} a_j,
\]
which is the desired conclusion. \( \square \)

Limit superior and limit inferior are the largest and smallest subsequential limits. If the subsequence \( \{x_{n_k}\} \) in the previous proposition is convergent, then \( \liminf_{n \to \infty} x_n = \lim_{k \to \infty} x_{n_k} = \limsup_{n \to \infty} x_n \). Therefore,
\[
\liminf_{n \to \infty} x_n \leq \lim_{k \to \infty} x_{n_k} \leq \limsup_{n \to \infty} x_n.
\]

Similarly, we get the following useful test for convergence of a bounded sequence. We leave the proof as an exercise.

**Proposition 2.3.7.** A bounded sequence \( \{x_n\} \) is convergent and converges to \( x \) if and only if every convergent subsequence \( \{x_{n_k}\} \) converges to \( x \).

### 2.3.3 Bolzano–Weierstrass theorem

While it is not true that a bounded sequence is convergent, the Bolzano–Weierstrass theorem tells us that we can at least find a convergent subsequence. The version of Bolzano–Weierstrass that we present in this section is the Bolzano–Weierstrass for sequences of real numbers.

**Theorem 2.3.8** (Bolzano–Weierstrass). Suppose a sequence \( \{x_n\} \) of real numbers is bounded. Then there exists a convergent subsequence \( \{x_{n_k}\} \).

**Proof.** We use **Theorem 2.3.4.** It says that there exists a subsequence whose limit is \( \limsup_{n \to \infty} x_n \). \( \square \)

The reader might complain right now that **Theorem 2.3.4** is strictly stronger than the Bolzano–Weierstrass theorem as presented above. That is true. However, **Theorem 2.3.4** only applies to the real line, but Bolzano–Weierstrass applies in more general contexts (that is, in \( \mathbb{R}^n \)) with pretty much the exact same statement.

As the theorem is so important to analysis, we present an explicit proof. The idea of the following proof also generalizes to different contexts.
Alternate proof of Bolzano–Weierstrass. As the sequence is bounded, then there exist two numbers \( a_1 < b_1 \) such that \( a_1 \leq x_n \leq b_1 \) for all \( n \in \mathbb{N} \). We will define a subsequence \( \{x_{n_k}\} \) and two sequences \( \{a_i\} \) and \( \{b_i\} \), such that \( \{a_i\} \) is monotone increasing, \( \{b_i\} \) is monotone decreasing, \( a_i \leq x_{n_k} \leq b_i \) and such that \( \lim a_i = \lim b_i \). That \( x_{n_k} \) converges then follows by the squeeze lemma.

We define the sequences inductively. We will always have that \( a_i < b_i \), and that \( x_n \in [a_i, b_i] \) for infinitely many \( n \in \mathbb{N} \). We have already defined \( a_1 \) and \( b_1 \). We take \( n_1 := 1 \), that is \( x_{n_1} = x_1 \). Suppose that up to some \( k \in \mathbb{N} \), we have defined the subsequence \( x_{n_1}, x_{n_2}, \ldots, x_{n_k} \), and the sequences \( a_1, a_2, \ldots, a_k \) and \( b_1, b_2, \ldots, b_k \). Let \( y := \frac{a_k + b_k}{2} \). Clearly \( a_k < y < b_k \). If there exist infinitely many \( j \in \mathbb{N} \) such that \( x_j \in [a_k, y] \), then set \( a_{k+1} := a_k, b_{k+1} := y \), and pick \( n_{k+1} > n_k \) such that \( x_{n_{k+1}} \in [a_k, y] \). If there are not infinitely many \( j \) such that \( x_j \in [a_k, y] \), then it must be true that there are infinitely many \( j \in \mathbb{N} \) such that \( x_j \in [y, b_k] \). In this case pick \( a_{k+1} := y, b_{k+1} := b_k \), and pick \( n_{k+1} > n_k \) such that \( x_{n_{k+1}} \in [y, b_k] \).

We now have the sequences defined. What is left to prove is that \( \lim a_i = \lim b_i \). The limits exist as the sequences are monotone. In the construction, \( b_i - a_i \) is cut in half in each step. Therefore, \( b_{i+1} - a_{i+1} = \frac{b_i - a_i}{2} \). By induction,

\[
b_i - a_i = \frac{b_1 - a_1}{2^{i-1}}.
\]

Let \( x := \lim a_i \). As \( \{a_i\} \) is monotone,

\[
x = \sup\{a_i : i \in \mathbb{N}\}.
\]

Let \( y := \lim b_i = \inf\{b_i : i \in \mathbb{N}\} \). Since \( a_i < b_i \) for all \( i \), then \( x \leq y \). As the sequences are monotone, then for all \( i \), we have (why?)

\[
y - x \leq b_i - a_i = \frac{b_1 - a_1}{2^{i-1}}.
\]

Because \( \frac{b_1 - a_1}{2^{i-1}} \) is arbitrarily small and \( y - x \geq 0 \), we have \( y - x = 0 \). Finish by the squeeze lemma. \( \Box \)

Yet another proof of the Bolzano–Weierstrass theorem is to show the following claim, which is left as a challenging exercise. \textit{Claim: Every sequence has a monotone subsequence.}

### 2.3.4 Infinite limits

Just as for infima and suprema, it is possible to allow certain limits to be infinite. That is, we write \( \lim x_n = \infty \) or \( \lim x_n = -\infty \) for certain divergent sequences.

**Definition 2.3.9.** We say \( \{x_n\} \) diverges to infinity\(^\ast\) if for every \( K \in \mathbb{R} \), there exists an \( M \in \mathbb{N} \) such that for all \( n \geq M \), we have \( x_n > K \). In this case we write

\[
\lim_{n \to \infty} x_n := \infty.
\]

Similarly, if for every \( K \in \mathbb{R} \) there exists an \( M \in \mathbb{N} \) such that for all \( n \geq M \), we have \( x_n < K \), we say \( \{x_n\} \) diverges to minus infinity and we write

\[
\lim_{n \to \infty} x_n := -\infty.
\]

\(^\ast\)Sometimes it is said that \( \{x_n\} \) converges to infinity.
With this definition and allowing $\infty$ and $-\infty$, we can write $\lim x_n$ for any monotone sequence.

**Proposition 2.3.10.** Suppose $\{x_n\}$ is a monotone unbounded sequence. Then

$$\lim_{n \to \infty} x_n = \begin{cases} 
\infty & \text{if } \{x_n\} \text{ is increasing,} \\
-\infty & \text{if } \{x_n\} \text{ is decreasing.}
\end{cases}$$

**Proof.** The case of monotone increasing follows from Exercise 2.3.14 part c) below. Let us do monotone decreasing. Suppose $\{x_n\}$ is decreasing and unbounded, that is, for every $K \in \mathbb{R}$, there is an $M \in \mathbb{N}$ such that $x_M < K$. By monotonicity $x_n \leq x_M < K$ for all $n \geq M$. Therefore, $\lim x_n = -\infty$. \qed

**Example 2.3.11:**

$$\lim_{n \to \infty} n = \infty, \quad \lim_{n \to \infty} n^2 = \infty, \quad \lim_{n \to \infty} -n = -\infty.$$  

We leave verification to the reader.

We may also allow $\liminf$ and $\limsup$ to take on the values $\infty$ and $-\infty$, so that we can apply $\liminf$ and $\limsup$ to absolutely any sequence, not just a bounded one. Unfortunately, the sequences $\{a_n\}$ and $\{b_n\}$ are not sequences of real numbers but of extended real numbers. In particular, $a_n$ can equal $\infty$ for some $n$, and $b_n$ can equal $-\infty$. So we have no definition for the limits. But since the extended real numbers are still an ordered set, we can take suprema and infima.

**Definition 2.3.12.** Let $\{x_n\}$ be an unbounded sequence of real numbers. Define sequences of extended real numbers by $a_n := \sup\{x_k : k \geq n\}$ and $b_n := \inf\{x_k : k \geq n\}$. Define

$$\limsup_{n \to \infty} x_n := \inf\{a_n : n \in \mathbb{N}\}, \quad \text{and} \quad \liminf_{n \to \infty} x_n := \sup\{b_n : n \in \mathbb{N}\}.$$  

This definition agrees with the definition for bounded sequences whenever $\lim a_n$ or $\lim b_n$ makes sense including possibly $\infty$ and $-\infty$.

**Proposition 2.3.13.** Let $\{x_n\}$ be an unbounded sequence. Define $\{a_n\}$ and $\{b_n\}$ as above. Then $\{a_n\}$ is decreasing, and $\{b_n\}$ is increasing. If $a_n$ is a real number for every $n$, then $\limsup x_n = \lim a_n$. If $b_n$ is a real number for every $n$, then $\liminf x_n = \lim b_n$.

**Proof.** As before, $a_n = \sup\{x_k : k \geq n\} \geq \sup\{x_k : k \geq n + 1\} = a_{n+1}$. So $\{a_n\}$ is decreasing. Similarly, $\{b_n\}$ is increasing.

If the sequence $\{a_n\}$ is a sequence of real numbers, then $\lim a_n = \inf\{a_n : n \in \mathbb{N}\}$. This follows from Proposition 2.1.10 if $\{a_n\}$ is bounded and Proposition 2.3.10 if $\{a_n\}$ is unbounded. We proceed similarly with $\{b_n\}$. \qed

The definition behaves as expected with $\limsup$ and $\liminf$, see exercises 2.3.13 and 2.3.14.

**Example 2.3.14:** Suppose $x_n := 0$ for odd $n$ and $x_n := n$ for even $n$. Then $a_n = \infty$ for all $n$, since for every $M$, there exists an even $k$ such that $x_k = k \geq M$. On the other hand, $b_n = 0$ for all $n$, as for every $n$, the set $\{b_k : k \geq n\}$ consists of 0 and nonnegative numbers. So, $\lim x_n$ does not exist, $\limsup x_n = \infty$, $\liminf x_n = 0$.  

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2.3.5 Exercises

Exercise 2.3.1: Suppose \( \{x_n\} \) is a bounded sequence. Define \( a_n \) and \( b_n \) as in Definition 2.3.1. Show that \( \{a_n\} \) and \( \{b_n\} \) are bounded.

Exercise 2.3.2: Suppose \( \{x_n\} \) is a bounded sequence. Define \( b_n \) as in Definition 2.3.1. Show that \( \{b_n\} \) is an increasing sequence.

Exercise 2.3.3: Finish the proof of Proposition 2.3.6. That is, suppose \( \{x_n\} \) is a bounded sequence and \( \{x_{n_k}\} \) is a subsequence. Prove \( \liminf_{n \to \infty} x_n \leq \liminf_{k \to \infty} x_{n_k} \).

Exercise 2.3.4: Prove Proposition 2.3.7.

Exercise 2.3.5: a) Let \( x_n := \frac{(-1)^n}{n} \). Find \( \limsup x_n \) and \( \liminf x_n \).

b) Let \( x_n := \frac{(n-1)(-1)^n}{n} \). Find \( \limsup x_n \) and \( \liminf x_n \).

Exercise 2.3.6: Let \( \{x_n\} \) and \( \{y_n\} \) be bounded sequences such that \( x_n \leq y_n \) for all \( n \). Then show that

\[
\limsup_{n \to \infty} x_n \leq \limsup_{n \to \infty} y_n
\]

and

\[
\liminf_{n \to \infty} x_n \leq \liminf_{n \to \infty} y_n.
\]

Exercise 2.3.7: Let \( \{x_n\} \) and \( \{y_n\} \) be bounded sequences.

a) Show that \( \{x_n + y_n\}\) is bounded.

b) Show that

\[
(\liminf_{n \to \infty} x_n) + (\liminf_{n \to \infty} y_n) \leq \liminf_{n \to \infty} (x_n + y_n).
\]

Hint: Find a subsequence \( \{x_{n_k} + y_{n_k}\} \) of \( \{x_n + y_n\} \) that converges. Then find a subsequence \( \{x_{n_{m_k}}\} \) of \( \{x_{n_k}\} \) that converges. Then apply what you know about limits.

c) Find an explicit \( \{x_n\} \) and \( \{y_n\} \) such that

\[
(\liminf_{n \to \infty} x_n) + (\liminf_{n \to \infty} y_n) < \liminf_{n \to \infty} (x_n + y_n).
\]

Hint: Look for examples that do not have a limit.

Exercise 2.3.8: Let \( \{x_n\} \) and \( \{y_n\}\) be bounded sequences (from the previous exercise we know that \( \{x_n + y_n\} \) is bounded).

a) Show that

\[
(\limsup_{n \to \infty} x_n) + (\limsup_{n \to \infty} y_n) \geq \limsup_{n \to \infty} (x_n + y_n).
\]

Hint: See previous exercise.

b) Find an explicit \( \{x_n\} \) and \( \{y_n\} \) such that

\[
(\limsup_{n \to \infty} x_n) + (\limsup_{n \to \infty} y_n) > \limsup_{n \to \infty} (x_n + y_n).
\]

Hint: See previous exercise.
Exercise 2.3.9: If \( S \subset \mathbb{R} \) is a set, then \( x \in \mathbb{R} \) is a cluster point if for every \( \varepsilon > 0 \), the set \( (x - \varepsilon, x + \varepsilon) \cap S \setminus \{x\} \) is not empty. That is, if there are points of \( S \) arbitrarily close to \( x \). For example, \( S := \{1/n : n \in \mathbb{N}\} \) has a unique (only one) cluster point 0, but 0 \( \notin S \). Prove the following version of the Bolzano–Weierstrass theorem:

**Theorem.** Let \( S \subset \mathbb{R} \) be a bounded infinite set, then there exists at least one cluster point of \( S \).

**Hint:** If \( S \) is infinite, then \( S \) contains a countably infinite subset. That is, there is a sequence \( \{x_n\} \) of distinct numbers in \( S \).

Exercise 2.3.10 (Challenging):

a) Prove that every sequence contains a monotone subsequence. **Hint:** Call \( n \in \mathbb{N} \) a peak if \( a_m \leq a_n \) for all \( m \geq n \). There are two possibilities: Either the sequence has at most finitely many peaks, or it has infinitely many peaks.

b) Conclude the Bolzano–Weierstrass theorem.

Exercise 2.3.11: Prove a stronger version of Proposition 2.3.7. Suppose \( \{x_n\} \) is a sequence such that every subsequence \( \{x_{n_i}\} \) has a subsequence \( \{x_{n_{m_i}}\} \) that converges to \( x \).

a) First show that \( \{x_n\} \) is bounded.

b) Now show that \( \{x_n\} \) converges to \( x \).

Exercise 2.3.12: Let \( \{x_n\} \) be a bounded sequence.

a) Prove that there exists an \( s \) such that for every \( r > s \), there exists an \( M \in \mathbb{N} \) such that for all \( n \geq M \), we have \( x_n < r \).

b) If \( s \) is a number as in a), then prove \( \limsup x_n \leq s \).

c) Show that if \( S \) is the set of all \( s \) as in a), then \( \limsup x_n = \inf S \).

Exercise 2.3.13 (Easy): Suppose \( \{x_n\} \) is such that \( \liminf x_n = -\infty \), \( \limsup x_n = \infty \).

a) Show that \( \{x_n\} \) is not convergent, and also that neither \( \lim x_n = \infty \) nor \( \lim x_n = -\infty \) is true.

b) Find an example of such a sequence.

Exercise 2.3.14: Let \( \{x_n\} \) be a sequence.

a) Show that \( \lim x_n = \infty \) if and only if \( \liminf x_n = \infty \).

b) Then show that \( \lim x_n = -\infty \) if and only if \( \limsup x_n = -\infty \).

c) If \( \{x_n\} \) is monotone increasing, show that either \( \lim x_n \) exists and is finite or \( \lim x_n = \infty \). In either case, \( \lim x_n = \sup \{x_n : n \in \mathbb{N}\} \).

Exercise 2.3.15: Prove the following stronger version of Lemma 2.2.12, the ratio test. Suppose \( \{x_n\} \) is a sequence such that \( x_n \neq 0 \) for all \( n \).

a) Prove that if

\[
\limsup_{n \to \infty} \frac{|x_{n+1}|}{|x_n|} < 1,
\]

then \( \{x_n\} \) converges to 0.

b) Prove that if

\[
\liminf_{n \to \infty} \frac{|x_{n+1}|}{|x_n|} > 1,
\]

then \( \{x_n\} \) is unbounded.
Exercise 2.3.16: Suppose \( \{x_n\} \) is a bounded sequence, \( a_n := \sup\{x_k : k \geq n\} \) as before. Suppose that for some \( \ell \in \mathbb{N} \), \( a_\ell \notin \{x_k : k \geq \ell\} \). Then show that \( a_j = a_\ell \) for all \( j \geq \ell \), and hence \( \limsup x_n = a_\ell \).

Exercise 2.3.17: Suppose \( \{x_n\} \) is a sequence, and \( a_n := \sup\{x_k : k \geq n\} \) and \( b_n := \sup\{x_k : k \geq n\} \) as before.

a) Prove that if \( a_\ell = \infty \) for some \( \ell \in \mathbb{N} \), then \( \limsup x_n = \infty \).

b) Prove that if \( b_\ell = -\infty \) for some \( \ell \in \mathbb{N} \), then \( \liminf x_n = -\infty \).

Exercise 2.3.18: Suppose \( \{x_n\} \) is a sequence such that both \( \liminf x_n \) and \( \limsup x_n \) are finite. Prove that \( \{x_n\} \) is bounded.

Exercise 2.3.19: Suppose \( \{x_n\} \) is a bounded sequence, and \( \varepsilon > 0 \) is given. Prove that there exists an \( M \) such that for all \( k \geq M \),

\[
 x_k - \left( \lim_{n \to \infty} x_n \right) < \varepsilon \quad \text{and} \quad (\liminf_{n \to \infty} x_n) - x_k < \varepsilon.
\]

Exercise 2.3.20: Extend Theorem 2.3.4 to unbounded sequences: Suppose that \( \{x_n\} \) is a sequence. If \( \limsup x_n = \infty \), then prove that there exists a subsequence \( \{x_{n_i}\} \) converging to \( \infty \). Then prove the same result for \( -\infty \), and then prove both statements for \( \liminf \).
2.4 Cauchy sequences

Note: 0.5–1 lecture

Often we wish to describe a certain number by a sequence that converges to it. In this case, it is impossible to use the number itself in the proof that the sequence converges. It would be nice if we could check for convergence without knowing the limit.

Definition 2.4.1. A sequence \( \{x_n\} \) is a Cauchy sequence if for every \( \varepsilon > 0 \) there exists an \( M \in \mathbb{N} \) such that for all \( n \geq M \) and all \( k \geq M \), we have

\[ |x_n - x_k| < \varepsilon. \]

Informally, being Cauchy means that the terms of the sequence are eventually all arbitrarily close to each other. We might expect such a sequence to be convergent, and we would be correct due to \( \mathbb{R} \) having the least-upper-bound property. Before we prove this fact, we look at some examples.

Example 2.4.2: The sequence \( \{1/n\} \) is a Cauchy sequence.

Proof: Given \( \varepsilon > 0 \), find \( M \) such that \( M > 2/\varepsilon \). Then for \( n, k \geq M \), we have \( 1/n < \varepsilon/2 \) and \( 1/k < \varepsilon/2 \). Therefore, for \( n, k \geq M \), we have

\[ \left| \frac{1}{n} - \frac{1}{k} \right| < \frac{1}{n} + \frac{1}{k} < \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon. \]

Example 2.4.3: The sequence \( \{n + 1/n\} \) is a Cauchy sequence.

Proof: Given \( \varepsilon > 0 \), find \( M \) such that \( M > 2/\varepsilon \). Then for \( n, k \geq M \), we have \( 1/n < \varepsilon/2 \) and \( 1/k < \varepsilon/2 \). Therefore, for \( n, k \geq M \), we have

\[ \left| \frac{n + 1}{n} - \frac{k + 1}{k} \right| = \frac{k(n + 1) - n(k + 1)}{nk} = \frac{kn + k - nk - n}{nk} = \frac{k - n}{nk} \leq \frac{k}{nk} + \frac{-n}{nk} = 1 + \frac{1}{k} < \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon. \]

Proposition 2.4.4. A Cauchy sequence is bounded.

Proof. Suppose \( \{x_n\} \) is Cauchy. Pick an \( M \) such that for all \( n, k \geq M \), we have \( |x_n - x_k| < 1 \). In particular, for all \( n \geq M \),

\[ |x_n - x_M| < 1. \]

By the reverse triangle inequality, \(|x_n| - |x_M| \leq |x_n - x_M| < 1\). Hence for \(n \geq M\),
\[|x_n| < 1 + |x_M|.
\]
Let
\[B := \max\{|x_1|, |x_2|, \ldots, |x_{M-1}|, 1 + |x_M|\}.
\]
Then \(|x_n| \leq B\) for all \(n \in \mathbb{N}\).

\[\text{Theorem 2.4.5.} \quad \text{A sequence of real numbers is Cauchy if and only if it converges.}\]

\[\text{Proof.} \quad \text{Let } \varepsilon > 0 \text{ be given and suppose } \{x_n\} \text{ converges to } x. \text{ Then there exists an } M \text{ such that for } n \geq M,
\]
\[|x_n - x| < \frac{\varepsilon}{2}.
\]
Hence for \(n \geq M\) and \(k \geq M\),
\[|x_n - x_k| = |x_n - x + x - x_k| \leq |x_n - x| + |x - x_k| < \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon.
\]

Alright, that direction was easy. Now suppose \(\{x_n\}\) is Cauchy. We have shown that \(\{x_n\}\) is bounded. For a bounded sequence, \(\liminf\) and \(\limsup\) exist, and this is where we use the least-upper-bound property. If we show that
\[\liminf_{n \to \infty} x_n = \limsup_{n \to \infty} x_n,
\]
then \(\{x_n\}\) must be convergent by Proposition 2.3.5.

Define \(a := \limsup x_n\) and \(b := \liminf x_n\). By Theorem 2.3.4, there exist subsequences \(\{x_{n_i}\}\) and \(\{x_{m_i}\}\), such that
\[\lim_{i \to \infty} x_{n_i} = a \quad \text{and} \quad \lim_{i \to \infty} x_{m_i} = b.
\]
Given an \(\varepsilon > 0\), there exists an \(M_1\) such that \(|x_{n_i} - a| < \varepsilon/3\) for all \(i \geq M_1\) and an \(M_2\) such that \(|x_{m_i} - b| < \varepsilon/3\) for all \(i \geq M_2\). There also exists an \(M_3\) such that \(|x_n - x_k| < \varepsilon/3\) for all \(n, k \geq M_3\). Let \(M := \max\{M_1, M_2, M_3\}\). If \(i \geq M\), then \(n_i \geq M\) and \(m_i \geq M\). Hence,
\[|a - b| = |a - x_{n_i} + x_{n_i} - x_{m_i} + x_{m_i} - b|
\leq |a - x_{n_i}| + |x_{n_i} - x_{m_i}| + |x_{m_i} - b|
< \frac{\varepsilon}{3} + \frac{\varepsilon}{3} + \frac{\varepsilon}{3} = \varepsilon.
\]
As \(|a - b| < \varepsilon\) for all \(\varepsilon > 0\), then \(a = b\) and the sequence converges.

\[\text{Remark 2.4.6.} \quad \text{The statement of this proposition is sometimes used to define the completeness property of the real numbers. We say a set is Cauchy-complete (or sometimes just complete) if every Cauchy sequence converges. Above, we proved that as } \mathbb{R} \text{ has the least-upper-bound property, then } \mathbb{R} \text{ is Cauchy-complete. One can construct } \mathbb{R} \text{ via “completing” } \mathbb{Q} \text{ by “throwing in” just enough points to make all Cauchy sequences converge (we omit the details). The resulting field has the least-upper-bound property. The advantage of using Cauchy sequences to define completeness is that this idea generalizes to more abstract settings such as metric spaces, see chapter 7.}\]
The Cauchy criterion is stronger than \(|x_{n+1} - x_n|\) (or \(|x_{n+j} - x_n|\) for a fixed \(j\)) going to zero as \(n\) goes to infinity. When we get to the partial sums of the harmonic series (see Example 2.5.11 in the next section), we will have a sequence such that \(x_{n+1} - x_n = 1/n\), yet \(\{x_n\}\) is divergent. In fact, for that sequence, \(\lim_{n \to \infty} |x_{n+j} - x_n| = 0\) for every \(j \in \mathbb{N}\) (confer Exercise 2.5.12). The key point in the definition of Cauchy is that \(n\) and \(k\) vary independently and can be arbitrarily far apart.

### 2.4.1 Exercises

**Exercise 2.4.1:** Prove that \(\{\frac{n^2-1}{n}\}\) is Cauchy using directly the definition of Cauchy sequences.

**Exercise 2.4.2:** Let \(\{x_n\}\) be a sequence such that there exists a positive \(C < 1\) and for all \(n\),

\[ |x_{n+1} - x_n| \leq C |x_n - x_{n-1}|. \]

Prove that \(\{x_n\}\) is Cauchy. Hint: You can freely use the formula (for \(C \neq 1\))

\[ 1 + C + C^2 + \cdots + C^n = \frac{1 - C^{n+1}}{1 - C}. \]

**Exercise 2.4.3** (Challenging): Suppose \(F\) is an ordered field that contains the rational numbers \(\mathbb{Q}\), such that \(\mathbb{Q}\) is dense, that is: Whenever \(x, y \in F\) are such that \(x < y\), then there exists a \(q \in \mathbb{Q}\) such that \(x < q < y\). Say a sequence \(\{x_n\}_{n=1}^{\infty}\) of rational numbers is Cauchy if given every \(\varepsilon \in \mathbb{Q}\) with \(\varepsilon > 0\), there exists an \(M\) such that for all \(n, k \geq M\), we have \(|x_n - x_k| < \varepsilon\). Suppose every Cauchy sequence of rational numbers has a limit in \(F\). Prove that \(F\) has the least-upper-bound property.

**Exercise 2.4.4:** Let \(\{x_n\}\) and \(\{y_n\}\) be sequences such that \(\lim y_n = 0\). Suppose that for all \(k \in \mathbb{N}\) and for all \(m \geq k\), we have

\[ |x_m - x_k| \leq y_k. \]

Show that \(\{x_n\}\) is Cauchy.

**Exercise 2.4.5:** Suppose a Cauchy sequence \(\{x_n\}\) is such that for every \(M \in \mathbb{N}\), there exists a \(k \geq M\) and an \(n \geq M\) such that \(x_k < 0\) and \(x_n > 0\). Using simply the definition of a Cauchy sequence and of a convergent sequence, show that the sequence converges to 0.

**Exercise 2.4.6:** Suppose \(|x_n - x_k| \leq n/k^2\) for all \(n\) and \(k\). Show that \(\{x_n\}\) is Cauchy.

**Exercise 2.4.7:** Suppose \(\{x_n\}\) is a Cauchy sequence such that for infinitely many \(n\), \(x_n = c\). Using only the definition of Cauchy sequence prove that \(\lim x_n = c\).

**Exercise 2.4.8:** True or false, prove or find a counterexample: If \(\{x_n\}\) is a Cauchy sequence, then there exists an \(M\) such that for all \(n \geq M\), we have \(|x_{n+1} - x_n| \leq |x_n - x_{n-1}|\).
2.5 Series

Note: 2 lectures

A fundamental object in mathematics is that of a series. In fact, when the foundations of analysis were being developed, the motivation was to understand series. Understanding series is important in applications of analysis. For example, solutions to differential equations are often given as series, and differential equations are the basis for understanding almost all of modern science.

2.5.1 Definition

Definition 2.5.1. Given a sequence \( \{x_n\} \), we write the formal object

\[
\sum_{n=1}^{\infty} x_n \quad \text{or sometimes just} \quad \sum x_n
\]

and call it a series. A series converges if the sequence \( \{s_k\} \) defined by

\[
s_k := \sum_{n=1}^{k} x_n = x_1 + x_2 + \cdots + x_k,
\]

converges. The numbers \( s_k \) are called partial sums. If \( x := \lim s_k \), we write

\[
\sum_{n=1}^{\infty} x_n = x.
\]

In this case, we cheat a little and treat \( \sum_{n=1}^{\infty} x_n \) as a number.

If the sequence \( \{s_k\} \) diverges, we say the series is divergent. In this case, \( \sum x_n \) is simply a formal object and not a number.

In other words, for a convergent series, we have

\[
\sum_{n=1}^{\infty} x_n = \lim_{k \to \infty} \sum_{n=1}^{k} x_n.
\]

We only have this equality if the limit on the right actually exists. If the series does not converge, the right-hand side does not make sense (the limit does not exist). Therefore, be careful as \( \sum x_n \) means two different things (a notation for the series itself or the limit of the partial sums), and you must use context to distinguish.

Remark 2.5.2. It is sometimes convenient to start the series at an index different from 1. For instance, we can write

\[
\sum_{n=0}^{\infty} r^n = \sum_{n=1}^{\infty} r^{n-1}.
\]

The left-hand side is more convenient to write.
Remark 2.5.3. It is common to write the series \( \sum x_n \) as
\[
x_1 + x_2 + x_3 + \cdots
\]
with the understanding that the ellipsis indicates a series and not a simple sum. We do not use this notation as it is the sort of informal notation that leads to mistakes in proofs.

Example 2.5.4: The series
\[
\sum_{n=1}^{\infty} \frac{1}{2^n}
\]
converges and the limit is 1. That is,
\[
\sum_{n=1}^{\infty} \frac{1}{2^n} = \lim_{k \to \infty} \sum_{n=1}^{k} \frac{1}{2^n} = 1.
\]

Proof: First we prove the following equality
\[
\left( \sum_{n=1}^{k} \frac{1}{2^n} \right) + \frac{1}{2^k} = 1.
\]
The equality is immediate when \( k = 1 \). The proof for general \( k \) follows by induction, which we leave to the reader. See Figure 2.7 for an illustration.

Let \( s_k \) be the partial sum. We write
\[
|1 - s_k| = \left|1 - \sum_{n=1}^{k} \frac{1}{2^n}\right| = \left|1 - \frac{1}{2^k}\right| = \frac{1}{2^k}.
\]
The sequence \( \left\{ \frac{1}{2^k} \right\} \), and therefore \( \left\{ |1 - s_k| \right\} \), converges to zero. So, \( \{s_k\} \) converges to 1.

Proposition 2.5.5. Suppose \(-1 < r < 1\). Then the geometric series \( \sum_{n=0}^{\infty} r^n \) converges, and
\[
\sum_{n=0}^{\infty} r^n = \frac{1}{1-r}.
\]
Details of the proof are left as an exercise. The proof consists of showing
\[
\sum_{n=0}^{k-1} r^n = \frac{1 - r^k}{1-r},
\]
and then taking the limit as \( k \) goes to \( \infty \). Geometric series is one of the most important series, and in fact it is one of the few series for which we can so explicitly find the limit.

As for sequences we can talk about a tail of a series.
**Proposition 2.5.6.** Let $\sum x_n$ be a series. Let $M \in \mathbb{N}$. Then

$$\sum_{n=1}^{\infty} x_n \text{ converges if and only if } \sum_{n=M}^{\infty} x_n \text{ converges.}$$

**Proof.** We look at partial sums of the two series (for $k \geq M$)

$$\sum_{n=1}^{k} x_n = \left(\sum_{n=1}^{M-1} x_n\right) + \left(\sum_{n=M}^{k} x_n\right).$$

Note that $\sum_{n=1}^{M-1} x_n$ is a fixed number. Use Proposition 2.2.5 to finish the proof. \qed

### 2.5.2 Cauchy series

**Definition 2.5.7.** A series $\sum x_n$ is said to be *Cauchy* or a *Cauchy series* if the sequence of partial sums $\{s_n\}$ is a Cauchy sequence.

A sequence of real numbers converges if and only if it is Cauchy. Therefore, a series is convergent if and only if it is Cauchy. The series $\sum x_n$ is Cauchy if for every $\varepsilon > 0$, there exists an $M \in \mathbb{N}$, such that for every $n \geq M$ and $k \geq M$, we have

$$\left| \left(\sum_{j=1}^{k} x_j\right) - \left(\sum_{j=1}^{n} x_j\right) \right| < \varepsilon.$$ 

Without loss of generality we assume $n < k$. Then we write

$$\left| \left(\sum_{j=1}^{k} x_j\right) - \left(\sum_{j=1}^{n} x_j\right) \right| = \left| \sum_{j=n+1}^{k} x_j \right| < \varepsilon.$$ 

We have proved the following simple proposition.

**Proposition 2.5.8.** The series $\sum x_n$ is Cauchy if for every $\varepsilon > 0$, there exists an $M \in \mathbb{N}$ such that for every $n \geq M$ and every $k > n$, we have

$$\left| \sum_{j=n+1}^{k} x_j \right| < \varepsilon.$$ 

### 2.5.3 Basic properties

**Proposition 2.5.9.** Let $\sum x_n$ be a convergent series. Then the sequence $\{x_n\}$ is convergent and

$$\lim_{n \to \infty} x_n = 0.$$ 

**Proof.** Let $\varepsilon > 0$ be given. As $\sum x_n$ is convergent, it is Cauchy. Thus we find an $M$ such that for every $n \geq M$, we have

$$\varepsilon > \left| \sum_{j=n+1}^{n+1} x_j \right| = |x_{n+1}|.$$ 

Hence for every $n \geq M + 1$, we have $|x_n| < \varepsilon$. \qed
Example 2.5.10: If \( r \geq 1 \) or \( r \leq -1 \), then the geometric series \( \sum_{n=0}^{\infty} r^n \) diverges.

Proof: \( |r^n| = |r|^n \geq 1^n = 1 \). So the terms do not go to zero and the series cannot converge.

So if a series converges, the terms of the series go to zero. The implication, however, goes only one way. Let us give an example.

Example 2.5.11: The series \( \sum \frac{1}{n} \) diverges (despite the fact that \( \lim \frac{1}{n} = 0 \)). This is the famous harmonic series*.

Proof: We will show that the sequence of partial sums is unbounded, and hence cannot converge.

Write the partial sums \( s_n \) for \( n = 2^k \) as:

\[
\begin{align*}
  s_1 &= 1, \\
  s_2 &= (1) + \left( \frac{1}{2} \right), \\
  s_4 &= (1) + \left( \frac{1}{2} \right) + \left( \frac{1}{3} + \frac{1}{4} \right), \\
  s_8 &= (1) + \left( \frac{1}{2} \right) + \left( \frac{1}{3} + \frac{1}{4} \right) + \left( \frac{1}{5} + \frac{1}{6} + \frac{1}{7} + \frac{1}{8} \right), \\
  & \vdots \\
  s_{2^k} &= 1 + \sum_{j=1}^{k} \left( \sum_{m=2^{j-1}+1}^{2^j} \frac{1}{m} \right).
\end{align*}
\]

Notice \( \frac{1}{3} + \frac{1}{4} \geq \frac{1}{4} + \frac{1}{4} = \frac{1}{2} \) and \( \frac{1}{5} + \frac{1}{6} + \frac{1}{7} + \frac{1}{8} \geq \frac{1}{8} + \frac{1}{8} + \frac{1}{8} + \frac{1}{8} = \frac{1}{2} \). More generally

\[
\sum_{m=2^{k-1}+1}^{2^k} \frac{1}{m} \geq \sum_{m=2^{k-1}+1}^{2^k} \frac{1}{2^k} = \left( 2^{k-1} \right) \frac{1}{2^k} = \frac{1}{2}.
\]

Therefore,

\[
s_{2^k} = 1 + \sum_{j=1}^{k} \left( \sum_{m=2^{j-1}+1}^{2^j} \frac{1}{m} \right) \geq 1 + \sum_{j=1}^{k} \frac{1}{2} = 1 + \frac{k}{2}.
\]

As \( \{ \frac{k}{2} \} \) is unbounded by the Archimedean property, that means that \( \{ s_{2^k} \} \) is unbounded, and therefore \( \{ s_n \} \) is unbounded. Hence \( \{ s_n \} \) diverges, and consequently \( \sum \frac{1}{n} \) diverges.

Convergent series are linear. That is, we can multiply them by constants and add them and these operations are done term by term.

**Proposition 2.5.12** (Linearity of series). Let \( \alpha \in \mathbb{R} \) and \( \sum x_n \) and \( \sum y_n \) be convergent series. Then

(i) \( \sum \alpha x_n \) is a convergent series and

\[
\sum_{n=1}^{\infty} \alpha x_n = \alpha \sum_{n=1}^{\infty} x_n.
\]

*The divergence of the harmonic series was known long before the theory of series was made rigorous. The proof we give is the earliest proof and was given by Nicole Oresme (1323?–1382).
(ii) $\sum (x_n + y_n)$ is a convergent series and

$$\sum_{n=1}^{\infty} (x_n + y_n) = \left( \sum_{n=1}^{\infty} x_n \right) + \left( \sum_{n=1}^{\infty} y_n \right).$$

**Proof.** For the first item, we simply write the $k$th partial sum

$$\sum_{n=1}^{k} \alpha x_n = \alpha \left( \sum_{n=1}^{k} x_n \right).$$

We look at the right-hand side and note that the constant multiple of a convergent sequence is convergent. Hence, we take the limit of both sides to obtain the result.

For the second item we also look at the $k$th partial sum

$$\sum_{n=1}^{k} (x_n + y_n) = \left( \sum_{n=1}^{k} x_n \right) + \left( \sum_{n=1}^{k} y_n \right).$$

We look at the right-hand side and note that the sum of convergent sequences is convergent. Hence, we take the limit of both sides to obtain the proposition.

An example of a useful application of the first item is the following formula. If $|r| < 1$ and $j \in \mathbb{N}$, then

$$\sum_{n=j}^{\infty} r^n = \frac{r^j}{1-r}.$$ 

The formula follows by using the geometric series and multiplying by $r^j$:

$$r^j \sum_{n=0}^{\infty} r^n = \sum_{n=0}^{\infty} r^{n+j} = \sum_{n=j}^{\infty} r^n.$$ 

Multiplying series is not as simple as adding, see the next section. It is not true, of course, that we multiply term by term. That strategy does not work even for finite sums: $(a+b)(c+d) \neq ac + bd$.

### 2.5.4 Absolute convergence

As monotone sequences are easier to work with than arbitrary sequences, it is usually easier to work with series $\sum x_n$, where $x_n \geq 0$ for all $n$. The sequence of partial sums is then monotone increasing and converges if it is bounded above. Let us formalize this statement as a proposition.

**Proposition 2.5.13.** If $x_n \geq 0$ for all $n$, then $\sum x_n$ converges if and only if the sequence of partial sums is bounded above.

As the limit of a monotone increasing sequence is the supremum, then when $x_n \geq 0$ for all $n$, we have the inequality

$$\sum_{n=1}^{k} x_n \leq \sum_{n=1}^{\infty} x_n.$$
If we allow infinite limits, the inequality still holds even when the series diverges to infinity, although in that case it is not terribly useful.

We will see that the following common criterion for convergence of series has big implications for how the series can be manipulated.

**Definition 2.5.14.** A series \( \sum x_n \) **converges absolutely** if the series \( \sum |x_n| \) converges. If a series converges, but does not converge absolutely, we say it is **conditionally convergent**.

**Proposition 2.5.15.** If the series \( \sum x_n \) converges absolutely, then it converges.

**Proof.** A series is convergent if and only if it is Cauchy. Hence suppose \( \sum |x_n| \) is Cauchy. That is, for every \( \varepsilon > 0 \), there exists an \( M \) such that for all \( k \geq M \) and all \( n > k \), we have

\[
\sum_{j=k+1}^{n} |x_j| = \left| \sum_{j=k+1}^{n} x_j \right| < \varepsilon.
\]

We apply the triangle inequality for a finite sum to obtain

\[
\left| \sum_{j=k+1}^{n} x_j \right| \leq \sum_{j=k+1}^{n} |x_j| < \varepsilon.
\]

Hence \( \sum x_n \) is Cauchy, and therefore it converges.

If \( \sum x_n \) converges absolutely, the limits of \( \sum x_n \) and \( \sum |x_n| \) are generally different. Computing one does not help us compute the other. However, the computation above leads to a useful inequality for absolutely convergent series, a series version of the triangle inequality, a proof of which we leave as an exercise:

\[
\left| \sum_{j=1}^{\infty} x_j \right| \leq \sum_{j=1}^{\infty} |x_j|.
\]

Absolutely convergent series have many wonderful properties. For example, absolutely convergent series can be rearranged arbitrarily, or we can multiply such series together easily. Conditionally convergent series on the other hand often do not behave as one would expect. See the next section.

We leave as an exercise to show that

\[
\sum_{n=1}^{\infty} \frac{(-1)^n}{n}
\]

converges, although the reader should finish this section before trying. On the other hand, we proved

\[
\sum_{n=1}^{\infty} \frac{1}{n}
\]

diverges. Therefore, \( \sum \frac{(-1)^n}{n} \) is a conditionally convergent series.
2.5.5 Comparison test and the $p$-series

We noted above that for a series to converge the terms not only have to go to zero, but they have to go to zero “fast enough.” If we know about convergence of a certain series, we can use the following comparison test to see if the terms of another series go to zero “fast enough.”

**Proposition 2.5.16** (Comparison test). Let $\sum x_n$ and $\sum y_n$ be series such that $0 \leq x_n \leq y_n$ for all $n \in \mathbb{N}$.

(i) If $\sum y_n$ converges, then so does $\sum x_n$.

(ii) If $\sum x_n$ diverges, then so does $\sum y_n$.

**Proof.** As the terms of the series are all nonnegative, the sequences of partial sums are both monotone increasing. Since $x_n \leq y_n$ for all $n$, the partial sums satisfy for all $k$

$$\sum_{n=1}^{k} x_n \leq \sum_{n=1}^{k} y_n. \tag{2.1}$$

If the series $\sum y_n$ converges, the partial sums for the series are bounded. Therefore, the right-hand side of (2.1) is bounded for all $k$; there exists some $B \in \mathbb{R}$ such that $\sum_{n=1}^{k} y_n \leq B$ for all $k$, and so

$$\sum_{n=1}^{k} x_n \leq \sum_{n=1}^{k} y_n \leq B.$$

Hence the partial sums for $\sum x_n$ are also bounded. Since the partial sums are a monotone increasing sequence they are convergent. The first item is thus proved.

On the other hand if $\sum x_n$ diverges, the sequence of partial sums must be unbounded since it is monotone increasing. That is, the partial sums for $\sum x_n$ are eventually bigger than any real number. Putting this together with (2.1) we see that for every $B \in \mathbb{R}$, there is a $k$ such that

$$B \leq \sum_{n=1}^{k} x_n \leq \sum_{n=1}^{k} y_n.$$

Hence the partial sums for $\sum y_n$ are also unbounded, and $\sum y_n$ also diverges. \qed

A useful series to use with the comparison test is the $p$-series*.

**Proposition 2.5.17** ($p$-series or the $p$-test). For $p \in \mathbb{R}$, the series

$$\sum_{n=1}^{\infty} \frac{1}{n^p}$$

converges if and only if $p > 1$.

---

*We have not yet defined $x^p$ for $x > 0$ and an arbitrary $p \in \mathbb{R}$. The definition is $x^p := \exp(p \ln x)$. We will define the logarithm and the exponential in §5.4. For now you can just think of rational $p$ where $x^{k/m} = (x^{1/m})^k$. See also Exercise 1.2.17.
Proof. First suppose $p \leq 1$. As $n \geq 1$, we have $\frac{1}{n^p} \geq \frac{1}{n}$. Since $\sum \frac{1}{n}$ diverges, $\sum \frac{1}{n^p}$ must diverge for all $p \leq 1$ by the comparison test.

Now suppose $p > 1$. We proceed as we did for the harmonic series, but instead of showing that the sequence of partial sums is unbounded, we show that it is bounded. The terms of the series are positive, so the sequence of partial sums is monotone increasing and converges if it is bounded above. Let $s_n$ denote the $n$th partial sum.

$$s_1 = 1,$$

$$s_3 = (1) + \left( \frac{1}{2^p} + \frac{1}{3^p} \right),$$

$$s_7 = (1) + \left( \frac{1}{2^p} + \frac{1}{3^p} \right) + \left( \frac{1}{4^p} + \frac{1}{5^p} + \frac{1}{6^p} + \frac{1}{7^p} \right),$$

$$\vdots$$

$$s_{2^k-1} = 1 + \sum_{j=1}^{k-1} \left( \sum_{m=2^j}^{2^{j+1}-1} \frac{1}{m^p} \right).$$

Instead of estimating from below, we estimate from above. As $p$ is positive, then $2^p < 3^p$, and hence $\frac{1}{2^p} + \frac{1}{3^p} < \frac{1}{2^p} + \frac{1}{3^p}$. Similarly, $\frac{1}{4^p} + \frac{1}{5^p} + \frac{1}{6^p} + \frac{1}{7^p} < \frac{1}{4^p} + \frac{1}{4^p} + \frac{1}{4^p} + \frac{1}{4^p}$. Therefore, for all $k \geq 2$,

$$s_{2^k-1} = 1 + \sum_{j=1}^{k-1} \left( \sum_{m=2^j}^{2^{j+1}-1} \frac{1}{m^p} \right) < 1 + \sum_{j=1}^{k-1} \left( \sum_{m=2^j}^{2^{j+1}-1} \frac{1}{(2j)^p} \right) = 1 + \sum_{j=1}^{k-1} \left( \frac{2j}{(2j)^p} \right) = 1 + \sum_{j=1}^{k-1} \left( \frac{1}{2^{p-1}} \right)^j.$$

As $p > 1$, then $\frac{1}{2^{p-1}} < 1$. Proposition 2.5.5 says that

$$\sum_{j=1}^{\infty} \left( \frac{1}{2^{p-1}} \right)^j$$

converges. Thus,

$$s_{2^k-1} < 1 + \sum_{j=1}^{k-1} \left( \frac{1}{2^{p-1}} \right)^j \leq 1 + \sum_{j=1}^{\infty} \left( \frac{1}{2^{p-1}} \right)^j.$$

For every $n$ there is a $k \geq 2$ such that $n \leq 2^k - 1$, and as $\{s_n\}$ is a monotone sequence, $s_n \leq s_{2^k-1}$. So for all $n$,

$$s_n < 1 + \sum_{j=1}^{\infty} \left( \frac{1}{2^{p-1}} \right)^j.$$

Thus the sequence of partial sums is bounded, and the series converges. \qed
Neither the $p$-series test nor the comparison test tell us what the sum converges to. They only tell us that a limit of the partial sums exists. For instance, while we know that $\sum 1/n^2$ converges, it is far harder to find* that the limit is $\pi^2/6$. If we treat $\sum 1/n^p$ as a function of $p$, we get the so-called Riemann $\zeta$ function. Understanding the behavior of this function contains one of the most famous unsolved problems in mathematics today and has applications in seemingly unrelated areas such as modern cryptography.

**Example 2.5.18:** The series $\sum \frac{1}{n^2 + 1}$ converges.

Proof: First, $\frac{1}{n^2 + 1} < \frac{1}{n^2}$ for all $n \in \mathbb{N}$. The series $\sum \frac{1}{n^2}$ converges by the $p$-series test. Therefore, by the comparison test, $\sum \frac{1}{n^2 + 1}$ converges.

### 2.5.6 Ratio test

Suppose $r > 0$. The ratio of two subsequent terms in the geometric series $\sum r^n$ is $\frac{r^{n+1}}{r^n} = r$, and the series converges whenever $r < 1$. Just as for sequences, this fact can be generalized to more arbitrary series as long as we have such a ratio “in the limit.” We then compare the tail of a series to the geometric series.

**Proposition 2.5.19** (Ratio test). Let $\sum x_n$ be a series, $x_n \neq 0$ for all $n$, and such that

$$L := \lim_{n \to \infty} \frac{|x_{n+1}|}{|x_n|}$$

exists.

(i) If $L < 1$, then $\sum x_n$ converges absolutely.

(ii) If $L > 1$, then $\sum x_n$ diverges.

Although the test as stated is often sufficient, it can be strengthened a bit, see Exercise 2.5.6.

**Proof.** If $L > 1$, then Lemma 2.2.12 says that the sequence $\{x_n\}$ diverges. Since it is a necessary condition for the convergence of series that the terms go to zero, we know that $\sum x_n$ must diverge.

Thus suppose $L < 1$. We will argue that $\sum |x_n|$ must converge. The proof is similar to that of Lemma 2.2.12. Of course $L \geq 0$. Pick $r$ such that $L < r < 1$. As $r - L > 0$, there exists an $M \in \mathbb{N}$ such that for all $n \geq M$,

$$\left| \frac{|x_{n+1}|}{|x_n|} - L \right| < r - L.$$

Therefore,

$$\frac{|x_{n+1}|}{|x_n|} < r.$$

For $n > M$ (that is for $n \geq M + 1$), write

$$|x_n| = |x_M| \left| \frac{x_{M+1}}{x_M} \right| \left| \frac{x_{M+2}}{x_{M+1}} \right| \cdots \left| \frac{x_n}{x_{n-1}} \right| < |x_M| r r \cdots r = |x_M| r^{n-M} = \left( |x_M| r^{-M} \right) r^n.$$

*Demonstration of this fact is what made the Swiss mathematician Leonhard Paul Euler (1707–1783) famous.
For \( k > M \), write the partial sum as

\[
\sum_{n=1}^{k} |x_n| = \left( \sum_{n=1}^{M} |x_n| \right) + \left( \sum_{n=M+1}^{k} |x_n| \right)
\]
\[< \left( \sum_{n=1}^{M} |x_n| \right) + \left( |x_M| r^{-M} \right) \left( \sum_{n=M+1}^{k} r^n \right)\]
\[= \left( \sum_{n=1}^{M} |x_n| \right) + (|x_M| r^{-M}) \left( \sum_{n=M+1}^{k} r^n \right).
\]

As \( 0 < r < 1 \), the geometric series \( \sum_{n=0}^{\infty} r^n \) converges, so \( \sum_{n=M+1}^{\infty} r^n \) converges as well. We take the limit as \( k \) goes to infinity on the right-hand side above to obtain

\[
\sum_{n=1}^{k} |x_n| \leq \left( \sum_{n=1}^{M} |x_n| \right) + (|x_M| r^{-M}) \left( \sum_{n=M+1}^{\infty} r^n \right).
\]

The right-hand side is a number that does not depend on \( k \). Hence the sequence of partial sums of \( \sum |x_n| \) is bounded and \( \sum |x_n| \) is convergent. Thus \( \sum x_n \) is absolutely convergent.

**Example 2.5.20:** The series

\[
\sum_{n=1}^{\infty} \frac{2^n}{n!}
\]

converges absolutely.

**Proof:** We write

\[
\lim_{n \to \infty} \frac{2^{n+1}/(n+1)!}{2^n/n!} = \lim_{n \to \infty} \frac{2}{n+1} = 0.
\]

Therefore, the series converges absolutely by the ratio test.

### 2.5.7 Exercises

**Exercise 2.5.1:** Suppose the \( k \)th partial sum of \( \sum_{n=1}^{\infty} x_n \) is \( s_k = \frac{k}{k+1} \). Find the series, that is find \( x_n \), prove that the series converges, and then find the limit.

**Exercise 2.5.2:** Prove Proposition 2.5.5, that is for \( -1 < r < 1 \) prove

\[
\sum_{n=0}^{\infty} r^n = \frac{1}{1-r}.
\]

**Hint:** See Example 0.3.8.

**Exercise 2.5.3:** Decide the convergence or divergence of the following series.

a) \( \sum_{n=1}^{\infty} \frac{3}{9n+1} \)  

b) \( \sum_{n=1}^{\infty} \frac{1}{2n-1} \)  

c) \( \sum_{n=1}^{\infty} \frac{(-1)^n}{n^2} \)  

d) \( \sum_{n=1}^{\infty} \frac{1}{n(n+1)} \)  

e) \( \sum_{n=1}^{\infty} ne^{-n^2} \)
Exercise 2.5.4:

a) Prove that if \( \sum_{n=1}^{\infty} x_n \) converges, then \( \sum_{n=1}^{\infty} (x_{2n} + x_{2n+1}) \) also converges.

b) Find an explicit example where the converse does not hold.

Exercise 2.5.5: For \( j = 1, 2, \ldots, n \), let \( \{x_{j,k}\}_{k=1}^{\infty} \) denote \( n \) sequences. Suppose that for each \( j \in \mathbb{N} \),

\[
\sum_{k=1}^{\infty} x_{j,k}
\]

is convergent. Prove

\[
\sum_{j=1}^{n} \left( \sum_{k=1}^{\infty} x_{j,k} \right) = \sum_{k=1}^{\infty} \left( \sum_{j=1}^{n} x_{j,k} \right) .
\]

Exercise 2.5.6: Prove the following stronger version of the ratio test: Let \( \sum x_n \) be a series.

a) If there is an \( N \) and a \( \rho < 1 \) such that \( \frac{|x_{n+1}|}{|x_n|} < \rho \) for all \( n \geq N \), then the series converges absolutely.

(Remark: Equivalently the condition can be stated as \( \limsup_{n \to \infty} \left| \frac{x_{n+1}}{x_n} \right| < 1 \).)

b) If there is an \( N \) such that \( \frac{|x_{n+1}|}{|x_n|} \geq 1 \) for all \( n \geq N \), then the series diverges.

Exercise 2.5.7 (Challenging): Suppose \( \{x_n\} \) is a decreasing sequence and \( \sum x_n \) converges. Prove \( \lim_{n \to \infty} nx_n = 0. \)

Exercise 2.5.8: Show that \( \sum_{n=1}^{\infty} \frac{(-1)^n}{n} \) converges. Hint: Consider the sum of two subsequent entries.

Exercise 2.5.9:

a) Prove that if \( \sum x_n \) and \( \sum y_n \) converge absolutely, then \( \sum x_n y_n \) converges absolutely.

b) Find an explicit example where the converse does not hold.

c) Find an explicit example where all three series are absolutely convergent, are not just finite sums, and \( (\sum x_n)(\sum y_n) \neq \sum x_n y_n \). That is, show that series are not multiplied term-by-term.

Exercise 2.5.10: Prove the triangle inequality for series: If \( \sum x_n \) converges absolutely, then

\[
\left| \sum_{n=1}^{\infty} x_n \right| \leq \sum_{n=1}^{\infty} |x_n| .
\]

Exercise 2.5.11: Prove the limit comparison test. That is, prove that if \( a_n > 0 \) and \( b_n > 0 \) for all \( n \), and

\[
0 < \lim_{n \to \infty} \frac{a_n}{b_n} < \infty,
\]

then either \( \sum a_n \) and \( \sum b_n \) both converge or both diverge.

Exercise 2.5.12: Let \( x_n := \sum_{j=1}^{\infty} j. \) Show that for every \( k \), we get \( \lim_{n \to \infty} |x_{n+k} - x_n| = 0 \), yet \( \{x_n\} \) is not Cauchy.

Exercise 2.5.13: Let \( s_k \) be the \( k \)th partial sum of \( \sum x_n \).

a) Suppose that there exists an \( m \in \mathbb{N} \) such that \( \lim_{k \to \infty} s_{mk} \) exists and \( \lim_{n \to \infty} x_n = 0 \). Show that \( \sum x_n \) converges.

b) Find an example where \( \lim_{k \to \infty} s_{2k} \) exists and \( \lim_{n \to \infty} x_n \neq 0 \) (and therefore \( \sum x_n \) diverges).

c) (Challenging) Find an example where \( \lim_{n \to \infty} x_n = 0 \), and there exists a subsequence \( \{s_{k_j}\} \) such that \( \lim_{j \to \infty} s_{k_j} \) exists, but \( \sum x_n \) still diverges.
Exercise 2.5.14: Suppose $\sum x_n$ converges and $x_n \geq 0$ for all $n$. Prove that $\sum x_n^2$ converges.

Exercise 2.5.15 (Challenging): Suppose $\{x_n\}$ is a decreasing sequence of positive numbers. The proof of convergence/divergence for the $p$-series generalizes. Prove the so-called Cauchy condensation principle:

$$\sum_{n=1}^{\infty} x_n \text{ converges if and only if } \sum_{n=1}^{\infty} 2^n x_{2^n} \text{ converges.}$$

Exercise 2.5.16: Use the Cauchy condensation principle (see Exercise 2.5.15) to decide the convergence of

a) $\sum \frac{\ln n}{n^2}$  

b) $\sum \frac{1}{n\ln n}$  

c) $\sum \frac{1}{n(\ln n)^2}$  

d) $\sum \frac{1}{n(n\ln n)(\ln\ln n)^2}$

For the series to be well-defined you need to start some of the series at $n = 2$. Note that only the tails of some of these series satisfy the hypotheses of the principle; you should argue why that is sufficient.

Hint: Feel free to use the identity $\ln(2^n) = n\ln 2$.

Exercise 2.5.17 (Challenging): Prove Abel’s theorem:

Theorem. Suppose $\sum x_n$ is a series whose partial sums are a bounded sequence, $\{\lambda_n\}$ is a sequence with $\lim \lambda_n = 0$, and $\sum |\lambda_{n+1} - \lambda_n|$ is convergent. Then $\sum \lambda_n x_n$ is convergent.
2.6 More on series

Note: up to 2–3 lectures (optional, can safely be skipped or covered partially)

2.6.1 Root test

A test similar to the ratio test is the so-called root test. In fact, the proof of this test is similar and somewhat easier. Again, the idea is to generalize what happens for the geometric series.

Proposition 2.6.1 (Root test). Let $\sum x_n$ be a series and let

$$L := \limsup_{n \to \infty} \left| x_n \right|^{1/n}.$$

(i) If $L < 1$, then $\sum x_n$ converges absolutely.
(ii) If $L > 1$, then $\sum x_n$ diverges.

Proof. If $L > 1$, then there exists* a subsequence $\{x_{n_k}\}$ such that $L = \lim_{k \to \infty} |x_{n_k}|^{1/n_k}$. Let $r$ be such that $L > r > 1$. There exists an $M$ such that for all $k \geq M$, we have $|x_{n_k}|^{1/n_k} > r > 1$, or in other words $|x_{n_k}| > r^{n_k} > 1$. The subsequence $\{|x_{n_k}|\}$, and therefore also $\{|x_n|\}$, cannot possibly converge to zero, and so the series diverges.

Now suppose $L < 1$. Pick $r$ such that $L < r < 1$. By definition of limit supremum, there is an $M$ such that for all $n \geq M$,

$$\sup\{|x_k|^{1/k} : k \geq n\} < r.$$

Therefore, for all $n \geq M$,

$$|x_n|^{1/n} < r, \quad \text{or in other words} \quad |x_n| < r^n.$$

Let $k > M$, and estimate the $k$th partial sum:

$$\sum_{n=1}^{k} |x_n| = \left( \sum_{n=1}^{M} |x_n| \right) + \left( \sum_{n=M+1}^{k} |x_n| \right) \leq \left( \sum_{n=1}^{M} |x_n| \right) + \left( \sum_{n=M+1}^{k} r^n \right).$$

As $0 < r < 1$, the geometric series $\sum_{n=M+1}^{\infty} r^n$ converges to $\frac{r^{M+1}}{1-r}$. As everything is positive,

$$\sum_{n=1}^{k} |x_n| \leq \left( \sum_{n=1}^{M} |x_n| \right) + \frac{r^{M+1}}{1-r}.$$

Thus the sequence of partial sums of $\sum |x_n|$ is bounded, and the series converges. Therefore, $\sum x_n$ converges absolutely.

*In case $L = \infty$, see Exercise 2.3.20. Alternatively, note that if $L = \infty$, then $\{|x_n|^{1/n}\}$ and thus $\{x_n\}$ is unbounded.
2.6.2 Alternating series test

The tests we have seen so far only addressed absolute convergence. The following test gives a large supply of conditionally convergent series.

**Proposition 2.6.2** (Alternating series). Let \( \{x_n\} \) be a monotone decreasing sequence of positive real numbers such that \( \lim x_n = 0 \). Then
\[
\sum_{n=1}^{\infty} (-1)^n x_n
\]
converges.

**Proof.** Let \( s_m := \sum_{k=1}^{m} (-1)^k x_k \) be the \( m \)th partial sum. Then write
\[
s_{2n} = \sum_{k=1}^{2n} (-1)^k x_k = (-x_1 + x_2) + \cdots + (-x_{2n-1} + x_{2n}) = \sum_{k=1}^{n} (-x_{2k-1} + x_{2k}).
\]
The sequence \( \{x_k\} \) is decreasing and so \( (-x_{2k-1} + x_{2k}) \leq 0 \) for all \( k \). Therefore, the subsequence \( \{s_{2n}\} \) of partial sums is a decreasing sequence. Similarly, \( (x_{2k} - x_{2k+1}) \geq 0 \), and so
\[
s_{2n} = -x_1 + (x_2 - x_3) + \cdots + (x_{2n-2} - x_{2n-1}) + x_{2n} \geq -x_1.
\]
The sequence \( \{s_{2n}\} \) is decreasing and bounded below, so it converges. Let \( a := \lim s_{2n} \).

We wish to show that \( \lim s_m = a \) (and not just for the subsequence). Notice
\[
s_{2n+1} = s_{2n} + x_{2n+1}.
\]
Given \( \varepsilon > 0 \), pick \( M \) such that \( |s_{2n} - a| < \varepsilon / 2 \) whenever \( 2n \geq M \). Since \( \lim x_n = 0 \), we also make \( M \) possibly larger to obtain \( x_{2n+1} < \varepsilon / 2 \) whenever \( 2n \geq M \). If \( 2n \geq M \), we have \( |s_{2n} - a| < \varepsilon / 2 < \varepsilon \), so we just need to check the situation for \( s_{2n+1} \):
\[
|s_{2n+1} - a| = |s_{2n} - a + x_{2n+1}| \leq |s_{2n} - a| + x_{2n+1} < \varepsilon / 2 + \varepsilon / 2 = \varepsilon.
\]

Notably, there exist conditionally convergent series where the absolute values of the terms go to zero arbitrarily slowly. The series
\[
\sum_{n=1}^{\infty} \frac{(-1)^n}{n^p}
\]
converges for arbitrarily small \( p > 0 \), but it does not converge absolutely when \( p \leq 1 \).

2.6.3 Rearrangements

Absolutely convergent series behave as we imagine they should. For example, absolutely convergent series can be summed in any order whatsoever. Nothing of the sort holds for conditionally convergent series (see Example 2.6.4 and Exercise 2.6.3).

Consider a series
\[
\sum_{n=1}^{\infty} x_n.
\]
Given a bijective function $\sigma : \mathbb{N} \to \mathbb{N}$, the corresponding rearrangement is the following series:

$$\sum_{k=1}^{\infty} x_{\sigma(k)}.$$

We simply sum the series in a different order.

**Proposition 2.6.3.** Let $\sum x_n$ be an absolutely convergent series converging to a number $x$. Let $\sigma : \mathbb{N} \to \mathbb{N}$ be a bijection. Then $\sum x_{\sigma(n)}$ is absolutely convergent and converges to $x$.

In other words, a rearrangement of an absolutely convergent series converges (absolutely) to the same number.

**Proof.** Let $\varepsilon > 0$ be given. As $\sum x_n$ is absolutely convergent, take $M$ such that

$$\left| \sum_{n=1}^{M} x_n - x \right| < \frac{\varepsilon}{2} \quad \text{and} \quad \sum_{n=M+1}^{\infty} |x_n| < \frac{\varepsilon}{2}.$$

As $\sigma$ is a bijection, there exists a number $K$ such that for each $n \leq M$, there exists $k \leq K$ such that $\sigma(k) = n$. In other words $\{1, 2, \ldots, M\} \subset \sigma(\{1, 2, \ldots, K\})$.

For $N \geq K$, let $Q := \max \{1, 2, \ldots, N\}$. Compute

$$\left| \sum_{n=1}^{N} x_{\sigma(n)} - x \right| = \left| \sum_{n=1}^{M} x_n + \sum_{n=M+1}^{N} x_{\sigma(n)} - x \right|$$

$$\leq \left| \sum_{n=1}^{M} x_n - x \right| + \sum_{n=M+1}^{N} |x_{\sigma(n)}|$$

$$\leq \left| \sum_{n=1}^{M} x_n - x \right| + \sum_{n=M+1}^{Q} |x_n|$$

$$< \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon.$$

So $\sum x_{\sigma(n)}$ converges to $x$. To see that the convergence is absolute, we apply the argument above to $\sum |x_n|$ to show that $\sum |x_{\sigma(n)}|$ converges. \qed

**Example 2.6.4:** Let us show that the alternating harmonic series $\sum \frac{(-1)^{n+1}}{n}$, which does not converge absolutely, can be rearranged to converge to anything. The odd terms and the even terms diverge to plus infinity and minus infinity respectively (prove this!):

$$\sum_{m=1}^{\infty} \frac{1}{2m-1} = \infty, \quad \text{and} \quad \sum_{m=1}^{\infty} \frac{-1}{2m} = -\infty.$$

Let $a_n := \frac{(-1)^{n+1}}{n}$ for simplicity, let an arbitrary number $L \in \mathbb{R}$ be given, and set $\sigma(1) := 1$. Suppose we have defined $\sigma(n)$ for all $n \leq N$. If

$$\sum_{n=1}^{N} a_{\sigma(n)} \leq L,$$
2.6. MORE ON SERIES

then let \( \sigma(N + 1) := k \) be the smallest odd \( k \in \mathbb{N} \) that we have not used yet, that is, \( \sigma(n) \neq k \) for all \( n \leq N \). Otherwise, let \( \sigma(N + 1) := k \) be the smallest even \( k \) that we have not yet used.

By construction \( \sigma: \mathbb{N} \to \mathbb{N} \) is one-to-one. It is also onto, because if we keep adding either odd (resp. even) terms, eventually we pass \( L \) and switch to the evens (resp. odds). So we switch infinitely many times.

Finally, let \( N \) be the \( N \) where we just pass \( L \) and switch. For example, suppose we have just switched from odd to even (so we start subtracting), and let \( N' > N \) be where we first switch back from even to odd. Then

\[
L + \frac{1}{\sigma(N)} \geq \sum_{n=1}^{N-1} a_{\sigma(n)} > \sum_{n=1}^{N'-1} a_{\sigma(n)} > L - \frac{1}{\sigma(N')}. 
\]

And similarly for switching in the other direction. Therefore, the sum up to \( N' - 1 \) is within \( \frac{1}{\min\{\sigma(N)^{N}, \sigma(N')\}} \) of \( L \). As we switch infinitely many times we obtain that \( \sigma(N) \to \infty \) and \( \sigma(N') \to \infty \), and hence

\[
\sum_{n=1}^{\infty} a_{\sigma(n)} = \sum_{n=1}^{\infty} (-1)^{\sigma(n)+1} = L.
\]

Here is an example to illustrate the proof. Suppose \( L = 1.2 \), then the order is

\[
1 + 1/3 - 1/2 + 1/5 + 1/7 + 1/9 - 1/4 + 1/11 + 1/13 - 1/6 + 1/15 + 1/17 + 1/19 - 1/8 + \cdots.
\]

At this point we are no more than \( 1/8 \) from the limit.

2.6.4 Multiplication of series

As we have already mentioned, multiplication of series is somewhat harder than addition. If at least one of the series converges absolutely, then we can use the following theorem. For this result, it is convenient to start the series at 0, rather than at 1.

Theorem 2.6.5 (Mertens’ theorem*). Suppose \( \sum_{n=0}^{\infty} a_n \) and \( \sum_{n=0}^{\infty} b_n \) are two convergent series, converging to \( A \) and \( B \) respectively. If at least one of the series converges absolutely, then the series \( \sum_{n=0}^{\infty} c_n \) where

\[
c_n = a_0 b_n + a_1 b_{n-1} + \cdots + a_n b_0 = \sum_{j=0}^{n} a_j b_{n-j},
\]

converges to \( AB \).

The series \( \sum c_n \) is called the Cauchy product of \( \sum a_n \) and \( \sum b_n \).

Proof. Suppose \( \sum a_n \) converges absolutely, and let \( \varepsilon > 0 \) be given. In this proof instead of picking complicated estimates just to make the final estimate come out as less than \( \varepsilon \), let us simply obtain an estimate that depends on \( \varepsilon \) and can be made arbitrarily small.

Write

\[
A_m := \sum_{n=0}^{m} a_n, \quad B_m := \sum_{n=0}^{m} b_n.
\]

*Proved by the German mathematician Franz Mertens (1840–1927).
We rearrange the $m$th partial sum of $\sum c_n$:

$$\left| \left( \sum_{n=0}^{m} c_n \right) - AB \right| = \left| \left( \sum_{n=0}^{m} \sum_{j=0}^{n} a_j b_{n-j} \right) - AB \right|$$

$$= \left| \sum_{n=0}^{m} B_n a_{m-n} \right| - AB$$

$$= \left| \sum_{n=0}^{m} (B_n - B) a_{m-n} \right| + BA_m - AB$$

$$\leq \left( \sum_{n=0}^{m} |B_n - B| |a_{m-n}| \right) + |B| |A_m - A|$$

We can surely make the second term on the right-hand side go to zero. The trick is to handle the first term. Pick $K$ such that for all $m \geq K$, we have $|A_m - A| < \varepsilon$ and also $|B_m - B| < \varepsilon$. Finally, as $\sum a_n$ converges absolutely, make sure that $K$ is large enough such that for all $m \geq K$,

$$\sum_{n=K}^{m} |a_n| < \varepsilon.$$ 

As $\sum b_n$ converges, then we have that $B_{\text{max}} := \sup\{|B_n - B| : n = 0, 1, 2, \ldots\}$ is finite. Take $m \geq 2K$, then in particular $m - K + 1 > K$. So

$$\sum_{n=0}^{m} |B_n - B| |a_{m-n}| = \left( \sum_{n=0}^{m-K} |B_n - B| |a_{m-n}| \right) + \left( \sum_{n=m-K+1}^{m} |B_n - B| |a_{m-n}| \right)$$

$$\leq \left( \sum_{n=K}^{m} |a_n| \right) B_{\text{max}} + \left( \sum_{n=0}^{K-1} \varepsilon |a_n| \right)$$

$$\leq \varepsilon B_{\text{max}} + \varepsilon \left( \sum_{n=0}^{\infty} |a_n| \right).$$

Therefore, for $m \geq 2K$, we have

$$\left| \left( \sum_{n=0}^{m} c_n \right) - AB \right| \leq \left( \sum_{n=0}^{m} |B_n - B| |a_{m-n}| \right) + |B| |A_m - A|$$

$$\leq \varepsilon B_{\text{max}} + \varepsilon \left( \sum_{n=0}^{\infty} |a_n| \right) + |B| \varepsilon = \varepsilon \left( B_{\text{max}} + \left( \sum_{n=0}^{\infty} |a_n| \right) + |B| \right).$$

The expression in the parenthesis on the right-hand side is a fixed number. Hence, we can make the right-hand side arbitrarily small by picking a small enough $\varepsilon > 0$. So $\sum_{n=0}^{\infty} c_n$ converges to $AB$.  

**Example 2.6.6:** If both series are only conditionally convergent, the Cauchy product series need not even converge. Suppose we take $a_n = b_n = (-1)^n \frac{1}{\sqrt{n+1}}$. The series $\sum_{n=0}^{\infty} a_n = \sum_{n=0}^{\infty} b_n$ converges
by the alternating series test; however, it does not converge absolutely as can be seen from the $p$-test. Let us look at the Cauchy product.

$$c_n = (-1)^n \left( \frac{1}{\sqrt{n+1}} + \frac{1}{\sqrt{2n}} + \frac{1}{\sqrt{3(n-1)}} + \cdots + \frac{1}{\sqrt{n+1}} \right) = (-1)^n \sum_{j=0}^{n} \frac{1}{\sqrt{(j+1)(n-j+1)}}.$$  

Therefore,

$$|c_n| = \sum_{j=0}^{n} \frac{1}{\sqrt{(j+1)(n-j+1)}} \geq \sum_{j=0}^{n} \frac{1}{\sqrt{(n+1)(n+1)}} = 1.$$  

The terms do not go to zero and hence $\sum c_n$ cannot converge.

### 2.6.5 Power series

Fix $x_0 \in \mathbb{R}$. A *power series* about $x_0$ is a series of the form

$$\sum_{n=0}^{\infty} a_n (x-x_0)^n.$$  

A power series is really a function of $x$, and many important functions in analysis can be written as a power series. We use the convention that $0^0 = 1$ (if $x = x_0$ and $n = 0$).

We say that a power series is *convergent* if there is at least one $x \neq x_0$ that makes the series converge. If $x = x_0$, then the series always converges since all terms except the first are zero. If the series does not converge for any point $x \neq x_0$, we say that the series is *divergent*.

**Example 2.6.7:** The series

$$\sum_{n=0}^{\infty} \frac{1}{n!} x^n$$

is absolutely convergent for all $x \in \mathbb{R}$ using the ratio test: For any $x \in \mathbb{R}$

$$\lim_{n \to \infty} \left| \frac{(1/(n+1)!)}{(1/n)!} x^{n+1} \right| = \lim_{n \to \infty} \frac{|x|}{n+1} = 0.$$  

Recall from calculus that this series converges to $e^x$.

**Example 2.6.8:** The series

$$\sum_{n=1}^{\infty} \frac{1}{n} x^n$$

converges absolutely for all $x \in (-1, 1)$ via the ratio test:

$$\lim_{n \to \infty} \left| \frac{(1/(n+1)!)}{(1/n)!} x^{n+1} \right| = \lim_{n \to \infty} \frac{|x|}{n+1} = |x| < 1.$$  

The series converges at $x = -1$, as $\sum_{n=1}^{\infty} \frac{(-1)^n}{n}$ converges by the alternating series test. But the power series does not converge absolutely at $x = -1$, because $\sum_{n=1}^{\infty} \frac{1}{n}$ does not converge. The series diverges at $x = 1$. When $|x| > 1$, then the series diverges via the ratio test.
Example 2.6.9: The series
\[ \sum_{n=1}^{\infty} n^r x^n \]
diverges for all \( x \neq 0 \). Let us apply the root test
\[ \limsup_{n \to \infty} |n^r x^n|^{\frac{1}{n}} = \limsup_{n \to \infty} n|x| = \infty. \]
Therefore, the series diverges for all \( x \neq 0 \).

Convergence of power series in general works analogously to one of the three examples above.

Proposition 2.6.10. Let \( \sum_{n=0}^{\infty} a_n (x-x_0)^n \) be a power series. If the series is convergent, then either it converges at all \( x \in \mathbb{R} \), or there exists a number \( \rho \), such that the series converges absolutely on the interval \( (x_0 - \rho, x_0 + \rho) \) and diverges when \( x < x_0 - \rho \) or \( x > x_0 + \rho \).

The number \( \rho \) is called the radius of convergence of the power series. We write \( \rho = \infty \) if the series converges for all \( x \), and we write \( \rho = 0 \) if the series is divergent. At the endpoints, that is if \( x = x_0 + \rho \) or \( x = x_0 - \rho \), the proposition says nothing, and the series might or might not converge. See Figure 2.8. In Example 2.6.8 the radius of convergence is \( \rho = 1 \). In Example 2.6.7 the radius of convergence is \( \rho = \infty \), and in Example 2.6.9 the radius of convergence is \( \rho = 0 \).

\[ \begin{array}{c}
\text{diverges} \quad ? \quad \text{converges absolutely} \quad ? \quad \text{diverges} \\
\hline
x_0 - \rho \quad x_0 \quad x_0 + \rho
\end{array} \]

Figure 2.8: Convergence of a power series.

Proof. Write
\[ R := \limsup_{n \to \infty} |a_n|^{\frac{1}{n}}. \]
We use the root test to prove the proposition:
\[ L = \limsup_{n \to \infty} |a_n(x-x_0)^n|^{\frac{1}{n}} = |x-x_0| \limsup_{n \to \infty} |a_n|^{\frac{1}{n}} = |x-x_0| R. \]
In particular, if \( R = \infty \), then \( L = \infty \) for every \( x \neq x_0 \), and the series diverges by the root test. On the other hand, if \( R = 0 \), then \( L = 0 \) for every \( x \), and the series converges absolutely for all \( x \).

Suppose \( 0 < R < \infty \). The series converges absolutely if \( 1 > L = R |x-x_0| \), or in other words when
\[ |x-x_0| < 1/R. \]
The series diverges when \( 1 < L = R |x-x_0| \), or
\[ |x-x_0| > 1/R. \]
Letting \( \rho := 1/R \) completes the proof. \( \square \)
2.6. MORE ON SERIES

It may be useful to restate what we have learned in the proof as a separate proposition.

**Proposition 2.6.11.** Let \( \sum_{n=0}^{\infty} a_n(x-x_0)^n \) be a power series, and let

\[
R := \limsup_{n \to \infty} |a_n|^{1/n}.
\]

If \( R = \infty \), the power series is divergent. If \( R = 0 \), then the power series converges everywhere. Otherwise, the radius of convergence \( \rho = 1/R \).

Often, radius of convergence is written as \( \rho = 1/R \) in all three cases, with the understanding of what \( \rho \) should be if \( R = 0 \) or \( R = \infty \).

Convergent power series can be added and multiplied together, and multiplied by constants. The proposition has a straight forward proof using what we know about series in general, and power series in particular. We leave the proof to the reader.

**Proposition 2.6.12.** Let \( \sum_{n=0}^{\infty} a_n(x-x_0)^n \) and \( \sum_{n=0}^{\infty} b_n(x-x_0)^n \) be two convergent power series with radius of convergence at least \( \rho > 0 \) and \( \alpha \in \mathbb{R} \). Then for all \( x \) such that \( |x-x_0| < \rho \), we have

\[
\left( \sum_{n=0}^{\infty} a_n(x-x_0)^n \right) + \left( \sum_{n=0}^{\infty} b_n(x-x_0)^n \right) = \sum_{n=0}^{\infty} (a_n + b_n)(x-x_0)^n,
\]

\[
\alpha \left( \sum_{n=0}^{\infty} a_n(x-x_0)^n \right) = \sum_{n=0}^{\infty} \alpha a_n(x-x_0)^n,
\]

and

\[
\left( \sum_{n=0}^{\infty} a_n(x-x_0)^n \right) \left( \sum_{n=0}^{\infty} b_n(x-x_0)^n \right) = \sum_{n=0}^{\infty} c_n(x-x_0)^n,
\]

where \( c_n = a_0b_n + a_1b_{n-1} + \cdots + a_nb_0 \).

That is, after performing the algebraic operations, the radius of convergence of the resulting series is at least \( \rho \). For all \( x \) with \( |x-x_0| < \rho \), we have two convergent series so their term by term addition and multiplication by constants follows by what we learned in the last section. For multiplication of two power series, the series are absolutely convergent inside the radius of convergence and that is why for those \( x \) we can apply Mertens’ theorem. Note that after applying an algebraic operation the radius of convergence could increase. See the exercises.

Let us look at some examples of power series. Polynomials are simply finite power series. That is, a polynomial is a power series where the \( a_n \) are zero for all \( n \) large enough. We expand a polynomial as a power series about any point \( x_0 \) by writing the polynomial as a polynomial in \( (x-x_0) \). For example, \( 2x^2 - 3x + 4 \) as a power series around \( x_0 = 1 \) is

\[ 2x^2 - 3x + 4 = 3 + (x-1) + 2(x-1)^2. \]

We can also expand *rational functions* (that is, ratios of polynomials) as power series, although we will not completely prove this fact here. Notice that a series for a rational function only defines
the function on an interval even if the function is defined elsewhere. For example, for the geometric series, we have that for \( x \in (-1, 1) \)
\[
\frac{1}{1-x} = \sum_{n=0}^{\infty} x^n.
\]
The series diverges when \( |x| > 1 \), even though \( \frac{1}{1-x} \) is defined for all \( x \neq 1 \).

We can use the geometric series together with rules for addition and multiplication of power series to expand rational functions as power series around \( x_0 \), as long as the denominator is not zero at \( x_0 \). We state without proof that this is always possible, and we give an example of such a computation using the geometric series.

**Example 2.6.13:** Let us expand \( \frac{x}{1+2x+x^2} \) as a power series around the origin \((x_0 = 0)\) and find the radius of convergence.

Write \( 1+2x+x^2 = (1+x)^2 = (1-(-x))^2 \), and suppose \( |x| < 1 \). Compute
\[
\frac{x}{1+2x+x^2} = x \left( \sum_{n=0}^{\infty} (-1)^n x^n \right)^2
\]
\[
= x \left( \sum_{n=0}^{\infty} c_n x^n \right)
\]
\[
= \sum_{n=0}^{\infty} c_n x^{n+1}.
\]

Using the formula for the product of series, we obtain \( c_0 = 1, c_1 = -1 - 1 = -2, c_2 = 1 + 1 + 1 = 3 \), etc. Hence, for \( |x| < 1 \),
\[
\frac{x}{1+2x+x^2} = \sum_{n=1}^{\infty} (-1)^{n+1} nx^n.
\]
The radius of convergence is at least 1. We leave it to the reader to verify that the radius of convergence is exactly equal to 1.

You can use the method of partial fractions you know from calculus. For example, to find the power series for \( \frac{x^3+x}{x^2-1} \) at 0, write
\[
\frac{x^3+x}{x^2-1} = x + \frac{1}{1+x} - \frac{1}{1-x} = x + \sum_{n=0}^{\infty} (-1)^n x^n - \sum_{n=0}^{\infty} x^n.
\]

### 2.6.6 Exercises

**Exercise 2.6.1:** Decide the convergence or divergence of the following series.

\begin{align*}
  a) \sum_{n=1}^{\infty} \frac{1}{2^{2n+1}} & \quad b) \sum_{n=1}^{\infty} \frac{(-1)^n(n-1)}{n} & \quad c) \sum_{n=1}^{\infty} \frac{(-1)^n}{n^{1/10}} & \quad d) \sum_{n=1}^{\infty} \frac{n^n}{(n+1)^{2n}}
\end{align*}
Exercise 2.6.2: Suppose both \( \sum_{n=0}^{\infty} a_n \) and \( \sum_{n=0}^{\infty} b_n \) converge absolutely. Show that the product series, \( \sum_{n=0}^{\infty} c_n \) where \( c_n = a_0b_n + a_1b_{n-1} + \cdots + a_nb_0 \), also converges absolutely.

Exercise 2.6.3 (Challenging): Let \( \sum a_n \) be conditionally convergent. Show that given an arbitrary \( x \in \mathbb{R} \) there exists a rearrangement of \( \sum a_n \) such that the rearranged series converges to \( x \). Hint: See Example 2.6.4.

Exercise 2.6.4:

a) Show that the alternating harmonic series \( \sum (-1)^{n+1} \) has a rearrangement such that whenever \( x < y \), there exists a partial sum \( s_n \) of the rearranged series such that \( x < s_n < y \).

b) Show that the rearrangement you found does not converge. See Example 2.6.4.

c) Show that for every \( x \in \mathbb{R} \), there exists a subsequence of partial sums \( \{s_{n_k}\} \) of your rearrangement such that \( \lim s_{n_k} = x \).

Exercise 2.6.5: For the following power series, find if they are convergent or not, and if so find their radius of convergence.

\[
\begin{align*}
& a) \sum_{n=0}^{\infty} 2^n x^n \\
& b) \sum_{n=0}^{\infty} n x^n \\
& c) \sum_{n=0}^{\infty} n! x^n \\
& d) \sum_{n=0}^{\infty} \frac{1}{(2n)!} (x-10)^n \\
& e) \sum_{n=0}^{\infty} x^{2n} \\
& f) \sum_{n=0}^{\infty} n! x^n!
\end{align*}
\]

Exercise 2.6.6: Suppose \( \sum a_n x^n \) converges for \( x = 1 \).

a) What can you say about the radius of convergence?

b) If you further know that at \( x = 1 \) the convergence is not absolute, what can you say?

Exercise 2.6.7: Expand \( \frac{x}{4-x^2} \) as a power series around \( x_0 = 0 \) and compute its radius of convergence.

Exercise 2.6.8:

a) Find an example where the radii of convergence of \( \sum a_n x^n \) and \( \sum b_n x^n \) are both 1, but the radius of convergence of the sum of the two series is infinite.

b) (Trickier) Find an example where the radii of convergence of \( \sum a_n x^n \) and \( \sum b_n x^n \) are both 1, but the radius of convergence of the product of the two series is infinite.

Exercise 2.6.9: Figure out how to compute the radius of convergence using the ratio test. That is, suppose \( \sum a_n x^n \) is a power series and \( R := \lim_{n \to \infty} \left| \frac{a_{n+1}}{a_n} \right| \) exists or is \( \infty \). Find the radius of convergence and prove your claim.

Exercise 2.6.10:

a) Prove that \( \lim n^{1/n} = 1 \) using the following procedure: Write \( n^{1/n} = 1 + b_n \) and note \( b_n > 0 \). Then show that \( (1 + b_n)^n \geq \frac{n(n-1)}{2} b_n^2 \) and use this to show that \( \lim b_n = 0 \).

b) Use the result of part a) to show that if \( \sum a_n x^n \) is a convergent power series with radius of convergence \( R \), then \( \sum n a_n x^n \) is also convergent with the same radius of convergence.

There are different notions of summability (convergence) of a series than just the one we have seen. A common one is Cesàro summability\(^*\). Let \( \sum a_n \) be a series and let \( s_n \) be the \( n \)th partial sum. The series is said to be Cesàro summable to \( a \) if

\[
a = \lim_{n \to \infty} \frac{s_1 + s_2 + \cdots + s_n}{n}.
\]

\(^*\)Named for the Italian mathematician Ernesto Cesàro (1859–1906).
Exercise 2.6.11 (Challenging):

a) If $\sum a_n$ is convergent to $a$ (in the usual sense), show that $\sum a_n$ is Cesàro summable (see above) to $a$.

b) Show that in the sense of Cesàro $\sum (-1)^n$ is summable to $1/2$.

c) Let $a_n := k$ when $n = k^3$ for some $k \in \mathbb{N}$, $a_n := -k$ when $n = k^3 + 1$ for some $k \in \mathbb{N}$, otherwise let $a_n := 0$.
   Show that $\sum a_n$ diverges in the usual sense, (partial sums are unbounded), but it is Cesàro summable to 0 (seems a little paradoxical at first sight).

Exercise 2.6.12 (Challenging): Show that the monotonicity in the alternating series test is necessary. That is, find a sequence of positive real numbers $\{x_n\}$ with $\lim x_n = 0$ but such that $\sum (-1)^n x_n$ diverges.

Exercise 2.6.13: Find a series such that $\sum x_n$ converges but $\sum x_n^2$ diverges. Hint: Compare Exercise 2.5.14.

Exercise 2.6.14: Suppose $\{c_n\}$ is a sequence. Prove that for every $r \in (0, 1)$, there exists a strictly increasing sequence $\{n_k\}$ of natural numbers ($n_{k+1} > n_k$) such that

$$\sum_{k=1}^{\infty} c_k x^{n_k}$$

converges absolutely for all $x \in [-r, r]$. 

Chapter 3

Continuous Functions

3.1 Limits of functions

Note: 2–3 lectures

Before we define continuity of functions, we must visit a somewhat more general notion of a limit. Given a function \( f : S \rightarrow \mathbb{R} \), we want to see how \( f(x) \) behaves as \( x \) tends to a certain point.

3.1.1 Cluster points

First, we return to a concept we have previously seen in an exercise. When moving within the set \( S \) we can only approach points that have elements of \( S \) arbitrarily near.

**Definition 3.1.1.** Let \( S \subset \mathbb{R} \) be a set. A number \( x \in \mathbb{R} \) is called a cluster point of \( S \) if for every \( \varepsilon > 0 \), the set \((x - \varepsilon, x + \varepsilon) \cap S \setminus \{x\}\) is not empty.

That is, \( x \) is a cluster point of \( S \) if there are points of \( S \) arbitrarily close to \( x \). Another way of phrasing the definition is to say that \( x \) is a cluster point of \( S \) if for every \( \varepsilon > 0 \), there exists a \( y \in S \) such that \( y \neq x \) and \( |x - y| < \varepsilon \). Note that a cluster point of \( S \) need not lie in \( S \).

Let us see some examples.

(i) The set \( \{1/n : n \in \mathbb{N}\} \) has a unique cluster point zero.

(ii) The cluster points of the open interval \((0, 1)\) are all points in the closed interval \([0, 1]\).

(iii) The set of cluster points of \( \mathbb{Q} \) is the whole real line \( \mathbb{R} \).

(iv) The set of cluster points of \([0, 1) \cup \{2\}\) is the interval \([0, 1]\).

(v) The set \( \mathbb{N} \) has no cluster points in \( \mathbb{R} \).

**Proposition 3.1.2.** Let \( S \subset \mathbb{R} \). Then \( x \in \mathbb{R} \) is a cluster point of \( S \) if and only if there exists a convergent sequence of numbers \( \{x_n\} \) such that \( x_n \neq x \) and \( x_n \in S \) for all \( n \), and \( \lim x_n = x \).

**Proof.** First suppose \( x \) is a cluster point of \( S \). For every \( n \in \mathbb{N} \), pick \( x_n \) to be an arbitrary point of \((x - 1/n, x + 1/n) \cap S \setminus \{x\}\), which is nonempty because \( x \) is a cluster point of \( S \). Then \( x_n \) is within \( 1/n \) of \( x \), that is, \( |x - x_n| < 1/n \).

As \( \{1/n\} \) converges to zero, \( \{x_n\} \) converges to \( x \).
On the other hand, if we start with a sequence of numbers \( \{x_n\} \) in \( S \) converging to \( x \) such that \( x_n \neq x \) for all \( n \), then for every \( \varepsilon > 0 \) there is an \( M \) such that, in particular, \( |x_M - x| < \varepsilon \). That is, \( x_M \in (x - \varepsilon, x + \varepsilon) \cap S \setminus \{x\} \). 

3.1.2 Limits of functions

If a function \( f \) is defined on a set \( S \) and \( c \) is a cluster point of \( S \), then we define the limit of \( f(x) \) as \( x \) gets close to \( c \). It is irrelevant for the definition whether \( f \) is defined at \( c \) or not. Even if the function is defined at \( c \), the limit of the function as \( x \) goes to \( c \) can very well be different from \( f(c) \).

**Definition 3.1.3.** Let \( f : S \to \mathbb{R} \) be a function and \( c \) a cluster point of \( S \subset \mathbb{R} \). Suppose there exists an \( L \in \mathbb{R} \) and for every \( \varepsilon > 0 \), there exists a \( \delta > 0 \) such that whenever \( x \in S \setminus \{c\} \) and \( |x - c| < \delta \), we have

\[
|f(x) - L| < \varepsilon.
\]

We then say \( f(x) \) converges to \( L \) as \( x \) goes to \( c \). We say \( L \) is the limit of \( f(x) \) as \( x \) goes to \( c \). We write

\[
\lim_{x \to c} f(x) := L,
\]

or

\[
f(x) \to L \quad \text{as} \quad x \to c.
\]

If no such \( L \) exists, then we say that the limit does not exist or that \( f \) diverges at \( c \).

Again the notation and language we are using above assumes the limit is unique even though we have not yet proved uniqueness. Let us do that now.

**Proposition 3.1.4.** Let \( c \) be a cluster point of \( S \subset \mathbb{R} \) and let \( f : S \to \mathbb{R} \) be a function such that \( f(x) \) converges as \( x \) goes to \( c \). Then the limit of \( f(x) \) as \( x \) goes to \( c \) is unique.

**Proof.** Let \( L_1 \) and \( L_2 \) be two numbers that both satisfy the definition. Take an \( \varepsilon > 0 \) and find a \( \delta_1 > 0 \) such that \( |f(x) - L_1| < \varepsilon/2 \) for all \( x \in S \setminus \{c\} \) with \( |x - c| < \delta_1 \). Also find \( \delta_2 > 0 \) such that \( |f(x) - L_2| < \varepsilon/2 \) for all \( x \in S \setminus \{c\} \) with \( |x - c| < \delta_2 \). Put \( \delta := \min\{\delta_1, \delta_2\} \). Suppose \( x \in S \), \( |x - c| < \delta \), and \( x \neq c \). As \( \delta > 0 \) and \( c \) is a cluster point, such an \( x \) exists. Then

\[
|L_1 - L_2| = |L_1 - f(x) + f(x) - L_2| \leq |L_1 - f(x)| + |f(x) - L_2| < \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon.
\]

As \( |L_1 - L_2| < \varepsilon \) for arbitrary \( \varepsilon > 0 \), then \( L_1 = L_2 \). 

**Example 3.1.5:** Consider \( f : \mathbb{R} \to \mathbb{R} \) defined by \( f(x) := x^2 \). Then for any \( c \in \mathbb{R} \),

\[
\lim_{x \to c} f(x) = \lim_{x \to c} x^2 = c^2.
\]

**Proof:** Let \( c \in \mathbb{R} \) be fixed, and suppose \( \varepsilon > 0 \) is given. Write

\[
\delta := \min\left\{ 1, \frac{\varepsilon}{2|c| + 1} \right\}.
\]
3.1. LIMITS OF FUNCTIONS

Take \( x \neq c \) such that \( |x - c| < \delta \). In particular, \( |x - c| < 1 \). By reverse triangle inequality, we get
\[
|x| - |c| \leq |x - c| < 1.
\]
Adding 2\(|c|\) to both sides, we obtain \( |x| + |c| < 2|c| + 1 \). We compute
\[
|f(x) - c^2| = |x^2 - c^2|
\]
\[
= |(x + c)(x - c)|
\]
\[
= |x + c||x - c|
\]
\[
\leq (|x| + |c|)|x - c|
\]
\[
< (2|c| + 1)|x - c|
\]
\[
< (2|c| + 1)\frac{\varepsilon}{2|c| + 1} = \varepsilon.
\]

**Example 3.1.6:** Define \( f : [0, 1) \to \mathbb{R} \) by
\[
f(x) := \begin{cases} x & \text{if } x > 0, \\ 1 & \text{if } x = 0. \end{cases}
\]
Then
\[
\lim_{x \to 0} f(x) = 0,
\]
even though \( f(0) = 1 \).

Proof: Let \( \varepsilon > 0 \) be given. Let \( \delta := \varepsilon \). For \( x \in [0, 1) \), \( x \neq 0 \), and \( |x - 0| < \delta \), we get
\[
|f(x) - 0| = |x| < \delta = \varepsilon.
\]

3.1.3 Sequential limits

Let us connect the limit as defined above with limits of sequences.

**Lemma 3.1.7.** Let \( S \subset \mathbb{R} \), let \( c \) be a cluster point of \( S \), let \( f : S \to \mathbb{R} \) be a function, and let \( L \in \mathbb{R} \).

Then \( f(x) \to L \) as \( x \to c \) if and only if for every sequence \( \{x_n\} \) of numbers such that \( x_n \in S \setminus \{c\} \) for all \( n \), and such that \( \lim x_n = c \), we have that the sequence \( \{f(x_n)\} \) converges to \( L \).

Proof. Suppose \( f(x) \to L \) as \( x \to c \), and \( \{x_n\} \) is a sequence such that \( x_n \in S \setminus \{c\} \) and \( \lim x_n = c \). We wish to show that \( \{f(x_n)\} \) converges to \( L \). Let \( \varepsilon > 0 \) be given. Find a \( \delta > 0 \) such that if \( x \in S \setminus \{c\} \) and \( |x - c| < \delta \), then \( |f(x) - L| < \varepsilon \). As \( \{x_n\} \) converges to \( c \), find an \( M \) such that for \( n \geq M \), we have that \( |x_n - c| < \delta \). Therefore, for \( n \geq M \),
\[
|f(x_n) - L| < \varepsilon.
\]
Thus \( \{f(x_n)\} \) converges to \( L \).

For the other direction, we use proof by contrapositive. Suppose it is not true that \( f(x) \to L \) as \( x \to c \). The negation of the definition is that there exists an \( \varepsilon > 0 \) such that for every \( \delta > 0 \) there exists an \( x \in S \setminus \{c\} \), where \( |x - c| < \delta \) and \( |f(x) - L| \geq \varepsilon \).

Let us use \( 1/n \) for \( \delta \) in the statement above to construct a sequence \( \{x_n\} \). We have that there exists an \( \varepsilon > 0 \) such that for every \( n \), there exists a point \( x_n \in S \setminus \{c\} \), where \( |x_n - c| < 1/n \) and \( |f(x_n) - L| \geq \varepsilon \). The sequence \( \{x_n\} \) just constructed converges to \( c \), but the sequence \( \{f(x_n)\} \) does not converge to \( L \). And we are done.
It is possible to strengthen the reverse direction of the lemma by simply stating that \( \{f(x_n)\} \) converges without requiring a specific limit. See Exercise 3.1.11.

**Example 3.1.8:** \( \lim_{x \to 0} \sin(1/x) \) does not exist, but \( \lim_{x \to 0} x \sin(1/x) = 0 \). See Figure 3.1.

![Figure 3.1](image)

**Figure 3.1:** Graphs of \( \sin(1/x) \) and \( x \sin(1/x) \). Note that the computer cannot properly graph \( \sin(1/x) \) near zero as it oscillates too fast.

Proof: We start with \( \sin(1/x) \). Define a sequence by \( x_n = \frac{1}{\pi n + \pi/2} \). It is not hard to see that \( \lim x_n = 0 \). Furthermore, 
\[
\sin(1/x_n) = \sin(\pi n + \pi/2) = (-1)^n.
\]
Therefore, \( \{\sin(1/x_n)\} \) does not converge. By Lemma 3.1.7, \( \lim_{x \to 0} \sin(1/x) \) does not exist.

Now consider \( x \sin(1/x) \). Let \( \{x_n\} \) be a sequence such that \( x_n \neq 0 \) for all \( n \), and such that \( \lim x_n = 0 \). Notice that \( |\sin(t)| \leq 1 \) for all \( t \in \mathbb{R} \). Therefore,
\[
|x_n \sin(1/x_n) - 0| = |x_n| |\sin(1/x_n)| \leq |x_n|.
\]
As \( x_n \) goes to 0, then \( |x_n| \) goes to zero, and hence \( \{x_n \sin(1/x_n)\} \) converges to zero. By Lemma 3.1.7, \( \lim_{x \to 0} x \sin(1/x) = 0 \).

Keep in mind the phrase “for every sequence” in the lemma. For example, take \( \sin(1/x) \) and the sequence given by \( x_n := 1/\pi n \). Then \( \{\sin(1/x_n)\} \) is the constant zero sequence, and therefore converges to zero, but the limit of \( \sin(1/x) \) as \( x \to 0 \) does not exist.

Using Lemma 3.1.7, we can start applying everything we know about sequential limits to limits of functions. Let us give a few important examples.

**Corollary 3.1.9.** Let \( S \subset \mathbb{R} \) and let \( c \) be a cluster point of \( S \). Suppose \( f : S \to \mathbb{R} \) and \( g : S \to \mathbb{R} \) are functions such that the limits of \( f(x) \) and \( g(x) \) as \( x \) goes to \( c \) both exist, and
\[
f(x) \leq g(x) \quad \text{for all } x \in S.
\]
Then
\[
\lim_{x \to c} f(x) \leq \lim_{x \to c} g(x).
\]
Proof. Take \( \{x_n\} \) be a sequence of numbers in \( S \setminus \{c\} \) that converges to \( c \). Let

\[
L_1 := \lim_{x \to c} f(x), \quad \text{and} \quad L_2 := \lim_{x \to c} g(x).
\]

Lemma 3.1.7 says that \( \{f(x_n)\} \) converges to \( L_1 \) and \( \{g(x_n)\} \) converges to \( L_2 \). We also have \( f(x_n) \leq g(x_n) \). We obtain \( L_1 \leq L_2 \) using Lemma 2.2.3.

By applying constant functions, we get the following corollary. The proof is left as an exercise.

**Corollary 3.1.10.** Let \( S \subset \mathbb{R} \) and let \( c \) be a cluster point of \( S \). Suppose \( f : S \to \mathbb{R} \) is a function such that the limit of \( f(x) \) as \( x \) goes to \( c \) exists. Suppose there are two real numbers \( a \) and \( b \) such that

\[
a \leq f(x) \leq b \quad \text{for all } x \in S.
\]

Then

\[
a \leq \lim_{x \to c} f(x) \leq b.
\]

Using Lemma 3.1.7 in the same way as above, we also get the following corollaries, whose proofs are again left as exercises.

**Corollary 3.1.11.** Let \( S \subset \mathbb{R} \) and let \( c \) be a cluster point of \( S \). Suppose \( f : S \to \mathbb{R} \), \( g : S \to \mathbb{R} \), and \( h : S \to \mathbb{R} \) are functions such that

\[
f(x) \leq g(x) \leq h(x) \quad \text{for all } x \in S.
\]

Suppose the limits of \( f(x) \) and \( h(x) \) as \( x \) goes to \( c \) both exist, and

\[
\lim_{x \to c} f(x) = \lim_{x \to c} h(x).
\]

Then the limit of \( g(x) \) as \( x \) goes to \( c \) exists and

\[
\lim_{x \to c} g(x) = \lim_{x \to c} f(x) = \lim_{x \to c} h(x).
\]

**Corollary 3.1.12.** Let \( S \subset \mathbb{R} \) and let \( c \) be a cluster point of \( S \). Suppose \( f : S \to \mathbb{R} \) and \( g : S \to \mathbb{R} \) are functions such that the limits of \( f(x) \) and \( g(x) \) as \( x \) goes to \( c \) both exist. Then

(i) \( \lim_{x \to c} (f(x) + g(x)) = \left( \lim_{x \to c} f(x) \right) + \left( \lim_{x \to c} g(x) \right) \).

(ii) \( \lim_{x \to c} (f(x) - g(x)) = \left( \lim_{x \to c} f(x) \right) - \left( \lim_{x \to c} g(x) \right) \).

(iii) \( \lim_{x \to c} (f(x)g(x)) = \left( \lim_{x \to c} f(x) \right) \left( \lim_{x \to c} g(x) \right) \).

(iv) If \( \lim_{x \to c} g(x) \neq 0 \), and \( g(x) \neq 0 \) for all \( x \in S \setminus \{c\} \), then

\[
\lim_{x \to c} \frac{f(x)}{g(x)} = \frac{\lim_{x \to c} f(x)}{\lim_{x \to c} g(x)}.
\]

**Corollary 3.1.13.** Let \( S \subset \mathbb{R} \) and let \( c \) be a cluster point of \( S \). Suppose \( f : S \to \mathbb{R} \) is a function such that the limit of \( f(x) \) as \( x \) goes to \( c \) exists. Then

\[
\lim_{x \to c} |f(x)| = \left| \lim_{x \to c} f(x) \right|.
\]
### 3.1.4 Limits of restrictions and one-sided limits

Sometimes we work with the function defined on a subset.

**Definition 3.1.14.** Let \( f: S \to \mathbb{R} \) be a function and \( A \subseteq S \). Define the function \( f|_A: A \to \mathbb{R} \) by

\[
f|_A(x) := f(x) \quad \text{for } x \in A.
\]

The function \( f|_A \) is called the **restriction** of \( f \) to \( A \).

The function \( f|_A \) is simply the function \( f \) taken on a smaller domain. The following proposition is the analogue of taking a tail of a sequence.

**Proposition 3.1.15.** Let \( S \subseteq \mathbb{R}, \ c \in \mathbb{R}, \) and let \( f: S \to \mathbb{R} \) be a function. Suppose \( A \subseteq S \) is such that there is some \( \alpha > 0 \) such that \( (A \setminus \{c\}) \cap (c - \alpha, c + \alpha) = (S \setminus \{c\}) \cap (c - \alpha, c + \alpha) \).

(i) The point \( c \) is a cluster point of \( A \) if and only if \( c \) is a cluster point of \( S \).

(ii) Supposing \( c \) is a cluster point of \( S \), then \( f(x) \to L \) as \( x \to c \) if and only if \( f|_A(x) \to L \) as \( x \to c \).

**Proof.** First, let \( c \) be a cluster point of \( A \). Since \( A \subseteq S \), then \( (A \setminus \{c\}) \cap (c - \epsilon, c + \epsilon) \) is nonempty for every \( \epsilon > 0 \), and \( (S \setminus \{c\}) \cap (c - \epsilon, c + \epsilon) \) is nonempty for every \( \epsilon > 0 \). Thus \( c \) is a cluster point of \( S \). Then for \( \epsilon > 0 \) such that \( \epsilon < \alpha \), we get that \( (A \setminus \{c\}) \cap (c - \epsilon, c + \epsilon) = (S \setminus \{c\}) \cap (c - \epsilon, c + \epsilon) \), which is nonempty. This is true for all \( \epsilon < \alpha \) and hence \( (A \setminus \{c\}) \cap (c - \epsilon, c + \epsilon) \) must be nonempty for all \( \epsilon > 0 \). Thus \( c \) is a cluster point of \( A \).

Now suppose \( c \) is a cluster point of \( S \) and \( f(x) \to L \) as \( x \to c \). That is, for every \( \epsilon > 0 \) there is a \( \delta > 0 \) such that if \( x \in S \setminus \{c\} \) and \( |x - c| < \delta \), then \( |f(x) - L| < \epsilon \). Because \( A \subseteq S \), if \( x \) is in \( A \setminus \{c\} \), then \( x \) is in \( S \setminus \{c\} \), and hence \( f|_A(x) \to L \) as \( x \to c \).

Finally suppose \( f|_A(x) \to L \) as \( x \to c \). For every \( \epsilon > 0 \) there is a \( \delta' > 0 \) such that if \( x \in A \setminus \{c\} \) and \( |x - c| < \delta' \), then \( |f|_A(x) - L| < \epsilon \). Take \( \delta := \min\{\delta', \alpha\} \). Now suppose \( x \in S \setminus \{c\} \) and \( |x - c| < \delta \). As \( |x - c| < \alpha \), then \( x \in A \setminus \{c\} \), and as \( |x - c| < \delta' \), we have \( |f(x) - L| = |f|_A(x) - L| < \epsilon \).

The hypothesis of the proposition is necessary. For an arbitrary restriction we generally only get implication in only one direction, see **Exercise 3.1.6**.

The usual notation for the limit is

\[
\lim_{x \to c} f(x) := \lim_{x \to c} f|_A(x).
\]

The most common use of restriction with respect to limits are the **one-sided limits**.

**Definition 3.1.16.** Let \( f: S \to \mathbb{R} \) be function and let \( c \) be a cluster point of \( S \cap (c, \infty) \). Then if the limit of the restriction of \( f \) to \( S \cap (c, \infty) \) as \( x \to c \) exists, define

\[
\lim_{x \to c^+} f(x) := \lim_{x \to c} f|_{S \cap (c, \infty)}(x).
\]

Similarly, if \( c \) is a cluster point of \( S \cap (-\infty, c) \) and the limit of the restriction as \( x \to c \) exists, define

\[
\lim_{x \to c^-} f(x) := \lim_{x \to c} f|_{S \cap (-\infty, c)}(x).
\]

*There are a plethora of notations for one-sided limits. E.g. for \( \lim_{x \to c^-} f(x) \) one sees \( \lim_{x \to c} f(x) \), \( \lim_{x \to c} f(x)_- \), or \( \lim_{x / c} f(x) \).*
The proposition above does not apply to one-sided limits. It is possible to have one-sided limits, but no limit at a point. For example, define \( f : \mathbb{R} \to \mathbb{R} \) by \( f(x) = 1 \) for \( x < 0 \) and \( f(x) = 0 \) for \( x \geq 0 \). We leave it to the reader to verify that

\[
\lim_{x \to 0^{-}} f(x) = 1, \quad \lim_{x \to 0^{+}} f(x) = 0, \quad \lim_{x \to 0} f(x) \ \text{does not exist.}
\]

We have the following replacement.

**Proposition 3.1.17.** Let \( S \subseteq \mathbb{R} \) be such that \( c \) is a cluster point of both \( S \cap (-\infty, c) \) and \( S \cap (c, \infty) \), let \( f : S \to \mathbb{R} \) be a function, and let \( L \in \mathbb{R} \). Then \( c \) is a cluster point of \( S \) and

\[
\lim_{x \to c^{-}} f(x) = L \quad \text{if and only if} \quad \lim_{x \to c^{+}} f(x) = \lim_{x \to c^{-}} f(x) = L.
\]

That is, a limit exists if both one-sided limits exist and are equal, and vice versa. The proof is a straightforward application of the definition of limit and is left as an exercise. The key point is that \( (S \cap (-\infty, c)) \cup (S \cap (c, \infty)) = S \setminus \{c\} \).

### 3.1.5 Exercises

**Exercise 3.1.1:** Find the limit (and prove it of course) or prove that the limit does not exist

\[
\begin{align*}
a) & \lim_{x \to c} \sqrt{x}, \text{ for } c \geq 0 \\
b) & \lim_{x \to c} x^2 + 1, \text{ for } c \in \mathbb{R} \\
c) & \lim_{x \to 0} x^2 \cos(1/x) \\
d) & \lim_{x \to 0} \sin(1/x) \cos(1/x) \\
e) & \lim_{x \to 0} \sin(x) \cos(1/x)
\end{align*}
\]

**Exercise 3.1.2:** Prove Corollary 3.1.10.

**Exercise 3.1.3:** Prove Corollary 3.1.11.

**Exercise 3.1.4:** Prove Corollary 3.1.12.

**Exercise 3.1.5:** Let \( A \subset S \). Show that if \( c \) is a cluster point of \( A \), then \( c \) is a cluster point of \( S \). Note the difference from Proposition 3.1.15.

**Exercise 3.1.6:** Let \( A \subset S \). Suppose \( c \) is a cluster point of \( A \) and it is also a cluster point of \( S \). Let \( f : S \to \mathbb{R} \) be a function. Show that if \( f(x) \to L \) as \( x \to c \), then \( f|_A(x) \to L \) as \( x \to c \). Note the difference from Proposition 3.1.15.

**Exercise 3.1.7:** Find an example of a function \( f : [-1, 1] \to \mathbb{R} \), where for \( A := [0, 1] \), we have \( f|_A(x) \to 0 \) as \( x \to 0 \), but the limit of \( f(x) \) as \( x \to 0 \) does not exist. Note why you cannot apply Proposition 3.1.15.

**Exercise 3.1.8:** Find example functions \( f \) and \( g \) such that the limit of neither \( f(x) \) nor \( g(x) \) exists as \( x \to 0 \), but such that the limit of \( f(x) + g(x) \) exists as \( x \to 0 \).

**Exercise 3.1.9:** Let \( c_1 \) be a cluster point of \( A \subset \mathbb{R} \) and \( c_2 \) be a cluster point of \( B \subset \mathbb{R} \). Suppose \( f : A \to B \) and \( g : B \to \mathbb{R} \) are functions such that \( f(x) \to c_2 \) as \( x \to c_1 \) and \( g(y) \to L \) as \( y \to c_2 \). If \( c_2 \in B \), also suppose that \( g(c_2) = L \). Let \( h(x) := g(f(x)) \) and show \( h(x) \to L \) as \( x \to c_1 \). Hint: Note that \( f(x) \) could equal \( c_2 \) for many \( x \in A \), see also Exercise 3.1.14.

**Exercise 3.1.10** (note*): Let \( c \) be a cluster point of \( A \subset \mathbb{R} \), and \( f : A \to \mathbb{R} \) be a function. Suppose for every sequence \( \{x_n\} \) in \( A \), such that \( \lim x_n = c \), the sequence \( \{f(x_n)\}_{n=1}^\infty \) is Cauchy. Prove that \( \lim_{x \to c} f(x) \) exists.

*This exercise is almost identical to the next one. It will be replaced in the next major edition.
Exercise 3.1.11: Prove the following stronger version of one direction of Lemma 3.1.7: Let \( S \subset \mathbb{R} \), \( c \) be a cluster point of \( S \), and \( f : S \to \mathbb{R} \) be a function. Suppose that for every sequence \( \{x_n\} \) in \( S \setminus \{c\} \) such that \( \lim x_n = c \) the sequence \( \{f(x_n)\} \) is convergent. Then show that the limit of \( f(x) \) as \( x \to c \) exists.

Exercise 3.1.12: Prove Proposition 3.1.17.

Exercise 3.1.13: Suppose \( S \subset \mathbb{R} \) and \( c \) is a cluster point of \( S \). Suppose \( f : S \to \mathbb{R} \) is bounded. Show that there exists a sequence \( \{x_n\} \) with \( x_n \in S \setminus \{c\} \) and \( \lim x_n = c \) such that \( \{f(x_n)\} \) converges.

Exercise 3.1.14 (Challenging): Show that the hypothesis that \( g(c_2) = L \) in Exercise 3.1.9 is necessary. That is, find \( f \) and \( g \) such that \( f(x) \to c_2 \) as \( x \to c_1 \) and \( g(y) \to L \) as \( y \to c_2 \), but \( g(f(x)) \) does not go to \( L \) as \( x \to c_1 \).

Exercise 3.1.15: Show that the condition of being a cluster point is necessary to have a reasonable definition of a limit. That is, suppose \( c \) is not a cluster point of \( S \subset \mathbb{R} \), and \( f : S \to \mathbb{R} \) is a function. Show that every \( L \) would satisfy the definition of limit at \( c \) without the condition on \( c \) being a cluster point.

Exercise 3.1.16:

a) Prove Corollary 3.1.13.

b) Find an example showing that the converse of the corollary does not hold.
3.2 Continuous functions

Note: 2–2.5 lectures

You undoubtedly heard of continuous functions in your schooling. A high-school criterion for this concept is that a function is continuous if we can draw its graph without lifting the pen from the paper. While that intuitive concept may be useful in simple situations, we require rigor. The following definition took three great mathematicians (Bolzano, Cauchy, and finally Weierstrass) to get correctly and its final form dates only to the late 1800s.

3.2.1 Definition and basic properties

Definition 3.2.1. Let $S \subset \mathbb{R}$, $c \in S$, and let $f : S \to \mathbb{R}$ be a function. We say that $f$ is continuous at $c$ if for every $\varepsilon > 0$ there is a $\delta > 0$ such that whenever $x \in S$ and $|x - c| < \delta$, then $|f(x) - f(c)| < \varepsilon$.

When $f : S \to \mathbb{R}$ is continuous at all $c \in S$, then we simply say $f$ is a continuous function.

Figure 3.2: For $|x - c| < \delta$, the graph of $f(x)$ should be within the gray region.

If $f$ is continuous for all $c \in A$, we say $f$ is continuous on $A \subset S$. A straightforward exercise (Exercise 3.2.7) shows that this implies that $f|_A$ is continuous, although the converse does not hold.

Continuity may be the most important definition to understand in analysis, and it is not an easy one. See Figure 3.2. Note that $\delta$ not only depends on $\varepsilon$, but also on $c$; we need not pick one $\delta$ for all $c \in S$. It is no accident that the definition of continuity is similar to the definition of a limit of a function. The main feature of continuous functions is that these are precisely the functions that behave nicely with limits.

Proposition 3.2.2. Consider a function $f : S \to \mathbb{R}$ defined on a set $S \subset \mathbb{R}$ and let $c \in S$. Then:

(i) If $c$ is not a cluster point of $S$, then $f$ is continuous at $c$.

(ii) If $c$ is a cluster point of $S$, then $f$ is continuous at $c$ if and only if the limit of $f(x)$ as $x \to c$ exists and

$$\lim_{x \to c} f(x) = f(c).$$

(iii) The function $f$ is continuous at $c$ if and only if for every sequence $\{x_n\}$ where $x_n \in S$ and $\lim x_n = c$, the sequence $\{f(x_n)\}$ converges to $f(c)$.
Then there exists an $\varepsilon$ such that $S \cap (c - \varepsilon, c + \varepsilon) = \{c\}$. For any $\varepsilon > 0$, simply pick this given $\delta$. The only $x \in S$ such that $|x - c| < \delta$ is $x = c$. Then $|f(x) - f(c)| = |f(c) - f(c)| = 0 < \varepsilon$.

Let us move to the second item. Suppose $c$ is a cluster point of $S$. Let us first suppose that $\lim_{x \to c} f(x) = f(c)$. Then for every $\varepsilon > 0$, there is a $\delta > 0$ such that if $x \in S \setminus \{c\}$ and $|x - c| < \delta$, then $|f(x) - f(c)| < \varepsilon$. Also $|f(c) - f(c)| = 0 < \varepsilon$, so the definition of continuity at $c$ is satisfied. On the other hand, suppose $f$ is continuous at $c$. For every $\varepsilon > 0$, there exists a $\delta > 0$ such that for $x \in S$ where $|x - c| < \delta$, we have $|f(x) - f(c)| < \varepsilon$. Then the statement is, of course, still true if $x \in S \setminus \{c\} \subset S$. Therefore, $\lim_{x \to c} f(x) = f(c)$.

For the third item, first suppose $f$ is continuous at $c$. Let $\{x_n\}$ be a sequence such that $x_n \in S$ and $\lim x_n = c$. Let $\varepsilon > 0$ be given. Find a $\delta > 0$ such that $|f(x) - f(c)| < \varepsilon$ for all $x \in S$ where $|x - c| < \delta$. Find an $M \in \mathbb{N}$ such that for $n \geq M$, we have $|x_n - c| < \delta$. Then for $n \geq M$, we have that $|f(x_n) - f(c)| < \varepsilon$, so $\{f(x_n)\}$ converges to $f(c)$.

We prove the other direction of the third item by contrapositive. Suppose $f$ is not continuous at $c$. Then there exists an $\varepsilon > 0$ such that for every $\delta > 0$, there exists an $x \in S$ such that $|x - c| < \delta$ and $|f(x) - f(c)| \geq \varepsilon$. Let us define a sequence $\{x_n\}$ as follows. Let $x_n \in S$ be such that $|x_n - c| < 1/n$ and $|f(x_n) - f(c)| \geq \varepsilon$. Now $\{x_n\}$ is a sequence of numbers in $S$ such that $\lim x_n = c$ and such that $|f(x_n) - f(c)| \geq \varepsilon$ for all $n \in \mathbb{N}$. Thus $\{f(x_n)\}$ does not converge to $f(c)$. It may or may not converge, but it definitely does not converge to $f(c)$.

The last item in the proposition is particularly powerful. It allows us to quickly apply what we know about limits of sequences to continuous functions and even to prove that certain functions are continuous. It can also be strengthened, see Exercise 3.2.13.

**Example 3.2.3:** The function $f : (0, \infty) \to \mathbb{R}$ defined by $f(x) := \frac{1}{x}$ is continuous.

Proof: Fix $c \in (0, \infty)$. Let $\{x_n\}$ be a sequence in $(0, \infty)$ such that $\lim x_n = c$. Then we know that $f(c) = \frac{1}{c} = \lim_{n \to \infty} x_n = \lim_{n \to \infty} f(x_n)$.

Thus $f$ is continuous at $c$. As $f$ is continuous at all $c \in (0, \infty)$, $f$ is continuous.

We have previously shown $\lim_{x \to c} x^2 = c^2$ directly. Therefore the function $x^2$ is continuous. We can use the continuity of algebraic operations with respect to limits of sequences, which we proved in the previous chapter, to prove a much more general result.

**Proposition 3.2.4.** Let $f : \mathbb{R} \to \mathbb{R}$ be a polynomial. That is $f(x) = a_d x^d + a_{d-1} x^{d-1} + \cdots + a_1 x + a_0$, for some constants $a_0, a_1, \ldots, a_d$. Then $f$ is continuous.

Proof. Fix $c \in \mathbb{R}$. Let $\{x_n\}$ be a sequence such that $\lim x_n = c$. Then

$$f(c) = a_d c^d + a_{d-1} c^{d-1} + \cdots + a_1 c + a_0 = a_d (\lim x_n)^d + a_{d-1} (\lim x_n)^{d-1} + \cdots + a_1 (\lim x_n) + a_0$$

$$= \lim_{n \to \infty} \left( a_d x_n^d + a_{d-1} x_n^{d-1} + \cdots + a_1 x_n + a_0 \right) = \lim_{n \to \infty} f(x_n).$$

Thus $f$ is continuous at $c$. As $f$ is continuous at all $c \in \mathbb{R}$, $f$ is continuous.
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By similar reasoning, or by appealing to Corollary 3.1.12, we can prove the following proposition. The proof is left as an exercise.

**Proposition 3.2.5.** Let \( f : S \rightarrow \mathbb{R} \) and \( g : S \rightarrow \mathbb{R} \) be functions continuous at \( c \in S \).

(i) The function \( h : S \rightarrow \mathbb{R} \) defined by \( h(x) := f(x) + g(x) \) is continuous at \( c \).

(ii) The function \( h : S \rightarrow \mathbb{R} \) defined by \( h(x) := f(x) - g(x) \) is continuous at \( c \).

(iii) The function \( h : S \rightarrow \mathbb{R} \) defined by \( h(x) := f(x)g(x) \) is continuous at \( c \).

(iv) If \( g(x) \neq 0 \) for all \( x \in S \), the function \( h : S \rightarrow \mathbb{R} \) defined by \( h(x) := \frac{f(x)}{g(x)} \) is continuous at \( c \).

**Example 3.2.6:** The functions \( \sin(x) \) and \( \cos(x) \) are continuous. In the following computations we use the sum-to-product trigonometric identities. We also use the simple facts that \( |\sin(x)| \leq |x| \), \( |\cos(x)| \leq 1 \), and \( |\sin(x)| \leq 1 \).

\[
|\sin(x) - \sin(c)| = \left| 2 \sin \left( \frac{x-c}{2} \right) \cos \left( \frac{x+c}{2} \right) \right| \\
= 2 \left| \sin \left( \frac{x-c}{2} \right) \right| \left| \cos \left( \frac{x+c}{2} \right) \right| \\
\leq 2 \left| \sin \left( \frac{x-c}{2} \right) \right| \\
\leq 2 \frac{|x-c|}{2} = |x-c|
\]

\[
|\cos(x) - \cos(c)| = \left| -2 \sin \left( \frac{x-c}{2} \right) \sin \left( \frac{x+c}{2} \right) \right| \\
= 2 \left| \sin \left( \frac{x-c}{2} \right) \right| \left| \sin \left( \frac{x+c}{2} \right) \right| \\
\leq 2 \left| \sin \left( \frac{x-c}{2} \right) \right| \\
\leq 2 \frac{|x-c|}{2} = |x-c|
\]

The claim that \( \sin \) and \( \cos \) are continuous follows by taking an arbitrary sequence \( \{x_n\} \) converging to \( c \), or by applying the definition of continuity directly. Details are left to the reader.

### 3.2.2 Composition of continuous functions

You probably already realized that one of the basic tools in constructing complicated functions out of simple ones is composition. Recall that for two functions \( f \) and \( g \), the composition \( f \circ g \) is defined by \((f \circ g)(x) := f(g(x))\). A composition of continuous functions is again continuous.

**Proposition 3.2.7.** Let \( A, B \subset \mathbb{R} \) and \( f : B \rightarrow \mathbb{R} \) and \( g : A \rightarrow B \) be functions. If \( g \) is continuous at \( c \in A \) and \( f \) is continuous at \( g(c) \), then \( f \circ g : A \rightarrow \mathbb{R} \) is continuous at \( c \).
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Proof. Let \( \{x_n\} \) be a sequence in \( A \) such that \( \lim x_n = c \). As \( g \) is continuous at \( c \), we have \( \{g(x_n)\} \) converges to \( g(c) \). As \( f \) is continuous at \( g(c) \), we have \( \{f(g(x_n))\} \) converges to \( f(g(c)) \). Thus \( f \circ g \) is continuous at \( c \).

\[ \square \]

**Example 3.2.8**: Claim: \((\sin(1/x))^2\) is a continuous function on \((0, \infty)\).

Proof: The function \( 1/x \) is continuous on \((0, \infty) \) and \( \sin(x) \) is continuous on \((0, \infty) \) (actually on \( \mathbb{R} \), but \((0, \infty) \) is the range for \( 1/x \)). Hence the composition \( \sin(1/x) \) is continuous. Also, \( x^2 \) is continuous on the interval \((-1, 1) \) (the range of \( \sin \)). Thus the composition \( (\sin(1/x))^2 \) is continuous on \((0, \infty) \).

### 3.2.3 Discontinuous functions

When \( f \) is not continuous at \( c \), we say \( f \) is *discontinuous* at \( c \), or that it has a *discontinuity* at \( c \). The following proposition is a useful test and follows immediately from third item of Proposition 3.2.2.

**Proposition 3.2.9.** Let \( f : S \to \mathbb{R} \) be a function and \( c \in S \). Suppose there exists a sequence \( \{x_n\} \), \( x_n \in S \), and \( \lim x_n = c \) such that \( \{f(x_n)\} \) does not converge to \( f(c) \). Then \( f \) is discontinuous at \( c \).

Again, saying that \( \{f(x_n)\} \) does not converge to \( f(c) \) means that it either does not converge at all, or it converges to something other than \( f(c) \).

**Example 3.2.10:** The function \( f : \mathbb{R} \to \mathbb{R} \) defined by

\[
f(x) := \begin{cases} 
-1 & \text{if } x < 0, \\
1 & \text{if } x \geq 0,
\end{cases}
\]

is not continuous at 0.

Proof: Take the sequence \( \{-1/n\} \), which converges to 0. Then \( f(-1/n) = -1 \) for every \( n \), and so \( \lim f(-1/n) = -1 \), but \( f(0) = 1 \). Thus the function is not continuous at 0. See Figure 3.3.

![Figure 3.3: Graph of the jump discontinuity. The values of \( f(-1/n) \) and \( f(0) \) are marked as black dots.](image)

Notice that \( f(1/n) = 1 \) for all \( n \in \mathbb{N} \). Hence, \( \lim f(1/n) = f(0) = 1 \). So \( \{f(x_n)\} \) may converge to \( f(0) \) for some specific sequence \( \{x_n\} \) going to 0, despite the function being discontinuous at 0.

Finally, consider \( f\left(\frac{(-1)^n}{n}\right) = (-1)^n \). This sequence diverges.

**Example 3.2.11:** For an extreme example, take the so-called *Dirichlet function*.*

\[
f(x) := \begin{cases} 
1 & \text{if } x \text{ is rational}, \\
0 & \text{if } x \text{ is irrational}.
\end{cases}
\]

The function \( f \) is discontinuous at all \( c \in \mathbb{R} \).

*Named after the German mathematician Johann Peter Gustav Lejeune Dirichlet (1805–1859).*
3.2. CONTINUOUS FUNCTIONS

Proof: Suppose \( c \) is rational. Take a sequence \( \{x_n\} \) of irrational numbers such that \( \lim x_n = c \) (why can we?). Then \( f(x_n) = 0 \) and so \( \lim f(x_n) = 0 \), but \( f(c) = 1 \). If \( c \) is irrational, take a sequence of rational numbers \( \{x_n\} \) that converges to \( c \) (why can we?). Then \( \lim f(x_n) = 1 \), but \( f(c) = 0 \).

Let us test the limits of our intuition. Can there exist a function continuous at all irrational numbers, but discontinuous at all rational numbers? There are rational numbers arbitrarily close to any irrational number. Perhaps strangely, the answer is yes, such a function exists. The following example is called the Thomae function* or the popcorn function.

Example 3.2.12: Define \( f : (0, 1) \to \mathbb{R} \) as

\[
f(x) := \begin{cases} 
1/k & \text{if } x = m/k, \text{ where } m, k \in \mathbb{N} \text{ and } m \text{ and } k \text{ have no common divisors}, \\
0 & \text{if } x \text{ is irrational}.
\end{cases}
\]

See the graph of \( f \) in Figure 3.4. We claim that \( f \) is continuous at all irrational \( c \) and discontinuous at all rational \( c \).

![Figure 3.4: Graph of the “popcorn function.”](image)

Proof: Suppose \( c = m/k \) is rational. Take a sequence of irrational numbers \( \{x_n\} \) such that \( \lim x_n = c \). Then \( \lim f(x_n) = \lim 0 = 0 \), but \( f(c) = 1/k \neq 0 \). So \( f \) is discontinuous at \( c \).

Now let \( c \) be irrational, so \( f(c) = 0 \). Take a sequence \( \{x_n\} \) in \( (0, 1) \) such that \( \lim x_n = c \). Given \( \varepsilon > 0 \), find \( K \in \mathbb{N} \) such that \( 1/k < \varepsilon \) by the Archimedean property. If \( m/k \in (0, 1) \) is in lowest terms (no common divisors), then \( 0 < m < k \). So there are only finitely many rational numbers in \( (0, 1) \) whose denominator \( k \) in lowest terms is less than \( K \). As \( \lim x_n = c \), every number not equal to \( c \) can appear at most finitely many times in \( \{x_n\} \). Hence, there is an \( M \) such that for \( n \geq M \), all the numbers \( x_n \) that are rational have a denominator larger than or equal to \( K \). Thus for \( n \geq M \),

\[
|f(x_n) - 0| = f(x_n) \leq 1/k < \varepsilon.
\]

Therefore, \( f \) is continuous at irrational \( c \).

Let us end on an easier example.

Example 3.2.13: Define \( g : \mathbb{R} \to \mathbb{R} \) by \( g(x) := 0 \) if \( x \neq 0 \) and \( g(0) := 1 \). Then \( g \) is not continuous at zero, but continuous everywhere else (why?). The point \( x = 0 \) is called a removable discontinuity. That is because if we would change the definition of \( g \), by insisting that \( g(0) \) be 0, we would obtain

*Named after the German mathematician Carl Johannes Thomae (1840–1921).
a continuous function. On the other hand, let \( f \) be the function of Example 3.2.10. Then \( f \) does not have a removable discontinuity at 0. No matter how we would define \( f(0) \) the function would still fail to be continuous. The difference is that \( \lim_{x \to 0} g(x) \) exists while \( \lim_{x \to 0} f(x) \) does not.

We stay with this example to show another phenomenon. Let \( A := \{0\} \), then \( g|_A \) is continuous (why?), while \( g \) is not continuous on \( A \). Similarly, if \( B := \mathbb{R} \setminus \{0\} \), then \( g|_B \) is also continuous, and \( g \) is in fact continuous on \( B \).

### 3.2.4 Exercises

**Exercise 3.2.1:** Using the definition of continuity directly prove that \( f : \mathbb{R} \to \mathbb{R} \) defined by \( f(x) := x^2 \) is continuous.

**Exercise 3.2.2:** Using the definition of continuity directly prove that \( f : (0, \infty) \to \mathbb{R} \) defined by \( f(x) := \frac{1}{x} \) is continuous.

**Exercise 3.2.3:** Let \( f : \mathbb{R} \to \mathbb{R} \) be defined by

\[
  f(x) := \begin{cases} 
    x & \text{if } x \text{ is rational,} \\
    x^2 & \text{if } x \text{ is irrational.}
  \end{cases}
\]

Using the definition of continuity directly prove that \( f \) is continuous at \( 1 \) and discontinuous at \( 2 \).

**Exercise 3.2.4:** Let \( f : \mathbb{R} \to \mathbb{R} \) be defined by

\[
  f(x) := \begin{cases} 
    \sin(1/x) & \text{if } x \neq 0, \\
    0 & \text{if } x = 0.
  \end{cases}
\]

Is \( f \) continuous? Prove your assertion.

**Exercise 3.2.5:** Let \( f : \mathbb{R} \to \mathbb{R} \) be defined by

\[
  f(x) := \begin{cases} 
    x \sin(1/x) & \text{if } x \neq 0, \\
    0 & \text{if } x = 0.
  \end{cases}
\]

Is \( f \) continuous? Prove your assertion.

**Exercise 3.2.6:** Prove Proposition 3.2.5.

**Exercise 3.2.7:** Prove the following statement. Let \( S \subseteq \mathbb{R} \) and \( A \subseteq S \). Let \( f : S \to \mathbb{R} \) be a continuous function. Then the restriction \( f|_A \) is continuous.

**Exercise 3.2.8:** Suppose \( S \subseteq \mathbb{R} \), such that \( (c - \alpha, c + \alpha) \subseteq S \) for some \( c \in \mathbb{R} \) and \( \alpha > 0 \). Let \( f : S \to \mathbb{R} \) be a function and \( A := (c - \alpha, c + \alpha) \). Prove that if \( f|_A \) is continuous at \( c \), then \( f \) is continuous at \( c \).

**Exercise 3.2.9:** Give an example of functions \( f : \mathbb{R} \to \mathbb{R} \) and \( g : \mathbb{R} \to \mathbb{R} \) such that the function \( h \) defined by \( h(x) := f(x) + g(x) \) is continuous, but \( f \) and \( g \) are not continuous. Can you find \( f \) and \( g \) that are nowhere continuous, but \( h \) is a continuous function?

**Exercise 3.2.10:** Let \( f : \mathbb{R} \to \mathbb{R} \) and \( g : \mathbb{R} \to \mathbb{R} \) be continuous functions. Suppose that for all rational numbers \( r \), \( f(r) = g(r) \). Show that \( f(x) = g(x) \) for all \( x \).
**Exercise 3.2.11:** Let $f : \mathbb{R} \to \mathbb{R}$ be continuous. Suppose $f(c) > 0$. Show that there exists an $\alpha > 0$ such that for all $x \in (c - \alpha, c + \alpha)$, we have $f(x) > 0$.

**Exercise 3.2.12:** Let $f : \mathbb{Z} \to \mathbb{R}$ be a function. Show that $f$ is continuous.

**Exercise 3.2.13:** Let $f : S \to \mathbb{R}$ be a function and $c \in S$, such that for every sequence $\{x_n\}$ in $S$ with $\lim x_n = c$, the sequence $\{f(x_n)\}$ converges. Show that $f$ is continuous at $c$.

**Exercise 3.2.14:** Suppose $f : [-1, 0] \to \mathbb{R}$ and $g : [0, 1] \to \mathbb{R}$ are continuous and $f(0) = g(0)$. Define $h : [-1, 1] \to \mathbb{R}$ by $h(x) := f(x)$ if $x \leq 0$ and $h(x) := g(x)$ if $x > 0$. Show that $h$ is continuous.

**Exercise 3.2.15:** Suppose $g : \mathbb{R} \to \mathbb{R}$ is a continuous function such that $g(0) = 0$, and suppose $f : \mathbb{R} \to \mathbb{R}$ is such that $|f(x) - f(y)| \leq g(x - y)$ for all $x$ and $y$. Show that $f$ is continuous.

**Exercise 3.2.16** (Challenging): Suppose $f : \mathbb{R} \to \mathbb{R}$ is continuous at $0$ and such that $f(x + y) = f(x) + f(y)$ for every $x$ and $y$. Show that $f(x) = ax$ for some $a \in \mathbb{R}$. Hint: Show that $f(nx) = nf(x)$, then show $f$ is continuous on $\mathbb{R}$. Then show that $f(x)/x = f(1)$ for all rational $x$.

**Exercise 3.2.17:** Suppose $S \subset \mathbb{R}$ and let $f : S \to \mathbb{R}$ and $g : S \to \mathbb{R}$ be continuous functions. Define $p : S \to \mathbb{R}$ by $p(x) := \max\{f(x), g(x)\}$ and $q : S \to \mathbb{R}$ by $q(x) := \min\{f(x), g(x)\}$. Prove that $p$ and $q$ are continuous.

**Exercise 3.2.18** (Challenging): Suppose $f : [-1, 1] \to \mathbb{R}$ is a function continuous at all $x \in [-1, 1] \setminus \{0\}$. Show that for every $\varepsilon$ such that $0 < \varepsilon < 1$, there exists a function $g : [-1, 1] \to \mathbb{R}$ continuous on all of $[-1, 1]$, such that $f(x) = g(x)$ for all $x \in [-1, -\varepsilon] \cup [\varepsilon, 1]$, and $|g(x)| \leq |f(x)|$ for all $x \in [-1, 1]$.

**Exercise 3.2.19** (Challenging): A function $f : I \to \mathbb{R}$ is convex if whenever $a \leq x \leq b$ for $a, x, b$ in $I$, we have $f(x) \leq f(a) \frac{x - a}{b - a} + f(b) \frac{x - b}{b - a}$. In other words, if the line drawn between $(a, f(a))$ and $(b, f(b))$ is above the graph of $f$.

a) Prove that if $I = (\alpha, \beta)$ an open interval and $f : I \to \mathbb{R}$ is convex, then $f$ is continuous.

b) Find an example of a convex $f : [0, 1] \to \mathbb{R}$ which is not continuous.
3.3 Min-max and intermediate value theorems

Note: 1.5 lectures

Continuous functions on closed and bounded intervals are quite well behaved.

3.3.1 Min-max or extreme value theorem

Recall a function \( f : [a, b] \to \mathbb{R} \) is bounded if there exists a \( B \in \mathbb{R} \) such that \(|f(x)| \leq B\) for all \( x \in [a, b] \). We have the following lemma.

**Lemma 3.3.1.** A continuous function \( f : [a, b] \to \mathbb{R} \) is bounded.

**Proof.** Let us prove this claim by contrapositive. Suppose \( f \) is not bounded. Then for each \( n \in \mathbb{N} \), there is an \( x_n \in [a, b] \), such that
\[
|f(x_n)| \geq n.
\]
The sequence \( \{x_n\} \) is bounded as \( a \leq x_n \leq b \). By the Boundedness theorem, there is a convergent subsequence \( \{x_{n_i}\} \). Let \( x := \lim_{i \to \infty} x_{n_i} \). Since \( a \leq x_{n_i} \leq b \) for all \( i \), then \( a \leq x \leq b \). The sequence \( \{f(x_{n_i})\} \) is not bounded as \(|f(x_{n_i})| \geq n_i \geq i\). Thus \( f \) is not continuous at \( x \) as
\[
f(x) = f\left(\lim_{i \to \infty} x_{n_i}\right), \quad \text{but} \quad \lim_{i \to \infty} f(x_{n_i}) \text{ does not exist.}
\]

Notice a key point of the proof. Boundedness of \([a, b]\) allows us to use Bolzano–Weierstrass, while the fact that it is closed gives us that the limit is back in \([a, b]\). The technique is a common one: Find a sequence with a certain property, then use Bolzano–Weierstrass to make such a sequence that also converges.

Recall from calculus that \( f : S \to \mathbb{R} \) achieves an absolute minimum at \( c \in S \) if
\[
f(x) \geq f(c) \quad \text{for all } x \in S.
\]
On the other hand, \( f \) achieves an absolute maximum at \( c \in S \) if
\[
f(x) \leq f(c) \quad \text{for all } x \in S.
\]
If such a \( c \in S \) exists, then \( f \) achieves an absolute minimum (resp. absolute maximum) on \( S \).

---

**Figure 3.5:** \( f : [a, b] \to \mathbb{R} \) achieves an absolute maximum \( f(c) \) at \( c \), and an absolute minimum \( f(d) \) at \( d \).

---

If \( S \) is a closed and bounded interval, then a continuous \( f \) must achieve an absolute minimum and an absolute maximum on \( S \).
3.3. MIN-MAX AND INTERMEDIATE VALUE THEOREMS

**Theorem 3.3.2** (Minimum-maximum theorem / Extreme value theorem). A continuous function \( f: [a, b] \to \mathbb{R} \) on a closed and bounded interval \([a, b]\) achieves both an absolute minimum and an absolute maximum on \([a, b]\).

**Proof.** The lemma says that \( f \) is bounded, so the set \( f([a, b]) = \{ f(x) : x \in [a, b] \} \) has a supremum and an infimum. There exist sequences in the set \( f([a, b]) \) that approach its supremum and its infimum. That is, there are sequences \( \{ f(x_n) \} \) and \( \{ f(y_n) \} \), where \( x_n \) and \( y_n \) are in \([a, b]\), such that

\[
\lim_{n \to \infty} f(x_n) = \inf f([a, b]) \quad \text{and} \quad \lim_{n \to \infty} f(y_n) = \sup f([a, b]).
\]

We are not done yet; we need to find where the minima and the maxima are. The problem is that the sequences \( \{x_n\} \) and \( \{y_n\} \) need not converge. We know \( \{x_n\} \) and \( \{y_n\} \) are bounded (their elements belong to a bounded interval \([a, b]\)). Apply the Bolzano–Weierstrass theorem, to find convergent subsequences \( \{x_{n_i}\} \) and \( \{y_{m_i}\} \). Let

\[
x := \lim_{i \to \infty} x_{n_i} \quad \text{and} \quad y := \lim_{i \to \infty} y_{m_i}.
\]

As \( a \leq x_{n_i} \leq b \) for all \( i \), we have \( a \leq x \leq b \). Similarly, \( a \leq y \leq b \). So \( x \) and \( y \) are in \([a, b]\). A limit of a subsequence is the same as the limit of the sequence, and we can take a limit past the continuous function \( f \):

\[
\inf f([a, b]) = \lim_{n \to \infty} f(x_n) = \lim_{i \to \infty} f(x_{n_i}) = f\left( \lim_{i \to \infty} x_{n_i} \right) = f(x).
\]

Similarly,

\[
\sup f([a, b]) = \lim_{n \to \infty} f(y_n) = \lim_{i \to \infty} f(y_{m_i}) = f\left( \lim_{i \to \infty} y_{m_i} \right) = f(y).
\]

Therefore, \( f \) achieves an absolute minimum at \( x \) and \( f \) achieves an absolute maximum at \( y \). \( \square \)

**Example 3.3.3:** The function \( f(x) := x^2 + 1 \) defined on the interval \([-1, 2]\) achieves a minimum at \( x = 0 \) when \( f(0) = 1 \). It achieves a maximum at \( x = 2 \) where \( f(2) = 5 \). Do note that the domain of definition matters. If we instead took the domain to be \([-10, 10]\), then \( x = 2 \) would no longer be a maximum of \( f \). Instead the maximum would be achieved at either \( x = 10 \) or \( x = -10 \).

We show by examples that the different hypotheses of the theorem are truly needed.

**Example 3.3.4:** The function \( f(x) := x \), defined on the whole real line, achieves neither a minimum, nor a maximum. So it is important that we are looking at a bounded interval.

**Example 3.3.5:** The function \( f(x) := 1/x \), defined on \((0, 1)\) achieves neither a minimum, nor a maximum. The values of the function are unbounded as we approach 0. Also as we approach \( x = 1 \), the values of the function approach 1, but \( f(x) > 1 \) for all \( x \in (0, 1) \). There is no \( x \in (0, 1) \) such that \( f(x) = 1 \). So it is important that we are looking at a closed interval.

**Example 3.3.6:** Continuity is important. Define \( f: [0, 1] \to \mathbb{R} \) by \( f(x) := 1/x \) for \( x > 0 \) and let \( f(0) := 0 \). The function does not achieve a maximum. The problem is that the function is not continuous at 0.
Bolzano’s intermediate value theorem

Bolzano’s intermediate value theorem is one of the cornerstones of analysis. It is sometimes only called the intermediate value theorem, or just Bolzano’s theorem. To prove Bolzano’s theorem we prove the following simpler lemma.

**Lemma 3.3.7.** Let \( f: [a, b] \rightarrow \mathbb{R} \) be a continuous function. Suppose \( f(a) < 0 \) and \( f(b) > 0 \). Then there exists a number \( c \in (a, b) \) such that \( f(c) = 0 \).

**Proof.** We define two sequences \( \{a_n\} \) and \( \{b_n\} \) inductively:

(i) Let \( a_1 := a \) and \( b_1 := b \).

(ii) If \( f\left(\frac{a_n + b_n}{2}\right) \geq 0 \), let \( a_{n+1} := a_n \) and \( b_{n+1} := \frac{a_n + b_n}{2} \).

(iii) If \( f\left(\frac{a_n + b_n}{2}\right) < 0 \), let \( a_{n+1} := \frac{a_n + b_n}{2} \) and \( b_{n+1} := b_n \).

See Figure 3.6 for an example defining the first five steps. If \( a_n < b_n \), then \( a_n < \frac{a_n + b_n}{2} < b_n \). So \( a_{n+1} < b_{n+1} \). Thus by induction \( a_n < b_n \) for all \( n \). Furthermore, \( a_n \leq a_{n+1} \) and \( b_n \geq b_{n+1} \) for all \( n \), that is, the sequences are monotone. As \( a_n < b_n \leq b_1 = b \) and \( b_n \geq a_n \geq a_1 = a \) for all \( n \), the sequences are also bounded. Therefore, the sequences converge. Let \( c := \lim a_n \) and \( d := \lim b_n \), where also \( a \leq c \leq d \leq b \). We need to show that \( c = d \). Notice

\[
b_{n+1} - a_{n+1} = \frac{b_n - a_n}{2}.
\]

By induction,

\[
b_n - a_n = \frac{b_1 - a_1}{2^{n-1}} = 2^{1-n}(b-a).
\]
As $2^{1-n}(b - a)$ converges to zero, we take the limit as $n$ goes to infinity to get

$$d - c = \lim_{n \to \infty} (b_n - a_n) = \lim_{n \to \infty} 2^{1-n}(b - a) = 0.$$  

In other words, $d = c$.

By construction, for all $n$, we have

$$f(a_n) < 0 \quad \text{and} \quad f(b_n) \geq 0.$$  

Since $\lim a_n = \lim b_n = c$ and as $f$ is continuous, we may take limits in those inequalities:

$$f(c) = \lim f(a_n) \leq 0 \quad \text{and} \quad f(c) = \lim f(b_n) \geq 0.$$  

As $f(c) \geq 0$ and $f(c) \leq 0$, we conclude $f(c) = 0$. Thus also $c \neq a$ and $c \neq b$, so $a < c < b$.  

**Theorem 3.3.8** (Bolzano’s intermediate value theorem). Let $f : [a, b] \to \mathbb{R}$ be a continuous function. Suppose $y \in \mathbb{R}$ is such that $f(a) < y < f(b)$ or $f(a) > y > f(b)$. Then there exists a $c \in (a, b)$ such that $f(c) = y$.

The theorem says that a continuous function on a closed interval achieves all the values between the values at the endpoints.

**Proof.** If $f(a) < y < f(b)$, then define $g(x) := f(x) - y$. Then $g(a) < 0$ and $g(b) > 0$, and we apply **Lemma 3.3.7** to $g$ to find $c$. If $g(c) = 0$, then $f(c) = y$.

Similarly, if $f(a) > y > f(b)$, then define $g(x) := y - f(x)$. Again, $g(a) < 0$ and $g(b) > 0$, and we apply **Lemma 3.3.7** to find $c$. As before, if $g(c) = 0$, then $f(c) = y$.  

If a function is continuous, then the restriction to a subset is continuous; if $f : S \to \mathbb{R}$ is continuous and $[a, b] \subset S$, then $f|_{[a,b]}$ is also continuous. We generally apply the theorem to a function continuous on some large set $S$, but we restrict our attention to an interval.

The proof of the lemma tells us how to find the root $c$. The proof is not only useful for us pure mathematicians, it is a useful idea in applied mathematics, where it is called the _bisection method_.

**Example 3.3.9** (Bisection method): The polynomial $f(x) := x^3 - 2x^2 + x - 1$ has a real root in $(1, 2)$. We simply notice that $f(1) = -1$ and $f(2) = 1$. Hence there must exist a point $c \in (1, 2)$ such that $f(c) = 0$. To find a better approximation of the root we follow the proof of **Lemma 3.3.7**. We look at 1.5 and find that $f(1.5) = -0.625$. Therefore, there is a root of the polynomial in $(1.5, 2)$. Next we look at 1.75 and note that $f(1.75) \approx -0.016$. Hence there is a root of $f$ in $(1.75, 2)$. Next we look at 1.875 and find that $f(1.875) \approx 0.44$, thus there is a root in $(1.75, 1.875)$. We follow this procedure until we gain sufficient precision. In fact, the root is at $c \approx 1.7549$.

The technique above is the simplest method of finding roots of polynomials, which is perhaps the most common problem in applied mathematics. In general, finding roots is hard to do quickly, precisely, and automatically. There are other, faster methods of finding roots of polynomials, such as Newton’s method. One advantage of the method above is its simplicity. The moment we find an initial interval where the intermediate value theorem applies, we are guaranteed to find a root up to a desired precision in finitely many steps. Furthermore, the bisection method finds roots of any continuous function, not just a polynomial.
The theorem guarantees at least one \( c \) such that \( f(c) = y \), but there may be many different roots of the equation \( f(c) = y \). If we follow the procedure of the proof, we are guaranteed to find approximations to one such root. We need to work harder to find any other roots.

Polynomials of even degree may not have any real roots. There is no real number \( x \) such that \( x^2 + 1 = 0 \). Odd polynomials, on the other hand, always have at least one real root.

**Proposition 3.3.10.** Let \( f(x) \) be a polynomial of odd degree. Then \( f \) has a real root.

**Proof.** Suppose \( f \) is a polynomial of odd degree \( d \). We write

\[
f(x) = a_d x^d + a_{d-1} x^{d-1} + \cdots + a_1 x + a_0,
\]

where \( a_d \neq 0 \). We divide by \( a_d \) to obtain a monic polynomial*

\[
g(x) := x^d + b_{d-1} x^{d-1} + \cdots + b_1 x + b_0,
\]

where \( b_k = a_k / a_d \). Let us show that \( g(n) \) is positive for some large \( n \in \mathbb{N} \). We first compare the highest order term with the rest:

\[
\left| \frac{b_{d-1} n^{d-1} + \cdots + b_1 n + b_0}{n^d} \right| = \frac{b_{d-1} n^{d-1} + \cdots + b_1 n + b_0}{n^d} \\
\leq \frac{|b_{d-1}| n^{d-1} + \cdots + |b_1| n + |b_0|}{n^d} \\
\leq \frac{|b_{d-1}| n^{d-1} + \cdots + |b_1| n^{d-1} + |b_0| n^{d-1}}{n^d} \\
= n^{d-1} \left( |b_{d-1}| + \cdots + |b_1| + |b_0| \right) \\
= \frac{1}{n} \left( |b_{d-1}| + \cdots + |b_1| + |b_0| \right).
\]

Therefore,

\[
\lim_{n \to \infty} \frac{b_{d-1} n^{d-1} + \cdots + b_1 n + b_0}{n^d} = 0.
\]

Thus there exists an \( M \in \mathbb{N} \) such that

\[
\frac{b_{d-1} M^{d-1} + \cdots + b_1 M + b_0}{M^d} < 1,
\]

which implies

\[
-(b_{d-1} M^{d-1} + \cdots + b_1 M + b_0) < M^d.
\]

Therefore, \( g(M) > 0 \).

Next, consider \( g(-n) \) for \( n \in \mathbb{N} \). By a similar argument, there exists a \( K \in \mathbb{N} \) such that \( b_{d-1} (-K)^{d-1} + \cdots + b_1 (-K) + b_0 < K^d \) and therefore \( g(-K) < 0 \) (see Exercise 3.3.5). In the proof, make sure you use the fact that \( d \) is odd. In particular, if \( d \) is odd, then \((-n)^d = -(n^d)\).

We appeal to the intermediate value theorem to find a \( c \in [-K, M] \), such that \( g(c) = 0 \). As \( g(x) = f(x) / a_d \), then \( f(c) = 0 \), and the proof is done. \( \square \)

*The word monic means that the coefficient of \( x^d \) is 1.
Example 3.3.11: Interestingly, there exist discontinuous functions with the intermediate value property. The function

\[ f(x) := \begin{cases} 
  \sin(1/x) & \text{if } x \neq 0, \\
  0 & \text{if } x = 0,
\end{cases} \]

is not continuous at 0; however, \( f \) has the intermediate value property: Whenever \( a < b \) and \( y \) is such that \( f(a) < y < f(b) \) or \( f(a) > y > f(b) \), there exists a \( c \) such that \( f(c) = y \). Proof is left as Exercise 3.3.4.

The intermediate value theorem says that if \( f : [a, b] \to \mathbb{R} \) is continuous, then \( f([a, b]) \) contains all the values between \( f(a) \) and \( f(b) \). In fact, more is true. Combining all the results of this section one can prove the following useful corollary whose proof is left as an exercise.

Corollary 3.3.12. If \( f : [a, b] \to \mathbb{R} \) is continuous, then the direct image \( f([a, b]) \) is a closed and bounded interval or a single number.

3.3.3 Exercises

Exercise 3.3.1: Find an example of a discontinuous function \( f : [0, 1] \to \mathbb{R} \) where the conclusion of the intermediate value theorem fails.

Exercise 3.3.2: Find an example of a bounded discontinuous function \( f : [0, 1] \to \mathbb{R} \) that has neither an absolute minimum nor an absolute maximum.

Exercise 3.3.3: Let \( f : (0, 1) \to \mathbb{R} \) be a continuous function such that \( \lim_{x \to 0} f(x) = \lim_{x \to 1} f(x) = 0 \). Show that \( f \) achieves either an absolute minimum or an absolute maximum on \((0, 1)\) (but perhaps not both).

Exercise 3.3.4: Let

\[ f(x) := \begin{cases} 
  \sin(1/x) & \text{if } x \neq 0, \\
  0 & \text{if } x = 0.
\end{cases} \]

Show that \( f \) has the intermediate value property. That is, whenever \( a < b \), if there exists a \( y \) such that \( f(a) < y < f(b) \) or \( f(a) > y > f(b) \), then there exists a \( c \in (a, b) \) such that \( f(c) = y \).

Exercise 3.3.5: Suppose \( g(x) \) is a monic polynomial of odd degree \( d \), that is,

\[ g(x) = x^d + b_{d-1}x^{d-1} + \cdots + b_1x + b_0, \]

for some real numbers \( b_0, b_1, \ldots, b_{d-1} \). Show that there exists a \( K \in \mathbb{N} \) such that \( g(-K) < 0 \). Hint: Make sure to use the fact that \( d \) is odd. You will have to use that \((-n)^d = -(n^d)\).

Exercise 3.3.6: Suppose \( g(x) \) is a monic polynomial of positive even degree \( d \), that is,

\[ g(x) = x^d + b_{d-1}x^{d-1} + \cdots + b_1x + b_0, \]

for some real numbers \( b_0, b_1, \ldots, b_{d-1} \). Suppose \( g(0) < 0 \). Show that \( g \) has at least two distinct real roots.

Exercise 3.3.7: Prove Corollary 3.3.12: Suppose \( f : [a, b] \to \mathbb{R} \) is a continuous function. Prove that the direct image \( f([a, b]) \) is a closed and bounded interval or a single number.
**Exercise 3.3.8:** Suppose \( f : \mathbb{R} \to \mathbb{R} \) is continuous and periodic with period \( P > 0 \). That is, \( f(x + P) = f(x) \) for all \( x \in \mathbb{R} \). Show that \( f \) achieves an absolute minimum and an absolute maximum.

**Exercise 3.3.9** (Challenging): Suppose \( f(x) \) is a bounded polynomial, in other words, there is an \( M \) such that \( |f(x)| \leq M \) for all \( x \in \mathbb{R} \). Prove that \( f \) must be a constant.

**Exercise 3.3.10:** Suppose \( f : [0, 1] \to [0, 1] \) is continuous. Show that \( f \) has a fixed point, in other words, show that there exists an \( x \in [0, 1] \) such that \( f(x) = x \).

**Exercise 3.3.11:** Find an example of a continuous bounded function \( f : \mathbb{R} \to \mathbb{R} \) that does not achieve an absolute minimum nor an absolute maximum on \( \mathbb{R} \).

**Exercise 3.3.12:** Suppose \( f : \mathbb{R} \to \mathbb{R} \) is a continuous function such that \( x \leq f(x) \leq x + 1 \) for all \( x \in \mathbb{R} \). Find \( f(\mathbb{R}) \).

**Exercise 3.3.13:** True/False, prove or find a counterexample. If \( f : \mathbb{R} \to \mathbb{R} \) is a continuous function such that \( f|_{\mathbb{Z}} \) is bounded, then \( f \) is bounded.

**Exercise 3.3.14:** Suppose \( f : [0, 1] \to (0, 1) \) is a bijection. Prove that \( f \) is not continuous.

**Exercise 3.3.15:** Suppose \( f : \mathbb{R} \to \mathbb{R} \) is continuous.

a) Prove that if there is a \( c \) such that \( f(c)f(-c) < 0 \), then there is a \( d \in \mathbb{R} \) such that \( f(d) = 0 \).

b) Find a continuous function \( f \) such that \( f(\mathbb{R}) = \mathbb{R} \), but \( f(x)f(-x) \geq 0 \) for all \( x \in \mathbb{R} \).

**Exercise 3.3.16:** Suppose \( g(x) \) is a monic polynomial of even degree \( d \), that is,

\[
g(x) = x^d + b_d x^{d-1} + \cdots + b_1 x + b_0,
\]

for some real numbers \( b_0, b_1, \ldots, b_d \). Show that \( g \) achieves an absolute minimum on \( \mathbb{R} \).

**Exercise 3.3.17:** Suppose \( f(x) \) is a polynomial of degree \( d \) and \( f(\mathbb{R}) = \mathbb{R} \). Show that \( d \) is odd.
3.4 Uniform continuity

Note: 1.5–2 lectures (continuous extension can be optional)

3.4.1 Uniform continuity

We made a fuss of saying that the $\delta$ in the definition of continuity depended on the point $c$. There are situations when it is advantageous to have a $\delta$ independent of any point, and so we give a name to this concept.

Definition 3.4.1. Let $S \subseteq \mathbb{R}$, and let $f: S \to \mathbb{R}$ be a function. Suppose for every $\varepsilon > 0$ there exists a $\delta > 0$ such that whenever $x, c \in S$ and $|x - c| < \delta$, then $|f(x) - f(c)| < \varepsilon$. Then we say $f$ is uniformly continuous.

A uniformly continuous function must be continuous. The only difference in the definitions is that in uniform continuity, for a given $\varepsilon > 0$ we pick a $\delta > 0$ that works for all $c \in S$. That is, $\delta$ can no longer depend on $c$; it only depends on $\varepsilon$. The domain of definition of the function makes a difference now. A function that is not uniformly continuous on a larger set, may be uniformly continuous when restricted to a smaller set. We will say uniformly continuous on $X$ to mean that $f$ restricted to $X$ is uniformly continuous, or perhaps to just emphasize the domain. Note that $x$ and $c$ are not treated any differently in this definition.

Example 3.4.2: $f: [0, 1] \to \mathbb{R}$, defined by $f(x) := x^2$ is uniformly continuous.

Proof: Note that $0 \leq x, c \leq 1$. Then

$$|x^2 - c^2| = |x + c| |x - c| \leq (|x| + |c|) |x - c| \leq (1 + 1) |x - c|.$$ 

Therefore, given $\varepsilon > 0$, let $\delta := \varepsilon/2$. If $|x - c| < \delta$, then $|x^2 - c^2| < \varepsilon$.

On the other hand, $g: \mathbb{R} \to \mathbb{R}$, defined by $g(x) := x^2$ is not uniformly continuous.

Proof: Suppose it is uniformly continuous, then for every $\varepsilon > 0$, there would exist a $\delta > 0$ such that if $|x - c| < \delta$, then $|x^2 - c^2| < \varepsilon$. Take $x > 0$ and let $c := x + \delta/2$. Write

$$\varepsilon > |x^2 - c^2| = |x + c| |x - c| = (2x + \delta/2)\delta/2 \geq \delta x.$$ 

Therefore, $x < \varepsilon/\delta$ for all $x > 0$, which is a contradiction.

Example 3.4.3: The function $f: (0, 1) \to \mathbb{R}$, defined by $f(x) := 1/x$ is not uniformly continuous.

Proof: Given $\varepsilon > 0$, then $\varepsilon > |1/x - 1/y|$ holds if and only if

$$\varepsilon > |1/x - 1/y| = \frac{|y - x|}{|xy|} = \frac{|y - x|}{xy},$$

or

$$|x - y| < xy\varepsilon.$$ 

Suppose $\varepsilon < 1$, and we wish to see if a small $\delta > 0$ would work. If $x \in (0, 1)$ and $y = x + \delta/2 \in (0, 1)$, then $|x - y| = \delta/2 < \delta$. We plug $y$ into the inequality above to get $\delta/2 < x(x + \delta/2)\varepsilon < x$. If the definition of uniform continuity is satisfied, then the inequality $\delta/2 < x$ holds for all $x > 0$. But then $\delta \leq 0$. Therefore, there is no single $\delta > 0$ that works for all points.
The examples show that if $f$ is defined on an interval that is either not closed or not bounded, then $f$ can be continuous, but not uniformly continuous. For a closed and bounded interval $[a, b]$, we can, however, make the following statement.

**Theorem 3.4.4.** Let $f: [a, b] \to \mathbb{R}$ be a continuous function. Then $f$ is uniformly continuous.

**Proof.** We prove the statement by contrapositive. Suppose $f$ is not uniformly continuous. We will prove that there is some $c \in [a, b]$ where $f$ is not continuous. Let us negate the definition of uniformly continuous. There exists an $\varepsilon > 0$ such that for every $\delta > 0$, there exist points $x, y$ in $[a, b]$ with $|x - y| < \delta$ and $|f(x) - f(y)| \geq \varepsilon$.

So for the $\varepsilon > 0$ above, we find sequences $\{x_n\}$ and $\{y_n\}$ such that $|x_n - y_n| < 1/n$ and such that $|f(x_n) - f(y_n)| \geq \varepsilon$. By Bolzano–Weierstrass, there exists a convergent subsequence $\{x_{n_k}\}$. Let $c := \lim x_{n_k}$. As $a \leq x_{n_k} \leq b$ for all $k$, we have $a \leq c \leq b$. Estimate

$$|y_{n_k} - c| = |y_{n_k} - x_{n_k} + x_{n_k} - c| \leq |y_{n_k} - x_{n_k}| + |x_{n_k} - c| < 1/n_k + |x_{n_k} - c|.$$ 

As $1/n_k$ and $|x_{n_k} - c|$ both go to zero when $k$ goes to infinity, $\{y_{n_k}\}$ converges and the limit is $c$. We now show that $f$ is not continuous at $c$. Estimate

$$|f(x_{n_k}) - f(c)| = |f(x_{n_k}) - f(y_{n_k}) + f(y_{n_k}) - f(c)| \geq |f(x_{n_k}) - f(y_{n_k})| - |f(y_{n_k}) - f(c)| \geq \varepsilon - |f(y_{n_k}) - f(c)|.$$ 

Or in other words,

$$|f(x_{n_k}) - f(c)| + |f(y_{n_k}) - f(c)| \geq \varepsilon.$$

At least one of the sequences $\{f(x_{n_k})\}$ or $\{f(y_{n_k})\}$ cannot converge to $f(c)$, otherwise the left-hand side of the inequality would go to zero while the right-hand side is positive. Thus $f$ cannot be continuous at $c$. 

As before, note what is key in the proof: We can apply Bolzano–Weierstrass because the interval $[a, b]$ is bounded, and the limit of the subsequence is back in $[a, b]$ because the interval is closed.

### 3.4.2 Continuous extension

Before we get to continuous extension, we show the following useful lemma. It says that uniformly continuous functions behave nicely with respect to Cauchy sequences. The new issue here is that for a Cauchy sequence we no longer know where the limit ends up; it may not end up in the domain of the function.

**Lemma 3.4.5.** Let $f: S \to \mathbb{R}$ be a uniformly continuous function. Let $\{x_n\}$ be a Cauchy sequence in $S$. Then $\{f(x_n)\}$ is Cauchy.

**Proof.** Let $\varepsilon > 0$ be given. There is a $\delta > 0$ such that $|f(x) - f(y)| < \varepsilon$ whenever $x, y \in S$ and $|x - y| < \delta$. Find an $M \in \mathbb{N}$ such that for all $n, k \geq M$, we have $|x_n - x_k| < \delta$. Then for all $n, k \geq M$, we have $|f(x_n) - f(x_k)| < \varepsilon$. 

$\square$
3.4. UNIFORM CONTINUITY

An application of the lemma above is the following extension result. It says that a function on an open interval is uniformly continuous if and only if it can be extended to a continuous function on the closed interval.

**Proposition 3.4.6.** A function \( f: (a, b) \to \mathbb{R} \) is uniformly continuous if and only if the limits

\[
L_a := \lim_{x \to a} f(x) \quad \text{and} \quad L_b := \lim_{x \to b} f(x)
\]

exist and the function \( \tilde{f}: [a, b] \to \mathbb{R} \) defined by

\[
\tilde{f}(x) := \begin{cases} f(x) & \text{if } x \in (a, b), \\
L_a & \text{if } x = a, \\
L_b & \text{if } x = b,
\end{cases}
\]

is continuous.

**Proof.** One direction is not difficult. If \( \tilde{f} \) is continuous, then it is uniformly continuous by Theorem 3.4.4. As \( f \) is the restriction of \( \tilde{f} \) to \( (a, b) \), then \( f \) is also uniformly continuous (easy exercise).

Now suppose \( f \) is uniformly continuous. We must first show that the limits \( L_a \) and \( L_b \) exist. Let us concentrate on \( L_a \). Take a sequence \( \{x_n\} \) in \( (a, b) \) such that \( \lim x_n = a \). The sequence \( \{x_n\} \) is Cauchy, so by Lemma 3.4.5 the sequence \( \{f(x_n)\} \) is Cauchy and thus convergent. We have some number \( L_1 := \lim f(x_n) \). Take another sequence \( \{y_n\} \) in \( (a, b) \) such that \( \lim y_n = a \). By the same reasoning we get \( L_2 := \lim f(y_n) \). If we show that \( L_1 = L_2 \), then the limit \( L_a = \lim_{x \to a} f(x) \) exists.

Let \( \varepsilon > 0 \) be given. Find \( \delta > 0 \) such that \( |x - y| < \delta \) implies \( |f(x) - f(y)| < \varepsilon/3 \). Find \( M \in \mathbb{N} \) such that for \( n \geq M \), we have \( |a - x_n| < \delta/2 \), \( |a - y_n| < \delta/2 \), \( |f(x_n) - L_1| < \varepsilon/3 \), and \( |f(y_n) - L_2| < \varepsilon/3 \). Then for \( n \geq M \),

\[
|x_n - y_n| = |x_n - a + a - y_n| \leq |x_n - a| + |a - y_n| < \delta/2 + \delta/2 = \delta.
\]

So

\[
|L_1 - L_2| = |L_1 - f(x_n) + f(x_n) - f(y_n) + f(y_n) - L_2| \\
\leq |L_1 - f(x_n)| + |f(x_n) - f(y_n)| + |f(y_n) - L_2| \\
\leq \varepsilon/3 + \varepsilon/3 + \varepsilon/3 = \varepsilon.
\]

Therefore, \( L_1 = L_2 \). Thus \( L_a \) exists. To show that \( L_b \) exists is left as an exercise.

Now that we know that the limits \( L_a \) and \( L_b \) exist, we are done. If \( \lim_{x \to a} f(x) \) exists, then \( \lim_{x \to a} \tilde{f}(x) \) exists (see Proposition 3.1.15). Similarly with \( L_b \). Hence \( \tilde{f} \) is continuous at \( a \) and \( b \). And since \( f \) is continuous at \( c \in (a, b) \), then \( \tilde{f} \) is continuous at \( c \in (a, b) \).

A common application of this proposition (together with Proposition 3.1.17) is the following. Suppose \( f: (-1, 0) \cup (0, 1) \to \mathbb{R} \) is uniformly continuous, then \( \lim_{x \to 0} f(x) \) exists and the function has what is called a removable singularity, that is, we can extend the function to a continuous function on \((-1, 1)\).
3.4.3 Lipschitz continuous functions

**Definition 3.4.7.** A function \( f : S \to \mathbb{R} \) is **Lipschitz continuous**, if there exists a \( K \in \mathbb{R} \), such that

\[
|f(x) - f(y)| \leq K |x - y| \quad \text{for all } x \text{ and } y \text{ in } S.
\]

A large class of functions is Lipschitz continuous. Be careful, just as for uniformly continuous functions, the domain of definition of the function is important. See the examples below and the exercises. First, we justify the use of the word *continuous*.

**Proposition 3.4.8.** A Lipschitz continuous function is uniformly continuous.

**Proof.** Let \( f : S \to \mathbb{R} \) be a function and let \( K \) be a constant such that \( |f(x) - f(y)| \leq K |x - y| \) for all \( x, y \) in \( S \). Let \( \varepsilon > 0 \) be given. Take \( \delta := \varepsilon/K \). For all \( x \) and \( y \) in \( S \) such that \( |x - y| < \delta \), we have

\[
|f(x) - f(y)| \leq K |x - y| < K \delta = K \frac{\varepsilon}{K} = \varepsilon.
\]

Therefore, \( f \) is uniformly continuous. \( \square \)

We interpret Lipschitz continuity geometrically. Let \( f \) be a Lipschitz continuous function with some constant \( K \). We rewrite the inequality to say that for \( x \neq y \), we have

\[
\left| \frac{f(x) - f(y)}{x - y} \right| \leq K.
\]

The quantity \( \frac{f(x) - f(y)}{x - y} \) is the slope of the line between the points \((x, f(x))\) and \((y, f(y))\), that is, a secant line. Therefore, \( f \) is Lipschitz continuous if and only if every line that intersects the graph of \( f \) in at least two distinct points has slope less than or equal to \( K \). See Figure 3.7.

![Figure 3.7: The slope of a secant line. A function is Lipschitz if \( |\text{slope}| = \left| \frac{f(x) - f(y)}{x - y} \right| \leq K \) for all \( x \) and \( y \).](image)

**Example 3.4.9:** The functions \( \sin(x) \) and \( \cos(x) \) are Lipschitz continuous. In Example 3.2.6 we have seen the following two inequalities.

\[
|\sin(x) - \sin(y)| \leq |x - y| \quad \text{and} \quad |\cos(x) - \cos(y)| \leq |x - y|.
\]

Hence sine and cosine are Lipschitz continuous with \( K = 1 \).

*Named after the German mathematician Rudolf Otto Sigismund Lipschitz (1832–1903).
Example 3.4.10: The function \( f : [1, \infty) \to \mathbb{R} \) defined by \( f(x) := \sqrt{x} \) is Lipschitz continuous. Proof:

\[
|\sqrt{x} - \sqrt{y}| = \left| \frac{x - y}{\sqrt{x} + \sqrt{y}} \right| = \frac{|x - y|}{\sqrt{x} + \sqrt{y}}.
\]

As \( x \geq 1 \) and \( y \geq 1 \), we see that \( \frac{1}{\sqrt{x} + \sqrt{y}} \leq \frac{1}{2} \). Therefore,

\[
|\sqrt{x} - \sqrt{y}| \leq \frac{1}{2} |x - y|.
\]

On the other hand, \( f : [0, \infty) \to \mathbb{R} \) defined by \( f(x) := \sqrt{x} \) is not Lipschitz continuous. Let us see why: Suppose

\[
|\sqrt{x} - \sqrt{y}| \leq K |x - y|
\]

for some \( K \). Set \( y = 0 \) to obtain \( \sqrt{x} \leq Kx \). If \( K > 0 \), then for \( x > 0 \) we then get \( 1/K \leq \sqrt{x} \). This cannot possibly be true for all \( x > 0 \). Thus no such \( K > 0 \) exists and \( f \) is not Lipschitz continuous.

The last example is a function that is uniformly continuous but not Lipschitz continuous. To see that \( \sqrt{x} \) is uniformly continuous on \([0, \infty)\), note that it is uniformly continuous on \([0, 1]\) by Theorem 3.4.4. It is also Lipschitz (and therefore uniformly continuous) on \([1, \infty)\). It is not hard (exercise) to show that this means that \( \sqrt{x} \) is uniformly continuous on \([0, \infty)\).

### Exercises

**Exercise 3.4.1:** Let \( f : S \to \mathbb{R} \) be uniformly continuous. Let \( A \subseteq S \). Then the restriction \( f|_A \) is uniformly continuous.

**Exercise 3.4.2:** Let \( f : (a, b) \to \mathbb{R} \) be a uniformly continuous function. Finish the proof of Proposition 3.4.6 by showing that the limit \( \lim_{x \to b} f(x) \) exists.

**Exercise 3.4.3:** Show that \( f : (c, \infty) \to \mathbb{R} \) for some \( c > 0 \) and defined by \( f(x) := 1/x \) is Lipschitz continuous.

**Exercise 3.4.4:** Show that \( f : (0, \infty) \to \mathbb{R} \) defined by \( f(x) := 1/x \) is not Lipschitz continuous.

**Exercise 3.4.5:** Let \( A, B \) be intervals. Let \( f : A \to \mathbb{R} \) and \( g : B \to \mathbb{R} \) be uniformly continuous functions such that \( f(x) = g(x) \) for \( x \in A \cap B \). Define the function \( h : A \cup B \to \mathbb{R} \) by \( h(x) := f(x) \) if \( x \in A \) and \( h(x) := g(x) \) if \( x \in B \setminus A \).

a) Prove that if \( A \cap B \neq \emptyset \), then \( h \) is uniformly continuous.

b) Find an example where \( A \cap B = \emptyset \) and \( h \) is not even continuous.

**Exercise 3.4.6** (Challenging): Let \( f : \mathbb{R} \to \mathbb{R} \) be a polynomial of degree \( d \geq 2 \). Show that \( f \) is not Lipschitz continuous.

**Exercise 3.4.7:** Let \( f : (0, 1) \to \mathbb{R} \) be a bounded continuous function. Show that the function \( g(x) := x(1 - x)f(x) \) is uniformly continuous.

**Exercise 3.4.8:** Show that \( f : (0, \infty) \to \mathbb{R} \) defined by \( f(x) := \sin(1/x) \) is not uniformly continuous.

**Exercise 3.4.9** (Challenging): Let \( f : \mathbb{Q} \to \mathbb{R} \) be a uniformly continuous function. Show that there exists a uniformly continuous function \( \tilde{f} : \mathbb{R} \to \mathbb{R} \) such that \( f(x) = \tilde{f}(x) \) for all \( x \in \mathbb{Q} \).
Exercise 3.4.10:

a) Find a continuous \( f : (0,1) \to \mathbb{R} \) and a sequence \( \{x_n\} \) in \((0,1)\) that is Cauchy, but such that \( \{f(x_n)\} \) is not Cauchy.

b) Prove that if \( f : \mathbb{R} \to \mathbb{R} \) is continuous, and \( \{x_n\} \) is Cauchy, then \( \{f(x_n)\} \) is Cauchy.

Exercise 3.4.11: Prove:

a) If \( f : S \to \mathbb{R} \) and \( g : S \to \mathbb{R} \) are uniformly continuous, then \( h : S \to \mathbb{R} \) given by \( h(x) := f(x) + g(x) \) is uniformly continuous.

b) If \( f : S \to \mathbb{R} \) is uniformly continuous and \( a \in \mathbb{R} \), then \( h : S \to \mathbb{R} \) given by \( h(x) := af(x) \) is uniformly continuous.

Exercise 3.4.12: Prove:

a) If \( f : S \to \mathbb{R} \) and \( g : S \to \mathbb{R} \) are Lipschitz, then \( h : S \to \mathbb{R} \) given by \( h(x) := f(x) + g(x) \) is Lipschitz.

b) If \( f : S \to \mathbb{R} \) is Lipschitz and \( a \in \mathbb{R} \), then \( h : S \to \mathbb{R} \) given by \( h(x) := af(x) \) is Lipschitz.

Exercise 3.4.13:

a) If \( f : [0,1] \to \mathbb{R} \) is given by \( f(x) := x^m \) for an integer \( m \geq 0 \), show \( f \) is Lipschitz and find the best (the smallest) Lipschitz constant \( K \) (depending on \( m \) of course). Hint: \( (x-y)(x^{m-1}+x^{m-2}y+x^{m-3}y^2+\cdots+xy^{m-2}+y^{m-1}) = x^m - y^m \).

b) Using the previous exercise, show that if \( f : [0,1] \to \mathbb{R} \) is a polynomial, that is, \( f(x) := a_mx^m+a_{m-1}x^{m-1}+\cdots+a_0 \), then \( f \) is Lipschitz.

Exercise 3.4.14: Suppose for \( f : [0,1] \to \mathbb{R} \), we have \( |f(x) - f(y)| \leq K|x-y| \) for all \( x,y \) in \([0,1]\), and \( f(0) = f(1) = 0 \). Prove that \( |f(x)| \leq K/2 \) for all \( x \in [0,1] \). Further show by example that \( K/2 \) is the best possible, that is, there exists such a continuous function for which \( |f(x)| = K/2 \) for some \( x \in [0,1] \).

Exercise 3.4.15: Suppose \( f : \mathbb{R} \to \mathbb{R} \) is continuous and periodic with period \( P > 0 \). That is, \( f(x+P) = f(x) \) for all \( x \in \mathbb{R} \). Show that \( f \) is uniformly continuous.

Exercise 3.4.16: Suppose \( f : S \to \mathbb{R} \) and \( g : [0,\infty) \to [0,\infty) \) are functions, \( g \) is continuous at \( 0 \), \( g(0) = 0 \), and whenever \( x \) and \( y \) are in \( S \), we have \( |f(x) - f(y)| \leq g(|x-y|) \). Prove that \( f \) is uniformly continuous.

Exercise 3.4.17: Suppose \( f : [a,b] \to \mathbb{R} \) is a function such that for every \( c \in [a,b] \) there is a \( K_c > 0 \) and an \( \varepsilon_c > 0 \) for which \( |f(x) - f(y)| \leq K_c|x-y| \) for all \( x \) and \( y \) in \( (c-\varepsilon_c,c+\varepsilon_c) \cap [a,b] \). In other words, \( f \) is “locally Lipschitz.”

a) Prove that there exists a single \( K > 0 \) such that \( |f(x) - f(y)| \leq K|x-y| \) for all \( x,y \) in \([a,b]\).

b) Find a counterexample to the above if the interval is open, that is, find an \( f : (a,b) \to \mathbb{R} \) that is locally Lipschitz, but not Lipschitz.
3.5 Limits at infinity

Note: less than 1 lecture (optional, can safely be omitted unless §3.6 or §5.5 is also covered)

3.5.1 Limits at infinity

As for sequences, a continuous variable can also approach infinity. Let us make this notion precise.

Definition 3.5.1. We say $\infty$ is a cluster point of $S \subset \mathbb{R}$ if for every $M \in \mathbb{R}$, there exists an $x \in S$ such that $x \geq M$. Similarly, $-\infty$ is a cluster point of $S \subset \mathbb{R}$ if for every $M \in \mathbb{R}$, there exists an $x \in S$ such that $x \leq M$.

Let $f : S \to \mathbb{R}$ be a function, where $\infty$ is a cluster point of $S$. If there exists an $L \in \mathbb{R}$ such that for every $\varepsilon > 0$, there is an $M \in \mathbb{R}$ such that

$$|f(x) - L| < \varepsilon$$

whenever $x \in S$ and $x \geq M$, then we say $f(x)$ converges to $L$ as $x$ goes to $\infty$. We call $L$ the limit and write

$$\lim_{x \to \infty} f(x) := L.$$ 

Alternatively we write $f(x) \to L$ as $x \to \infty$.

Similarly, if $-\infty$ is a cluster point of $S$ and there exists an $L \in \mathbb{R}$ such that for every $\varepsilon > 0$, there is an $M \in \mathbb{R}$ such that

$$|f(x) - L| < \varepsilon$$

whenever $x \in S$ and $x \leq M$, then we say $f(x)$ converges to $L$ as $x$ goes to $-\infty$. We call $L$ the limit and write

$$\lim_{x \to -\infty} f(x) := L.$$ 

Alternatively we write $f(x) \to L$ as $x \to -\infty$.

We cheated a little bit again and said the limit. We leave it as an exercise for the reader to prove the following proposition.

Proposition 3.5.2. The limit at $\infty$ or $-\infty$ as defined above is unique if it exists.

Example 3.5.3: Let $f(x) := \frac{1}{|x|+1}$. Then

$$\lim_{x \to \infty} f(x) = 0 \quad \text{and} \quad \lim_{x \to -\infty} f(x) = 0.$$ 

Proof: Let $\varepsilon > 0$ be given. Find $M > 0$ large enough so that $\frac{1}{M+1} < \varepsilon$. If $x \geq M$, then $\frac{1}{x+1} \leq \frac{1}{M+1} < \varepsilon$. Since $\frac{1}{|x|+1} > 0$ for all $x$ the first limit is proved. The proof for $-\infty$ is left to the reader.
Example 3.5.4: Let \( f(x) := \sin(\pi x) \). Then \( \lim_{x \to \infty} f(x) \) does not exist. To prove this fact note that if \( x = 2n + \frac{1}{2} \) for some \( n \in \mathbb{N} \), then \( f(x) = 1 \), while if \( x = 2n + \frac{3}{2} \), then \( f(x) = -1 \). So they cannot both be within a small \( \varepsilon \) of a single real number.

We must be careful not to confuse continuous limits with limits of sequences. We could say \( \lim_{n \to \infty} \sin(\pi n) = 0 \), but \( \lim_{x \to \infty} \sin(\pi x) \) does not exist.

Of course the notation is ambiguous: Are we thinking of the sequence \( \{\sin(\pi n)\}_{n=1}^{\infty} \) or the function \( \sin(\pi x) \) of a real variable? We are simply using the convention that \( n \in \mathbb{N} \), while \( x \in \mathbb{R} \). When the notation is not clear, it is good to explicitly mention where the variable lives, or what kind of limit are you using. If there is possibility of confusion, one can write, for example,

\[
\lim_{n \to \infty} \sin(\pi n). 
\]

There is a connection of continuous limits to limits of sequences, but we must take all sequences going to infinity, just as before in Lemma 3.1.7.

Lemma 3.5.5. Suppose \( f: S \to \mathbb{R} \) is a function, \( \infty \) is a cluster point of \( S \subset \mathbb{R} \), and \( L \in \mathbb{R} \). Then

\[
\lim_{x \to \infty} f(x) = L
\]

if and only if

\[
\lim_{n \to \infty} f(x_n) = L
\]

for all sequences \( \{x_n\} \) in \( S \) such that \( \lim_{n \to \infty} x_n = \infty \).

The lemma holds for the limit as \( x \to -\infty \). Its proof is almost identical and is left as an exercise.

Proof. First suppose \( f(x) \to L \) as \( x \to \infty \). Given an \( \varepsilon > 0 \), there exists an \( M \) such that for all \( x \geq M \), we have \( |f(x) - L| < \varepsilon \). Let \( \{x_n\} \) be a sequence in \( S \) such that \( \lim_{n \to \infty} x_n = \infty \). Then there exists an \( N \) such that for all \( n \geq N \), we have \( x_n \geq M \). And thus \( |f(x_n) - L| < \varepsilon \).

We prove the converse by contrapositive. Suppose \( f(x) \) does not go to \( L \) as \( x \to \infty \). This means that there exists an \( \varepsilon > 0 \), such that for every \( n \in \mathbb{N} \), there exists an \( x \in S \), \( x \geq n \), let us call it \( x_n \), such that \( |f(x_n) - L| \geq \varepsilon \). Consider the sequence \( \{x_n\} \). Clearly \( \{f(x_n)\} \) does not converge to \( L \). It remains to note that \( \lim_{n \to \infty} x_n = \infty \), because \( x_n \geq n \) for all \( n \).

Using the lemma, we again translate results about sequential limits into results about continuous limits as \( x \) goes to infinity. That is, we have almost immediate analogues of the corollaries in §3.1.3. We simply allow the cluster point \( c \) to be either \( \infty \) or \( -\infty \), in addition to a real number. We leave it to the student to verify these statements.

### 3.5.2 Infinite limit

Just as for sequences, it is often convenient to distinguish certain divergent sequences, and talk about limits being infinite almost as if the limits existed.
3.5. LIMITS AT INFINITY

**Definition 3.5.6.** Let \( f : S \to \mathbb{R} \) be a function and suppose \( S \) has \( \infty \) as a cluster point. We say \( f(x) \) diverges to infinity as \( x \) goes to \( \infty \), if for every \( N \in \mathbb{R} \) there exists an \( M \in \mathbb{R} \) such that

\[
f(x) > N
\]

whenever \( x \in S \) and \( x \geq M \). We write

\[
\lim_{x \to \infty} f(x) := \infty,
\]

or we say that \( f(x) \to \infty \) as \( x \to \infty \).

A similar definition can be made for limits as \( x \to -\infty \) or as \( x \to c \) for a finite \( c \). Also similar definitions can be made for limits being \(-\infty\). Stating these definitions is left as an exercise. Note that sometimes converges to infinity is used. We can again use sequential limits, and an analogue of Lemma 3.1.7 is left as an exercise.

**Example 3.5.7:** Let us show that \( \lim_{x \to \infty} \frac{1 + x^2}{1 + x} = \infty \).

Proof: For \( x \geq 1 \), we have

\[
\frac{1 + x^2}{1 + x} \geq \frac{x^2}{x + x} = \frac{x}{2}.
\]

Given \( N \in \mathbb{R} \), take \( M = \max\{2N + 1, 1\} \). If \( x \geq M \), then \( x \geq 1 \) and \( x/2 > N \). So

\[
\frac{1 + x^2}{1 + x} \geq \frac{x}{2} > N.
\]

### 3.5.3 Compositions

Finally, just as for limits at finite numbers we can compose functions easily.

**Proposition 3.5.8.** Suppose \( f : A \to B \), \( g : B \to \mathbb{R} \), \( A, B \subset \mathbb{R} \), \( a \in \mathbb{R} \cup \{-\infty, \infty\} \) is a cluster point of \( A \), and \( b \in \mathbb{R} \cup \{-\infty, \infty\} \) is a cluster point of \( B \). Suppose

\[
\lim_{x \to a} f(x) = b \quad \text{and} \quad \lim_{y \to b} g(y) = c
\]

for some \( c \in \mathbb{R} \cup \{-\infty, \infty\} \). If \( b \in B \), then suppose \( g(b) = c \). Then

\[
\lim_{x \to a} g(f(x)) = c.
\]

The proof is straightforward, and left as an exercise. We already know the proposition when \( a, b, c \in \mathbb{R} \), see Exercises 3.1.9 and 3.1.14. Again the requirement that \( g \) is continuous at \( b \), if \( b \in B \), is necessary.

**Example 3.5.9:** Let \( h(x) := e^{-x^2 + x} \). Then

\[
\lim_{x \to \infty} h(x) = 0.
\]

Proof: The claim follows once we know

\[
\lim_{x \to \infty} -x^2 + x = -\infty
\]

and

\[
\lim_{y \to -\infty} e^y = 0,
\]

which is usually proved when the exponential function is defined.
3.5.4 Exercises

Exercise 3.5.1: Prove Proposition 3.5.2.

Exercise 3.5.2: Let \( f: [1, \infty) \to \mathbb{R} \) be a function. Define \( g: (0, 1] \to \mathbb{R} \) via \( g(x) := f(1/x) \). Using the definitions of limits directly, show that \( \lim_{x \to 0^+} g(x) \) exists if and only if \( \lim_{x \to \infty} f(x) \) exists, in which case they are equal.

Exercise 3.5.3: Prove Proposition 3.5.8.

Exercise 3.5.4: Let us justify terminology. Let \( f: \mathbb{R} \to \mathbb{R} \) be a function such that \( \lim_{x \to \infty} f(x) = \infty \) (diverges to infinity). Show that \( f(x) \) diverges (i.e. does not converge) as \( x \to \infty \).

Exercise 3.5.5: Come up with the definitions for limits of \( f(x) \) going to \( -\infty \) as \( x \to \infty \), \( x \to -\infty \), and as \( x \to c \) for a finite \( c \in \mathbb{R} \). Then state the definitions for limits of \( f(x) \) going to \( \infty \) as \( x \to -\infty \), and as \( x \to c \) for a finite \( c \in \mathbb{R} \).

Exercise 3.5.6: Suppose \( P(x) := x^n + a_{n-1}x^{n-1} + \cdots + a_1x + a_0 \) is a monic polynomial of degree \( n \geq 1 \) (monic means that the coefficient of \( x^n \) is 1).

a) Show that if \( n \) is even, then \( \lim_{x \to \infty} P(x) = \lim_{x \to -\infty} P(x) = \infty \).

b) Show that if \( n \) is odd, then \( \lim_{x \to \infty} P(x) = \infty \) and \( \lim_{x \to -\infty} P(x) = -\infty \) (see previous exercise).

Exercise 3.5.7: Let \( \{x_n\} \) be a sequence. Consider \( S := \mathbb{N} \subset \mathbb{R} \), and \( f: S \to \mathbb{R} \) defined by \( f(n) := x_n \). Show that the two notions of limit,

\[
\lim_{n \to \infty} x_n \quad \text{and} \quad \lim_{x \to \infty} f(x)
\]

are equivalent. That is, show that if one exists so does the other one, and in this case they are equal.

Exercise 3.5.8: Extend Lemma 3.5.5 as follows. Suppose \( S \subset \mathbb{R} \) has a cluster point \( c \in \mathbb{R} \), \( c = \infty \), or \( c = -\infty \). Let \( f: S \to \mathbb{R} \) be a function and suppose \( L = \infty \) or \( L = -\infty \). Show that

\[
\lim_{x \to c} f(x) = L \quad \text{if and only if} \quad \lim_{n \to \infty} f(x_n) = L \quad \text{for all sequences} \quad \{x_n\} \quad \text{such that} \quad \lim x_n = c.
\]

Exercise 3.5.9: Suppose \( f: \mathbb{R} \to \mathbb{R} \) is a 2-periodic function, that is \( f(x + 2) = f(x) \) for all \( x \). Define \( g: \mathbb{R} \to \mathbb{R} \) by

\[
g(x) := f\left(\frac{\sqrt{x^2 + 1} - 1}{x}\right)
\]

a) Find the function \( \varphi: (-1, 1) \to \mathbb{R} \) such that \( g(\varphi(t)) = f(t) \), that is \( \varphi^{-1}(x) = \frac{\sqrt{x^2 + 1} - 1}{x} \).

b) Show that \( f \) is continuous if and only if \( g \) is continuous and

\[
\lim_{x \to \infty} g(x) = \lim_{x \to -\infty} g(x) = f(1) = f(-1).
\]
3.6 Monotone functions and continuity

Note: 1 lecture (optional, can safely be omitted unless §4.4 is also covered, requires §3.5)

Definition 3.6.1. Let $S \subset \mathbb{R}$. We say $f: S \to \mathbb{R}$ is increasing (resp. strictly increasing) if $x, y \in S$ with $x < y$ implies $f(x) \leq f(y)$ (resp. $f(x) < f(y)$). We define decreasing and strictly decreasing in the same way by switching the inequalities for $f$.

If a function is either increasing or decreasing, we say it is monotone. If it is strictly increasing or strictly decreasing, we say it is strictly monotone.

Sometimes nondecreasing (resp. nonincreasing) is used for increasing (resp. decreasing) function to emphasize it is not strictly increasing (resp. strictly decreasing).

If $f$ is increasing, then $-f$ is decreasing and vice versa. Therefore, many results about monotone functions can just be proved for, say, increasing functions, and the results follow easily for decreasing functions.

3.6.1 Continuity of monotone functions

One-sided limits for monotone functions are computed by computing infima and suprema.

Proposition 3.6.2. Let $S \subset \mathbb{R}$, $c \in \mathbb{R}$, $f: S \to \mathbb{R}$ be increasing, and $g: S \to \mathbb{R}$ be decreasing. If $c$ is a cluster point of $S \cap (-\infty, c)$, then

$$
\lim_{x \to c^-} f(x) = \sup \{ f(x) : x < c, x \in S \} \quad \text{and} \quad \lim_{x \to c^-} g(x) = \inf \{ g(x) : x < c, x \in S \}.
$$

If $c$ is a cluster point of $S \cap (c, \infty)$, then

$$
\lim_{x \to c^+} f(x) = \inf \{ f(x) : x > c, x \in S \} \quad \text{and} \quad \lim_{x \to c^+} g(x) = \sup \{ g(x) : x > c, x \in S \}.
$$

If $\infty$ is a cluster point of $S$, then

$$
\lim_{x \to \infty} f(x) = \sup \{ f(x) : x \in S \} \quad \text{and} \quad \lim_{x \to \infty} g(x) = \inf \{ g(x) : x \in S \}.
$$

If $-\infty$ is a cluster point of $S$, then

$$
\lim_{x \to -\infty} f(x) = \inf \{ f(x) : x \in S \} \quad \text{and} \quad \lim_{x \to -\infty} g(x) = \sup \{ g(x) : x \in S \}.
$$

Namely, all the one-sided limits exist whenever they make sense. For monotone functions therefore, when we say the left-hand limit $x \to c^-$ exists, we mean that $c$ is a cluster point of $S \cap (-\infty, c)$, and same for the right-hand limit.

Proof. Let us assume $f$ is increasing, and we will show the first equality. The rest of the proof is very similar and is left as an exercise.

Let $a := \sup \{ f(x) : x < c, x \in S \}$. If $a = \infty$, then given an $M \in \mathbb{R}$, there exists an $x_M \in S$, $x_M < c$, such that $f(x_M) > M$. As $f$ is increasing, $f(x) \geq f(x_M) > M$ for all $x \in S$ with $x > x_M$. If we take $\delta := c - x_M > 0$, then we obtain the definition of the limit going to infinity.

Next suppose $a < \infty$. Let $\varepsilon > 0$ be given. Because $a$ is the supremum and $S \cap (-\infty, c)$ is nonempty, $a \in \mathbb{R}$ and there exists an $x_\varepsilon \in S$, $x_\varepsilon < c$, such that $f(x_\varepsilon) > a - \varepsilon$. As $f$ is increasing, if $x \in S$ and $x_\varepsilon < x < c$, we have $a - \varepsilon < f(x_\varepsilon) \leq f(x) \leq a$. Let $\delta := c - x_\varepsilon$. Then for $x \in S \cap (-\infty, c)$ with $|x - c| < \delta$, we have $|f(x) - a| < \varepsilon$. \qed
Suppose \( f : S \to \mathbb{R} \) is increasing, \( c \in S \), and that both one-sided limits exist. Since \( f(x) \leq f(c) \leq f(y) \) whenever \( x < c < y \), taking the limits we obtain

\[
\lim_{x \to c^-} f(x) \leq f(c) \leq \lim_{x \to c^+} f(x).
\]

Then \( f \) is continuous at \( c \) if and only if both limits are equal to each other (and hence equal to \( f(c) \)). See also Proposition 3.1.17. See Figure 3.8 to get an idea of what a discontinuity looks like.

**Corollary 3.6.3.** If \( I \subset \mathbb{R} \) is an interval and \( f : I \to \mathbb{R} \) is monotone and not constant, then \( f(I) \) is an interval if and only if \( f \) is continuous.

Assuming \( f \) is not constant is to avoid the technicality that \( f(I) \) is a single point: \( f(I) \) is a single point if and only if \( f \) is constant. A constant function is continuous.

**Proof.** Without loss of generality, suppose \( f \) is increasing.

First suppose \( f \) is continuous. Take two points \( f(x_1) < f(x_2) \) in \( f(I) \). As \( f \) is increasing, then \( x_1 < x_2 \). By the intermediate value theorem, given \( y \) with \( f(x_1) < y < f(x_2) \), we find a \( c \in (x_1, x_2) \subset I \) such that \( f(c) = y \), so \( y \in f(I) \). Hence, \( f(I) \) is an interval.

Let us prove the reverse direction by contrapositive. Suppose \( f \) is not continuous at \( c \in I \), and that \( c \) is not an endpoint of \( I \). Let

\[
a := \lim_{x \to c^-} f(x) = \sup\{ f(x) : x \in I, x < c \}, \quad b := \lim_{x \to c^+} f(x) = \inf\{ f(x) : x \in I, x > c \}.
\]

As \( c \) is a discontinuity, \( a < b \). If \( x < c \), then \( f(x) \leq a \), and if \( x > c \), then \( f(x) \geq b \). Therefore no point in \( (a, b) \setminus \{ f(c) \} \) is in \( f(I) \). However there exists \( x_1 \in I, x_1 < c \), so \( f(x_1) \leq a \), and there exists \( x_2 \in I, x_2 > c \), so \( f(x_2) \geq b \). Both \( f(x_1) \) and \( f(x_2) \) are in \( f(I) \), but there are points in between them that are not in \( f(I) \). So \( f(I) \) is not an interval. See Figure 3.8.

When \( c \in I \) is an endpoint, the proof is similar and is left as an exercise. \( \square \)

![Figure 3.8](image-url) 

**Figure 3.8:** Increasing function \( f : I \to \mathbb{R} \) discontinuity at \( c \).

A striking property of monotone functions is that they cannot have too many discontinuities.
3.6. MONOTONE FUNCTIONS AND CONTINUITY

Corollary 3.6.4. Let $I \subset \mathbb{R}$ be an interval and $f : I \to \mathbb{R}$ be monotone. Then $f$ has at most countably many discontinuities.

Proof. Let $E \subset I$ be the set of all discontinuities that are not endpoints of $I$. As there are only two endpoints, it is enough to show that $E$ is countable. Without loss of generality, suppose $f$ is increasing. We will define an injection $h : E \to \mathbb{Q}$. For each $c \in E$ the one-sided limits of $f$ both exist as $c$ is not an endpoint. Let

$$a := \lim_{x \to c^-} f(x) = \sup \{ f(x) : x \in I, x < c \}, \quad b := \lim_{x \to c^+} f(x) = \inf \{ f(x) : x \in I, x > c \}.$$

As $c$ is a discontinuity, we have $a < b$. There exists a rational number $q \in (a, b)$, so let $h(c) = q$. If $d \in E$ is another discontinuity, then if $d > c$, then there exist an $x \in I$ with $c < x < d$, and so $\lim_{x \to d^-} f(x) \geq b$. Hence the rational number we choose for $h(d)$ is different from $q$, since $q = h(c) < b$ and $h(d) > b$. Similarly if $d < c$. So after making such a choice for every $c \in E$, we have a one-to-one (injective) function into $\mathbb{Q}$. Therefore, $E$ is countable. □

Example 3.6.5: By $\lceil x \rceil$ denote the largest integer less than or equal to $x$. Define $f : [0, 1] \to \mathbb{R}$ by

$$f(x) := x + \sum_{n=0}^{\lceil 1/(1-x) \rceil} 2^{-n},$$

for $x < 1$ and $f(1) := 3$. It is left as an exercise to show that $f$ is strictly increasing, bounded, and has a discontinuity at all points $1 - 1/k$ for $k \in \mathbb{N}$. In particular, there are countably many discontinuities, but the function is bounded and defined on a closed bounded interval. See Figure 3.9.

![Figure 3.9: Increasing function with countably many discontinuities.](image)

Similarly, one can find an example of a function discontinuous on a dense set such as the rational numbers. See the exercises.

3.6.2 Continuity of inverse functions

A strictly monotone function $f$ is one-to-one (injective). To see this fact, notice that if $x \neq y$, then we can assume $x < y$. Either $f(x) < f(y)$ if $f$ is strictly increasing or $f(x) > f(y)$ if $f$ is strictly decreasing, so $f(x) \neq f(y)$. Hence, $f$ must have an inverse $f^{-1}$ defined on its range.

Proposition 3.6.6. If $I \subset \mathbb{R}$ is an interval and $f : I \to \mathbb{R}$ is strictly monotone, then the inverse $f^{-1} : f(I) \to I$ is continuous.
Proof. Let us suppose \( f \) is strictly increasing. The proof is almost identical for a strictly decreasing function. Since \( f \) is strictly increasing, so is \( f^{-1} \). That is, if \( f(x) < f(y) \), then we must have \( x < y \) and therefore \( f^{-1}(f(x)) < f^{-1}(f(y)) \).

Take \( c \in f(I) \). If \( c \) is not a cluster point of \( f(I) \), then \( f^{-1} \) is continuous at \( c \) automatically. So let \( c \) be a cluster point of \( f(I) \). Suppose both of the following one-sided limits exist:

\[
x_0 := \lim_{y \to c^-} f^{-1}(y) = \sup \{ f^{-1}(y) : y < c, y \in f(I) \} = \sup \{ x \in I : f(x) < c \},
\]

\[
x_1 := \lim_{y \to c^+} f^{-1}(y) = \inf \{ f^{-1}(y) : y > c, y \in f(I) \} = \inf \{ x \in I : f(x) > c \}.
\]

We have \( x_0 \leq x_1 \) as \( f^{-1} \) is increasing. For all \( x \in I \) where \( x > x_0 \), we have \( f(x) \geq c \). As \( f \) is strictly increasing, we must have \( f(x) > c \) for all \( x \in I \) where \( x > x_0 \). Therefore,

\[
\{ x \in I : x > x_0 \} \subseteq \{ x \in I : f(x) > c \}.
\]

The infimum of the left-hand set is \( x_0 \), and the infimum of the right-hand set is \( x_1 \), so we obtain \( x_0 \geq x_1 \). So \( x_1 = x_0 \), and \( f^{-1} \) is continuous at \( c \).

If one of the one-sided limits does not exist, the argument is similar and is left as an exercise. \( \square \)

**Example 3.6.7:** The proposition does not require \( f \) itself to be continuous. Let \( f : \mathbb{R} \to \mathbb{R} \) be defined by

\[
f(x) := \begin{cases} 
    x & \text{if } x < 0, \\
    x + 1 & \text{if } x \geq 0.
\end{cases}
\]

The function \( f \) is not continuous at 0. The image of \( I = \mathbb{R} \) is the set \((-\infty, 0) \cup [1, \infty)\), not an interval. Then \( f^{-1} : (\mathbb{R}) \to \mathbb{R} \) can be written as

\[
f^{-1}(y) = \begin{cases} 
    y & \text{if } y < 0, \\
    y - 1 & \text{if } y \geq 1.
\end{cases}
\]

It is not difficult to see that \( f^{-1} \) is a continuous function. See Figure 3.10 for the graphs.

---

**Figure 3.10:** Graph of \( f \) on the left and \( f^{-1} \) on the right.
Notice what happens with the proposition if \( f(I) \) is an interval. In that case, we could simply apply Corollary 3.6.3 to both \( f \) and \( f^{-1} \). That is, if \( f: I \to J \) is an onto strictly monotone function and \( I \) and \( J \) are intervals, then both \( f \) and \( f^{-1} \) are continuous. Furthermore, \( f(I) \) is an interval precisely when \( f \) is continuous.

### 3.6.3 Exercises

**Exercise 3.6.1:** Suppose \( f: [0, 1] \to \mathbb{R} \) is monotone. Prove \( f \) is bounded.

**Exercise 3.6.2:** Finish the proof of Proposition 3.6.2. Hint: You can halve your work by noticing that if \( g \) is decreasing, then \( -g \) is increasing.

**Exercise 3.6.3:** Finish the proof of Corollary 3.6.3.

**Exercise 3.6.4:** Prove the claims in Example 3.6.5.

**Exercise 3.6.5:** Finish the proof of Proposition 3.6.6.

**Exercise 3.6.6:** Suppose \( S \subset \mathbb{R} \), and \( f: S \to \mathbb{R} \) is an increasing function. Prove:

\( a) \) If \( c \) is a cluster point of \( S \cap (c, \infty) \), then \( \lim_{x \to c^+} f(x) < \infty \).

\( b) \) If \( c \) is a cluster point of \( S \cap (-\infty, c) \) and \( \lim_{x \to c^-} f(x) = \infty \), then \( S \subset (-\infty, c) \).

**Exercise 3.6.7:** Let \( I \subset \mathbb{R} \) be an interval and \( f: I \to \mathbb{R} \) a function. Suppose that for each \( c \in I \), there exist \( a, b \in \mathbb{R} \) with \( a > 0 \) such that \( f(x) \geq ax + b \) for all \( x \in I \) and \( f(c) = ac + b \). Show that \( f \) is strictly increasing.

**Exercise 3.6.8:** Suppose \( I \) and \( J \) are intervals and \( f: I \to J \) is a continuous, bijective (one-to-one and onto) function. Show that \( f \) is strictly monotone.

**Exercise 3.6.9:** Consider a monotone function \( f: I \to \mathbb{R} \) on an interval \( I \). Prove that there exists a function \( g: I \to \mathbb{R} \) such that \( \lim_{x \to c^-} g(x) = g(c) \) for all \( c \) in \( I \) except the smaller (left) endpoint of \( I \), and such that \( g(x) = f(x) \) for all but countably many \( x \in I \).

**Exercise 3.6.10:**

\( a) \) Let \( S \subset \mathbb{R} \) be a subset. If \( f: S \to \mathbb{R} \) is increasing and bounded, then show that there exists an increasing \( F: \mathbb{R} \to \mathbb{R} \) such that \( f(x) = F(x) \) for all \( x \in S \).

\( b) \) Find an example of a strictly increasing bounded \( f: S \to \mathbb{R} \) such that an increasing \( F \) as above is never strictly increasing.

**Exercise 3.6.11** (Challenging): Find an example of an increasing function \( f: [0, 1] \to \mathbb{R} \) that has a discontinuity at each rational number. Then show that the image \( f([0, 1]) \) contains no interval. Hint: Enumerate the rational numbers and define the function with a series.

**Exercise 3.6.12:** Suppose \( I \) is an interval and \( f: I \to \mathbb{R} \) is monotone. Show that \( \mathbb{R} \setminus f(I) \) is a countable union of disjoint intervals.

**Exercise 3.6.13:** Suppose \( f: [0, 1] \to (0, 1) \) is increasing. Show that for every \( \varepsilon > 0 \), there exists a strictly increasing \( g: [0, 1] \to (0, 1) \) such that \( g(0) = f(0), f(x) \leq g(x) \) for all \( x \), and \( g(1) - f(1) < \varepsilon \).

**Exercise 3.6.14:** Prove that the Dirichlet function \( f: [0, 1] \to \mathbb{R} \) defined by \( f(x) := 1 \) if \( x \) is rational and \( f(x) := 0 \) otherwise cannot be written as a difference of two increasing functions. That is, there do not exist increasing \( g \) and \( h \) such that \( f(x) = g(x) - h(x) \).

**Exercise 3.6.15:** Suppose \( f: (a, b) \to (c, d) \) is a strictly increasing onto function. Prove that there exists a \( g: (a, b) \to (c, d) \), which is also strictly increasing and onto, and \( g(x) < f(x) \) for all \( x \in (a, b) \).
Chapter 4

The Derivative

4.1 The derivative

Note: 1 lecture

The idea of a derivative is the following. If the graph of a function looks locally like a straight line, then we can then talk about the slope of this line. The slope tells us the rate at which the value of the function is changing at that particular point. Of course, we are leaving out any function that has corners or discontinuities. Let us be precise.

4.1.1 Definition and basic properties

Definition 4.1.1. Let $I$ be an interval, let $f: I \to \mathbb{R}$ be a function, and let $c \in I$. If the limit

$$L := \lim_{x \to c} \frac{f(x) - f(c)}{x - c}$$

exists, then we say $f$ is differentiable at $c$, that $L$ is the derivative of $f$ at $c$, and write $f'(c) := L$.

If $f$ is differentiable at all $c \in I$, then we simply say that $f$ is differentiable, and then we obtain a function $f': I \to \mathbb{R}$. The derivative is sometimes written as $\frac{df}{dx}$ or $\frac{d}{dx}(f(x))$.

The expression $\frac{f(x) - f(c)}{x - c}$ is called the difference quotient.

The graphical interpretation of the derivative is depicted in Figure 4.1. The left-hand plot gives the line through $(c, f(c))$ and $(x, f(x))$ with slope $\frac{f(x) - f(c)}{x - c}$, that is, the so-called secant line. When we take the limit as $x$ goes to $c$, we get the right-hand plot, where we see that the derivative of the function at the point $c$ is the slope of the line tangent to the graph of $f$ at the point $(c, f(c))$.

We allow $I$ to be a closed interval and we allow $c$ to be an endpoint of $I$. Some calculus books do not allow $c$ to be an endpoint of an interval, but all the theory still works by allowing it, and it makes our work easier.

Example 4.1.2: Let $f(x) := x^2$ defined on the whole real line. Let $c \in \mathbb{R}$ be arbitrary. We find that if $x \neq c$, 

$$\frac{x^2 - c^2}{x - c} = \frac{(x + c)(x - c)}{x - c} = (x + c).$$
CHAPTER 4. THE DERIVATIVE

Therefore,

\[ f'(c) = \lim_{x \to c} \frac{x^2 - c^2}{x - c} = \lim_{x \to c} (x + c) = 2c. \]

**Example 4.1.3:** Let \( f(x) := ax + b \) for numbers \( a, b \in \mathbb{R} \). Let \( c \in \mathbb{R} \) be arbitrary. For \( x \neq c \),

\[ \frac{f(x) - f(c)}{x - c} = \frac{a(x - c)}{x - c} = a. \]

Therefore,

\[ f'(c) = \lim_{x \to c} \frac{f(x) - f(c)}{x - c} = \lim_{x \to c} a = a. \]

In fact, every differentiable function “infinitesimally” behaves like the affine function \( ax + b \). You can guess many results and formulas for derivatives, if you work them out for affine functions first.

**Example 4.1.4:** The function \( f(x) := \sqrt{x} \) is differentiable for \( x > 0 \). To see this fact, fix \( c > 0 \), and take \( x \neq c, x > 0 \). Compute

\[ \frac{\sqrt{x} - \sqrt{c}}{x - c} = \frac{\sqrt{x} - \sqrt{c}}{(\sqrt{x} - \sqrt{c})(\sqrt{x} + \sqrt{c})} = \frac{1}{\sqrt{x} + \sqrt{c}}. \]

Therefore,

\[ f'(c) = \lim_{x \to c} \frac{\sqrt{x} - \sqrt{c}}{x - c} = \lim_{x \to c} \frac{1}{\sqrt{x} + \sqrt{c}} = \frac{1}{2\sqrt{c}}. \]

**Example 4.1.5:** The function \( f(x) := |x| \) is not differentiable at the origin. When \( x > 0 \),

\[ \frac{|x| - |0|}{x - 0} = \frac{x - 0}{x - 0} = 1. \]

When \( x < 0 \),

\[ \frac{|x| - |0|}{x - 0} = \frac{-x - 0}{x - 0} = -1. \]
A famous example of Weierstrass shows that there exists a continuous function that is not differentiable at any point. The construction of this function is beyond the scope of this chapter. On the other hand, a differentiable function is always continuous.

**Proposition 4.1.6.** Let $f : I \to \mathbb{R}$ be differentiable at $c \in I$, then it is continuous at $c$.

*Proof.* We know the limits
\[
\lim_{x \to c} \frac{f(x) - f(c)}{x - c} = f'(c) \quad \text{and} \quad \lim_{x \to c} (x - c) = 0
\]
exist. Furthermore,
\[
f(x) - f(c) = \left( \frac{f(x) - f(c)}{x - c} \right) (x - c).
\]
Therefore, the limit of $f(x) - f(c)$ exists and
\[
\lim_{x \to c} (f(x) - f(c)) = \left( \lim_{x \to c} \frac{f(x) - f(c)}{x - c} \right) \left( \lim_{x \to c} (x - c) \right) = f'(c) \cdot 0 = 0.
\]
Hence $\lim_{x \to c} f(x) = f(c)$, and $f$ is continuous at $c$. \hfill \Box

An important property of the derivative is linearity. The derivative is the approximation of a function by a straight line. The slope of a line through two points changes linearly when the $y$-coordinates are changed linearly. By taking the limit, it makes sense that the derivative is linear.

**Proposition 4.1.7.** Let $I$ be an interval, let $f : I \to \mathbb{R}$ and $g : I \to \mathbb{R}$ be differentiable at $c \in I$, and let $\alpha \in \mathbb{R}$.

(i) Define $h : I \to \mathbb{R}$ by $h(x) := \alpha f(x)$. Then $h$ is differentiable at $c$ and $h'(c) = \alpha f'(c)$.

(ii) Define $h : I \to \mathbb{R}$ by $h(x) := f(x) + g(x)$. Then $h$ is differentiable at $c$ and $h'(c) = f'(c) + g'(c)$.

*Proof.* First, let $h(x) := \alpha f(x)$. For $x \in I$, $x \neq c$,
\[
\frac{h(x) - h(c)}{x - c} = \frac{\alpha f(x) - \alpha f(c)}{x - c} = \alpha \frac{f(x) - f(c)}{x - c}.
\]
The limit as $x$ goes to $c$ exists on the right-hand side by Corollary 3.1.12. We get
\[
\lim_{x \to c} \frac{h(x) - h(c)}{x - c} = \alpha \lim_{x \to c} \frac{f(x) - f(c)}{x - c}.
\]
Therefore, $h$ is differentiable at $c$, and the derivative is computed as given.

Next, define $h(x) := f(x) + g(x)$. For $x \in I$, $x \neq c$, we have
\[
\frac{h(x) - h(c)}{x - c} = \frac{(f(x) + g(x)) - (f(c) + g(c))}{x - c} = \frac{f(x) - f(c)}{x - c} + \frac{g(x) - g(c)}{x - c}.
\]
The limit as $x$ goes to $c$ exists on the right-hand side by Corollary 3.1.12. We get
\[
\lim_{x \to c} \frac{h(x) - h(c)}{x - c} = \lim_{x \to c} \frac{f(x) - f(c)}{x - c} + \lim_{x \to c} \frac{g(x) - g(c)}{x - c}.
\]
Therefore, $h$ is differentiable at $c$, and the derivative is computed as given. \hfill \Box
It is not true that the derivative of a multiple of two functions is the multiple of the derivatives. Instead we get the so-called product rule or the Leibniz rule*.

**Proposition 4.1.8** (Product rule). Let \( I \) be an interval, let \( f: I \to \mathbb{R} \) and \( g: I \to \mathbb{R} \) be functions differentiable at \( c \). If \( h: I \to \mathbb{R} \) is defined by

\[
h(x) := f(x)g(x),
\]

then \( h \) is differentiable at \( c \) and

\[
h'(c) = f(c)g'(c) + f'(c)g(c).
\]

The proof of the product rule is left as an exercise. The key to the proof is the identity

\[
f(x)g(x) - f(c)g(c) = f(x)(g(x) - g(c)) + (f(x) - f(c))g(c),
\]

which is illustrated in Figure 4.2.

![Figure 4.2: The idea of product rule. The area of the entire rectangle \( f(x)g(x) \) differs from the area of the white rectangle \( f(c)g(c) \) by the area of the lightly shaded rectangle \( f(x)(g(x) - g(c)) \) plus the darker rectangle \( (f(x) - f(c))g(c) \). In other words, \( \Delta(f \cdot g) = f \cdot \Delta g + \Delta f \cdot g \).](image)

**Proposition 4.1.9** (Quotient Rule). Let \( I \) be an interval, let \( f: I \to \mathbb{R} \) and \( g: I \to \mathbb{R} \) be differentiable at \( c \) and \( g(x) \neq 0 \) for all \( x \in I \). If \( h: I \to \mathbb{R} \) is defined by

\[
h(x) := \frac{f(x)}{g(x)},
\]

then \( h \) is differentiable at \( c \) and

\[
h'(c) = \frac{f'(c)g(c) - f(c)g'(c)}{(g(c))^2}.
\]

Again, the proof is left as an exercise.

*Named for the German mathematician **Gottfried Wilhelm Leibniz** (1646–1716).
4.1.2 Chain rule

More complicated functions are often obtained by composition, which is differentiated via the chain rule. The rule also tells us how a derivative changes if we change variables.

**Proposition 4.1.10** (Chain Rule). Let $I_1, I_2$ be intervals, let $g : I_1 \to I_2$ be differentiable at $c \in I_1$, and $f : I_2 \to \mathbb{R}$ be differentiable at $g(c)$. If $h : I_1 \to \mathbb{R}$ is defined by

$$h(x) := (f \circ g)(x) = f(g(x)),$$

then $h$ is differentiable at $c$ and

$$h'(c) = f'(g(c))g'(c).$$

**Proof.** Let $d := g(c)$. Define $u : I_2 \to \mathbb{R}$ and $v : I_1 \to \mathbb{R}$ by

$$u(y) := \begin{cases} \frac{f(y) - f(d)}{y - d} & \text{if } y \neq d, \\ f'(d) & \text{if } y = d, \end{cases} \quad v(x) := \begin{cases} \frac{g(x) - g(c)}{x - c} & \text{if } x \neq c, \\ g'(c) & \text{if } x = c. \end{cases}$$

Because $f$ is differentiable at $d = g(c)$, we find that $u$ is continuous at $d$. Similarly, $v$ is continuous at $c$. For any $x$ and $y$,

$$f(y) - f(d) = u(y)(y - d) \quad \text{and} \quad g(x) - g(c) = v(x)(x - c).$$

Plug in to obtain

$$h(x) - h(c) = f(g(x)) - f(g(c)) = u(g(x))(g(x) - g(c)) = u(g(x))(v(x)(x - c)).$$

Therefore, if $x \neq c$,

$$\frac{h(x) - h(c)}{x - c} = u(g(x))v(x). \quad (4.1)$$

By continuity of $u$ and $v$ at $d$ and $c$ respectively, we find $\lim_{y \to d} u(y) = f'(d) = f'(g(c))$ and $\lim_{x \to c} v(x) = g'(c)$. The function $g$ is continuous at $c$, and so $\lim_{x \to c} g(x) = g(c)$. Hence the limit of the right-hand side of (4.1) as $x$ goes to $c$ exists and is equal to $f'(g(c))g'(c)$. Thus $h$ is differentiable at $c$ and the limit is $f'(g(c))g'(c)$. □

4.1.3 Exercises

**Exercise 4.1.1:** Prove the product rule. Hint: Prove and use $f(x)g(x) - f(c)g(c) = f(x)(g(x) - g(c)) + (f(x) - f(c))g(c)$.

**Exercise 4.1.2:** Prove the quotient rule. Hint: You can do this directly, but it may be easier to find the derivative of $1/x$ and then use the chain rule and the product rule.

**Exercise 4.1.3:** For $n \in \mathbb{Z}$, prove that $x^n$ is differentiable and find the derivative, unless, of course, $n < 0$ and $x = 0$. Hint: Use the product rule.

**Exercise 4.1.4:** Prove that a polynomial is differentiable and find the derivative. Hint: Use the previous exercise.
Exercise 4.1.5: Define $f : \mathbb{R} \to \mathbb{R}$ by
\[ f(x) := \begin{cases} x^2 & \text{if } x \in \mathbb{Q}, \\ 0 & \text{otherwise}. \end{cases} \]
Prove that $f$ is differentiable at 0, but discontinuous at all points except 0.

Exercise 4.1.6: Assume the inequality $|x - \sin(x)| \leq x^2$. Prove that $\sin$ is differentiable at 0, and find the derivative at 0.

Exercise 4.1.7: Using the previous exercise, prove that $\sin$ is differentiable at all $x$ and that the derivative is $\cos(x)$. Hint: Use the sum-to-product trigonometric identity as we did before.

Exercise 4.1.8: Let $f : I \to \mathbb{R}$ be differentiable. For $n \in \mathbb{Z}$, let $f^n$ be the function defined by $f^n(x) := (f(x))^n$. If $n < 0$, assume $f(x) \neq 0$ for all $x \in I$. Prove that $(f^n)'(x) = n(f(x))^{n-1}f'(x)$.

Exercise 4.1.9: Suppose $f : \mathbb{R} \to \mathbb{R}$ is a differentiable Lipschitz continuous function. Prove that $f'$ is a bounded function.

Exercise 4.1.10: Let $I_1, I_2$ be intervals. Let $f : I_1 \to I_2$ be a bijective function and $g : I_2 \to I_1$ be the inverse. Suppose that both $f$ is differentiable at $c \in I_1$ and $f'(c) \neq 0$ and $g$ is differentiable at $f(c)$. Use the chain rule to find a formula for $g'(f(c))$ (in terms of $f'(c)$).

Exercise 4.1.11: Suppose $f : I \to \mathbb{R}$ is bounded, $g : I \to \mathbb{R}$ is differentiable at $c \in I$, and $g'(c) = 0$. Show that $h(x) := f(x)g(x)$ is differentiable at $c$. Hint: You cannot apply the product rule.

Exercise 4.1.12: Suppose $f : I \to \mathbb{R}$, $g : I \to \mathbb{R}$, and $h : I \to \mathbb{R}$, are functions. Suppose $c \in I$ is such that $f(c) = g(c) = h(c)$, and $f$ and $g$ and $h$ are differentiable at $c$, and $g'(c) = h'(c)$. Furthermore suppose $h(x) \leq f(x) \leq g(x)$ for all $x \in I$. Prove $f$ is differentiable at $c$ and $f'(c) = g'(c) = h'(c)$.

Exercise 4.1.13: Suppose $f : (-1, 1) \to \mathbb{R}$ is a function such that $f(x) = xh(x)$ for a bounded function $h$.

a) Show that $g(x) := (f(x))^2$ is differentiable at the origin and $g'(0) = 0$.

b) Find an example of a continuous function $f : (-1, 1) \to \mathbb{R}$ with $f(0) = 0$, but such that $g(x) := (f(x))^2$ is not differentiable at the origin.

Exercise 4.1.14: Suppose $f : I \to \mathbb{R}$ is differentiable at $c \in I$. Prove there exist numbers $a$ and $b$ with the property that for every $\varepsilon > 0$, there is a $\delta > 0$, such that $|a + b(x - c) - f(x)| \leq \varepsilon |x - c|$, whenever $x \in I$ and $|x - c| < \delta$. In other words, show that there exists a function $g : I \to \mathbb{R}$ such that $\lim_{x \to c} g(x) = 0$ and $|a + b(x - c) - f(x)| = g(x) |x - c|$.

Exercise 4.1.15: Prove the following simple version of L’Hôpital’s rule. Suppose $f : (a, b) \to \mathbb{R}$ and $g : (a, b) \to \mathbb{R}$ are differentiable functions whose derivatives $f'$ and $g'$ are continuous functions. Suppose that at $c \in (a, b)$, $f(c) = 0$, $g(c) = 0$, $g'(x) \neq 0$ for all $x \in (a, b)$, and $g(x) \neq 0$ whenever $x \neq c$. Note that the limit of $f'(x)/g'(x)$ as $x$ goes to $c$ exists. Show that
\[ \lim_{x \to c} \frac{f(x)}{g(x)} = \lim_{x \to c} \frac{f'(x)}{g'(x)}. \]

Exercise 4.1.16: Suppose $f : (a, b) \to \mathbb{R}$ is differentiable at $c \in (a, b)$, $f(c) = 0$, and $f'(c) > 0$. Prove that there is a $\delta > 0$ such that $f(x) < 0$ whenever $c - \delta < x < c$ and $f(x) > 0$ whenever $c < x < c + \delta$. 
4.2 Mean value theorem

Note: 2 lectures (some applications may be skipped)

4.2.1 Relative minima and maxima

We talked about absolute maxima and minima. These are the tallest peaks and lowest valleys in the whole mountain range. What about peaks of individual mountains and bottoms of individual valleys? The derivative, being a local concept, is like walking around in a fog; it can’t tell you if you’re on the highest peak, but it can help you find all the individual peaks.

**Definition 4.2.1.** Let $S \subset \mathbb{R}$ be a set and let $f : S \to \mathbb{R}$ be a function. The function $f$ is said to have a **relative maximum** at $c \in S$ if there exists a $\delta > 0$ such that for all $x \in S$ where $|x - c| < \delta$, we have $f(x) \leq f(c)$. The definition of **relative minimum** is analogous.

**Lemma 4.2.2.** Suppose $f : [a, b] \to \mathbb{R}$ is differentiable at $c \in (a, b)$, and $f$ has a relative minimum or a relative maximum at $c$. Then $f'(c) = 0$.

**Proof.** We prove the statement for a maximum. For a minimum the statement follows by considering the function $-f$.

Let $c$ be a relative maximum of $f$. That is, there is a $\delta > 0$ such that as long as $|x - c| < \delta$, we have $f(x) - f(c) \leq 0$. We look at the difference quotient. If $c < x < c + \delta$, then

$$\frac{f(x) - f(c)}{x - c} \leq 0,$$

and if $c - \delta < y < c$, then

$$\frac{f(y) - f(c)}{y - c} \geq 0.$$

See Figure 4.3 for an illustration.

![Figure 4.3: Slopes of secants at a relative maximum.](image)

As $a < c < b$, there exist sequences $\{x_n\}$ and $\{y_n\}$ in $[a, b]$ and within $\delta$ of $c$, such that $x_n > c$, and $y_n < c$ for all $n \in \mathbb{N}$, and such that $\lim x_n = \lim y_n = c$. Since $f$ is differentiable at $c$,

$$0 \geq \lim_{n \to \infty} \frac{f(x_n) - f(c)}{x_n - c} = f'(c) = \lim_{n \to \infty} \frac{f(y_n) - f(c)}{y_n - c} \geq 0.$$


For a differentiable function, a point where $f'(c) = 0$ is called a **critical point**. When $f$ is not differentiable at some points, it is common to also say that $c$ is a critical point if $f'(c)$ does not exist. The theorem says that a relative minimum or maximum at an interior point of an interval must be a critical point. As you remember from calculus, finding minima and maxima of a function can be done by finding all the critical points together with the endpoints of the interval and simply checking at which of these points is the function biggest or smallest.

### 4.2.2 Rolle’s theorem

Suppose a function has the same value at both endpoints of an interval. Intuitively it ought to attain a minimum or a maximum in the interior of the interval, then at such a minimum or a maximum, the derivative should be zero. See **Figure 4.4** for the geometric idea. This is the content of the so-called Rolle’s theorem*.

![Figure 4.4: Point where the tangent line is horizontal, that is $f'(c) = 0$.](image)

**Theorem 4.2.3** (Rolle). Let $f: [a, b] \to \mathbb{R}$ be continuous function differentiable on $(a, b)$ such that $f(a) = f(b)$. Then there exists a $c \in (a, b)$ such that $f'(c) = 0$.

**Proof.** As $f$ is continuous on $[a, b]$, it attains an absolute minimum and an absolute maximum in $[a, b]$. We wish to apply Lemma 4.2.2, and so we need to find some $c \in (a, b)$ where $f$ attains a minimum or a maximum. Write $K := f(a) = f(b)$. If there exists an $x$ such that $f(x) > K$, then the absolute maximum is bigger than $K$ and hence occurs at some $c \in (a, b)$, and therefore $f'(c) = 0$. On the other hand, if there exists an $x$ such that $f(x) < K$, then the absolute minimum occurs at some $c \in (a, b)$, and so $f'(c) = 0$. If there is no $x$ such that $f(x) > K$ or $f(x) < K$, then $f(x) = K$ for all $x$ and then $f'(x) = 0$ for all $x \in [a, b]$, so any $c \in (a, b)$ works.

It is absolutely necessary for the derivative to exist for all $x \in (a, b)$. Consider the function $f(x) := |x|$ on $[-1, 1]$. Clearly $f(-1) = f(1)$, but there is no point $c$ where $f'(c) = 0$.

*Named after the French mathematician **Michel Rolle** (1652–1719).
4.2. MEAN VALUE THEOREM

4.2.3 Mean value theorem
We extend Rolle’s theorem to functions that attain different values at the endpoints.

Theorem 4.2.4 (Mean value theorem). Let \( f : [a, b] \to \mathbb{R} \) be a continuous function differentiable on \((a, b)\). Then there exists a point \( c \in (a, b) \) such that

\[
f(b) - f(a) = f'(c)(b - a).
\]

For a geometric interpretation of the mean value theorem, see Figure 4.5. The idea is that the value \( \frac{f(b) - f(a)}{b - a} \) is the slope of the line between the points \((a, f(a))\) and \((b, f(b))\). Then \( c \) is the point such that \( f'(c) = \frac{f(b) - f(a)}{b - a} \), that is, the tangent line at the point \((c, f(c))\) has the same slope as the line between \((a, f(a))\) and \((b, f(b))\). The theorem follows from Rolle’s theorem, by subtracting from \( f \) the affine linear function with the derivative \( \frac{f(b) - f(a)}{b - a} \) with the same values at \( a \) and \( b \) as \( f \). That is, we subtract the function whose graph is the straight line \((a, f(a))\) and \((b, f(b))\). Then we are looking for a point where this new function has derivative zero.

\[\text{Figure 4.5: Graphical interpretation of the mean value theorem.}\]

**Proof.** Define the function \( g : [a, b] \to \mathbb{R} \) by

\[
g(x) := f(x) - f(b) - \frac{f(b) - f(a)}{b - a} (x - b).
\]

The function \( g \) is differentiable on \((a, b)\), continuous on \([a, b]\), such that \( g(a) = 0 \) and \( g(b) = 0 \). Thus there exists a \( c \in (a, b) \) such that \( g'(c) = 0 \), that is,

\[
0 = g'(c) = f'(c) - \frac{f(b) - f(a)}{b - a}.
\]

In other words, \( f'(c)(b - a) = f(b) - f(a) \).

The proof generalizes. By considering \( g(x) := f(x) - f(b) - \frac{f(b) - f(a)}{\varphi(b) - \varphi(a)} (\varphi(x) - \varphi(b)) \), one can prove the following version. We leave the proof as an exercise.
Theorem 4.2.5 (Cauchy’s mean value theorem). Let \( f : [a, b] \to \mathbb{R} \) and \( \phi : [a, b] \to \mathbb{R} \) be continuous functions differentiable on \((a, b)\). Then there exists a point \( c \in (a, b) \) such that

\[
(f(b) - f(a)) \phi'(c) = f'(c)(\phi(b) - \phi(a)).
\]

The mean value theorem has the distinction of being one of the few theorems commonly cited in court. That is, when police measure the speed of cars by aircraft, or via cameras reading license plates, they measure the time the car takes to go between two points. The mean value theorem then says that the car must have somewhere attained the speed you get by dividing the difference in distance by the difference in time.

4.2.4 Applications

We now solve our very first differential equation.

Proposition 4.2.6. Let \( I \) be an interval and let \( f : I \to \mathbb{R} \) be a differentiable function such that \( f'(x) = 0 \) for all \( x \in I \). Then \( f \) is constant.

Proof. Take arbitrary \( x, y \in I \) with \( x < y \). As \( I \) is an interval, \([x, y] \subset I\). Then \( f \) restricted to \([x, y]\) satisfies the hypotheses of the mean value theorem. Therefore, there is a \( c \in (x, y) \) such that

\[
f(y) - f(x) = f'(c)(y - x).
\]

As \( f'(c) = 0 \), we have \( f(y) = f(x) \). Hence, the function is constant. \(\square\)

Now that we know what it means for the function to stay constant, let us look at increasing and decreasing functions. We say \( f : I \to \mathbb{R} \) is increasing (resp. strictly increasing) if \( x < y \) implies \( f(x) \leq f(y) \) (resp. \( f(x) < f(y) \)). We define decreasing and strictly decreasing in the same way by switching the inequalities for \( f \).

Proposition 4.2.7. Let \( I \) be an interval and let \( f : I \to \mathbb{R} \) be a differentiable function.

(i) \( f \) is increasing if and only if \( f'(x) \geq 0 \) for all \( x \in I \).

(ii) \( f \) is decreasing if and only if \( f'(x) \leq 0 \) for all \( x \in I \).

Proof. Let us prove the first item. Suppose \( f \) is increasing, then for all \( x, c \in I \) with \( x \neq c \), we have

\[
\frac{f(x) - f(c)}{x - c} \geq 0.
\]

Taking a limit as \( x \) goes to \( c \) we see that \( f'(c) \geq 0 \).

For the other direction, suppose \( f'(x) \geq 0 \) for all \( x \in I \). Take any \( x, y \in I \) where \( x < y \), and note that \([x, y] \subset I\). By the mean value theorem, there is some \( c \in (x, y) \) such that

\[
f(y) - f(x) = f'(c)(y - x).
\]

As \( f'(c) \geq 0 \), and \( y - x > 0 \), then \( f(y) - f(x) \geq 0 \) or \( f(x) \leq f(y) \) and so \( f \) is increasing.

We leave the decreasing part to the reader as exercise. \(\square\)
A similar but weaker statement is true for strictly increasing and decreasing functions.

**Proposition 4.2.8.** Let $I$ be an interval and let $f : I \to \mathbb{R}$ be a differentiable function.

(i) If $f'(x) > 0$ for all $x \in I$, then $f$ is strictly increasing.

(ii) If $f'(x) < 0$ for all $x \in I$, then $f$ is strictly decreasing.

The proof of (i) is left as an exercise. Then (ii) follows from (i) by considering $-f$ instead. The converse of this proposition is not true. The function $f(x) := x^3$ is strictly increasing, but $f'(0) = 0$.

Another application of the **mean value theorem** is the following result about location of extrema, sometimes called the **first derivative test**. The theorem is stated for an absolute minimum and maximum. To apply it to find relative minima and maxima, restrict $f$ to an interval $(c - \delta, c + \delta)$.

**Proposition 4.2.9.** Let $f : (a, b) \to \mathbb{R}$ be continuous. Let $c \in (a, b)$ and suppose $f$ is differentiable on $(a, c)$ and $(c, b)$.

(i) If $f'(x) \leq 0$ for $x \in (a, c)$ and $f'(x) \geq 0$ for $x \in (c, b)$, then $f$ has an absolute minimum at $c$.

(ii) If $f'(x) \geq 0$ for $x \in (a, c)$ and $f'(x) \leq 0$ for $x \in (c, b)$, then $f$ has an absolute maximum at $c$.

**Proof.** We prove the first item and leave the second to the reader. Take $x \in (a, c)$ and a sequence $\{y_n\}$ such that $x < y_n < c$ for all $n$ and $\lim y_n = c$. By the preceding proposition, $f$ is decreasing on $(a, c)$ so $f(x) \geq f(y_n)$. As $f$ is continuous at $c$, we take the limit to get $f(x) \geq f(c)$ for all $x \in (a, c)$.

Similarly, take $x \in (c, b)$ and $\{y_n\}$ a sequence such that $c < y_n < x$ and $\lim y_n = c$. The function is increasing on $(c, b)$ so $f(x) \geq f(y_n)$. By continuity of $f$ we get $f(x) \geq f(c)$ for all $x \in (c, b)$. Thus $f(x) \geq f(c)$ for all $x \in (a, b)$.

The converse of the proposition does not hold. See **Example 4.2.12** below.

Another often used application of the mean value theorem you have possibly seen in calculus is the following result on differentiability at the end points of an interval. The proof is **Exercise 4.2.13**.

**Proposition 4.2.10.**

(i) Suppose $f : [a, b) \to \mathbb{R}$ is continuous, differentiable in $(a, b)$, and $\lim_{x \to a} f'(x) = L$. Then $f$ is differentiable at $a$ and $f'(a) = L$.

(ii) Suppose $f : (a, b] \to \mathbb{R}$ is continuous, differentiable in $(a, b)$, and $\lim_{x \to b} f'(x) = L$. Then $f$ is differentiable at $b$ and $f'(b) = L$.

In fact, using the extension result **Proposition 3.4.6**, you do not need to assume that $f$ is defined at the end point. See **Exercise 4.2.14**.

### 4.2.5 Continuity of derivatives and the intermediate value theorem

Derivatives of functions satisfy an intermediate value property.

**Theorem 4.2.11** (Darboux). Let $f : [a, b] \to \mathbb{R}$ be differentiable. Suppose $y \in \mathbb{R}$ is such that $f'(a) < y < f'(b)$ or $f'(a) > y > f'(b)$. Then there exists a $c \in (a, b)$ such that $f'(c) = y$. 
The proof follows by subtracting \( f \) and a linear function with derivative \( y \). The new function \( g \) reduces the problem to the case \( y = 0 \), where \( g'(a) > 0 > g'(b) \). That is, \( g \) is increasing at \( a \) and decreasing at \( b \), so it must attain a maximum inside \((a,b)\), where the derivative is zero. See Figure 4.6.

**Proof.** Suppose \( f'(a) < y < f'(b) \). Define

\[
g(x) := yx - f(x).
\]

The function \( g \) is continuous on \([a,b]\), and so \( g \) attains a maximum at some \( c \in [a,b] \).

The function \( g \) is also differentiable on \([a,b]\). Compute \( g'(x) = y - f'(x) \). Thus \( g'(a) > 0 \). As the derivative is the limit of difference quotients and is positive, there must be some difference quotient that is positive. That is, there must exist an \( x > a \) such that

\[
\frac{g(x) - g(a)}{x-a} > 0,
\]

or \( g(x) > g(a) \). Thus \( g \) cannot possibly have a maximum at \( a \). Similarly, as \( g'(b) < 0 \), we find an \( x < b \) (a different \( x \)) such that \( \frac{g(x) - g(b)}{x-b} < 0 \) or that \( g(x) > g(b) \), thus \( g \) cannot possibly have a maximum at \( b \). Therefore, \( c \in (a,b) \), and Lemma 4.2.2 applies: As \( g \) attains a maximum at \( c \) we find \( g'(c) = 0 \) and so \( f'(c) = y \).

Similarly, if \( f'(a) > y > f'(b) \), consider \( g(x) := f(x) - yx \).

We have seen already that there exist discontinuous functions that have the intermediate value property. While it is hard to imagine at first, there also exist functions that are differentiable everywhere and the derivative is not continuous.

**Example 4.2.12:** Let \( f : \mathbb{R} \to \mathbb{R} \) be the function defined by

\[
f(x) := \begin{cases} 
(x \sin(1/x))^2 & \text{if } x \neq 0, \\
0 & \text{if } x = 0.
\end{cases}
\]

We claim that \( f \) is differentiable everywhere, but \( f' : \mathbb{R} \to \mathbb{R} \) is not continuous at the origin. Furthermore, \( f \) has a minimum at 0, but the derivative changes sign infinitely often near the origin. See Figure 4.7.
4.2. MEAN VALUE THEOREM

Figure 4.7: A function with a discontinuous derivative. The function $f$ is on the left and $f'$ is on the right. Notice that $f(x) \leq x^2$ on the left graph.

Proof: It is immediate from the definition that $f$ has an absolute minimum at 0; we know $f(x) \geq 0$ for all $x$ and $f(0) = 0$.

The function $f$ is differentiable for $x \neq 0$, and the derivative is $2 \sin(1/x)(x \sin(1/x) - \cos(1/x))$.

As an exercise, show that for $x_n = \frac{4}{(8n+1)\pi}$, we have $\lim f'(x_n) = -1$, and for $y_n = \frac{4}{(8n+3)\pi}$, we have $\lim f'(y_n) = 1$. Hence if $f'$ exists at 0, then it cannot be continuous.

Let us show that $f'$ exists at 0. We claim that the derivative is zero. In other words,

$$\left| \frac{f(x) - f(0)}{x - 0} - 0 \right|$$

goes to zero as $x$ goes to zero. For $x \neq 0$,

$$\left| \frac{f(x) - f(0)}{x - 0} - 0 \right| = \left| \frac{x^2 \sin^2(1/x)}{x} \right| = \left| x \sin^2(1/x) \right| \leq |x|.$$

And, of course, as $x$ tends to zero, $|x|$ tends to zero, and hence $\left| \frac{f(x) - f(0)}{x - 0} - 0 \right|$ goes to zero. Therefore, $f$ is differentiable at 0 and the derivative at 0 is 0. A key point in the calculation above is that $|f(x)| \leq x^2$, see also Exercises 4.1.11 and 4.1.12.

It is sometimes useful to assume the derivative of a differentiable function is continuous. If $f: I \rightarrow \mathbb{R}$ is differentiable and the derivative $f'$ is continuous on $I$, then we say $f$ is *continuously differentiable*. It is common to write $C^1(I)$ for the set of continuously differentiable functions on $I$.

4.2.6 Exercises

**Exercise 4.2.1:** Finish the proof of Proposition 4.2.7.

**Exercise 4.2.2:** Finish the proof of Proposition 4.2.9.

**Exercise 4.2.3:** Suppose $f: \mathbb{R} \rightarrow \mathbb{R}$ is a differentiable function such that $f'$ is a bounded function. Prove that $f$ is a Lipschitz continuous function.
Exercise 4.2.4: Suppose \( f: [a,b] \to \mathbb{R} \) is differentiable and \( c \in [a,b] \). Show there exists a sequence \( \{x_n\} \) converging to \( c \), \( x_n \neq c \) for all \( n \), such that
\[
f'(c) = \lim_{n \to \infty} f'(x_n).
\]
Do note this does not imply that \( f' \) is continuous (why?).

Exercise 4.2.5: Suppose \( f: \mathbb{R} \to \mathbb{R} \) is a function such that \( |f(x) - f(y)| \leq |x - y|^2 \) for all \( x \) and \( y \). Show that \( f(x) = C \) for some constant \( C \). Hint: Show that \( f \) is differentiable at all points and compute the derivative.

Exercise 4.2.6: Finish the proof of Proposition 4.2.8. That is, suppose \( I \) is an interval and \( f: I \to \mathbb{R} \) is a differentiable function such that \( f'(x) > 0 \) for all \( x \in I \). Show that \( f \) is strictly increasing.

Exercise 4.2.7: Suppose \( f: (a,b) \to \mathbb{R} \) is a differentiable function such that \( f'(x) \neq 0 \) for all \( x \in (a,b) \). Suppose there exists a point \( c \in (a,b) \) such that \( f'(c) > 0 \). Prove \( f'(x) > 0 \) for all \( x \in (a,b) \).

Exercise 4.2.8: Suppose \( f: (a,b) \to \mathbb{R} \) and \( g: (a,b) \to \mathbb{R} \) are differentiable functions such that \( f'(x) = g'(x) \) for all \( x \in (a,b) \), then show that there exists a constant \( C \) such that \( f(x) = g(x) + C \).

Exercise 4.2.9: Prove the following version of L'Hôpital's rule. Suppose \( f: (a,b) \to \mathbb{R} \) and \( g: (a,b) \to \mathbb{R} \) are differentiable functions and \( c \in (a,b) \). Suppose that \( f(c) = 0, g(c) = 0, g'(x) \neq 0 \) when \( x \neq c \), and that the limit of \( f'(x)/g'(x) \) as \( x \) goes to \( c \) exists. Show that
\[
\lim_{x \to c} \frac{f(x)}{g(x)} = \frac{\lim_{x \to c} f'(x)}{\lim_{x \to c} g'(x)}.
\]
Compare to Exercise 4.1.15. Note: Before you do anything else, prove that \( g(x) \neq 0 \) when \( x \neq c \).

Exercise 4.2.10: Let \( f: (a,b) \to \mathbb{R} \) be an unbounded differentiable function. Show \( f': (a,b) \to \mathbb{R} \) is unbounded.

Exercise 4.2.11: Prove the theorem Rolle actually proved in 1691: If \( f \) is a polynomial, \( f'(a) = f'(b) = 0 \) for some \( a < b \), and there is no \( c \in (a,b) \) such that \( f'(c) = 0 \), then there is at most one root of \( f \) in \( (a,b) \), that is at most one \( x \in (a,b) \) such that \( f(x) = 0 \). In other words, between any two consecutive roots of \( f' \) is at most one root of \( f \). Hint: Suppose there are two roots and see what happens.

Exercise 4.2.12: Suppose \( a,b \in \mathbb{R} \) and \( f: \mathbb{R} \to \mathbb{R} \) is differentiable, \( f'(x) = a \) for all \( x \), and \( f(0) = b \). Find \( f \) and prove that it is the unique differentiable function with this property.

Exercise 4.2.13:
\begin{itemize}
  \item a) Prove Proposition 4.2.10.
  \item b) Suppose \( f: (a,b) \to \mathbb{R} \) is continuous, and suppose \( f \) is differentiable everywhere except at \( c \in (a,b) \) and \( \lim_{x \to c} f'(x) = L \). Prove that \( f \) is differentiable at \( c \) and \( f'(c) = L \).
\end{itemize}

Exercise 4.2.14: Suppose \( f: (0,1) \to \mathbb{R} \) is differentiable and \( f' \) is bounded.
\begin{itemize}
  \item a) Show that there exists a continuous function \( g: [0,1] \to \mathbb{R} \) such that \( f(x) = g(x) \) for all \( x \neq 0 \). Hint: Proposition 3.4.6 and Exercise 4.2.3.
  \item b) Find an example where the \( g \) is not differentiable at \( x = 0 \). Hint: Consider something based on \( \sin(\ln(x)) \), and assume you know basic properties of \( \sin \) and \( \ln \) from calculus.
  \item c) Instead of assuming that \( f' \) is bounded, assume that \( \lim_{x \to 0} f'(x) = L \). Prove that not only does \( g \) exist but it is differentiable at 0 and \( g'(0) = L \).
\end{itemize}

Exercise 4.2.15: Prove Theorem 4.2.5.
4.3 Taylor’s theorem

Note: less than a lecture (optional section)

4.3.1 Derivatives of higher orders

When \( f : I \to \mathbb{R} \) is differentiable, we obtain a function \( f' : I \to \mathbb{R} \). The function \( f' \) is called the first derivative of \( f \). If \( f' \) is differentiable, we denote by \( f'' : I \to \mathbb{R} \) the derivative of \( f' \). The function \( f'' \) is called the second derivative of \( f \). We similarly obtain \( f''' \), \( f'''' \), and so on. With a larger number of derivatives the notation would get out of hand; we denote by \( f^{(n)} \) the \( n \)th derivative of \( f \).

When \( f \) possesses \( n \) derivatives, we say \( f \) is \( n \) times differentiable.

4.3.2 Taylor’s theorem

Taylor’s theorem* is a generalization of the mean value theorem. Mean value theorem says that up to a small error \( f(x) \) for \( x \) near \( x_0 \) can be approximated by \( f(x_0) \), that is

\[
f(x) = f(x_0) + f'(c)(x - x_0),
\]

where the “error” is measured in terms of the first derivative at some point \( c \) between \( x \) and \( x_0 \). Taylor’s theorem generalizes this result to higher derivatives. It tells us that up to a small error, any \( n \) times differentiable function can be approximated at a point \( x_0 \) by a polynomial. The error of this approximation behaves like \( (x - x_0)^n \) near the point \( x_0 \). To see why this is a good approximation notice that for a big \( n \), \( (x - x_0)^n \) is very small in a small interval around \( x_0 \).

Definition 4.3.1. For an \( n \) times differentiable function \( f \) defined near a point \( x_0 \in \mathbb{R} \), define the \( n \)th order Taylor polynomial for \( f \) at \( x_0 \) as

\[
P_n^{x_0}(x) := \sum_{k=0}^{n} \frac{f^{(k)}(x_0)}{k!}(x - x_0)^k
\]

\[
= f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2}(x - x_0)^2 + \frac{f'''(x_0)}{6}(x - x_0)^3 + \cdots + \frac{f^{(n)}(x_0)}{n!}(x - x_0)^n.
\]

See Figure 4.8 for the odd degree Taylor polynomials for the sine function at \( x_0 = 0 \). The even degree terms are all zero, as even derivatives of sine are again a sine, which are zero at the origin.

Taylor’s theorem says a function behaves like its \( n \)th Taylor polynomial. The mean value theorem is really Taylor’s theorem for the first derivative.

Theorem 4.3.2 (Taylor). Suppose \( f : [a, b] \to \mathbb{R} \) is a function with \( n \) continuous derivatives on \([a, b]\) and such that \( f^{(n+1)} \) exists on \((a, b)\). Given distinct points \( x_0 \) and \( x \) in \([a, b]\), we can find a point \( c \) between \( x_0 \) and \( x \) such that

\[
f(x) = P_n^{x_0}(x) + \frac{f^{(n+1)}(c)}{(n+1)!}(x - x_0)^{n+1}.
\]

*Named for the English mathematician Brook Taylor (1685–1731). It was first found by the Scottish mathematician James Gregory (1638–1675). The statement we give was proved by Joseph-Louis Lagrange (1736–1813).
The term $R_{x_0}^n(x) := \frac{f^{(n+1)}(c)}{(n+1)!} (x-x_0)^{n+1}$ is called the remainder term. This form of the remainder term is called the Lagrange form of the remainder. There are other ways to write the remainder term, but we skip those. Note that $c$ depends on both $x$ and $x_0$.

**Proof.** Find a number $M_{x,x_0}$ (depending on $x$ and $x_0$) solving the equation

$$f(x) = P_{x_0}^n(x) + M_{x,x_0}(x-x_0)^{n+1}.$$ 

Define a function $g(s)$ by

$$g(s) := f(s) - P_{x_0}^n(s) - M_{x,x_0}(s-x_0)^{n+1}.$$ 

We compute the $k$th derivative at $x_0$ of the Taylor polynomial $(P_{x_0}^n)^{(k)}(x_0) = f^{(k)}(x_0)$ for $k = 0, 1, 2, \ldots, n$ (the zeroth derivative of a function is the function itself). Therefore,

$$g(x_0) = g'(x_0) = g''(x_0) = \cdots = g^{(n)}(x_0) = 0.$$ 

In particular, $g(x_0) = 0$. On the other hand $g(x) = 0$. By the mean value theorem there exists an $x_1$ between $x_0$ and $x$ such that $g'(x_1) = 0$. Applying the mean value theorem to $g'$ we obtain that there exists $x_2$ between $x_0$ and $x_1$ (and therefore between $x_0$ and $x$) such that $g''(x_2) = 0$. We repeat the argument $n+1$ times to obtain a number $x_{n+1}$ between $x_0$ and $x_n$ (and therefore between $x_0$ and $x$) such that $g^{(n+1)}(x_{n+1}) = 0$.

Let $c := x_{n+1}$. We compute the $(n+1)$th derivative of $g$ to find

$$g^{(n+1)}(s) = f^{(n+1)}(s) - (n+1)! M_{x,x_0}.$$ 

Plugging in $c$ for $s$ we obtain $M_{x,x_0} = \frac{f^{(n+1)}(c)}{(n+1)!}$, and we are done. 

In the proof, we have computed $(P_{x_0}^n)^{(k)}(x_0) = f^{(k)}(x_0)$ for $k = 0, 1, 2, \ldots, n$. Therefore, the Taylor polynomial has the same derivatives as $f$ at $x_0$ up to the $n$th derivative. That is why the
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Taylor polynomial is a good approximation to $f$. Notice how in Figure 4.8 the Taylor polynomials are reasonably good approximations to the sine near $x = 0$.

We do not necessarily get good approximations by the Taylor polynomial everywhere. Consider expanding the function $f(x) := \frac{1}{1-x}$ around $0$, for $x < 1$, we get the graphs in Figure 4.9. The dotted lines are the first, second, and third degree approximations. The dashed line is the 20th degree polynomial, and yet the approximation only seems to get better with the degree for $x > -1$, and for smaller $x$, it in fact gets worse. The polynomials are the partial sums of the geometric series $\sum_{n=1}^\infty x^n$, and the series only converges on $(-1, 1)$. See the discussion of power series §2.6.

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{figure4.9}
\caption{The function $\frac{1}{1-x}$, and the Taylor polynomials $P_1^0$, $P_2^0$, $P_3^0$ (all dotted), and the polynomial $P_{20}^0$ (dashed).}
\end{figure}

If $f$ is infinitely differentiable, that is, if $f$ can be differentiated any number of times, then we define the Taylor series:

$$\sum_{k=0}^\infty \frac{f^{(k)}(x_0)}{k!}(x-x_0)^k.$$ 

There is no guarantee that this series converges for any $x \neq x_0$. And even where it does converge, there is no guarantee that it converges to the function $f$. Functions $f$ whose Taylor series at every point $x_0$ converges to $f$ in some open interval containing $x_0$ are called analytic functions. Most functions one tends to see in practice are analytic. See Exercise 5.4.11, for an example of a non-analytic function.

The definition of derivative says that a function is differentiable if it is locally approximated by a line. We mention in passing that there exists a converse to Taylor’s theorem, which we will neither state nor prove, saying that if a function is locally approximated in a certain way by a polynomial of degree $d$, then it has $d$ derivatives.

Taylor’s theorem gives us a quick proof of a version of the second derivative test. By a strict relative minimum of $f$ at $c$, we mean that there exists a $\delta > 0$ such that $f(x) > f(c)$ for all $x \in (c - \delta, c + \delta)$ where $x \neq c$. A strict relative maximum is defined similarly. Continuity of the second derivative is not needed, but the proof is more difficult and is left as an exercise. The proof also generalizes immediately into the $n$th derivative test, which is also left as an exercise.
Proposition 4.3.3 (Second derivative test). Suppose $f : (a,b) \to \mathbb{R}$ is twice continuously differentiable, $x_0 \in (a,b)$, $f'(x_0) = 0$ and $f''(x_0) > 0$. Then $f$ has a strict relative minimum at $x_0$.

Proof. As $f''$ is continuous, there exists a $\delta > 0$ such that $f''(c) > 0$ for all $c \in (x_0 - \delta, x_0 + \delta)$, see Exercise 3.2.11. Take $x \in (x_0 - \delta, x_0 + \delta)$, $x \neq x_0$. Taylor’s theorem says that for some $c$ between $x_0$ and $x$,

$$f(x) = f(x_0) + f'(x_0)(x-x_0) + \frac{f''(c)}{2}(x-x_0)^2 = f(x_0) + \frac{f''(c)}{2}(x-x_0)^2.$$

As $f''(c) > 0$, and $(x-x_0)^2 > 0$, then $f(x) > f(x_0).$ \hfill $\square$

4.3.3 Exercises

Exercise 4.3.1: Compute the $n$th Taylor polynomial at $0$ for the exponential function.

Exercise 4.3.2: Suppose $p$ is a polynomial of degree $d$. Given $x_0 \in \mathbb{R}$, show that the $d$th Taylor polynomial for $p$ at $x_0$ is equal to $p$.

Exercise 4.3.3: Let $f(x) := |x|^3$. Compute $f'(x)$ and $f''(x)$ for all $x$, but show that $f^{(3)}(0)$ does not exist.

Exercise 4.3.4: Suppose $f : \mathbb{R} \to \mathbb{R}$ has $n$ continuous derivatives. Show that for every $x_0 \in \mathbb{R}$, there exist polynomials $P$ and $Q$ of degree $n$ and an $\varepsilon > 0$ such that $P(x) \leq f(x) \leq Q(x)$ for all $x \in [x_0, x_0 + \varepsilon]$ and $Q(x) - P(x) = \lambda (x-x_0)^n$ for some $\lambda \geq 0$.

Exercise 4.3.5: If $f : [a,b] \to \mathbb{R}$ has $n+1$ continuous derivatives and $x_0 \in (a,b)$, prove $\lim_{x \to x_0} \frac{R_n(x)}{(x-x_0)^n} = 0$.

Exercise 4.3.6: Suppose $f : [a,b] \to \mathbb{R}$ has $n+1$ continuous derivatives and $x_0 \in (a,b)$. Prove: $f^{(k)}(x_0) = 0$ for all $k = 0, 1, 2, \ldots, n$ if and only if $\lim_{x \to x_0} \frac{f(x)}{(x-x_0)^n}$ exists.

Exercise 4.3.7: Suppose $a, b, c \in \mathbb{R}$ and $f : \mathbb{R} \to \mathbb{R}$ is differentiable, $f''(x) = a$ for all $x$, $f'(0) = b$, and $f(0) = c$. Find $f$ and prove that it is the unique differentiable function with this property.

Exercise 4.3.8 (Challenging): Show that a simple converse to Taylor’s theorem does not hold. Find a function $f : \mathbb{R} \to \mathbb{R}$ with no second derivative at $x = 0$ such that $|f(x)| \leq |x^3|$, that is, $f$ goes to zero at 0 faster than $x^3$, and while $f'(0)$ exists, $f''(0)$ does not.

Exercise 4.3.9: Suppose $f : (0,1) \to \mathbb{R}$ is differentiable and $f''$ is bounded.

a) Show that there exists a once differentiable function $g : [0,1) \to \mathbb{R}$ such that $f(x) = g(x)$ for all $x \neq 0$.

   Hint: See Exercise 4.2.14.

b) Find an example where the $g$ is not twice differentiable at $x = 0$.

Exercise 4.3.10: Prove the $n$th derivative test. Suppose $n \in \mathbb{N}$, $x_0 \in (a,b)$, and $f : (a,b) \to \mathbb{R}$ is $n$ times continuously differentiable, with $f^{(k)}(x_0) = 0$ for $k = 1, 2, \ldots, n-1$, and $f^{(n)}(x_0) \neq 0$. Prove:

a) If $n$ is odd, then $f$ has neither a relative minimum, nor a maximum at $x_0$.

b) If $n$ is even, then $f$ has a strict relative minimum at $x_0$ if $f^{(n)}(x_0) > 0$ and a strict relative maximum at $x_0$ if $f^{(n)}(x_0) < 0$.

Exercise 4.3.11: Prove the more general version of the second derivative test. Suppose $f : (a,b) \to \mathbb{R}$ is differentiable and $x_0 \in (a,b)$ is such that, $f'(x_0) = 0$, $f''(x_0)$ exists, and $f''(x_0) > 0$. Prove that $f$ has a strict relative minimum at $x_0$. Hint: Consider the limit definition of $f''(x_0)$. 


4.4. INVERSE FUNCTION THEOREM

4.4 Inverse function theorem

Note: less than 1 lecture (optional section, needed for §5.4, requires §3.6)

4.4.1 Inverse function theorem

We start with a simple example. Consider the function \( f(x) := ax \) for a number \( a \neq 0 \). Then \( f: \mathbb{R} \to \mathbb{R} \) is bijective, and the inverse is \( f^{-1}(y) = \frac{1}{a}y \). In particular, \( f'(x) = a \) and \( (f^{-1})'(y) = \frac{1}{a} \).

As differentiable functions are “infinitesimally like” linear functions, we expect the same behavior from the inverse function. The main idea of differentiating inverse functions is the following lemma.

**Lemma 4.4.1.** Let \( I, J \subset \mathbb{R} \) be intervals. If \( f: I \to J \) is strictly monotone (hence one-to-one), onto (\( f(I) = J \)), differentiable at \( x_0 \in I \), and \( f'(x_0) \neq 0 \), then the inverse \( f^{-1} \) is differentiable at \( y_0 = f(x_0) \) and

\[
(f^{-1})'(y_0) = \frac{1}{f'(f^{-1}(y_0))} = \frac{1}{f'(x_0)}.
\]

If \( f \) is continuously differentiable and \( f' \) is never zero, then \( f^{-1} \) is continuously differentiable.

**Proof.** By Proposition 3.6.6, \( f \) has a continuous inverse. For convenience call the inverse \( g: J \to I \). Let \( x_0, y_0 \) be as in the statement. For \( x \in I \) write \( y = f(x) \). If \( x \neq x_0 \) and so \( y \neq y_0 \), we find

\[
\frac{g(y) - g(y_0)}{y - y_0} = \frac{g(f(x)) - g(f(x_0))}{f(x) - f(x_0)} = \frac{x - x_0}{f(x) - f(x_0)}.
\]

See Figure 4.10 for the geometric idea.

---

**Figure 4.10:** Interpretation of the derivative of the inverse function.

Let

\[
Q(x) := \begin{cases} 
\frac{x-x_0}{f(x)-f(x_0)} & \text{if } x \neq x_0, \\
\frac{1}{f'(x_0)} & \text{if } x = x_0 \quad (\text{notice that } f'(x_0) \neq 0).
\end{cases}
\]
As $f$ is differentiable at $x_0$, \[
\lim_{x \to x_0} Q(x) = \lim_{x \to x_0} \frac{x - x_0}{f(x) - f(x_0)} = \frac{1}{f'(x_0)} = Q(x_0),
\]
that is, $Q$ is continuous at $x_0$. As $g(y)$ is continuous at $y_0$, the composition $Q(g(y)) = \frac{g(y) - g(y_0)}{y - y_0}$ is continuous at $y_0$ by Proposition 3.2.7. Therefore, \[
\frac{1}{f'(g(y_0))} = Q(g(y_0)) = \lim_{y \to y_0} Q(g(y)) = \lim_{y \to y_0} \frac{g(y) - g(y_0)}{y - y_0}.
\]
So $g$ is differentiable at $y_0$ and $g'(y_0) = \frac{1}{f'(g(y_0))}$.

If $f'$ is continuous and nonzero at all $x \in I$, then the lemma applies at all $x \in I$. As $g$ is also continuous (it is differentiable), the derivative $g'(y) = \frac{1}{f'(g(y))}$ must be continuous. \hfill \Box

What is usually called the inverse function theorem is the following result.

**Theorem 4.4.2** (Inverse function theorem). Let $f : (a, b) \to \mathbb{R}$ be a continuously differentiable function, $x_0 \in (a, b)$ a point where $f'(x_0) \neq 0$. Then there exists an open interval $I \subset (a, b)$ with $x_0 \in I$, the restriction $f|_I$ is injective with a continuously differentiable inverse $g : J \to I$ defined on an interval $J := f(I)$, and \[
g'(y) = \frac{1}{f'(g(y))} \quad \text{for all } y \in J.
\]

**Proof.** Without loss of generality, suppose $f'(x_0) > 0$. As $f'$ is continuous, there must exist an open interval $I = (x_0 - \delta, x_0 + \delta)$ such that $f'(x) > 0$ for all $x \in I$. See Exercise 3.2.11.

By Proposition 4.2.8, $f$ is strictly increasing on $I$, and hence the restriction $f|_I$ is bijective onto $J := f(I)$. As $f$ is continuous, then by the Corollary 3.6.3 (or directly via the intermediate value theorem) $f(I)$ is in interval. Now apply Lemma 4.4.1. \hfill \Box

If you tried to prove the existence of roots directly as in Example 1.2.3, you saw how difficult that endeavor is. However, with the machinery we have built for inverse functions it becomes an almost trivial exercise, and with the lemma above we prove far more than mere existence.

**Corollary 4.4.3.** Given $n \in \mathbb{N}$ and $x \geq 0$ there exists a unique number $y \geq 0$ (denoted $x^{1/n} := y$), such that $y^n = x$. Furthermore, the function $g : (0, \infty) \to (0, \infty)$ defined by $g(x) := x^{1/n}$ is continuously differentiable and \[
g'(x) = \frac{1}{nx^{(n-1)/n}} = \frac{1}{n} x^{(1-n)/n},
\]
using the convention $x^{m/n} := (x^{1/n})^m$.

**Proof.** For $x = 0$ the existence of a unique root is trivial.

Let $f : (0, \infty) \to (0, \infty)$ be defined by $f(y) := y^n$. The function $f$ is continuously differentiable and $f'(y) = ny^{n-1}$, see Exercise 4.1.3. For $y > 0$ the derivative $f'$ is strictly positive and so again by Proposition 4.2.8, $f$ is strictly increasing (this can also be proved directly). Given any $M > 1$, $f(M) = M^n \geq M$, and given any $1 > \varepsilon > 0$, $f(\varepsilon) = \varepsilon^n \leq \varepsilon$. For every $x$ with $\varepsilon < x < M$,
we have, by the intermediate value theorem, that $x \in f([\epsilon, M]) \subset f((0, \infty))$. As $M$ and $\epsilon$ were arbitrary, $f$ is onto $(0, \infty)$, and hence $f$ is bijective. Let $g$ be the inverse of $f$, and we obtain the existence and uniqueness of positive $n$th roots. Lemma 4.4.1 says $g$ has a continuous derivative and $g'(x) = \frac{1}{f'(g(x))} = \frac{1}{n(x^{1/n})^{n-1}}$.

Example 4.4.4: The corollary provides a good example of where the inverse function theorem gives us an interval smaller than $(a, b)$. Take $f : \mathbb{R} \to \mathbb{R}$ defined by $f(x) := x^2$. Then $f'(x) \neq 0$ as long as $x \neq 0$. If $x_0 > 0$, we can take $I = (0, \infty)$, but no larger.

Example 4.4.5: Another useful example is $f(x) := x^3$. The function $f : \mathbb{R} \to \mathbb{R}$ is one-to-one and onto, so $f^{-1}(y) = y^{1/3}$ exists on the entire real line including zero and negative $y$. The function $f$ has a continuous derivative, but $f^{-1}$ has no derivative at the origin. The point is that $f'(0) = 0$. See Figure 4.11 for a graph, notice the vertical tangent on the cube root at the origin. See also Exercise 4.4.4.

![Figure 4.11: Graphs of $x^3$ and $x^{1/3}$.](image)

4.4.2 Exercises

Exercise 4.4.1: Suppose $f : \mathbb{R} \to \mathbb{R}$ is continuously differentiable such that $f'(x) > 0$ for all $x$. Show that $f$ is invertible on the interval $J = f(\mathbb{R})$, the inverse is continuously differentiable, and $(f^{-1})'(y) > 0$ for all $y \in f(\mathbb{R})$.

Exercise 4.4.2: Suppose $I, J$ are intervals and a monotone onto $f : I \to J$ has an inverse $g : J \to I$. Suppose you already know that both $f$ and $g$ are differentiable everywhere and $f'$ is never zero. Using chain rule but not Lemma 4.4.1 prove the formula $g'(y) = \frac{1}{f'(g(y))}$.

Exercise 4.4.3: Let $n \in \mathbb{N}$ be even. Prove that every $x > 0$ has a unique negative nth root. That is, there exists a negative number $y$ such that $y^n = x$. Compute the derivative of the function $g(x) := y$.

Exercise 4.4.4: Let $n \in \mathbb{N}$ be odd and $n \geq 3$. Prove that every $x$ has a unique nth root. That is, there exists a number $y$ such that $y^n = x$. Prove that the function defined by $g(x) := y$ is differentiable except at $x = 0$ and compute the derivative. Prove that $g$ is not differentiable at $x = 0$. 
Exercise 4.4.5 (requires §4.3): Show that if in the inverse function theorem \( f \) has \( k \) continuous derivatives, then the inverse function \( g \) also has \( k \) continuous derivatives.

Exercise 4.4.6: Let \( f(x) := x + 2x^2 \sin(1/x) \) for \( x \neq 0 \) and \( f(0) := 0 \). Show that \( f \) is differentiable at all \( x \), that \( f'(0) > 0 \), but that \( f \) is not invertible on any open interval containing the origin.

Exercise 4.4.7:

a) Let \( f: \mathbb{R} \to \mathbb{R} \) be a continuously differentiable function and \( k > 0 \) be a number such that \( f'(x) \geq k \) for all \( x \in \mathbb{R} \). Show \( f \) is one-to-one and onto, and has a continuously differentiable inverse \( f^{-1}: \mathbb{R} \to \mathbb{R} \).

b) Find an example \( f: \mathbb{R} \to \mathbb{R} \) where \( f'(x) > 0 \) for all \( x \), but \( f \) is not onto.

Exercise 4.4.8: Suppose \( I, J \) are intervals and a monotone onto \( f: I \to J \) has an inverse \( g: J \to I \). Suppose \( x \in I \) and \( y := f(x) \in J \), and that \( g \) is differentiable at \( y \). Prove:

a) If \( g'(y) \neq 0 \), then \( f \) is differentiable at \( x \).

b) If \( g'(y) = 0 \), then \( f \) is not differentiable at \( x \).
Chapter 5

The Riemann Integral

5.1 The Riemann integral

Note: 1.5 lectures

An integral is a way to “sum” the values of a function. There is often confusion among students of calculus between integral and antiderivative. The integral is (informally) the area under the curve, nothing else. That we can compute an antiderivative using the integral is a nontrivial result we have to prove. In this chapter we define the Riemann integral* using the Darboux integral†, which is technically simpler than (but equivalent to) the traditional definition of Riemann.

5.1.1 Partitions and lower and upper integrals

We want to integrate a bounded function defined on an interval \([a, b]\). We first define two auxiliary integrals that are defined for all bounded functions. Only then can we talk about the Riemann integral and the Riemann integrable functions.

**Definition 5.1.1.** A partition \(P\) of the interval \([a, b]\) is a finite set of numbers \(\{x_0, x_1, x_2, \ldots, x_n\}\) such that
\[
a = x_0 < x_1 < x_2 < \cdots < x_{n-1} < x_n = b.
\]
We write
\[
\Delta x_i := x_i - x_{i-1}.
\]

Let \(f: [a, b] \to \mathbb{R}\) be a bounded function. Let \(P\) be a partition of \([a, b]\). Define
\[
m_i := \inf \left\{ f(x) : x_{i-1} \leq x \leq x_i \right\},
\]
\[
M_i := \sup \left\{ f(x) : x_{i-1} \leq x \leq x_i \right\},
\]
\[
L(P, f) := \sum_{i=1}^{n} m_i \Delta x_i,
\]
\[
U(P, f) := \sum_{i=1}^{n} M_i \Delta x_i.
\]

*Named after the German mathematician Georg Friedrich Bernhard Riemann (1826–1866).
†Named after the French mathematician Jean-Gaston Darboux (1842–1917).
We call $L(P, f)$ the lower Darboux sum and $U(P, f)$ the upper Darboux sum.

The geometric idea of Darboux sums is indicated in Figure 5.1. The lower sum is the area of the shaded rectangles, and the upper sum is the area of the entire rectangles, shaded plus unshaded parts. The width of the $i$th rectangle is $\Delta x_i$, the height of the shaded rectangle is $m_i$, and the height of the entire rectangle is $M_i$.

![Figure 5.1: Sample Darboux sums.](image)

**Proposition 5.1.2.** Let $f: [a, b] \to \mathbb{R}$ be a bounded function. Let $m, M \in \mathbb{R}$ be such that for all $x \in [a, b]$, we have $m \leq f(x) \leq M$. Then for every partition $P$ of $[a, b]$,

$$m(b-a) \leq L(P, f) \leq U(P, f) \leq M(b-a). \quad (5.1)$$

**Proof.** Let $P$ be a partition. Note that $m \leq m_i$ for all $i$ and $M_i \leq M$ for all $i$. Also $m_i \leq M_i$ for all $i$. Finally, $\sum_{i=1}^{n} \Delta x_i = (b-a)$. Therefore,

$$m(b-a) = m \left( \sum_{i=1}^{n} \Delta x_i \right) = \sum_{i=1}^{n} m \Delta x_i \leq \sum_{i=1}^{n} m_i \Delta x_i \leq \sum_{i=1}^{n} M_i \Delta x_i \leq \sum_{i=1}^{n} M \Delta x_i = M \left( \sum_{i=1}^{n} \Delta x_i \right) = M(b-a).$$

Hence we get (5.1). In particular, the set of lower and upper sums are bounded sets.

**Definition 5.1.3.** As the sets of lower and upper Darboux sums are bounded, we define

$$ \int_{a}^{b} f(x) \, dx := \sup \{ L(P, f) : P \text{ a partition of } [a, b] \},$$

$$ \int_{a}^{b} f(x) \, dx := \inf \{ U(P, f) : P \text{ a partition of } [a, b] \}.$$

We call $\int$ the lower Darboux integral and $\bar{\int}$ the upper Darboux integral. To avoid worrying about the variable of integration, we often simply write

$$\int_{a}^{b} f := \int_{a}^{b} f(x) \, dx \quad \text{and} \quad \bar{\int}_{a}^{b} f := \bar{\int}_{a}^{b} f(x) \, dx.$$
If integration is to make sense, then the lower and upper Darboux integrals should be the same number, as we want a single number to call the integral. However, these two integrals may differ for some functions.

**Example 5.1.4:** Take the Dirichlet function \( f: [0, 1] \to \mathbb{R} \), where \( f(x) := 1 \) if \( x \in \mathbb{Q} \) and \( f(x) := 0 \) if \( x \notin \mathbb{Q} \). Then
\[
\int_0^1 f = 0 \quad \text{and} \quad \int_0^1 f = 1.
\]
The reason is that for every \( i \), we have \( m_i = \inf \{ f(x) : x \in [x_{i-1}, x_i] \} = 0 \) and \( M_i = \sup \{ f(x) : x \in [x_{j-1}, x_j] \} = 1 \). Thus
\[
L(P, f) = \sum_{i=1}^n 0 \cdot \Delta x_i = 0,
\]
\[
U(P, f) = \sum_{i=1}^n 1 \cdot \Delta x_i = \sum_{i=1}^n \Delta x_i = 1.
\]

**Remark 5.1.5.** The same definition of \( \int_a^b \) and \( \int_a^b \) is used when \( f \) is defined on a larger set \( S \) such that \([a, b] \subset S\). In that case, we use the restriction of \( f \) to \([a, b]\) and we must ensure that the restriction is bounded on \([a, b]\).

To compute the integral, we often take a partition \( P \) and make it finer. That is, we cut intervals in the partition into yet smaller pieces.

**Definition 5.1.6.** Let \( P = \{x_0, x_1, \ldots, x_n\} \) and \( P = \{\tilde{x}_0, \tilde{x}_1, \ldots, \tilde{x}_\ell\} \) be partitions of \([a, b]\). We say \( P \) is a *refinement* of \( P \) if as sets \( P \subset P \).

That is, \( P \) is a refinement of a partition if it contains all the numbers in \( P \) and perhaps some other numbers in between. For example, \( \{0, 0.5, 1, 2\} \) is a partition of \([0, 2]\) and \( \{0, 0.2, 0.5, 1, 1.5, 1.75, 2\} \) is a refinement. The main reason for introducing refinements is the following proposition.

**Proposition 5.1.7.** Let \( f: [a, b] \to \mathbb{R} \) be a bounded function, and let \( P \) be a partition of \([a, b]\). Let \( P \) be a refinement of \( P \). Then
\[
L(P, f) \leq L(P, f) \quad \text{and} \quad U(P, f) \leq U(P, f).
\]

**Proof.** The tricky part of this proof is to get the notation correct. Let \( P = \{x_0, x_1, \ldots, x_n\} \) be a refinement of \( P = \{x_0, x_1, \ldots, x_n\} \). Then \( x_0 = \tilde{x}_0 \) and \( x_n = \tilde{x}_\ell \). In fact, there are integers \( k_0 < k_1 < \cdots < k_n \) such that \( x_j = \tilde{x}_{k_j} \) for \( j = 0, 1, 2, \ldots, n \).

Let \( \Delta \tilde{x}_{p} := \tilde{x}_p - \tilde{x}_{p-1} \). See Figure 5.2. We get
\[
\Delta x_j = x_j - x_{j-1} = \tilde{x}_{k_j} - \tilde{x}_{k_j-1} = \sum_{p=k_j+1}^{k_j} \tilde{x}_p - \tilde{x}_{p-1} = \sum_{p=k_{j-1}+1}^{k_j} \Delta \tilde{x}_p.
\]
Let \( m_j \) be as before and correspond to the partition \( P \). Let \( \tilde{m}_j := \inf \{ f(x) : \tilde{x}_{j-1} \leq x \leq \tilde{x}_j \} \). Now, \( m_j \leq \tilde{m}_p \) for \( k_{j-1} < p \leq k_j \). Therefore,

\[
m_j \Delta x_j = m_j \sum_{p=k_{j-1}+1}^{k_j} \Delta \tilde{x}_p = \sum_{p=k_{j-1}+1}^{k_j} m_j \Delta \tilde{x}_p \leq \sum_{p=k_{j-1}+1}^{k_j} \tilde{m}_p \Delta \tilde{x}_p.
\]

So

\[
L(P, f) = \sum_{j=1}^{n} m_j \Delta x_j \leq \sum_{j=1}^{n} \sum_{p=k_{j-1}+1}^{k_j} \tilde{m}_p \Delta \tilde{x}_p = \sum_{j=1}^{\ell} \tilde{m}_j \Delta \tilde{x}_j = L(\tilde{P}, f).
\]

The proof of \( U(\tilde{P}, f) \leq U(P, f) \) is left as an exercise.

Armed with refinements we prove the following. The key point of this next proposition is that the lower Darboux integral is less than or equal to the upper Darboux integral.

**Proposition 5.1.8.** Let \( f : [a, b] \to \mathbb{R} \) be a bounded function. Let \( m, M \in \mathbb{R} \) be such that for all \( x \in [a, b] \), we have \( m \leq f(x) \leq M \). Then

\[
m(b - a) \leq \int_a^b f \leq \int_a^b f \leq M(b - a).
\]

**Proof.** By Proposition 5.1.2, for every partition \( P \),

\[
m(b - a) \leq L(P, f) \leq U(P, f) \leq M(b - a).
\]

The inequality \( m(b - a) \leq L(P, f) \) implies \( m(b - a) \leq \int_a^b f \). The inequality \( U(P, f) \leq M(b - a) \) implies \( \int_a^b f \leq M(b - a) \).

The middle inequality in (5.2) is the main point of this proposition. Let \( P_1, P_2 \) be partitions of \([a, b]\). Define \( \tilde{P} := P_1 \cup P_2 \). The set \( \tilde{P} \) is a partition of \([a, b]\), which is a refinement of \( P_1 \) and a refinement of \( P_2 \). By Proposition 5.1.7, \( L(P_1, f) \leq L(\tilde{P}, f) \) and \( U(\tilde{P}, f) \leq U(P_2, f) \). So

\[
L(P_1, f) \leq L(\tilde{P}, f) \leq U(\tilde{P}, f) \leq U(P_2, f).
\]

In other words, for two arbitrary partitions \( P_1 \) and \( P_2 \), we have \( L(P_1, f) \leq U(P_2, f) \). Recall Proposition 1.2.7, and take the supremum and infimum over all partitions:

\[
\int_a^b f = \sup \{ L(P, f) : P \text{ a partition of } [a, b] \} \leq \inf \{ U(P, f) : P \text{ a partition of } [a, b] \} = \int_a^b f.
\]

\[\square\]
5.1.2 Riemann integral

We can finally define the Riemann integral. However, the Riemann integral is only defined on a certain class of functions, called the Riemann integrable functions.

**Definition 5.1.9.** Let \( f : [a, b] \to \mathbb{R} \) be a bounded function such that

\[
\int_a^b f(x) \, dx = \int_a^b f(x) \, dx.
\]

Then \( f \) is said to be **Riemann integrable**. The set of Riemann integrable functions on \([a, b]\) is denoted by \( \mathcal{R}[a, b] \). When \( f \in \mathcal{R}[a, b] \), we define

\[
\int_a^b f(x) \, dx := \int_a^b f(x) \, dx = \int_a^b f(x) \, dx.
\]

As before, we often write

\[
\int_a^b f := \int_a^b f(x) \, dx.
\]

The number \( \int_a^b f \) is called the **Riemann integral** of \( f \), or sometimes simply the **integral** of \( f \).

By definition, a Riemann integrable function is bounded. Appealing to **Proposition 5.1.8**, we immediately obtain the following proposition. See also **Figure 5.3**.

**Proposition 5.1.10.** Let \( f : [a, b] \to \mathbb{R} \) be a Riemann integrable function. Let \( m, M \in \mathbb{R} \) be such that \( m \leq f(x) \leq M \) for all \( x \in [a, b] \). Then

\[
m(b - a) \leq \int_a^b f \leq M(b - a).
\]

**Figure 5.3:** The area under the curve is bounded from above by the area of the entire rectangle, \( M(b - a) \), and from below by the area of the shaded part, \( m(b - a) \).

Often we use a weaker form of this proposition. That is, if \( |f(x)| \leq M \) for all \( x \in [a, b] \), then

\[
\left| \int_a^b f \right| \leq M(b - a).
\]
Example 5.1.11: We integrate constant functions using Proposition 5.1.8. If \( f(x) := c \) for some constant \( c \), then we take \( m = M = c \). In inequality (5.2) all the inequalities must be equalities. Thus \( f \) is integrable on \([a,b]\) and \( \int_a^b f = c(b - a) \).

Example 5.1.12: Let \( f : [0,2] \to \mathbb{R} \) be defined by

\[
 f(x) := \begin{cases} 
 1 & \text{if } x < 1, \\
 1/2 & \text{if } x = 1, \\
 0 & \text{if } x > 1.
\end{cases}
\]

We claim \( f \) is Riemann integrable and \( \int_0^2 f = 1 \).

Proof: Let \( 0 < \varepsilon < 1 \) be arbitrary. Let \( P : = \{0, 1 - \varepsilon, 1 + \varepsilon, 2\} \) be a partition. We use the notation from the definition of the Darboux sums. Then

\[
 m_1 = \inf \{ f(x) : x \in [0, 1 - \varepsilon] \} = 1, \quad M_1 = \sup \{ f(x) : x \in [0, 1 - \varepsilon] \} = 1, \\
 m_2 = \inf \{ f(x) : x \in [1 - \varepsilon, 1 + \varepsilon] \} = 0, \quad M_2 = \sup \{ f(x) : x \in [1 - \varepsilon, 1 + \varepsilon] \} = 1, \\
 m_3 = \inf \{ f(x) : x \in [1 + \varepsilon, 2] \} = 0, \quad M_3 = \sup \{ f(x) : x \in [1 + \varepsilon, 2] \} = 0.
\]

Furthermore, \( \Delta x_1 = 1 - \varepsilon, \Delta x_2 = 2\varepsilon \) and \( \Delta x_3 = 1 - \varepsilon \). See Figure 5.4.

![Figure 5.4: Darboux sums for the step function.](image)

We compute

\[
 L(P, f) = \sum_{i=1}^{3} m_i \Delta x_i = 1 \cdot (1 - \varepsilon) + 0 \cdot 2\varepsilon + 0 \cdot (1 - \varepsilon) = 1 - \varepsilon, \\
 U(P, f) = \sum_{i=1}^{3} M_i \Delta x_i = 1 \cdot (1 - \varepsilon) + 1 \cdot 2\varepsilon + 0 \cdot (1 - \varepsilon) = 1 + \varepsilon.
\]

Thus,

\[
 \int_0^2 f - \int_0^2 f \leq U(P, f) - L(P, f) = (1 + \varepsilon) - (1 - \varepsilon) = 2\varepsilon.
\]
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By Proposition 5.1.8, we have \( \int_0^2 f \leq \int_0^2 f \). As \( \epsilon \) was arbitrary, \( \int_0^2 f = \int_0^2 f \). So \( f \) is Riemann integrable. Finally,

\[ 1 - \epsilon = L(P, f) \leq \int_0^2 f \leq U(P, f) = 1 + \epsilon. \]

Hence, \( |\int_0^2 f - 1| \leq \epsilon \). As \( \epsilon \) was arbitrary, we conclude \( \int_0^2 f = 1 \).

It may be worthwhile to extract part of the technique of the example into a proposition.

**Proposition 5.1.13.** Let \( f : [a, b] \to \mathbb{R} \) be a bounded function. Then \( f \) is Riemann integrable if for every \( \epsilon > 0 \), there exists a partition \( P \) of \([a, b]\) such that

\[ U(P, f) - L(P, f) < \epsilon. \]

**Proof.** If for every \( \epsilon > 0 \) such a \( P \) exists, then

\[ 0 \leq \int_a^b f - \int_a^b f \leq U(P, f) - L(P, f) < \epsilon. \]

Therefore, \( \int_a^b f = \int_a^b f \), and \( f \) is integrable. \( \square \)

**Example 5.1.14:** Let us show \( \frac{1}{1+x} \) is integrable on \([0, b]\) for all \( b > 0 \). We will see later that continuous functions are integrable, but let us demonstrate how we do it directly.

Let \( \epsilon > 0 \) be given. Take \( n \in \mathbb{N} \) and pick \( x_j := jb/n \), to form the partition \( P := \{x_0, x_1, \ldots, x_n\} \) of \([0, b]\). We have \( \Delta x_j = b/n \) for all \( j \). As \( f \) is decreasing, for every subinterval \([x_{j-1}, x_j]\), we obtain

\[ m_j = \inf \left\{ \frac{1}{1+x} : x \in [x_{j-1}, x_j] \right\} = \frac{1}{1+x_j}, \quad M_j = \sup \left\{ \frac{1}{1+x} : x \in [x_{j-1}, x_j] \right\} = \frac{1}{1+x_{j-1}}. \]

Then

\[ U(P, f) - L(P, f) = \sum_{j=1}^n \Delta x_j (M_j - m_j) = \frac{b}{n} \sum_{j=1}^n \left( \frac{1}{1+(j-1)b/n} - \frac{1}{1+jb/n} \right) = \frac{b^2}{n(b+1)}. \]

The sum telescopes, the terms successively cancel each other, something we have seen before. Picking \( n \) to be such that \( \frac{b^2}{n(b+1)} < \epsilon \), the proposition is satisfied, and the function is integrable.

**Remark 5.1.15.** A way of thinking of the integral is that it adds up (integrates) lots of local information—it sums \( f(x) \, dx \) over all \( x \). The integral sign was chosen by Leibniz to be the long S to mean summation. Unlike derivatives, which are “local,” integrals show up in applications when one wants a “global” answer: total distance travelled, average temperature, total charge, etc.
CHAPTER 5. THE RIEMANN INTEGRAL

5.1.3 More notation

When \( f : S \to \mathbb{R} \) is defined on a larger set \( S \) and \([a, b] \subset S\), we say \( f \) is Riemann integrable on \([a, b]\) if the restriction of \( f \) to \([a, b]\) is Riemann integrable. In this case, we say \( f \in \mathcal{R}[a, b] \), and we write \( \int_{a}^{b} f \) to mean the Riemann integral of the restriction of \( f \) to \([a, b]\).

It is useful to define the integral \( \int_{a}^{b} f \) even if \( a \not< b \). Suppose \( b < a \) and \( f \in \mathcal{R}[b, a] \), then define

\[
\int_{b}^{a} f := -\int_{a}^{b} f.
\]

For any function \( f \), define

\[
\int_{a}^{a} f := 0.
\]

At times, the variable \( x \) may already have some other meaning. When we need to write down the variable of integration, we may simply use a different letter. For example,

\[
\int_{a}^{b} f(x) \, dx := \int_{a}^{b} f(s) \, ds.
\]

5.1.4 Exercises

Exercise 5.1.1: Define \( f : [0, 1] \to \mathbb{R} \) by \( f(x) := x^3 \) and let \( P := \{0, 0.1, 0.4, 1\} \). Compute \( L(P, f) \) and \( U(P, f) \).

Exercise 5.1.2: Let \( f : [0, 1] \to \mathbb{R} \) be defined by \( f(x) := x \). Show that \( f \in \mathcal{R}[0, 1] \) and compute \( \int_{0}^{1} f \) using the definition of the integral (but feel free to use the propositions of this section).

Exercise 5.1.3: Let \( f : [a, b] \to \mathbb{R} \) be a bounded function. Suppose there exists a sequence of partitions \( \{P_k\} \) of \([a, b]\) such that

\[
\lim_{k \to \infty} (U(P_k, f) - L(P_k, f)) = 0.
\]

Show that \( f \) is Riemann integrable and that

\[
\int_{a}^{b} f = \lim_{k \to \infty} U(P_k, f) = \lim_{k \to \infty} L(P_k, f).
\]

Exercise 5.1.4: Finish the proof of Proposition 5.1.7.

Exercise 5.1.5: Suppose \( f : [-1, 1] \to \mathbb{R} \) is defined as

\[
f(x) := \begin{cases} 
1 & \text{if } x > 0, \\
0 & \text{if } x \leq 0.
\end{cases}
\]

Prove that \( f \in \mathcal{R}[-1, 1] \) and compute \( \int_{-1}^{1} f \) using the definition of the integral (but feel free to use the propositions of this section).

Exercise 5.1.6: Let \( c \in (a, b) \) and let \( d \in \mathbb{R} \). Define \( f : [a, b] \to \mathbb{R} \) as

\[
f(x) := \begin{cases} 
d & \text{if } x = c, \\
0 & \text{if } x \neq c.
\end{cases}
\]

Prove that \( f \in \mathcal{R}[a, b] \) and compute \( \int_{a}^{b} f \) using the definition of the integral (but feel free to use the propositions of this section).
Exercise 5.1.7: Suppose \( f: [a, b] \to \mathbb{R} \) is Riemann integrable. Let \( \varepsilon > 0 \) be given. Then show that there exists a partition \( P = \{x_0, x_1, \ldots, x_n\} \) such that for every set of numbers \( \{c_1, c_2, \ldots, c_n\} \) with \( c_k \in [x_{k-1}, x_k] \) for all \( k \), we have
\[
\left| \int_a^b f - \sum_{k=1}^n f(c_k)\Delta x_k \right| < \varepsilon.
\]

Exercise 5.1.8: Let \( f: [a, b] \to \mathbb{R} \) be a Riemann integrable function. Let \( \alpha > 0 \) and \( \beta \in \mathbb{R} \). Then define \( g(x) := f(\alpha x + \beta) \) on the interval \( I = [\frac{a - \beta}{\alpha}, \frac{b - \beta}{\alpha}] \). Show that \( g \) is Riemann integrable on \( I \).

Exercise 5.1.9: Suppose \( f: [0, 1] \to \mathbb{R} \) and \( g: [0, 1] \to \mathbb{R} \) are such that for all \( x \in (0, 1) \), we have \( f(x) = g(x) \). Suppose \( f \) is Riemann integrable. Prove \( g \) is Riemann integrable and \( \int_0^1 f = \int_0^1 g \).

Exercise 5.1.10: Let \( f: [0, 1] \to \mathbb{R} \) be a bounded function. Let \( P_n = \{x_0, x_1, \ldots, x_n\} \) be a uniform partition of \( [0, 1] \), that is, \( x_i = i/n \). Is \( \{L(P_n, f)\}_{n=1}^\infty \) always monotone? Yes/No: Prove or find a counterexample.

Exercise 5.1.11 (Challenging): For a bounded function \( f: [0, 1] \to \mathbb{R} \), let \( R_n := (1/n)\sum_{j=1}^n f(j/n) \) (the uniform right-hand rule).

a) If \( f \) is Riemann integrable show \( \int_0^1 f = \lim R_n \).

b) Find an \( f \) that is not Riemann integrable, but \( \lim R_n \) exists.

Exercise 5.1.12 (Challenging): Generalize the previous exercise. Show that \( f \in \mathcal{R}[a, b] \) if and only if there exists an \( I \in \mathbb{R} \), such that for all \( \varepsilon > 0 \) there exists a \( \delta > 0 \) such that if \( P \) is a partition with \( \Delta x_i < \delta \) for all \( i \), then \( |L(P, f) - I| < \varepsilon \) and \( |U(P, f) - I| < \varepsilon \). If \( f \in \mathcal{R}[a, b] \), then \( I = \int_a^b f \).

Exercise 5.1.13: Using Exercise 5.1.12 and the idea of the proof in Exercise 5.1.7, show that Darboux integral is the same as the standard definition of Riemann integral, which you have most likely seen in calculus. That is, show that \( f \in \mathcal{R}[a, b] \) if and only if there exists an \( I \in \mathbb{R} \), such that for every \( \varepsilon > 0 \) there exists a \( \delta > 0 \) such that if \( P = \{x_0, x_1, \ldots, x_n\} \) is a partition with \( \Delta x_i < \delta \) for all \( i \), then \( \left| \sum_{i=1}^n f(c_i)\Delta x_i - I \right| < \varepsilon \) for every set \( \{c_1, c_2, \ldots, c_n\} \) with \( c_i \in [x_{i-1}, x_i] \). If \( f \in \mathcal{R}[a, b] \), then \( I = \int_a^b f \).

Exercise 5.1.14 (Challenging): Construct functions \( f \) and \( g \), where \( f: [0, 1] \to \mathbb{R} \) is Riemann integrable, \( g: [0, 1] \to [0, 1] \) is one-to-one and onto, and such that the composition \( f \circ g \) is not Riemann integrable.
5.2 Properties of the integral

Note: 2 lectures, integrability of functions with discontinuities can safely be skipped

5.2.1 Additivity

Adding a bunch of things in two parts and then adding those two parts should be the same as adding everything all at once. The corresponding property for integral is called the additive property of the integral. First, we prove the additivity property for the lower and upper Darboux integrals.

Lemma 5.2.1. Suppose $a < b < c$ and $f : [a, c] \to \mathbb{R}$ is a bounded function. Then

$$\int_a^c f = \int_a^b f + \int_b^c f$$

and

$$\int_a^c f = \int_a^b f + \int_b^c f.$$ 

Proof. If we have partitions $P_1 = \{x_0, x_1, \ldots, x_k\}$ of $[a, b]$ and $P_2 = \{x_k, x_{k+1}, \ldots, x_n\}$ of $[b, c]$, then the set $P := P_1 \cup P_2 = \{x_0, x_1, \ldots, x_n\}$ is a partition of $[a, c]$. We find

$$L(P, f) = \sum_{j=1}^{n} m_j \Delta x_j = \sum_{j=1}^{k} m_j \Delta x_j + \sum_{j=k+1}^{n} m_j \Delta x_j = L(P_1, f) + L(P_2, f).$$

When we take the supremum of the right-hand side over all $P_1$ and $P_2$, we are taking a supremum of the left-hand side over all partitions $P$ of $[a, c]$ that contain $b$. If $Q$ is a partition of $[a, c]$ and $P = Q \cup \{b\}$, then $P$ is a refinement of $Q$ and so $L(Q, f) \leq L(P, f)$. Therefore, taking a supremum only over the $P$ that contain $b$ is sufficient to find the supremum of $L(P, f)$ over all partitions $P$, see Exercise 1.1.9. Finally, recall Exercise 1.2.9 to compute

$$\int_a^c f = \sup \left\{ L(P, f) : P \text{ a partition of } [a, c] \right\}$$

$$= \sup \left\{ L(P, f) : P \text{ a partition of } [a, c], b \in P \right\}$$

$$= \sup \left\{ L(P_1, f) + L(P_2, f) : P_1 \text{ a partition of } [a, b], P_2 \text{ a partition of } [b, c] \right\}$$

$$= \sup \left\{ L(P_1, f) : P_1 \text{ a partition of } [a, b] \right\} + \sup \left\{ L(P_2, f) : P_2 \text{ a partition of } [b, c] \right\}$$

$$= \int_a^b f + \int_b^c f.$$ 

Similarly, for $P, P_1,$ and $P_2$ as above, we obtain

$$U(P, f) = \sum_{j=1}^{n} M_j \Delta x_j = \sum_{j=1}^{k} M_j \Delta x_j + \sum_{j=k+1}^{n} M_j \Delta x_j = U(P_1, f) + U(P_2, f).$$

We wish to take the infimum on the right over all $P_1$ and $P_2$, and so we are taking the infimum over all partitions $P$ of $[a, c]$ that contain $b$. If $Q$ is a partition of $[a, c]$ and $P = Q \cup \{b\}$, then $P$ is
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a refinement of $Q$ and so $U(Q, f) \geq U(P, f)$. Therefore, taking an infimum only over the $P$ that contain $b$ is sufficient to find the infimum of $U(P, f)$ for all $P$. We obtain

$$\int_a^c f = \int_a^b f + \int_b^c f.$$ 

$\square$

**Proposition 5.2.2.** Let $a < b < c$. A function $f : [a, c] \to \mathbb{R}$ is Riemann integrable if and only if $f$ is Riemann integrable on $[a, b]$ and $[b, c]$. If $f$ is Riemann integrable, then

$$\int_a^c f = \int_a^b f + \int_b^c f.$$ 

**Proof.** Suppose $f \in \mathcal{R}[a, c]$, then $\int_a^c f = \int_a^b f + \int_b^c f$. We apply the lemma to get

$$\int_a^c f = \int_a^b f + \int_b^c f \leq \int_a^b f + \int_b^c f = \int_a^c f = \int_a^c f.$$ 

Thus the inequality is an equality,

$$\int_a^b f + \int_b^c f = \int_a^b f + \int_b^c f.$$ 

As we also know $\int_a^b f \leq \int_a^b f$ and $\int_b^c f \leq \int_b^c f$, we conclude

$$\int_a^b f = \int_a^b f \quad \text{and} \quad \int_b^c f = \int_b^c f.$$ 

Thus $f$ is Riemann integrable on $[a, b]$ and $[b, c]$ and the desired formula holds.

Now assume $f$ is Riemann integrable on $[a, b]$ and on $[b, c]$. Again apply the lemma to get

$$\int_a^c f = \int_a^b f + \int_b^c f = \int_a^b f + \int_b^c f = \int_a^c f = \int_a^c f.$$ 

Therefore, $f$ is Riemann integrable on $[a, c]$, and the integral is computed as indicated. $\square$

An easy consequence of the additivity is the following corollary. We leave the details to the reader as an exercise.

**Corollary 5.2.3.** If $f \in \mathcal{R}[a, b]$ and $[c, d] \subset [a, b]$, then the restriction $f|_{[c, d]}$ is in $\mathcal{R}[c, d]$.

### 5.2.2 Linearity and monotonicity

A sum is a linear function of the summands. So is the integral.

**Proposition 5.2.4 (Linearity).** Let $f$ and $g$ be in $\mathcal{R}[a, b]$ and $\alpha \in \mathbb{R}$.

(i) $\alpha f$ is in $\mathcal{R}[a, b]$ and

$$\int_a^b \alpha f(x) \, dx = \alpha \int_a^b f(x) \, dx.$$ 

(ii) $f + g$ is in $\mathcal{R}[a, b]$ and

$$\int_a^b (f(x) + g(x)) \, dx = \int_a^b f(x) \, dx + \int_a^b g(x) \, dx.$$
Proof. Let us prove the first item for $\alpha \geq 0$. Let $P$ be a partition of $[a, b]$. Let $m_i := \inf \{f(x) : x \in [x_{i-1}, x_i] \}$ as usual. Since $\alpha$ is nonnegative, we can move the multiplication by $\alpha$ past the infimum, 

$$\inf \{ \alpha f(x) : x \in [x_{i-1}, x_i] \} = \alpha \inf \{ f(x) : x \in [x_{i-1}, x_i] \} = \alpha m_i.$$ 

Therefore,

$$L(P, \alpha f) = \sum_{i=1}^{n} \alpha m_i \Delta x_i = \alpha \sum_{i=1}^{n} m_i \Delta x_i = \alpha L(P, f).$$

Similarly,

$$U(P, \alpha f) = \alpha U(P, f).$$

Again, as $\alpha \geq 0$ we may move multiplication by $\alpha$ past the supremum. Hence,

$$\int_{a}^{b} \alpha f(x) \, dx = \sup \{ L(P, \alpha f) : P \text{ a partition of } [a, b] \}$$

$$= \sup \{ \alpha L(P, f) : P \text{ a partition of } [a, b] \}$$

$$= \alpha \sup \{ L(P, f) : P \text{ a partition of } [a, b] \}$$

$$= \alpha \int_{a}^{b} f(x) \, dx.$$

Similarly, we show

$$\int_{a}^{b} \alpha f(x) \, dx = \alpha \int_{a}^{b} f(x) \, dx.$$

The conclusion now follows for $\alpha \geq 0$.

To finish the proof of the first item, we need to show that $-f$ is Riemann integrable and $\int_{a}^{b} -f(x) \, dx = -\int_{a}^{b} f(x) \, dx$. The proof of this fact is left as Exercise 5.2.1.

The proof of the second item is left as Exercise 5.2.2. It is not difficult, but it is not as trivial as it may appear at first glance.

The second item in the proposition does not hold with equality for the Darboux integrals, but we do obtain inequalities. The proof of the following proposition is Exercise 5.2.16. It follows for upper and lower sums on a fixed partition by Exercise 1.3.7, that is, supremum of a sum is less than or equal to the sum of suprema and similarly for infima.

**Proposition 5.2.5.** Let $f : [a, b] \to \mathbb{R}$ and $g : [a, b] \to \mathbb{R}$ be bounded functions. Then

$$\int_{a}^{b} (f + g) \leq \int_{a}^{b} f + \int_{a}^{b} g,$$

and

$$\int_{a}^{b} (f + g) \geq \int_{a}^{b} f + \int_{a}^{b} g.$$

Adding up smaller numbers should give us a smaller result. That is true for an integral as well.

**Proposition 5.2.6 (Monotonicity).** Let $f : [a, b] \to \mathbb{R}$ and $g : [a, b] \to \mathbb{R}$ be bounded, and $f(x) \leq g(x)$ for all $x \in [a, b]$. Then

$$\int_{a}^{b} f \leq \int_{a}^{b} g$$

and

$$\int_{a}^{b} f \leq \int_{a}^{b} g.$$ 

Moreover, if $f$ and $g$ are in $\mathcal{R}[a, b]$, then

$$\int_{a}^{b} f \leq \int_{a}^{b} g.$$
Proof. Let $P = \{x_0, x_1, \ldots, x_n\}$ be a partition of $[a, b]$. Then let
\[
m_i := \inf \{f(x) : x \in [x_{i-1}, x_i]\} \quad \text{and} \quad \tilde{m}_i := \inf \{g(x) : x \in [x_{i-1}, x_i]\}.
\]
As $f(x) \leq g(x)$, then $m_i \leq \tilde{m}_i$. Therefore,
\[
L(P, f) = \sum_{i=1}^{n} m_i \Delta x_i \leq \sum_{i=1}^{n} \tilde{m}_i \Delta x_i = L(P, g).
\]
We take the supremum over all $P$ (see Proposition 1.3.7) to obtain
\[
\int_{a}^{b} f \leq \int_{a}^{b} g.
\]
Similarly, we obtain the same conclusion for the upper integrals. Finally, if $f$ and $g$ are Riemann integrable all the integrals are equal, and the conclusion follows. \qed

5.2.3 Continuous functions

Let us show that continuous functions are Riemann integrable. In fact, we can even allow some discontinuities. We start with a function continuous on the whole closed interval $[a, b]$.

Lemma 5.2.7. If $f : [a, b] \to \mathbb{R}$ is a continuous function, then $f \in \mathcal{R}[a, b]$.

Proof. As $f$ is continuous on a closed bounded interval, it is uniformly continuous. Let $\varepsilon > 0$ be given. Find a $\delta > 0$ such that $|x - y| < \delta$ implies $|f(x) - f(y)| < \frac{\varepsilon}{b - a}$.

Let $P = \{x_0, x_1, \ldots, x_n\}$ be a partition of $[a, b]$ such that $\Delta x_i < \delta$ for all $i = 1, 2, \ldots, n$. For example, take $n$ such that $\frac{b - a}{n} < \delta$, and let $x_i := \frac{i}{n}(b - a) + a$. Then for all $x, y \in [x_{i-1}, x_i]$, we have $|x - y| \leq \Delta x_i < \delta$, and so
\[
f(x) - f(y) \leq |f(x) - f(y)| < \frac{\varepsilon}{b - a}.
\]
As $f$ is continuous on $[x_{i-1}, x_i]$, it attains a maximum and a minimum on this interval. Let $x$ be a point where $f$ attains the maximum and $y$ be a point where $f$ attains the minimum. Then $f(x) = M_i$ and $f(y) = m_i$ in the notation from the definition of the integral. Therefore,
\[
M_i - m_i = f(x) - f(y) < \frac{\varepsilon}{b - a}.
\]
And so
\[
\int_{a}^{b} f \leq \int_{a}^{b} f - \frac{\varepsilon}{b - a} \sum_{i=1}^{n} \Delta x_i
\]
\[
= \left( \sum_{i=1}^{n} M_i \Delta x_i \right) - \left( \sum_{i=1}^{n} m_i \Delta x_i \right)
\]
\[
= \sum_{i=1}^{n} (M_i - m_i) \Delta x_i
\]
\[
< \frac{\varepsilon}{b - a} \sum_{i=1}^{n} \Delta x_i
\]
\[
= \frac{\varepsilon}{b - a} (b - a) = \varepsilon.
\]
As $\varepsilon > 0$ was arbitrary,
\[ \int_a^b f = \int_a^b f, \]
and $f$ is Riemann integrable on $[a, b]$.

The second lemma says that we need the function to only be “Riemann integrable inside the interval,” as long as it is bounded. It also tells us how to compute the integral.

**Lemma 5.2.8.** Let $f : [a, b] \to \mathbb{R}$ be a bounded function, $\{a_n\}$ and $\{b_n\}$ be sequences such that $a < a_n < b_n < b$ for all $n$, with $\lim a_n = a$ and $\lim b_n = b$. Suppose $f \in \mathcal{R}[a_n, b_n]$ for all $n$. Then $f \in \mathcal{R}[a, b]$ and
\[ \int_a^b f = \lim_{n \to \infty} \int_{a_n}^{b_n} f. \]

**Proof.** Let $M > 0$ be a real number such that $|f(x)| \leq M$. As $(b - a) \geq (b_n - a_n)$,
\[ -M(b - a) \leq -M(b_n - a_n) \leq \int_{a_n}^{b_n} f \leq M(b_n - a_n) \leq M(b - a). \]

Therefore, the sequence of numbers $\{\int_{a_n}^{b_n} f\}_{n=1}^{\infty}$ is bounded and by Bolzano–Weierstrass has a convergent subsequence indexed by $n_k$. Let us call $L$ the limit of the subsequence $\{\int_{a_n}^{b_n} f\}_{k=1}^{\infty}$.

**Lemma 5.2.1** says that the lower and upper integral are additive and the hypothesis says that $f$ is integrable on $[a_{n_k}, b_{n_k}]$. Therefore
\[ \int_a^b f = \int_a^{a_{n_k}} f + \int_{a_{n_k}}^{b_{n_k}} f + \int_{b_{n_k}}^b f \geq -M(a_{n_k} - a) + \int_{a_{n_k}}^{b_{n_k}} f - M(b - b_{n_k}). \]

We take the limit as $k$ goes to $\infty$ on the right-hand side,
\[ \int_a^b f \geq -M \cdot 0 + L - M \cdot 0 = L. \]

Next we use additivity of the upper integral,
\[ \int_a^b f = \int_a^{a_{n_k}} f + \int_{a_{n_k}}^{b_{n_k}} f + \int_{b_{n_k}}^b f \leq M(a_{n_k} - a) + \int_{a_{n_k}}^{b_{n_k}} f + M(b - b_{n_k}). \]

We take the same subsequence $\{\int_{a_n}^{b_n} f\}_{k=1}^{\infty}$ and take the limit to obtain
\[ \int_a^b f \leq M \cdot 0 + L + M \cdot 0 = L. \]

Thus $\int_a^b f = \int_a^b f = L$ and hence $f$ is Riemann integrable and $\int_a^b f = L$. In particular, no matter what subsequence we chose, the $L$ is the same number.

To prove the final statement of the lemma we use Proposition 2.3.7. We have shown that every convergent subsequence $\{\int_{a_{n_k}}^{b_{n_k}} f\}$ converges to $L = \int_a^b f$. Therefore, the sequence $\{\int_{a_n}^{b_n} f\}$ is convergent and converges to $\int_a^b f$. \qed
We say a function \( f : [a, b] \rightarrow \mathbb{R} \) has \textit{finitely many discontinuities} if there exists a finite set \( S = \{x_1, x_2, \ldots, x_n\} \subset [a, b] \), and \( f \) is continuous at all points of \( [a, b] \setminus S \).

**Theorem 5.2.9.** Let \( f : [a, b] \rightarrow \mathbb{R} \) be a bounded function with finitely many discontinuities. Then \( f \in \mathcal{R}[a, b] \).

**Proof.** We divide the interval into finitely many intervals \([a_i, b_i]\) so that \( f \) is continuous on the interior \((a_i, b_i)\). If \( f \) is continuous on \((a_i, b_i)\), then it is continuous and hence integrable on \([c_i, d_i]\) whenever \( a_i < c_i < d_i < b_i \). By Lemma 5.2.8 the restriction of \( f \) to \([a_i, b_i]\) is integrable. By additivity of the integral (and induction) \( f \) is integrable on the union of the intervals. \(\square\)

### 5.2.4 More on integrable functions

Sometimes it is convenient (or necessary) to change certain values of a function and then integrate. The next result says that if we change the values at finitely many points, the integral does not change.

**Proposition 5.2.10.** Let \( f : [a, b] \rightarrow \mathbb{R} \) be Riemann integrable. Let \( g : [a, b] \rightarrow \mathbb{R} \) be such that \( f(x) = g(x) \) for all \( x \in [a, b] \setminus S \), where \( S \) is a finite set. Then \( g \) is a Riemann integrable function and 
\[
\int_a^b g = \int_a^b f.
\]

**Sketch of proof.** Using additivity of the integral, we split up the interval \([a, b]\) into smaller intervals such that \( f(x) = g(x) \) holds for all \( x \) except at the endpoints (details are left to the reader).

Therefore, without loss of generality suppose \( f(x) = g(x) \) for all \( x \in (a, b) \). The proof follows by Lemma 5.2.8, and is left as Exercise 5.2.3. \(\square\)

Finally, monotone (increasing or decreasing) functions are always Riemann integrable. The proof is left to the reader as part of Exercise 5.2.14.

**Proposition 5.2.11.** Let \( f : [a, b] \rightarrow \mathbb{R} \) be a monotone function. Then \( f \in \mathcal{R}[a, b] \).

### 5.2.5 Exercises

**Exercise 5.2.1:** Finish the proof of the first part of Proposition 5.2.4. Let \( f \) be in \( \mathcal{R}[a, b] \). Prove that \( -f \) is in \( \mathcal{R}[a, b] \) and 
\[
\int_a^b -f(x) \, dx = -\int_a^b f(x) \, dx.
\]

**Exercise 5.2.2:** Prove the second part of Proposition 5.2.4. Let \( f \) and \( g \) be in \( \mathcal{R}[a, b] \). Prove, without using Proposition 5.2.5, that \( f + g \) is in \( \mathcal{R}[a, b] \) and 
\[
\int_a^b (f(x) + g(x)) \, dx = \int_a^b f(x) \, dx + \int_a^b g(x) \, dx.
\]

**Hint:** One way to do it is to use Proposition 5.1.7 to find a single partition \( P \) such that \( U(P, f) - L(P, f) < \epsilon/2 \) and \( U(P, g) - L(P, g) < \epsilon/2 \).
Exercise 5.2.3: Let $f: [a, b] \to \mathbb{R}$ be Riemann integrable, and $g: [a, b] \to \mathbb{R}$ be such that $f(x) = g(x)$ for all $x \in (a, b)$. Prove that $g$ is Riemann integrable and that

$$\int_a^b g = \int_a^b f.$$ 

Exercise 5.2.4: Prove the mean value theorem for integrals: If $f: [a, b] \to \mathbb{R}$ is continuous, then there exists a $c \in [a, b]$ such that $\int_a^b f = f(c)(b - a)$.

Exercise 5.2.5: Let $f: [a, b] \to \mathbb{R}$ be a continuous function such that $f(x) \geq 0$ for all $x \in [a, b]$ and $\int_a^b f = 0$. Prove that $f(x) = 0$ for all $x$.

Exercise 5.2.6: Let $f: [a, b] \to \mathbb{R}$ be a continuous function and $\int_a^b f = 0$. Prove that there exists a $c \in [a, b]$ such that $f(c) = 0$. (Compare with the previous exercise.)

Exercise 5.2.7: Let $f: [a, b] \to \mathbb{R}$ and $g: [a, b] \to \mathbb{R}$ be continuous functions such that $\int_a^b f = \int_a^b g$. Show that there exists a $c \in [a, b]$ such that $f(c) = g(c)$.

Exercise 5.2.8: Let $f \in \mathcal{R}[a, b]$. Let $\alpha, \beta, \gamma$ be arbitrary numbers in $[a, b]$ (not necessarily ordered in any way). Prove

$$\int_\alpha^\gamma f = \int_\alpha^\beta f + \int_\beta^\gamma f.$$ 

Recall what $\int_a^b f$ means if $b \leq a$.

Exercise 5.2.9: Prove Corollary 5.2.3.

Exercise 5.2.10: Suppose $f: [a, b] \to \mathbb{R}$ is bounded and has finitely many discontinuities. Show that as a function of $x$ the expression $|f(x)|$ is bounded with finitely many discontinuities and is thus Riemann integrable. Then show

$$\left| \int_a^b f(x) \, dx \right| \leq \int_a^b |f(x)| \, dx.$$ 

Exercise 5.2.11 (Hard): Show that the Thomae or popcorn function (see Example 3.2.12) is Riemann integrable. Therefore, there exists a function discontinuous at all rational numbers (a dense set) that is Riemann integrable.

In particular, define $f: [0, 1] \to \mathbb{R}$ by

$$f(x) := \begin{cases} \frac{1}{k} & \text{if } x = \frac{m}{k} \text{ where } m, k \in \mathbb{N} \text{ and } m \text{ and } k \text{ have no common divisors}, \\ 0 & \text{if } x \text{ is irrational}. \end{cases}$$

Show $\int_0^1 f = 0$.

If $I \subset \mathbb{R}$ is a bounded interval, then the function

$$\varphi_I(x) := \begin{cases} 1 & \text{if } x \in I, \\ 0 & \text{otherwise}, \end{cases}$$

is called an elementary step function.

Exercise 5.2.12: Let $I$ be an arbitrary bounded interval (you should consider all types of intervals: closed, open, half-open) and $a < b$, then using only the definition of the integral show that the elementary step function $\varphi_I$ is integrable on $[a, b]$, and find the integral in terms of $a, b$, and the endpoints of $I$. 
A function $f$ is called a step function if it can be written as

$$f(x) = \sum_{k=1}^{n} \alpha_k \varphi_{I_k}(x)$$

for some real numbers $\alpha_1, \alpha_2, \ldots, \alpha_n$ and some bounded intervals $I_1, I_2, \ldots, I_n$.

**Exercise 5.2.13:** Using Exercise 5.2.12, show that a step function is integrable on every interval $[a, b]$. Furthermore, find the integral in terms of $a$, $b$, the endpoints of $I_k$ and the $\alpha_k$.

**Exercise 5.2.14:** Let $f : [a, b] \to \mathbb{R}$ be a function.

a) Show that if $f$ is increasing, then it is Riemann integrable. Hint: Use a uniform partition; each subinterval of same length.

b) Use part a) to show that if $f$ is decreasing, then it is Riemann integrable.

c) Suppose $h = f - g$ where $f$ and $g$ are increasing functions on $[a, b]$. Show that $h$ is Riemann integrable.

**Exercise 5.2.15** (Challenging): Suppose $f \in \mathcal{R}[a, b]$, then the function that takes $x$ to $|f(x)|$ is also Riemann integrable on $[a, b]$. Then show the same inequality as Exercise 5.2.10.

**Exercise 5.2.16:** Suppose $f : [a, b] \to \mathbb{R}$ and $g : [a, b] \to \mathbb{R}$ are bounded.

a) Show $\int_a^b (f + g) \leq \int_a^b f + \int_a^b g$ and $\int_a^b (f + g) \geq \int_a^b f + \int_a^b g$.

b) Find example $f$ and $g$ where the inequality is strict. Hint: $f$ and $g$ should not be Riemann integrable.

**Exercise 5.2.17:** Suppose $f : [a, b] \to \mathbb{R}$ is continuous and $g : \mathbb{R} \to \mathbb{R}$ is Lipschitz continuous. Define

$$h(x) := \int_a^b g(t - x) f(t) \, dt.$$

Prove that $h$ is Lipschitz continuous.

---

*Such an $h$ is said to be of bounded variation.*
5.3 Fundamental theorem of calculus

Note: 1.5 lectures

In this chapter we discuss and prove the fundamental theorem of calculus. The entirety of integral calculus is built upon this theorem, ergo the name. The theorem relates the seemingly unrelated concepts of integral and derivative. It tells us how to compute the antiderivative of a function using the integral and vice versa.

5.3.1 First form of the theorem

Theorem 5.3.1. Let \( F: [a, b] \rightarrow \mathbb{R} \) be a continuous function, differentiable on \((a, b)\). Let \( f \in \mathcal{R}[a, b] \) be such that \( f(x) = F'(x) \) for \( x \in (a, b) \). Then

\[
\int_{a}^{b} f = F(b) - F(a).
\]

It is not hard to generalize the theorem to allow a finite number of points in \([a, b]\) where \( F \) is not differentiable, as long as it is continuous. This generalization is left as an exercise.

Proof. Let \( P = \{x_0, x_1, \ldots, x_n\} \) be a partition of \([a, b]\). For each interval \([x_{i-1}, x_i]\), use the mean value theorem to find a \( c_i \in (x_{i-1}, x_i) \) such that

\[
f(c_i)\Delta x_i = F'(c_i)(x_i - x_{i-1}) = F(x_i) - F(x_{i-1}).
\]

See Figure 5.5, and notice that the area of all three shaded rectangles is \( F(x_{i+1}) - F(x_{i-2}) \). The idea is that by taking smaller and smaller subintervals we prove that this area is the integral of \( f \).

![Figure 5.5: Mean value theorem on subintervals of a partition approximating area under the curve.](image)

Using the notation from the definition of the integral, we have \( m_i \leq f(c_i) \leq M_i \), and so

\[
m_i\Delta x_i \leq F(x_i) - F(x_{i-1}) \leq M_i\Delta x_i.
\]
We sum over \( i = 1, 2, \ldots, n \) to get
\[
\sum_{i=1}^{n} m_i \Delta x_i \leq \sum_{i=1}^{n} (F(x_i) - F(x_{i-1})) \leq \sum_{i=1}^{n} M_i \Delta x_i.
\]
In the middle sum, all the terms except the first and last cancel and we end up with
\[
F(x_n) - F(x_0) = \int_a^b f(x) \, dx.
\]
The sums on the left and on the right are the lower and the upper sum respectively. So
\[
L(P, f) \leq F(b) - F(a) \leq U(P, f).
\]
We take the supremum of \( L(P, f) \) over all partitions \( P \) and the left inequality yields
\[
\int_a^b f \leq F(b) - F(a).
\]
Similarly, taking the infimum of \( U(P, f) \) over all partitions \( P \) yields
\[
F(b) - F(a) \leq \int_a^b f.
\]
As \( f \) is Riemann integrable, we have
\[
\int_a^b f = \int_a^b f \leq F(b) - F(a) \leq \int_a^b f = \int_a^b f.
\]
The inequalities must be equalities and we are done.

The theorem is used to compute integrals. Suppose we know that the function \( f(x) \) is a derivative of some other function \( F(x) \), then we can find an explicit expression for \( \int_a^b f \).

**Example 5.3.2:** To compute
\[
\int_0^1 x^2 \, dx,
\]
we notice \( x^2 \) is the derivative of \( \frac{x^3}{3} \). The fundamental theorem says
\[
\int_0^1 x^2 \, dx = \frac{1^3}{3} - \frac{0^3}{3} = \frac{1}{3}.
\]

**5.3.2 Second form of the theorem**

The second form of the fundamental theorem gives us a way to solve the differential equation \( F'(x) = f(x) \), where \( f \) is a known function and we are trying to find an \( F \) that satisfies the equation.

**Theorem 5.3.3.** Let \( f : [a, b] \to \mathbb{R} \) be a Riemann integrable function. Define
\[
F(x) := \int_a^x f.
\]
First, \( F \) is continuous on \( [a, b] \). Second, if \( f \) is continuous at \( c \in [a, b] \), then \( F \) is differentiable at \( c \) and \( F'(c) = f(c) \).
When $c$ with $x > y$. Then

$$|F(x) - F(y)| = \left| \int_a^x f - \int_a^y f \right| = \left| \int_y^x f \right| \leq M |x - y|.$$  

By symmetry, the same also holds if $x < y$. So $F$ is Lipschitz continuous and hence continuous.

Now suppose $f$ is continuous at $c$. Let $\varepsilon > 0$ be given. Let $\delta > 0$ be such that for $x \in [a, b], |x - c| < \delta$ implies $|f(x) - f(c)| < \varepsilon$. In particular, for such $x$, we have

$$f(c) - \varepsilon < f(x) < f(c) + \varepsilon.$$  

Thus if $x > c$, then

$$(f(c) - \varepsilon)(x - c) \leq \int_c^x f \leq (f(c) + \varepsilon)(x - c).$$  

When $c > x$, then the inequalities are reversed. Therefore, assuming $c \neq x$, we get

$$f(c) - \varepsilon \leq \frac{\int_c^x f}{x - c} \leq f(c) + \varepsilon.$$  

As

$$\frac{F(x) - F(c)}{x - c} = \frac{\int_a^x f - \int_a^c f}{x - c} = \frac{\int_c^x f}{x - c},$$  

we have

$$\left| \frac{F(x) - F(c)}{x - c} - f(c) \right| \leq \varepsilon.$$  

The result follows. It is left to the reader to see why is it OK that we just have a non-strict inequality.

Of course, if $f$ is continuous on $[a, b]$, then it is automatically Riemann integrable, $F$ is differentiable on all of $[a, b]$ and $F'(x) = f(x)$ for all $x \in [a, b]$.

Remark 5.3.4. The second form of the fundamental theorem of calculus still holds if we let $d \in [a, b]$ and define

$$F(x) := \int_d^x f.$$  

That is, we can use any point of $[a, b]$ as our base point. The proof is left as an exercise.

Let us look at what a simple discontinuity can do. Take $f(x) := -1$ if $x < 0$, and $f(x) := 1$ if $x \geq 0$. Let $F(x) := \int_0^x f$. It is not difficult to see that $F(x) = |x|$. Notice that $f$ is discontinuous at 0 and $F$ is not differentiable at 0. However, the converse in the theorem does not hold. Let $g(x) := 0$ if $x \neq 0$, and $g(0) := 1$. Letting $G(x) := \int_0^x g$, we find that $G(x) = 0$ for all $x$. So $g$ is discontinuous at 0, but $G'(0)$ exists and is equal to 0.

A common misunderstanding of the integral for calculus students is to think of integrals whose solution cannot be given in closed-form as somehow deficient. This is not the case. Most integrals we write down are not computable in closed-form. Even some integrals that we consider in closed-form are not really such. We define the natural logarithm as the antiderivative of $1/x$ such that $\ln 1 = 0$:

$$\ln x := \int_1^x \frac{1}{s} ds.$$
How does a computer find the value of ln(x)? One way to do it is to numerically approximate this integral. Morally, we did not really “simplify” $\int_1^x \frac{1}{s} ds$ by writing down ln(x). We simply gave the integral a name. If we require numerical answers, it is possible we end up doing the calculation by approximating an integral anyway. In the next section, we even define the exponential using the logarithm, which we define in terms of the integral.

Another common function defined by an integral that cannot be evaluated symbolically in terms of elementary functions is the erf function, defined as

$$\text{erf}(x) := \frac{2}{\sqrt{\pi}} \int_0^x e^{-s^2} ds.$$  

This function comes up often in applied mathematics. It is simply the antiderivative of $(2/\sqrt{\pi}) e^{-x^2}$ that is zero at zero. The second form of the fundamental theorem tells us that we can write the function as an integral. If we wish to compute any particular value, we numerically approximate the integral.

### 5.3.3 Change of variables

A theorem often used in calculus to solve integrals is the change of variables theorem, you may have called it \textit{u-substitution}. Let us prove it now. Recall a function is continuously differentiable if it is differentiable and the derivative is continuous.

**Theorem 5.3.5** (Change of variables). \textit{Let $g : [a, b] \to \mathbb{R}$ be a continuously differentiable function, let $f : [c, d] \to \mathbb{R}$ be continuous, and suppose $g([a, b]) \subset [c, d]$. Then}

$$\int_a^b f(g(x)) g'(x) \, dx = \int_{g(a)}^{g(b)} f(s) \, ds.$$  

**Proof.** As $g$, $g'$, and $f$ are continuous, $f(g(x)) g'(x)$ is a continuous function of $[a, b]$, therefore it is Riemann integrable. Similarly, $f$ is integrable on every subinterval of $[c, d]$.

Define $F : [c, d] \to \mathbb{R}$ by

$$F(y) := \int_{g(a)}^{y} f(s) \, ds.$$  

By the second form of the fundamental theorem of calculus (see Remark 5.3.4 and Exercise 5.3.4), $F$ is a differentiable function and $F'(y) = f(y)$. Apply the chain rule,

$$(F \circ g)'(x) = F'(g(x)) g'(x) = f(g(x)) g'(x).$$  

Note that $F(g(a)) = 0$ and use the first form of the fundamental theorem to obtain

$$\int_{g(a)}^{g(b)} f(s) \, ds = F(g(b)) - F(g(a))$$

$$= \int_a^b (F \circ g)'(x) \, dx = \int_a^b f(g(x)) g'(x) \, dx.$$  

The change of variables theorem is often used to solve integrals by changing them to integrals that we know or that we can solve using the fundamental theorem of calculus.
Example 5.3.6: The derivative of $\sin(x)$ is $\cos(x)$. Using $g(x) := x^2$, we solve

$$\int_0^\sqrt{\pi} x \cos(x^2) \, dx = \int_0^\pi \frac{\cos(s)}{2} \, ds = \frac{1}{2} \int_0^\pi \cos(s) \, ds = \frac{\sin(\pi) - \sin(0)}{2} = 0.$$  

However, beware that we must satisfy the hypotheses of the theorem. The following example demonstrates why we should not just move symbols around mindlessly. We must be careful that those symbols really make sense.

Example 5.3.7: Consider

$$\int_{-1}^1 \frac{\ln|x|}{x} \, dx.$$  

It may be tempting to take $g(x) := \ln|x|$. Compute $g'(x) = 1/x$ and try to write

$$\int_{g(-1)}^{g(1)} s \, ds = \int_0^0 s \, ds = 0.$$  

This “solution” is incorrect, and it does not say that we can solve the given integral. First problem is that $\frac{\ln|x|}{x}$ is not continuous on $[-1, 1]$. It is not defined at 0, and cannot be made continuous by defining a value at 0. Second, $\frac{\ln|x|}{x}$ is not even Riemann integrable on $[-1, 1]$ (it is unbounded). The integral we wrote down simply does not make sense. Finally, $g$ is not continuous on $[-1, 1]$, let alone continuously differentiable.

5.3.4 Exercises

Exercise 5.3.1: Compute $\frac{d}{dx} \left( \int_{-x}^{x} e^s \, ds \right)$.

Exercise 5.3.2: Compute $\frac{d}{dx} \left( \int_0^{x^2} \sin(s^2) \, ds \right)$.

Exercise 5.3.3: Suppose $F : [a, b] \to \mathbb{R}$ is continuous and differentiable on $[a, b] \setminus S$, where $S$ is a finite set. Suppose there exists an $f \in \mathcal{R}[a, b]$ such that $f(x) = F'(x)$ for $x \in [a, b] \setminus S$. Show that $\int_a^b f = F(b) - F(a)$.

Exercise 5.3.4: Let $f : [a, b] \to \mathbb{R}$ be a continuous function. Let $c \in [a, b]$ be arbitrary. Define

$$F(x) := \int_c^x f.$$  

Prove that $F$ is differentiable and that $F'(x) = f(x)$ for all $x \in [a, b]$.

Exercise 5.3.5: Prove integration by parts. That is, suppose $F$ and $G$ are continuously differentiable functions on $[a, b]$. Then prove

$$\int_a^b F(x)G'(x) \, dx = F(b)G(b) - F(a)G(a) - \int_a^b F'(x)G(x) \, dx.$$  

Exercise 5.3.6: Suppose $F$ and $G$ are continuously differentiable functions defined on $[a, b]$ such that $F'(x) = G'(x)$ for all $x \in [a, b]$. Using the fundamental theorem of calculus, show that $F$ and $G$ differ by a constant. That is, show that there exists a $C \in \mathbb{R}$ such that $F(x) - G(x) = C$.

*Compare this hypothesis to Exercise 4.2.8.*
The next exercise shows how we can use the integral to “smooth out” a non-differentiable function.

**Exercise 5.3.7:** Let \( f : [a, b] \to \mathbb{R} \) be a continuous function. Let \( \varepsilon > 0 \) be a constant. For \( x \in [a + \varepsilon, b - \varepsilon] \), define
\[
g(x) := \frac{1}{2\varepsilon} \int_{x-\varepsilon}^{x+\varepsilon} f.
\]

a) Show that \( g \) is differentiable and find the derivative.
b) Let \( f \) be differentiable and fix \( x \in (a, b) \) (let \( \varepsilon \) be small enough). What happens to \( g'(x) \) as \( \varepsilon \) gets smaller?
c) Find \( g \) for \( f(x) := |x|, \varepsilon = 1 \) (you can assume \([a, b]\) is large enough).

**Exercise 5.3.8:** Suppose \( f : [a, b] \to \mathbb{R} \) is continuous and \( \int_a^x f = \int_x^b f \) for all \( x \in [a, b] \). Show that \( f(x) = 0 \) for all \( x \in [a, b] \).

**Exercise 5.3.9:** Suppose \( f : [a, b] \to \mathbb{R} \) is continuous and \( \int_a^x f = 0 \) for all rational \( x \) in \([a, b]\). Show that \( f(x) = 0 \) for all \( x \in [a, b] \).

**Exercise 5.3.10:** A function \( f \) is an odd function if \( f(-x) = -f(x) \), and \( f \) is an even function if \( f(x) = f(-x) \). Let \( a > 0 \). Assume \( f \) is continuous. Prove:
a) If \( f \) is odd, then \( \int_{-a}^a f = 0 \).
b) If \( f \) is even, then \( \int_{-a}^a f = 2 \int_0^a f \).

**Exercise 5.3.11:**
a) Show that \( f(x) := \sin(\frac{1}{x}) \) is integrable on every interval (you can define \( f(0) \) to be anything).
b) Compute \( \int_1^{-1} \sin(\frac{1}{x}) \, dx \) (mind the discontinuity).

**Exercise 5.3.12** (uses §3.6):
a) Suppose \( f : [a, b] \to \mathbb{R} \) is increasing, by Proposition 5.2.11, \( f \) is Riemann integrable. Suppose \( f \) has a discontinuity at \( c \in (a, b) \), show that \( F(x) := \int_a^x f \) is not differentiable at \( c \).
b) In Exercise 3.6.11, you constructed an increasing function \( f : [0, 1] \to \mathbb{R} \) that is discontinuous at every \( x \in [0, 1] \cap \mathbb{Q} \). Use this \( f \) to construct a function \( F(x) \) that is continuous on \([0, 1]\), but not differentiable at every \( x \in [0, 1] \cap \mathbb{Q} \).
5.4 The logarithm and the exponential

Note: 1 lecture (optional, requires the optional sections §3.5, §3.6, §4.4)

We now have the tools required to properly define the exponential and the logarithm that you
know from calculus so well. We start with exponentiation. If \( n \) is a positive integer, it is obvious to
define \( x^n := x \cdot x \cdots x \), \( n \) times.

It makes sense to define \( x^0 := 1 \). For negative integers, let \( x^{-n} := 1/x^n \). If \( x > 0 \), define \( x^{1/n} \) as the
unique positive \( n \)th root. Finally, for a rational number \( n/m \) (in lowest terms), define
\[ x^{n/m} := (x^{1/m})^n. \]

It is not difficult to show we get the same number no matter what representation of \( n/m \) we use, so
we do not need to use lowest terms.

However, what do we mean by \( \sqrt{2} \)? Or \( x^y \) in general? In particular, what is \( e^x \) for all \( x \)? And
how do we solve \( y = e^x \) for \( x \)? This section answers these questions and more.

5.4.1 The logarithm

It is convenient to define the logarithm first. Let us show that a unique function with the right
properties exists, and only then will we call it the logarithm.

**Proposition 5.4.1.** There exists a unique function \( L : (0, \infty) \to \mathbb{R} \) such that

(i) \( L(1) = 0 \).

(ii) \( L \) is differentiable and \( L'(x) = 1/x \).

(iii) \( L \) is strictly increasing, bijective, and
\[ \lim_{x \to 0} L(x) = -\infty, \quad \text{and} \quad \lim_{x \to \infty} L(x) = \infty. \]

(iv) \( L(xy) = L(x) + L(y) \) for all \( x, y \in (0, \infty) \).

(v) If \( q \) is a rational number and \( x > 0 \), then \( L(x^q) = qL(x) \).

**Proof.** To prove existence, we define a candidate and show it satisfies all the properties. Let
\[ L(x) := \int_1^x \frac{1}{t} \, dt. \]

Obviously, (i) holds. Property (ii) holds via the second form of the fundamental theorem of
calculus (Theorem 5.3.3).

To prove property (iv), we change variables \( u = yt \) to obtain
\[ L(x) = \int_1^x \frac{1}{t} \, dt = \int_y^{xy} \frac{1}{u} \, du = \int_1^{xy} \frac{1}{u} \, du - \int_1^y \frac{1}{u} \, du = L(xy) - L(y). \]
Let us prove (iii). Property (ii) together with the fact that $L'(x) = 1/x > 0$ for $x > 0$, implies that $L$ is strictly increasing and hence one-to-one. Let us show $L$ is onto. As $1/t \geq 1/2$ when $t \in [1,2]$, 

$$L(2) = \int_1^2 \frac{1}{t} \, dt \geq 1/2.$$ 

By induction, (iv) implies that for $n \in \mathbb{N}$ 

$$L(2^n) = L(2) + L(2) + \cdots + L(2) = nL(2).$$

Given $y > 0$, by the Archimedean property of the real numbers (notice $L(2) > 0$), there is an $n \in \mathbb{N}$ such that $L(2^n) > y$. By the intermediate value theorem there is an $x_1 \in (1, 2^n)$ such that $L(x_1) = y$. We get $(0, \infty)$ is in the image of $L$. As $L$ is increasing, $L(x) > y$ for all $x > 2^n$, and so 

$$\lim_{x \to \infty} L(x) = \infty.$$ 

Next $0 = L(1/x) = L(x) + L(1/x)$, and so $L(x) = -L(1/x)$. Using $x = 2^{-n}$, we obtain as above that $L$ achieves all negative numbers. And 

$$\lim_{x \to 0} L(x) = \lim_{x \to 0} -L(1/x) = \lim_{x \to \infty} -L(x) = -\infty.$$ 

In the limits, note that only $x > 0$ are in the domain of $L$. 

Let us prove (v). Fix $x > 0$. As above, (iv) implies $L(x^n) = nL(x)$ for all $n \in \mathbb{N}$. We already found that $L(x) = -L(1/x)$, so $L(x^{-n}) = -L(x^n) = -nL(x)$. Then for $m \in \mathbb{N}$ 

$$L(x) = L\left(\left(x^{1/m}\right)^m\right) = mL(x^{1/m}).$$

Putting everything together for $n \in \mathbb{Z}$ and $m \in \mathbb{N}$, we have $L(x^{n/m}) = nL(x^{1/m}) = (n/m)L(x)$. 

Uniqueness follows using properties (i) and (ii). Via the first form of the fundamental theorem of calculus (Theorem 5.3.1), 

$$L(x) = \int_1^x \frac{1}{t} \, dt$$

is the unique function such that $L(1) = 0$ and $L'(x) = 1/x$. 

Having proved that there is a unique function with these properties, we simply define the logarithm or sometimes called the natural logarithm: 

$$\ln(x) := L(x).$$

Mathematicians usually write $\log(x)$ instead of $\ln(x)$, which is more familiar to calculus students. For all practical purposes, there is only one logarithm: the natural logarithm. See Exercise 5.4.2.
5.4.2 The exponential

Just as with the logarithm we define the exponential via a list of properties.

**Proposition 5.4.2.** There exists a unique function $E : \mathbb{R} \to (0, \infty)$ such that

(i) $E(0) = 1$.

(ii) $E$ is differentiable and $E'(x) = E(x)$.

(iii) $E$ is strictly increasing, bijective, and

$$\lim_{x \to -\infty} E(x) = 0, \quad \text{and} \quad \lim_{x \to \infty} E(x) = \infty.$$ 

(iv) $E(x + y) = E(x)E(y)$ for all $x, y \in \mathbb{R}$.

(v) If $q \in \mathbb{Q}$, then $E(qx) = E(x)^q$.

**Proof.** Again, we prove existence of such a function by defining a candidate and proving that it satisfies all the properties. The $L = \ln$ defined above is invertible. Let $E$ be the inverse function of $L$. Property (i) is immediate.

Property (ii) follows via the inverse function theorem, in particular Lemma 4.4.1: $L$ satisfies all the hypotheses of the lemma, and hence

$$E'(x) = \frac{1}{L'(E(x))} = E(x).$$

Let us look at property (iii). The function $E$ is strictly increasing since $E'(x) = E(x) > 0$. As $E$ is the inverse of $L$, it must also be bijective. To find the limits, we use that $E$ is strictly increasing and onto $(0, \infty)$. For every $M > 0$, there is an $x_0$ such that $E(x_0) = M$ and $E(x) \geq M$ for all $x \geq x_0$. Similarly, for every $\varepsilon > 0$, there is an $x_0$ such that $E(x_0) = \varepsilon$ and $E(x) < \varepsilon$ for all $x < x_0$. Therefore,

$$\lim_{n \to -\infty} E(x) = 0, \quad \text{and} \quad \lim_{n \to \infty} E(x) = \infty.$$ 

To prove property (iv), we use the corresponding property for the logarithm. Take $x, y \in \mathbb{R}$. As $L$ is bijective, find $a$ and $b$ such that $x = L(a)$ and $y = L(b)$. Then

$$E(x + y) = E(L(a) + L(b)) = E(L(ab)) = ab = E(x)E(y).$$

Property (v) also follows from the corresponding property of $L$. Given $x \in \mathbb{R}$, let $a$ be such that $x = L(a)$ and

$$E(qx) = E(qL(a)) = E(L(a^q)) = a^q = E(x)^q.$$ 

Uniqueness follows from (i) and (ii). Let $E$ and $F$ be two functions satisfying (i) and (ii).

$$\frac{d}{dx} \left( F(x)E(-x) \right) = F'(x)E(-x) - E'(-x)F(x) = F(x)E(-x) - E(-x)F(x) = 0.$$ 

Therefore, by Proposition 4.2.6, $F(x)E(-x) = F(0)E(-0) = 1$ for all $x \in \mathbb{R}$. Next, $1 = E(0) = E(x-x) = E(x)E(-x)$. Then

$$0 = 1 - 1 = F(x)E(-x) - E(x)E(-x) = (F(x) - E(x))E(-x).$$
Finally, \( E(-x) \neq 0^* \) for all \( x \in \mathbb{R} \). So \( F(x) - E(x) = 0 \) for all \( x \), and we are done. \( \square \)

Having proved \( E \) is unique, we define the exponential function as

\[
\exp(x) := E(x).
\]

If \( y \in \mathbb{Q} \) and \( x > 0 \), then

\[
x^y = \exp(\ln(x^y)) = \exp(y\ln(x)).
\]

We can now make sense of exponentiation \( x^y \) for arbitrary \( y \in \mathbb{R} \); if \( x > 0 \) and \( y \) is irrational, define

\[
x^y := \exp(y\ln(x)).
\]

As \( \exp \) is continuous, then \( x^y \) is a continuous function of \( y \). Therefore, we would obtain the same result had we taken a sequence of rational numbers \( \{y_n\} \) approaching \( y \) and defined \( x^y = \lim x^{y_n} \).

Define the number \( e \), sometimes called Euler’s number or the base of the natural logarithm, as

\[
e := \exp(1).
\]

Let us justify the notation \( e^x \) for \( \exp(x) \):

\[
e^x = \exp(x\ln(e)) = \exp(x).
\]

The properties of the logarithm and the exponential extend to irrational powers. The proof is immediate.

**Proposition 5.4.3.** Let \( x, y \in \mathbb{R} \).

(i) \( \exp(xy) = (\exp(x))^y \).

(ii) If \( x > 0 \), then \( \ln(x^y) = y\ln(x) \).

**Remark 5.4.4.** There are other equivalent ways to define the exponential and the logarithm. A common way is to define \( E \) as the solution to the differential equation \( E'(x) = E(x) \), \( E(0) = 1 \). See Example 6.3.3, for a sketch of that approach. Yet another approach is to define the exponential function by power series, see Example 6.2.14.

**Remark 5.4.5.** We proved the uniqueness of the functions \( L \) and \( E \) from just the properties \( L(1) = 0 \), \( L'(x) = 1/x \) and the equivalent condition for the exponential \( E'(x) = E(x) \), \( E(0) = 1 \). Existence also follows from just these properties. Alternatively, uniqueness also follows from the laws of exponents, see the exercises.

\(^*E \) is a function into \((0, \infty)\) after all. However, \( E(-x) \neq 0 \) also follows from \( E(x)E(-x) = 1 \). Therefore, we can prove uniqueness of \( E \) given (i) and (ii), even for functions \( E : \mathbb{R} \to \mathbb{R} \).
5.4.3 Exercises

Exercise 5.4.1: Given a real number \( y \) and \( b > 0 \), define \( f : (0, \infty) \to \mathbb{R} \) and \( g : \mathbb{R} \to \mathbb{R} \) as \( f(x) := x^y \) and \( g(x) := b^x \). Show that \( f \) and \( g \) are differentiable and find their derivative.

Exercise 5.4.2: Let \( b > 0, b \neq 1 \) be given.

a) Show that for every \( y > 0 \), there exists a unique number \( x \) such that \( y = b^x \). Define the logarithm base \( b \), \( \log_b : (0, \infty) \to \mathbb{R} \), by \( \log_b(y) := x \).

b) Show that \( \log_b(x) = \frac{\ln(x)}{\ln(b)} \).

c) Prove that if \( c > 0, c \neq 1 \), then \( \log_b(x) = \log_c(x) \cdot \frac{\ln(b)}{\ln(c)} \).

d) Prove \( \log_b(xy) = \log_b(x) + \log_b(y) \), and \( \log_b(x^y) = y \log_b(x) \).

Exercise 5.4.3 (requires §4.3): Use Taylor’s theorem to study the remainder term and show that for all \( x \in \mathbb{R} \)
\[
e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!}.
\]

Hint: Do not differentiate the series term by term (unless you would prove that it works).

Exercise 5.4.4: Use the geometric sum formula to show (for \( t \neq -1 \))
\[
1 - t + t^2 - \cdots + (-1)^n t^n = \frac{1}{1+t} - \frac{(-1)^n+1}{1+t}.
\]

Using this fact show
\[
\ln(1+x) = \sum_{n=1}^{\infty} \frac{(-1)^n+1}{n} x^n
\]
for all \( x \in (-1,1] \) (note that \( x = 1 \) is included). Finally, find the limit of the alternating harmonic series
\[
\sum_{n=1}^{\infty} \frac{(-1)^n+1}{n} = 1 - \frac{1}{2} + \frac{1}{3} - \frac{1}{4} + \cdots
\]

Exercise 5.4.5: Show
\[
e^x = \lim_{n \to \infty} \left(1 + \frac{x}{n}\right)^n.
\]

Hint: Take the logarithm.

Note: The expression \( (1 + \frac{x}{n})^n \) arises in compound interest calculations. It is the amount of money in a bank account after 1 year if 1 dollar was deposited initially at interest \( x \) and the interest was compounded \( n \) times during the year. The exponential \( e^x \) is the result of continuous compounding.

Exercise 5.4.6:

a) Prove that for \( n \in \mathbb{N} \),
\[
\sum_{k=2}^{n} \frac{1}{k} \leq \ln(n) \leq \sum_{k=1}^{n-1} \frac{1}{k}.
\]

b) Prove that the limit
\[
\gamma := \lim_{n \to \infty} \left( \sum_{k=1}^{n} \frac{1}{k} - \ln(n) \right)
\]
exists. This constant is known as the Euler–Mascheroni constant\(^*\). It is not known if this constant is rational or not. It is approximately \( \gamma \approx 0.5772 \).

\(^*\)Named for the Swiss mathematician Leonhard Paul Euler (1707–1783) and the Italian mathematician Lorenzo Mascheroni (1750–1800).
Exercise 5.4.7: Show
\[ \lim_{x \to \infty} \frac{\ln(x)}{x} = 0. \]

Exercise 5.4.8: Show that \( e^x \) is convex, in other words, show that if \( a \leq x \leq b \), then \( e^x \leq e^{\frac{b-x}{b-a}} + e^{\frac{x-a}{b-a}} \).

Exercise 5.4.9: Using the logarithm find
\[ \lim_{n \to \infty} n^{1/n}. \]

Exercise 5.4.10: Show that \( E(x) = e^x \) is the unique continuous function such that \( E(x+y) = E(x)E(y) \) and \( E(1) = e \). Similarly, prove that \( L(x) = \ln(x) \) is the unique continuous function defined on positive \( x \) such that \( L(xy) = L(x) + L(y) \) and \( L(e) = 1 \).

Exercise 5.4.11 (requires §4.3): Since \( (e^x)' = e^x \), it is easy to see that \( e^x \) is infinitely differentiable (has derivatives of all orders). Define the function \( f: \mathbb{R} \to \mathbb{R} \).
\[
 f(x) := \begin{cases} 
 e^{-1/x} & \text{if } x > 0, \\
 0 & \text{if } x \leq 0.
\end{cases}
\]

a) Prove that for every \( m \in \mathbb{N} \),
\[
 \lim_{x \to 0^+} \frac{e^{-1/x}}{x^m} = 0.
\]

b) Prove that \( f \) is infinitely differentiable.

c) Compute the Taylor series for \( f \) at the origin, that is,
\[
 \sum_{k=0}^{\infty} \frac{f^{(k)}(0)}{k!} x^k.
\]

Show that it converges, but show that it does not converge to \( f(x) \) for any given \( x > 0 \).
5.5 Improper integrals

Note: 2–3 lectures (optional section, can safely be skipped, requires the optional §3.5)

Often it is necessary to integrate over the entire real line, or an unbounded interval of the form \([a, \infty)\) or \((-\infty, b]\). We may also wish to integrate unbounded functions defined on a open bounded interval \((a, b)\). For such intervals or functions, the Riemann integral is not defined, but we will write down the integral anyway in the spirit of Lemma 5.2.8. These integrals are called improper integrals and are limits of integrals rather than integrals themselves.

**Definition 5.5.1.** Suppose \(f: [a, b) \to \mathbb{R}\) is a function (not necessarily bounded) that is Riemann integrable on \([a, c)\) for all \(c < b\). We define

\[
\int_a^b f := \lim_{c \to b^-} \int_a^c f
\]

if the limit exists.

Suppose \(f: [a, \infty) \to \mathbb{R}\) is a function such that \(f\) is Riemann integrable on \([a, c)\) for all \(c < \infty\). We define

\[
\int_a^\infty f := \lim_{c \to \infty} \int_a^c f
\]

if the limit exists.

If the limit exists, we say the improper integral converges. If the limit does not exist, we say the improper integral diverges.

We similarly define improper integrals for the left-hand endpoint, we leave this to the reader.

For a finite endpoint \(b\), if \(f\) is bounded, then Lemma 5.2.8 says that we defined nothing new. What is new is that we can apply this definition to unbounded functions. The following set of examples is so useful that we state it as a proposition.

**Proposition 5.5.2** (\(p\)-test for integrals). The improper integral

\[
\int_1^\infty \frac{1}{x^p} \, dx
\]

converges to \(\frac{1}{p-1}\) if \(p > 1\) and diverges if \(0 < p \leq 1\).

The improper integral

\[
\int_0^1 \frac{1}{x^p} \, dx
\]

converges to \(\frac{1}{1-p}\) if \(0 < p < 1\) and diverges if \(p \geq 1\).

**Proof.** The proof follows by application of the fundamental theorem of calculus. Let us do the proof for \(p > 1\) for the infinite right endpoint and leave the rest to the reader. Hint: You should handle \(p = 1\) separately.

Suppose \(p > 1\). Then using the fundamental theorem,

\[
\int_1^b \frac{1}{x^p} \, dx = \int_1^b x^{-p} \, dx = \left[ \frac{x^{-p+1}}{-p+1} \right]_1^b = \frac{b^{-p+1}}{-p+1} - \frac{1^{-p+1}}{-p+1} = \frac{-1}{(p-1)b^{p-1}} + \frac{1}{p-1}.
\]

As \(p > 1\), then \(p-1 > 0\). Take the limit as \(b \to \infty\) to obtain that \(\frac{1}{b^{p-1}}\) goes to 0. The result follows.
We state the following proposition on “tails” for just one type of improper integral, though the proof is straightforward and the same for other types of improper integrals.

**Proposition 5.5.3.** Let \( f : [a, \infty) \rightarrow \mathbb{R} \) be a function that is Riemann integrable on \([a, b]\) for all \( b > a \). For every \( b > a \), the integral \( \int_b^\infty f \) converges if and only if \( \int_a^\infty f \) converges, in which case

\[
\int_a^\infty f = \int_a^b f + \int_b^\infty f.
\]

**Proof.** Let \( c > b \). Then

\[
\int_a^c f = \int_a^b f + \int_b^c f.
\]

Taking the limit \( c \to \infty \) finishes the proof.

Nonnegative functions are easier to work with as the following proposition demonstrates. The exercises will show that this proposition holds only for nonnegative functions. Analogues of this proposition exist for all the other types of improper limits and are left to the student.

**Proposition 5.5.4.** Suppose \( f : [a, \infty) \rightarrow \mathbb{R} \) is nonnegative (\( f(x) \geq 0 \) for all \( x \)) and such that \( f \) is Riemann integrable on \([a, b]\) for all \( b > a \).

(i) \[
\int_a^\infty f = \sup \left\{ \int_a^x f : x \geq a \right\}.
\]

(ii) Suppose \( \{x_n\} \) is a sequence with \( \lim x_n = \infty \). Then \( \int_a^\infty f \) converges if and only if \( \lim \int_a^{x_n} f \) exists, in which case

\[
\int_a^\infty f = \lim_{n \to \infty} \int_a^{x_n} f.
\]

In the first item we allow for the value of \( \infty \) in the supremum indicating that the integral diverges to infinity.

**Proof.** We start with the first item. As \( f \) is nonnegative, \( \int_a^x f \) is increasing as a function of \( x \). If the supremum is infinite, then for every \( M \in \mathbb{R} \) we find \( N \) such that \( \int_a^N f \geq M \). As \( \int_a^x f \) is increasing, \( \int_a^x f \geq M \) for all \( x \geq N \). So \( \int_a^\infty f \) diverges to infinity.

Next suppose the supremum is finite, say \( A := \sup \{ \int_a^x f : x \geq a \} \). For every \( \varepsilon > 0 \), we find an \( N \) such that \( A - \int_a^{x_N} f < \varepsilon \). As \( \int_a^x f \) is increasing, then \( A - \int_a^x f < \varepsilon \) for all \( x \geq N \) and hence \( \int_a^\infty f \) converges to \( A \).

Let us look at the second item. If \( \int_a^\infty f \) converges, then every sequence \( \{x_n\} \) going to infinity works. The trick is proving the other direction. Suppose \( \{x_n\} \) is such that \( \lim x_n = \infty \) and

\[
\lim_{n \to \infty} \int_a^{x_n} f = A
\]

converges. Given \( \varepsilon > 0 \), pick \( N \) such that for all \( n \geq N \), we have \( A - \varepsilon < \int_a^{x_n} f < A + \varepsilon \). Because \( \int_a^x f \) is increasing as a function of \( x \), we have that for all \( x \geq x_N \)

\[
A - \varepsilon < \int_a^{x_N} f \leq \int_a^x f.
\]
As \( \{x_n\} \) goes to \( \infty \), then for any given \( x \), there is an \( x_m \) such that \( m \geq N \) and \( x \leq x_m \). Then

\[
\int_a^x f \leq \int_a^{x_m} f < A + \epsilon.
\]

In particular, for all \( x \geq x_N \), we have \( |\int_a^x f - A| < \epsilon \).

\textbf{Proposition 5.5.5} (Comparison test for improper integrals). Let \( f : [a, \infty) \to \mathbb{R} \) and \( g : [a, \infty) \to \mathbb{R} \) be functions that are Riemann integrable on \( [a, b] \) for all \( b > a \). Suppose that for all \( x \geq a \),

\[
|f(x)| \leq g(x).
\]

(i) If \( \int_a^\infty g \) converges, then \( \int_a^\infty f \) converges, and in this case \( |\int_a^\infty f| \leq \int_a^\infty g \).

(ii) If \( \int_a^\infty f \) diverges, then \( \int_a^\infty g \) diverges.

\textbf{Proof.} We start with the first item. For every \( b \) and \( c \), such that \( a \leq b \leq c \), we have \( -g(x) \leq f(x) \leq g(x) \), and so

\[
\int_b^c g - \int_b^c f \leq \int_b^c g.
\]

In other words, \( \int_b^c f \leq \int_b^c g \).

Let \( \epsilon > 0 \) be given. Because of \textbf{Proposition 5.5.3},

\[
\int_a^\infty g = \int_a^b g + \int_b^\infty g.
\]

As \( \int_a^b g \) goes to \( \int_a^\infty g \) as \( b \) goes to infinity, \( \int_b^\infty g \) goes to 0 as \( b \) goes to infinity. Choose \( B \) such that

\[
\int_B^\infty g < \epsilon.
\]

As \( g \) is nonnegative, if \( B \leq b < c \), then \( \int_b^c g < \epsilon \) as well. Let \( \{x_n\} \) be a sequence going to infinity. Let \( M \) be such that \( x_n \geq B \) for all \( n \geq M \). Take \( n, m \geq M \), with \( x_n \leq x_m \),

\[
\left| \int_{x_n}^{x_m} f - \int_{x_n}^{x_m} f \right| = \left| \int_{x_n}^{x_m} f \right| \leq \int_{x_n}^{x_m} g < \epsilon.
\]

Therefore, the sequence \( \{\int_{x_n}^{x_m} f\}_{n=1}^{\infty} \) is Cauchy and hence converges.

We need to show that the limit is unique. Suppose \( \{x_n\} \) is a sequence converging to infinity such that \( \{\int_{a}^{x_n} f\} \) converges to \( L_1 \), and \( \{y_n\} \) is a sequence converging to infinity is such that \( \{\int_{a}^{y_n} f\} \) converges to \( L_2 \). Then there must be some \( n \) such that \( |\int_{a}^{x_n} f - L_1| < \epsilon \) and \( |\int_{a}^{y_n} f - L_2| < \epsilon \). We can also suppose \( x_n \geq B \) and \( y_n \geq B \). Then

\[
|L_1 - L_2| \leq |L_1 - \int_{a}^{x_n} f| + |\int_{a}^{x_n} f - \int_{a}^{y_n} f| + |\int_{a}^{y_n} f - L_2| < \epsilon + |\int_{a}^{y_n} f| + \epsilon < 3\epsilon.
\]

As \( \epsilon > 0 \) was arbitrary, \( L_1 = L_2 \), and hence \( \int_a^\infty f \) converges. Above we have shown that \( |\int_a^c f| \leq \int_a^c g \) for all \( c > a \). By taking the limit \( c \to \infty \), the first item is proved.

The second item is simply a contrapositive of the first item. \( \square \)
Example 5.5.6: The improper integral

\[
\int_0^\infty \frac{\sin(x^2)(x+2)}{x^3+1} \, dx
\]
converges.

Proof: Observe we simply need to show that the integral converges when going from 1 to infinity. For \(x \geq 1\) we obtain

\[
\left| \frac{\sin(x^2)(x+2)}{x^3+1} \right| \leq \frac{x+2}{x^3+1} \leq \frac{x+2}{x^3} \leq \frac{3}{x^2}.
\]

Then

\[
\int_1^\infty \frac{3}{x^2} \, dx = 3 \int_1^\infty \frac{1}{x^2} \, dx = 3.
\]

So using the comparison test and the tail test, the original integral converges.

Example 5.5.7: You should be careful when doing formal manipulations with improper integrals. The integral

\[
\int_2^\infty \frac{2}{x^2-1} \, dx
\]
converges via the comparison test using \(1/x^2\) again. However, if you succumb to the temptation to write

\[
\frac{2}{x^2-1} = \frac{1}{x-1} - \frac{1}{x+1}
\]

and try to integrate each part separately, you will not succeed. It is not true that you can split the improper integral in two; you cannot split the limit.

\[
\int_2^\infty \frac{2}{x^2-1} \, dx = \lim_{b \to \infty} \int_2^b \frac{2}{x^2-1} \, dx
\]

\[
= \lim_{b \to \infty} \left( \int_2^b \frac{1}{x-1} \, dx - \int_2^b \frac{1}{x+1} \, dx \right)
\]

\[
\neq \int_2^\infty \frac{1}{x-1} \, dx - \int_2^\infty \frac{1}{x+1} \, dx.
\]

The last line in the computation does not even make sense. Both of the integrals there diverge to infinity, since we can apply the comparison test appropriately with \(1/x\). We get \(\infty - \infty\).

Now suppose we need to take limits at both endpoints.

Definition 5.5.8. Suppose \(f: (a, b) \to \mathbb{R}\) is a function that is Riemann integrable on \([c, d]\) for all \(c, d\) such that \(a < c < d < b\), then we define

\[
\int_a^b f := \lim_{c \to a^+} \lim_{d \to b^-} \int_c^d f
\]

if the limits exist.
Suppose $f: \mathbb{R} \to \mathbb{R}$ is a function such that $f$ is Riemann integrable on all bounded intervals $[a, b]$. Then we define

$$\int_{-\infty}^{\infty} f := \lim_{c \to -\infty} \lim_{d \to \infty} \int_{c}^{d} f$$

if the limits exist.

We similarly define improper integrals with one infinite and one finite improper endpoint, we leave this to the reader.

One ought to always be careful about double limits. The definition given above says that we first take the limit as $d$ goes to $b$ or $\infty$ for a fixed $c$, and then we take the limit in $c$. We will have to prove that in this case it does not matter which limit we compute first.

**Example 5.5.9:** Let us see an example:

$$\int_{-\infty}^{\infty} \frac{1}{1 + x^2} \, dx = \lim_{a \to -\infty} \lim_{b \to \infty} \int_{a}^{b} \frac{1}{1 + x^2} \, dx = \lim_{a \to -\infty} \lim_{b \to \infty} (\arctan(b) - \arctan(a)) = \pi.$$

In the definition the order of the limits can always be switched if they exist. Let us prove this fact only for the infinite limits.

**Proposition 5.5.10.** If $f: \mathbb{R} \to \mathbb{R}$ is a function integrable on every bounded interval $[a, b]$. Then

$$\lim_{a \to -\infty} \lim_{b \to \infty} \int_{a}^{b} f \quad \text{converges} \quad \text{if and only if} \quad \lim_{b \to \infty} \lim_{a \to -\infty} \int_{a}^{b} f \quad \text{converges},$$

in which case the two expressions are equal. If either of the expressions converges, then the improper integral converges and

$$\lim_{a \to -\infty} \int_{a}^{\infty} f = \int_{-\infty}^{\infty} f.$$

**Proof.** Without loss of generality, assume $a < 0$ and $b > 0$. Suppose the first expression converges. Then

$$\lim_{a \to -\infty} \lim_{b \to \infty} \int_{a}^{b} f = \lim_{a \to -\infty} \lim_{b \to \infty} \left( \int_{a}^{0} f + \int_{0}^{b} f \right) = \left( \lim_{a \to -\infty} \int_{a}^{0} f \right) + \left( \lim_{b \to \infty} \int_{0}^{b} f \right)$$

$$= \lim_{b \to \infty} \left( \lim_{a \to -\infty} \int_{a}^{0} f \right) + \int_{0}^{b} f = \lim_{b \to \infty} \lim_{a \to -\infty} \left( \int_{a}^{0} f + \int_{0}^{b} f \right).$$

Similar computation shows the other direction. Therefore, if either expression converges, then the improper integral converges and

$$\int_{-\infty}^{\infty} f = \lim_{a \to -\infty} \lim_{b \to \infty} \int_{a}^{b} f = \left( \lim_{a \to -\infty} \int_{a}^{0} f \right) + \left( \lim_{b \to \infty} \int_{0}^{b} f \right)$$

$$= \left( \lim_{a \to -\infty} \int_{-a}^{0} f \right) + \left( \lim_{a \to \infty} \int_{0}^{a} f \right) = \lim_{a \to -\infty} \left( \int_{-a}^{0} f + \int_{0}^{a} f \right) = \lim_{a \to \infty} \int_{-a}^{a} f.$$ 

\[\square\]
Example 5.5.11: On the other hand, you must be careful to take the limits independently before you know convergence. Let \( f(x) = \frac{x}{|x|} \) for \( x \neq 0 \) and \( f(0) = 0 \). If \( a < 0 \) and \( b > 0 \), then

\[
\int_a^b f = \int_a^0 f + \int_0^b f = a + b.
\]

For every fixed \( a < 0 \), the limit as \( b \to \infty \) is infinite. So even the first limit does not exist, and the improper integral \( \int_{-\infty}^{\infty} f \) does not converge. On the other hand, if \( a > 0 \), then

\[
\int_{-a}^a f = (-a) + a = 0.
\]

Therefore,

\[
\lim_{a \to \infty} \int_{-a}^a f = 0.
\]

Example 5.5.12: An example to keep in mind for improper integrals is the so-called sinc function*. This function comes up quite often in both pure and applied mathematics. Define

\[
sinc(x) = \begin{cases} 
\frac{\sin(x)}{x} & \text{if } x \neq 0, \\
1 & \text{if } x = 0.
\end{cases}
\]

![Figure 5.6: The sinc function.](image)

It is not difficult to show that the sinc function is continuous at zero, but that is not important right now. What is important is that

\[
\int_{-\infty}^{\infty} \sin(x) \, dx = \pi, \quad \text{while} \quad \int_{-\infty}^{\infty} |\sin(x)| \, dx = \infty.
\]

The integral of the sinc function is a continuous analogue of the alternating harmonic series \( \sum (-1)^n/n \), while the absolute value is like the regular harmonic series \( \sum 1/n \). In particular, the fact that the integral converges must be done directly rather than using comparison test.

*Shortened from Latin: sinus cardinalis
We will not prove the first statement exactly. Let us simply prove that the integral of the sinc function converges, but we will not worry about the exact limit. Because \( \frac{\sin(-x)}{-x} = \frac{\sin(x)}{x} \), it is enough to show that
\[
\int_{-\infty}^{\infty} \frac{\sin(x)}{x} \, dx
\]
converges. We also avoid \( x = 0 \) this way to make our life simpler.

For every \( n \in \mathbb{N} \), we have that for \( x \in [\pi 2n, \pi (2n+1)] \),
\[
\frac{\sin(x)}{\pi (2n+1)} \leq \frac{\sin(x)}{x} \leq \frac{\sin(x)}{\pi 2n},
\]
as \( \sin(x) \geq 0 \). For \( x \in [\pi (2n+1), \pi (2n+2)] \),
\[
\frac{\sin(x)}{\pi (2n+1)} \leq \frac{\sin(x)}{x} \leq \frac{\sin(x)}{\pi (2n+2)},
\]
as \( \sin(x) \leq 0 \).

Via the fundamental theorem of calculus,
\[
\frac{2}{\pi (2n+1)} = \int_{\pi 2n}^{\pi (2n+1)} \frac{\sin(x)}{\pi (2n+1)} \, dx \leq \int_{\pi 2n}^{\pi (2n+1)} \frac{\sin(x)}{x} \, dx \leq \int_{\pi 2n}^{\pi (2n+1)} \frac{\sin(x)}{\pi 2n} \, dx = \frac{1}{\pi n}.
\]
Similarly,
\[
\frac{-2}{\pi (2n+1)} \leq \int_{\pi (2n+1)}^{\pi (2n+2)} \frac{\sin(x)}{x} \, dx \leq \frac{-1}{\pi (n+1)}.
\]
Adding the two together we find
\[
0 = \frac{2}{\pi (2n+1)} + \frac{-2}{\pi (2n+1)} \leq \int_{2\pi n}^{2\pi (n+1)} \frac{\sin(x)}{x} \, dx \leq \frac{1}{\pi n} + \frac{-1}{\pi (n+1)} = \frac{1}{\pi n(n+1)}.
\]
See Figure 5.7.

---

**Figure 5.7:** Bound of \( \int_{2\pi n}^{2\pi (n+1)} \frac{\sin(x)}{x} \, dx \) using the shaded integral (signed area \( \frac{1}{\pi n} + \frac{-1}{\pi (n+1)} \)).
For \( k \in \mathbb{N} \),
\[
\int_{2\pi}^{2k\pi} \frac{\sin(x)}{x} \, dx = \sum_{n=1}^{k-1} \int_{2\pi n}^{2\pi(n+1)} \frac{\sin(x)}{x} \, dx \leq \sum_{n=1}^{k-1} \frac{1}{\pi n(n+1)}.
\]

We find the partial sums of a series with positive terms. The series converges as \( \sum \frac{1}{\pi n(n+1)} \) is a convergent series. Thus as a sequence,
\[
\lim_{k \to \infty} \int_{2\pi}^{2k\pi} \frac{\sin(x)}{x} \, dx = L \leq \frac{1}{\pi} \sum_{n=1}^{\infty} \frac{1}{n(n+1)} < \infty.
\]

Let \( M > 2\pi \) be arbitrary, and let \( k \in \mathbb{N} \) be the largest integer such that \( 2k\pi \leq M \). For \( x \in [2k\pi, M] \), we have \( -\frac{1}{2k\pi} \leq \frac{\sin(x)}{x} \leq \frac{1}{2k\pi} \), and so
\[
\left| \int_{2k\pi}^{M} \frac{\sin(x)}{x} \, dx \right| \leq \frac{M - 2k\pi}{2k\pi} \leq \frac{1}{k}.
\]

As \( k \) is the largest \( k \) such that \( 2k\pi \leq M \), then as \( M \in \mathbb{R} \) goes to infinity, so does \( k \in \mathbb{N} \).

Then
\[
\int_{2\pi}^{M} \frac{\sin(x)}{x} \, dx = \int_{2\pi}^{2k\pi} \frac{\sin(x)}{x} \, dx + \int_{2k\pi}^{M} \frac{\sin(x)}{x} \, dx.
\]

As \( M \) goes to infinity, the first term on the right-hand side goes to \( L \), and the second term on the right-hand side goes to zero. Hence
\[
\int_{2\pi}^{\infty} \frac{\sin(x)}{x} \, dx = L.
\]

The double-sided integral of \( \text{sinc} \) also exists as noted above. We leave the other statement—that the integral of the absolute value of the \( \text{sinc} \) function diverges—as an exercise.

### 5.5.1 Integral test for series

The fundamental theorem of calculus can be used in proving a series is summable and to estimate its sum.

**Proposition 5.5.13 (Integral test).** Suppose \( f : [k, \infty) \to \mathbb{R} \) is a decreasing nonnegative function where \( k \in \mathbb{Z} \). Then
\[
\sum_{n=k}^{\infty} f(n) \text{ converges if and only if } \int_{k}^{\infty} f(x) \, dx \text{ converges.}
\]

In this case
\[
\int_{k}^{\infty} f(x) \, dx \leq \sum_{n=k}^{\infty} f(n) \leq f(k) + \int_{k}^{\infty} f(x) \, dx.
\]

See Figure 5.8, for an illustration with \( k = 1 \). By Proposition 5.2.11, \( f \) is integrable on every interval \([k, b]\) for all \( b > k \), so the statement of the theorem makes sense without additional hypotheses of integrability.
CHAPTER 5. THE RIEMANN INTEGRAL

Figure 5.8: The area under the curve, \( \int_{1}^{\infty} f \), is bounded below by the area of the shaded rectangles, \( f(2) + f(3) + f(4) + \cdots \), and bounded above by the area entire rectangles, \( f(1) + f(2) + f(3) + \cdots \).

Proof. Let \( \ell, m \in \mathbb{Z} \) be such that \( m > \ell \geq k \). Because \( f \) is decreasing, we have \( \int_{n+1}^{m} f \leq f(n) \leq \int_{n}^{m-1} f \). Therefore,

\[
\int_{\ell}^{m} f = \sum_{n=\ell}^{m-1} \int_{n}^{n+1} f \leq \sum_{n=\ell}^{m-1} f(n) \leq f(\ell) + \sum_{n=\ell+1}^{m-1} \int_{n-1}^{n} f \leq f(\ell) + \int_{\ell}^{m-1} f. \tag{5.3}
\]

Suppose first that \( \int_{k}^{\infty} f \) converges and let \( \varepsilon > 0 \) be given. As before, since \( f \) is positive, then there exists an \( L \in \mathbb{N} \) such that if \( \ell \geq L \), then \( \int_{\ell}^{m} f < \varepsilon / 2 \) for all \( m \geq \ell \). The function \( f \) must decrease to zero (why?), so make \( L \) large enough so that for \( \ell \geq L \), we have \( f(\ell) < \varepsilon / 2 \). Thus, for \( m > \ell \geq L \), we have via (5.3),

\[
\sum_{n=\ell}^{m} f(n) \leq f(\ell) + \int_{\ell}^{m} f < \varepsilon / 2 + \varepsilon / 2 = \varepsilon.
\]

The series is therefore Cauchy and thus converges. The estimate in the proposition is obtained by letting \( m \) go to infinity in (5.3) with \( \ell = k \).

Conversely, suppose \( \int_{k}^{\infty} f \) diverges. As \( f \) is positive, then by Proposition 5.5.4, the sequence \( \{\int_{k}^{m} f\}_{m=k}^{\infty} \) diverges to infinity. Using (5.3) with \( \ell = k \), we find

\[
\int_{k}^{m} f \leq \sum_{n=k}^{m-1} f(n).
\]

As the left-hand side goes to infinity as \( m \to \infty \), so does the right-hand side. \( \square \)

Example 5.5.14: The integral test can be used not only to show that a series converges, but to estimate its sum to arbitrary precision. Let us show \( \sum_{n=1}^{\infty} \frac{1}{n^2} \) exists and estimate its sum to within 0.01. As this series is the \( p \)-series for \( p = 2 \), we already proved it converges (let us pretend we do not know that), but we only roughly estimated its sum.

The fundamental theorem of calculus says that for all \( k \in \mathbb{N} \),

\[
\int_{k}^{\infty} \frac{1}{x^2} \, dx = \frac{1}{k}.
\]
In particular, the series must converge. But we also have
\[
\frac{1}{k} = \int_k^\infty \frac{1}{x^2} \, dx \leq \sum_{n=k}^\infty \frac{1}{n^2} \leq \frac{1}{k^2} + \int_k^\infty \frac{1}{x^2} \, dx = \frac{1}{k^2} + \frac{1}{k}.
\]

Adding the partial sum up to \(k - 1\) we get
\[
\frac{1}{k} + \frac{1}{k-1} \sum_{n=1}^{k-1} \frac{1}{n^2} \leq \sum_{n=1}^\infty \frac{1}{n^2} \leq \frac{1}{k^2} + \frac{1}{k} + \frac{1}{k} + \frac{1}{k-1} \sum_{n=1}^{k-1} \frac{1}{n^2}.
\]

In other words, \(1/k + \sum_{n=1}^{k-1} 1/n^2\) is an estimate for the sum to within \(1/k^2\). Therefore, if we wish to find the sum to within 0.01, we note \(1/10^2 = 0.01\). We obtain
\[
1.6397 \ldots \approx \frac{1}{10} + \frac{1}{100} \sum_{n=1}^9 \frac{1}{n^2} \leq \sum_{n=1}^\infty \frac{1}{n^2} \leq \frac{1}{10} + \frac{1}{100} \sum_{n=1}^9 \frac{1}{n^2} \approx 1.6497 \ldots.
\]

The actual sum is \(\pi^2/6 \approx 1.6449 \ldots\).

### 5.5.2 Exercises

**Exercise 5.5.1:** Finish the proof of Proposition 5.5.2.

**Exercise 5.5.2:** Find out for which \(a \in \mathbb{R}\) does \(\sum_{n=1}^\infty e^{an}\) converge. When the series converges, find an upper bound for the sum.

**Exercise 5.5.3:**

a) Estimate \(\sum_{n=1}^\infty \frac{1}{n(n+1)}\) correct to within 0.01 using the integral test.

b) Compute the limit of the series exactly and compare. Hint: The sum telescopes.

**Exercise 5.5.4:** Prove
\[
\int_{-\infty}^\infty |\text{sinc}(x)| \, dx = \infty.
\]

Hint: Again, it is enough to show this on just one side.

**Exercise 5.5.5:** Can you interpret
\[
\int_{-1}^1 \frac{1}{\sqrt{|x|}} \, dx
\]

as an improper integral? If so, compute its value.

**Exercise 5.5.6:** Take \(f : [0, \infty) \to \mathbb{R}\), Riemann integrable on every interval \([0, b]\), and such that there exist \(M, a, \) and \(T\), such that \(|f(t)| \leq Me^{-at}\) for all \(t \geq T\). Show that the Laplace transform of \(f\) exists. That is, for every \(s > a\) the following integral converges:
\[
F(s) := \int_0^\infty f(t)e^{-st} \, dt.
\]
Exercise 5.5.7: Let \( f : \mathbb{R} \to \mathbb{R} \) be a Riemann integrable function on every interval \([a, b]\), and such that \( \int_{-\infty}^{\infty} |f(x)| \, dx < \infty \). Show that the Fourier sine and cosine transforms exist. That is, for every \( \omega \geq 0 \) the following integrals converge
\[
F^s(\omega) := \frac{1}{\pi} \int_{-\infty}^{\infty} f(t) \sin(\omega t) \, dt, \quad F^c(\omega) := \frac{1}{\pi} \int_{-\infty}^{\infty} f(t) \cos(\omega t) \, dt.
\]
Furthermore, show that \( F^s \) and \( F^c \) are bounded functions.

Exercise 5.5.8: Suppose \( f : [0, \infty) \to \mathbb{R} \) is Riemann integrable on every interval \([0, b]\). Show that \( \int_0^{\infty} f \) converges if and only if for every \( \varepsilon > 0 \) there exists an \( M \) such that if \( M \leq a < b \), then \( \left| \int_a^b f \right| < \varepsilon \).

Exercise 5.5.9: Suppose \( f : [0, \infty) \to \mathbb{R} \) is nonnegative and decreasing. Prove:
\begin{enumerate}[(a)]
  \item If \( \int_0^{\infty} f < \infty \), then \( \lim_{x \to \infty} f(x) = 0 \).
  \item The converse does not hold.
\end{enumerate}

Exercise 5.5.10: Find an example of an unbounded continuous function \( f : [0, \infty) \to \mathbb{R} \) that is nonnegative and such that \( \int_0^{\infty} f \) converges. Note that \( \lim_{x \to \infty} f(x) \) will not exist; compare previous exercise. Hint: On each interval \([k, k+1] \), \( k \in \mathbb{N} \), define a function whose integral over this interval is less than say \( 2^{-k} \).

Exercise 5.5.11 (More challenging): Find an example of a function \( f : [0, \infty) \to \mathbb{R} \) integrable on all intervals such that \( \lim_{n \to \infty} \int_0^f f \) converges as a limit of a sequence (so \( n \in \mathbb{N} \)), but such that \( \int_0^{\infty} f \) does not exist. Hint: For all \( n \in \mathbb{N} \), divide \([n, n+1] \) into two halves. On one half make the function negative, on the other make the function positive.

Exercise 5.5.12: Suppose \( f : [1, \infty) \to \mathbb{R} \) is such that \( g(x) := x^2 f(x) \) is a bounded function. Prove that \( \int_1^{\infty} f \) converges.

It is sometimes desirable to assign a value to integrals that normally cannot be interpreted even as improper integrals, e.g. \( \int_{-1}^{1} 1/\sqrt{x} \, dx \). Suppose \( f : [a, b] \to \mathbb{R} \) is a function and \( a < c < b \), where \( f \) is Riemann integrable on the intervals \([a, c - \varepsilon] \) and \([c + \varepsilon, b] \) for all \( \varepsilon > 0 \). Define the Cauchy principal value of \( \int_a^b f \) as
\[
p.v. \int_a^b f := \lim_{\varepsilon \to 0^+} \left( \int_a^{c-\varepsilon} f + \int_{c+\varepsilon}^b f \right),
\]
if the limit exists.

Exercise 5.5.13:
\begin{enumerate}[(a)]
  \item Compute \( p.v. \int_{-1}^{1} 1/\sqrt{x} \, dx \).
  \item Compute \( \lim_{\varepsilon \to 0^+} \left( \int_{-1}^{-\varepsilon} 1/\sqrt{x} \, dx + \int_{2\varepsilon}^{1} 1/\sqrt{x} \, dx \right) \) and show it is not equal to the principal value.
  \item Show that if \( f \) is integrable on \([a, b] \), then \( p.v. \int_a^b f = \int_a^b f \) (for an arbitrary \( c \in (a, b) \)).
  \item Suppose \( f : [-1, 1] \to \mathbb{R} \) is an odd function \( (f(-x) = -f(x)) \) that is integrable on \([-1, -\varepsilon] \) and \([\varepsilon, 1] \) for all \( \varepsilon > 0 \). Prove that \( p.v. \int_{-1}^{1} f = 0 \)
  \item Suppose \( f : [-1, 1] \to \mathbb{R} \) is continuous and differentiable at 0. Show that \( p.v. \int_{-1}^{1} \frac{f(x)}{x} \, dx \) exists.
Exercise 5.5.14: Let \( f : \mathbb{R} \to \mathbb{R} \) and \( g : \mathbb{R} \to \mathbb{R} \) be continuous functions, where \( g(x) = 0 \) for all \( x \notin [a, b] \) for some interval \([a, b]\).

a) Show that the convolution
\[
(g * f)(x) := \int_{-\infty}^{\infty} f(t)g(x-t) \, dt
\]
is well-defined for all \( x \in \mathbb{R} \).

b) Suppose \( \int_{-\infty}^{\infty} |f(x)| \, dx < \infty \). Prove that
\[
\lim_{x \to -\infty} (g * f)(x) = 0, \quad \text{and} \quad \lim_{x \to \infty} (g * f)(x) = 0.
\]
Chapter 6

Sequences of Functions

6.1 Pointwise and uniform convergence

Note: 1–1.5 lecture

Up till now, when we talked about limits of sequences we talked about sequences of numbers. A very useful concept in analysis is a sequence of functions. For example, a solution to some differential equation might be found by finding only approximate solutions. Then the actual solution is some sort of limit of those approximate solutions.

When talking about sequences of functions, the tricky part is that there are multiple notions of a limit. Let us describe two common notions of a limit of a sequence of functions.

6.1.1 Pointwise convergence

Definition 6.1.1. For every $n \in \mathbb{N}$, let $f_n : S \to \mathbb{R}$ be a function. The sequence $\{f_n\}_{n=1}^\infty$ converges pointwise to $f : S \to \mathbb{R}$ if for every $x \in S$, we have

$$f(x) = \lim_{n \to \infty} f_n(x).$$

Limits of sequences of numbers are unique, and so if a sequence $\{f_n\}$ converges pointwise, the limit function $f$ is unique. It is common to say that $f_n : S \to \mathbb{R}$ converges to $f$ on $T \subseteq S$ for some $f : T \to \mathbb{R}$. In that case we mean $f(x) = \lim f_n(x)$ for every $x \in T$. In other words, the restrictions of $f_n$ to $T$ converge pointwise to $f$.

Example 6.1.2: On $[-1, 1]$, the sequence of functions defined by $f_n(x) := x^{2n}$ converges pointwise to $f : [-1, 1] \to \mathbb{R}$, where

$$f(x) = \begin{cases} 1 & \text{if } x = -1 \text{ or } x = 1, \\ 0 & \text{otherwise.} \end{cases}$$

See Figure 6.1.

To see this is so, first take $x \in (-1, 1)$. Then $0 \leq x^2 < 1$. We have seen before that

$$|x^{2n} - 0| = (x^2)^n \to 0 \quad \text{as} \quad n \to \infty.$$ 

Therefore, $\lim f_n(x) = 0$. 
When \( x = 1 \) or \( x = -1 \), then \( x^{2n} = 1 \) for all \( n \) and hence \( \lim f_n(x) = 1 \). For all other \( x \), the sequence \( \{f_n(x)\} \) does not converge.

Often, functions are given as a series. In this case, we use the notion of pointwise convergence to find the values of the function.

**Example 6.1.3:** We write

\[
\sum_{k=0}^{\infty} x^k
\]

to denote the limit of the functions

\[
f_n(x) := \sum_{k=0}^{n} x^k.
\]

When studying series, we saw that for \((-1, 1)\) the \( f_n \) converge pointwise to

\[
\frac{1}{1-x}.
\]

The subtle point here is that while \( \frac{1}{1-x} \) is defined for all \( x \neq 1 \), and \( f_n \) are defined for all \( x \) (even at \( x = 1 \)), convergence only happens on \((-1, 1)\). Therefore, when we write

\[
f(x) := \sum_{k=0}^{\infty} x^k
\]

we mean that \( f \) is defined on \((-1, 1)\) and is the pointwise limit of the partial sums.

**Example 6.1.4:** Let \( f_n(x) := \sin(nx) \). Then \( f_n \) does not converge pointwise to any function on any interval. It may converge at certain points, such as when \( x = 0 \) or \( x = \pi \). It is left as an exercise that in any interval \([a,b]\), there exists an \( x \) such that \( \sin(nx) \) does not have a limit as \( n \) goes to infinity. See Figure 6.2.

Before we move to uniform convergence, let us reformulate pointwise convergence in a different way. We leave the proof to the reader—it is a simple application of the definition of convergence of a sequence of real numbers.
Proposition 6.1.5. Let \( f_n : S \to \mathbb{R} \) and \( f : S \to \mathbb{R} \) be functions. Then \( \{f_n\} \) converges pointwise to \( f \) if and only if for every \( x \in S \), and every \( \varepsilon > 0 \), there exists an \( N \in \mathbb{N} \) such that for all \( n \geq N \), we have
\[
|f_n(x) - f(x)| < \varepsilon.
\]

The key point here is that \( N \) can depend on \( x \), not just on \( \varepsilon \). That is, for each \( x \) we can pick a different \( N \). If we could pick one \( N \) for all \( x \), we would have what is called uniform convergence.

### 6.1.2 Uniform convergence

Definition 6.1.6. Let \( f_n : S \to \mathbb{R} \) and \( f : S \to \mathbb{R} \) be functions. The sequence \( \{f_n\} \) converges uniformly to \( f \) if for every \( \varepsilon > 0 \), there exists an \( N \in \mathbb{N} \) such that for all \( n \geq N \), we have
\[
|f_n(x) - f(x)| < \varepsilon \quad \text{for all } x \in S.
\]

In uniform convergence, \( N \) cannot depend on \( x \). Given \( \varepsilon > 0 \), we must find an \( N \) that works for all \( x \in S \). See Figure 6.3 for an illustration.

Uniform convergence implies pointwise convergence, and the proof follows by Proposition 6.1.5:

Proposition 6.1.7. Let \( \{f_n\} \) be a sequence of functions \( f_n : S \to \mathbb{R} \). If \( \{f_n\} \) converges uniformly to \( f : S \to \mathbb{R} \), then \( \{f_n\} \) converges pointwise to \( f \).
The converse does not hold.

**Example 6.1.8:** The functions \( f_n(x) := x^{2n} \) do not converge uniformly on \([-1, 1]\), even though they converge pointwise. To see this, suppose for contradiction that the convergence is uniform. For \( \varepsilon := 1/2 \), there would have to exist an \( N \) such that \( x^{2N} = |x^{2N} - 0| < 1/2 \) for all \( x \in (-1, 1) \) (as \( f_n(x) \) converges to 0 on \((-1, 1)\)). But that means that for every sequence \( \{x_k\} \) in \((-1, 1)\) such that \( \lim x_k = 1 \), we have \( x_k^{2N} < 1/2 \) for all \( k \). On the other hand, \( x^{2N} \) is a continuous function of \( x \) (it is a polynomial). Therefore, we obtain a contradiction

\[
1 = 1^{2N} = \lim_{k \to \infty} x_k^{2N} \leq 1/2.
\]

However, if we restrict our domain to \([-a,a]\) where \( 0 < a < 1 \), then \( \{f_n\} \) converges uniformly to 0 on \([-a,a]\). Note that \( a^{2n} \to 0 \) as \( n \to \infty \). Given \( \varepsilon > 0 \), pick \( N \in \mathbb{N} \) such that \( a^{2n} < \varepsilon \) for all \( n \geq N \). When \( x \in [-a,a] \), we have \( |x| \leq a \). So for all \( n \geq N \) and all \( x \in [-a,a] \), \n
\[
|x^{2n}| = |x|^{2n} \leq a^{2n} < \varepsilon.
\]

### 6.1.3 Convergence in uniform norm

For bounded functions, there is another more abstract way to think of uniform convergence. To every bounded function we assign a certain nonnegative number (called the uniform norm). This number measures the “distance” of the function from 0. We can then “measure” how far two functions are from each other. We then translate a statement about uniform convergence into a statement about a certain sequence of real numbers converging to zero.

**Definition 6.1.9.** Let \( f : S \to \mathbb{R} \) be a bounded function. Define

\[
\|f\|_u := \sup \{|f(x)| : x \in S\}.
\]

We call \( \|\cdot\|_u \) the **uniform norm**.

To use this notation\(^*\) and this concept, the domain \( S \) must be fixed. Some authors use the notation \( \|f\|_S \) to emphasize the dependence on \( S \).

**Proposition 6.1.10.** A sequence of bounded functions \( f_n : S \to \mathbb{R} \) converges uniformly to \( f : S \to \mathbb{R} \), if and only if

\[
\lim_{n \to \infty} \|f_n - f\|_u = 0.
\]

**Proof.** First suppose \( \lim \|f_n - f\|_u = 0 \). Let \( \varepsilon > 0 \) be given. Then there exists an \( N \) such that for \( n \geq N \), we have \( \|f_n - f\|_u < \varepsilon \). As \( \|f_n - f\|_u \) is the supremum of \( |f_n(x) - f(x)| \), we see that for all \( x \in S \), we have \( |f_n(x) - f(x)| \leq \|f_n - f\|_u < \varepsilon \).

On the other hand, suppose \( \{f_n\} \) converges uniformly to \( f \). Let \( \varepsilon > 0 \) be given. Then find \( N \) such that \( |f_n(x) - f(x)| < \varepsilon \) for all \( x \in S \). Taking the supremum we see that \( \|f_n - f\|_u \leq \varepsilon \). Hence \( \lim \|f_n - f\|_u = 0 \).

\(^*\)The notation nor terminology is not completely standardized. The norm is also called the **sup norm** or **infinity norm**, and in addition to \( \|f\|_u \) and \( \|f\|_S \) it is sometimes written as \( \|f\|_w \) or \( \|f\|_{w,S} \).
Sometimes it is said that \( \{ f_n \} \) converges to \( f \) in uniform norm instead of converges uniformly if \( \| f_n - f \| \to 0 \). The proposition says that the two notions are the same thing.

**Example 6.1.11:** Let \( f_n : [0, 1] \to \mathbb{R} \) be defined by \( f_n(x) := \frac{nx + \sin(nx^2)}{n} \). We claim \( \{ f_n \} \) converges uniformly to \( f(x) := x \). Let us compute:

\[
\| f_n - f \|_u = \sup \left\{ \left| \frac{nx + \sin(nx^2)}{n} - x \right| : x \in [0, 1] \right\} \\
= \sup \left\{ \left| \frac{\sin(nx^2)}{n} \right| : x \in [0, 1] \right\} \\
\leq \sup \left\{ \frac{1}{n} : x \in [0, 1] \right\} \\
= \frac{1}{n}.
\]

Using uniform norm, we define Cauchy sequences in a similar way as we define Cauchy sequences of real numbers.

**Definition 6.1.12.** Let \( f_n : S \to \mathbb{R} \) be bounded functions. The sequence is Cauchy in the uniform norm or uniformly Cauchy if for every \( \varepsilon > 0 \), there exists an \( N \in \mathbb{N} \) such that for all \( m, k \geq N \),

\[
\| f_m - f_k \|_u < \varepsilon.
\]

**Proposition 6.1.13.** Let \( f_n : S \to \mathbb{R} \) be bounded functions. Then \( \{ f_n \} \) is Cauchy in the uniform norm if and only if there exists an \( f : S \to \mathbb{R} \) and \( \{ f_n \} \) converges uniformly to \( f \).

**Proof.** Let us first suppose \( \{ f_n \} \) is Cauchy in the uniform norm. Let us define \( f \). Fix \( x \), then the sequence \( \{ f_n(x) \} \) is Cauchy because

\[
|f_m(x) - f_k(x)| \leq \| f_m - f_k \|_u.
\]

Thus \( \{ f_n(x) \} \) converges to some real number. Define \( f : S \to \mathbb{R} \) by

\[
f(x) := \lim_{n \to \infty} f_n(x).
\]

The sequence \( \{ f_n \} \) converges pointwise to \( f \). To show that the convergence is uniform, let \( \varepsilon > 0 \) be given. Find an \( N \) such that for all \( m, k \geq N \), we have \( \| f_m - f_k \|_u < \varepsilon/2 \). In other words, for all \( x \), we have \( |f_m(x) - f_k(x)| < \varepsilon/2 \). For any fixed \( x \), take the limit as \( k \) goes to infinity. Then \( |f_m(x) - f_k(x)| \) goes to \( |f_m(x) - f(x)| \). Consequently for all \( x \),

\[
|f_m(x) - f(x)| \leq \varepsilon/2 < \varepsilon.
\]

And hence \( \{ f_n \} \) converges uniformly.

Next, we prove the other direction. Suppose \( \{ f_n \} \) converges uniformly to \( f \). Given \( \varepsilon > 0 \), find \( N \) such that for all \( n \geq N \), we have \( |f_n(x) - f(x)| < \varepsilon/4 \) for all \( x \in S \). Therefore, for all \( m, k \geq N \),

\[
|f_m(x) - f_k(x)| = |f_m(x) - f(x) + f(x) - f_k(x)| \leq |f_m(x) - f(x)| + |f(x) - f_k(x)| < \varepsilon/4 + \varepsilon/4.
\]

Take supremum over all \( x \) to obtain

\[
\| f_m - f_k \|_u \leq \varepsilon/2 < \varepsilon.
\]
6.1.4 Exercises

Exercise 6.1.1: Let $f$ and $g$ be bounded functions on $[a, b]$. Prove

$$\|f + g\|_a \leq \|f\|_a + \|g\|_a.$$ 

Exercise 6.1.2:

a) Find the pointwise limit $\frac{e^{x/n}}{n}$ for $x \in \mathbb{R}$.

b) Is the limit uniform on $\mathbb{R}$?

c) Is the limit uniform on $[0, 1]$?

Exercise 6.1.3: Suppose $f_n : S \to \mathbb{R}$ are functions that converge uniformly to $f : S \to \mathbb{R}$. Suppose $A \subset S$. Show that the sequence of restrictions $\{f_n|_A\}$ converges uniformly to $f|_A$.

Exercise 6.1.4: Suppose $\{f_n\}$ and $\{g_n\}$ defined on some set $A$ converge to $f$ and $g$ respectively pointwise. Show that $\{f_n + g_n\}$ converges pointwise to $f + g$.

Exercise 6.1.5: Suppose $\{f_n\}$ and $\{g_n\}$ defined on some set $A$ converge to $f$ and $g$ respectively uniformly on $A$. Show that $\{f_n + g_n\}$ converges uniformly to $f + g$.

Exercise 6.1.6: Find an example of a sequence of functions $\{f_n\}$ and $\{g_n\}$ that converge uniformly to some $f$ and $g$ on some set $A$, but such that $\{f_ng_n\}$ (the multiple) does not converge uniformly to $fg$ on $A$. Hint: Let $A := \mathbb{R}$, let $f(x) := g(x) := x$. You can even pick $f_n = g_n$.

Exercise 6.1.7: Suppose there exists a sequence of functions $\{g_n\}$ uniformly converging to $0$ on $A$. Now suppose we have a sequence of functions $\{f_n\}$ and a function $f$ on $A$ such that

$$|f_n(x) - f(x)| \leq g_n(x)$$

for all $x \in A$. Show that $\{f_n\}$ converges uniformly to $f$ on $A$.

Exercise 6.1.8: Let $\{f_n\}$, $\{g_n\}$ and $\{h_n\}$ be sequences of functions on $[a, b]$. Suppose $\{f_n\}$ and $\{h_n\}$ converge uniformly to some function $f : [a, b] \to \mathbb{R}$ and suppose $f_n(x) \leq g_n(x) \leq h_n(x)$ for all $x \in [a, b]$. Show that $\{g_n\}$ converges uniformly to $f$.

Exercise 6.1.9: Let $f_n : [0, 1] \to \mathbb{R}$ be a sequence of increasing functions (that is, $f_n(x) \geq f_n(y)$ whenever $x \geq y$). Suppose $f_n(0) = 0$ and $\lim_{n \to \infty} f_n(1) = 0$. Show that $\{f_n\}$ converges uniformly to $0$.

Exercise 6.1.10: Let $\{f_n\}$ be a sequence of functions defined on $[0, 1]$. Suppose there exists a sequence of distinct numbers $x_n \in [0, 1]$ such that

$$f_n(x_n) = 1.$$ 

Prove or disprove the following statements:

a) True or false: There exists $\{f_n\}$ as above that converges to $0$ pointwise.

b) True or false: There exists $\{f_n\}$ as above that converges to $0$ uniformly on $[0, 1]$.

Exercise 6.1.11: Fix a continuous $h : [a, b] \to \mathbb{R}$. Let $f(x) := h(x)$ for $x \in [a, b]$, $f(x) := h(a)$ for $x < a$ and $f(x) := h(b)$ for all $x > b$. First show that $f : \mathbb{R} \to \mathbb{R}$ is continuous. Now let $f_n$ be the function $g$ from Exercise 5.3.7 with $\epsilon = 1/n$, defined on the interval $[a, b]$. That is,

$$f_n(x) := \frac{n}{2} \int_{x-1/n}^{x+1/n} f.$$ 

Show that $\{f_n\}$ converges uniformly to $h$ on $[a, b]$. 

Exercise 6.1.12: Prove that if a sequence of functions $f_n: S \to \mathbb{R}$ converge uniformly to a bounded function $f: S \to \mathbb{R}$, then there exists an $N$ such that for all $n \geq N$, the $f_n$ are bounded.

Exercise 6.1.13: Suppose there is a single constant $B$ and a sequence of functions $f_n: S \to \mathbb{R}$ that are bounded by $B$, that is $|f_n(x)| \leq B$ for all $x \in S$. Suppose that $\{f_n\}$ converges pointwise to $f: S \to \mathbb{R}$. Prove that $f$ is bounded.

Exercise 6.1.14 (requires §2.6): In Example 6.1.3 we saw $\sum_{k=0}^{\infty} x^k$ converges pointwise to $\frac{1}{1-x}$ on $(-1,1)$.

a) Show that whenever $0 \leq c < 1$, the series $\sum_{k=0}^{\infty} x^k$ converges uniformly on $[-c,c]$.

b) Show that the series $\sum_{k=0}^{\infty} x^k$ does not converge uniformly on $(-1,1)$. 


6.2 Interchange of limits

Note: 1–2.5 lectures, subsections on derivatives and power series (which requires §2.6) optional.

Large parts of modern analysis deal mainly with the question of the interchange of two limiting operations. When we have a chain of two limits, we cannot always just swap the limits. For instance,

\[ 0 = \lim_{n \to \infty} \left( \lim_{k \to \infty} \frac{n}{n + k} \right) \neq \lim_{k \to \infty} \left( \lim_{n \to \infty} \frac{n}{n + k} \right) = 1. \]

When talking about sequences of functions, interchange of limits comes up quite often. We look at several instances: continuity of the limit, the integral of the limit, the derivative of the limit, and the convergence of power series.

6.2.1 Continuity of the limit

If we have a sequence \( \{f_n\} \) of continuous functions, is the limit continuous? Suppose \( f \) is the (pointwise) limit of \( \{f_n\} \). If \( \lim x_k = x \), we are interested in the following interchange of limits, where the equality to prove is marked with a question mark. Equality is not always true, in fact, the limits to the left of the question mark might not even exist.

\[
\lim_{k \to \infty} f(x_k) = \lim_{k \to \infty} \left( \lim_{n \to \infty} f_n(x_k) \right) \overset{?}{=} \lim_{n \to \infty} \left( \lim_{k \to \infty} f_n(x_k) \right) = \lim_{n \to \infty} f_n(x) = f(x).
\]

We wish to find conditions on the sequence \( \{f_n\} \) so that the equation above holds. If we only require pointwise convergence, then the limit of a sequence of functions need not be continuous, and the equation above need not hold.

**Example 6.2.1:** Define \( f_n : [0, 1] \to \mathbb{R} \) as

\[
f_n(x) := \begin{cases} 
1 - nx & \text{if } x < 1/n, \\
0 & \text{if } x \geq 1/n.
\end{cases}
\]

See Figure 6.4.

Each function \( f_n \) is continuous. Fix an \( x \in (0, 1] \). If \( n \geq 1/x \), then \( x \geq 1/n \). Therefore for \( n \geq 1/x \), we have \( f_n(x) = 0 \), and so

\[
\lim_{n \to \infty} f_n(x) = 0.
\]

On the other hand, if \( x = 0 \), then

\[
\lim_{n \to \infty} f_n(0) = \lim_{n \to \infty} 1 = 1.
\]

Thus the pointwise limit of \( f_n \) is the function \( f : [0, 1] \to \mathbb{R} \) defined by

\[
f(x) := \begin{cases} 
1 & \text{if } x = 0, \\
0 & \text{if } x > 0.
\end{cases}
\]

The function \( f \) is not continuous at 0.
If we, however, require the convergence to be uniform, the limits can be interchanged.

**Theorem 6.2.2.** Suppose $S \subset \mathbb{R}$. Let $\{f_n\}$ be a sequence of continuous functions $f_n: S \to \mathbb{R}$ converging uniformly to $f: S \to \mathbb{R}$. Then $f$ is continuous.

**Proof.** Let $x \in S$ be fixed. Let $\{x_n\}$ be a sequence in $S$ converging to $x$.

Let $\varepsilon > 0$ be given. As $\{f_k\}$ converges uniformly to $f$, we find a $k \in \mathbb{N}$ such that

$$|f_k(y) - f(y)| < \varepsilon/3$$

for all $y \in S$. As $f_k$ is continuous at $x$, we find an $N \in \mathbb{N}$ such that for all $m \geq N$,

$$|f_k(x_m) - f_k(x)| < \varepsilon/3.$$

Thus for all $m \geq N$,

$$|f(x_m) - f(x)| = |f(x_m) - f_k(x_m) + f_k(x_m) - f_k(x) + f_k(x) - f(x)|$$

$$\leq |f_k(x_m) - f_k(x_m)| + |f_k(x_m) - f_k(x)| + |f_k(x) - f(x)|$$

$$< \varepsilon/3 + \varepsilon/3 + \varepsilon/3 = \varepsilon.$$

Therefore, $\{f(x_m)\}$ converges to $f(x)$ and $f$ is continuous at $x$. As $x$ was arbitrary, $f$ is continuous everywhere.

### 6.2.2 Integral of the limit

Again, if we simply require pointwise convergence, then the integral of a limit of a sequence of functions need not be equal to the limit of the integrals.

**Example 6.2.3:** Define $f_n: [0, 1] \to \mathbb{R}$ as

$$f_n(x) := \begin{cases} 
0 & \text{if } x = 0, \\
- n^2 x & \text{if } 0 < x < \frac{1}{n}, \\
0 & \text{if } x \geq \frac{1}{n}.
\end{cases}$$

See Figure 6.5.
Each $f_n$ is Riemann integrable (it is continuous on $(0, 1]$ and bounded), and the fundamental theorem of calculus says that

$$\int_0^1 f_n = \int_0^{1/n} (n - n^2 x) \, dx = \frac{1}{2}.$$  

Let us compute the pointwise limit of $\{f_n\}$. Fix an $x \in (0, 1]$. For $n \geq 1/x$, we have $x \geq 1/n$ and so $f_n(x) = 0$. Therefore,

$$\lim_{n \to \infty} f_n(x) = 0.$$  

We also have $f_n(0) = 0$ for all $n$. Therefore, the pointwise limit of $\{f_n\}$ is the zero function. Thus

$$\frac{1}{2} = \lim_{n \to \infty} \int_0^1 f_n(x) \, dx \neq \int_0^1 \left( \lim_{n \to \infty} f_n(x) \right) \, dx = \int_0^1 0 \, dx = 0.$$  

But if we require the convergence to be uniform, the limits can be interchanged.*

**Theorem 6.2.4.** Let $\{f_n\}$ be a sequence of Riemann integrable functions $f_n : [a, b] \to \mathbb{R}$ converging uniformly to $f : [a, b] \to \mathbb{R}$. Then $f$ is Riemann integrable, and

$$\int_a^b f = \lim_{n \to \infty} \int_a^b f_n.$$  

**Proof.** Let $\varepsilon > 0$ be given. As $f_n$ goes to $f$ uniformly, we find an $M \in \mathbb{N}$ such that for all $n \geq M$, we have $|f_n(x) - f(x)| < \frac{\varepsilon}{2(b - a)}$ for all $x \in [a, b]$. In particular, by reverse triangle inequality, $|f(x)| < \frac{\varepsilon}{2(b - a)} + |f_n(x)|$ for all $x$. Hence $f$ is bounded, as $f_n$ is bounded. Note that $f_n$ is integrable

*Weaker conditions are sufficient for this kind of theorem, but to prove such a generalization requires more sophisticated machinery than we cover here: the Lebesgue integral. In particular, the theorem holds with pointwise convergence as long as $f$ is integrable and there is an $M$ such that $\|f_n\|_u \leq M$ for all $n$. 
and compute
\[
\begin{align*}
\int_a^b f - \int_a^b f_n &= \int_a^b (f(x) - f_n(x) + f_n(x)) \, dx - \int_a^b (f(x) - f_n(x) + f_n(x)) \, dx \\
&\leq \int_a^b (f(x) - f_n(x)) \, dx + \int_a^b f_n(x) \, dx - \int_a^b (f(x) - f_n(x)) \, dx - \int_a^b f_n(x) \, dx \\
&= \int_a^b (f(x) - f_n(x)) \, dx + \int_a^b f_n(x) \, dx - \int_a^b (f(x) - f_n(x)) \, dx - \int_a^b f_n(x) \, dx \\
&= \int_a^b (f(x) - f_n(x)) \, dx - \int_a^b (f(x) - f_n(x)) \, dx \\
&\leq \frac{\varepsilon}{2(b-a)} (b-a) + \frac{\varepsilon}{2(b-a)} (b-a) = \varepsilon.
\end{align*}
\]

The first inequality is Proposition 5.2.5. The second inequality follows from Proposition 5.1.8 and the fact that for all \(x \in [a, b]\), we have \(\frac{-\varepsilon}{2(b-a)} < f(x) - f_n(x) < \frac{\varepsilon}{2(b-a)}\). As \(\varepsilon > 0\) was arbitrary, \(f\) is Riemann integrable.

Finally, we compute \(\int_a^b f\). We apply Proposition 5.1.10 in the calculation. Again, for all \(n \geq M\) (where \(M\) is the same as above),
\[
\left| \int_a^b f - \int_a^b f_n \right| = \left| \int_a^b (f(x) - f_n(x)) \, dx \right| \\
\leq \frac{\varepsilon}{2(b-a)} (b-a) = \frac{\varepsilon}{2} < \varepsilon.
\]
Therefore, \(\{\int_a^b f_n\}\) converges to \(\int_a^b f\). 

\[\square\]

**Example 6.2.5:** Suppose we wish to compute
\[
\lim_{n \to \infty} \int_0^1 \frac{nx + \sin(nx^2)}{n} \, dx.
\]
It is impossible to compute the integrals for any particular \(n\) using calculus as \(\sin(nx^2)\) has no closed-form antiderivative. However, we can compute the limit. We have shown before that \(\frac{nx + \sin(nx^2)}{n}\) converges uniformly on \([0, 1]\) to \(x\). By Theorem 6.2.4, the limit exists and
\[
\lim_{n \to \infty} \int_0^1 \frac{nx + \sin(nx^2)}{n} \, dx = \int_0^1 x \, dx = 1/2.
\]

**Example 6.2.6:** If convergence is only pointwise, the limit need not even be Riemann integrable. On \([0, 1]\) define
\[
f_n(x) := \begin{cases} 
1 & \text{if } x = p/q \text{ in lowest terms and } q \leq n, \\
0 & \text{otherwise.}
\end{cases}
\]
The function \(f_n\) differs from the zero function at finitely many points; there are only finitely many fractions in \([0, 1]\) with denominator less than or equal to \(n\). So \(f_n\) is integrable and \(\int_0^1 f_n = \int_0^1 0 = 0.\)
It is an easy exercise to show that \( \{f_n\} \) converges pointwise to the Dirichlet function

\[
f(x) := \begin{cases} 1 & \text{if } x \in \mathbb{Q}, \\ 0 & \text{otherwise}, \end{cases}
\]

which is not Riemann integrable.

**Example 6.2.7:** In fact, if the convergence is only pointwise, the limit of bounded functions is not even necessarily bounded. Define \( f_n : [0, 1] \to \mathbb{R} \) by

\[
f_n(x) := \begin{cases} 0 & \text{if } x < 1/n, \\ 1/x & \text{else.} \end{cases}
\]

For every \( n \) we get that \( |f_n(x)| \leq n \) for all \( x \in [0, 1] \) so the functions are bounded. However \( f_n \) converge pointwise to

\[
f(x) := \begin{cases} 0 & \text{if } x = 0, \\ 1/x & \text{else,} \end{cases}
\]

which is unbounded.

### 6.2.3 Derivative of the limit

While uniform convergence is enough to swap limits with integrals, it is not, however, enough to swap limits with derivatives, unless you also have uniform convergence of the derivatives themselves.

**Example 6.2.8:** Let \( f_n(x) := \frac{\sin(nx)}{n} \). Then \( f_n \) converges uniformly to 0. See Figure 6.6. The derivative of the limit is 0. But \( f'_n(x) = \cos(nx) \), which does not converge even pointwise, for example \( f'_n(\pi) = (-1)^n \). Furthermore, \( f'_n(0) = 1 \) for all \( n \), which does converge, but not to 0.

![Figure 6.6: Graphs of \( \frac{\sin(nx)}{n} \) for \( n = 1, 2, \ldots, 10 \), with higher \( n \) in lighter gray.](image)

**Example 6.2.9:** Let \( f_n(x) := \frac{1}{1+nx^2} \). If \( x \neq 0 \), then \( \lim_{n \to \infty} f_n(x) = 0 \), but \( \lim_{n \to \infty} f_n(0) = 1 \). Hence, \( \{f_n\} \) converges pointwise to a function that is not continuous at 0. We compute

\[
f'_n(x) = \frac{-2nx}{(1+nx^2)^2}.
\]
For every \( x \), \( \lim_{n \to \infty} f'_n(x) = 0 \), so the derivatives converge pointwise to 0, but the reader can check that the convergence is not uniform on any interval containing 0. The limit of \( f_n \) is not differentiable at 0—it is not even continuous at 0.

See the exercises for more examples. Using the fundamental theorem of calculus, we find an answer for continuously differentiable functions. The following theorem is true even if we do not assume continuity of the derivatives, but the proof is more difficult.

**Theorem 6.2.10.** Let \( I \) be a bounded interval and let \( f_n : I \to \mathbb{R} \) be continuously differentiable functions. Suppose \( \{f'_n\} \) converges uniformly to \( g : I \to \mathbb{R} \), and suppose \( \{f_n(c)\}_{n=1}^\infty \) is a convergent sequence for some \( c \in I \). Then \( \{f_n\} \) converges uniformly to a continuously differentiable function \( f : I \to \mathbb{R} \), and \( f' = g \).

**Proof.** Define \( f(c) := \lim_{n \to \infty} f_n(c) \). As \( f'_n \) are continuous and hence Riemann integrable, then via the fundamental theorem of calculus, we find that for \( x \in I \),

\[
 f_n(x) = f_n(c) + \int_c^x f'_n.
\]

As \( \{f'_n\} \) converges uniformly on \( I \), it converges uniformly on \([c,x]\) (or \([x,c]\) if \( x < c \)). Thus, the limit as \( n \to \infty \) on the right-hand side exists. Define \( f \) at the remaining points (where \( x \neq c \)) by this limit:

\[
 f(x) := \lim_{n \to \infty} f_n(c) + \lim_{n \to \infty} \int_c^x f'_n = f(c) + \int_c^x g.
\]

The function \( g \) is continuous, being the uniform limit of continuous functions. Hence \( f \) is differentiable and \( f'(x) = g(x) \) for all \( x \in I \) by the second form of the fundamental theorem.

It remains to prove uniform convergence. Suppose \( I \) has a lower bound \( a \) and upper bound \( b \). Let \( \varepsilon > 0 \) be given. Take \( M \) such that for all \( n \geq M \), we have \( |f(c) - f_n(c)| < \varepsilon/2 \) and \( |g(x) - f'_n(x)| < \varepsilon/(2(b-a)) \) for all \( x \in I \). Then

\[
 |f(x) - f_n(x)| = \left| \left( f(c) + \int_c^x g \right) - \left( f_n(c) + \int_c^x f'_n \right) \right| \\
 \leq |f(c) - f_n(c)| + \left| \int_c^x g - \int_c^x f'_n \right| \\
 = |f(c) - f_n(c)| + \left| \int_c^x (g(s) - f'_n(s)) \, ds \right| \\
 < \frac{\varepsilon}{2} + \frac{\varepsilon}{2(b-a)}(b-a) = \varepsilon.
\]

The proof goes through without boundedness of \( I \), except for the uniform convergence of \( f_n \) to \( f \). As an example suppose \( I = \mathbb{R} \) and let \( f_n(x) := x/n \). Then \( f'_n(x) = 1/n \), which converges uniformly to 0. However, \( \{f_n\} \) converges to 0 only pointwise.

### 6.2.4 Convergence of power series

In §2.6 we saw that a power series converges absolutely inside its radius of convergence, so it converges pointwise. Let us show that it (and all its derivatives) also converges uniformly. This fact allows us to swap several types of limits. Not only is the limit continuous, we can integrate and even differentiate convergent power series term by term.
Proposition 6.2.11. Let $\sum_{n=0}^{\infty} c_n(x-a)^n$ be a convergent power series with a radius of convergence $\rho$, where $0 < \rho \leq \infty$. Then the series converges uniformly in $[a-r,a+r]$ whenever $0 < r < \rho$.

In particular, the series defines a continuous function on $(a-\rho,a+\rho)$ (if $\rho < \infty$), or $\mathbb{R}$ (if $\rho = \infty$).

Proof. Let $I := (a-\rho,a+\rho)$ if $\rho < \infty$, or let $I := \mathbb{R}$ if $\rho = \infty$. Take $0 < r < \rho$. The series converges absolutely for every $x \in I$, in particular if $x = a + r$. So $\sum_{n=0}^{\infty} |c_n| r^n$ converges. Given $\varepsilon > 0$, find $M$ such that for all $k \geq M$,

$$\sum_{n=k+1}^{\infty} |c_n| r^n < \varepsilon.$$ 

For all $x \in [a-r,a+r]$ and all $m > k$,

$$\left| \sum_{n=0}^{m} c_n(x-a)^n - \sum_{n=0}^{k} c_n(x-a)^n \right| = \left| \sum_{n=k+1}^{m} c_n(x-a)^n \right| \leq \sum_{n=k+1}^{m} |c_n||x-a|^n \leq \sum_{n=k+1}^{m} |c_n| r^n \leq \sum_{n=k+1}^{\infty} |c_n| r^n < \varepsilon.$$ 

The partial sums are therefore uniformly Cauchy on $[a-r,a+r]$ and hence converge uniformly on that set.

Moreover, the partial sums are polynomials, which are continuous, and so their uniform limit on $[a-r,a+r]$ is a continuous function. As $r < \rho$ was arbitrary, the limit function is continuous on all of $I$.

As we said, we will show that power series can be differentiated and integrated term by term. The differentiated or integrated series is again a power series, and we will show it has the same radius of convergence. Therefore, any power series defines an infinitely differentiable function.

We first prove that we can antidifferentiate, as integration only needs uniform limits.

Corollary 6.2.12. Let $\sum_{n=0}^{\infty} c_n(x-a)^n$ be a convergent power series with a radius of convergence $0 < \rho \leq \infty$. Let $I := (a-\rho,a+\rho)$ if $\rho < \infty$ or $I := \mathbb{R}$ if $\rho = \infty$. Let $f : I \to \mathbb{R}$ be the limit. Then

$$\int_a^x f = \sum_{n=1}^{\infty} \frac{c_{n-1}}{n} (x-a)^n,$$

where the radius of convergence of this series is at least $\rho$.

Proof. Take $0 < r < \rho$. The partial sums $\sum_{n=0}^{k} c_n(x-a)^n$ converge uniformly on $[a-r,a+r]$. For every fixed $x \in [a-r,a+r]$, the convergence is also uniform on $[a,x]$ (or $[x,a]$ if $x < a$). Hence,

$$\int_a^x f = \int_a^x \lim_{k \to \infty} \sum_{n=0}^{k} c_n(s-a)^n \, ds = \lim_{k \to \infty} \int_a^x \sum_{n=0}^{k} c_n(s-a)^n \, ds = \lim_{k \to \infty} \sum_{n=1}^{k+1} \frac{c_{n-1}}{n} (x-a)^n.$$ 

Corollary 6.2.13. Let $\sum_{n=0}^{\infty} c_n(x-a)^n$ be a convergent power series with a radius of convergence $0 < \rho \leq \infty$. Let $I := (a-\rho,a+\rho)$ if $\rho < \infty$ or $I := \mathbb{R}$ if $\rho = \infty$. Let $f : I \to \mathbb{R}$ be the limit. Then $f$ is a differentiable function, and

$$f'(x) = \sum_{n=0}^{\infty} (n+1) c_{n+1} (x-a)^n,$$

where the radius of convergence of this series is $\rho$. 

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Proof. Take $0 < r < \rho$. The series converges uniformly on $[a - r, a + r]$, but we need uniform convergence of the derivative. Let

$$R := \limsup_{n \to \infty} |c_n|^{1/n}.$$

As the series is convergent $R < \infty$, and the radius of convergence is $1/R$ (or $\infty$ if $R = 0$).

Let $\varepsilon > 0$ be given. In Example 6.2.14, we saw $\lim n^{1/n} = 1$. Hence there exists an $N$ such that for all $n \geq N$, we have $n^{1/n} < 1 + \varepsilon$. So

$$R = \limsup_{n \to \infty} |c_n|^{1/n} \leq \limsup_{n \to \infty} |nc_n|^{1/n} \leq (1 + \varepsilon) \limsup_{n \to \infty} |c_n|^{1/n} = (1 + \varepsilon)R.$$

As $\varepsilon$ was arbitrary, $\limsup_{n \to \infty} |nc_n|^{1/n} = R$. Therefore, $\sum_{n=1}^\infty nc_n(x - a)^n$ has radius of convergence $\rho$. By dividing by $(x - a)$, we find $\sum_{n=1}^\infty (n+1)c_{n+1}(x - a)^n$ has radius of convergence $\rho$ as well.

Consequently, the partial sums $\sum_{n=0}^{r} c_n(x - a)^n$, which are derivatives of the partial sums $\sum_{n=0}^{r} c_n(x - a)^n$, converge uniformly on $[a - r, a + r]$. Furthermore, the series clearly converges at $x = a$. We may thus apply Theorem 6.2.10, and we are done as $r < \rho$ was arbitrary.

Example 6.2.14: We could have used this result to define the exponential function. That is, the power series

$$f(x) := \sum_{n=0}^\infty \frac{x^n}{n!}$$

has radius of convergence $\rho = \infty$. Furthermore, $f(0) = 1$, and by differentiating term by term, we find that $f'(x) = f(x)$.

Example 6.2.15: The series

$$\sum_{n=1}^\infty nx^n$$

converges to $\frac{x}{(1-x)^2}$ on $(-1, 1)$.

Proof: On $(-1, 1)$, $\sum_{n=0}^\infty x^n$ converges to $\frac{1}{1-x}$. The derivative $\sum_{n=1}^\infty nx^{n-1}$ then converges on the same interval to $\frac{1}{(1-x)^2}$. Multiplying by $x$ obtains the result.

6.2.5 Exercises

Exercise 6.2.1: Find an explicit example of a sequence of differentiable functions on $[-1, 1]$ that converge uniformly to a function $f$ such that $f$ is not differentiable. Hint: There are many possibilities, simplest is perhaps to combine $|x|$ and $\frac{2x}{n} + \frac{1}{m}$, another is to consider $\sqrt{x^2 + (1/n)^2}$. Show that these functions are differentiable, converge uniformly, and then show that the limit is not differentiable.

Exercise 6.2.2: Let $f_n(x) := \frac{x}{n}$. Show that $\{f_n\}$ converges uniformly to a differentiable function $f$ on $[0, 1]$ (find $f$). However, show that $f'(1) \neq \lim_{n \to \infty} f_n'(1)$.

Exercise 6.2.3: Let $f : [0, 1] \to \mathbb{R}$ be a Riemann integrable (hence bounded) function. Find $\lim_{n \to \infty} \int_0^1 \frac{f(x)}{n} \, dx$. 


Exercise 6.2.4: Show \( \lim_{n \to \infty} \int_1^2 e^{-nx^2} \, dx = 0 \). Feel free to use what you know about the exponential function from calculus.

Exercise 6.2.5: Find an example of a sequence of continuous functions on \((0, 1)\) that converges pointwise to a continuous function on \((0, 1)\), but the convergence is not uniform.

Note: In the previous exercise, \((0, 1)\) was picked for simplicity. For a more challenging exercise, replace \((0, 1)\) with \([0, 1]\).

Exercise 6.2.6: True/False; prove or find a counterexample to the following statement: If \( \{f_n\} \) is a sequence of everywhere discontinuous functions on \([0, 1]\) that converge uniformly to a function \(f\), then \(f\) is everywhere discontinuous.

Exercise 6.2.7: For a continuously differentiable function \(f: [a, b] \to \mathbb{R}\), define

\[
\|f\|_{C^1} := \|f\|_u + \|f'\|_u.
\]

Suppose \( \{f_n\} \) is a sequence of continuously differentiable functions such that for every \(\varepsilon > 0\), there exists an \(M\) such that for all \(n, k \geq M\), we have

\[
\|f_n - f_k\|_{C^1} < \varepsilon.
\]

Show that \( \{f_n\} \) converges uniformly to some continuously differentiable function \(f: [a, b] \to \mathbb{R}\).

Exercise 6.2.8: Suppose \( \{f_n\} \) is a sequence of Riemann integrable functions on \([0, 1]\) that converges uniformly to \(0\). Show that

\[
\lim_{n \to \infty} \|f_n\|_{L^1} = 0.
\]

Exercise 6.2.9: Find a sequence \( \{f_n\} \) of Riemann integrable functions on \([0, 1]\) converging pointwise to 0, but

\[
\lim_{n \to \infty} \|f_n\|_{L^1} \text{ does not exist (is } \infty).\]

Exercise 6.2.10 (Hard): Prove Dini’s theorem: Let \( f_n: [a, b] \to \mathbb{R} \) be a sequence of continuous functions such that

\[
0 \leq f_{n+1}(x) \leq f_n(x) \leq \cdots \leq f_1(x) \quad \text{for all } n \in \mathbb{N}.
\]

Suppose \( \{f_n\} \) converges pointwise to \(0\). Show that \( \{f_n\} \) converges to zero uniformly.

Exercise 6.2.11: Suppose \( f_n: [a, b] \to \mathbb{R} \) is a sequence of continuous functions that converges pointwise to a continuous \(f: [a, b] \to \mathbb{R}\). Suppose that for every \(x \in [a, b]\), the sequence \( \{|f_n(x) - f(x)|\} \) is monotone. Show that the sequence \( \{f_n\} \) converges uniformly.
Exercise 6.2.12: Find a sequence of Riemann integrable functions \( f_n : [0, 1] \to \mathbb{R} \) such that \( \{ f_n \} \) converges to zero pointwise, and such that

a) \( \{ \int_0^1 f_n \}_{n=1}^{\infty} \) increases without bound.

b) \( \{ \int_0^1 f_n \}_{n=1}^{\infty} \) is the sequence \(-1, 1, -1, 1, -1, 1, \ldots\).

It is possible to define a joint limit of a double sequence \( \{ x_{n,m} \} \) of real numbers (that is a function from \( \mathbb{N} \times \mathbb{N} \) to \( \mathbb{R} \)). We say \( L \) is the joint limit of \( \{ x_{n,m} \} \) and write

\[
\lim_{m \to \infty} x_{n,m} = L, \quad \text{or} \quad \lim_{(n,m) \to \infty} x_{n,m} = L,
\]

if for every \( \varepsilon > 0 \), there exists an \( M \) such that if \( n \geq M \) and \( m \geq M \), then \( |x_{n,m} - L| < \varepsilon \).

Exercise 6.2.13: Suppose the joint limit (see above) of \( \{ x_{n,m} \} \) is \( L \), and suppose that for all \( n \), \( \lim_{m \to \infty} x_{n,m} \) exists, and for all \( m \), \( \lim_{n \to \infty} x_{n,m} \) exists. Then show \( \lim_{m \to \infty} \lim_{n \to \infty} x_{n,m} = \lim_{n \to \infty} \lim_{m \to \infty} x_{n,m} = L \).

Exercise 6.2.14: A joint limit (see above) does not mean the iterated limits exist. Consider \( x_{n,m} := \frac{(-1)^{m+n}}{\min(n,m)} \).

a) Show that for no \( n \) does \( \lim_{m \to \infty} x_{n,m} \) exist, and for no \( m \) does \( \lim_{n \to \infty} x_{n,m} \) exist. So neither \( \lim_{m \to \infty} \lim_{n \to \infty} x_{n,m} \) nor \( \lim_{n \to \infty} \lim_{m \to \infty} x_{n,m} \) makes any sense at all.

b) Show that the joint limit of \( \{ x_{n,m} \} \) exists and equals 0.

Exercise 6.2.15: We say that a sequence of functions \( f_n : \mathbb{R} \to \mathbb{R} \) converges uniformly on compact subsets if for every \( k \in \mathbb{N} \), the sequence \( \{ f_n \} \) converges uniformly on \([-k,k] \).

a) Prove that if \( f_n : \mathbb{R} \to \mathbb{R} \) is a sequence of continuous functions converging uniformly on compact subsets, then the limit is continuous.

b) Prove that if \( f_n : \mathbb{R} \to \mathbb{R} \) is a sequence of Riemann integrable on every closed and bounded interval \([a,b] \), and converging uniformly on compact subsets to an \( f : \mathbb{R} \to \mathbb{R} \), then for every interval \([a,b] \), we have \( f \in \mathcal{R}[a,b] \), and \( \int_a^b f = \lim_{n \to \infty} \int_a^b f_n \).

Exercise 6.2.16 (Challenging): Find a sequence of continuous functions \( f_n : [0,1] \to \mathbb{R} \) that converge to the popcorn function \( f : [0,1] \to \mathbb{R} \), that is the function such that \( f(p/q) := 1/q \) (if \( p/q \) is in lowest terms) and \( f(x) := 0 \) if \( x \) is not rational (note that \( f(0) = f(1) = 1 \)), see Example 3.2.12. So a pointwise limit of continuous functions can have a dense set of discontinuities. See also the next exercise.

Exercise 6.2.17 (Challenging): The Dirichlet function \( f : [0,1] \to \mathbb{R} \), that is the function such that \( f(x) := 1 \) if \( x \in \mathbb{Q} \) and \( f(x) := 0 \) if \( x \notin \mathbb{Q} \), is not the pointwise limit of continuous functions, although this is difficult to show. Prove, however, that \( f \) is a pointwise limit of functions that are themselves pointwise limits of continuous functions themselves.

Exercise 6.2.18:

a) Find a sequence of Lipschitz continuous functions on \([0,1]\) whose uniform limit is \( \sqrt{x} \), which is a non-Lipschitz function.

b) On the other hand, show that if \( f_n : S \to \mathbb{R} \) are Lipschitz with a uniform constant \( K \) (meaning all of them satisfy the definition with the same constant) and \( \{ f_n \} \) converges pointwise to \( f : S \to \mathbb{R} \), then the limit \( f \) is a Lipschitz continuous function with Lipschitz constant \( K \).
CHAPTER 6. SEQUENCES OF FUNCTIONS

Exercise 6.2.19 (requires §2.6): If \( \sum_{n=0}^{\infty} c_n(x-a)^n \) has radius of convergence \( \rho \), show that the term by term integral \( \sum_{n=1}^{\infty} \frac{c_{n-1}}{n} (x-a)^n \) has radius of convergence \( \rho \). Note that we only proved above that the radius of convergence was at least \( \rho \).

Exercise 6.2.20 (requires §2.6 and §4.3): Suppose \( f(x) := \sum_{n=0}^{\infty} c_n(x-a)^n \) converges in \((a-\rho, a+\rho)\).

a) Suppose that \( f^{(k)}(a) = 0 \) for all \( k = 0, 1, 2, 3, \ldots \). Prove that \( c_n = 0 \) for all \( n \), or in other words, \( f(x) = 0 \) for all \( x \in (a-\rho, a+\rho) \).

b) Using part a) prove a version of the so-called “identity theorem for analytic functions”: If there exists an \( \varepsilon > 0 \) such that \( f(x) = 0 \) for all \( x \in (a-\varepsilon, a+\varepsilon) \), then \( f(x) = 0 \) for all \( x \in (a-\rho, a+\rho) \).

Exercise 6.2.21: Let \( f_n(x) := \frac{x}{1+(nx)} \). Notice that \( f_n \) are differentiable functions.

a) Show that \( \{f_n\} \) converges uniformly to 0.

b) Show that \( |f'_n(x)| \leq 1 \) for all \( x \) and all \( n \).

c) Show that \( \{f'_n\} \) converges pointwise to a function discontinuous at the origin.

d) Let \( \{a_n\} \) be an enumeration of the rational numbers. Define

\[
g_n(x) := \sum_{k=1}^{n} 2^{-k} f_n(x-a_k).\]

Show that \( \{g_n\} \) converges uniformly to 0.

e) Show that \( \{g'_n\} \) converges pointwise to a function \( \psi \) that is discontinuous at every rational number and continuous at every irrational number. In particular, \( \lim_{n \to \infty} g'_n(x) \neq 0 \) for every rational number \( x \).
6.3 Picard’s theorem

Note: 1–2 lectures (can be safely skipped)

A first semester course in analysis should have a pièce de résistance caliber theorem. We pick a theorem whose proof combines everything we have learned. It is more sophisticated than the fundamental theorem of calculus, the first highlight theorem of this course. The theorem we are talking about is Picard’s theorem* on existence and uniqueness of a solution to an ordinary differential equation. Both the statement and the proof are beautiful examples of what one can do with the material we mastered so far. It is also a good example of how analysis is applied as differential equations are indispensable in science of every stripe.

6.3.1 First order ordinary differential equation

Modern science is described in the language of differential equations. That is, equations involving not only the unknown, but also its derivatives. The simplest nontrivial form of a differential equation is the so-called first order ordinary differential equation

\[ y' = F(x, y). \]

Generally, we also specify an initial condition \( y(x_0) = y_0 \). The solution of the equation is a function \( y(x) \) such that \( y(x_0) = y_0 \) and \( y'(x) = F(x, y(x)) \).

When \( F \) involves only the \( x \) variable, the solution is given by the fundamental theorem of calculus. On the other hand, when \( F \) depends on both \( x \) and \( y \) we need far more firepower. It is not always true that a solution exists, and if it does, that it is the unique solution. Picard’s theorem gives us certain sufficient conditions for existence and uniqueness.

6.3.2 The theorem

We need a definition of continuity in two variables. A point in the plane \( \mathbb{R}^2 = \mathbb{R} \times \mathbb{R} \) is denoted by an ordered pair \((x, y)\). For simplicity, we give the following sequential definition of continuity.

**Definition 6.3.1.** Let \( U \subset \mathbb{R}^2 \) be a set, \( F: U \to \mathbb{R} \) a function, and \((x, y) \in U \) a point. The function \( F \) is continuous at \((x, y)\) if for every sequence \( \{(x_n, y_n)\}_{n=1}^\infty \) of points in \( U \) such that \( \lim x_n = x \) and \( \lim y_n = y \), we have

\[ \lim_{n \to \infty} F(x_n, y_n) = F(x, y). \]

We say \( F \) is continuous if it is continuous at all points in \( U \).

**Theorem 6.3.2** (Picard’s theorem on existence and uniqueness). Let \( I, J \subset \mathbb{R} \) be closed bounded intervals, let \( I^\circ \) and \( J^\circ \) be their interiors†, and let \((x_0, y_0) \in I^\circ \times J^\circ \). Suppose \( F: I \times J \to \mathbb{R} \) is continuous and Lipschitz in the second variable, that is, there exists an \( L \in \mathbb{R} \) such that

\[ |F(x, y) - F(x, z)| \leq L |y - z| \quad \text{for all } y, z \in J, x \in I. \]

Then there exists an \( h > 0 \) and a unique differentiable function \( f: [x_0 - h, x_0 + h] \to J \subset \mathbb{R} \) such that

\[ f'(x) = F(x, f(x)) \quad \text{and} \quad f(x_0) = y_0. \quad (6.1) \]

†Named for the French mathematician Charles Émile Picard (1856–1941).

‡By interior of \([a, b]\) we mean \((a, b)\).
Proof. Suppose we could find a solution $f$. Using the fundamental theorem of calculus we integrate the equation $f'(x) = F(x, f(x))$, $f(x_0) = y_0$, and write (6.1) as the integral equation

$$f(x) = y_0 + \int_{x_0}^{x} F(t, f(t)) \, dt. \tag{6.2}$$

The idea of our proof is that we try to plug in approximations to a solution to the right-hand side of (6.2) to get better approximations on the left-hand side of (6.2). We hope that in the end the sequence converges and solves (6.2) and hence (6.1). The technique below is called Picard iteration, and the individual functions $f_k$ are called the Picard iterates.

Without loss of generality, suppose $x_0 = 0$ (exercise below). Another exercise tells us that $F$ is bounded as it is continuous. Therefore pick some $M > 0$ so that $|F(x, y)| \leq M$ for all $(x, y) \in I \times J$. Pick $\alpha > 0$ such that $[-\alpha, \alpha] \subset I$ and $[y_0 - \alpha, y_0 + \alpha] \subset J$. Define

$$h := \min \left\{ \alpha, \frac{\alpha}{M + L\alpha} \right\}.$$

Observe $[-h, h] \subset I$.

Set $f_0(x) := y_0$. We define $f_k$ inductively. Assuming $f_{k-1}([-h, h]) \subset [y_0 - \alpha, y_0 + \alpha]$, we see $F(t, f_{k-1}(t))$ is a well-defined function of $t$ for $t \in [-h, h]$. Further if $f_{k-1}$ is continuous on $[-h, h]$, then $F(t, f_{k-1}(t))$ is continuous as a function of $t$ on $[-h, h]$ (left as an exercise). Define

$$f_k(x) := y_0 + \int_{0}^{x} F(t, f_{k-1}(t)) \, dt,$$

and $f_k$ is continuous on $[-h, h]$ by the fundamental theorem of calculus. To see that $f_k$ maps $[-h, h]$ to $[y_0 - \alpha, y_0 + \alpha]$, we compute for $x \in [-h, h]$

$$|f_k(x) - y_0| = \left| \int_{0}^{x} F(t, f_{k-1}(t)) \, dt \right| \leq M|x| \leq Mh \leq M \frac{\alpha}{M + L\alpha} \leq \alpha.$$

We now define $f_{k+1}$ and so on, and we have defined a sequence $\{f_k\}$ of functions. We need to show that it converges to a function $f$ that solves the equation (6.2) and therefore (6.1).

We wish to show that the sequence $\{f_k\}$ converges uniformly to some function on $[-h, h]$. First, for $t \in [-h, h]$, we have the following bound

$$|F(t, f_n(t)) - F(t, f_k(t))| \leq L |f_n(t) - f_k(t)| \leq L \|f_n - f_k\|_u,$$

where $\|f_n - f_k\|_u$ is the uniform norm, that is the supremum of $|f_n(t) - f_k(t)|$ for $t \in [-h, h]$. Now note that $|x| \leq h \leq \frac{\alpha}{M + L\alpha}$. Therefore

$$|f_n(x) - f_k(x)| = \left| \int_{0}^{x} F(t, f_{n-1}(t)) \, dt - \int_{0}^{x} F(t, f_{k-1}(t)) \, dt \right|$$

$$= \left| \int_{0}^{x} F(t, f_{n-1}(t)) - F(t, f_{k-1}(t)) \, dt \right|$$

$$\leq L \|f_{n-1} - f_{k-1}\|_u |x|$$

$$\leq \frac{L\alpha}{M + L\alpha} \|f_{n-1} - f_{k-1}\|_u.$$
Let $C := \frac{L\alpha}{M + L\alpha}$ and note that $C < 1$. Taking supremum on the left-hand side we get

$$\|f_n - f_k\|_u \leq C \|f_{n-1} - f_{k-1}\|_u.$$ 

Without loss of generality, suppose $n \geq k$. Then by induction we can show

$$\|f_n - f_k\|_u \leq C^k \|f_{n-k} - f_0\|_u.$$ 

For $x \in [-h, h]$, we have

$$|f_{n-k}(x) - f_0(x)| = |f_{n-k}(x) - y_0| \leq \alpha.$$ 

Therefore,

$$\|f_n - f_k\|_u \leq C^k \|f_{n-k} - f_0\|_u \leq C^k \alpha.$$ 

As $C < 1$, $\{f_n\}$ is uniformly Cauchy and by Proposition 6.1.13 we obtain that $\{f_n\}$ converges uniformly on $[-h, h]$ to some function $f : [-h, h] \to \mathbb{R}$. The function $f$ is the uniform limit of continuous functions and therefore continuous. Furthermore, since $f_n([-h, h]) \subset \{y_0 - \alpha, y_0 + \alpha\}$ for all $n$, then $f([-h, h]) \subset \{y_0 - \alpha, y_0 + \alpha\}$ (why?).

We now need to show that $f$ solves (6.2). First, as before we notice

$$|F(t, f_n(t)) - F(t, f(t))| \leq L |f_n(t) - f(t)| \leq L \|f_n - f\|_u.$$ 

As $\|f_n - f\|_u$ converges to 0, then $F(t, f_n(t))$ converges uniformly to $F(t, f(t))$ for $t \in [-h, h]$. Hence, for $x \in [-h, h]$ the convergence is uniform for $t \in [0, x]$ (or $[x, 0]$ if $x < 0$). Therefore,

$$y_0 + \int_0^x F(t, f(t)) \, dt = y_0 + \int_0^x F(t, \lim_{n \to \infty} f_n(t)) \, dt$$

$$= y_0 + \int_0^x \lim_{n \to \infty} F(t, f_n(t)) \, dt$$ (by continuity of $F$)

$$= \lim_{n \to \infty} \left( y_0 + \int_0^x F(t, f_n(t)) \, dt \right)$$ (by uniform convergence)

$$= \lim_{n \to \infty} f_{n+1}(x) = f(x).$$

We apply the fundamental theorem of calculus (Theorem 5.3.3) to show that $f$ is differentiable and its derivative is $F(x, f(x))$. It is obvious that $f(0) = y_0$.

Finally, what is left to do is to show uniqueness. Suppose $g : [-h, h] \to J \subset \mathbb{R}$ is another solution. As before we use the fact that $|F(t, f(t)) - F(t, g(t))| \leq L \|f - g\|_u$. Then

$$|f(x) - g(x)| = \left| y_0 + \int_0^x F(t, f(t)) \, dt - \left( y_0 + \int_0^x F(t, g(t)) \, dt \right) \right|$$

$$= \left| \int_0^x F(t, f(t)) - F(t, g(t)) \, dt \right|$$

$$\leq L \|f - g\|_u |x| \leq Lh \|f - g\|_u \leq \frac{L\alpha}{M + L\alpha} \|f - g\|_u.$$ 

As before, $C = \frac{L\alpha}{M + L\alpha} < 1$. By taking supremum over $x \in [-h, h]$ on the left-hand side we obtain

$$\|f - g\|_u \leq C \|f - g\|_u.$$ 

This is only possible if $\|f - g\|_u = 0$. Therefore, $f = g$, and the solution is unique. \qed
6.3.3 Examples

Let us look at some examples. The proof of the theorem gives us an explicit way to find an $h$ that works. It does not, however, give us the best $h$. It is often possible to find a much larger $h$ for which the conclusion of the theorem holds.

The proof also gives us the Picard iterates as approximations to the solution. So the proof actually tells us how to obtain the solution, not just that the solution exists.

Example 6.3.3: Consider

\[ f'(x) = f(x), \quad f(0) = 1. \]

That is, we suppose $F(x, y) = y$, and we are looking for a function $f$ such that $f'(x) = f(x)$. Let us forget for the moment that we solved this equation in §5.4.

We pick any $I$ that contains 0 in the interior. We pick an arbitrary $J$ that contains 1 in its interior. We can use $L = 1$. The theorem guarantees an $h > 0$ such that there exists a unique solution $f: [-h, h] \to \mathbb{R}$. This solution is usually denoted by

\[ e^x := f(x). \]

We leave it to the reader to verify that by picking $I$ and $J$ large enough the proof of the theorem guarantees that we are able to pick $\alpha$ such that we get any $h$ we want as long as $h < 1/2$. We omit the calculation.

Of course, we know this function exists as a function for all $x$, so an arbitrary $h$ ought to work. By same reasoning as above, no matter what $x_0$ and $y_0$ are, the proof guarantees an arbitrary $h$ as long as $h < 1/2$. Fix such an $h$. We get a unique function defined on $[x_0 - h, x_0 + h]$. After defining the function on $[-h, h]$ we find a solution on the interval $[0, 2h]$ and notice that the two functions must coincide on $[0, h]$ by uniqueness. We thus iteratively construct the exponential for all $x \in \mathbb{R}$. Therefore Picard’s theorem could be used to prove the existence and uniqueness of the exponential.

Let us compute the Picard iterates. We start with the constant function $f_0(x) := 1$. Then

\[
\begin{align*}
    f_1(x) &= 1 + \int_0^x f_0(s) \, ds = 1 + x, \\
    f_2(x) &= 1 + \int_0^x f_1(s) \, ds = 1 + \int_0^x (1 + s) \, ds = 1 + x + \frac{x^2}{2}, \\
    f_3(x) &= 1 + \int_0^x f_2(s) \, ds = 1 + \int_0^x \left( 1 + s + \frac{s^2}{2} \right) \, ds = 1 + x + \frac{x^2}{2} + \frac{x^3}{6}.
\end{align*}
\]

We recognize the beginning of the Taylor series for the exponential.

Example 6.3.4: Consider the equation

\[ f'(x) = (f(x))^2 \quad \text{and} \quad f(0) = 1. \]

From elementary differential equations we know

\[ f(x) = \frac{1}{1 - x}. \]
is the solution. The solution is only defined on \((-\infty, 1)\). That is, we are able to use \(h < 1\), but never a larger \(h\). The function that takes \(y\) to \(y^2\) is not Lipschitz as a function on all of \(\mathbb{R}\). As we approach \(x = 1\) from the left, the solution becomes larger and larger. The derivative of the solution grows as \(y^2\), and so the \(L\) required has to be larger and larger as \(y_0\) grows. If we apply the theorem with \(x_0\) close to 1 and \(y_0 = \frac{1}{1-x_0}\) we find that the \(h\) that the proof guarantees is smaller and smaller as \(x_0\) approaches 1.

The \(h\) from the proof is not the best \(h\). By picking \(\alpha\) correctly, the proof of the theorem guarantees \(h = 1 - \sqrt{3}/2 \approx 0.134\) (we omit the calculation) for \(x_0 = 0\) and \(y_0 = 1\), even though we saw above that any \(h < 1\) should work.

**Example 6.3.5:** Consider the equation

\[
f'(x) = 2\sqrt{|f(x)|}, \quad f(0) = 0.
\]

The function \(F(x, y) = 2\sqrt{|y|}\) is continuous, but not Lipschitz in \(y\) (why?). The equation does not satisfy the hypotheses of the theorem. The function

\[
f(x) = \begin{cases} 
  x^2 & \text{if } x \geq 0, \\
  -x^2 & \text{if } x < 0,
\end{cases}
\]

is a solution, but \(f(x) = 0\) is also a solution. A solution exists, but is not unique.

**Example 6.3.6:** Consider \(y' = \phi(x)\) where \(\phi(x) := 0\) if \(x \in \mathbb{Q}\) and \(\phi(x) := 1\) if \(x \not\in \mathbb{Q}\). In other words, the \(F(x, y) = \phi(x)\) is discontinuous. The equation has no solution regardless of the initial conditions. A solution would have derivative \(\phi\), but \(\phi\) does not have the intermediate value property at any point (why?). No solution exists by Darboux’s theorem.

The examples show that without the Lipschitz condition, a solution might exist but not be a unique, and without continuity of \(F\), we may not have a solution at all. It is in fact a theorem, the Peano existence theorem, that if \(F\) is continuous a solution exists (but may not be unique).

**Remark 6.3.7.** It is possible to weaken what we mean by “solution to \(y' = F(x, y)\)” by focusing on the integral equation \(f(x) = y_0 + \int_0^x F(t, f(t)) \, dt\). For example, let \(H\) be the Heaviside function*, that is \(H(t) := 0\) for \(t < 0\) and \(H(t) := 1\) for \(t \geq 0\). Then \(y' = H(t), y(0) = 0\), is a common equation. The “solution” is the ramp function \(f(x) := 0\) if \(x < 0\) and \(f(x) := x\) if \(x \geq 0\), since this function satisfies \(f(x) = \int_0^x H(t) \, dt\). Notice, however, that \(f'(0)\) does not exist, so \(f\) is only a so-called weak solution to the differential equation.

### 6.3.4 Exercises

**Exercise 6.3.1:** Let \(I, J \subset \mathbb{R}\) be intervals. Let \(F : I \times J \to \mathbb{R}\) be a continuous function of two variables and suppose \(f : I \to J\) be a continuous function. Show that \(F(x, f(x))\) is a continuous function on \(I\).

**Exercise 6.3.2:** Let \(I, J \subset \mathbb{R}\) be closed bounded intervals. Show that if \(F : I \times J \to \mathbb{R}\) is continuous, then \(F\) is bounded.

---

*Named for the English engineer, mathematician, and physicist Oliver Heaviside (1850–1825).
**Exercise 6.3.3:** We proved Picard’s theorem under the assumption that \( x_0 = 0 \). Prove the full statement of Picard’s theorem for an arbitrary \( x_0 \).

**Exercise 6.3.4:** Let \( f'(x) = xf(x) \) be our equation. Start with the initial condition \( f(0) = 2 \) and find the Picard iterates \( f_0, f_1, f_2, f_3, f_4 \).

**Exercise 6.3.5:** Suppose \( F : I \times J \to \mathbb{R} \) is a function that is continuous in the first variable, that is, for every fixed \( y \) the function that takes \( x \) to \( F(x,y) \) is continuous. Further, suppose \( F \) is Lipschitz in the second variable, that is, there exists a number \( L \) such that

\[
|F(x,y) - F(x,z)| \leq L|y - z| \quad \text{for all } y, z \in J, x \in I.
\]

Show that \( F \) is continuous as a function of two variables. Therefore, the hypotheses in the theorem could be made even weaker.

**Exercise 6.3.6:** A common type of equation one encounters are linear first order differential equations, that is equations of the form

\[
y' + p(x)y = q(x), \quad y(x_0) = y_0.
\]

Prove Picard’s theorem for linear equations. Suppose \( I \) is an interval, \( x_0 \in I \), and \( p : I \to \mathbb{R} \) and \( q : I \to \mathbb{R} \) are continuous. Show that there exists a unique differentiable \( f : I \to \mathbb{R} \), such that \( y = f(x) \) satisfies the equation and the initial condition. Hint: Assume existence of the exponential function and use the integrating factor formula for existence of \( f \) (prove that it works and then that it is unique):

\[
f(x) := e^{- \int_{x_0}^{x} p(s) \, ds} \left( \int_{x_0}^{x} e^{\int_{s_0}^{s} p(t) \, dt} q(t) \, dt + y_0 \right).
\]

**Exercise 6.3.7:** Consider the equation \( f'(x) = f(x) \), from Example 6.3.3. Show that given any \( x_0 \), any \( y_0 \), and any positive \( h < \frac{1}{2} \), we can pick \( \alpha > 0 \) large enough that the proof of Picard’s theorem guarantees a solution for the initial condition \( f(x_0) = y_0 \) in the interval \([x_0 - h, x_0 + h]\).

**Exercise 6.3.8:** Consider the equation \( y' = y^{1/3} \).

a) Show that for the initial condition \( y(1) = 1 \), Picard’s theorem applies. Find an \( \alpha > 0 \), \( M \), \( L \), and \( h \) that would work in the proof.

b) Show that for the initial condition \( y(1) = 0 \), Picard’s theorem does not apply.

c) Find a solution for \( y(1) = 0 \) anyway.

**Exercise 6.3.9:** Consider the equation \( xy' = 2y \).

a) Show that \( y = Cx^2 \) is a solution for every constant \( C \).

b) Show that for every \( x_0 \neq 0 \) and every \( y_0 \), Picard’s theorem applies for the initial condition \( y(x_0) = y_0 \).

c) Show that \( y(0) = y_0 \) is solvable if and only if \( y_0 = 0 \).
Chapter 7
Metric Spaces

7.1 Metric spaces

Note: 1.5 lectures

As mentioned in the introduction, the main idea in analysis is to take limits. In chapter 2 we learned to take limits of sequences of real numbers. And in chapter 3 we learned to take limits of functions as a real number approached some other real number.

We want to take limits in more complicated contexts. For example, we want to have sequences of points in 3-dimensional space. We wish to define continuous functions of several variables. We even want to define functions on spaces that are a little harder to describe, such as the surface of the earth. We still want to talk about limits there.

Finally, we have seen the limit of a sequence of functions in chapter 6. We wish to unify all these notions so that we do not have to reprove theorems over and over again in each context. The concept of a metric space is an elementary yet powerful tool in analysis. And while it is not sufficient to describe every type of limit we find in modern analysis, it gets us very far indeed.

Definition 7.1.1. Let $X$ be a set, and let $d : X \times X \to \mathbb{R}$ be a function such that for all $x, y, z \in X$

(i) $d(x, y) \geq 0$. \hfill (nonnegativity)
(ii) $d(x, y) = 0$ if and only if $x = y$. \hfill (identity of indiscernibles)
(iii) $d(x, y) = d(y, x)$. \hfill (symmetry)
(iv) $d(x, z) \leq d(x, y) + d(y, z)$. \hfill (triangle inequality)

The pair $(X, d)$ is called a metric space. The function $d$ is called the metric or the distance function. Sometimes we write just $X$ as the metric space instead of $(X, d)$ if the metric is clear from context.

The geometric idea is that $d$ is the distance between two points. Items (i)–(iii) have obvious geometric interpretation: Distance is always nonnegative, the only point that is distance 0 away from $x$ is $x$ itself, and finally that the distance from $x$ to $y$ is the same as the distance from $y$ to $x$. The triangle inequality (iv) has the interpretation given in Figure 7.1.

For the purposes of drawing, it is convenient to draw figures and diagrams in the plane with the metric being the euclidean distance. However, that is only one particular metric space. Just because a certain fact seems to be clear from drawing a picture does not mean it is true in every metric space.
You might be getting sidetracked by intuition from euclidean geometry, whereas the concept of a metric space is a lot more general.

Let us give some examples of metric spaces.

**Example 7.1.2:** The set of real numbers $\mathbb{R}$ is a metric space with the metric

$$d(x, y) := |x - y|.$$ 

Items (i)–(iii) of the definition are easy to verify. The triangle inequality (iv) follows immediately from the standard triangle inequality for real numbers:

$$d(x, z) = |x - z| = |x - y + y - z| \leq |x - y| + |y - z| = d(x, y) + d(y, z).$$

This metric is the *standard metric on $\mathbb{R}$*. If we talk about $\mathbb{R}$ as a metric space without mentioning a specific metric, we mean this particular metric.

**Example 7.1.3:** We can also put a different metric on the set of real numbers. For example, take the set of real numbers $\mathbb{R}$ together with the metric

$$d(x, y) := \frac{|x - y|}{|x - y| + 1}.$$ 

Items (i)–(iii) are again easy to verify. The triangle inequality (iv) is a little bit more difficult. Note that $d(x, y) = \varphi(|x - y|)$ where $\varphi(t) = \frac{t}{t+1}$ and $\varphi$ is an increasing function (positive derivative). Hence

$$d(x, z) = \varphi(|x - z|)$$

$$= \varphi(|x - y + y - z|)$$

$$\leq \varphi(|x - y| + |y - z|)$$

$$= \frac{|x - y| + |y - z|}{|x - y| + |y - z| + 1}$$

$$= \frac{|x - y|}{|x - y| + 1} + \frac{|y - z|}{|y - z| + 1} + \frac{|y - z|}{|x - y| + 1}$$

$$\leq \frac{|x - y|}{|x - y| + 1} + \frac{|y - z|}{|y - z| + 1} = d(x, y) + d(y, z).$$
The function $d$ is thus a metric, and we have an example of a nonstandard metric on $\mathbb{R}$. With this metric, $d(x,y) < 1$ for all $x, y \in \mathbb{R}$. That is, every two points are less than 1 unit apart.

An important metric space is the $n$-dimensional euclidean space $\mathbb{R}^n = \mathbb{R} \times \mathbb{R} \times \cdots \times \mathbb{R}$. We use the following notation for points: $x = (x_1, x_2, \ldots, x_n) \in \mathbb{R}^n$. We will not write $\vec{x}$ nor $x$ for a vector, we simply give it a name such as $x$ and we will remember that $x$ is a vector. We also write simply $0 \in \mathbb{R}^n$ to mean the point $(0, 0, \ldots, 0)$. Before making $\mathbb{R}^n$ a metric space, we prove an important inequality, the so-called Cauchy–Schwarz inequality.

**Lemma 7.1.4 (Cauchy–Schwarz inequality*)**. If $x = (x_1, x_2, \ldots, x_n) \in \mathbb{R}^n$, $y = (y_1, y_2, \ldots, y_n) \in \mathbb{R}^n$, then

$$\left(\sum_{j=1}^{n} x_j y_j\right)^2 \leq \left(\sum_{j=1}^{n} x_j^2\right)\left(\sum_{j=1}^{n} y_j^2\right).$$

**Proof.** Any square of a real number is nonnegative. Hence any sum of squares is nonnegative:

$$0 \leq \sum_{j=1}^{n} \sum_{k=1}^{n} (x_j y_k - x_k y_j)^2 = \sum_{j=1}^{n} \sum_{k=1}^{n} (x_j^2 y_k^2 + x_k^2 y_j^2 - 2x_j x_k y_j y_k) = \left(\sum_{j=1}^{n} x_j^2\right)\left(\sum_{j=1}^{n} y_j^2\right) + \left(\sum_{j=1}^{n} y_j^2\right)\left(\sum_{j=1}^{n} x_j^2\right) - 2\left(\sum_{j=1}^{n} x_j y_j\right)\left(\sum_{j=1}^{n} x_j y_j\right).$$

We relabel and divide by 2 to obtain

$$0 \leq \left(\sum_{j=1}^{n} x_j^2\right)\left(\sum_{j=1}^{n} y_j^2\right) - \left(\sum_{j=1}^{n} x_j y_j\right)^2,$$

which is precisely what we wanted. 

**Example 7.1.5:** Let us construct the standard metric for $\mathbb{R}^n$. Define

$$d(x,y) := \sqrt{(x_1 - y_1)^2 + (x_2 - y_2)^2 + \cdots + (x_n - y_n)^2} = \sqrt{\sum_{j=1}^{n} (x_j - y_j)^2}.$$  

For $n = 1$, the real line, this metric agrees with what we defined above. For $n > 1$, the only tricky part of the definition to check, as before, is the triangle inequality. It is less messy to work with the

*Sometimes it is called the Cauchy–Bunyakovsky–Schwarz inequality. Karl Hermann Amandus Schwarz (1843–1921) was a German mathematician and Viktor Yakovlevich Bunyakovsky (1804–1889) was a Ukrainian mathematician. What we stated should really be called the Cauchy inequality, as Bunyakovsky and Schwarz provided proofs for infinite-dimensional versions.*
square of the metric. In the following estimate, note the use of the Cauchy–Schwarz inequality.

\[ (d(x, z))^2 = \sum_{j=1}^{n} (x_j - z_j)^2 \]

\[ = \sum_{j=1}^{n} (x_j - y_j + y_j - z_j)^2 \]

\[ = \sum_{j=1}^{n} \left( (x_j - y_j)^2 + (y_j - z_j)^2 + 2(x_j - y_j)(y_j - z_j) \right) \]

\[ = \sum_{j=1}^{n} (x_j - y_j)^2 + \sum_{j=1}^{n} (y_j - z_j)^2 + 2 \sum_{j=1}^{n} (x_j - y_j)(y_j - z_j) \]

\[ \leq \sum_{j=1}^{n} (x_j - y_j)^2 + \sum_{j=1}^{n} (y_j - z_j)^2 + 2 \sqrt{\sum_{j=1}^{n} (x_j - y_j)^2 \sum_{j=1}^{n} (y_j - z_j)^2} \]

\[ = \left( \sqrt{\sum_{j=1}^{n} (x_j - y_j)^2} + \sqrt{\sum_{j=1}^{n} (y_j - z_j)^2} \right)^2 = (d(x, y) + d(y, z))^2. \]

Because the square root is an increasing function, the inequality is preserved when we take the square root of both sides, and we obtain the triangle inequality.

**Example 7.1.6**: The set of complex numbers \( \mathbb{C} \) is the set of numbers \( z = x + iy \), where \( x \) and \( y \) are in \( \mathbb{R} \). By imposing \( i^2 = -1 \), we make \( \mathbb{C} \) into a field. For the purposes of taking limits, the set \( \mathbb{C} \) is regarded as the metric space \( \mathbb{R}^2 \), where \( z = x + iy \in \mathbb{C} \) corresponds to \( (x, y) \in \mathbb{R}^2 \). For \( z = x + iy \) define the *complex modulus* by \( |z| := \sqrt{x^2 + y^2} \). Then for two complex numbers \( z_1 = x_1 + iy_1 \) and \( z_2 = x_2 + iy_2 \), the distance is

\[ d(z_1, z_2) = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2} = |z_1 - z_2|. \]

Furthermore, when working with complex numbers it is often convenient to write the metric in terms of the so-called *complex conjugate*: The conjugate of \( z = x + iy \) is \( \bar{z} := x - iy \). Then \( |z|^2 = x^2 + y^2 = z\bar{z} \), and so \( |z_1 - z_2|^2 = (z_1 - z_2)(\bar{z_1} - \bar{z_2}). \)

**Example 7.1.7**: An example to keep in mind is the so-called *discrete metric*. For any set \( X \), define

\[ d(x, y) := \begin{cases} 1 & \text{if } x \neq y, \\ 0 & \text{if } x = y. \end{cases} \]

That is, all points are equally distant from each other. When \( X \) is a finite set, we can draw a diagram, see for example Figure 7.2. Of course, in the diagram the distances are not the normal euclidean distances in the plane. Things become subtle when \( X \) is an infinite set such as the real numbers.

While this particular example seldom comes up in practice, it gives a useful “smell test.” If you make a statement about metric spaces, try it with the discrete metric. To show that \( (X, d) \) is indeed a metric space is left as an exercise.
7.1. METRIC SPACES

Example 7.1.8: Let $C([a, b], \mathbb{R})$ be the set of continuous real-valued functions on the interval $[a, b]$. Define the metric on $C([a, b], \mathbb{R})$ as

$$d(f, g) := \sup_{x \in [a, b]} |f(x) - g(x)|.$$  

Let us check the properties. First, $d(f, g)$ is finite as $|f(x) - g(x)|$ is a continuous function on a closed bounded interval $[a, b]$, and so is bounded. It is clear that $d(f, g) \geq 0$, it is the supremum of nonnegative numbers. If $f = g$, then $|f(x) - g(x)| = 0$ for all $x$, and hence $d(f, g) = 0$. Conversely, if $d(f, g) = 0$, then for every $x$, we have $|f(x) - g(x)| \leq d(f, g) = 0$, and hence $f(x) = g(x)$ for all $x$, and so $f = g$. That $d(f, g) = d(g, f)$ is equally trivial. To show the triangle inequality we use the standard triangle inequality;

$$d(f, g) = \sup_{x \in [a, b]} |f(x) - g(x)| = \sup_{x \in [a, b]} |f(x) - h(x) + h(x) - g(x)|$$

$$\leq \sup_{x \in [a, b]} (|f(x) - h(x)| + |h(x) - g(x)|)$$

$$\leq \sup_{x \in [a, b]} |f(x) - h(x)| + \sup_{x \in [a, b]} |h(x) - g(x)| = d(f, h) + d(h, g).$$

When treating $C([a, b], \mathbb{R})$ as a metric space without mentioning a metric, we mean this particular metric. Notice that $d(f, g) = \|f - g\|_u$, the uniform norm of Definition 6.1.9.

This example may seem esoteric at first, but it turns out that working with spaces such as $C([a, b], \mathbb{R})$ is really the meat of a large part of modern analysis. Treating sets of functions as metric spaces allows us to abstract away a lot of the grubby detail and prove powerful results such as Picard’s theorem with less work.

Example 7.1.9: Another useful example of a metric space is the sphere with a metric usually called the great circle distance. Let $S^2$ be the unit sphere in $\mathbb{R}^3$, that is $S^2 := \{x \in \mathbb{R}^3 : x_1^2 + x_2^2 + x_3^2 = 1\}$. Take $x$ and $y$ in $S^2$, draw a line through the origin and $x$, and another line through the origin and $y$, and let $\theta$ be the angle that the two lines make. Then define $d(x, y) := \theta$. See Figure 7.3. The law of cosines from vector calculus says $d(x, y) = \arccos(x_1y_1 + x_2y_2 + x_3y_3)$. It is relatively easy to see that this function satisfies the first three properties of a metric. Triangle inequality is harder to prove, and requires a bit more trigonometry and linear algebra than we wish to indulge in right now, so let us leave it without proof.
This distance is the shortest distance between points on a sphere if we are allowed to travel on the sphere only. It is easy to generalize to arbitrary diameters. If we take a sphere of radius $r$, we let the distance be $d(x,y) := r \theta$. As an example, this is the standard distance you would use if you compute a distance on the surface of the earth, such as computing the distance a plane travels from London to Los Angeles.

Oftentimes it is useful to consider a subset of a larger metric space as a metric space itself. We obtain the following proposition, which has a trivial proof.

**Proposition 7.1.10.** Let $(X,d)$ be a metric space and $Y \subset X$. Then the restriction $d|_{Y \times Y}$ is a metric on $Y$.

**Definition 7.1.11.** If $(X,d)$ is a metric space, $Y \subset X$, and $d' := d|_{Y \times Y}$, then $(Y,d')$ is said to be a *subspace* of $(X,d)$.

It is common to simply write $d$ for the metric on $Y$, as it is the restriction of the metric on $X$. Sometimes we say $d'$ is the *subspace metric* and $Y$ has the *subspace topology*.

A subset of the real numbers is bounded whenever all its elements are at most some fixed distance from 0. When dealing with an arbitrary metric space there may not be some natural fixed point 0, but for the purposes of boundedness it does not matter.

**Definition 7.1.12.** Let $(X,d)$ be a metric space. A subset $S \subset X$ is said to be *bounded* if there exists a $p \in X$ and a $B \in \mathbb{R}$ such that $$d(p,x) \leq B$$ for all $x \in S$.

We say $(X,d)$ is bounded if $X$ itself is a bounded subset.

For example, the set of real numbers with the standard metric is not a bounded metric space. It is not hard to see that a subset of the real numbers is bounded in the sense of chapter 1 if and only if it is bounded as a subset of the metric space of real numbers with the standard metric.

On the other hand, if we take the real numbers with the discrete metric, then we obtain a bounded metric space. In fact, any set with the discrete metric is bounded.

There are other equivalent ways we could generalize boundedness, and are left as exercises. Suppose $X$ is nonempty to avoid a technicality. Then $S \subset X$ being bounded is equivalent to either

(i) For every $p \in X$, there exists a $B > 0$ such that $d(p,x) \leq B$ for all $x \in S$.

(ii) $\text{diam}(S) := \sup\{d(x,y) : x,y \in S\} < \infty$.

The quantity $\text{diam}(S)$ is called the *diameter* of a set and is usually only defined for a nonempty set.
7.1. Exercises

Exercise 7.1.1: Show that for every set \( X \), the discrete metric \( (x, y) = 1 \) if \( x \neq y \) and \( d(x, x) = 0 \) does give a metric space \( (X, d) \).

Exercise 7.1.2: Let \( X := \{0\} \) be a set. Can you make it into a metric space?

Exercise 7.1.3: Let \( X := \{a, b\} \) be a set. Can you make it into two distinct metric spaces? (define two distinct metrics on it)

Exercise 7.1.4: Let the set \( X := \{A, B, C\} \) represent 3 buildings on campus. Suppose we wish our distance to be the time it takes to walk from one building to the other. It takes 5 minutes either way between buildings A and B. However, building C is on a hill and it takes 10 minutes from A and 15 minutes from B to get to C. On the other hand it takes 5 minutes to go from C to A and 7 minutes to go from C to B, as we are going downhill. Do these distances define a metric? If so, prove it, if not, say why not.

Exercise 7.1.5: Suppose \( (X, d) \) is a metric space and \( \varphi : [0, \infty) \rightarrow \mathbb{R} \) is an increasing function such that \( \varphi(t) \geq 0 \) for all \( t \) and \( \varphi(t) = 0 \) if and only if \( t = 0 \). Also suppose \( \varphi \) is subadditive, that is, \( \varphi(s + t) \leq \varphi(s) + \varphi(t) \). Show that with \( d'(x, y) := \varphi(d(x, y)) \), we obtain a new metric space \( (X, d') \).

Exercise 7.1.6: Let \( (X, d_X) \) and \( (Y, d_Y) \) be metric spaces.

a) Show that \( (X \times Y, d) \) with \( d((x_1, y_1), (x_2, y_2)) := d_X(x_1, x_2) + d_Y(y_1, y_2) \) is a metric space.

b) Show that \( (X \times Y, d) \) with \( d((x_1, y_1), (x_2, y_2)) := \max\{d_X(x_1, x_2), d_Y(y_1, y_2)\} \) is a metric space.

Exercise 7.1.7: Let \( X \) be the set of continuous functions on \([0, 1]\). Let \( \varphi : [0, 1] \rightarrow (0, \infty) \) be continuous. Define

\[
d(f, g) := \int_0^1 |f(x) - g(x)| \varphi(x) \, dx.
\]

Show that \( (X, d) \) is a metric space.

Exercise 7.1.8: Let \( (X, d) \) be a metric space. For nonempty bounded subsets \( A \) and \( B \) let

\[
d(x, B) := \inf\{d(x, b) : b \in B\} \quad \text{and} \quad d(A, B) := \sup\{d(a, B) : a \in A\}.
\]

Now define the Hausdorff metric as

\[
d_H(A, B) := \max\{d(A, B), d(B, A)\}.
\]

Note: \( d_H \) can be defined for arbitrary nonempty subsets if we allow the extended reals.

a) Let \( Y \subset \mathcal{P}(X) \) be the set of bounded nonempty subsets. Prove that \( (Y, d_H) \) is a so-called pseudometric space: \( d_H \) satisfies the metric properties (i), (iii), (iv), and further \( d_H(A, A) = 0 \) for all \( A \in Y \).

b) Show by example that \( d \) itself is not symmetric, that is \( d(A, B) \neq d(B, A) \).

c) Find a metric space \( X \) and two different nonempty bounded subsets \( A \) and \( B \) such that \( d_H(A, B) = 0 \).

Exercise 7.1.9: Let \( (X, d) \) be a nonempty metric space and \( S \subset X \) a subset. Prove:

a) \( S \) is bounded if and only if for every \( p \in X \), there exists a \( B > 0 \) such that \( d(p, x) \leq B \) for all \( x \in S \).

b) A nonempty \( S \) is bounded if and only if \( \text{diam}(S) := \sup\{d(x, y) : x, y \in S\} < \infty \).
Exercise 7.1.10:

a) Working in \( \mathbb{R} \), compute \( \text{diam}(a, b) \).

b) Working in \( \mathbb{R}^n \), for every \( r > 0 \), let \( B_r := \{ x_1^2 + x_2^2 + \cdots + x_n^2 < r^2 \} \). Compute \( \text{diam}(B_r) \).

c) Suppose \((X, d)\) is a metric space with at least two points, \( d \) is the discrete metric, and \( p \in X \). Compute \( \text{diam}\{p\} \) and \( \text{diam}(X) \), then conclude that \((X, d)\) is bounded.

Exercise 7.1.11:

a) Find a metric \( d \) on \( \mathbb{N} \), such that \( \mathbb{N} \) is an unbounded set in \((\mathbb{N}, d)\).

b) Find a metric \( d \) on \( \mathbb{N} \), such that \( \mathbb{N} \) is a bounded set in \((\mathbb{N}, d)\).

c) Find a metric \( d \) on \( \mathbb{N} \) such that for every \( n \in \mathbb{N} \) and every \( \epsilon > 0 \), there exists an \( m \in \mathbb{N} \) such that \( d(n, m) < \epsilon \).

Exercise 7.1.12: Let \( C^1([a, b], \mathbb{R}) \) be the set of once continuously differentiable functions on \([a, b]\). Define

\[
d(f, g) := \|f - g\|_u + \|f' - g'\|_u,
\]

where \( \|\cdot\|_u \) is the uniform norm. Prove that \( d \) is a metric.

Exercise 7.1.13: Consider \( \ell^2 \) the set of sequences \( \{x_n\} \) of real numbers such that \( \sum_{n=1}^{\infty} x_n^2 < \infty \).

a) Prove the Cauchy–Schwarz inequality for two sequences \( \{x_n\} \) and \( \{y_n\} \) in \( \ell^2 \): Prove that \( \sum_{n=1}^{\infty} x_n y_n \) converges (absolutely) and

\[
\left( \sum_{n=1}^{\infty} x_n y_n \right)^2 \leq \left( \sum_{n=1}^{\infty} x_n^2 \right) \left( \sum_{n=1}^{\infty} y_n^2 \right).
\]

b) Prove that \( \ell^2 \) is a metric space with the metric \( d(x, y) := \sqrt{\sum_{n=1}^{\infty} (x_n - y_n)^2} \). Hint: Don’t forget to show that the series for \( d(x, y) \) always converges to some finite number.
7.2 Open and closed sets

Note: 2 lectures

7.2.1 Topology

Before we get to convergence, we define the so-called topology. That is, we define open and closed sets in a metric space. And before doing that, we define two special open and closed sets.

Definition 7.2.1. Let \((X, d)\) be a metric space, \(x \in X\), and \(\delta > 0\). Define the open ball, or simply ball, of radius \(\delta\) around \(x\) as

\[ B(x, \delta) := \{ y \in X : d(x, y) < \delta \}. \]

Define the closed ball as

\[ C(x, \delta) := \{ y \in X : d(x, y) \leq \delta \}. \]

When dealing with different metric spaces, it is sometimes vital to emphasize which metric space the ball is in. We do this by writing \(B_X(x, \delta) := B(x, \delta)\) or \(C_X(x, \delta) := C(x, \delta)\).

Example 7.2.2: Take the metric space \(\mathbb{R}\) with the standard metric. For \(x \in \mathbb{R}\) and \(\delta > 0\),

\[ B(x, \delta) = (x - \delta, x + \delta) \quad \text{and} \quad C(x, \delta) = [x - \delta, x + \delta]. \]

Example 7.2.3: Be careful when working on a subspace. Consider the metric space \([0, 1]\) as a subspace of \(\mathbb{R}\). Then in \([0, 1]\),

\[ B(0, 1/2) = B_{[0,1]}(0, 1/2) = \{ y \in [0, 1] : |0 - y| < 1/2 \} = [0, 1/2). \]

This is different from \(B_{\mathbb{R}}(0, 1/2) = (-1/2, 1/2)\). The important thing to keep in mind is which metric space we are working in.

Definition 7.2.4. Let \((X, d)\) be a metric space. A subset \(V \subset X\) is open if for every \(x \in V\), there exists a \(\delta > 0\) such that \(B(x, \delta) \subset V\). See Figure 7.4. A subset \(E \subset X\) is closed if the complement \(E^c = X \setminus E\) is open. When the ambient space \(X\) is not clear from context, we say \(V\) is open in \(X\) and \(E\) is closed in \(X\).

If \(x \in V\) and \(V\) is open, then we say \(V\) is an open neighborhood of \(x\) (or sometimes just neighborhood).

Intuitively, an open set \(V\) is a set that does not include its “boundary.” Wherever we are in \(V\), we are allowed to “wiggle” a little bit and stay in \(V\). Similarly, a set \(E\) is closed if everything not in \(E\) is some distance away from \(E\). The open and closed balls are examples of open and closed sets (this must still be proved). But not every set is either open or closed. Generally, most subsets are neither.

Example 7.2.5: The set \((0, \infty) \subset \mathbb{R}\) is open: Given any \(x \in (0, \infty)\), let \(\delta := x\). Then \(B(x, \delta) = (0, 2x) \subset (0, \infty)\).

The set \([0, \infty) \subset \mathbb{R}\) is closed: Given \(x \in (-\infty, 0) = [0, \infty)^c\), let \(\delta := -x\). Then \(B(x, \delta) = (-2x, 0) \subset (-\infty, 0) = [0, \infty)^c\).
CHAPTER 7. METRIC SPACES

The set \([0, 1) \subset \mathbb{R}\) is neither open nor closed. First, every ball in \(\mathbb{R}\) around 0, \(B(0, \delta) = (-\delta, \delta)\), contains negative numbers and hence is not contained in \([0, 1)\). So \([0, 1)\) is not open. Second, every ball in \(\mathbb{R}\) around 1, \(B(1, \delta) = (1 - \delta, 1 + \delta)\), contains numbers strictly less than 1 and greater than 0 (e.g. \(1 - \delta/2\) as long as \(\delta < 2\)). Thus \([0, 1)^c = \mathbb{R} \setminus [0, 1)\) is not open, and \([0, 1)\) is not closed.

**Proposition 7.2.6.** Let \((X, d)\) be a metric space.

(i) \(\emptyset\) and \(X\) are open.

(ii) If \(V_1, V_2, \ldots, V_k\) are open subsets of \(X\), then

\[
\bigcap_{j=1}^{k} V_j
\]

is also open. That is, a finite intersection of open sets is open.

(iii) If \(\{V_\lambda\}_{\lambda \in I}\) is an arbitrary collection of open subsets of \(X\), then

\[
\bigcup_{\lambda \in I} V_\lambda
\]

is also open. That is, a union of open sets is open.

The index set \(I\) in (iii) can be arbitrarily large. By \(\bigcup_{\lambda \in I} V_\lambda\), we simply mean the set of all \(x\) such that \(x \in V_\lambda\) for at least one \(\lambda \in I\).

**Proof.** The sets \(\emptyset\) and \(X\) are obviously open in \(X\).

Let us prove (ii). If \(x \in \bigcap_{j=1}^{k} V_j\), then \(x \in V_j\) for all \(j\). As \(V_j\) are all open, for every \(j\) there exists a \(\delta_j > 0\) such that \(B(x, \delta_j) \subset V_j\). Take \(\delta := \min\{\delta_1, \delta_2, \ldots, \delta_k\}\) and notice \(\delta > 0\). We have \(B(x, \delta) \subset B(x, \delta_j) \subset V_j\) for every \(j\) and so \(B(x, \delta) \subset \bigcap_{j=1}^{k} V_j\). Consequently the intersection is open.

Let us prove (iii). If \(x \in \bigcup_{\lambda \in I} V_\lambda\), then \(x \in V_\lambda\) for some \(\lambda \in I\). As \(V_\lambda\) is open, there exists a \(\delta > 0\) such that \(B(x, \delta) \subset V_\lambda\). But then \(B(x, \delta) \subset \bigcup_{\lambda \in I} V_\lambda\), and so the union is open.

**Example 7.2.7:** The main thing to notice is the difference between items (ii) and (iii). Item (ii) is not true for an arbitrary intersection. For example, \(\bigcap_{n=1}^{\infty} (-1/n, 1/n) = \{0\}\), which is not open.

The proof of the following analogous proposition for closed sets is left as an exercise.
Proposition 7.2.8. Let \((X, d)\) be a metric space.

(i) \(\emptyset\) and \(X\) are closed.

(ii) If \(\{E_\lambda\}_{\lambda \in I}\) is an arbitrary collection of closed subsets of \(X\), then
\[
\bigcap_{\lambda \in I} E_\lambda
\]

is also closed. That is, an intersection of closed sets is closed.

(iii) If \(E_1, E_2, \ldots, E_k\) are closed subsets of \(X\), then
\[
\bigcup_{j=1}^k E_j
\]

is also closed. That is, a finite union of closed sets is closed.

Despite the naming, we have not yet shown that the open ball is open and the closed ball is closed. Let us show these facts now to justify the terminology.

Proposition 7.2.9. Let \((X, d)\) be a metric space, \(x \in X\), and \(\delta > 0\). Then \(B(x, \delta)\) is open and \(C(x, \delta)\) is closed.

Proof. Let \(y \in B(x, \delta)\). Let \(\alpha := \delta - d(x, y)\). As \(\alpha > 0\), consider \(z \in B(y, \alpha)\). Then
\[
d(x, z) \leq d(x, y) + d(y, z) < d(x, y) + \alpha = d(x, y) + \delta - d(x, y) = \delta.
\]
Therefore, \(z \in B(x, \delta)\) for every \(z \in B(y, \alpha)\). So \(B(y, \alpha) \subset B(x, \delta)\) and \(B(x, \delta)\) is open. See Figure 7.5.

![Figure 7.5](image)

Figure 7.5: Proof that \(B(x, \delta)\) is open: \(B(y, \alpha) \subset B(x, \delta)\) with the triangle inequality illustrated.

The proof that \(C(x, \delta)\) is closed is left as an exercise.

Again, be careful about which metric space we are in. The set \([0, 1/2]\) is an open ball in \([0, 1]\), and so \([0, 1/2]\) is an open set in \([0, 1]\). On the other hand, \([0, 1/2]\) is neither open nor closed in \(\mathbb{R}\).

Proposition 7.2.10. Let \(a, b\) be two real numbers, \(a < b\). Then \((a, b)\), \((a, \infty)\), and \((-\infty, b)\) are open in \(\mathbb{R}\). Also \([a, b]\), \([a, \infty)\), and \((-\infty, b]\) are closed in \(\mathbb{R}\).

The proof is left as an exercise. Keep in mind that there are many other open and closed sets in the set of real numbers.
Proposition 7.2.11. Suppose \((X, d)\) is a metric space, and \(Y \subset X\). Then \(U \subset Y\) is open in \(Y\) (in the subspace topology) if and only if there exists an open set \(V \subset X\) (so open in \(X\)), such that \(V \cap Y = U\).

For example, let \(X := \mathbb{R}\), \(Y := [0, 1]\), \(U := [0, 1/2)\). We saw that \(U\) is an open set in \(Y\). We may take \(V := (-1/2, 1/2)\).

Proof. Suppose \(V \subset X\) is open and \(x \in V \cap Y\). Let \(U := V \cap Y\). As \(V\) is open, there exists a \(\delta > 0\) such that \(B_X(x, \delta) \subset V\). Then

\[ B_Y(x, \delta) = B_X(x, \delta) \cap Y \subset V \cap Y = U. \]

The proof of the opposite direction, that is, that if \(U \subset Y\) is open in the subspace topology there exists a \(V\) is left as Exercise 7.2.12.

A hint for finishing the proof (the exercise) is that a useful way to think about an open set is as a union of open balls. If \(U\) is open, then for each \(x \in U\), there is a \(\delta_x > 0\) (depending on \(x\)) such that \(B(x, \delta_x) \subset U\). Then \(U = \bigcup_{x \in U} B(x, \delta_x)\).

In case of an open subset of an open set or a closed subset of a closed set, matters are simpler.

Proposition 7.2.12. Suppose \((X, d)\) is a metric space, \(V \subset X\) is open, and \(E \subset X\) is closed.

(i) \(U \subset V\) is open in the subspace topology if and only if \(U\) is open in \(X\).

(ii) \(F \subset E\) is closed in the subspace topology if and only if \(F\) is closed in \(X\).

Proof. Let us prove (i) and leave (ii) to an exercise.

If \(U \subset V\) is open in the subspace topology, by Proposition 7.2.11, there exists a set \(W \subset X\) open in \(X\), such that \(U = W \cap V\). Intersection of two open sets is open so \(U\) is open in \(X\).

Now suppose \(U\) is open in \(X\). Then \(U = U \cap V\). So \(U\) is open in \(V\) again by Proposition 7.2.11.

7.2.2 Connected sets

Let us generalize the idea of an interval to general metric spaces. One of the main features of an interval in \(\mathbb{R}\) is that it is connected—that we can continuously move from one point of it to another point without jumping. For example, in \(\mathbb{R}\) we usually study functions on intervals, and in more general metric spaces we usually study functions on connected sets.

Definition 7.2.13. A nonempty* metric space \((X, d)\) is connected if the only subsets of \(X\) that are both open and closed (so-called clopen subsets) are \(\emptyset\) and \(X\) itself. If a nonempty \((X, d)\) is not connected we say it is disconnected.

When we apply the term connected to a nonempty subset \(A \subset X\), we mean that \(A\) with the subspace topology is connected.

*Some authors do not exclude the empty set from the definition, and the empty set would then be connected. We avoid the empty set for essentially the same reason why 1 is neither a prime nor a composite number: Our connected sets have exactly two clopen subsets and disconnected sets have more than two. The empty set has exactly one.
In other words, a nonempty $X$ is connected if whenever we write $X = X_1 \cup X_2$ where $X_1 \cap X_2 = \emptyset$ and $X_1$ and $X_2$ are open, then either $X_1 = \emptyset$ or $X_2 = \emptyset$. So to show $X$ is disconnected, we need to find nonempty disjoint open sets $X_1$ and $X_2$ whose union is $X$. For subsets, we state this idea as a proposition. The proposition is illustrated in Figure 7.6.

**Proposition 7.2.14.** Let $(X, d)$ be a metric space. A nonempty set $S \subset X$ is disconnected if and only if there exist open sets $U_1$ and $U_2$ in $X$, such that $U_1 \cap U_2 \cap S = \emptyset$, $U_1 \cap S \neq \emptyset$, $U_2 \cap S \neq \emptyset$, and $S = (U_1 \cap S) \cup (U_2 \cap S)$.

![Figure 7.6: Disconnected subset. Notice that $U_1 \cap U_2$ need not be empty, but $U_1 \cap U_2 \cap S = \emptyset$.](image)

**Proof.** First suppose $S$ is disconnected: There are nonempty disjoint $S_1$ and $S_2$ that are open in $S$ and $S = S_1 \cup S_2$. **Proposition 7.2.11** says there exist $U_1$ and $U_2$ that are open in $X$ such that $U_1 \cap S = S_1$ and $U_2 \cap S = S_2$.

For the other direction start with the $U_1$ and $U_2$. Then $U_1 \cap S$ and $U_2 \cap S$ are open in $S$ by **Proposition 7.2.11**. Via the discussion before the proposition, $S$ is disconnected.

**Example 7.2.15:** Let $S \subset \mathbb{R}$ be such that $x < z < y$ with $x, y \in S$ and $z \notin S$. Claim: $S$ is disconnected.

**Proof:** Notice $$((-\infty, z) \cap S) \cup ((z, \infty) \cap S) = S.$$

**Proposition 7.2.16.** A nonempty set $S \subset \mathbb{R}$ is connected if and only if it is an interval or a single point.

**Proof.** Suppose $S$ is connected. If $S$ is a single point, then we are done. So suppose $x < y$ and $x, y \in S$. If $z \in \mathbb{R}$ is such that $x < z < y$, then $(-\infty, z) \cap S$ is nonempty and $(z, \infty) \cap S$ is nonempty. The two sets are disjoint. As $S$ is connected, we must have they their union is not $S$, so $z \notin S$. By **Proposition 1.4.1**, $S$ is an interval.

If $S$ is a single point, it is connected. Therefore, suppose $S$ is an interval. Consider open subsets $U_1$ and $U_2$ of $\mathbb{R}$, such that $U_1 \cap S$ and $U_2 \cap S$ are nonempty, and $S = (U_1 \cap S) \cup (U_2 \cap S)$. We will show that $U_1 \cap S$ and $U_2 \cap S$ contain a common point, so they are not disjoint, proving that $S$ is connected. Suppose $x \in U_1 \cap S$ and $y \in U_2 \cap S$. Without loss of generality, assume $x < y$. As $S$ is an interval, $[x, y] \subset S$. Note that $U_2 \cap [x, y] \neq \emptyset$, and let $z := \inf(U_2 \cap [x, y])$. We wish to show that $z \in U_1$. If $z = x$, then $z \in U_1$. If $z > x$, then for every $\varepsilon > 0$, the ball $B(z, \varepsilon) = (z - \varepsilon, z + \varepsilon)$ contains points of $[x, y]$ not in $U_2$, as $z$ is the infimum of such points. So $z \notin U_2$ as $U_2$ is open. Therefore, $z \in U_1$ as every point of $[x, y]$ is in $U_1$ or $U_2$. As $U_1$ is open, $B(z, \delta) \subset U_1$ for a small enough $\delta > 0$. As $z$ is the infimum of the nonempty set $U_2 \cap [x, y]$, there must exist some $w \in U_2 \cap [x, y]$ such that $w \in [z, z + \delta] \subset B(z, \delta) \subset U_1$. Therefore, $w \in U_1 \cap U_2 \cap [x, y]$. So $U_1 \cap S$ and $U_2 \cap S$ are not disjoint, and $S$ is connected. See Figure 7.7.
Example 7.2.17: Oftentimes a ball \( B(x, \delta) \) is connected, but this is not necessarily true in every metric space. For a simplest example, take a two point space \( \{a, b\} \) with the discrete metric. Then \( B(a, 2) = \{a, b\} \), which is not connected as \( B(a, 1) = \{a\} \) and \( B(b, 1) = \{b\} \) are open and disjoint.

7.2.3 Closure and boundary

Sometimes we wish to take a set and throw in everything that we can approach from the set. This concept is called the closure.

Definition 7.2.18. Let \( (X, d) \) be a metric space and \( A \subset X \). The closure \( \overline{A} \) of \( A \) is the set

\[
\overline{A} := \bigcap \{ E \subset X : E \text{ is closed and } A \subset E \}.
\]

That is, \( \overline{A} \) is the intersection of all closed sets that contain \( A \).

Proposition 7.2.19. Let \( (X, d) \) be a metric space and \( A \subset X \). The closure \( \overline{A} \) is closed, and \( A \subset \overline{A} \). Furthermore, if \( A \) is closed, then \( \overline{A} = A \).

Proof. The closure is an intersection of closed sets, so \( \overline{A} \) is closed. There is at least one closed set containing \( A \), namely \( X \) itself, so \( A \subset \overline{A} \). If \( A \) is closed, then \( A \) is a closed set that contains \( A \). So \( \overline{A} \subset A \), and thus \( A = \overline{A} \). \( \square \)

Example 7.2.20: The closure of \( (0, 1) \) in \( \mathbb{R} \) is \( [0, 1] \). Proof: If \( E \) is closed and contains \( (0, 1) \), then \( E \) contains 0 and 1 (why?). Thus \( [0, 1] \subset E \). But \( [0, 1] \) is also closed. Hence, the closure \( (0, 1) = [0, 1] \).

Example 7.2.21: Be careful to notice what ambient metric space you are working with. If \( X = (0, \infty) \), then the closure of \( (0, 1) \) in \( (0, \infty) \) is \( (0, 1] \). Proof: Similarly as above, \( (0, 1] \) is closed in \( (0, \infty) \) (why?). Any closed set \( E \) that contains \( (0, 1) \) must contain 1 (why?). Therefore, \( (0, 1] \subset E \), and hence \( (0, 1) = (0, 1] \) when working in \( (0, \infty) \).

Let us justify the statement that the closure is everything that we can “approach” from the set.

Proposition 7.2.22. Let \( (X, d) \) be a metric space and \( A \subset X \). Then \( x \in \overline{A} \) if and only if for every \( \delta > 0 \), \( B(x, \delta) \cap A \neq \emptyset \).

Proof. Let us prove the two contrapositives. Let us show that \( x \notin \overline{A} \) if and only if there exists a \( \delta > 0 \) such that \( B(x, \delta) \cap A = \emptyset \).

First suppose \( x \notin \overline{A} \). We know \( \overline{A} \) is closed. Thus there is a \( \delta > 0 \) such that \( B(x, \delta) \subset \overline{A}^c \). As \( A \subset \overline{A} \) we see that \( B(x, \delta) \subset \overline{A}^c \subset A^c \) and hence \( B(x, \delta) \cap A = \emptyset \).

On the other hand, suppose there is a \( \delta > 0 \), such that \( B(x, \delta) \cap A = \emptyset \). In other words, \( A \subset B(x, \delta)^c \). As \( B(x, \delta)^c \) is a closed set, as \( x \notin B(x, \delta)^c \), and as \( \overline{A} \) is the intersection of closed sets containing \( A \), we have \( x \notin \overline{A} \). \( \square \)
We can also talk about the interior of a set (points we cannot approach from the complement), and the boundary of a set (points we can approach both from the set and its complement).

**Definition 7.2.23.** Let \((X,d)\) be a metric space and \(A \subset X\). The **interior** of \(A\) is the set
\[
A^\circ := \{ x \in A : \text{there exists a } \delta > 0 \text{ such that } B(x, \delta) \subset A \}.
\]
The **boundary** of \(A\) is the set
\[
\partial A := \overline{A} \setminus A^\circ.
\]

Alternatively, the interior is the union of open sets lying in \(A\), see Exercise 7.2.14. By definition, \(A^\circ \subset A\); however, the points of the boundary may or may not be in \(A\).

**Example 7.2.24:** Suppose \(A := (0, 1]\) and \(X := \mathbb{R}\). Then \(\overline{A} = [0, 1]\), \(A^\circ = (0, 1)\), and \(\partial A = \{0, 1\}\).

**Example 7.2.25:** Consider \(X := \{a, b\}\) with the discrete metric, and let \(A := \{a\}\). Then \(\overline{A} = A^\circ = A\) and \(\partial A = \emptyset\).

**Proposition 7.2.26.** Let \((X,d)\) be a metric space and \(A \subset X\). Then \(A^\circ\) is open and \(\partial A\) is closed.

**Proof.** Given \(x \in A^\circ\), there is a \(\delta > 0\) such that \(B(x, \delta) \subset A\). If \(z \in B(x, \delta)\), then as open balls are open, there is an \(\varepsilon > 0\) such that \(B(z, \varepsilon) \subset B(x, \delta) \subset A\). So \(z \in A^\circ\). Therefore, \(B(x, \delta) \subset A^\circ\), and \(A^\circ\) is open.

As \(A^\circ\) is open, then \(\partial A = \overline{A} \setminus A^\circ = \overline{A} \cap (A^\circ)^c\) is closed.

The boundary is the set of points that are close to both the set and its complement. See Figure 7.8 for the a diagram of the next proposition.

**Proposition 7.2.27.** Let \((X,d)\) be a metric space and \(A \subset X\). Then \(x \in \partial A\) if and only if for every \(\delta > 0\), \(B(x, \delta) \cap A\) and \(B(x, \delta) \cap A^c\) are both nonempty.

![Figure 7.8: Boundary is the set where every ball contains points in the set and also its complement.](image)

**Proof.** Suppose \(x \in \partial A = \overline{A} \setminus A^\circ\) and let \(\delta > 0\) be arbitrary. By Proposition 7.2.22, \(B(x, \delta)\) contains a point of \(A\). If \(B(x, \delta)\) contained no points of \(A^c\), then \(x\) would be in \(A^\circ\). Hence \(B(x, \delta)\) contains a point of \(A^c\) as well.

Let us prove the other direction by contrapositive. Suppose \(x \notin \partial A\), so \(x \notin \overline{A}\) or \(x \in A^\circ\). If \(x \notin \overline{A}\), then \(B(x, \delta) \subset A^c\) for some \(\delta > 0\) as \(\overline{A}\) is closed. So \(B(x, \delta) \cap A\) is empty, because \(\overline{A} \subset A^c\). If \(x \in A^\circ\), then \(B(x, \delta) \subset A\) for some \(\delta > 0\), so \(B(x, \delta) \cap A^c\) is empty. 

\(\square\)
We obtain the following immediate corollary about closures of $A$ and $A^c$. We simply apply Proposition 7.2.22.

**Corollary 7.2.28.** Let $(X,d)$ be a metric space and $A \subset X$. Then $\partial A = \overline{A} \cap \overline{A^c}$.

### 7.2.4 Exercises

**Exercise 7.2.1:** Prove Proposition 7.2.8. Hint: Apply Proposition 7.2.6 to the complements of the sets.

**Exercise 7.2.2:** Finish the proof of Proposition 7.2.9 by proving that $C(x,\delta)$ is closed.

**Exercise 7.2.3:** Prove Proposition 7.2.10.

**Exercise 7.2.4:** Suppose $(X,d)$ is a nonempty metric space with the discrete topology. Show that $X$ is connected if and only if it contains exactly one element.

**Exercise 7.2.5:** Take $\mathbb{Q}$ with the standard metric, $d(x,y) = |x - y|$, as our metric space. Prove that $\mathbb{Q}$ is totally disconnected, that is, show that for every $x, y \in \mathbb{Q}$ with $x \neq y$, there exists an two open sets $U$ and $V$, such that $x \in U$, $y \in V$, $U \cap V = \emptyset$, and $U \cup V = \mathbb{Q}$.

**Exercise 7.2.6:** Show that in a metric space, every open set can be written as a union of closed sets.

**Exercise 7.2.7:** Prove that in a metric space,

a) $E$ is closed if and only if $\partial E \subset E$.

b) $U$ is open if and only if $\partial U \cap U = \emptyset$.

**Exercise 7.2.8:** Prove that in a metric space,

a) $A$ is open if and only if $A^\circ = A$.

b) $U \subset A^\circ$ for every open set $U$ such that $U \subset A$.

**Exercise 7.2.9:** Let $X$ be a set and $d, d'$ be two metrics on $X$. Suppose there exists an $\alpha > 0$ and $\beta > 0$ such that $\alpha d(x,y) \leq d'(x,y) \leq \beta d(x,y)$ for all $x, y \in X$. Show that $U$ is open in $(X,d)$ if and only if $U$ is open in $(X,d')$. That is, the topologies of $(X,d)$ and $(X,d')$ are the same.

**Exercise 7.2.10:** Suppose $\{S_i\}, i \in \mathbb{N}$, is a collection of connected subsets of a metric space $(X,d)$, and there exists an $x \in X$ such that $x \in S_i$ for all $i \in \mathbb{N}$. Show that $\bigcup_{i=1}^{\infty} S_i$ is connected.

**Exercise 7.2.11:** Let $A$ be a connected set in a metric space.

a) Is $\overline{A}$ connected? Prove or find a counterexample.

b) Is $A^\circ$ connected? Prove or find a counterexample.

Hint: Think of sets in $\mathbb{R}^2$.

**Exercise 7.2.12:** Finish the proof of Proposition 7.2.11. Suppose $(X,d)$ is a metric space and $Y \subset X$. Show that with the subspace metric on $Y$, if a set $U \subset Y$ is open (in $Y$), then there exists an open set $V \subset X$ such that $U = V \cap Y$.

**Exercise 7.2.13:** Let $(X,d)$ be a metric space.

a) For every $x \in X$ and $\delta > 0$, show $\overline{B(x,\delta)} \subset C(x,\delta)$.

b) Is it always true that $\overline{B(x,\delta)} = C(x,\delta)$? Prove or find a counterexample.
Exercise 7.2.14: Let \((X, d)\) be a metric space and \(A \subset X\). Show that \(A^\circ = \bigcup \{ V : V \text{ is open and } V \subset A \}\).

Exercise 7.2.15: Finish the proof of Proposition 7.2.12.

Exercise 7.2.16: Let \((X, d)\) be a metric space. Show that there exists a bounded metric \(d'\) such that \((X, d')\) has the same open sets, that is, the topology is the same.

Exercise 7.2.17: Let \((X, d)\) be a metric space.

a) Prove that for every \(x \in X\), there either exists a \(\delta > 0\) such that \(B(x, \delta) = \{ x \}\), or \(B(x, \delta)\) is infinite for every \(\delta > 0\).

b) Find an explicit example of \((X, d)\), \(X\) infinite, where for every \(\delta > 0\) and every \(x \in X\), the ball \(B(x, \delta)\) is finite.

c) Find an explicit example of \((X, d)\) where for every \(\delta > 0\) and every \(x \in X\), the ball \(B(x, \delta)\) is countably infinite.

d) Prove that if \(X\) is uncountable, then there exists an \(x \in X\) and a \(\delta > 0\) such that \(B(x, \delta)\) is uncountable.

Exercise 7.2.18: For every \(x \in \mathbb{R}^n\) and every \(\delta > 0\) define the “rectangle” \(R(x, \delta) := (x_1 - \delta, x_1 + \delta) \times (x_2 - \delta, x_2 + \delta) \times \cdots \times (x_n - \delta, x_n + \delta)\). Show that these sets generate the same open sets as the balls in standard metric. That is, show that a set \(U \subset \mathbb{R}^n\) is open in the sense of the standard metric if and only if for every point \(x \in U\), there exists a \(\delta > 0\) such that \(R(x, \delta) \subset U\).
7.3 Sequences and convergence

Note: 1 lecture

7.3.1 Sequences

The notion of a sequence in a metric space is very similar to a sequence of real numbers. The related definitions are essentially the same as those for real numbers in the sense of chapter 2, where \( \mathbb{R} \) with the standard metric \( d(x, y) = |x - y| \) is replaced by an arbitrary metric space \((X, d)\).

**Definition 7.3.1.** A sequence in a metric space \((X, d)\) is a function \(x: \mathbb{N} \to X\). As before we write \(x_n\) for the \(n\)th element in the sequence and use the notation \(\{x_n\}_{n=1}^{\infty}\), or more precisely \(\{x_n\}_{n=1}^{\infty}\).

A sequence \(\{x_n\}\) is **bounded** if there exists a point \(p \in X\) and \(B \in \mathbb{R}\) such that \(d(p, x_n) \leq B\) for all \(n \in \mathbb{N}\).

In other words, the sequence \(\{x_n\}\) is bounded whenever the set \(\{x_n : n \in \mathbb{N}\}\) is bounded.

If \(\{n_j\}_{j=1}^{\infty}\) is a sequence of natural numbers such that \(n_{j+1} > n_j\) for all \(j\), then the sequence \(\{x_{n_j}\}_{j=1}^{\infty}\) is said to be a **subsequence** of \(\{x_n\}\).

Similarly we define convergence. Again, we cheat a little and use the definite article in front of the word limit before we prove that the limit is unique. See Figure 7.9, for an idea of the definition.

**Definition 7.3.2.** A sequence \(\{x_n\}\) in a metric space \((X, d)\) is said to converge to a point \(p \in X\) if for every \(\varepsilon > 0\), there exists an \(M \in \mathbb{N}\) such that \(d(x_n, p) < \varepsilon\) for all \(n \geq M\). The point \(p\) is said to be the **limit** of \(\{x_n\}\). We write

\[
\lim_{n \to \infty} x_n := p.
\]

A sequence that converges is **convergent**. Otherwise, the sequence is **divergent**.

---

Let us prove that the limit is unique. The proof is almost identical (word for word) to the proof of the same fact for sequences of real numbers, Proposition 2.1.6. Proofs of many results we know for sequences of real numbers can be adapted to the more general settings of metric spaces. We must replace \(|x - y|\) with \(d(x, y)\) in the proofs and apply the triangle inequality correctly.
Proposition 7.3.3. A convergent sequence in a metric space has a unique limit.

Proof. Suppose the sequence \( \{x_n\} \) has limits \( x \) and \( y \). Take an arbitrary \( \varepsilon > 0 \). From the definition find an \( M_1 \) such that for all \( n \geq M_1 \), \( d(x_n, x) < \varepsilon/2 \). Similarly find an \( M_2 \) such that for all \( n \geq M_2 \), we have \( d(x_n, y) < \varepsilon/2 \). Now take an \( n \) such that \( n \geq M_1 \) and also \( n \geq M_2 \), and estimate

\[
d(y, x) \leq d(y, x_n) + d(x_n, x) < \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon.
\]

As \( d(y, x) < \varepsilon \) for all \( \varepsilon > 0 \), then \( d(x, y) = 0 \) and \( y = x \). Hence the limit (if it exists) is unique. \( \square \)

The proofs of the following propositions are left as exercises.

Proposition 7.3.4. A convergent sequence in a metric space is bounded.

Proposition 7.3.5. A sequence \( \{x_n\} \) in a metric space \( (X, d) \) converges to \( p \in X \) if and only if there exists a sequence \( \{a_n\} \) of real numbers such that

\[
d(x_n, p) \leq a_n \quad \text{for all } n \in \mathbb{N},
\]

and

\[
\lim_{n \to \infty} a_n = 0.
\]

Proposition 7.3.6. Let \( \{x_n\} \) be a sequence in a metric space \( (X, d) \).

(i) If \( \{x_n\} \) converges to \( p \in X \), then every subsequence \( \{x_{n_k}\} \) converges to \( p \).

(ii) If for some \( K \in \mathbb{N} \) the \( K \)-tail \( \{x_n\}_{n=K+1}^\infty \) converges to \( p \in X \), then \( \{x_n\} \) converges to \( p \).

Example 7.3.7: Take \( C([0,1], \mathbb{R}) \) be the set of continuous functions with the metric being the uniform metric. We saw that we obtain a metric space. If we look at the definition of convergence, we notice that it is identical to uniform convergence. That is, \( \{f_n\} \) converges uniformly if and only if it converges in the metric space sense.

Remark 7.3.8. It is perhaps surprising that on the set of functions \( f : [a,b] \to \mathbb{R} \) (continuous or not) there is no metric that gives pointwise convergence. Although the proof of this fact is beyond the scope of this book.

7.3.2 Convergence in euclidean space

In the euclidean space \( \mathbb{R}^n \), a sequence converges if and only if every component converges:

Proposition 7.3.9. Let \( \{x_j\}_{j=1}^\infty \) be a sequence in \( \mathbb{R}^n \), where we write \( x_j = (x_{j,1}, x_{j,2}, \ldots, x_{j,n}) \in \mathbb{R}^n \). Then \( \{x_j\}_{j=1}^\infty \) converges if and only if \( \{x_{j,k}\}_{j=1}^\infty \) converges for every \( k = 1, 2, \ldots, n \), in which case

\[
\lim_{j \to \infty} x_j = \left( \lim_{j \to \infty} x_{j,1}, \lim_{j \to \infty} x_{j,2}, \ldots, \lim_{j \to \infty} x_{j,n} \right).
\]
Proof. Suppose the sequence \( \{x_j\}_{j=1}^{\infty} \) converges to \( y = (y_1, y_2, \ldots, y_n) \in \mathbb{R}^n \). Given \( \varepsilon > 0 \), there exists an \( M \), such that for all \( j \geq M \), we have

\[
d(y, x_j) < \varepsilon.
\]

Fix some \( k = 1, 2, \ldots, n \). For all \( j \geq M \),

\[
|y_k - x_j| = \sqrt{(y_k - x_j)^2} \leq \left( \sum_{\ell=1}^{n} (y_{\ell} - x_{j,\ell})^2 \right)^{1/2} = d(y, x_j) < \varepsilon.
\]

Hence the sequence \( \{x_{j,k}\}_{j=1}^{\infty} \) converges to \( y_k \).

For the other direction, suppose \( \{x_{j,k}\}_{j=1}^{\infty} \) converges to \( y_k \) for every \( k = 1, 2, \ldots, n \). Given \( \varepsilon > 0 \), pick an \( M \), such that if \( j \geq M \), then \( |y_k - x_j| < \varepsilon/\sqrt{n} \) for all \( k = 1, 2, \ldots, n \). Then

\[
d(y, x_j) = \left( \sum_{k=1}^{n} (y_k - x_{j,k})^2 \right)^{1/2} < \left( \sum_{k=1}^{n} \left( \frac{\varepsilon}{\sqrt{n}} \right)^2 \right)^{1/2} = \sqrt{\sum_{k=1}^{n} \frac{\varepsilon^2}{n}} = \varepsilon.
\]

That is, the sequence \( \{x_j\} \) converges to \( y = (y_1, y_2, \ldots, y_n) \in \mathbb{R}^n \).

Example 7.3.10: As we said, the set \( \mathbb{C} \) of complex numbers \( z = x + iy \) is considered as the metric space \( \mathbb{R}^2 \). The proposition says that the sequence \( \{z_j\}_{j=1}^{\infty} = \{x_j + iy_j\}_{j=1}^{\infty} \) converges to \( z = x + iy \) if and only if \( \{x_j\} \) converges to \( x \) and \( \{y_j\} \) converges to \( y \).

### 7.3.3 Convergence and topology

The topology—the set of open sets of a space—encodes which sequences converge.

**Proposition 7.3.11.** Let \((X, d)\) be a metric space and \( \{x_n\} \) a sequence in \( X \). Then \( \{x_n\} \) converges to \( x \in X \) if and only if for every open neighborhood \( U \) of \( x \), there exists an \( M \in \mathbb{N} \) such that for all \( n \geq M \), we have \( x_n \in U \).

**Proof.** Suppose \( \{x_n\} \) converges to \( x \). Let \( U \) be an open neighborhood of \( x \), then there exists an \( \varepsilon > 0 \) such that \( B(x, \varepsilon) \subset U \). As the sequence converges, find an \( M \in \mathbb{N} \) such that for all \( n \geq M \), we have \( d(x, x_n) < \varepsilon \), or in other words \( x_n \in B(x, \varepsilon) \subset U \).

Let us prove the other direction. Given \( \varepsilon > 0 \), let \( U := B(x, \varepsilon) \) be the neighborhood of \( x \). Then there is an \( M \in \mathbb{N} \) such that for \( n \geq M \), we have \( x_n \in U = B(x, \varepsilon) \), or in other words, \( d(x, x_n) < \varepsilon \). □

A closed set contains the limits of its convergent sequences.

**Proposition 7.3.12.** Let \((X, d)\) be a metric space, \( E \subset X \) a closed set, and \( \{x_n\} \) a sequence in \( E \) that converges to some \( x \in X \). Then \( x \in E \).

**Proof.** Let us prove the contrapositive. Suppose \( \{x_n\} \) is a sequence in \( X \) that converges to \( x \in E^c \). As \( E^c \) is open, **Proposition 7.3.11** says that there is an \( M \) such that for all \( n \geq M \), \( x_n \in E^c \). So \( \{x_n\} \) is not a sequence in \( E \).

To take a closure of a set \( A \), we take \( A \), and we throw in points that are limits of sequences in \( A \).
7.3. SEQUENCES AND CONVERGENCE

Proposition 7.3.13. Let \((X,d)\) be a metric space and \(A \subset X\). Then \(x \in \bar{A}\) if and only if there exists a sequence \(\{x_n\}\) of elements in \(A\) such that \(\lim x_n = x\).

Proof. Let \(x \in \bar{A}\). For every \(n \in \mathbb{N}\), Proposition 7.2.22 guarantees a point \(x_n \in B(x,1/n) \cap A\). As \(d(x,x_n) < 1/n\), we have \(\lim x_n = x\).

For the other direction, see Exercise 7.3.1.

\(\square\)

7.3.4 Exercises

Exercise 7.3.1: Finish the proof of Proposition 7.3.13: Let \((X,d)\) be a metric space and \(A \subset X\). Let \(x \in X\) be such that there exists a sequence \(\{x_n\}\) in \(A\) that converges to \(x\). Prove that \(x \in \bar{A}\).

Exercise 7.3.2:

a) Show that \(d(x,y) := \min\{1,|x-y|\}\) defines a metric on \(\mathbb{R}\).

b) Show that a sequence converges in \((\mathbb{R},d)\) if and only if it converges in the standard metric.

c) Find a bounded sequence in \((\mathbb{R},d)\) that contains no convergent subsequence.

Exercise 7.3.3: Prove Proposition 7.3.4.

Exercise 7.3.4: Prove Proposition 7.3.5.

Exercise 7.3.5: Suppose \(\{x_n\}_{n=1}^\infty\) converges to \(x\). Suppose \(f : \mathbb{N} \to \mathbb{N}\) is a one-to-one function. Show that \(\{x_{f(n)}\}_{n=1}^\infty\) converges to \(x\).

Exercise 7.3.6: Let \((X,d)\) be a metric space where \(d\) is the discrete metric. Suppose \(\{x_n\}\) is a convergent sequence in \(X\). Show that there exists a \(K \in \mathbb{N}\) such that for all \(n \geq K\), we have \(x_n = x_K\).

Exercise 7.3.7: A set \(S \subset X\) is said to be dense in \(X\) if \(X \subset \bar{S}\) or in other words if for every \(x \in X\), there exists a sequence \(\{x_n\}\) in \(S\) that converges to \(x\). Prove that \(\mathbb{R}^n\) contains a countable dense subset.

Exercise 7.3.8 (Tricky): Suppose \(\{U_n\}_{n=1}^\infty\) is a decreasing \((U_{n+1} \subset U_n\) for all \(n\)) sequence of open sets in a metric space \((X,d)\) such that \(\bigcap_{n=1}^\infty U_n = \{p\}\) for some \(p \in X\). Suppose \(\{x_n\}\) is a sequence of points in \(X\) such that \(x_n \in U_n\). Does \(\{x_n\}\) necessarily converge to \(p\)? Prove or construct a counterexample.

Exercise 7.3.9: Let \(E \subset X\) be closed and let \(\{x_n\}\) be a sequence in \(X\) converging to \(p \in X\). Suppose \(x_n \in E\) for infinitely many \(n \in \mathbb{N}\). Show \(p \in E\).

Exercise 7.3.10: Take \(\mathbb{R}^* = \{-\infty\} \cup \mathbb{R} \cup \{\infty\}\) be the extended reals. Define \(d(x,y) := \left|\frac{x}{1+|x|} - \frac{y}{1+|y|}\right|\) if \(x, y \in \mathbb{R}\), define \(d(\infty, x) := \left|1 - \frac{x}{1+|x|}\right|\), \(d(-\infty, x) := \left|1 + \frac{x}{1+|x|}\right|\) for all \(x \in \mathbb{R}\), and let \(d(\infty, -\infty) := 2\).

a) Show that \((\mathbb{R}^*,d)\) is a metric space.

b) Suppose \(\{x_n\}\) is a sequence of real numbers such that for every \(M \in \mathbb{R}\), there exists an \(N\) such that \(x_n \geq M\) for all \(n \geq N\). Show that \(\lim x_n = \infty\) in \((\mathbb{R}^*,d)\).

c) Show that a sequence of real numbers converges to a real number in \((\mathbb{R}^*,d)\) if and only if it converges in \(\mathbb{R}\) with the standard metric.

Exercise 7.3.11: Suppose \(\{V_n\}_{n=1}^\infty\) is a sequence of open sets in \((X,d)\) such that \(V_{n+1} \supset V_n\) for all \(n\). Let \(\{x_n\}\) be a sequence such that \(x_n \in V_{n+1} \setminus V_n\) and suppose \(\{x_n\}\) converges to \(p \in X\). Show that \(p \in \partial V\) where \(V = \bigcup_{n=1}^\infty V_n\).
Exercise 7.3.12: Prove Proposition 7.3.6.

Exercise 7.3.13: Let \((X,d)\) be a metric space and \(\{x_n\}\) a sequence in \(X\). Prove that \(\{x_n\}\) converges to \(p \in X\) if and only if every subsequence of \(\{x_n\}\) has a subsequence that converges to \(p\).

Exercise 7.3.14: Consider \(\mathbb{R}^n\), and let \(d\) be the standard euclidean metric. Let \(d'(x,y) := \sum_{\ell=1}^{n} |x_\ell - y_\ell|\) and \(d''(x,y) := \max\{|x_1 - y_1|, |x_2 - y_2|, \ldots, |x_n - y_n|\}\).

a) Use Exercise 7.1.6, to show that \((\mathbb{R}^n, d')\) and \((\mathbb{R}^n, d'')\) are metric spaces.

b) Let \(\{x_j\}_{j=1}^{\infty}\) be a sequence in \(\mathbb{R}^n\) and \(p \in \mathbb{R}^n\). Prove that the following statements are equivalent:
   1. \(\{x_j\}\) converges to \(p\) in \((\mathbb{R}^n, d)\).
   2. \(\{x_j\}\) converges to \(p\) in \((\mathbb{R}^n, d')\).
   3. \(\{x_j\}\) converges to \(p\) in \((\mathbb{R}^n, d'')\).
7.4 Completeness and compactness

Note: 2 lectures

7.4.1 Cauchy sequences and completeness

Just like with sequences of real numbers, we define Cauchy sequences.

**Definition 7.4.1.** Let \((X,d)\) be a metric space. A sequence \(\{x_n\}\) in \(X\) is a **Cauchy sequence** if for every \(\varepsilon > 0\), there exists an \(M \in \mathbb{N}\) such that for all \(n \geq M\) and all \(k \geq M\), we have

\[
d(x_n, x_k) < \varepsilon.
\]

The definition is again simply a translation of the concept from the real numbers to metric spaces. A sequence of real numbers is Cauchy in the sense of chapter 2 if and only if it is Cauchy in the sense above, provided we equip the real numbers with the standard metric \(d(x, y) = |x - y|\).

**Proposition 7.4.2.** A convergent sequence in a metric space is Cauchy.

**Proof.** Suppose \(\{x_n\}\) converges to \(x\). Given \(\varepsilon > 0\), there is an \(M\) such that for all \(n \geq M\), we have \(d(x_n, x) < \varepsilon/2\). Hence for all \(n, k \geq M\), we have \(d(x_n, x_k) \leq d(x_n, x) + d(x, x_k) < \varepsilon/2 + \varepsilon/2 = \varepsilon\).

**Definition 7.4.3.** Let \((X,d)\) be a metric space. We say \(X\) is **complete** or **Cauchy-complete** if every Cauchy sequence \(\{x_n\}\) in \(X\) converges to an \(x \in X\).

**Proposition 7.4.4.** The space \(\mathbb{R}^n\) with the standard metric is a complete metric space.

For \(\mathbb{R} = \mathbb{R}^1\), completeness was proved in chapter 2. The proof of completeness in \(\mathbb{R}^n\) is a reduction to the one-dimensional case.

**Proof.** Let \(\{x_j\}^\infty_{j=1}\) be a Cauchy sequence in \(\mathbb{R}^n\), where we write \(x_j = (x_{j,1}, x_{j,2}, \ldots, x_{j,n}) \in \mathbb{R}^n\). As the sequence is Cauchy, given \(\varepsilon > 0\), there exists an \(M\) such that for all \(i, j \geq M\),

\[
d(x_i, x_j) < \varepsilon.
\]

Fix some \(k = 1, 2, \ldots, n\). For \(i, j \geq M\),

\[
|x_{i,k} - x_{j,k}| = \sqrt{(x_{i,k} - x_{j,k})^2} \leq \sqrt{\sum_{\ell=1}^{n} (x_{i,\ell} - x_{j,\ell})^2} = d(x_i, x_j) < \varepsilon.
\]

Hence the sequence \(\{x_{j,k}\}^\infty_{j=1}\) is Cauchy. As \(\mathbb{R}\) is complete the sequence converges; there exists a \(y_k \in \mathbb{R}\) such that \(y_k = \lim_{j \to \infty} x_{j,k}\). Write \(y = (y_1, y_2, \ldots, y_n) \in \mathbb{R}^n\). By Proposition 7.3.9, \(\{x_j\}\) converges to \(y \in \mathbb{R}^n\), and hence \(\mathbb{R}^n\) is complete.

A subset of \(\mathbb{R}^n\) with the subspace metric need not be complete. For example, \((0, 1]\) with the subspace metric is not complete as \(\{1/n\}\) is a Cauchy sequence in \((0, 1]\) with no limit in \((0, 1]\).

In the language of metric spaces, the results on continuity of section §6.2, say that the metric space \(C([a, b], \mathbb{R})\) of Example 7.1.8 is complete. The proof follows by “unrolling the definitions,” and is left as Exercise 7.4.7.
Proposition 7.4.5. The space of continuous functions $C([a, b], \mathbb{R})$ with the uniform norm as metric is a complete metric space.

As we saw above, a subspace of a complete metric space need not be complete. However, a closed subspace of a complete metric space is complete. After all, one way to think of a closed set is that it contains all points reachable from the set via a sequence. The proof is Exercise 7.4.16.

Proposition 7.4.6. Suppose $(X, d)$ is a complete metric space and $E \subset X$ is closed, then $E$ is a complete metric space with the subspace topology.

7.4.2 Compactness

Definition 7.4.7. Let $(X, d)$ be a metric space and $K \subset X$. The set $K$ is said to be compact if for every collection of open sets $\{U_\lambda\}_{\lambda \in I}$ such that

$$K \subset \bigcup_{\lambda \in I} U_\lambda,$$

there exists a finite subset $\{\lambda_1, \lambda_2, \ldots, \lambda_k\} \subset I$ such that

$$K \subset \bigcup_{j=1}^k U_{\lambda_j}.$$

A collection of open sets $\{U_\lambda\}_{\lambda \in I}$ as above is said to be an open cover of $K$. A way to say that $K$ is compact is to say that every open cover of $K$ has a finite subcover.

Example 7.4.8: Let $\mathbb{R}$ be the metric space with the standard metric.

The set $\mathbb{R}$ is not compact. Proof: Take the sets $U_j := (-j, j)$. Any $x \in \mathbb{R}$ is in some $U_j$ (by the Archimedean property), so we have an open cover. Suppose we have a finite subcover $\mathbb{R} \subset U_{j_1} \cup U_{j_2} \cup \cdots \cup U_{j_k}$, and suppose $j_1 < j_2 < \cdots < j_k$. Then $\mathbb{R} \subset U_{j_k}$, but that is a contradiction as $j_k \in \mathbb{R}$ on one hand and $j_k \notin U_{j_k} = (-j_k, j_k)$ on the other.

The set $(0, 1) \subset \mathbb{R}$ is also not compact. Proof: Take the sets $U_j := (1/j, 1 - 1/j)$ for $j = 3, 4, 5, \ldots$. As above $(0, 1) = \bigcup_{j=3}^\infty U_j$. And similarly as above, if there exists a finite subcover, then there is one $U_j$ such that $(0, 1) \subset U_j$, which again leads to a contradiction.

The set $\{0\} \subset \mathbb{R}$ is compact. Proof: Given an open cover $\{U_\lambda\}_{\lambda \in I}$, there must exist a $\lambda_0$ such that $0 \in U_{\lambda_0}$ as it is a cover. But then $U_{\lambda_0}$ gives a finite subcover.

We will prove below that $[0, 1]$, and in fact every closed and bounded interval $[a, b]$, is compact.

Proposition 7.4.9. Let $(X, d)$ be a metric space. A compact set $K \subset X$ is closed and bounded.

Proof. First, we prove that a compact set is bounded. Fix $p \in X$. We have the open cover

$$K \subset \bigcup_{n=1}^\infty B(p, n) = X.$$

If $K$ is compact, then there exists some set of indices $n_1 < n_2 < \cdots < n_k$ such that

$$K \subset \bigcup_{j=1}^k B(p, n_j) = B(p, n_k).$$
As $K$ is contained in a ball, $K$ is bounded. See the left-hand side of Figure 7.10.

Next, we show a set that is not closed is not compact. Suppose $\overline{K} \neq K$, that is, there is a point $x \in \overline{K} \setminus K$. If $y \neq x$, then $y \notin C(x, 1/n)$ for $n \in \mathbb{N}$ such that $1/n < d(x, y)$. Furthermore, $x \notin K$, so

$$K \subset \bigcup_{n=1}^{\infty} C(x, 1/n)^c.$$  

A closed ball is closed, so its complement $C(x, 1/n)^c$ is open, and we have an open cover. If we take any finite collection of indices $n_1 < n_2 < \ldots < n_k$, then

$$\bigcup_{j=1}^{k} C(x, 1/n_j)^c = C(x, 1/n_k)^c.$$  

As $x$ is in the closure of $K$, then $C(x, 1/n_k) \cap K \neq \emptyset$. So there is no finite subcover and $K$ is not compact. See the right-hand side of Figure 7.10.  

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**Figure 7.10**: Proving compact set is bounded (left) and closed (right).

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We prove below that in a finite-dimensional euclidean space, every closed bounded set is compact. So closed bounded sets of $\mathbb{R}^n$ are examples of compact sets. It is not true that in every metric space, closed and bounded is equivalent to compact. A simple example is an incomplete metric space such as $(0, 1)$ with the subspace metric from $\mathbb{R}$. There are many complete and very useful metric spaces where closed and bounded is not enough to give compactness: $C([a, b], \mathbb{R})$ is a complete metric space, but the closed unit ball $C(0, 1)$ is not compact, see Exercise 7.4.8. However, see also Exercise 7.4.12.

A useful property of compact sets in a metric space is that every sequence in the set has a convergent subsequence converging to a point in the set. Such sets are called *sequentially compact*. Let us prove that in the context of metric spaces, a set is compact if and only if it is sequentially compact. First we prove a lemma.

**Lemma 7.4.10** (*Lebesgue covering lemma*). Let $(X, d)$ be a metric space and $K \subset X$. Suppose every sequence in $K$ has a subsequence convergent in $K$. Given an open cover $\{U_\lambda\}_{\lambda \in I}$ of $K$, there exists a $\delta > 0$ such that for every $x \in K$, there exists a $\lambda \in I$ with $B(x, \delta) \subset U_\lambda$.

---

*Named after the French mathematician Henri Léon Lebesgue (1875–1941). The number $\delta$ is sometimes called the Lebesgue number of the cover.*
Proof. We prove the lemma by contrapositive. If the conclusion is not true, then there is an open cover \( \{U_\lambda\}_{\lambda \in I} \) of \( K \) with the following property. For every \( n \in \mathbb{N} \), there exists an \( x_n \in K \) such that \( B(x_n, 1/n) \) is not a subset of any \( U_\lambda \). Take any \( x \in K \). There is a \( \lambda \in I \) such that \( x \in U_\lambda \). As \( U_\lambda \) is open, there is an \( \varepsilon > 0 \) such that \( B(x, \varepsilon) \subseteq U_\lambda \). Take \( M \) such that \( 1/M < \varepsilon/2 \). If \( y \in B(x, \varepsilon/2) \) and \( n \geq M \), then

\[
B(y, 1/n) \subseteq B(y, 1/M) \subseteq B(y, \varepsilon/2) \subseteq B(x, \varepsilon) \subseteq U_\lambda,
\]

where \( B(y, \varepsilon/2) \subseteq B(x, \varepsilon) \) follows by triangle inequality. See Figure 7.11. Thus \( y \neq x_n \). In other words, for all \( n \geq M \), \( x_n \notin B(x, \varepsilon/2) \). The sequence cannot have a subsequence converging to \( x \). As \( x \in K \) was arbitrary we are done.

Figure 7.11: Proof of Lebesgue covering lemma. Note that \( B(y, \varepsilon/2) \subseteq B(x, \varepsilon) \) by triangle inequality.

It is important to recognize what the lemma says. It says that if \( K \) is sequentially compact, then given any cover there is a single \( \delta > 0 \). The \( \delta \) depends on the cover, but, of course, it does not depend on \( x \).

For example, let \( K := [-10, 10] \) and for \( n \in \mathbb{Z} \) let \( U_n := (n, n+2) \) define sets in an open cover. Take \( x \in K \). There is an \( n \in \mathbb{Z} \), such that \( n \leq x < n + 1 \). If \( n \leq x < n + 1/2 \), then \( B(x, 1/2) \subseteq U_{n-1} \). If \( n + 1/2 \leq x < n + 1 \), then \( B(x, 1/2) \subseteq U_n \). So \( \delta = 1/2 \). If instead we let \( U'_n := (n, n+2/3) \), then we again obtain an open cover, but now the best \( \delta \) we can find is \( 1/4 \).

On the other hand, \( \mathbb{N} \subset \mathbb{R} \) is not sequentially compact. It is an exercise to find a cover for which no \( \delta > 0 \) works.

**Theorem 7.4.11.** Let \( (X, d) \) be a metric space. Then \( K \subset X \) is compact if and only if every sequence in \( K \) has a subsequence converging to a point in \( K \).

Proof. Claim: Let \( K \subset X \) be a subset of \( X \) and \( \{x_n\} \) a sequence in \( K \). Suppose that for each \( x \in K \), there is a ball \( B(x, \alpha_x) \) for some \( \alpha_x > 0 \) such that \( x_n \in B(x, \alpha_x) \) for only finitely many \( n \in \mathbb{N} \). Then \( K \) is not compact.

Proof of the claim: Notice

\[
K \subset \bigcup_{x \in K} B(x, \alpha_x).
\]

Any finite collection of these balls contains at most finitely many elements of \( \{x_n\} \), and so there must be an \( x_n \in K \) not in their union. Therefore, \( K \) is not compact and the claim is proved.
7.4. COMPLETENESS AND COMPACTNESS

So suppose that $K$ is compact and $\{x_n\}$ is a sequence in $K$. Then there exists an $x \in K$ such that for all $\delta > 0$, $B(x, \delta)$ contains $x_k$ for infinitely many $k \in \mathbb{N}$. We define the subsequence inductively. The ball $B(x, 1)$ contains some $x_k$, so let $n_1 := k$. Suppose $n_{j-1}$ is defined. There must exist a $k > n_{j-1}$ such that $x_k \in B(x, 1/j)$. Define $n_j := k$. We now possess a subsequence $\{x_{n_j}\}_{j=1}^\infty$. Since $d(x, x_{n_j}) < 1/j$, Proposition 7.3.5 says $\lim x_{n_j} = x$.

For the other direction, suppose every sequence in $K$ has a subsequence converging in $K$. Take an open cover $\{U_\lambda\}_{\lambda \in I}$ of $K$. Using the Lebesgue covering lemma above, find a $\delta > 0$ such that for every $x \in K$, there is a $\lambda \in I$ with $B(x, \delta) \subset U_\lambda$.

Pick $x_1 \in K$ and find $\lambda_1 \in I$ such that $B(x_1, \delta) \subset U_{\lambda_1}$. If $K \subset U_{\lambda_1}$, we stop as we have found a finite subcover. Otherwise, there must be a point $x_2 \in K \setminus U_{\lambda_1}$. Note that $d(x_2, x_1) \geq \delta$. There must exist some $\lambda_2 \in I$ such that $B(x_2, \delta) \subset U_{\lambda_2}$. We work inductively. Suppose $\lambda_{n-1}$ is defined. Either $U_{\lambda_1} \cup U_{\lambda_2} \cup \cdots \cup U_{\lambda_{n-1}}$ is a finite cover of $K$, in which case we stop, or there must be a point $x_n \in K \setminus (U_{\lambda_1} \cup U_{\lambda_2} \cup \cdots \cup U_{\lambda_{n-1}})$. Note that $d(x_n, x_j) \geq \delta$ for all $j = 1, 2, \ldots, n-1$. Next, there must be some $\lambda_n \in I$ such that $B(x_n, \delta) \subset U_{\lambda_n}$. See Figure 7.12.

![Figure 7.12: Covering $K$ by $U_\lambda$. The points $x_1, x_2, x_3, x_4$, the three sets $U_{\lambda_1}, U_{\lambda_2}, U_{\lambda_3}$, and the first three balls of radius $\delta$ are drawn.](image)

Either at some point we obtain a finite subcover of $K$, or we obtain an infinite sequence $\{x_n\}$ as above. For contradiction, suppose that there is no finite subcover and we have the sequence $\{x_n\}$. For all $n$ and $k$, $n \neq k$, we have $d(x_n, x_k) \geq \delta$, so no subsequence of $\{x_n\}$ can be Cauchy. Hence, no subsequence of $\{x_n\}$ can be convergent, which is a contradiction.

\begin{example}

The Bolzano–Weierstrass theorem for sequences of real numbers (Theorem 2.3.8) says that every bounded sequence in $\mathbb{R}$ has a convergent subsequence. Therefore, every sequence in a closed interval $[a, b] \subset \mathbb{R}$ has a convergent subsequence. The limit is also in $[a, b]$ as limits preserve non-strict inequalities. Hence a closed bounded interval $[a, b] \subset \mathbb{R}$ is (sequentially) compact.

\end{example}

\begin{proposition}

Let $(X, d)$ be a metric space and let $K \subset X$ be compact. If $E \subset K$ is a closed set, then $E$ is compact.

Because $K$ is closed, $E$ is closed in $K$ if and only if it is closed in $X$. See Proposition 7.2.12.

\begin{proof}

Let $\{x_n\}$ be a sequence in $E$. It is also a sequence in $K$. Therefore, it has a convergent subsequence $\{x_{n_j}\}$ that converges to some $x \in K$. As $E$ is closed the limit of a sequence in $E$ is also in $E$ and so $x \in E$. Thus $E$ must be compact.

\end{proof}

Theorem 7.4.14 (Heine–Borel\textsuperscript{*}). A closed bounded subset \( K \subset \mathbb{R}^n \) is compact.

So subsets of \( \mathbb{R}^n \) are compact if and only if they are closed and bounded, a condition that is much easier to check. Let us reiterate that the Heine–Borel theorem only holds for \( \mathbb{R}^n \) and not for metric spaces in general. In general, compact implies closed and bounded, but not vice versa.

Proof. For \( \mathbb{R} = \mathbb{R}^1 \) if \( K \subset \mathbb{R} \) is closed and bounded, then every sequence \( \{x_k\} \) in \( K \) is bounded, so it has a convergent subsequence by Bolzano–Weierstrass theorem (Theorem 2.3.8). As \( K \) is closed, the limit of the subsequence must be an element of \( K \). So \( K \) is compact. Let us carry out the proof for \( n = 2 \) and leave arbitrary \( n \) as an exercise. As \( K \subset \mathbb{R}^2 \) is bounded, there exists a set \( B = [a, b] \times [c, d] \subset \mathbb{R}^2 \) such that \( K \subset B \). We will show that \( B \) is compact. Then \( K \), being a closed subset of a compact \( B \), is also compact.

Let \( \{(x_k, y_k)\}_{k=1}^{\infty} \) be a sequence in \( B \). That is, \( a \leq x_k \leq b \) and \( c \leq y_k \leq d \) for all \( k \). A bounded sequence of real numbers has a convergent subsequence so there is a subsequence \( \{x_{k_j}\}_{j=1}^{\infty} \) that is convergent. The subsequence \( \{y_{k_j}\}_{j=1}^{\infty} \) is also a bounded sequence so there exists a subsequence \( \{y_{k_{ji}}\}_{i=1}^{\infty} \) that is convergent. A subsequence of a convergent sequence is still convergent, so \( \{x_{k_{ji}}\}_{i=1}^{\infty} \) is convergent. Let 
\[
  x := \lim_{i \to \infty} x_{k_{ji}} \quad \text{and} \quad y := \lim_{i \to \infty} y_{k_{ji}}.
\]

By Proposition 7.3.9, \( \{(x_{k_{ji}}, y_{k_{ji}})\}_{i=1}^{\infty} \) converges to \((x, y)\). Furthermore, as \( a \leq x_k \leq b \) and \( c \leq y_k \leq d \) for all \( k \), we know that \((x, y) \in B\). \( \square \)

Example 7.4.15: The discrete metric provides interesting counterexamples again. Let \((X, d)\) be a metric space with the discrete metric, that is, \( d(x, y) = 1 \) if \( x \neq y \). Suppose \( X \) is an infinite set. Then

(i) \((X, d)\) is a complete metric space.

(ii) Any subset \( K \subset X \) is closed and bounded.

(iii) A subset \( K \subset X \) is compact if and only if it is a finite set.

(iv) The conclusion of the Lebesgue covering lemma is always satisfied with e.g. \( \delta = 1/2 \), even for noncompact \( K \subset X \).

The proofs of the statements above are either trivial or are relegated to the exercises below.

Remark 7.4.16. A subtle issue with Cauchy sequences, completeness, compactness, and convergence is that compactness and convergence only depend on the topology, that is, on which sets are the open sets. On the other hand, Cauchy sequences and completeness depend on the actual metric. See Exercise 7.4.19.

7.4.3 Exercises

Exercise 7.4.1: Let \((X, d)\) be a metric space and \( A \) a finite subset of \( X \). Show that \( A \) is compact.

Exercise 7.4.2: Let \( A := \{1/n : n \in \mathbb{N}\} \subset \mathbb{R} \).

a) Show that \( A \) is not compact directly using the definition.

b) Show that \( A \cup \{0\} \) is compact directly using the definition.

\textsuperscript{*}Named after the German mathematician Heinrich Eduard Heine (1821–1881), and the French mathematician Félix Édouard Justin Émile Borel (1871–1956).
Exercise 7.4.3: Let \((X, d)\) be a metric space with the discrete metric.

a) Prove that \(X\) is complete.

b) Prove that \(X\) is compact if and only if \(X\) is a finite set.

Exercise 7.4.4:

a) Show that the union of finitely many compact sets is a compact set.

b) Find an example where the union of infinitely many compact sets is not compact.

Exercise 7.4.5: Prove Theorem 7.4.14 for arbitrary dimension. Hint: The trick is to use the correct notation.

Exercise 7.4.6: Show that a compact set \(K\) (in any metric space) is itself a complete metric space (using the subspace metric).

Exercise 7.4.7: Let \(C([a, b], \mathbb{R})\) be the metric space as in Example 7.1.8. Show that \(C([a, b], \mathbb{R})\) is a complete metric space.

Exercise 7.4.8 (Challenging): Let \(C([0, 1], \mathbb{R})\) be the metric space of Example 7.1.8. Let \(0\) denote the zero function. Then show that the closed ball \(C(0, 1)\) is not compact (even though it is closed and bounded). Hints: Construct a sequence of distinct continuous functions \(\{f_n\}\) such that \(d(f_n, 0) = 1\) and \(d(f_n, f_k) = 1\) for all \(n \neq k\). Show that the set \(\{f_n : n \in \mathbb{N}\} \subset C(0, 1)\) is closed but not compact. See Chapter 6 for inspiration.

Exercise 7.4.9 (Challenging): Show that there exists a metric on \(\mathbb{R}\) that makes \(\mathbb{R}\) into a compact set.

Exercise 7.4.10: Suppose \((X, d)\) is complete and suppose we have a countably infinite collection of nonempty compact sets \(E_1 \supset E_2 \supset E_3 \supset \cdots\). Prove \(\bigcap_{j=1}^{\infty} E_j \neq \emptyset\).

Exercise 7.4.11 (Challenging): Let \(C([0, 1], \mathbb{R})\) be the metric space of Example 7.1.8. Let \(K\) be the set of \(f \in C([0, 1], \mathbb{R})\) such that \(f\) is equal to a quadratic polynomial, i.e. \(f(x) = a + bx + cx^2\), and such that \(|f(x)| \leq 1\) for all \(x \in [0, 1]\), that is \(f \in C(0, 1)\). Show that \(K\) is compact.

Exercise 7.4.12 (Challenging): Let \((X, d)\) be a complete metric space. Show that \(K \subset X\) is compact if and only if \(K\) is closed and such that for every \(\varepsilon > 0\) there exists a finite set of points \(x_1, x_2, \ldots, x_n\) with \(K \subset \bigcup_{j=1}^{n} B(x_j, \varepsilon)\). Note: Such a set \(K\) is said to be totally bounded, so in a complete metric space a set is compact if and only if it is closed and totally bounded.

Exercise 7.4.13: Take \(\mathbb{N} \subset \mathbb{R}\) using the standard metric. Find an open cover of \(\mathbb{N}\) such that the conclusion of the Lebesgue covering lemma does not hold.

Exercise 7.4.14: Prove the general Bolzano–Weierstrass theorem: Any bounded sequence \(\{x_k\}\) in \(\mathbb{R}^n\) has a convergent subsequence.

Exercise 7.4.15: Let \(X\) be a metric space and \(C \subset \mathcal{P}(X)\) the set of nonempty compact subsets of \(X\). Using the Hausdorff metric from Exercise 7.1.8, show that \((C, d_H)\) is a metric space. That is, show that if \(L\) and \(K\) are nonempty compact subsets, then \(d_H(L, K) = 0\) if and only if \(L = K\).

Exercise 7.4.16: Prove Proposition 7.4.6. That is, let \((X, d)\) be a complete metric space and \(E \subset X\) a closed set. Show that \(E\) with the subspace metric is a complete metric space.

Exercise 7.4.17: Let \((X, d)\) be an incomplete metric space. Show that there exists a closed and bounded set \(E \subset X\) that is not compact.
**Exercise 7.4.18**: Let \((X,d)\) be a metric space and \(K \subset X\). Prove that \(K\) is compact as a subset of \((X,d)\) if and only if \(K\) is compact as a subset of itself with the subspace metric.

**Exercise 7.4.19**: Consider two metrics on \(\mathbb{R}\). Let \(d(x,y) := |x - y|\) be the standard metric, and let \(d'(x,y) := \left| \frac{x}{1+|x|} - \frac{y}{1+|y|} \right|\).

a) Show that \((\mathbb{R},d')\) is a metric space (if you have done Exercise 7.3.10, the computation is the same).
b) Show that the topology is the same, that is, a set is open in \((\mathbb{R},d)\) if and only if it is open in \((\mathbb{R},d')\).
c) Show that a set is compact in \((\mathbb{R},d)\) if and only if it is compact in \((\mathbb{R},d')\).
d) Show that a sequence converges in \((\mathbb{R},d)\) if and only if it converges in \((\mathbb{R},d')\).
e) Find a sequence of real numbers that is Cauchy in \((\mathbb{R},d')\) but not Cauchy in \((\mathbb{R},d)\).
f) While \((\mathbb{R},d)\) is complete, show that \((\mathbb{R},d')\) is not complete.

**Exercise 7.4.20**: Let \((X,d)\) be a complete metric space. We say a set \(S \subset X\) is relatively compact if the closure \(\overline{S}\) is compact. Prove that \(S \subset X\) is relatively compact if and only if given any sequence \(\{x_n\}\) in \(S\), there exists a subsequence \(\{x_{n_k}\}\) that converges (in \(X\)).
7.5 Continuous functions

Note: 1.5–2 lectures

7.5.1 Continuity

Definition 7.5.1. Let $(X,d_X)$ and $(Y,d_Y)$ be metric spaces and $c \in X$. Then $f: X \to Y$ is continuous at $c$ if for every $\varepsilon > 0$ there is a $\delta > 0$ such that whenever $x \in X$ and $d_X(x,c) < \delta$, then $d_Y(f(x),f(c)) < \varepsilon$.

When $f: X \to Y$ is continuous at all $c \in X$, then we simply say that $f$ is a continuous function.

The definition agrees with the definition from chapter 3 when $f$ is a real-valued function on the real line—as long as we take the standard metric on $\mathbb{R}$, of course.

Proposition 7.5.2. Let $(X,d_X)$ and $(Y,d_Y)$ be metric spaces. Then $f: X \to Y$ is continuous at $c \in X$ if and only if for every sequence $\{x_n\}$ in $X$ converging to $c$, the sequence $\{f(x_n)\}$ converges to $f(c)$.

Proof. Suppose $f$ is continuous at $c$. Let $\{x_n\}$ be a sequence in $X$ converging to $c$. Given $\varepsilon > 0$, there is a $\delta > 0$ such that $d_X(x,c) < \delta$ implies $d_Y(f(x),f(c)) < \varepsilon$. So take $M$ such that for all $n \geq M$, we have $d_X(x_n,c) < \delta$, then $d_Y(f(x_n),f(c)) < \varepsilon$. Hence $\{f(x_n)\}$ converges to $f(c)$.

On the other hand, suppose $f$ is not continuous at $c$. Then there exists an $\varepsilon > 0$, such that for every $n \in \mathbb{N}$ there exists an $x_n \in X$, with $d_X(x_n,c) < 1/n$ such that $d_Y(f(x_n),f(c)) \geq \varepsilon$. Then $\{x_n\}$ converges to $c$, but $\{f(x_n)\}$ does not converge to $f(c)$.

Example 7.5.3: Suppose $f: \mathbb{R}^2 \to \mathbb{R}$ is a polynomial. That is,

$$f(x,y) = \sum_{j=0}^{d} \sum_{k=0}^{d-j} a_{jk} x^j y^k = a_0 + a_1 x + a_0 x + a_2 x^2 + a_{11} xy + a_{02} y^2 + \cdots + a_{0d} y^d,$$

for some $d \in \mathbb{N}$ (the degree) and $a_{jk} \in \mathbb{R}$. Then we claim $f$ is continuous. Let $\{(x_n,y_n)\}_{n=1}^{\infty}$ be a sequence in $\mathbb{R}^2$ that converges to $(x,y) \in \mathbb{R}^2$. We proved that this means $\lim x_n = x$ and $\lim y_n = y$. By Proposition 2.2.5, we have

$$\lim_{n \to \infty} f(x_n,y_n) = \lim_{n \to \infty} \sum_{j=0}^{d} \sum_{k=0}^{d-j} a_{jk} x_n^j y_n^k = \sum_{j=0}^{d} \sum_{k=0}^{d-j} a_{jk} x^j y^k = f(x,y).$$

So $f$ is continuous at $(x,y)$, and as $(x,y)$ was arbitrary $f$ is continuous everywhere. Similarly, a polynomial in $n$ variables is continuous.

Be careful about taking limits separately. Consider the function defined by $f(x,y) := \frac{xy}{x^2+y^2}$ outside the origin and $f(0,0) := 0$. See Figure 7.13. In Exercise 7.5.2, you are asked to prove that $f$ is not continuous at the origin. However, for every $y$, the function $g(x) := f(x,y)$ is continuous, and for every $x$, the function $h(y) := f(x,y)$ is continuous.

Example 7.5.4: Let $X$ be a metric space and $f: X \to \mathbb{C}$ a complex-valued function. We write $f(p) = g(p) + ih(p)$, where $g: X \to \mathbb{R}$ and $h: X \to \mathbb{R}$ are the real and imaginary parts of $f$. Then $f$ is continuous at $c \in X$ if and only if $g$ and $h$ are continuous at $c$. This fact follows because $\{f(p_n) = g(p_n) + ih(p_n)\}_{n=1}^{\infty}$ converges to $f(p) = g(p) + ih(p)$ if and only if $\{g(p_n)\}$ converges to $g(p)$ and $\{h(p_n)\}$ converges to $h(p)$.
7.5.2 Compactness and continuity

Continuous maps do not map closed sets to closed sets. For example, \( f : (0, 1) \to \mathbb{R} \) defined by \( f(x) := x \) takes the set \((0, 1)\), which is closed in \((0, 1)\), to the set \((0, 1)\), which is not closed in \(\mathbb{R}\). On the other hand, continuous maps do preserve compact sets.

**Lemma 7.5.5.** Let \((X, d_X)\) and \((Y, d_Y)\) be metric spaces and \(f : X \to Y\) a continuous function. If \(K \subset X\) is a compact set, then \(f(K)\) is a compact set.

**Proof.** A sequence in \(f(K)\) can be written as \(\{f(x_n)\}_{n=1}^{\infty}\), where \(\{x_n\}_{n=1}^{\infty}\) is a sequence in \(K\). The set \(K\) is compact and therefore there is a subsequence \(\{x_{n_j}\}_{j=1}^{\infty}\) that converges to some \(x \in K\). By continuity,

\[
\lim_{j \to \infty} f(x_{n_j}) = f(x) \in f(K).
\]

So every sequence in \(f(K)\) has a subsequence convergent to a point in \(f(K)\), and \(f(K)\) is compact by Theorem 7.4.11.

As before, \(f : X \to \mathbb{R}\) achieves an **absolute minimum** at \(c \in X\) if

\[
f(x) \geq f(c) \quad \text{for all } x \in X.
\]

On the other hand, \(f\) achieves an **absolute maximum** at \(c \in X\) if

\[
f(x) \leq f(c) \quad \text{for all } x \in X.
\]

**Theorem 7.5.6.** Let \((X, d)\) be a nonempty compact metric space and let \(f : X \to \mathbb{R}\) be continuous. Then \(f\) is bounded and in fact \(f\) achieves an absolute minimum and an absolute maximum on \(X\).

**Proof.** As \(X\) is compact and \(f\) is continuous, \(f(X) \subset \mathbb{R}\) is compact. Hence \(f(X)\) is closed and bounded. In particular, \(\sup f(X) \in f(X)\) and \(\inf f(X) \in f(X)\), because both the sup and the inf can be achieved by sequences in \(f(X)\) and \(f(X)\) is closed. Therefore, there is some \(x \in X\) such that \(f(x) = \sup f(X)\) and some \(y \in X\) such that \(f(y) = \inf f(X)\).
7.5.3 Continuity and topology

Let us see how to define continuity in terms of the topology, that is, the open sets. We have already seen that topology determines which sequences converge, and so it is no wonder that the topology also determines continuity of functions.

**Lemma 7.5.7.** Let \((X,d_X)\) and \((Y,d_Y)\) be metric spaces. A function \(f : X \to Y\) is continuous at \(c \in X\) if and only if for every open neighborhood \(U\) of \(f(c)\) in \(Y\), the set \(f^{-1}(U)\) contains an open neighborhood of \(c\) in \(X\). See Figure 7.14.

![Figure 7.14](image)

**Figure 7.14:** For every neighborhood \(U\) of \(f(c)\), the set \(f^{-1}(U)\) contains an open neighborhood \(W\) of \(c\).

**Proof.** First suppose that \(f\) is continuous at \(c\). Let \(U\) be an open neighborhood of \(f(c)\) in \(Y\), then \(B_Y(f(c),\varepsilon) \subset U\) for some \(\varepsilon > 0\). By continuity of \(f\), there exists a \(\delta > 0\) such that whenever \(x\) is such that \(d_X(x,c) < \delta\), then \(d_Y(f(x),f(c)) < \varepsilon\). In other words,

\[
B_X(c,\delta) \subset f^{-1}(B_Y(f(c),\varepsilon)) \subset f^{-1}(U),
\]

and \(B_X(c,\delta)\) is an open neighborhood of \(c\).

For the other direction, let \(\varepsilon > 0\) be given. If \(f^{-1}(B_Y(f(c),\varepsilon))\) contains an open neighborhood \(W\) of \(c\), it contains a ball. That is, there is some \(\delta > 0\) such that

\[
B_X(c,\delta) \subset W \subset f^{-1}(B_Y(f(c),\varepsilon)).
\]

That means precisely that if \(d_X(x,c) < \delta\), then \(d_Y(f(x),f(c)) < \varepsilon\), and so \(f\) is continuous at \(c\).

**Theorem 7.5.8.** Let \((X,d_X)\) and \((Y,d_Y)\) be metric spaces. A function \(f : X \to Y\) is continuous if and only if for every open \(U \subset Y\), \(f^{-1}(U)\) is open in \(X\).

The proof follows from **Lemma 7.5.7** and is left as an exercise.

**Example 7.5.9:** Let \(f : X \to Y\) be a continuous function. **Theorem 7.5.8** tells us that if \(E \subset Y\) is closed, then \(f^{-1}(E) = X \setminus f^{-1}(E^c)\) is also closed. Therefore, if we have a continuous function \(f : X \to \mathbb{R}\), then the zero set of \(f\), that is, \(f^{-1}(0) = \{x \in X : f(x) = 0\}\), is closed. We have just proved the most basic result in algebraic geometry, the study of zero sets of polynomials: The zero set of a polynomial is closed.

Similarly the set where \(f\) is nonnegative, that is, \(f^{-1}([0,\infty)) = \{x \in X : f(x) \geq 0\}\), is closed. On the other hand the set where \(f\) is positive, \(f^{-1}((0,\infty)) = \{x \in X : f(x) > 0\}\), is open.
7.5.4 Uniform continuity

As for continuous functions on the real line, in the definition of continuity it is sometimes convenient to be able to pick one \( \delta \) for all points.

**Definition 7.5.10.** Let \((X, d_X)\) and \((Y, d_Y)\) be metric spaces. Then \( f: X \to Y \) is uniformly continuous if for every \( \varepsilon > 0 \) there is a \( \delta > 0 \) such that whenever \( p, q \in X \) and \( d_X(p, q) < \delta \), we have \( d_Y(f(p), f(q)) < \varepsilon \).

A uniformly continuous function is continuous, but not necessarily vice versa as we have seen.

**Theorem 7.5.11.** Let \((X, d_X)\) and \((Y, d_Y)\) be metric spaces. Suppose \( f: X \to Y \) is continuous and \( X \) is compact. Then \( f \) is uniformly continuous.

**Proof.** Let \( \varepsilon > 0 \) be given. For each \( c \in X \), pick \( \delta_c > 0 \) such that \( d_Y(f(x), f(c)) < \varepsilon/2 \) whenever \( x \in B(c, \delta_c) \). The balls \( B(c, \delta_c) \) cover \( X \), and the space \( X \) is compact. Apply the Lebesgue covering lemma to obtain a \( \delta > 0 \) such that for every \( x \in X \), there is a \( c \in X \) for which \( B(x, \delta) \subset B(c, \delta_c) \).

If \( p, q \in X \) where \( d_X(p, q) < \delta \), find a \( c \in X \) such that \( B(p, \delta) \subset B(c, \delta_c) \). Then \( q \in B(c, \delta_c) \). By the triangle inequality and the definition of \( \delta_c \), we have

\[
d_Y(f(p), f(q)) \leq d_Y(f(p), f(c)) + d_Y(f(c), f(q)) < \varepsilon/2 + \varepsilon/2 = \varepsilon. \quad \blacksquare
\]

As an application of uniform continuity, let us prove a useful criterion for continuity of functions defined by integrals. Let \( f(x, y) \) be a function of two variables and define

\[
g(y) := \int_a^b f(x, y) \, dx.
\]

Question is, is \( g \) is continuous? We are really asking when do two limiting operations commute, which is not always possible, so some extra hypothesis is necessary. A useful sufficient (but not necessary) condition is that \( f \) is continuous on a closed rectangle.

**Proposition 7.5.12.** If \( f: [a, b] \times [c, d] \to \mathbb{R} \) is a continuous function, then \( g: [c, d] \to \mathbb{R} \) defined by

\[
g(y) := \int_a^b f(x, y) \, dx \quad \text{is continuous.}
\]

**Proof.** Fix \( y \in [c, d] \), and let \( \{y_n\} \) be a sequence in \([c, d]\) converging to \( y \). Let \( \varepsilon > 0 \) be given. As \( f \) is continuous on \([a, b] \times [c, d]\), which is compact, \( f \) is uniformly continuous. In particular, there exists a \( \delta > 0 \) such that whenever \( \tilde{y} \in [c, d] \) and \( |\tilde{y} - y| < \delta \), we have \( |f(x, \tilde{y}) - f(x, y)| < \frac{\varepsilon}{b-a} \) for all \( x \in [a, b] \). So suppose \( |\tilde{y} - y| < \delta \). Then

\[
|g(\tilde{y}) - g(y)| = \left| \int_a^b f(x, \tilde{y}) \, dx - \int_a^b f(x, y) \, dx \right|
= \left| \int_a^b \left( f(x, \tilde{y}) - f(x, y) \right) \, dx \right| \leq (b-a) \frac{\varepsilon}{b-a} = \varepsilon. \quad \blacksquare
\]

In applications, if we are interested in continuity at \( y_0 \), we just need to apply the proposition in \([a, b] \times [y_0 - \varepsilon, y_0 + \varepsilon]\) for some small \( \varepsilon > 0 \). For example, if \( f \) is continuous in \([a, b] \times \mathbb{R}\), then \( g \) is continuous on \( \mathbb{R} \).
7.5. CONTINUOUS FUNCTIONS

Example 7.5.13: Useful examples of uniformly continuous functions are again the so-called Lipschitz continuous functions. That is, if \((X,d_X)\) and \((Y,d_Y)\) are metric spaces, then \(f: X \to Y\) is called Lipschitz or \(K\)-Lipschitz if there exists a \(K \in \mathbb{R}\) such that

\[
d_Y(f(p), f(q)) \leq Kd_X(p, q) \quad \text{for all } p, q \in X.
\]

A Lipschitz function is uniformly continuous: Take \(\delta = \varepsilon / K\). A function can be uniformly continuous but not Lipschitz, as we already saw: \(\sqrt{x}\) on \([0, 1]\) is uniformly continuous but not Lipschitz.

It is worth mentioning that, if a function is Lipschitz, it tends to be easiest to simply show it is Lipschitz even if we are only interested in knowing continuity.

7.5.5 Cluster points and limits of functions

While we haven’t started the discussion of continuity with them and we won’t need them until volume II, let us also translate the idea of a limit of a function from the real line to metric spaces. Again we need to start with cluster points.

Definition 7.5.14. Let \((X,d)\) be a metric space and \(S \subset X\). A point \(p \in X\) is called a cluster point of \(S\) if for every \(\varepsilon > 0\), the set \(B(p, \varepsilon) \cap S \setminus \{p\}\) is not empty.

It is not enough that \(p\) is in the closure of \(S\), it must be in the closure of \(S \setminus \{p\}\) (exercise). So, \(p\) is a cluster point if and only if there exists a sequence in \(S \setminus \{p\}\) that converges to \(p\).

Definition 7.5.15. Let \((X,d_X)\), \((Y,d_Y)\) be metric spaces, \(S \subset X\), \(p \in X\) a cluster point of \(S\), and \(f: S \to Y\) a function. Suppose there exists an \(L \in Y\) and for every \(\varepsilon > 0\), there exists a \(\delta > 0\) such that whenever \(x \in S \setminus \{p\}\) and \(d_X(x,p) < \delta\), then

\[
d_Y(f(x), L) < \varepsilon.
\]

Then we say \(f(x)\) converges to \(L\) as \(x\) goes to \(p\), and \(L\) is the limit of \(f(x)\) as \(x\) goes to \(p\). We write

\[
\lim_{x \to p} f(x) := L.
\]

If \(f(x)\) does not converge as \(x\) goes to \(p\), we say \(f\) diverges at \(p\).

As usual, we used the definite article without showing that the limit is unique. The proof is a direct translation of the proof from chapter 3, so we leave it as an exercise.

Proposition 7.5.16. Let \((X,d_X)\) and \((Y,d_Y)\) be metric spaces, \(S \subset X\), \(p \in X\) a cluster point of \(S\), and let \(f: S \to Y\) be a function such that \(f(x)\) converges as \(x\) goes to \(p\). Then the limit of \(f(x)\) as \(x\) goes to \(p\) is unique.

In any metric space, just like in \(\mathbb{R}\), continuous limits may be replaced by sequential limits. The proof is again a direct translation of the proof from chapter 3, and we leave it as an exercise. The upshot is that we really only need to prove things for sequential limits.

Lemma 7.5.17. Let \((X,d_X)\) and \((Y,d_Y)\) be metric spaces, \(S \subset X\), \(p \in X\) a cluster point of \(S\), and let \(f: S \to Y\) be a function.

Then \(f(x)\) converges to \(L \in Y\) as \(x\) goes to \(p\) if and only if for every sequence \(\{x_n\}\) in \(S \setminus \{p\}\) such that \(\lim x_n = p\), the sequence \(\{f(x_n)\}\) converges to \(L\).
By applying Proposition 7.5.2 or the definition directly we find (exercise) as in chapter 3, that for cluster points \( p \) of \( S \subset X \), the function \( f : S \rightarrow Y \) is continuous at \( p \) if and only if

\[
\lim_{x \to p} f(x) = f(p).
\]

### 7.5.6 Exercises

**Exercise 7.5.1:** Consider \( \mathbb{N} \subset \mathbb{R} \) with the standard metric. Let \((X,d)\) be a metric space and \( f : X \rightarrow \mathbb{N} \) a continuous function.

a) Prove that if \( X \) is connected, then \( f \) is constant (the range of \( f \) is a single value).

b) Find an example where \( X \) is disconnected and \( f \) is not constant.

**Exercise 7.5.2:** Let \( f : \mathbb{R}^2 \rightarrow \mathbb{R} \) be defined by \( f(0,0) := 0 \) and \( f(x,y) := \frac{xy}{x^2+y^2} \) if \( (x,y) \neq (0,0) \).

a) Show that for every fixed \( x \), the function that takes \( y \) to \( f(x,y) \) is continuous. Similarly for every fixed \( y \), the function that takes \( x \) to \( f(x,y) \) is continuous.

b) Show that \( f \) is not continuous.

**Exercise 7.5.3:** Suppose \((X,d_X), (Y,d_Y)\) are metric spaces and \( f : X \rightarrow Y \) is continuous. Let \( A \subset X \).

a) Show that \( f(A) \subset f(A) \).

b) Show that the subset can be proper.

**Exercise 7.5.4:** Prove Theorem 7.5.8. Hint: Use Lemma 7.5.7.

**Exercise 7.5.5:** Suppose \( f : X \rightarrow Y \) is continuous for metric spaces \((X,d_X)\) and \((Y,d_Y)\). Show that if \( X \) is connected, then \( f(X) \) is connected.

**Exercise 7.5.6:** Prove the following version of the intermediate value theorem. Let \((X,d)\) be a connected metric space and \( f : X \rightarrow \mathbb{R} \) a continuous function. Suppose that there exist \( x_0,x_1 \in X \) and \( y \in \mathbb{R} \) such that \( f(x_0) < y < f(x_1) \). Then prove that there exists \( a \in X \) such that \( f(a) = y \). Hint: See Exercise 7.5.5.

**Exercise 7.5.7:** A continuous \( f : X \rightarrow Y \) between metric spaces \((X,d_X)\) and \((Y,d_Y)\) is said to be proper if for every compact set \( K \subset Y \), the set \( f^{-1}(K) \) is compact. Suppose a continuous \( f : (0,1) \rightarrow (0,1) \) is proper and \( \{x_n\} \) is a sequence in \((0,1)\) converging to 0. Show that \( \{f(x_n)\} \) has no subsequence that converges in \((0,1)\).

**Exercise 7.5.8:** Let \((X,d_X)\) and \((Y,d_Y)\) be metric spaces and \( f : X \rightarrow Y \) be a one-to-one and onto continuous function. Suppose \( X \) is compact. Prove that the inverse \( f^{-1} : Y \rightarrow X \) is continuous.

**Exercise 7.5.9:** Take the metric space of continuous functions \( C([0,1],\mathbb{R}) \). Let \( k : [0,1] \times [0,1] \rightarrow \mathbb{R} \) be a continuous function. Given \( f \in C([0,1],\mathbb{R}) \) define

\[
\varphi_f(x) := \int_0^1 k(x,y)f(y)dy.
\]

a) Show that \( T(f) := \varphi_f \) defines a function \( T : C([0,1],\mathbb{R}) \rightarrow C([0,1],\mathbb{R}) \).

b) Show that \( T \) is continuous.
Exercise 7.5.10: Let \((X, d)\) be a metric space.

a) If \(p \in X\), show that \(f : X \to \mathbb{R}\) defined by \(f(x) := d(x, p)\) is continuous.

b) Define a metric on \(X \times X\) as in Exercise 7.1.6 part b, and show that \(g : X \times X \to \mathbb{R}\) defined by \(g(x, y) := d(x, y)\) is continuous.

c) Show that if \(K_1\) and \(K_2\) are compact subsets of \(X\), then there exists a \(p \in K_1\) and \(q \in K_2\) such that \(d(p, q)\) is minimal, that is, \(d(p, q) = \inf\{d(x, y) : x \in K_1, y \in K_2\}\).

Exercise 7.5.11: Let \((X, d)\) be a compact metric space, let \(C(X, \mathbb{R})\) be the set of real-valued continuous functions. Define

\[
d(f, g) := \|f - g\|_u := \sup_{x \in X} |f(x) - g(x)|.
\]

a) Show that \(d\) makes \(C(X, \mathbb{R})\) into a metric space.

b) Show that for every \(x \in X\), the evaluation function \(E_x : C(X, \mathbb{R}) \to \mathbb{R}\) defined by \(E_x(f) := f(x)\) is a continuous function.

Exercise 7.5.12: Let \(C([a, b], \mathbb{R})\) be the set of continuous functions and \(C^1([a, b], \mathbb{R})\) the set of once continuously differentiable functions on \([a, b]\). Define

\[
d_C(f, g) := \|f - g\|_u \quad \text{and} \quad d_{C^1}(f, g) := \|f - g\|_u + \|f' - g'\|_u,
\]

where \(\|\cdot\|_u\) is the uniform norm. By Example 7.1.8 and Exercise 7.1.12 we know that \(C([a, b], \mathbb{R})\) with \(d_C\) is a metric space and so is \(C^1([a, b], \mathbb{R})\) with \(d_{C^1}\).

a) Prove that the derivative operator \(D : C^1([a, b], \mathbb{R}) \to C([a, b], \mathbb{R})\) defined by \(D(f) := f'\) is continuous.

b) On the other hand if we consider the metric \(d_C\) on \(C^1([a, b], \mathbb{R})\), then prove the derivative operator is no longer continuous. Hint: Consider \(\sin(nx)\).

Exercise 7.5.13: Let \((X, d)\) be a metric space, \(S \subset X\), and \(p \in X\). Prove that \(p\) is a cluster point of \(S\) if and only if \(p \in S \setminus \{p\}\).

Exercise 7.5.14: Prove Proposition 7.5.16.

Exercise 7.5.15: Prove Lemma 7.5.17.

Exercise 7.5.16: Let \((X, d_X)\) and \((Y, d_Y)\) be metric spaces, \(S \subset X\), \(p \in X\) a cluster point of \(S\), and let \(f : S \to Y\) be a function. Prove that \(f : S \to Y\) is continuous at \(p\) if and only if

\[
\lim_{x \to p} f(x) = f(p).
\]

Exercise 7.5.17: Define

\[
f(x, y) := \begin{cases} 
\frac{2xy}{x^2 + y^2} & \text{if } (x, y) \neq (0, 0), \\
0 & \text{if } (x, y) = (0, 0).
\end{cases}
\]

a) Show that for every fixed \(y\) the function that takes \(x\) to \(f(x, y)\) is continuous and hence Riemann integrable.

b) For every fixed \(x\), the function that takes \(y\) to \(f(x, y)\) is continuous.

c) Show that \(f\) is not continuous at \((0, 0)\).

d) Now show that \(g(y) := \int_0^1 f(x, y) \, dx\) is not continuous at \(y = 0\).

Note: Feel free to use what you know about \(\arctan\) from calculus, in particular that \(\frac{d}{dx} \arctan(x) = \frac{1}{1+x^2}\).
Exercise 7.5.18: Prove a stronger version of Proposition 7.5.12: If \( f : (a,b) \times (c,d) \to \mathbb{R} \) is a bounded continuous function, then \( g : (c,d) \to \mathbb{R} \) defined by

\[
g(y) := \int_{a}^{b} f(x,y) \, dx \quad \text{is continuous.}
\]

Hint: First integrate over \([a + 1/n, b - 1/n]\).
## 7.6 Fixed point theorem and Picard’s theorem again

**Note**: 1 lecture (optional, does not require §6.3)

In this section we prove the fixed point theorem for contraction mappings. As an application we prove Picard’s theorem, which we proved without metric spaces in §6.3. The proof we present here is similar, but the proof goes a lot smoother with metric spaces and the fixed point theorem.

### 7.6.1 Fixed point theorem

**Definition 7.6.1.** Let \((X, d_X)\) and \((Y, d_Y)\) be metric spaces. A mapping \(f : X \to Y\) is said to be a **contraction** (or a contractive map) if it is a \(k\)-Lipschitz map for some \(k < 1\), i.e. if there exists a \(k < 1\) such that

\[
d_Y(f(p), f(q)) \leq k d_X(p, q) \quad \text{for all } p, q \in X.
\]

Given a map \(f : X \to X\), a point \(x \in X\) is called a **fixed point** if \(f(x) = x\).

**Theorem 7.6.2** (Contraction mapping principle or Banach fixed point theorem\(^*\)). Let \((X, d)\) be a nonempty complete metric space and \(f : X \to X\) a contraction. Then \(f\) has a unique fixed point.

The words **complete** and **contraction** are necessary. See Exercise 7.6.6.

**Proof.** Pick \(x_0 \in X\). Define a sequence \(\{x_n\}\) by \(x_{n+1} := f(x_n)\). Then

\[
d(x_{n+1}, x_n) = d(f(x_n), f(x_{n-1})) \leq kd(x_n, x_{n-1}).
\]

Repeating \(n\) times, we get \(d(x_{n+1}, x_n) \leq k^n d(x_1, x_0)\). For \(m > n\,

\[
d(x_m, x_n) \leq \sum_{i=n}^{m-1} d(x_{i+1}, x_i)
\]

\[
\leq \sum_{i=n}^{m-1} k^i d(x_1, x_0)
\]

\[
= k^n d(x_1, x_0) \sum_{i=0}^{m-n-1} k^i
\]

\[
\leq k^n d(x_1, x_0) \sum_{i=0}^{\infty} k^i = k^n d(x_1, x_0) \frac{1}{1-k}.
\]

In particular, the sequence is Cauchy (why?). Since \(X\) is complete, we let \(x := \lim x_n\), and we claim that \(x\) is our unique fixed point.

Fixed point? The function \(f\) is a contraction, so it is Lipschitz continuous:

\[
f(x) = f(\lim x_n) = \lim f(x_n) = \lim x_{n+1} = x.
\]

Unique? Let \(x\) and \(y\) be fixed points.

\[
d(x, y) = d(f(x), f(y)) \leq k d(x, y).
\]

As \(k < 1\) this means that \(d(x, y) = 0\) and hence \(x = y\). The theorem is proved.

\(^*\)Named after the Polish mathematician Stefan Banach (1892–1945) who first stated the theorem in 1922.
The proof is constructive. Not only do we know a unique fixed point exists. We also know how to find it. Start with any point \( x_0 \in X \), then iterate \( f(x_0), f(f(x_0)), f(f(f(x_0))) \), etc. We can even find how far away from the fixed point we are, see the exercises. The idea of the proof is therefore useful in real-world applications.

### 7.6.2 Picard’s theorem

Let us start with the metric space to which we will apply the fixed point theorem. That is, the space \( C([a,b], \mathbb{R}) \) of Example 7.1.8, the space of continuous functions \( f: [a,b] \to \mathbb{R} \) with the metric

\[
d(f,g) := \|f - g\|_a = \sup_{x \in [a,b]} |f(x) - g(x)|.
\]

Convergence in this metric is convergence in uniform norm, or in other words, uniform convergence. Therefore, \( C([a,b], \mathbb{R}) \) is a complete metric space, see Proposition 7.4.5.

Consider now the ordinary differential equation

\[
\frac{dy}{dx} = F(x,y).
\]

Given some \( x_0, y_0 \), we are looking for a function \( y = f(x) \) such that \( f(x_0) = y_0 \) and such that

\[
f'(x) = F(x,f(x)).
\]

To avoid having to come up with many names, we often simply write \( y' = F(x,y) \) for the equation and \( y(x) \) for the solution.

The simplest example is the equation \( y' = y, y(0) = 1 \). The solution is the exponential \( y(x) = e^x \). A somewhat more complicated example is \( y' = -2xy, y(0) = 1 \), whose solution is the Gaussian \( y(x) = e^{-x^2} \).

A subtle issue is how long does the solution exist. Consider the equation \( y' = y^2, y(0) = 1 \). Then \( y(x) = \frac{1}{1-x} \) is a solution. While \( F \) is a reasonably “nice” function and in particular it exists for all \( x \) and \( y \), the solution “blows up” at \( x = 1 \). For more examples related to Picard’s theorem, see §6.3.

It may be strange that we are looking in \( C([a,b], \mathbb{R}) \) for a differentiable function, but the idea is to consider the corresponding integral equation

\[
f(x) = y_0 + \int_0^x F(t,f(t)) \, dt.
\]

To solve this integral equation we only need a continuous function, and in some sense our task should be easier—we have more candidate functions to try. This way of thinking is quite typical when solving differential equations.

**Theorem 7.6.3** (Picard’s theorem on existence and uniqueness). Let \( I, J \subset \mathbb{R} \) be closed and bounded intervals, let \( I^\circ \) and \( J^\circ \) be their interiors, and let \( (x_0,y_0) \in I^\circ \times J^\circ \). Suppose \( F: I \times J \to \mathbb{R} \) is continuous and Lipschitz in the second variable, that is, there exists an \( L \in \mathbb{R} \) such that

\[
|F(x,y) - F(x,z)| \leq L|y - z| \quad \text{for all } y,z \in J, x \in I.
\]

Then there exists an \( h > 0 \) and a unique differentiable function \( f: [x_0-h,x_0+h] \to J \subset \mathbb{R} \) such that

\[
f'(x) = F(x,f(x)) \quad \text{and} \quad f(x_0) = y_0.
\]
Proof. Without loss of generality assume \( x_0 = 0 \) (exercise). As \( I \times J \) is compact and \( F(x,y) \) is continuous, it is bounded. So find an \( M > 0 \), such that \( |F(x,y)| \leq M \) for all \((x,y) \in I \times J \). Pick \( \alpha > 0 \) such that \([ -\alpha, \alpha ] \subset I \) and \([ y_0 - \alpha, y_0 + \alpha ] \subset J \). Let

\[
h := \min \left\{ \alpha, \frac{\alpha}{M + L\alpha} \right\}.
\]

Note \([-h,h] \subset I \). Let

\[
Y := \left\{ f \in C([-h,h], \mathbb{R}) : f([-h,h]) \subset J \right\}.
\]

That is, \( Y \) is the space of continuous functions on \([-h, h] \) with values in \( J \), in other words, exactly those functions where \( F(x, f(x)) \) makes sense. The metric used is the standard metric given above.

It is left as an exercise to show that \( Y \) is complete (because \( J \) is closed). The space \( C([-h,h], \mathbb{R}) \) is complete, and a closed subset of a complete metric space is a complete metric space with the subspace metric, see Proposition 7.4.6. So \( Y \) with the subspace metric is a complete metric space.

Define a mapping \( T : Y \to C([-h,h], \mathbb{R}) \) by

\[
T(f)(x) := y_0 + \int_0^x F(t, f(t)) \, dt.
\]

It is an exercise to check that \( T \) is well-defined, and that \( T(f) \) really is in \( C([-h, h], \mathbb{R}) \).

Let \( f \in Y \) and \( |x| \leq h \). As \( F \) is bounded by \( M \), we have

\[
|T(f)(x) - y_0| = \left| \int_0^x F(t, f(t)) \, dt \right| 
\leq |x| M \leq h M \leq \frac{\alpha M}{M + L\alpha} \leq \alpha.
\]

So \( T(f)([-h,h]) \subset [y_0 - \alpha, y_0 + \alpha] \subset J \), and \( T(f) \in Y \). In other words, \( T(Y) \subset Y \). From now on, we consider \( T \) as a mapping of \( Y \) to \( Y \).

We claim \( T : Y \to Y \) is a contraction. First, for \( x \in [-h,h] \) and \( f, g \in Y \), we have

\[
|F(x, f(x)) - F(x, g(x))| \leq L|f(x) - g(x)| \leq Ld(f, g).
\]

Therefore,

\[
|T(f)(x) - T(g)(x)| = \left| \int_0^x F(t, f(t)) - F(t, g(t)) \, dt \right|
\leq |x| Ld(f, g) \leq h Ld(f, g) \leq \frac{L\alpha}{M + L\alpha} d(f, g).
\]

We chose \( M > 0 \) and so \( \frac{L\alpha}{M + L\alpha} < 1 \). The claim that \( T \) is a contraction is proved by taking supremum over \( x \in [-h,h] \) of the left-hand side above to obtain \( d(T(f), T(g)) \leq \frac{L\alpha}{M + L\alpha} d(f, g) \).

We apply the fixed point theorem (Theorem 7.6.2) to find a unique \( f \in Y \) such that \( T(f) = f \), that is,

\[
f(x) = y_0 + \int_0^x F(t, f(t)) \, dt.
\]

By the fundamental theorem of calculus (Theorem 5.3.3), \( T(f) = f \) is differentiable, its derivative is \( F(x, f(x)) \) and \( T(f)(0) = y_0 \). Differentiable functions are continuous, so \( f \) is the unique differentiable function \( f : [-h,h] \to J \) such that \( f'(x) = F(x, f(x)) \) and \( f(0) = y_0 \). \( \square \)
7.6.3 Exercises

For more exercises related to Picard’s theorem see §6.3.

**Exercise 7.6.1:** Suppose $J$ is a closed and bounded interval, and let $Y := \{ f \in C([-h, h], \mathbb{R}) : f([-h, h]) \subset J \}$. Show that $Y \subset C([-h, h], \mathbb{R})$ is closed. Hint: $J$ is closed.

**Exercise 7.6.2:** In the proof of Picard’s theorem, show that if $f : [-h, h] \to J$ is continuous, then $F(t, f(t))$ is continuous on $[-h, h]$ as a function of $t$. Use this to show that

$$T(f)(x) := y_0 + \int_0^x F(t, f(t)) \, dt$$

is well-defined and that $T(f) \in C([-h, h], \mathbb{R})$.

**Exercise 7.6.3:** Prove that in the proof of Picard’s theorem, the statement “Without loss of generality assume $x_0 = 0$” is justified. That is, prove that if we know the theorem with $x_0 = 0$, the theorem is true as stated.

**Exercise 7.6.4:** Let $F : \mathbb{R} \to \mathbb{R}$ be defined by $F(x) := kx + b$ where $0 < k < 1$, $b \in \mathbb{R}$.

a) Show that $F$ is a contraction.

b) Find the fixed point and show directly that it is unique.

**Exercise 7.6.5:** Let $f : [0, 1/4] \to [0, 1/4]$ be defined by $f(x) := x^2$.

a) Show that $f$ is a contraction, and find the best (smallest) $k$ from the definition that works.

b) Find the fixed point and show directly that it is unique.

**Exercise 7.6.6:**

a) Find an example of a contraction $f : X \to X$ of a non-complete metric space $X$ with no fixed point.

b) Find a 1-Lipschitz map $f : X \to X$ of a complete metric space $X$ with no fixed point.

**Exercise 7.6.7:** Consider $y' = y^2$, $y(0) = 1$. Use the iteration scheme from the proof of the contraction mapping principle. Start with $f_0(x) = 1$. Find a few iterates (at least up to $f_2$). Prove that the pointwise limit of $f_n$ is $1/(1-x)$, that is, for every $x$ with $|x| < h$ for some $h > 0$, prove that $\lim_{n \to \infty} f_n(x) = 1/(1-x)$.

**Exercise 7.6.8:** Suppose $f : X \to X$ is a contraction for $k < 1$. Suppose you use the iteration procedure with $x_{n+1} := f(x_n)$ as in the proof of the fixed point theorem. Suppose $x$ is the fixed point of $f$.

a) Show that $d(x, x_n) \leq k^n d(x_1, x_0) \frac{1}{1-k}$ for all $n \in \mathbb{N}$.

b) Suppose $d(y_1, y_2) \leq 16$ for all $y_1, y_2 \in X$, and $k = 1/2$. Find an $N$ such that starting at any given point $x_0 \in X$, $d(x, x_n) \leq 2^{-16}$ for all $n \geq N$.

**Exercise 7.6.9:** Let $f(x) := x - \frac{x^2}{2}$ (you may recognize Newton’s method for $\sqrt{2}$).

a) Prove $f([1, \infty)) \subset [1, \infty)$.

b) Prove that $f : [1, \infty) \to [1, \infty)$ is a contraction.

c) Apply the fixed point theorem to find an $x \geq 1$ such that $f(x) = x$, and show that $x = \sqrt{2}$.

**Exercise 7.6.10:** Suppose $f : X \to X$ is a contraction, and $(X, d)$ is a metric space with the discrete metric, that is, $d(x, y) = 1$ whenever $x \neq y$. Show that $f$ is constant, that is, there exists a $c \in X$ such that $f(x) = c$ for all $x \in X$.

**Exercise 7.6.11:** Suppose $(X, d)$ is a nonempty complete metric space, $f : X \to X$ is a mapping, and denote by $f^n$ the nth iterate of $f$. Suppose for every $n$ there exists a $k_n > 0$ such that $d\left(f^n(x), f^n(y)\right) \leq k_n d(x, y)$ for all $x, y \in X$, where $\sum_{j=1}^{\infty} k_n < \infty$. Prove that $f$ has a unique fixed point in $X$. 
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