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BABEȘ-BOLYAI TUDOMÁNYEGYETEM  
BABEȘ-BOLYAI UNIVERSITÁT  
BABEȘ-BOLYAI UNIVERSITY  
TRADITIO ET EXCELLENTIA

# Mathematical Analysis

1st Year Computer Science

Mihai Nechita \*

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\*[math.ubbcluj.ro/~mihai.nechita](mailto:math.ubbcluj.ro/~mihai.nechita). If you find any typos, please let me know on Teams or by email.

## ※ Real numbers

Let us start with some standard notation:  $\emptyset$  is the empty set;  $\mathbb{N} = \{1, 2, \dots\}$  the set of natural numbers;  $\mathbb{Z} = \{\dots, -1, 0, 1, \dots\} = \{m - n \mid m, n \in \mathbb{N}\}$  the set of integers;  $\mathbb{Q} = \{\frac{m}{n} \mid m, n \in \mathbb{Z}, n \neq 0\}$  the set of rational numbers;  $\mathbb{R}$  the set of real numbers.

You are very much used with the real numbers. However, it is not trivial to define them in a rigorous way, see for example [1] or [2] for different methods of constructing  $\mathbb{R}$  from  $\mathbb{Q}$ . We will straightforwardly start working with the real numbers – for this reason some of their properties, e.g. [Theorem 1.5](#), will simply be given as definitions.

**Definition 1.1.** Let  $A$  be a subset of  $\mathbb{R}$ , denoted as  $A \subseteq \mathbb{R}$ . We define  $x \in \mathbb{R}$  to be

a *lower bound* for  $A$  if  $x \leq a, \forall a \in A$ ; an *upper bound* for  $A$  if  $x \geq a, \forall a \in A$ .

We define

$lb(A) := \{x \in \mathbb{R} \mid x \leq a, \forall a \in A\}$  the set of lower bounds of  $A$ ,

$ub(A) := \{x \in \mathbb{R} \mid x \geq a, \forall a \in A\}$  the set of upper bounds of  $A$ .

We define  $x \in \mathbb{R}$  to be

the *minimum* of  $A$  if  $x \in lb(A) \cap A$ ; the *maximum* of  $A$  if  $x \in ub(A) \cap A$ ,

denoted by  $\min(A)$ , respectively  $\max(A)$ . In other words, we have that

$\min(A) \in A$  and  $\min(A) \leq a, \forall a \in A$ ;  $\max(A) \in A$  and  $\max(A) \geq a, \forall a \in A$ .

Note that there are sets which do not have minimum or maximum, e.g.  $(0, 1)$ .

**Definition 1.2.** A set  $A \subseteq \mathbb{R}$  is defined to be

- bounded (from) below if  $lb(A) \neq \emptyset$ ;
- bounded (from) above if  $ub(A) \neq \emptyset$ ;
- bounded if it is both bounded below and above;
- unbounded if it is not bounded.

**Definition 1.3.** We say that  $x \in \mathbb{R}$  is the *supremum* of  $A \subseteq \mathbb{R}$ ,  $x := \sup(A)$ , if and only if:

1.  $x \geq a, \forall a \in A$ , that is  $x \in ub(A)$ .
2. if  $u$  is an upper bound for  $A$ , then  $x \leq u$ .

The supremum is *the least upper bound*, i.e.  $\sup(A) := \min(ub(A))$ .

**Definition 1.4.** We say that  $x \in \mathbb{R}$  is the *infimum* of  $A \subseteq \mathbb{R}$ ,  $x := \inf(A)$ , if and only if:

1.  $x \leq a$ ,  $\forall a \in A$ , that is  $x \in lb(A)$ .
2. if  $u$  is a lower bound for  $A$ , then  $x \geq u$ .

The infimum is *the greatest lower bound*, i.e.  $\inf(A) := \max(lb(A))$ .

**Definition 1.5** (Completeness Axiom). Every set  $A \subseteq \mathbb{R}$  that is bounded above has a supremum. Similarly, every set  $A \subseteq \mathbb{R}$  that is bounded below has an infimum.

Note that if  $A$  has a maximum, then  $\sup(A) = \max(A)$ . Similarly, if  $A$  has a minimum, then  $\inf(A) = \min(A)$ . Also, if  $\sup(A) \in A$ , then  $\max(A) = \sup(A)$ .

**Example 1.6.** (a)  $A = \{\frac{1}{n} \mid n \in \mathbb{N}\}$ ,  $\sup(A) = 1 = \max(A)$ ,  $\inf(A) = 0$ ,  $\nexists \min(A)$ .

(b)  $A = \{x \in \mathbb{Q} \mid x^2 \leq 2\}$ ,  $\sup(A) = \sqrt{2}$ ,  $\nexists \max(A)$ ,  $\inf(A) = -\sqrt{2}$ ,  $\nexists \min(A)$ .

**Theorem 1.7.** Let  $A \subseteq \mathbb{R}$  be a bounded set. For  $\sup(A)$  and  $\inf(A)$  the following are true:

$$\forall \varepsilon > 0, \exists x \in A \text{ such that } \sup(A) - \varepsilon < x,$$

$$\forall \varepsilon > 0, \exists x \in A \text{ such that } x < \inf(A) + \varepsilon.$$

*Proof.* By definition, for any  $y < \sup(A)$ , say  $y = \sup(A) - \varepsilon$  with  $\varepsilon > 0$ , we have that  $y \notin ub(A)$ . Hence there exists  $x \in A$  such that  $y < x$ . Similar proof for  $\inf(A)$ .  $\square$

**Proposition 1.8.** Let  $A \subseteq B \subseteq \mathbb{R}$  be (nonempty) bounded sets. Then

$$\inf(B) \leq \inf(A) \leq \sup(A) \leq \sup(B)$$

and

$$\begin{aligned} \sup(A \cup B) &= \max\{\sup(A), \sup(B)\}, \\ \inf(A \cup B) &= \min\{\inf(A), \inf(B)\}. \end{aligned}$$

*Proof.* It follows directly from the definitions.  $\square$

**Definition 1.9.** Define the *extended real line*  $\overline{\mathbb{R}} := \mathbb{R} \cup \{-\infty, \infty\}$ , where  $\infty$  and  $-\infty$  are such that

$$\forall x \in \mathbb{R}, -\infty < x < \infty.$$

If a set  $A$  is not bounded above, we define  $\sup(A) := \infty$ .

If a set  $A$  is not bounded below, we define  $\inf(A) := -\infty$ .

[Seminar] The empty set  $\emptyset$  is bounded by any number. In  $\overline{\mathbb{R}}$ ,  $\sup(\emptyset) = -\infty$ ,  $\inf(\emptyset) = \infty$ .

**Definition 1.10.** A set  $V \subseteq \mathbb{R}$  is a *neighborhood (vecinity)* of  $x \in \mathbb{R}$  if

$$\exists \varepsilon > 0 \text{ such that } (x - \varepsilon, x + \varepsilon) \subseteq V.$$

A set  $V \subseteq \mathbb{R}$  is a *neighborhood* of  $\infty$  if  $\exists a \in \mathbb{R}$  such that  $(a, \infty) \subseteq V$ .

A set  $V \subseteq \mathbb{R}$  is a *neighborhood* of  $-\infty$  if  $\exists a \in \mathbb{R}$  such that  $(-\infty, a) \subseteq V$ .

We denote all the neighborhoods of  $x$  by  $\mathcal{V}(x) := \{V \subseteq \mathbb{R} \mid V \text{ is a neighborhood of } x\}$ .

**Definition 1.11.** Let  $A \subseteq \mathbb{R}$ . The following set is called the *interior* of  $A$

$$\text{int}(A) := \{x \in \mathbb{R} \mid \exists V \in \mathcal{V}(x) \text{ such that } V \subseteq A\},$$

and the following set is called the *closure* of  $A$

$$\text{cl}(A) := \{x \in \mathbb{R} \mid \forall V \in \mathcal{V}(x), V \cap A \neq \emptyset\}.$$

**Proposition 1.12.** For any  $A \subseteq \mathbb{R}$ , it holds that  $\text{int}(A) \subseteq A \subseteq \text{cl}(A)$ .

*Proof.* To prove that  $\text{int}(A) \subseteq A$  we prove that if  $x \in \text{int}(A)$ , then  $x \in A$ . Let  $x \in \text{int}(A)$ , then  $\exists \varepsilon > 0$  such that  $(x - \varepsilon, x + \varepsilon) \subseteq A$ . Since  $x \in (x - \varepsilon, x + \varepsilon)$ , we have that  $x \in A$ .

To prove that  $A \subseteq \text{cl}(A)$  we show that if  $x \in A$ , then  $x \in \text{cl}(A)$ . Let  $x \in A$ . Then for any  $V \in \mathcal{V}(x)$  it holds that  $x \in V$ , giving that  $x \in V \cap A$ . Hence  $x \in \text{cl}(A)$  since  $V \cap A \neq \emptyset$ .  $\square$

**Definition 1.13.** If  $A = \text{int}(A)$ , then  $A$  is called *open*. If  $A = \text{cl}(A)$ , then  $A$  is called *closed*.

**Remark 1.14.** To prove that a set  $A$  is open, it is sufficient to prove that  $A \subseteq \text{int}(A)$ .

To prove that a set  $A$  is closed, it is sufficient to prove that  $\text{cl}(A) \subseteq A$ .

**Proposition 1.15.** The following statements are true:

- The complement of an open set is closed.
- The complement of a closed set is open.

*Proof.* Let us prove the first statement, the other one being similar. Consider  $A$  an open set, i.e.  $A = \text{int}(A)$ , and denote by  $A^c = \{x \in \mathbb{R} \mid x \notin A\}$  its complement. To prove that  $A^c$  is closed, we prove that  $\text{cl}(A^c) \subseteq A^c$ . Consider  $x \in \text{cl}(A^c)$  and let's assume that  $x \notin A^c$ , i.e.  $x \in A$ , aiming to obtain a contradiction. Since  $A$  is open, there exists  $V \in \mathcal{V}(x)$  such that  $V \subseteq A$ , giving that  $V \cap A^c = \emptyset$ : contradiction with  $x \in \text{cl}(A^c)$ . Hence the assumption  $x \notin A^c$  is false, and we have that if  $x \in \text{cl}(A^c)$ , then  $x \in A^c$ . In other words,  $\text{cl}(A^c) \subseteq A^c$ .  $\square$

**Proposition 1.16.** Any union of open sets is open. Any intersection of closed sets is closed. Any finite intersection of open sets is open. Any finite union of closed sets is closed.

*Proof.* (Optional) Left as extra homework. □

### Exercises. Seminar 1

1. Find the lower and the upper bounds, then sup, inf, max, min for each of the following:

- (a)  $[-3, 2) \cup \{3\}$ . (c)  $(-5, 5) \cap \mathbb{Z}$ .  
 (b)  $(-1, 1] \cup (2, \infty)$ . (d)  $\emptyset$ .

2. Find the sup, inf, max, min for each of the following sets:

- (a)  $\{x \in \mathbb{Q} \mid x^2 < 3\}$ . (c)  $\{\frac{n}{n+1} \mid n \in \mathbb{N}\}$ .  
 (b)  $\{x^2 - 4x + 3 \mid x \in \mathbb{R}\}$ . (d)  $\{2^{-k} + 3^{-m} \mid k, m \in \mathbb{N}\}$ .

3. Suppose that  $S$  is nonempty and bounded above. Show that the set  $-S := \{-x \mid x \in S\}$  is bounded below and  $\inf(-S) = -\sup(S)$ .

4. Let  $f : D \rightarrow \mathbb{R}$  and  $g : D \rightarrow \mathbb{R}$  be two functions defined on a nonempty set  $D$ . Prove that

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

Give examples where the above inequalities are strict.

5. Which of the following sets are neighborhoods of 0?

$$[-1, 1] \cup \{2\}; \quad (-1, 1) \cap \mathbb{Q}; \quad \bigcap_{n=1}^{\infty} \left[-\frac{1}{n}, \frac{1}{n}\right].$$

6. Let  $x \in \mathbb{R}$  and  $U, V \in \mathcal{V}(x)$ . Prove that  $U \cap V \in \mathcal{V}(x)$ .

7. Find the interior and the closure for each of the following sets:

- (a)  $(1, 2]$ . (c)  $(-1, 1] \cup (2, \infty)$ .  
 (b)  $[-3, 2) \cup \{3\}$ . (d)  $(-5, 5) \cap \mathbb{Z}$ .

### Exercises. Homework 1

1. Let  $a, b \in \mathbb{R}$  with  $a > 0$ . If  $S \subset \mathbb{R}$  is nonempty and bounded above, prove that

$$\sup_{x \in S} (ax + b) = a \sup(S) + b.$$

2. Let  $a, b \in \mathbb{R}$ . Prove that there exist neighborhoods  $U \in \mathcal{V}(a)$  and  $V \in \mathcal{V}(b)$  such that

$$U \cap V = \emptyset.$$

3. Let  $A = (0, 1) \cap \mathbb{Q}$ . Show rigorously (using the definitions) that

$$\inf A = 0, \sup A = 1, \text{int}A = \emptyset \text{ and } \text{cl}A = [0, 1].$$

## ※ Sequences

A set  $\{x_n \mid n \in \mathbb{N}\}$  is called a sequence and is denoted by  $(x_n)_{n \in \mathbb{N}}$  or simply  $(x_n)$ . A sequence  $(x_n)$  is bounded above (or below) if the set  $\{x_n \mid n \in \mathbb{N}\}$  is bounded above (or below). A sequence  $(x_n)$  is increasing if  $x_{n+1} \geq x_n$ ,  $\forall n \in \mathbb{N}$ , and decreasing if  $x_{n+1} \leq x_n$ ,  $\forall n \in \mathbb{N}$ . A sequence is monotone if it is either increasing or decreasing.

**Definition 2.1.** A sequence  $(x_n)$  has a limit  $\ell \in \overline{\mathbb{R}}$ , and we write  $\lim_{n \rightarrow \infty} x_n = \ell$  or  $x_n \rightarrow \ell$ , if

$$\forall V \in \mathcal{V}(\ell), \exists N_V \in \mathbb{N} \text{ such that } x_n \in V, \forall n \geq N_V.$$

If  $\ell \in \mathbb{R}$ , we say that  $(x_n)$  converges to  $\ell$ :  $\forall \varepsilon > 0$ ,  $\exists N_\varepsilon \in \mathbb{N}$  such that  $|x_n - \ell| < \varepsilon$ ,  $\forall n \geq N_\varepsilon$ .

$x_n \rightarrow \infty$  if  $\forall a > 0$ ,  $\exists N_a \in \mathbb{N}$  such that  $x_n > a$ ,  $\forall n \geq N_a$ .

$x_n \rightarrow -\infty$  if  $\forall a < 0$ ,  $\exists N_a \in \mathbb{N}$  such that  $x_n < a$ ,  $\forall n \geq N_a$ .

**Example 2.2.** Prove that  $\lim_{n \rightarrow \infty} \frac{1}{n} = 0$ . Let  $\varepsilon > 0$ ; we need to find an index  $N_\varepsilon \in \mathbb{N}$  such that  $\frac{1}{n} < \varepsilon$ ,  $\forall n \geq N_\varepsilon$ . Since  $\frac{1}{n} \leq \frac{1}{N_\varepsilon}$  for  $n \geq N_\varepsilon$ , it is enough to have  $\frac{1}{N_\varepsilon} < \varepsilon$  or  $N_\varepsilon > \frac{1}{\varepsilon}$ . We conclude by taking  $N_\varepsilon > \frac{1}{\varepsilon}$ , e.g.  $N_\varepsilon = [1/\varepsilon] + 1$ .

**Proposition 2.3.** A sequence  $(x_n)$  converges to  $\ell \in \mathbb{R}$  if and only if  $\lim_{n \rightarrow \infty} |x_n - \ell| = 0$ .

**Proposition 2.4.** Any convergent sequence is bounded.

*Proof.* Let  $\ell \in \mathbb{R}$  be the limit. For  $\varepsilon = 1$ , there exists  $N_1 \in \mathbb{N}$  such that  $|x_n - \ell| < 1$ ,  $\forall n \geq N_1$ . All the terms after index  $N_1$  are in  $(\ell - 1, \ell + 1)$ . Since there are finitely many terms before index  $N_1$ , we conclude that the sequence is bounded.  $\square$

**Theorem 2.5** (Weierstrass). Any monotone and bounded sequence is convergent.

*Proof.* Assume that the sequence is increasing, for example. Let  $S = \{x_n \mid n \in \mathbb{N}\}$  and consider  $\sup(S) \in \mathbb{R}$  (we know that  $S$  is bounded). From [Theorem 1.7](#) we have that

$$\forall \varepsilon > 0, \exists N_\varepsilon \in \mathbb{N} \text{ such that } \sup(S) - \varepsilon < x_{N_\varepsilon}.$$

As  $(x_n)$  is increasing,  $\sup(S) - \varepsilon < x_{N_\varepsilon} \leq x_n$   $\forall n \geq N_\varepsilon$ . Hence  $\sup(S) - x_n < \varepsilon$ ,  $\forall n \geq N_\varepsilon$ . The sequence converges to  $\sup(S)$  by [Theorem 2.1](#). Similarly, a decreasing and bounded sequence converges to its infimum.  $\square$

**Proposition 2.6.** Any monotone sequence has a limit in  $\overline{\mathbb{R}}$ .

*Proof.* If the sequence is bounded and monotone, then it is convergent by the Weierstrass theorem. If the sequence is unbounded and monotone, then its limit will be infinite.  $\square$

**Theorem 2.7** (Squeeze/Sandwich theorem). Let  $(x_n), (y_n), (z_n)$  be sequences for which there is an  $n_0 \in \mathbb{N}$  such that

$$x_n \leq y_n \leq z_n, \forall n \geq n_0,$$

and

$$\lim_{n \rightarrow \infty} x_n = \lim_{n \rightarrow \infty} z_n.$$

Then

$$\lim_{n \rightarrow \infty} x_n = \lim_{n \rightarrow \infty} y_n = \lim_{n \rightarrow \infty} z_n.$$

*Proof.* Let  $\ell := \lim_{n \rightarrow \infty} x_n = \lim_{n \rightarrow \infty} z_n$  and assume first that  $\ell \in \mathbb{R}$ . Let  $\varepsilon > 0$ . Then

$$\exists N_1 \in \mathbb{N} \text{ such that } |x_n - \ell| < \varepsilon, \forall n \geq N_1$$

and

$$\exists N_2 \in \mathbb{N} \text{ such that } |z_n - \ell| < \varepsilon, \forall n \geq N_2.$$

Taking  $N_\varepsilon := \max\{N_1, N_2\}$ , we have that

$$|y_n - \ell| \leq \max\{|x_n - \ell|, |z_n - \ell|\} < \varepsilon, \forall n \geq N_\varepsilon,$$

hence the conclusion. When  $\ell$  is infinite the proof is similar.  $\square$

**Theorem 2.8** (Cantor's nested intervals). Let  $(a_n)$  be increasing and  $(b_n)$  decreasing such that  $a_n \leq a_{n+1} \leq b_{n+1} \leq b_n, \forall n \in \mathbb{N}$ . Consider the closed intervals  $I_n := [a_n, b_n]$ , with  $I_{n+1} \subseteq I_n$ . If  $\lim_{n \rightarrow \infty} (b_n - a_n) = 0$ , then there exists  $x \in \mathbb{R}$  such that

$$\bigcap_{n=1}^{\infty} I_n = \{x\}.$$

*Proof.* Consider the bounded sets  $A := \{a_n \mid n \in \mathbb{N}\}$  and  $B := \{b_n \mid n \in \mathbb{N}\}$ . For any  $k \in \mathbb{N}$ , we have that

$$a_k \leq \sup(A) \leq b_k$$

and

$$b_k \geq \inf(B) \geq a_k.$$

Hence by the squeeze [Theorem 2.7](#) we have that  $\sup(A) = \inf(B)$  and  $\bigcap_{n=1}^{\infty} I_n = \{\sup(A)\}$ .  $\square$

**Theorem 2.9** (Bolzano-Weierstrass). Any bounded sequence has a convergent subsequence.



*Proof.* Consider the bounded set  $A := \{x_n \mid n \in \mathbb{N}\}$ . Let  $a_1 := \inf(A)$  and  $b_1 := \sup(A)$ , and define  $I_1 := [a_1, b_1]$ . Bisect  $I_1$  and notice that at least one of the two halves must contain infinitely many terms from the sequence. Take  $I_2 := [a_2, b_2]$  to be the half that does. Continuing this procedure we obtain for each  $k \in \mathbb{N}$  an interval  $I_k := [a_k, b_k]$  containing (at least) a term  $x_{n_k} \in A$ , such that  $I_{k+1} \subseteq I_k$  and  $b_k - a_k \rightarrow 0$ .

From Cantor's nested intervals [Theorem 2.8](#) we have that there exists  $x \in \mathbb{R}$  such that  $\bigcap_{n=1}^{\infty} I_n = \{x\}$ , and hence the subsequence  $(x_{n_k})$  converges to  $x$ .  $\square$

**Definition 2.10.** For a sequence  $(x_n)$  we define the set of its *limit points* by

$$\text{LIM}(x_n) := \{x \in \overline{\mathbb{R}} \mid \text{there exists a subsequence } (x_{n_k}) \text{ s.t. } x_{n_k} \rightarrow x\},$$

and

$$\begin{aligned} \liminf_{n \rightarrow \infty} x_n &:= \inf(\text{LIM}(x_n)), \\ \limsup_{n \rightarrow \infty} x_n &:= \sup(\text{LIM}(x_n)). \end{aligned}$$

**Example 2.11.** For  $x_n = \frac{(-1)^n n}{n+1}$ ,  $\text{LIM}(x_n) = \{-1, 1\}$ ,  $\liminf_{n \rightarrow \infty} x_n = -1$ ,  $\limsup_{n \rightarrow \infty} x_n = 1$ .

**Proposition 2.12.**  $\lim_{n \rightarrow \infty} x_n = \ell \in \overline{\mathbb{R}}$  if and only if  $\liminf_{n \rightarrow \infty} x_n = \limsup_{n \rightarrow \infty} x_n = \ell$ .

**Definition 2.13** (Cauchy sequence). A sequence  $(x_n)$  is called *Cauchy* (or *fundamental*) if

$$\forall \varepsilon > 0, \exists N_\varepsilon \in \mathbb{N} \text{ such that } |x_m - x_n| < \varepsilon, \forall m, n \geq N_\varepsilon.$$

**Proposition 2.14.** Any Cauchy sequence is bounded.

*Proof.* For  $\varepsilon = 1$ , there exists  $N_1 \in \mathbb{N}$  such that  $|x_m - x_n| < 1, \forall m, n \geq N_1$ . In particular,  $|x_n - x_{N_1}| < 1, \forall n \geq N_1$ , hence all the terms after index  $N_1$  are in  $(x_{N_1} - 1, x_{N_1} + 1)$ . There are finitely many terms before index  $N_1$ , thus we conclude that the sequence is bounded.  $\square$

**Theorem 2.15.** A sequence is convergent if and only if it is Cauchy.

*Proof.* Let's consider first a convergent sequence  $(x_n)$  with  $x_n \rightarrow \ell$ . For any  $\varepsilon > 0$ , there exists  $N_\varepsilon \in \mathbb{N}$  such that  $|x_n - \ell| < \frac{\varepsilon}{2}$ , for any  $n \geq N_\varepsilon$ . Then  $|x_m - x_n| \leq |x_m - \ell| + |x_n - \ell| < \varepsilon$ , for any  $n \geq N_\varepsilon$ . Hence the sequence  $(x_n)$  is Cauchy.

Assume now that  $(x_n)$  is a Cauchy sequence. From the previous proposition we have that  $(x_n)$  must be bounded, and thus it has a convergent subsequence  $(x_{n_k}), x_{n_k} \rightarrow x \in \mathbb{R}$ . Let  $\varepsilon > 0$ . There exists thus  $K_\varepsilon \in \mathbb{N}$  such that  $|x_{n_k} - x| < \varepsilon, \forall k \geq K_\varepsilon$ . Also, there exists  $N_\varepsilon \in \mathbb{N}$  such that  $|x_m - x_n| < \varepsilon, \forall m, n \geq N_\varepsilon$ . In particular,  $|x_{n_k} - x_n| < \varepsilon, \forall k, n \geq N_\varepsilon$ . Hence  $|x_n - x| \leq |x_n - x_{n_k}| + |x_{n_k} - x| < 2\varepsilon, \forall n \geq \max\{K_\varepsilon, N_\varepsilon\}$ , meaning that  $x_n \rightarrow x$ .  $\square$

**Example 2.16.** The sequence defined by  $x_n = 1 + \frac{1}{2} + \dots + \frac{1}{n}$  is not convergent. Indeed, one can see, for example, that

$$x_{2n} - x_n = \frac{1}{n+1} + \dots + \frac{1}{2n} > \frac{n}{2n},$$

hence  $x_{2n} - x_n > \frac{1}{2}$  for any  $n \in \mathbb{N}$ . Thus  $(x_n)$  is not Cauchy, hence it is not convergent.

**Proposition 2.17** (Ratio test). Let  $(x_n)$  be a sequence with positive terms such that

$$\lim_{n \rightarrow \infty} \frac{x_{n+1}}{x_n} = \ell \in \overline{\mathbb{R}}.$$

If  $\ell < 1$ , then  $\lim_{n \rightarrow \infty} x_n = 0$ . If  $\ell > 1$ , then  $\lim_{n \rightarrow \infty} x_n = \infty$ . If  $\ell = 1$ , the test is inconclusive.

*Proof.* Let us consider the case  $\lim_{n \rightarrow \infty} \frac{x_{n+1}}{x_n} = \ell < 1$ . Let  $\varepsilon > 0$  such that  $q := \ell + \varepsilon < 1$ . Then there exists  $N \in \mathbb{N}$  such that  $\frac{x_{n+1}}{x_n} - \ell < \varepsilon$ ,  $\forall n \geq N$ . We have that  $x_{n+1} < x_n \cdot q$ ,  $\forall n \geq N$ , hence  $x_n < q^{n-N} x_N$ , giving that  $0 < x_n < q^n \frac{x_N}{q^N}$ . Since  $q < 1$ ,  $q^n \rightarrow 0$ , and  $x_n \rightarrow 0$  by the squeeze theorem. The proof is similar when  $\ell > 1$ .  $\square$

**Lemma 2.18** (Stolz-Cesàro). Let  $(a_n), (b_n)$  be two sequences such that (i)  $a_n \rightarrow 0$  and  $b_n \rightarrow 0$  with  $(b_n)$  decreasing; or (ii)  $b_n \rightarrow \infty$  with  $(b_n)$  increasing.

$$\text{If } \lim_{n \rightarrow \infty} \frac{a_{n+1} - a_n}{b_{n+1} - b_n} = \ell, \text{ then } \lim_{n \rightarrow \infty} \frac{a_n}{b_n} = \ell.$$

*Proof.* (Optional) See [3, Theorem 2.5.6].  $\square$

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### Exercises. Seminar 2

1. Prove using the  $\varepsilon$ -definition that  $\lim_{n \rightarrow \infty} \frac{1}{\sqrt{n}} = 0$ .
2. Study if the sequence  $(x_n)$  is bounded, monotone, and convergent, for each of the following:
  - (a)  $x_n = \sqrt{n+1} - \sqrt{n}$ .
  - (b)  $x_n = \frac{1}{1 \cdot 2} + \dots + \frac{1}{n(n+1)}$ .
  - (c)  $x_n = \frac{2^n}{n!}$ .
3. Find the limit for each of the following sequences:
  - (a)  $\sqrt{n}(\sqrt{n+1} - \sqrt{n})$ .
  - (b)  $\frac{2^n + (-1)^n}{3^n}$ .
  - (c)  $\sqrt[n]{n}$ .
  - (d)  $\frac{(an+1)^2}{4n^2 - 2n + 1}, a \in \mathbb{R}$ .
  - (e)  $(a_1^n + a_2^n + \dots + a_k^n)^{\frac{1}{n}}$  with  $a_i > 0$ .
4. Consider the sequence  $(e_n)$  given by  $e_n = \left(1 + \frac{1}{n}\right)^n$ . Prove that  $(e_n)$  is increasing and bounded, hence convergent. Its limit is denoted by  $e$ .

5. Find the limit for each of the following sequences: (a)  $\left(\frac{2n+1}{2n-1}\right)^n$ . (b)  $n(\ln(n+2) - \ln(n+1))$ .

6. (a) Let  $(x_n)$  be convergent. What can you say about the sequence  $(a_n)$  of averages,

$$a_n = \frac{x_1 + x_2 + \dots + x_n}{n}?$$

Can the averages converge, but the sequence not?

(b) Compare  $1 + \frac{1}{2} + \dots + \frac{1}{n}$  with  $n$  and  $\ln n$  by taking the ratio, respectively.

7. Let  $(x_n)$  be a sequence such that  $\lim_{n \rightarrow \infty} \frac{x_{n+1}}{x_n} = \ell$ . Prove that  $\lim_{n \rightarrow \infty} \sqrt[n]{x_n} = \ell$ .

8. Find the limit for each of the following sequences: (a)  $\frac{n}{\sqrt[n]{n!}}$ . (b)  $\frac{1^p + 2^p + 3^p + \dots + n^p}{n^{p+1}}$ ,  $p \in \mathbb{N}$ .

### Exercises. Homework 2

1. Prove using the  $\varepsilon$ -definition that  $\lim_{n \rightarrow \infty} \frac{n+1}{2n+3} = \frac{1}{2}$ .

2. Find the limit for each of the following sequences:

$$(a) (1 + 2 + \dots + n)^{1/n}. \quad (b) \left(\frac{\ln(n+1)}{\ln n}\right)^n. \quad (c) \frac{n^n}{1 + 2^2 + 3^3 + \dots + n^n}.$$

3. For  $x_n = \frac{\sin(n)}{n}$  study if  $(x_n)$  is bounded, monotone, and convergent. Find its limit.

4. Prove that the sequence  $(x_n)$  given by

$$x_n = 1 + \frac{1}{2} + \dots + \frac{1}{n} - \ln n$$

is decreasing and bounded, hence convergent. Its limit is typically denoted by  $\gamma$  and is known as the Euler-Mascheroni constant.

## ※ Series. Power series

For a sequence  $(x_n)$ , the sum  $\sum_{n=1}^{\infty} x_n$  is called a *series* and  $s_n := \sum_{k=1}^n x_k$  is called the *partial sum of the series*. The summation in a series can start from any index, not necessarily 0 or 1. A series is also often written as  $\sum_{n \geq 1} x_n$ .

**Definition 3.1.** The series  $\sum_{n=1}^{\infty} x_n$  converges iff the sequence of partial sums  $(s_n)$  converges.

**Example 3.2.** The *geometric series*  $\sum_{n=0}^{\infty} q^n$  converges iff  $|q| < 1$ , with sum  $\frac{1}{1-q}$ .

**Example 3.3.** The *harmonic series*  $\sum_{n=1}^{\infty} \frac{1}{n} = \infty$  diverges since  $(s_n)$  is not a Cauchy sequence.

**Example 3.4** (Euler's number).  $e = \sum_{n=0}^{\infty} \frac{1}{n!}$ .

*Proof.* (Optional) Let  $s_n = 1 + 1 + \frac{1}{2!} + \dots + \frac{1}{n!}$ . Start from  $\lim_{n \rightarrow \infty} \left(1 + \frac{1}{n}\right)^n = e$  and expand

$$\left(1 + \frac{1}{n}\right)^n = 1 + 1 + \frac{1}{2!}\left(1 - \frac{1}{n}\right) + \dots + \frac{1}{n!}\left(1 - \frac{1}{n}\right)\left(1 - \frac{2}{n}\right) \cdot \dots \cdot \left(1 - \frac{n-1}{n}\right) \leq s_n.$$

We have that

$$\left(1 + \frac{1}{n}\right)^n \leq s_n.$$

Consider now an index  $k \geq n$ . We have that

$$\left(1 + \frac{1}{k}\right)^k \geq 1 + 1 + \frac{1}{2!}\left(1 - \frac{1}{k}\right) + \dots + \frac{1}{n!}\left(1 - \frac{1}{k}\right)\left(1 - \frac{2}{k}\right) \cdot \dots \cdot \left(1 - \frac{n-1}{k}\right)$$

and taking  $k \rightarrow \infty$  we obtain that  $e \geq s_n$ . We conclude with the squeeze theorem for

$$\left(1 + \frac{1}{n}\right)^n \leq s_n \leq e,$$

obtaining that the series  $\sum_{n=0}^{\infty} \frac{1}{n!}$  converges and its sum is  $e$ . □

**Proposition 3.5.** If the series  $\sum_{n=1}^{\infty} x_n$  is convergent, then  $\lim_{n \rightarrow \infty} x_n = 0$ .

*Proof.* Consider the partial sum  $s_n$ . We have that  $x_n = s_n - s_{n-1}$ , hence the conclusion. □

It thus follows that if  $\lim_{n \rightarrow \infty} x_n \neq 0$ , then the series  $\sum_{n=1}^{\infty} x_n$  is divergent.

**Example 3.6.** Series like  $\sum_{n=1}^{\infty} \frac{1}{n(n+1)} = \sum_{n=1}^{\infty} \left( \frac{1}{n} - \frac{1}{n+1} \right) = 1$  are called *telescoping series*. The partial sum of a telescoping series can be easily computed since after cancellations the only remaining terms are the first one and the last one.

If the sequence  $(x_n)$  has only nonnegative terms  $x_n \geq 0$ , then the sequence of partial sums  $(s_n)$  is increasing. The series  $\sum_{n=1}^{\infty} x_n$  then converges iff  $(s_n)$  is bounded.

**Theorem 3.7** (Comparison test). Let  $\sum_{n=1}^{\infty} x_n, \sum_{n=1}^{\infty} y_n$  be series with nonnegative terms. If there is an  $n_0 \in \mathbb{N}$  such that

$$x_n \leq y_n, \forall n \geq n_0, \text{ then}$$

- (a) If  $\sum_{n=1}^{\infty} y_n$  converges, then  $\sum_{n=1}^{\infty} x_n$  also converges.
- (b) If  $\sum_{n=1}^{\infty} x_n$  diverges, then  $\sum_{n=1}^{\infty} y_n$  also diverges.

*Proof.* Consider the sequences of partial sums. In case (a), both sequences are bounded. In case (b), both sequences are unbounded.  $\square$

**Example 3.8.** If  $p \leq 1$ , then  $\sum_{n=1}^{\infty} \frac{1}{n^p} = \infty$  since  $\frac{1}{n^p} \geq \frac{1}{n}$  and  $\sum_{n=1}^{\infty} \frac{1}{n} = \infty$ . E.g.  $\sum_{n=1}^{\infty} \frac{1}{\sqrt{n}} = \infty$ .

**Theorem 3.9.** Let  $\sum_{n=1}^{\infty} x_n, \sum_{n=1}^{\infty} y_n$  be series with nonnegative terms. If

$$\lim_{n \rightarrow \infty} \frac{x_n}{y_n} = \ell, \text{ then}$$

- if  $\ell \in (0, \infty)$ , then the series  $\sum_{n=1}^{\infty} x_n$  and  $\sum_{n=1}^{\infty} y_n$  have the same nature.
- if  $\ell = 0$ , then if the series  $\sum_{n=1}^{\infty} y_n$  converges, the series  $\sum_{n=1}^{\infty} x_n$  also converges.
- if  $\ell = \infty$ , then if the series  $\sum_{n=1}^{\infty} y_n$  diverges, the series  $\sum_{n=1}^{\infty} x_n$  also diverges.

**Theorem 3.10** (Ratio test). Let  $\sum_{n=1}^{\infty} x_n$  be a series with nonnegative terms such that

$$\lim_{n \rightarrow \infty} \frac{x_{n+1}}{x_n} = \ell.$$

- If  $\ell < 1$ , then the series  $\sum_{n=1}^{\infty} x_n$  is convergent.
- If  $\ell > 1$ , then the series  $\sum_{n=1}^{\infty} x_n$  is divergent.

The test is *inconclusive* when  $\ell = 1$ .

*Proof.* The idea is that  $\sum_{n \geq 1} x_n$  behaves like a geometric series with ratio  $\ell$ . We will only give a proof when  $\ell < 1$ , the other case being similar.

Take  $\varepsilon > 0$  such that  $q := \ell + \varepsilon < 1$ . There exists  $N \in \mathbb{N}$  such that

$$\frac{x_{n+1}}{x_n} - \ell < \varepsilon, \forall n \geq N,$$

giving that  $x_{n+1} < x_n \cdot q$ ,  $\forall n \geq N$ . Hence  $x_n < q^{n-N} x_N$ , that is  $x_n < q^n \frac{x_N}{q^N}$ . Since  $q < 1$ , the series converges by comparison with the geometric series  $\sum_{n \geq 1} q^n$ .  $\square$

**Theorem 3.11** (Root test). Let  $\sum_{n=1}^{\infty} x_n$  be a series with nonnegative terms such that

$$\lim_{n \rightarrow \infty} \sqrt[n]{x_n} = \ell.$$

- If  $\ell < 1$ , then the series  $\sum_{n=1}^{\infty} x_n$  is convergent.
- If  $\ell > 1$ , then the series  $\sum_{n=1}^{\infty} x_n$  is divergent.

The test is *inconclusive* when  $\ell = 1$ .

*Proof.* As in the ratio test, there exists  $q \in (\ell, 1)$  and  $N \in \mathbb{N}$  such that  $\sqrt[n]{x_n} \leq q$ ,  $\forall n \geq N$ . Then  $\sum_{n \geq N} x_n \leq \sum_{n \geq N} q^n$ , which is convergent since  $q < 1$ .  $\square$

**Example 3.12.** The series  $\sum_{n \geq 0} \frac{x^n}{n!}$  converges for any  $x > 0$ . We will see later that  $\sum_{n \geq 0} \frac{x^n}{n!} = e^x$ . We have that  $\frac{x_{n+1}}{x_n} = \frac{x}{n+1} \rightarrow 0 < 1$ , hence the series converges by the ratio test.

**Theorem 3.13** (Cauchy condensation test). Let  $(x_n)$  be a decreasing sequence with  $x_n > 0$ . Then the series  $\sum_{n=1}^{\infty} x_n$  and  $\sum_{n=0}^{\infty} 2^n x_{2^n}$  have the same nature.

*Proof.* Let  $S_n = x_1 + x_2 + \dots + x_n$  and  $T_n = x_1 + 2x_2 + \dots + 2^n x_n$ . Since  $x_n > 0$ , the two series will have the same nature if and only if  $S_n$  and  $T_n$  are both bounded/unbounded.

For any  $n \in \mathbb{N}$  there exists  $k \in \mathbb{N}$  s.t.  $2^k \leq n \leq 2^{k+1} - 1$ . Since  $(x_n)$  is decreasing and positive, we can group the terms in the following ways

$$\begin{aligned} S_n &= x_1 + x_2 + \dots + x_n \leq x_1 + x_2 + \dots + x_{2^{k+1}-1} \\ &\leq x_1 + (x_2 + x_3) + \dots + (x_{2^k} + \dots + x_{2^{k+1}-1}) \\ &\leq T_k, \end{aligned}$$

and

$$\begin{aligned} S_n &= x_1 + x_2 + \dots + x_n \geq x_1 + x_2 + \dots + x_{2^k} \\ &\geq x_1 + x_2 + (x_3 + x_4) + \dots + (x_{2^{k-1}+1} + \dots + x_{2^k}) \\ &\geq \frac{x_1}{2} + \frac{1}{2} T_k. \end{aligned}$$

We obtained that  $0 \leq \frac{1}{2} T_k \leq S_n \leq T_k$ , hence  $(S_n)$  bounded if and only if  $(T_n)$  is bounded.  $\square$

**Example 3.14.** The series  $\sum_{n=1}^{\infty} \frac{1}{n^p}$  converges if and only if  $p > 1$ .

*Proof.* By the Cauchy condensation test,  $\sum_{n=1}^{\infty} \frac{1}{n^p}$  has the same nature as  $\sum_{n=1}^{\infty} \frac{2^n}{2^{np}} = \sum_{n=1}^{\infty} (2^{1-p})^n$ , which converges if and only if  $2^{1-p} < 1$ , i.e for  $p > 1$ .  $\square$

**Theorem 3.15** (Kummer's test). Let  $(x_n)$  be a positive sequence and consider another positive sequence  $(c_n)$ .

(a) If

$$\lim_{n \rightarrow \infty} \left( c_n \frac{x_n}{x_{n+1}} - c_{n+1} \right) > 0,$$

then  $\sum_{n \geq 1} x_n$  is convergent.

(b) If  $\sum_{n \geq 1} \frac{1}{c_n} = \infty$  and

$$\lim_{n \rightarrow \infty} \left( c_n \frac{x_n}{x_{n+1}} - c_{n+1} \right) < 0,$$

then  $\sum_{n \geq 1} x_n$  is divergent.

*Proof.* (Optional) Let us start with (a). Since that limit is positive, there exist  $r > 0$  and  $n_0 \in \mathbb{N}$  such that

$$c_n x_n - c_{n+1} x_{n+1} \geq r x_{n+1}, \quad \forall n \geq n_0.$$

Denote by  $s_n = x_1 + \dots + x_n$ . Adding all these inequalities for  $k \in \{n_0, \dots, n\}$  we have that

$$c_{n_0} x_{n_0} - c_{n+1} x_{n+1} \geq r(s_{n+1} - s_{n_0}),$$

which gives  $s_{n+1} \leq s_{n_0} + \frac{1}{r} c_{n_0} x_{n_0}$ . Hence  $(s_n)$  is bounded and the series converges.

Let us now consider (b). Since the limit is negative, there exists  $n_0 \in \mathbb{N}$  such that

$$c_n x_n < c_{n+1} x_{n+1}, \quad \forall n \geq n_0.$$

Hence for  $n > n_0$ , we have that  $c_{n_0} x_{n_0} < c_n x_n$ , which gives

$$\frac{1}{c_n} < \frac{1}{c_{n_0} x_{n_0}} x_n.$$

Since  $\sum_{n \geq 1} \frac{1}{c_n} = \infty$ , we conclude that  $\sum_{n \geq 1} x_n = \infty$ . □

Many convergence tests can be obtained by taking particular sequences in Kummer's test. We will restrict to the following one.

**Theorem 3.16** (Raabe-Duhamel). Let  $\sum_{n \geq 1} x_n$  be a series with positive terms such that

$$\lim_{n \rightarrow \infty} n \left( \frac{x_n}{x_{n+1}} - 1 \right) = R.$$

- If  $R > 1$ , then the series  $\sum_{n=1}^{\infty} x_n$  is convergent.
- If  $R < 1$ , then the series  $\sum_{n=1}^{\infty} x_n$  is divergent.

*Proof.* Take  $c_n = n$  in Kummer's test ([Theorem 3.15](#)). □



**Example 3.17.** Study the convergence of the series  $\sum_{n \geq 0} \frac{n!}{a(a+1)\dots(a+n)}$ , with  $a > 0$ .

*Proof.* The ratio test is inconclusive since  $\frac{x_{n+1}}{x_n} = \frac{n+1}{a+n+1} \rightarrow 1$ . Let us then try the Raabe-Duhamel test:

$$\lim_{n \rightarrow \infty} n \left( \frac{x_n}{x_{n+1}} - 1 \right) = \lim_{n \rightarrow \infty} n \left( \frac{a+n+1}{n+1} - 1 \right) = a.$$

Hence if  $a > 1$  the series converges; and if  $a < 1$  the series diverges. When  $a = 1$  the series is  $\sum_{n \geq 0} \frac{1}{n+1} = \infty$ .  $\square$

### Exercises. Seminar 3

1. Find the sum for each of the following series:

(a)  $\sum_{n \geq 1} \frac{2}{3^n}$ .

(c)  $\sum_{n \geq 1} \frac{1}{4n^2 - 1}$ .

(e)  $\sum_{n \geq 1} \frac{n}{2^n}$ .

(b)  $\sum_{n \geq 1} \frac{2n+1}{n!}$ .

(d)  $\sum_{n \geq 1} \frac{1}{n(n+1)(n+2)}$ .

(f)  $\sum_{n \geq 1} \frac{1}{\sqrt[3]{n}}$ .

2. Study the convergence of the following series:

(a)  $\sum_{n \geq 2} \frac{1}{\ln n}$ .

(c)  $\sum_{n \geq 1} \frac{\ln(1 + \frac{1}{n})}{n}$ .

(e)  $\sum_{n \geq 1} \left( \frac{n}{n+1} \right)^{n^2}$ .

(b)  $\sum_{n \geq 1} \frac{1}{n\sqrt{n+1}}$ .

(d)  $\sum_{n \geq 1} \frac{n!}{(n+1)^n}$ .

(f)  $\sum_{n \geq 2} \frac{1}{n \ln(n)}$ .

3. Study the convergence of the following series:

(a)  $\sum_{n \geq 1} \frac{1 \cdot 3 \cdot \dots \cdot (2n-1)}{2 \cdot 4 \cdot \dots \cdot 2n}$ .

(b)  $\sum_{n \geq 1} a^{\ln n}, a > 0$ .

(c)  $\sum_{n \geq 1} \frac{a^n n!}{n^n}, a > 0$ .

### Exercises. Homework 3

1. Find the sum for each of the following series:

(a)  $\sum_{n \geq 2} \ln \left( 1 - \frac{1}{n^2} \right)$ .

(b)  $\sum_{n \geq 1} \frac{n+1}{3^n}$ .

(c)  $\sum_{n \geq 1} \frac{n}{n^4 + n^2 + 1}$ .

2. Study the convergence of the following series:

(a)  $\sum_{n \geq 1} \frac{x^n}{n^p}, x > 0, p \in \mathbb{N}.$

(b)  $\sum_{n \geq 2} \frac{1}{(\ln n)^{\ln n}}.$

(c)  $\sum_{n \geq 1} (\sqrt[n]{n} - 1).$

(d)  $\sum_{n \geq 1} \frac{1 \cdot 3 \cdot \dots \cdot (2n-1)}{2 \cdot 4 \cdot \dots \cdot 2n} \cdot \frac{1}{2n+1}.$

3. Start with an equilateral triangle of side 1. For each side, remove the middle third and add there another equilateral triangle. Repeat this process at each iteration (see the figure).

How many sides are there at iteration  $n$ ? What is the limit of the perimeter and the area?



A series  $\sum_{n \geq 1} x_n$  is called an *alternating series* if  $x_n x_{n+1} \leq 0$ ,  $\forall n \in \mathbb{N}$ . A fundamental class of alternating series are series of the form  $\sum_{n \geq 1} (-1)^n a_n$  or  $\sum_{n \geq 1} (-1)^{n+1} a_n$ , with  $a_n > 0$ .

**Example 3.18.** The series  $\sum_{n \geq 1} \frac{(-1)^{n+1}}{n} = 1 - \frac{1}{2} + \frac{1}{3} - \frac{1}{4} + \dots$  converges to  $\ln 2$ .

*Proof.* Let us prove convergence by considering the partial sums  $s_n = \sum_{k=1}^n \frac{(-1)^{k+1}}{k}$ . Notice that  $s_{2k+2} - s_{2k} = \frac{1}{2k+1} - \frac{1}{2k+2} > 0$  and that  $s_{2k+3} - s_{2k+1} = \frac{1}{2k+3} - \frac{1}{2k+2} < 0$ . This means that the subsequence  $(s_{2k})$  is increasing, while the subsequence  $(s_{2k+1})$  is decreasing. Notice also that  $s_{2k+1} - s_{2k} = \frac{1}{2k+1}$  and  $s_{2k} < s_{2k+1}$ , so both subsequences are also bounded and converge to the same limit. To find the sum of the alternating series, recall (from the seminar) that

$$\lim_{n \rightarrow \infty} 1 + \frac{1}{2} + \dots + \frac{1}{n} - \ln n = \gamma \in (0, 1), \text{ hence}$$

$$\begin{aligned} s_{2n} &= 1 - \frac{1}{2} + \frac{1}{3} - \dots - \frac{1}{2n} = 1 + \frac{1}{2} + \dots + \frac{1}{2n} - 2\left(\frac{1}{2} + \dots + \frac{1}{2n}\right) \\ &= \underbrace{1 + \frac{1}{2} + \dots + \frac{1}{2n} - \ln(2n)}_{\rightarrow \gamma} - \underbrace{\left(1 + \frac{1}{2} + \dots + \frac{1}{n} - \ln n\right)}_{\rightarrow \gamma} + \ln 2 \rightarrow \ln 2. \end{aligned}$$

□

**Theorem 3.19** (Leibniz test). Let  $(x_n)$  be a decreasing sequence with  $x_n \rightarrow 0$ . Then the series  $\sum_{n \geq 1} (-1)^n x_n$  is convergent.

*Proof.* Consider the partial sum  $s_n = \sum_{k=1}^n (-1)^k x_k$ . We will prove that  $(s_n)$  is convergent by showing that it is a Cauchy sequence. For  $n, p \in \mathbb{N}$  consider

$$\begin{aligned} |s_{n+p} - s_n| &= |(-1)^{n+1} x_{n+1} + \dots + (-1)^{n+p} x_{n+p}| \\ &= \underbrace{|x_{n+1} - x_{n+2}|}_{\geq 0} + \underbrace{|x_{n+3} - x_{n+4}|}_{\geq 0} + \dots + (-1)^{p-2} x_{n+p-1} + (-1)^{p-1} x_{n+p}| \\ &= x_{n+1} - \underbrace{x_{n+2} + x_{n+3} - x_{n+4}}_{\leq 0} + \dots \pm x_{n+p-1} \mp x_{n+p} \\ &\leq x_{n+1}. \end{aligned}$$

Since  $x_n \rightarrow 0$ ,  $|s_{n+p} - s_n|$  can be made arbitrarily small, so  $(s_n)$  is Cauchy. □

**Definition 3.20.** A series  $\sum_{n \geq 1} x_n$  is called *absolutely convergent* if  $\sum_{n \geq 1} |x_n|$  is convergent.

**Proposition 3.21.** Any absolutely convergent series is also convergent.

*Proof.* If  $\sum_{k=1}^n |x_k|$  gives a Cauchy sequence, then  $\sum_{k=1}^n x_k$  also gives a Cauchy sequence.  $\square$

**Theorem 3.22 (Cauchy).** Let  $\sum_{n \geq 1} x_n$  be an *absolutely convergent series* and let  $\sigma : \mathbb{N} \rightarrow \mathbb{N}$  be a bijection. Then  $\sum_{n \geq 1} x_{\sigma(n)}$  is also absolutely convergent and  $\sum_{n \geq 1} x_{\sigma(n)} = \sum_{n \geq 1} x_n$ . In other words, any rearrangement of an absolutely convergent series has the same sum.

*Proof.* (Optional) See [2][Theorem 7.4.3].  $\square$

**Definition 3.23.** A series  $\sum_{n \geq 1} x_n$  is called *conditionally convergent* (or *semi-convergent*) if  $\sum_{n \geq 1} x_n$  converges, but  $\sum_{n \geq 1} |x_n|$  diverges.

**Theorem 3.24 (Riemann).** Let  $\sum_{n \geq 1} x_n$  be a *conditionally convergent series* and let  $x \in \overline{\mathbb{R}}$ . Then there exists a bijection  $\sigma : \mathbb{N} \rightarrow \mathbb{N}$  such that  $\sum_{n \geq 1} x_{\sigma(n)} = x$ . In other words, a conditionally convergent series can be rearranged to converge to any value or diverge to  $\pm\infty$ .

*Proof.* (Optional) See [2][Theorem 8.2.8].  $\square$

**Example 3.25.** Rearranging the terms in the alternating harmonic series one can obtain a different sum. Indeed, consider  $\sum_{n \geq 1} \frac{(-1)^{n+1}}{n} = 1 - \frac{1}{2} + \frac{1}{3} - \frac{1}{4} + \dots = \ln 2$ , and reorder the terms in the following way: one positive, two negative. Then

$$1 - \frac{1}{2} - \frac{1}{4} + \frac{1}{3} - \frac{1}{6} - \frac{1}{8} + \dots = \frac{1}{2} - \frac{1}{4} + \frac{1}{6} - \frac{1}{8} + \dots = \frac{1}{2} \left( 1 - \frac{1}{2} + \frac{1}{3} - \frac{1}{4} + \dots \right) = \frac{1}{2} \ln 2.$$

**Definition 3.26.** Let  $(a_n)$  be a sequence of real numbers and let  $c \in \mathbb{R}$ . The series

$$\sum_{n=0}^{\infty} a_n (x - c)^n$$

is called a *power series* centered at  $c$ .

**Theorem 3.27.** Consider the power series  $\sum_{n=0}^{\infty} a_n (x - c)^n$ . There exists a unique  $R \in [0, \infty]$ , called the *radius of convergence* of the power series, such that the power series

- converges absolutely when  $|x - c| < R$ .
- diverges when  $|x - c| > R$ .

**Theorem 3.28.** If the limit

$$\lim_{n \rightarrow \infty} \sqrt[n]{|a_n|} = L \in [0, \infty]$$

exists, then the power series  $\sum_{n=0}^{\infty} a_n(x - c)^n$  has the radius of convergence

$$R = \begin{cases} \frac{1}{L}, & \text{if } L \in (0, \infty) \\ 0, & \text{if } L = \infty \\ \infty, & \text{if } L = 0. \end{cases}$$

*Proof.* It follows from the root test for series with positive terms. □

**Corollary 3.29.** If the limit

$$\lim_{n \rightarrow \infty} \frac{|a_{n+1}|}{|a_n|} = L \in [0, \infty]$$

exists, then the power series  $\sum_{n=0}^{\infty} a_n(x - c)^n$  has the radius of convergence

$$R = \begin{cases} \frac{1}{L}, & \text{if } L \in (0, \infty) \\ 0, & \text{if } L = \infty \\ \infty, & \text{if } L = 0. \end{cases}$$

*Proof.* It follows from  $\lim_{n \rightarrow \infty} \sqrt[n]{|a_n|} = \lim_{n \rightarrow \infty} \frac{|a_{n+1}|}{|a_n|} = L$ . □

**Definition 3.30.** The convergence set of a power series is

$$C := \{x \in \mathbb{R} \mid \sum_{n=0}^{\infty} a_n(x - c)^n \text{ converges}\}.$$

**Remark 3.31.** The convergence set  $C$  contains the open interval  $(c - R, c + R)$  and possibly the endpoints  $\{c - R, c + R\}$ .

**Example 3.32.** The power series  $\sum_{n \geq 0} x^n$  has radius of convergence  $R = 1$ , it converges absolutely for  $|x| < 1$  and diverges when  $|x| > 1$  (by the root test or the ratio test). The convergence set is  $(-1, 1)$  and for  $x \in (-1, 1)$  we have that

$$\sum_{n \geq 0} x^n = \frac{1}{1 - x}, \quad \sum_{n \geq 0} (-x)^n = \frac{1}{1 + x}.$$

**Example 3.33.** The power series  $\sum_{n \geq 1} \frac{x^n}{n}$  has radius of convergence  $R = 1$ , it converges absolutely for  $|x| < 1$  and diverges when  $|x| > 1$  (by the root test or the ratio test). Moreover, the series converges for  $x = -1$  (alternating harmonic series) and diverges for  $x = 1$  (harmonic series), hence its convergence set is  $C = [-1, 1)$ .

**Theorem 3.34.** Consider a power series with radius of convergence  $R$ , given by

$$s(x) = \sum_{n=0}^{\infty} a_n(x - c)^n.$$

Then for any  $x \in (c - R, c + R)$ , the power series can be differentiated term by term and

$$s'(x) = \sum_{n=1}^{\infty} n a_n(x - c)^{n-1},$$

and for any  $t \in (c - R, c + R)$  the power series can be integrated term by term

$$\int_c^t s(x) dx = \sum_{n=0}^{\infty} \frac{a_n}{n+1} (t - c)^{n+1}.$$

**Example 3.35.** The power series  $\sum_{n \geq 0} \frac{x^n}{n!}$  converges absolutely for any  $x \in \mathbb{R}$  (ratio test). Let  $\exp(x) := \sum_{n \geq 0} \frac{x^n}{n!}$  and differentiate term by term, then  $\exp'(x) = \exp(x)$  and  $\exp(0) = 1$ .

#### Exercises. Seminar 4

1. Study the convergence and the absolute convergence of the following series:

$$(a) \sum_{n \geq 1} \frac{(-1)^{n+1}}{\sqrt{n(n+1)}}. \quad (b) \sum_{n \geq 1} (-1)^n \sin \frac{1}{n}. \quad (c) \sum_{n \geq 1} \frac{\sin n}{n^2}.$$

2. Prove by differentiating the geometric series that, for  $|x| < 1$ ,

$$\sum_{n=1}^{\infty} n x^n = \frac{x}{(1-x)^2}, \quad \sum_{n=2}^{\infty} n(n-1)x^n = \frac{2x^2}{(1-x)^3}.$$

3. Prove by integrating the geometric series that, for  $|x| < 1$ ,

$$\sum_{n=1}^{\infty} \frac{x^n}{n} = -\ln(1-x), \quad \sum_{n=1}^{\infty} (-1)^{n+1} \frac{x^n}{n} = \ln(1+x).$$

4. Prove that  $\sum_{n=0}^{\infty} (-1)^n \frac{x^{2n+1}}{2n+1} = x - \frac{x^3}{3} + \frac{x^5}{5} - \dots = \arctan x$ , for  $x \in [-1, 1]$ .

5. Find the radius of convergence and the convergence set for each of the following series:

$$\begin{array}{lll} \text{(a)} \sum_{n \geq 0} \frac{x^n}{2^n}. & \text{(c)} \sum_{n \geq 1} \frac{(x-2)^n}{(n+1)3^n}. & \text{(e)} \sum_{n \geq 1} \frac{(x-1)^n}{n^p}, p > 0. \\ \text{(b)} \sum_{n \geq 0} 2^n (x-1)^n. & \text{(d)} \sum_{n \geq 0} n! x^n. & \text{(f)} \sum_{n \geq 1} \frac{(-1)^n x^n}{\sqrt{n}}. \end{array}$$

**Exercises. Homework 4**

1. Find the radius of convergence and the convergence set for each of the following series:

$$\begin{array}{lll} \text{(a)} \sum_{n \geq 0} \frac{n x^n}{2^n}. & \text{(b)} \sum_{n \geq 1} \frac{x^{2n}}{\sqrt{n}}. & \text{(c)} \sum_{n \geq 1} (-1)^n \frac{(x-1)^n}{n}. \end{array}$$

2. Study the convergence and compute the sum for the series  $\sum_{n \geq 2} \frac{x^n}{n(n-1)}$ .

3. For this question you have to use Python.

Illustrate that  $\sum_{n \geq 1} \frac{(-1)^{n+1}}{n} = \ln 2$  by computing partial sums.

Change the order of summation in this series – for example by first adding  $p$  positive terms, then  $q$  negative terms, and so on – and show numerically that the rearrangement gives a different sum (depending on  $p, q$ ).

## ※ Limits, continuity, differentiability

**Definition 4.1.** Let  $A \subseteq \mathbb{R}$ . We say that  $x_0 \in \overline{\mathbb{R}}$  is an *accumulation point* (or *cluster point*) if

$$\forall V \in \mathcal{V}(x_0), V \cap (A \setminus \{x_0\}) \neq \emptyset.$$

We denote by  $A'$  the set of the accumulation points of  $A$ . We say that  $x_0 \in A$  is an *isolated point* if  $x_0 \in A \setminus A'$ .

**Remark 4.2.**  $\text{cl}(A) = A' \cup \{\text{isolated points}\}$ .

**Proposition 4.3.** Let  $A \subseteq \mathbb{R}$  and  $x_0 \in \overline{\mathbb{R}}$ , then  $x_0 \in A'$  if and only if there exists a sequence  $(x_n)$  in  $A \setminus \{x_0\}$  such that  $\lim_{n \rightarrow \infty} x_n = x_0$ .

*Proof.* Assume that  $x_0 \in A'$ , with  $x_0 \in \mathbb{R}$ , and consider the neighborhoods  $(x_0 - \frac{1}{n}, x_0 + \frac{1}{n})$ . Then each neighborhood must contain an  $x_n \in A \setminus \{x_0\}$  with  $|x_n - x_0| < \frac{1}{n}$ , hence  $x_n \rightarrow x_0$ . If  $x_0$  is infinite, the neighborhoods can be taken  $(-\infty, -n)$  or  $(n, \infty)$ , respectively.

Assume now that there exists a sequence  $(x_n)$  in  $A \setminus \{x_0\}$  such that  $\lim_{n \rightarrow \infty} x_n = x_0$ . Then for any  $V \in \mathcal{V}(x_0)$ , there exists  $N_V \in \mathbb{N}$  such that  $x_n \in V$ , for any  $n \geq N_V$ . In particular,  $x_{N_V} \in V \cap (A \setminus \{x_0\})$ , for any  $V \in \mathcal{V}(x_0)$ , hence  $x_0 \in A'$ .  $\square$

**Example 4.4.** For  $A = \{\frac{1}{n} \mid n \in \mathbb{N}\}$ , each element  $x \in A$  is an isolated point and  $A' = \{0\}$ .

**Definition 4.5.** Let  $A \subseteq \mathbb{R}$ ,  $f : A \rightarrow \mathbb{R}$  and  $x_0 \in A'$ . We say that  $\lim_{x \rightarrow x_0} f(x) = \ell \in \overline{\mathbb{R}}$  if

$$\forall V \in \mathcal{V}(\ell), \exists U \in \mathcal{V}(x_0) \text{ s.t. } f(x) \in V, \forall x \in U \cap (A \setminus \{x_0\}).$$

**Remark 4.6** ( $\varepsilon$ - $\delta$ ). Let  $A \subseteq \mathbb{R}$ ,  $f : A \rightarrow \mathbb{R}$  and  $x_0 \in A'$  finite. If  $\lim_{x \rightarrow x_0} f(x) = \ell \in \mathbb{R}$ , then

$$\forall \varepsilon > 0, \exists \delta > 0 \text{ s.t. } |f(x) - \ell| < \varepsilon, \forall x \in A \text{ with } |x - x_0| < \delta.$$

**Theorem 4.7.** Let  $A \subseteq \mathbb{R}$ ,  $f : A \rightarrow \mathbb{R}$  and  $x_0 \in A'$ . Then  $\lim_{x \rightarrow x_0} f(x) = \ell \in \overline{\mathbb{R}}$  iff

for any sequence  $(x_n)$  in  $A \setminus \{x_0\}$  with  $\lim_{n \rightarrow \infty} x_n = x_0$ , we have that  $\lim_{n \rightarrow \infty} f(x_n) = \ell$ .

**Theorem 4.8.** Let  $A \subseteq \mathbb{R}$ ,  $f : A \rightarrow \mathbb{R}$ ,  $x_0 \in \mathbb{R}$  s.t.  $x_0 \in (A \cap (-\infty, x_0))'$  and  $x_0 \in (A \cap (x_0, \infty))'$ . Then

$$\lim_{x \rightarrow x_0} f(x) = \ell \text{ iff } \lim_{\substack{x \rightarrow x_0 \\ x < x_0}} f(x) = \lim_{\substack{x \rightarrow x_0 \\ x > x_0}} f(x) = \ell.$$

**Example 4.9.** (a)  $\text{sgn} : \mathbb{R} \rightarrow \mathbb{R}$ ,  $\text{sgn}(x) = \begin{cases} +1, & \text{if } x \geq 0 \\ -1, & \text{if } x < 0. \end{cases}$  has no limit at 0.



(b)  $f : \mathbb{R}^* \rightarrow \mathbb{R}$ ,  $f(x) = \sin(\frac{1}{x})$  has no limit at 0 since  $f(\frac{1}{2n\pi}) = 0$ ,  $f(\frac{1}{2n\pi + \pi/2}) = 1$ .

(c)  $f : \mathbb{R} \rightarrow \mathbb{R}$ ,  $f(x) = \begin{cases} 1, & \text{if } x \in \mathbb{Q} \\ 0, & \text{if } x \in \mathbb{R} \setminus \mathbb{Q} \end{cases}$  has no limit at any  $x \in \mathbb{R}$ .

**Definition 4.10.** Let  $A \subseteq \mathbb{R}$ ,  $f : A \rightarrow \mathbb{R}$  and  $x_0 \in A$ . We say that  $f$  is *continuous* at  $x_0$  if

$$\forall V \in \mathcal{V}(f(x_0)), \exists U \in \mathcal{V}(x_0) \text{ s.t. } f(x) \in V, \forall x \in U \cap A.$$

**Remark 4.11.** If  $x_0 \in A \cap A'$  is an accumulation point, then  $f$  is continuous at  $x_0$  if

$$\lim_{x \rightarrow x_0} f(x) = f(x_0).$$

**Remark 4.12.** If  $x_0$  is an isolated point of  $A$ , then  $\exists U \in \mathcal{V}(x_0)$  with  $U \cap A = \{x_0\}$ , and since  $f(x_0) \in V$ ,  $\forall V \in \mathcal{V}(f(x_0))$ , we have that  $f$  is continuous at  $x_0$ .

**Theorem 4.13.** Let  $A \subseteq \mathbb{R}$ ,  $f : A \rightarrow \mathbb{R}$  and  $x_0 \in A \cap A'$ . The following are equivalent:

- (a)  $f$  is continuous at  $x_0$ .
- (b)  $\forall \varepsilon > 0, \exists \delta > 0$  s.t.  $|f(x) - f(x_0)| < \varepsilon$ ,  $\forall x \in A$  with  $|x - x_0| < \delta$ .
- (c) for any sequence  $(x_n)$  in  $A$  with  $\lim_{n \rightarrow \infty} x_n = x_0$ , we have that  $\lim_{n \rightarrow \infty} f(x_n) = f(x_0)$ .

**Remark 4.14.** Elementary operations – e.g. sums, products or compositions – of continuous functions are continuous (when defined).

**Definition 4.15.** For  $f : A \rightarrow \mathbb{R}$  denote by  $f(A) := \{f(x) \mid x \in A\}$  the image of  $A$ . We say that  $f$  is *bounded* if  $f(A)$  is *bounded*, i.e.  $\inf(f(A)), \sup(f(A))$  are finite.

**Theorem 4.16 (Weierstrass).** Let  $f : [a, b] \rightarrow \mathbb{R}$  be a continuous function. Then  $f$  is bounded and it attains its bounds, i.e. there exist  $\min(f(A)), \max(f(A))$ .

*Proof.* Let us first prove that  $f$  is bounded. Assuming that this is not the case, we have that for any  $n \in \mathbb{N}$  there exists  $x_n \in [a, b]$  such that  $|f(x_n)| > n$ . Since the sequence  $(x_n)$  is bounded, we have that it has a convergent subsequence  $(x_{n_k})$ , see [Theorem 2.9](#); denote its limit by  $x$ . We have that  $x_{n_k} \rightarrow x$  and  $f$  is continuous, hence  $f(x_{n_k}) \rightarrow f(x)$ . But  $|f(x_{n_k})| > n_k \rightarrow \infty$ , contradiction. Hence  $f$  is bounded on  $[a, b]$ .

To prove that  $f$  attains its bounds, let's consider the upper bound and show that there exists  $x_M \in [a, b]$  such that  $f(x_M) = \sup(f(A))$ , i.e.  $f(x_M) = \max(f(A)) = \sup(f(A))$ . From [Theorem 1.7](#), we obtain a sequence  $(x_n)$  in  $[a, b]$  such that  $f(x_n) \rightarrow \sup(f(A))$ . Since the sequence  $(x_n)$  is bounded, it has a convergent subsequence  $(x_{n_k})$ ; let's call its limit  $x_M \in [a, b]$ . Since  $f$  is continuous, it follows that  $f(x_{n_k}) \rightarrow f(x_M)$ , but we know that  $f(x_{n_k}) \rightarrow \sup(f(A))$ , hence  $f(x_M) = \sup(f(A))$  and  $f$  reaches its upper bound.  $\square$

**Theorem 4.17** (Intermediate value property). Let  $f : [a, b] \rightarrow \mathbb{R}$  be a continuous function. Then  $f$  has the intermediate value property, i.e. if  $y \in \mathbb{R}$  is in between  $f(a)$  and  $f(b)$ , there exists  $c \in (a, b)$  such that  $f(c) = y$ .

*Proof.* Assume that  $f(a) < y < f(b)$  and consider the set  $S := \{x \in [a, b] \mid f(x) \leq y\}$ . Take

$$c := \sup(S)$$

Let  $\varepsilon > 0$ , then  $\exists \delta > 0$  such that  $|f(x) - f(c)| < \varepsilon$ , whenever  $|x - c| < \delta$ . Since  $c = \sup(S)$ , we have from [Theorem 1.7](#) that there exists  $x_1 \in S$  such that  $c - \delta < x_1 \leq c$ . From continuity we have that  $f(c) < f(x_1) + \varepsilon \leq y + \varepsilon$ . Also, for  $x_2 \in (c, c + \delta)$ , we have from continuity that  $f(c) > f(x_2) - \varepsilon$ . From the definition of the supremum,  $x_2 \notin S$  hence  $f(x_2) > y$  and  $f(c) > y - \varepsilon$ . We conclude that  $y - \varepsilon < f(c) < y + \varepsilon$ , for any  $\varepsilon > 0$ . Hence  $f(c) = y$ .  $\square$

**Definition 4.18.** Let  $A \subseteq \mathbb{R}$ ,  $f : A \rightarrow \mathbb{R}$  and  $x_0 \in A \cap A'$ . The *derivative* of  $f$  at  $x_0$  is

$$f'(x_0) := \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0} \in \overline{\mathbb{R}}$$

If  $f'(x_0) \in \mathbb{R}$  (finite) we say that  $f$  is *differentiable* at  $x_0$ .

**Remark 4.19.**  $f'(x_0)$  represents the gradient of the tangent to the curve  $y = f(x)$  at the point  $(x_0, f(x_0))$ . The equation of the tangent is  $f(x) - f(x_0) = f'(x_0)(x - x_0)$ .

**Theorem 4.20.** Let  $A \subseteq \mathbb{R}$ ,  $f : A \rightarrow \mathbb{R}$  and  $x_0 \in A \cap A'$ . If  $f$  is differentiable at  $x_0$ , then  $f$  is continuous at  $x_0$ .

*Proof.* Since  $f(x) = f(x_0) + \frac{f(x) - f(x_0)}{x - x_0}(x - x_0)$ , we have that  $\lim_{x \rightarrow x_0} f(x) = f(x_0) + 0 = f(x_0)$ .  $\square$

**Example 4.21.**  $f : \mathbb{R} \rightarrow \mathbb{R}$ ,  $f(x) = |x|$  is not differentiable in 0 since  $\nexists \lim_{x \rightarrow 0} \frac{|x|}{x}$ .

**Theorem 4.22** (Calculus Rules).

- $(cf)'(x) = cf'(x)$ , for any constant  $c \in \mathbb{R}$ .
- $(f + g)'(x) = f'(x) + g'(x)$ .
- $(fg)'(x) = f'(x)g(x) + f(x)g'(x)$ . (Product Rule)
- $\left(\frac{f}{g}\right)'(x) = \frac{f'(x)g(x) - f(x)g'(x)}{g^2(x)}$ . (Quotient Rule)
- $(f \circ g)'(x) = f'(g(x))g'(x)$ . (Chain Rule)

**Proposition 4.23** (l'Hôpital's rule). Let  $I$  be an open interval,  $x_0 \in \overline{\mathbb{R}}$  and  $f, g : I \setminus \{x_0\} \rightarrow \mathbb{R}$  differentiable. If  $\lim_{x \rightarrow x_0} f(x) = \lim_{x \rightarrow x_0} g(x) = 0$  or  $\pm\infty$ , and  $\lim_{x \rightarrow x_0} \frac{f'(x)}{g'(x)} \in \overline{\mathbb{R}}$ , then

$$\lim_{x \rightarrow x_0} \frac{f(x)}{g(x)} = \lim_{x \rightarrow x_0} \frac{f'(x)}{g'(x)}.$$

**Definition 4.24.**  $f : A \rightarrow \mathbb{R}$  has a local extremum (minimum or maximum) at  $x_0 \in A$  if

$$\exists V \in \mathcal{V}(x_0) \text{ s.t. } f(x_0) \leq f(x) \text{ or } f(x_0) \geq f(x), \forall x \in V \cap A.$$

**Theorem 4.25** (Fermat). Let  $f : (a, b) \rightarrow \mathbb{R}$  and  $x_0 \in (a, b)$ . If  $f$  is differentiable at  $x_0$  and  $x_0$  is a local extremum, then  $f'(x_0) = 0$ .

*Proof.* The lateral derivatives at  $x_0$  are equal. Since  $x_0$  is a local extremum, one of them is  $\geq 0$ , the other  $\leq 0$ . Hence  $f'(x_0) = 0$ .  $\square$

**Theorem 4.26** (Rolle). Let  $f : (a, b) \rightarrow \mathbb{R}$  with  $f(a) = f(b)$ . If  $f$  is continuous on  $[a, b]$  and differentiable on  $(a, b)$ , then there exists  $c \in (a, b)$  s.t.  $f'(c) = 0$ .

*Proof.* Since  $f$  is continuous, it is bounded and it attains its bounds. Denote by  $x_m$  and  $x_M$  the minimum and maximum points of  $f$  on  $[a, b]$ . If at least one of  $x_m$  and  $x_M$  belongs to  $(a, b)$ , then  $f'(x_m) = 0$  or  $f'(x_M) = 0$ . Otherwise,  $x_m, x_M \in \{a, b\}$  and  $f(x_m) = f(x_M)$ , hence the function is constant and its derivative is zero on  $(a, b)$ .  $\square$

**Theorem 4.27** (Mean value theorem). Let  $f : (a, b) \rightarrow \mathbb{R}$  be continuous on  $[a, b]$  and differentiable on  $(a, b)$ . Then there exists  $c \in (a, b)$  s.t.

$$f'(c) = \frac{f(b) - f(a)}{b - a}.$$

*Proof.* Consider the function  $g : (a, b) \rightarrow \mathbb{R}$ ,  $g(x) := f(x) - x \frac{f(b) - f(a)}{b - a}$ . Since  $g(a) = g(b)$ , the conclusion follows from Rolle's theorem.  $\square$

**Theorem 4.28** (Monotony). Let  $f : (a, b) \rightarrow \mathbb{R}$  be differentiable on  $(a, b)$ . Then

$$f \text{ is increasing iff } f' \geq 0,$$

$$f \text{ is decreasing iff } f' \leq 0.$$

*Proof.*  $\Rightarrow$  follows from the definition of the derivative;  $\Leftarrow$  from the mean value theorem.  $\square$

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**Exercises. Seminar 5**

1. Find the accumulation points for each of the following sets:

$$[0, 1) \cup \{2\}, \mathbb{Z}, \mathbb{Q}, \{0.1, 0.11, \dots\}.$$

2. Find a function  $f : \mathbb{R} \rightarrow \mathbb{R}$  that is discontinuous everywhere, with  $|f|$  continuous everywhere.
3. Prove that a continuous function  $f : [a, b] \rightarrow [a, b]$  has at least one fixed point  $x^* = f(x^*)$ .
4. Study the continuity and the differentiability of the functions  $f$  and  $f'$ , where  $f : \mathbb{R} \rightarrow \mathbb{R}$ ,

$$f(x) = \begin{cases} x^2 \sin \frac{1}{x}, & \text{if } x \neq 0 \\ 0, & \text{if } x = 0. \end{cases}$$

5. Prove (from scratch) that  $\lim_{x \rightarrow 0} \frac{\sin x}{x} = 1$ . Then prove that  $(\sin x)' = \cos x$ ,  $(\cos x)' = -\sin x$ .
6. Compute the following limits:

(a)  $\lim_{x \rightarrow \infty} \frac{\lfloor x \rfloor}{x}.$

(d)  $\lim_{\substack{x \rightarrow 0 \\ x > 0}} x^x.$

(b)  $\lim_{x \rightarrow \infty} x(\ln(x+2) - \ln(x+1)).$

(e)  $\lim_{\substack{x \rightarrow 0 \\ x > 0}} (\sin x)^x.$

(c)  $\lim_{x \rightarrow 0} (\cos x)^{\frac{1}{x^2}}.$

(f)  $\lim_{x \rightarrow \infty} x((1 + \frac{1}{x})^x - e).$

7. Find the  $n^{\text{th}}$  derivative of the following functions:

(a)  $f : (-1, \infty) \rightarrow \mathbb{R}, f(x) = \ln(1+x).$

(c)  $f : \mathbb{R} \rightarrow \mathbb{R}, f(x) = x^2 \sin x.$

(b)  $f : \mathbb{R} \rightarrow \mathbb{R}, f(x) = \sin x.$

(d)  $f : \mathbb{R} \rightarrow \mathbb{R}, f(x) = e^{2x} x^3.$

**Exercises. Homework 5**

1. Let  $f : \mathbb{R} \rightarrow \mathbb{R}$  be a differentiable function for which we want to find a local minimum using the following *gradient descent* method: define

$$x_{n+1} = x_n - \eta f'(x_n),$$

where  $x_1 \in \mathbb{R}$  is a starting value and  $\eta > 0$  is the step size (also called learning rate).

Notice that if  $f'(x_n) > 0$ , then  $x_{n+1} < x_n$ ; if  $f'(x_n) < 0$ , then  $x_{n+1} > x_n$ ; if  $f'(x_n) = 0$ , then  $x_{n+1} = x_n$  and the algorithm would end. If the sequence  $(x_n)$  has a limit  $x$ , then  $f'(x) = 0$ .

Using Python explore the following:

- (a) Take a particular convex function  $f$  and show that for a small value of  $\eta$  the method converges to the minimum of  $f$ .
- (b) Show that by using a larger  $\eta$  the method can converge faster (in fewer steps).
- (c) Show that taking  $\eta$  too large might lead to the divergence of the method.
- (d) Take a particular nonconvex function  $f$  and show that, depending on the starting value, the method can end up in a local minimum, not in the global minimum.

## ※ Taylor series

Let  $I \subseteq \mathbb{R}$  be an interval,  $f : I \rightarrow \mathbb{R}$  and  $x_0 \in I$  a point where  $f$  is differentiable  $n$  times ( $n \in \mathbb{N}$ ). We want to construct a polynomial  $P : \mathbb{R} \rightarrow \mathbb{R}$  that matches the function  $f$  and all its derivatives up to order  $n$  at the point  $x_0$ . That is, we want to have

$$P(x_0) = f(x_0), P'(x_0) = f'(x_0), P''(x_0) = f''(x_0), \dots, P^{(n)}(x_0) = f^{(n)}(x_0).$$

Let us consider a polynomial  $P$  of degree at most  $n$  of the following form

$$P(x) = a_0 + a_1(x - x_0) + a_2(x - x_0)^2 + \dots + a_n(x - x_0)^n.$$

By imposing the conditions at  $x_0$  and differentiating  $P$  we have that

$$\begin{aligned} f(x_0) &= P(x_0) = a_0, \\ f'(x_0) &= P'(x_0) = a_1, \\ f''(x_0) &= P''(x_0) = 2a_2, \\ &\vdots \\ f^{(n)}(x_0) &= P^{(n)}(x_0) = n!a_n. \end{aligned}$$

We thus see that there exists a unique such polynomial  $P$  of degree at most  $n$  given by

$$P(x) = f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2}(x - x_0)^2 + \dots + \frac{f^{(n)}(x_0)}{n!}(x - x_0)^n,$$

that matches the function  $f$  and all its derivatives up to order  $n$  at the point  $x_0$ .

**Definition 5.1.** Let  $I \subseteq \mathbb{R}$  be an interval,  $f : I \rightarrow \mathbb{R}$  and  $x_0 \in I$  where  $f$  is differentiable  $n$  times. The polynomial  $T_n : \mathbb{R} \rightarrow \mathbb{R}$ ,

$$T_n(x) := f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2}(x - x_0)^2 + \dots + \frac{f^{(n)}(x_0)}{n!}(x - x_0)^n$$

is called the *Taylor polynomial* of degree  $n$  centered around  $x_0$ . □

The Taylor polynomial  $T_n$  gives a good approximation of  $f$  around  $x_0$ , i.e. when  $x \approx x_0$ ,

$$f(x) \approx T_n(x).$$

The simplest approximations are: the *linear approximation* of  $f$  around  $x_0$  given by  $T_1$ , i.e.

$$f(x) \approx f(x_0) + f'(x_0)(x - x_0)$$

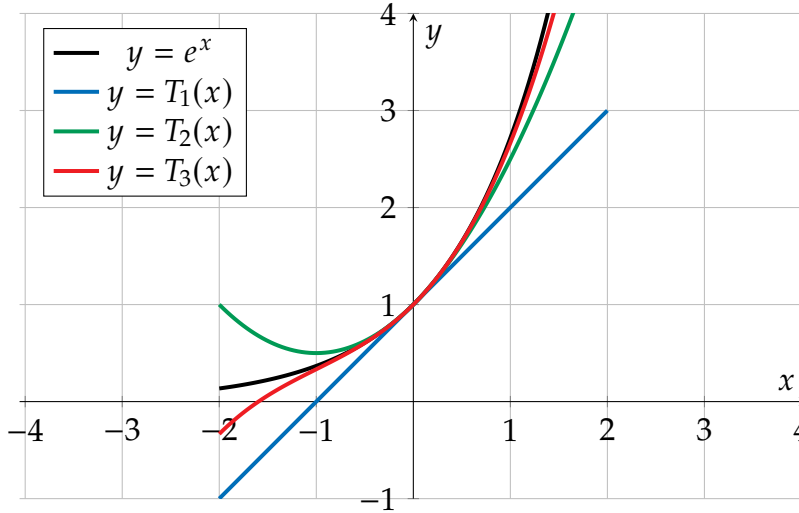
and the *quadratic approximation* of  $f$  around  $x_0$  given by  $T_2$ , i.e.

$$f(x) \approx f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2}(x - x_0)^2.$$

The closer  $x$  is to  $x_0$  and the higher the degree of  $T_n$  is, the better  $T_n(x)$  approximates  $f(x)$ .

**Example 5.2.** Let  $f : \mathbb{R} \rightarrow \mathbb{R}$ ,  $f(x) = e^x$  and  $x_0 = 0$ . Then  $f(0) = f'(0) = \dots = f^{(n)}(0) = 1$  and

$$T_n(x) = 1 + x + \frac{x^2}{2} + \dots + \frac{x^n}{n!}.$$



**Definition 5.3.** Let  $I \subseteq \mathbb{R}$  be an interval,  $f : I \rightarrow \mathbb{R}$  and  $x_0 \in I$  where  $f$  is differentiable  $n$  times. We define  $R_n : \mathbb{R} \rightarrow \mathbb{R}$  to be the remainder when approximating  $f$  by  $T_n$ ,

$$R_n(x) := f(x) - T_n(x).$$

**Theorem 5.4** (Taylor-Lagrange). Let  $I \subseteq \mathbb{R}$  be an interval,  $f : I \rightarrow \mathbb{R}$  differentiable  $n + 1$  times. Then for any  $x, x_0 \in I$ , there exists  $c \in (x_0, x)$  or  $c \in (x, x_0)$  such that

$$R_n(x) = \frac{f^{(n+1)}(c)}{(n+1)!} (x - x_0)^{n+1},$$

called *the remainder in Lagrange's form*. Taylor's formula with Lagrange remainder is

$$f(x) = T_n(x) + \frac{f^{(n+1)}(c)}{(n+1)!} (x - x_0)^{n+1}.$$

**Remark 5.5.** There exist other forms of the remainder, but we will only use this one. Its main advantage is that assuming that all the derivatives of  $f$  are bounded by  $M > 0$ ,

$$|f(x) - T_n(x)| = |R_n(x)| \leq \frac{M}{(n+1)!} |x - x_0|^{n+1} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

**Definition 5.6.** Let  $I \subseteq \mathbb{R}$  be an interval and let  $f : I \rightarrow \mathbb{R}$  be infinitely differentiable. For  $x_0 \in I$  and  $x \in \mathbb{R}$ , the series

$$\sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} (x - x_0)^n$$

is called the *Taylor series* of  $f$  around  $x_0$ . If the series converges and

$$f(x) = \sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} (x - x_0)^n,$$

we say that  $f$  can be expanded in a Taylor series around  $x_0$  (also called Taylor expansion).

**Remark 5.7.** The partial sum of a Taylor series is the Taylor polynomial  $T_n(x)$ . A Taylor series converges to  $f(x)$  if and only if the remainder  $f(x) - T_n(x) = R_n(x) \rightarrow 0$  as  $n \rightarrow \infty$ .

**Remark 5.8.** The Taylor series around 0 is called the *MacLaurin series*,

$$\sum_{n=0}^{\infty} \frac{f^{(n)}(0)}{n!} x^n.$$

**Example 5.9.** Let  $f : \mathbb{R} \rightarrow \mathbb{R}$ ,  $f(x) = e^x$  and  $x_0 = 0$ . Then

$$T_n(x) = 1 + x + \frac{x^2}{2} + \dots + \frac{x^n}{n!}.$$

Consider Taylor's formula  $f(x) = T_n(x) + R_n(x)$  with the Lagrange remainder, for which there exists  $c$  in between 0 and  $x$  such that

$$R_n(x) = \frac{e^c}{(n+1)!} x^{n+1} \rightarrow 0$$

since  $\frac{|x|^n}{n!} \rightarrow 0$  as  $n \rightarrow \infty$ . It follows that  $e^x$  can be expanded as a Taylor series around 0:

$$e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!} = 1 + x + \frac{x^2}{2} + \dots + \frac{x^n}{n!} + \dots, \forall x \in \mathbb{R}.$$

**Example 5.10.** The functions  $\sin$  and  $\cos$  can be expanded in a Taylor series around 0.

$$\sin x = \sum_{n=0}^{\infty} \frac{(-1)^n}{(2n+1)!} x^{2n+1} = x - \frac{x^3}{3!} + \frac{x^5}{5!} - \dots$$

and

$$\cos x = \sum_{n=0}^{\infty} \frac{(-1)^n}{(2n)!} x^{2n} = 1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \dots$$

**Example 5.11.** The function  $f : \mathbb{R} \rightarrow \mathbb{R}$ ,

$$f(x) = \begin{cases} e^{-1/x^2}, & \text{if } x \neq 0 \\ 0, & \text{if } x = 0 \end{cases}$$

is infinitely differentiable at 0, but  $f$  is not expandable in a Taylor series around 0.



**Example 5.12** (Convex/concave). Let  $f : I \rightarrow \mathbb{R}$  be two times differentiable, with a critical point at  $x_0$ , i.e.  $f'(x_0) = 0$ . Then from Taylor's formula we have that

$$f(x) = f(x_0) + \frac{f''(x_0)}{2}(x - x_0)^2 + R_2(x).$$

When  $x$  is very close to  $x_0$ , the quadratic approximation is very accurate and the remainder  $R_2(x)$  is very small. Thus the behaviour of  $f(x)$  around  $x_0$  is dictated by the quadratic term  $f''(x_0)(x - x_0)^2$  and we see that:

- If  $f''(x_0) > 0$  (convexity), then  $f(x) > f(x_0)$  and  $x_0$  is a local minimum.
- If  $f''(x_0) < 0$  (concavity), then  $f(x) < f(x_0)$  and  $x_0$  is a local maximum.

**Theorem 5.13** (Local optimality conditions). Let  $I \subseteq \mathbb{R}$  be an open interval,  $f : I \rightarrow \mathbb{R}$  and  $x_0 \in I$  a point where  $f$  is differentiable  $n$  times and

$$f'(x_0) = f''(x_0) = \dots = f^{(n-1)}(x_0) = 0 \text{ and } f^{(n)}(x_0) \neq 0.$$

1. If  $n$  is even and  $f^{(n)}(x_0) > 0$ , then  $x_0$  is a *local minimum* of  $f$ .
2. If  $n$  is even and  $f^{(n)}(x_0) < 0$ , then  $x_0$  is a *local maximum* of  $f$ .
3. If  $n$  is odd, then  $x_0$  is not a local extremum point of  $f$ .

*Proof.* It follows from the Taylor approximation  $f(x) = f(x_0) + \frac{f^{(n)}(x_0)}{n!}(x - x_0)^n + R_n(x)$ .  $\square$

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### Exercises. Seminar 6

1. Recall the Taylor series for  $\sin$  and  $\cos$ :

$$\begin{aligned} \sin x &= \sum_{n=0}^{\infty} \frac{(-1)^n}{(2n+1)!} x^{2n+1} = x - \frac{x^3}{3!} + \frac{x^5}{5!} - \dots, \quad \forall x \in \mathbb{R}, \\ \cos x &= \sum_{n=0}^{\infty} \frac{(-1)^n}{(2n)!} x^{2n} = 1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \dots, \quad \forall x \in \mathbb{R}. \end{aligned}$$

- (a) Prove that  $(\sin x)' = \cos x$  and  $(\cos x)' = -\sin x$ ,  $\forall x \in \mathbb{R}$ .
- (b) Deduce that

$$x - \frac{x^3}{6} < \sin x < x, \quad \forall x > 0$$

and

$$1 - \frac{x^2}{2} < \cos x < 1 - \frac{x^2}{2} + \frac{x^4}{24}, \quad \forall x \in \mathbb{R}.$$

- (c) Prove formally Euler's formula  $e^{ix} = \cos x + i \sin x$ .

2. (a) For  $\alpha \in \mathbb{R}$  and  $|x| < 1$ , prove the generalized binomial expansion (binomial series)

$$(1+x)^\alpha = \sum_{n=0}^{\infty} \binom{\alpha}{n} x^n, \quad \binom{\alpha}{k} := \frac{\alpha(\alpha-1)\dots(\alpha-k+1)}{k!}, \quad \binom{\alpha}{0} = 1.$$

- (b) Find the first four terms in the binomial series of  $\sqrt{1+x}$  and  $1/\sqrt{1+x}$ .

3. Find the MacLaurin series and its radius of convergence for the following functions:

(a)  $a^x, a > 0$ .

(c)  $\sin^2(x)$ .

(b)  $(1+x)\ln(1+x)$ .

(d)  $\arctan x$ .

4. For each function  $f: \mathbb{R} \rightarrow \mathbb{R}$  given below check that  $f'(0) = 0$  and find the first  $n \in \mathbb{N}$  such that  $f^{(n)}(0) \neq 0$ . Then, deduce whether 0 is a local extremum point of  $f$  or not; in the affirmative, specify if 0 is a global extremum point or just a local one.

(a)  $f(x) = e^x + e^{-x} - x^2$ .

(b)  $f(x) = \cos(x^2)$ .

(c)  $f(x) = 6 \sin x - 6x + x^3$ .

### Exercises. Homework 6

1. Using Taylor series, compute the following limits:

(a)  $\lim_{x \rightarrow 0} \frac{\sin x - x + x^3/6}{x^5}$ .

(b)  $\lim_{x \rightarrow 0} \frac{e^{x^2} - \cos x - 3x^2/2}{x^4}$ .

2. Prove that the Taylor series of  $\ln(1+x)$  around 0 is

$$\ln(1+x) = \sum_{n=1}^{\infty} (-1)^{n+1} \frac{x^n}{n}.$$

3. Using Taylor series, prove that the forward difference  $(f(x+h) - f(x))/h$  approximates the derivative  $f'(x)$  with an error of order  $h$  (first order approximation), i.e.

$$f'(x) = \frac{f(x+h) - f(x)}{h} + O(h),$$

and that the centered difference  $(f(x+h) - f(x-h))/2h$  approximates the derivative  $f'(x)$  with an error of order  $h^2$  (second order approximation), i.e.

$$f'(x) = \frac{f(x+h) - f(x-h)}{2h} + O(h^2).$$

Here the Big Oh notation  $e_1(h) = O(e_2(h))$  means that  $\lim_{h \rightarrow 0} \frac{e_1(h)}{e_2(h)} \in (0, \infty)$ .

Python:

Choose a particular function  $f$ , a point  $x$  and compute exactly  $f'(x)$ . By taking a range of small values  $h$ , show that the errors when approximating  $f'(x)$  with the finite differences above are proportional to  $h$  and  $h^2$ , respectively.

Explain why after  $h$  becomes small enough the errors start to increase.

## ※ Riemann integrals. Improper integrals

Let  $[a, b]$  be a compact interval and let  $f : [a, b] \rightarrow \mathbb{R}$ . The points  $a = x_0 < x_1 < \dots < x_n = b$  define a partition of the interval  $[a, b]$

$$\mathcal{P} = \{[x_{k-1}, x_k] \mid k = \overline{1, n}\},$$

whose norm is given by

$$\|\mathcal{P}\| = \max_{k=\overline{1, n}} \{x_k - x_{k-1}\}.$$

Consider also a set of intermediate points  $c_k \in [x_k, x_{k-1}]$  attached to the partition  $\mathcal{P}$ .

**Definition 6.1.** For  $f : [a, b] \rightarrow \mathbb{R}$  and a partition  $\mathcal{P}$  of  $[a, b]$ , the Riemann sum is given by

$$\sigma(f, \mathcal{P}) := \sum_{k=1}^n f(c_k)(x_k - x_{k-1}).$$

**Remark 6.2.** The Riemann sum collects the areas of the rectangles defined by the partition  $\mathcal{P}$  (and the intermediate points). In the limit one obtains the area below the curve  $y = f(x)$ .

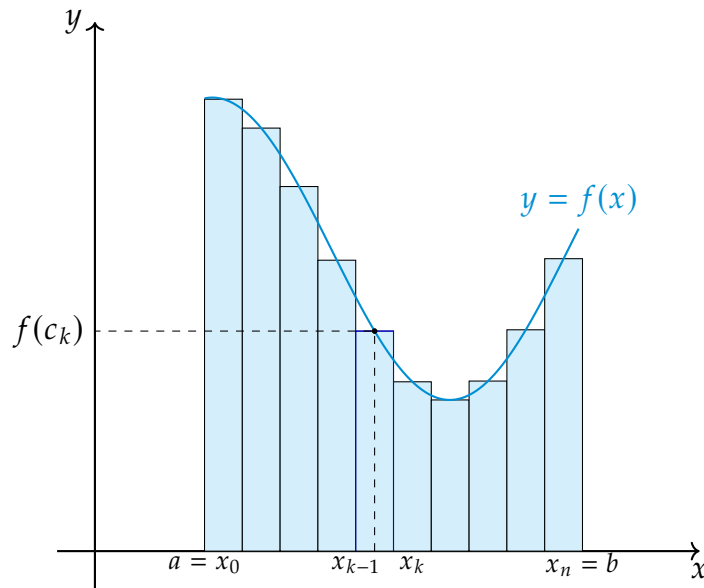


Figure 1: Area under a curve approximated through rectangles. Riemann sum.

**Definition 6.3.** We say that  $f : [a, b] \rightarrow \mathbb{R}$  is *Riemann integrable* if there exists  $I \in \mathbb{R}$  s.t. for any partition  $\mathcal{P}$  of  $[a, b]$  the Riemann sum  $\sigma(f, \mathcal{P})$  converges to  $I$  as  $\|\mathcal{P}\| \rightarrow 0$ , i.e.

$$\lim_{\|\mathcal{P}\| \rightarrow 0} \sigma(f, \mathcal{P}) = I =: \int_a^b f(x) dx.$$

**Proposition 6.4.** Let  $f, g : [a, b] \rightarrow \mathbb{R}$  be Riemann integrable and  $\alpha \in \mathbb{R}$ . Then

- $\int_a^b \alpha f(x) dx = \alpha \int_a^b f(x) dx$ .
- $f + g$  is Riemann integrable and  $\int_a^b (f(x) + g(x)) dx = \int_a^b f(x) dx + \int_a^b g(x) dx$ .
- If  $f \leq g$ , then  $\int_a^b f(x) dx \leq \int_a^b g(x) dx$ .

**Proposition 6.5.** Let  $f : [a, b] \rightarrow \mathbb{R}$  be Riemann integrable and  $c \in (a, b)$ . Then

$$\int_a^b f(x) dx = \int_a^c f(x) dx + \int_c^b f(x) dx.$$

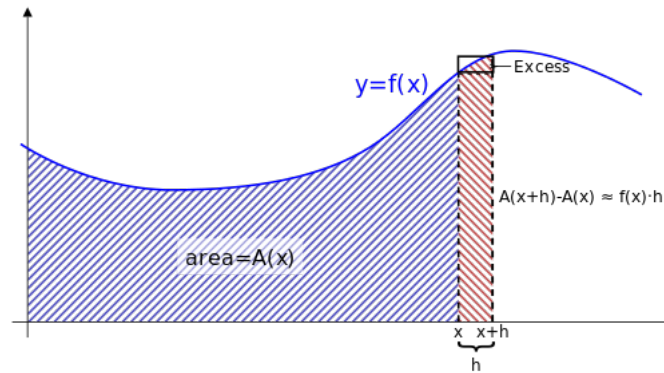


Figure 2: The derivative of the area function  $A$  is  $f$ . Source: wikipedia.

**Theorem 6.6.** Let  $f : [a, b] \rightarrow \mathbb{R}$  be Riemann integrable. Then the function  $A : [a, b] \rightarrow \mathbb{R}$ ,  $A(x) := \int_a^x f(t) dt$  is continuous. Furthermore, if  $f$  is continuous, then  $A$  is differentiable and  $A'(x) = f(x)$ .

**Theorem 6.7** (Fundamental theorem of calculus). Let  $f : [a, b] \rightarrow \mathbb{R}$  be Riemann integrable and  $F : [a, b] \rightarrow \mathbb{R}$  an antiderivative (primitive) of  $f$ , i.e.  $F' = f$ , then

$$\int_a^b f(x) dx = F(b) - F(a).$$

**Definition 6.8** (Trapezium rule). Let  $f : [a, b] \rightarrow \mathbb{R}$  be Riemann integrable and consider  $a = x_0 < x_1 < \dots < x_n = b$ . The area below the curve  $y = f(x)$  can be approximated by

$$\sum_{k=1}^n \frac{f(x_{k-1}) + f(x_k)}{2} (x_k - x_{k-1}).$$

Note that  $\frac{f(x_{k-1})+f(x_k)}{2}(x_k-x_{k-1})$  is the area of the trapezium determined by  $x_{k-1}, x_k, f(x_{k-1}), f(x_k)$ . In the case of a uniform partition with  $x_k - x_{k-1} = \frac{b-a}{n}, \forall k \in \overline{1, n}$ , we have that

$$\int_a^b f(x) dx \approx \frac{b-a}{n} \left( \frac{1}{2}f(a) + \sum_{i=1}^{n-1} f(x_i) + \frac{1}{2}f(b) \right).$$

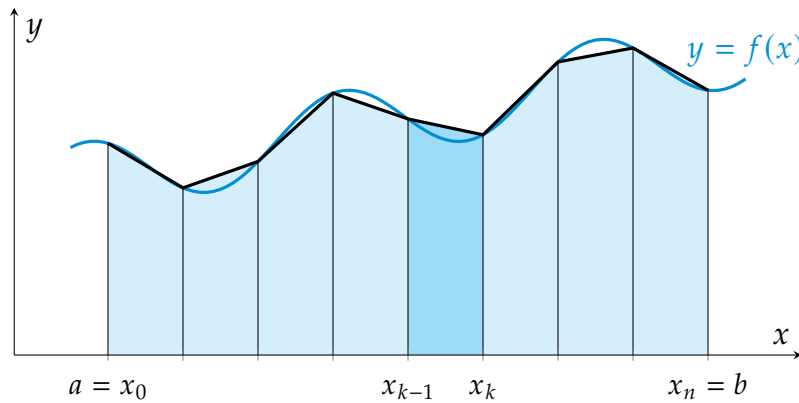


Figure 3: Trapezium rule.

**Definition 6.9.** Let  $a, b \in \mathbb{R}$ . If the following limits exist, we define the *improper integrals*

- If  $f : [a, \infty) \rightarrow \mathbb{R}$  is Riemann integrable on any compact interval in the domain,

$$\int_a^\infty f(x) dx := \lim_{t \rightarrow \infty} \int_a^t f(x) dx.$$

- If  $f : [a, b) \rightarrow \mathbb{R}$  is Riemann integrable on any compact interval included in the domain,

$$\int_a^{b-0} f(x) dx := \lim_{t \nearrow b} \int_a^t f(x) dx.$$

- If  $f : (a, b] \rightarrow \mathbb{R}$  is Riemann integrable on any compact interval included in the domain,

$$\int_{a+0}^b f(x) dx := \lim_{t \searrow a} \int_t^b f(x) dx.$$

The notation  $\int_a^{b-0} \dots, \int_{a+0}^b \dots$  emphasizes that the integrals are improper, but we can also simply write  $\int_a^b \dots$  even when dealing with an improper integral.

**Definition 6.10.** We say that an improper integral is convergent if it is finite (finite limit).

Note that an improper integral represents the area of an infinite region.

**Example 6.11.** Let  $a > 0$  and  $p \in \mathbb{R}$ . The improper integral

$$\int_a^\infty \frac{1}{x^p} dx$$

converges when  $p > 1$  and diverges when  $p \leq 1$ . Indeed, for  $p = 1$  the integral diverges ( $\ln(\infty)$ ) and for  $p \neq 1$ ,

$$\int_a^\infty \frac{1}{x^p} dx = \lim_{t \rightarrow \infty} \int_a^t f(x) dx = \lim_{t \rightarrow \infty} \frac{t^{-p+1}}{-p+1} - \frac{a^{-p+1}}{-p+1},$$

which converges when  $-p + 1 < 0$ , i.e.  $p > 1$ , and diverges when  $p < 1$ .

**Example 6.12.** Let  $0 < a < b$  and  $p \in \mathbb{R}$ . The improper integrals

$$\int_a^b \frac{1}{(b-x)^p} dx, \int_a^b \frac{1}{(x-a)^p} dx$$

converge when  $p < 1$  and diverge when  $p \geq 1$ . Indeed, for  $p = 1$  the integrals diverge ( $\ln(0)$ ) and for  $p \neq 1$  the first integral, for example, is

$$\int_a^b \frac{1}{(b-x)^p} dx = -\lim_{t \nearrow b} \frac{(b-t)^{-p+1}}{-p+1} + \frac{(b-a)^{-p+1}}{-p+1},$$

which converges when  $-p + 1 > 0$ , i.e.  $p < 1$ , and diverges when  $p > 1$ .

**Theorem 6.13.** Let  $a < b \leq \infty$  and  $f, g : [a, b) \rightarrow [0, \infty)$ . If there exists  $c \in (a, b)$  s.t.  $f(x) \leq g(x), \forall x \geq c$ , then

- If  $\int_a^b g(x) dx$  converges, then  $\int_a^b f(x) dx$  converges.
- If  $\int_a^b f(x) dx$  diverges, then  $\int_a^b g(x) dx$  diverges.
- If  $\lim_{x \nearrow b} \frac{f(x)}{g(x)} \in (0, \infty)$ , then  $\int_a^b f(x) dx$  and  $\int_a^b g(x) dx$  have the same nature.

**Theorem 6.14** (Integral test for series). Let  $f : [1, \infty) \rightarrow [0, \infty)$  be decreasing, then

$\int_1^\infty f(x) dx$  and  $\sum_{n=1}^\infty f(n)$  have the same nature.

*Proof.* Let  $N \in \mathbb{N}$  and write  $\int_1^N f(x) dx = \sum_{n=1}^{N-1} \int_n^{n+1} f(x) dx$ . Since  $f$  is decreasing we have that

$$\sum_{n=1}^{N-1} f(n+1) \leq \int_1^N f(x) dx \leq \sum_{n=1}^{N-1} f(n).$$

The conclusion follows by letting  $N \rightarrow \infty$  and using the comparison test.  $\square$

### Exercises. Seminar 7

1. Compute the following limits using a Riemann sum:

$$(a) \lim_{n \rightarrow \infty} \sum_{k=1}^n \frac{1}{n+k}. \quad (b) \lim_{n \rightarrow \infty} \sum_{k=1}^n \frac{k^p}{n^{p+1}}, p \in \mathbb{N}. \quad (c) \lim_{n \rightarrow \infty} \frac{\sqrt[n]{n!}}{n}.$$

2. Study the Riemann integrability of the function  $f : [0, 1] \rightarrow \mathbb{R}$ ,

$$f(x) = \begin{cases} 1, & \text{if } x \in \mathbb{Q}, \\ 0, & \text{if } x \in \mathbb{R} \setminus \mathbb{Q}. \end{cases}$$

3. Compute the following improper integrals:

$$(a) \int_1^2 \frac{1}{x(x-2)} dx. \quad (c) \int_0^1 \frac{\ln x}{\sqrt{x}} dx. \\ (b) \int_0^\infty x e^{-x^2} dx. \quad (d) \int_0^\infty e^{-x} \sin x dx.$$

4. Study the convergence of the following improper integrals:

$$(a) \int_1^\infty \frac{1}{x\sqrt{1+x^2}} dx. \quad (b) \int_0^{\frac{\pi}{2}} \frac{1}{\cos x} dx. \quad (c) \int_1^\infty \frac{\ln x}{x\sqrt{x^2-1}} dx.$$

5. Using the integral test, study the convergence of the following series:

$$(a) \sum_{n \geq 2} \frac{1}{n(\ln n)^2}. \quad (b) \sum_{n \geq 2} \frac{\ln n}{n^2}. \quad (c) \sum_{n \geq 2} \frac{1}{n \ln n \ln(\ln n)}.$$

### Exercises. Homework 7

1. Compute the following limits using a Riemann sum:



$$(a) \lim_{n \rightarrow \infty} \frac{\sqrt[n]{e} + 2\sqrt[n]{e^2} + \dots + n\sqrt[n]{e^n}}{n^2}.$$

$$(b) \lim_{n \rightarrow \infty} \sqrt[n]{\sin \frac{\pi}{2n} \sin \frac{2\pi}{2n} \dots \sin \frac{(n-1)\pi}{2n}}.$$

2. Let  $\Gamma(\alpha) = \int_0^\infty x^{\alpha-1} e^{-x} dx$ , for  $\alpha > 0$ . Prove that

$$\Gamma(\alpha + 1) = \alpha \Gamma(\alpha),$$

and that

$$\Gamma(n) = (n-1)!, \quad n \in \mathbb{N}^*.$$

3. Python: The integral  $\int_{-\infty}^\infty e^{-x^2} dx$  represents the area under the bell curve  $y = e^{-x^2}$  and it is related to the normal (Gaussian) probability distribution. It is essential in probability theory and has a wide range of applications. Considering intervals of the form  $[-a, a]$ , for increasing values of  $a > 0$ , show numerically (e.g. trapezium rule) that  $\int_{-\infty}^\infty e^{-x^2} dx = \sqrt{\pi}$ .

## ※ The Euclidean space $\mathbb{R}^n$

Elements in  $\mathbb{R}^n$  are vectors with  $n$  components. We will write  $x = (x_1, \dots, x_n) \in \mathbb{R}^n$  most of the time, apart from situations where matrices will also be involved – in this case we

will adopt the linear algebra notation of writing  $x \in \mathbb{R}^n$  as a column vector  $x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}_{n \times 1}$ , since this allows to multiply matrices  $\begin{bmatrix} & \end{bmatrix}_{m \times n}$  with vectors  $\begin{bmatrix} \end{bmatrix}_{n \times 1}$  to get vectors  $\begin{bmatrix} \end{bmatrix}_{m \times 1}$ .

As you've seen in your Algebra course,  $\mathbb{R}^n$  is a vector space: two vectors  $x, y \in \mathbb{R}^n$  can be added component wise  $x + y := (x_1 + y_1, \dots, x_n + y_n)$ , and a vector can be multiplied by a scalar  $\alpha \in \mathbb{R}$  to get  $\alpha x := (\alpha x_1, \dots, \alpha x_n)$ . We will denote by  $e_i$  the canonical basis vector with a 1 in the  $i$ th component and 0's everywhere else, giving  $x = x_1 e_1 + \dots + x_n e_n$ .

**Definition 7.1.** A map  $\langle \cdot, \cdot \rangle : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$  is called a *scalar product* (or *inner product*) if

- (a)  $\langle x, y \rangle = \langle y, x \rangle$ , for any  $x, y \in \mathbb{R}^n$ .
- (b)  $\langle \alpha x + \beta y, z \rangle = \alpha \langle x, z \rangle + \beta \langle y, z \rangle$ , for any  $x, y, z \in \mathbb{R}^n$  and  $\alpha, \beta \in \mathbb{R}$ .
- (c)  $\langle x, x \rangle > 0$ , for any  $x \in \mathbb{R}^n \setminus \{0\}$ .

**Definition 7.2.** The *dot product* of two vectors  $x, y \in \mathbb{R}^n$  is given by

$$x \cdot y := x_1 y_1 + \dots + x_n y_n.$$

The dot product is the most important scalar product. In matrix notation, it is written as

$$x \cdot y = x^T y = [x_1 \dots x_n]_{1 \times n} \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix}_{n \times 1} = x_1 y_1 + \dots + x_n y_n.$$

**Definition 7.3.** Two vectors  $x, y \in \mathbb{R}^n$  are perpendicular (or orthogonal) iff  $x \cdot y = 0$ .

**Example 7.4.** A symmetric positive definite matrix  $M \in \mathbb{R}^{n \times n}$  defines a scalar product  $\langle x, y \rangle = x^T M y$ .

**Definition 7.5.** A function  $\| \cdot \| : \mathbb{R}^n \rightarrow [0, \infty)$  is called a *norm* if

- (a)  $\|x\| = 0$  if and only if  $x = 0$ .
- (b)  $\|\alpha x\| = |\alpha| \|x\|$ , for any  $x \in \mathbb{R}^n$  and  $\alpha \in \mathbb{R}$ .
- (c)  $\|x + y\| \leq \|x\| + \|y\|$ , for any  $x, y \in \mathbb{R}^n$  (triangle inequality).

**Proposition 7.6.** Any scalar product generates a norm on  $\mathbb{R}^n$  given by  $\|x\| = \sqrt{\langle x, x \rangle}$ .

**Theorem 7.7** (Cauchy-Schwarz inequality). For any  $x, y \in \mathbb{R}^n$  it holds that

$$|\langle x, y \rangle| \leq \|x\| \|y\|.$$

Here the norm  $\|\cdot\|$  is generated by the scalar product  $\langle \cdot, \cdot \rangle$ .

*Proof.* Consider  $f : \mathbb{R} \rightarrow \mathbb{R}$ ,  $f(t) = \langle tx + y, tx + y \rangle = \|tx + y\|^2 \geq 0$ . Since  $f(t) = t^2\|x\|^2 + 2t\langle x, y \rangle + \|y\|^2$  is quadratic in  $t$ , we must have that  $\Delta = 4\langle x, y \rangle^2 - 4\|x\|^2\|y\|^2 \leq 0$   $\square$

**Definition 7.8.** The Euclidean norm is generated by the dot product and it is given by

$$\|x\| = \sqrt{x \cdot x} = \sqrt{x_1^2 + \dots + x_n^2}.$$

This represents the length of the vector  $x \in \mathbb{R}^n$  measured using the Euclidean norm.

**Theorem 7.9.** For  $n \in \{2, 3\}$  the dot product of  $x, y \in \mathbb{R}^n$  is

$$x \cdot y = \|x\| \|y\| \cos \angle(x, y).$$

*Proof.* Consider the triangle with sides determined by the vectors  $x, y$  and  $x - y$ . From the cosine rule we have that

$$\|x - y\|^2 = \|x\|^2 + \|y\|^2 - 2\|x\| \|y\| \cos \angle(x, y).$$

Since  $\|x - y\|^2 = (x - y) \cdot (x - y) = x \cdot x + y \cdot y - 2x \cdot y$ , we obtain that  $x \cdot y = \|x\| \|y\| \cos \angle(x, y)$ .  $\square$

**Example 7.10.** (a)  $\|x\|_1 := |x_1| + \dots + |x_n|$  is a norm (so-called Manhattan norm).

(b)  $\|x\|_p := (|x_1|^p + \dots + |x_n|^p)^{\frac{1}{p}}$ ,  $p > 1$ , is a norm.

(c)  $\|x\|_\infty := \max\{|x_1|, \dots, |x_n|\}$  is a norm.

**Definition 7.11.** A function  $d : \mathbb{R}^n \times \mathbb{R}^n \rightarrow [0, \infty)$  is called a *distance* (or *metric*) if

(a)  $d(x, y) = 0$  if and only if  $x = y$ .

(b)  $d(x, y) = d(y, x)$ , for any  $x, y \in \mathbb{R}^n$ .

(c)  $d(x, z) \leq d(x, y) + d(y, z)$ , for any  $x, y, z \in \mathbb{R}^n$  (triangle inequality).

**Proposition 7.12.** Any norm generates a distance on  $\mathbb{R}^n$  given by  $d(x, y) = \|x - y\|$ .

**Definition 7.13.** The Euclidean distance is generated by the Euclidean norm and it is given by

$$d(x, y) = \|x - y\| = \sqrt{(x_1 - y_1)^2 + \dots + (x_n - y_n)^2}.$$

We will be using the Euclidean norm and distance, unless we specify otherwise.

**Neighborhoods. Interior. Closure. Boundary.**

**Definition 7.14.** A set  $A \subseteq \mathbb{R}^n$  is called *bounded* if there exists  $r > 0$  such that

$$\|x\| \leq r, \forall x \in A.$$

**Definition 7.15.** Let  $x_0 \in \mathbb{R}^n$  and  $r > 0$ . The open ball of centre  $x_0$  and radius  $r$  is given by

$$B(x_0, r) := \{x \in \mathbb{R}^n \mid \|x - x_0\| < r\},$$

and the closed ball of centre  $x_0$  and radius  $r$  is given by

$$\bar{B}(x_0, r) := \{x \in \mathbb{R}^n \mid \|x - x_0\| \leq r\}.$$

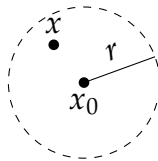


Figure 4: Open ball  $B(x_0, r)$ .

**Definition 7.16.** A set  $V \subseteq \mathbb{R}^n$  is a *neighborhood (vecinity)* of  $x \in \mathbb{R}^n$  if

$$\exists r > 0 \text{ such that } B(x, r) \subseteq V.$$

We denote all the neighborhoods of  $x$  by  $\mathcal{V}(x) := \{V \subseteq \mathbb{R}^n \mid V \text{ is a neighborhood of } x\}$ .

**Definition 7.17.** Let  $A \subseteq \mathbb{R}^n$ . The following set is called the *interior* of  $A$

$$\text{int}(A) := \{x \in \mathbb{R}^n \mid \exists V \in \mathcal{V}(x) \text{ such that } V \subseteq A\},$$

the following set is called the *closure* of  $A$

$$\text{cl}(A) := \{x \in \mathbb{R}^n \mid \forall V \in \mathcal{V}(x), V \cap A \neq \emptyset\},$$

and the following set is called the *boundary* of  $A$

$$\text{bd}(A) := \{x \in \mathbb{R}^n \mid \forall V \in \mathcal{V}(x), V \cap A \neq \emptyset \text{ and } V \cap A^c \neq \emptyset\}.$$

**Example 7.18.** Let  $A = \{(x, y) \in \mathbb{R}^2 \mid x^2 + y^2 < 1\}$ . Then

$$\text{int}(A) = A,$$

$$\text{cl}(A) = \{(x, y) \in \mathbb{R}^2 \mid x^2 + y^2 \leq 1\},$$

$$\text{bd}(A) = \{(x, y) \in \mathbb{R}^2 \mid x^2 + y^2 = 1\}.$$

**Proposition 7.19.** For any  $A \subseteq \mathbb{R}^n$ , it holds that  $\text{cl}(A) = A \cup \text{bd}(A)$ .

**Definition 7.20.** If  $A = \text{int}(A)$ , then  $A$  is called *open*. If  $A = \text{cl}(A)$ , then  $A$  is called *closed*.

**Proposition 7.21.** For any  $A \subseteq \mathbb{R}^n$ , it holds that  $\text{int}(A) \subseteq A \subseteq \text{cl}(A)$ .

*Proof.* Similar to [Theorem 1.12](#). □

**Remark 7.22.** To prove that a set  $A$  is open, it is sufficient to prove that  $A \subseteq \text{int}(A)$ .

To prove that a set  $A$  is closed, it is sufficient to prove that  $\text{cl}(A) \subseteq A$ .

**Proposition 7.23.** The following statements are true:

- The complement of an open set is closed.
- The complement of a closed set is open.

*Proof.* Similar to [Theorem 1.15](#). □

A sequence  $(x^k)$  in  $\mathbb{R}^n$  indexed by  $k \in \mathbb{N}$  has vector elements  $x^1, x^2, \dots, x^k, \dots$ . Notice that the index  $k$  appears as superscript (in order to avoid confusion with the coordinates of the vectors).

**Definition 7.24.** A sequence  $(x^k)$  converges to  $x \in \mathbb{R}^n$  if  $\lim_{k \rightarrow \infty} \|x^k - x\| = 0$ . We write  $\lim_{k \rightarrow \infty} x^k = x$ .

**Example 7.25.** Let  $x^k = (\frac{1}{k}, \frac{k}{k+1})$ , then  $\lim_{k \rightarrow \infty} x^k = (0, 1)$ .

**Theorem 7.26.** A sequence  $(x^k)$  converges to  $x \in \mathbb{R}^n$  if and only if  $\lim_{k \rightarrow \infty} x_i^k = x_i, \forall i = \overline{1, n}$ .

*Proof.* Consider first  $i \in \{1, \dots, n\}$ . We have that

$$|x_i^k - x_i| = \sqrt{(x_i^k - x_i)^2} \leq \sqrt{(x_1^k - x_1)^2 + \dots + (x_n^k - x_n)^2} = \|x^k - x\|,$$

hence if  $(x^k)$  converges to  $x \in \mathbb{R}^n$ , i.e.  $\|x^k - x\| \rightarrow 0$ , then  $|x_i^k - x_i| \rightarrow 0$  and  $x_i^k \rightarrow x_i$ .

Let us now prove the converse statement and assume that  $\lim_{k \rightarrow \infty} x_i^k = x_i, \forall i = \overline{1, n}$ . Then

$$\|x^k - x\| = \sqrt{(x_1^k - x_1)^2 + \dots + (x_n^k - x_n)^2} \rightarrow 0,$$

hence  $x^k \rightarrow x$ . □

Note that this is telling us that a sequence of vectors converges if and only if the components of the vectors converge, respectively.

**Exercises. Seminar 8**

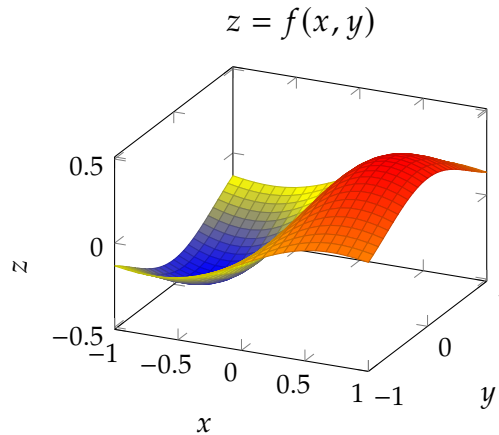
1. Prove that for any  $x, y \in \mathbb{R}^n$  the following identities hold:
  - (a)  $\|x + y\|^2 + \|x - y\|^2 = 2(\|x\|^2 + \|y\|^2)$  (the parallelogram identity).
  - (b)  $\langle x, y \rangle = \frac{1}{4}(\|x + y\|^2 - \|x - y\|^2)$ .
2. Find the orthogonal projection of a vector  $v \in \mathbb{R}^2$  onto a vector  $a \in \mathbb{R}^2$ .
3. Show that  $\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$  is a rotation matrix with angle  $\theta$  in  $\mathbb{R}^2$ .
4. Consider the  $p$ -norm  $\|x\|_p := (|x_1|^p + \dots + |x_n|^p)^{\frac{1}{p}}$ ,  $p \geq 1$  and  $\|x\|_\infty := \max\{|x_1|, \dots, |x_n|\}$ . Draw the unit ball in  $\mathbb{R}^2$  for the  $p$ -norm with  $p \in \{1, 2, \infty\}$ .
5. Find the interior, the closure and the boundary for each of the following sets:
  - (a)  $[0, 1) \times (1, 2]$ .
  - (b)  $\{(x, y) \in \mathbb{R}^2 \mid |x| < |y|\}$ .
  - (c)  $\{(x, y) \in \mathbb{R}^2 \mid |x| + |y| < 1\}$ .
  - (d)  $\{(x, y) \in \mathbb{R}^2 \mid (x-1)^2 + y^2 \leq 1, x \leq 1\}$ .
6. Draw the level sets  $L_c = \{(x, y) \in \mathbb{R}^2 \mid f(x, y) = c\}$  and sketch the graph for the following:
  - (a)  $f(x, y) = x + y$  and  $c \in \{0, \pm 1\}$ .
  - (b)  $f(x, y) = 2x^2 + y^2$  and  $c \in \{0, 1, 4\}$ .
  - (c)  $f(x, y) = \sqrt{x^2 + y^2}$  and  $c \in \{0, 1, 2\}$ .
  - (d)  $f(x, y) = x^2 - y^2$  and  $c \in \{0, \pm 1\}$ .

**Exercises. Homework 8**

1. For  $x, y \in \mathbb{R}^n$  prove that the following statements are equivalent:
  - (a)  $\langle x, y \rangle = 0$ .
  - (b)  $\|x + y\| = \|x - y\|$ .
  - (c)  $\|x + y\|^2 = \|x\|^2 + \|y\|^2$ .
2. Python: Represent the unit ball in  $\mathbb{R}^2$  for the  $p$ -norm with  $p \in \{1.25, 1.5, 3, 8\}$  by random sampling and plotting the points that are inside the unit ball.

## ※ Functions of several variables. Limits and continuity

We will now introduce functions of several variables, focusing on those having real (scalar) values. This means we will mostly consider functions  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  mapping vectors in  $\mathbb{R}^n$  into real numbers. As you already know, when  $n = 1$  the graph of a function is a curve in  $\mathbb{R}^2$ . When  $n = 2$ , the graph of a function  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  is given by points with coordinates  $(x, y, f(x, y))$  – this represents a surface in  $\mathbb{R}^3$  (an example is shown in the figure below).



What about when  $n \geq 3$ ? The graph of the function,  $\{(x, f(x)) \in \mathbb{R}^{n+1} \mid x \in A \subseteq \mathbb{R}^n\}$ , would be a set in  $\mathbb{R}^{n+1}$  and we are able to visualize only its projections in lower dimensional spaces ( $\mathbb{R}^3$  or  $\mathbb{R}^2$ ). Apart from the graph, another way of visualizing a function is through

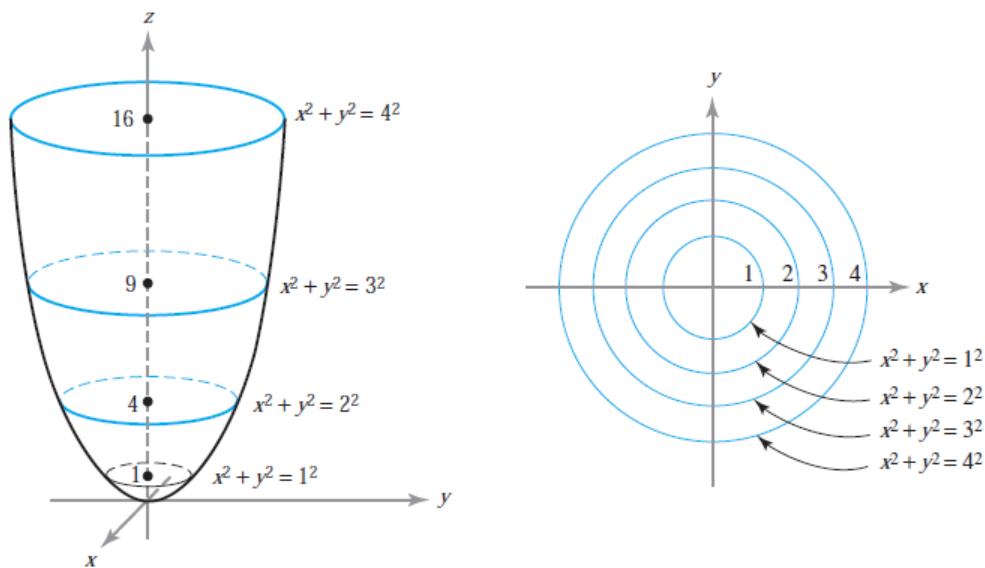


Figure 5: Graph and level curves for  $f(x, y) = x^2 + y^2$ . Source: [4, page 80].

its level sets, which are given by

$$L_c := \{x \in A \subseteq \mathbb{R}^n \mid f(x) = c\},$$

for a constant  $c \in \mathbb{R}$ . If  $n = 2$ , the set  $L_c = \{(x, y) \in A \mid f(x, y) = c\}$  describes a *level curve* (see the figure above). If  $n = 3$ , the set  $L_c = \{(x, y, z) \in A \mid f(x, y, z) = c\}$  describes a *level surface*.

**Limits of functions of several variables. Continuity.**

Using neighbourhoods in  $\mathbb{R}^n$ , we can define limits and continuity exactly as in [Section 4](#).

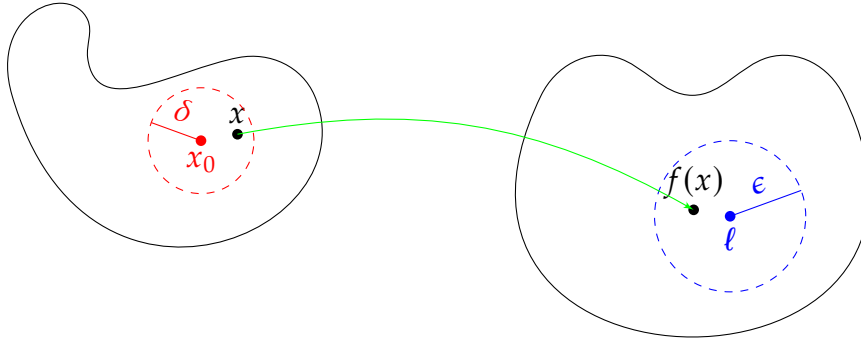
**Definition 8.1.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  and  $x_0 \in A'$ . We say that  $\lim_{x \rightarrow x_0} f(x) = \ell \in \overline{\mathbb{R}}$  if

$$\forall V \in \mathcal{V}(\ell), \exists U \in \mathcal{V}(x_0) \text{ s.t. } f(x) \in V, \forall x \in U \cap (A \setminus \{x_0\}).$$

**Remark 8.2.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  and  $x_0 \in A'$ . We say that  $\lim_{x \rightarrow x_0} f(x) = \ell \in \mathbb{R}$  if

$$\forall \varepsilon > 0, \exists \delta > 0 \text{ s.t. } |f(x) - \ell| < \varepsilon, \forall x \in A \text{ with } \|x - x_0\| < \delta.$$

A similar definition can be given when  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$ , illustrated below. □



**Theorem 8.3.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  and  $x_0 \in A'$ . Then  $\lim_{x \rightarrow x_0} f(x) = \ell \in \overline{\mathbb{R}}$  if and only if

for any sequence  $(x^k)$  in  $A \setminus \{x_0\}$  with  $\lim_{k \rightarrow \infty} x^k = x_0$ , we have that  $\lim_{k \rightarrow \infty} f(x^k) = \ell \in \overline{\mathbb{R}}$ . □

Let us now consider some limits in  $\mathbb{R}^2$  and explore some methods of computing them.

**Example 8.4.** (a)  $\lim_{(x,y) \rightarrow (0,0)} \frac{x^2}{\sqrt{x^2 + y^2}} = 0.$

We will try a simple strategy: to bound the function and use the squeeze theorem.

Since  $0 \leq \frac{x^2}{\sqrt{x^2 + y^2}} \leq \frac{x^2 + y^2}{\sqrt{x^2 + y^2}} = \sqrt{x^2 + y^2}$  and  $\sqrt{x^2 + y^2} \rightarrow 0$ , as  $(x, y) \rightarrow (0, 0)$ , we have that the limit is zero.



$$(b) \lim_{(x,y) \rightarrow (0,0)} \frac{\sin(x^2 + y^2)}{x^2 + y^2} = 1.$$

Here we can use a remarkable limit since  $t := x^2 + y^2 \rightarrow 0$  as  $(x, y) \rightarrow (0, 0)$ , hence the limit equals  $\lim_{t \rightarrow 0} \frac{\sin t}{t} = 1$ .

$$(c) \lim_{(x,y) \rightarrow (0,0)} \frac{xy}{x^2 + y^2} \text{ does not exist.}$$

As  $(x, y) \rightarrow (0, 0)$  we are having points in  $\mathbb{R}^2$  that converge towards the origin. The points can approach the origin along any path – if the limit exists, we will always get the same thing. One important strategy is thus to approach the origin along different paths: if the function converges to different values, then the limit doesn't exist! The simplest paths we could consider are lines that pass through the origin, i.e. points  $(x, mx)$ . For our example,  $\lim_{(x,mx) \rightarrow (0,0)} \frac{xy}{x^2 + y^2} = \lim_{(x,mx) \rightarrow (0,0)} \frac{mx^2}{(1 + m^2)x^2} = \frac{m}{1 + m^2}$ . The limit value depends on the gradient  $m$ , e.g. for  $m = 0$  we get 0 and for  $m = 1$  we get  $1/2$ , so the limit does not exist.

**Remark 8.5.** If  $\lim_{(x,y) \rightarrow (x_0,y_0)} f(x, y)$  exists, then  $\lim_{x \rightarrow x_0} \lim_{y \rightarrow y_0} f(x, y) = \lim_{y \rightarrow y_0} \lim_{x \rightarrow x_0} f(x, y)$ .

**Definition 8.6.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  and  $x_0 \in A$ . We say that  $f$  is continuous at  $x_0$  if

$$\forall V \in \mathcal{V}(f(x_0)), \exists U \in \mathcal{V}(x_0) \text{ s.t. } f(x) \in V, \forall x \in U \cap A.$$

**Remark 8.7.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  and  $x_0 \in A \cap A'$  an accumulation point. Then  $f$  is continuous at  $x_0$  if and only if

$$\lim_{x \rightarrow x_0} f(x) = f(x_0).$$

**Remark 8.8.** If  $x_0 \in A$  is an isolated point, then  $f$  is continuous at  $x_0$ .

**Theorem 8.9.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  and  $x_0 \in A \cap A'$ . The following are equivalent:

- (a)  $f$  is continuous at  $x_0$ .
- (b)  $\forall \varepsilon > 0, \exists \delta > 0$  s.t.  $|f(x) - f(x_0)| < \varepsilon, \forall x \in A$  with  $\|x - x_0\| < \delta$ .
- (c) for any sequence  $(x^k)$  in  $A$  with  $\lim_{k \rightarrow \infty} x^k = x_0$ , we have that  $\lim_{k \rightarrow \infty} f(x^k) = f(x_0)$ .

**Proposition 8.10** (Any norm is continuous).  $f : \mathbb{R}^n \rightarrow \mathbb{R}, f(x) = \|x\|$  is continuous on  $\mathbb{R}^n$ .

*Proof.* Let  $x_0 \in \mathbb{R}^n$  and  $(x^k)$  with  $x^k \rightarrow x_0$ . We have that  $\|x^k - x^0\| \rightarrow 0$ . By the triangle inequality  $|\|x^k\| - \|x^0\|| \leq \|x^k - x^0\| \rightarrow 0$ , hence  $\|x^k\| - \|x^0\| \rightarrow 0$ , i.e.  $\|x^k\| \rightarrow \|x^0\|$ .  $\square$

**Theorem 8.11** (Weierstrass). Let  $A \subseteq \mathbb{R}^n$  be closed and bounded, and  $f : A \rightarrow \mathbb{R}$  a continuous function. Then  $f$  is bounded and it attains its bounds, i.e. there exist  $\min(f(A)), \max(f(A))$ .  $\square$

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**Exercises. Seminar 9**

1. Study the limits of the following functions when  $(x, y) \rightarrow (0, 0)$ :

$$(a) \frac{x^2 - y^2}{x^2 + y^2}, \quad (b) \frac{x + y}{x^2 + y^2}, \quad (c) \frac{x^3 + y^3}{x^2 + y^2}, \quad (d) \frac{\sin x - \sin y}{x - y}.$$

2. Compute the partial derivatives (and specify where they exist) for the following functions:

$$(a) f(x, y) = e^{-(x^2+y^2)}, \quad (c) f(x, y) = \|(x, y)\| = \sqrt{x^2 + y^2},$$

$$(b) f(x, y) = \cos x \cos y - \sin x \sin y, \quad (d) f(x, y, z) = x^2 y z + y e^z.$$

3. Let  $f : \mathbb{R}^2 \rightarrow \mathbb{R}, f(x, y) = xy$ . Using the definition, prove that  $Df(x_0, y_0) = (y_0, x_0)$ .

4. Prove that

$$f(x, y) = \begin{cases} \frac{xy}{\sqrt{x^2 + y^2}}, & (x, y) \neq (0, 0) \\ 0, & (x, y) = (0, 0). \end{cases}$$

is continuous and has partial derivatives, but it is not differentiable in the origin.

5. Find the gradient of the function  $f$  at the point  $a$  for the following:

$$(a) f(x, y) = e^{-x} \sin(x + 2y), a = (0, \frac{\pi}{4}). \quad (c) f(x, y, z) = e^{xyz}, a = (0, 0, 0).$$

$$(b) f(x, y) = \arctan(\frac{y}{x}), a = (1, 1). \quad (d) f(x, y, z) = \sqrt{x^2 + y^2 + z^2}, a = (1, 1, 1).$$

## ※ Partial derivatives and differentiability in $\mathbb{R}^n$

The idea of a partial derivative is to vary the function along a single variable while keeping fixed all the other variables.

**Definition 9.1.** Let  $A \subseteq \mathbb{R}^n$  be an open set and  $f : A \rightarrow \mathbb{R}$ . The *partial derivative* of  $f$  with respect to  $x_i$  at the point  $x = (x_1, \dots, x_n) \in A$  is given by

$$\begin{aligned} \frac{\partial f}{\partial x_i}(x) &= \partial_i f(x) := \lim_{h \rightarrow 0} \frac{f(x_1, \dots, x_{i-1}, x_i + h, x_{i+1}, \dots, x_n) - f(x_1, \dots, x_n)}{h} \\ &= \lim_{h \rightarrow 0} \frac{f(x + he_i) - f(x)}{h}. \end{aligned}$$

Note that  $\frac{\partial f}{\partial x_i}$  is the derivative of  $f$  with respect to  $x_i$ , with the other variables held fixed.

**Definition 9.2.** For a function  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  that has partial derivatives at  $x \in A$  with respect to all its variables, the *gradient* at  $x$  is given by the vector  $\nabla f(x) \in \mathbb{R}^n$ ,

$$\nabla f(x) := \left( \frac{\partial f}{\partial x_1}(x), \dots, \frac{\partial f}{\partial x_n}(x) \right).$$

**Example 9.3.** For  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $f(x, y) = x^2y + y^2$  we have that  $\nabla f(x, y) = (\frac{\partial f}{\partial x}(x, y), \frac{\partial f}{\partial y}(x, y)) = (2xy, x^2 + 2y)$ .

**Example 9.4** (With partial derivatives but discontinuous). Let  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ ,

$$f(x, y) = \begin{cases} \frac{xy}{x^2 + y^2}, & (x, y) \neq (0, 0) \\ 0, & (x, y) = (0, 0). \end{cases}$$

Since  $f(x, 0) - f(0, 0) = 0$ ,  $\frac{\partial f}{\partial x}(0, 0) = \lim_{x \rightarrow 0} \frac{f(x, 0) - f(0, 0)}{x} = 0 = \frac{\partial f}{\partial y}(0, 0)$ , so  $f$  has partial derivatives zero at  $(0, 0)$ . But  $\lim_{(x, mx) \rightarrow (0, 0)} \frac{mx^2}{(m^2 + 1)x^2} = \frac{m}{(m^2 + 1)}$  depends on  $m$ , so  $f$  doesn't have a limit at  $(0, 0)$ , which means that  $f$  is discontinuous at  $(0, 0)$ .  $\square$

As the above example shows, a function that has partial derivatives at a point doesn't have to be continuous. This means that if we want to have good properties for differentiable functions, we have to find a better way of defining differentiability.

Let us recall an important idea for differentiable functions in  $\mathbb{R}$ :  $f(x_0) + f'(x_0)(x - x_0)$  is the linear approximation to  $f(x)$ . This comes from the definition of the derivative

$$\lim_{x \rightarrow x_0} \frac{f(x) - f(x_0) - f'(x_0)(x - x_0)}{x - x_0} = 0,$$

which can also be written as

$$f(x) = f(x_0) + f'(x_0)(x - x_0) + R(x - x_0), \text{ with } \frac{R(x - x_0)}{\|x - x_0\|} \rightarrow 0,$$

where  $R(x - x_0)$  is the remainder of the linear approximation.

**Definition 9.5.** Let  $A \subseteq \mathbb{R}^n$  be an open set and  $f : A \rightarrow \mathbb{R}$ . We say that  $f$  is *differentiable* at  $x_0 \in A$  if there exists a vector  $Df(x_0) \in \mathbb{R}^n$ , called the *differential/derivative* of  $f$  at  $x_0$ , s.t.

$$\lim_{x \rightarrow x_0} \frac{f(x) - f(x_0) - Df(x_0) \cdot (x - x_0)}{\|x - x_0\|} = 0.$$

Here  $Df(x_0) \cdot (x - x_0)$  is the dot product of two vectors. With  $h = x - x_0$ , we have that

$$\lim_{h \rightarrow 0} \frac{f(x_0 + h) - f(x_0) - Df(x_0) \cdot h}{\|h\|} = 0.$$

Note that differentiability is equivalent to

$$f(x) = f(x_0) + Df(x_0) \cdot (x - x_0) + R(x - x_0), \text{ with } \frac{R(x - x_0)}{\|x - x_0\|} \rightarrow 0,$$

where  $R(x - x_0)$  is the remainder of the linear approximation.

**Definition 9.6.** Let  $A \subseteq \mathbb{R}^n$  be an open set. If  $f : A \rightarrow \mathbb{R}^m$ ,  $f = (f_1, \dots, f_m)$ , then  $f$  is differentiable at  $x_0$  if there exists a matrix  $Df(x_0) \in \mathbb{R}^{m \times n}$ , called the *differential/derivative* of  $f$  at  $x_0$ , s.t.

$$\lim_{x \rightarrow x_0} \frac{\|f(x) - f(x_0) - Df(x_0)(x - x_0)\|_m}{\|x - x_0\|_n} = 0.$$

Here  $Df(x_0)(x - x_0)$  is a matrix–vector product:  $[ \quad ]_{m \times n} [ \quad ]_{n \times 1} = [ \quad ]_{m \times 1}$ .

**Example 9.7.** Constant functions have zero derivative and linear functions have a constant derivative.

- If  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is constant, then  $Df(x) = 0$  since  $f(x) = f(x_0)$  for any  $x, x_0 \in \mathbb{R}^n$ .
- If  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $f(x) = a \cdot x$  with  $a \in \mathbb{R}^n$ , then  $Df(x) = a$  since  $f(x) - f(x_0) - a \cdot (x - x_0) = 0$ .
- If  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ ,  $f(x) = Ax$  with  $A \in \mathbb{R}^{m \times n}$ , then  $Df(x) = A$ ;  $f(x) - f(x_0) - A(x - x_0) = 0$ .

**Theorem 9.8.** Let  $A \subseteq \mathbb{R}^n$  be an open set. If  $f : A \rightarrow \mathbb{R}^m$  is differentiable at  $x_0$ , then  $f$  is continuous at  $x_0$ .

*Proof.* Since  $f$  is differentiable,  $f(x) = f(x_0) + Df(x_0)(x - x_0) + \frac{R(x - x_0)}{\|x - x_0\|} \|x - x_0\|$ . Letting  $x \rightarrow x_0$  we use that  $\frac{R(x - x_0)}{\|x - x_0\|} \rightarrow 0$  to obtain that  $f(x) \rightarrow f(x_0)$ .  $\square$

**Theorem 9.9.** Let  $A \subseteq \mathbb{R}^n$  be an open set and  $x \in A$ . If  $f : A \rightarrow \mathbb{R}$  is differentiable at  $x$ , then the partial derivatives exist at  $x$  and

$$Df(x) = \nabla f(x) = \left( \frac{\partial f}{\partial x_1}(x), \dots, \frac{\partial f}{\partial x_n}(x) \right).$$

*Proof.* Differentiability at  $x$  gives

$$\lim_{h \rightarrow 0} \frac{f(x+h) - f(x) - Df(x) \cdot h}{\|h\|} = 0.$$

Let us take the vector  $h$  in the direction of  $e_i$ , with a non-zero value on the  $i$ th component only, i.e.  $h = (0, \dots, 0, h_i, 0, \dots, 0) = h_i e_i$ . Then we have that

$$\lim_{h_i \rightarrow 0} \frac{f(x + h_i e_i) - f(x) - Df(x) \cdot h_i e_i}{|h_i|} = 0 = \lim_{h_i \rightarrow 0} \frac{f(x + h_i e_i) - f(x) - Df(x) \cdot h_i e_i}{h_i},$$

which gives that

$$\frac{\partial f}{\partial x_i}(x) = \lim_{h_i \rightarrow 0} \frac{f(x + h_i e_i) - f(x)}{h_i} = Df(x) \cdot e_i = Df(x)_i,$$

hence the  $i$ th component of  $Df(x)$  is  $\frac{\partial f}{\partial x_i}(x)$ , which means that  $Df(x) = \nabla f(x)$ .  $\square$

**Theorem 9.10.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  and  $x \in A$ . If all the partial derivatives exist and are continuous at  $x$ , then  $f$  is differentiable at  $x$ .

*Proof.* (Optional) See [1][Theorem 9.21].  $\square$

It is possible for a function to have partial derivatives, but not be differentiable if the partial derivatives are not continuous. The function in [Theorem 9.4](#) has partial derivatives at  $(0, 0)$ , but it is discontinuous there, so it is not differentiable at that point.

**Theorem 9.11.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^m$ ,  $f = (f_1, \dots, f_m)$  be differentiable at  $x \in A$ , then

$$Df(x) = J = \begin{bmatrix} \nabla f_1(x) \\ \vdots \\ \nabla f_m(x) \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1}(x) & \dots & \frac{\partial f_1}{\partial x_n}(x) \\ \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1}(x) & \dots & \frac{\partial f_m}{\partial x_n}(x) \end{bmatrix}_{m \times n}.$$

This matrix is called the *Jacobian matrix* and is typically denoted by  $J$ .

**Theorem 9.12** (Calculus rules). Let  $A \subseteq \mathbb{R}^n$  and  $f, g : A \rightarrow \mathbb{R}$  differentiable at  $x \in A$ . Then

$$1. \nabla(f + g)(x) = \nabla f(x) + \nabla g(x).$$

$$2. \nabla(fg)(x) = g(x)\nabla f(x) + f(x)\nabla g(x).$$

$$3. \nabla\left(\frac{f}{g}\right)(x) = \frac{g(x)\nabla f(x) - f(x)\nabla g(x)}{g^2(x)}.$$

**Theorem 9.13** (Chain rule). Let  $g : \mathbb{R}^n \rightarrow \mathbb{R}^m$ ,  $f : \mathbb{R}^m \rightarrow \mathbb{R}^p$  differentiable at  $x$  and  $g(x)$ , respectively. Then

$$D(f \circ g)(x) = Df(g(x))Dg(x).$$

In terms of matrix dimensions:  $[ \quad ]_{p \times n} = [ \quad ]_{p \times m} [ \quad ]_{m \times n}$ .

*Proof.* (Optional) Considering  $E(h) := \|f(g(x+h)) - f(g(x)) - Df(g(x))Dg(x)h\|$  we aim to prove that  $\lim_{h \rightarrow 0} \frac{E(h)}{\|h\|} = 0$ . Since

$$\begin{aligned} E(h) &= \|f(g(x+h)) - f(g(x)) - Df(g(x))(g(x+h) - g(x)) \\ &\quad + Df(g(x))(g(x+h) - g(x)) - Df(g(x))Dg(x)h\|, \end{aligned}$$

using the triangle inequality we have that

$$\begin{aligned} E(h) &\leq \|f(g(x+h)) - f(g(x)) - Df(g(x))(g(x+h) - g(x))\| \\ &\quad + \|Df(g(x))\| \|g(x+h) - g(x) - Dg(x)h\|. \end{aligned}$$

At this point, after a few intermediate steps, one can now divide by  $\|h\|$ , take  $h \rightarrow 0$  and use the differentiability of  $f$  at  $g(x)$ , and of  $g$  at  $x$ , together with the fact that  $\|Df(g(x))\|$  is bounded and independent of  $h$ .  $\square$

**Example 9.14** (Chain rule). Let  $g : \mathbb{R} \rightarrow \mathbb{R}^n$ ,  $g(t) = (g_1(t), \dots, g_n(t))$ ,  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $f \circ g : \mathbb{R} \rightarrow \mathbb{R}$ . Since  $Df(x) = \nabla f(x)$  and  $Dg(t) = g'(t) = (g'_1(t), \dots, g'_n(t))$  we have that

$$\begin{aligned} D(f \circ g)(t) &= (f \circ g)'(t) = \nabla f(g(t)) \cdot g'(t) \\ &= \sum_{i=1}^n \frac{\partial f}{\partial x_i}(g(t)) g'_i(t). \end{aligned}$$

If  $n = 2$  and  $g(t) = (x(t), y(t))$ , then  $(f \circ g)(t) = f(x(t), y(t)) =: h(t)$  and

$$\frac{dh}{dt} = \frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt}.$$

**Example 9.15** (Chain rule). Let  $g : \mathbb{R}^n \rightarrow \mathbb{R}^m$ ,  $g(x) = (g_1(x), \dots, g_m(x))$ ,  $f : \mathbb{R}^m \rightarrow \mathbb{R}$ ,  $f \circ g : \mathbb{R}^n \rightarrow \mathbb{R}$ . Since

$$Df(x) = \nabla f(x) = \left[ \frac{\partial f}{\partial x_1}(x) \quad \dots \quad \frac{\partial f}{\partial x_m}(x) \right]_{1 \times m}$$

and

$$Dg(x) = \begin{bmatrix} \nabla g_1(x) \\ \vdots \\ \nabla g_m(x) \end{bmatrix} = \begin{bmatrix} \frac{\partial g_1}{\partial x_1}(x) & \dots & \frac{\partial g_1}{\partial x_n}(x) \\ \vdots & & \vdots \\ \frac{\partial g_m}{\partial x_1}(x) & \dots & \frac{\partial g_m}{\partial x_n}(x) \end{bmatrix}_{m \times n}.$$

we have that

$$\begin{aligned} D(f \circ g)(x) &= Df(g(x))Dg(x) \\ &= \begin{bmatrix} \frac{\partial f}{\partial x_1}(g(x)) & \dots & \frac{\partial f}{\partial x_m}(g(x)) \end{bmatrix}_{1 \times m} \begin{bmatrix} \frac{\partial g_1}{\partial x_1}(x) & \dots & \frac{\partial g_1}{\partial x_n}(x) \\ \vdots & & \vdots \\ \frac{\partial g_m}{\partial x_1}(x) & \dots & \frac{\partial g_m}{\partial x_n}(x) \end{bmatrix}_{m \times n}. \end{aligned}$$

### Exercises. Seminar 10

- For each of the following functions, compute  $\frac{df}{dt}$  directly and using the chain rule:
  - $f(x, y) = \ln(x^2 + y^2)$ ,  
 $x = t, y = t^2$ .
  - $f(x, y, z) = \sqrt{x^2 + y^2 + z^2}$ ,  
 $x = \cos t, y = \sin t, z = t > 0$ .
- Let  $f(u, v, w) = u^2 + v^2 - w$  and  $u(x, y, z) = x^2y, v(x, y, z) = y^2, w(x, y, z) = e^{-xz}$ . Find  $\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}, \frac{\partial f}{\partial z}$  directly and using the chain rule.
- For  $f: \mathbb{R}^2 \rightarrow \mathbb{R}, f(x, y) = x^2 + xy$  find:
  - the gradient of  $f$  and the direction of steepest descent at the point  $(1, 0)$ .
  - the directional derivative at the point  $(1, 0)$  in the direction of  $e_1 + e_2 = (1, 1)$ .
- Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}, f(x) = \frac{1}{2}\|x\|^2$ . Find the gradient of  $f$ . Find the directional derivative  $D_v f(x)$  in two ways: using the definition and using the gradient.
- Let  $D = \text{diag}(d_1, \dots, d_n)$  be a diagonal  $n \times n$  matrix and consider the quadratic function  $f: \mathbb{R}^n \rightarrow \mathbb{R}, f(x) = \frac{1}{2}x^T D x$ . Prove that  $\nabla f(x) = Dx$  and  $H(x) = D$ .
- Find the equation of the tangent line to the ellipse  $\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$  at an arbitrary point  $(x_0, y_0)$ .

### Exercises. Homework 10

1. Find the tangent plane to the unit sphere  $x^2 + y^2 + z^2 = 1$  at an arbitrary point  $(x_0, y_0, z_0)$ .
2. Let  $x, y, f : \mathbb{R}^2 \rightarrow \mathbb{R}$  and  $f(x, y) = f(x(u, v), y(u, v))$ . Prove that

$$\frac{\partial f}{\partial u} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial u} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial u} \quad \text{and} \quad \frac{\partial f}{\partial v} = \frac{\partial f}{\partial x} \frac{\partial x}{\partial v} + \frac{\partial f}{\partial y} \frac{\partial y}{\partial v}.$$

3. Consider the quadratic function  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $f(x, y) = \frac{1}{2}(x^2 + by^2)$  with  $b > 0$  and the following gradient descent method for finding its minimum

$$(x_{k+1}, y_{k+1}) = (x_k, y_k) - s_k \nabla f(x_k, y_k),$$

with step size  $s_k > 0$  (also called learning rate in the machine learning literature).

- (a) Find the step size  $s_k$  using exact line search, such that it minimizes the function

$$\varphi(s_k) = f(x_{k+1}, y_{k+1}) = f((x_k, y_k) - s_k \nabla f(x_k, y_k)).$$

Note that for this optimal step size  $s_k$  we have that  $\varphi'(s_k) = 0$ .

- (b) [Python] For  $b = 1, \frac{1}{2}, \frac{1}{5}, \frac{1}{10}$  plot the gradient descent iterations and some relevant contour lines of  $f$  to show convergence towards the minimum. What do you notice as  $b$  gets smaller?



## ※ Directional derivatives. Gradient descent

**Definition 10.1.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  and a vector  $v \in \mathbb{R}^n$ . The derivative of  $f$  in the direction of  $v$  at  $x \in A$  (*directional derivative*) is given by

$$Df_v(x) := \lim_{h \rightarrow 0} \frac{f(x + hv) - f(x)}{h}.$$

Note that here  $h$  is a scalar. The directional derivative  $Df_v(x)$  is also denoted by  $\partial_v f(x)$ .

**Theorem 10.2.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  and  $v \in \mathbb{R}^n$ . If  $f$  is differentiable at  $x \in A$ , then

$$Df_v(x) = \nabla f(x) \cdot v.$$

*Proof.* From differentiability we have that

$$\lim_{h \rightarrow 0} \frac{f(x + hv) - f(x) - \nabla f(x) \cdot hv}{\|hv\|} = 0.$$

Since  $\|hv\| = |h|\|v\|$ , this gives that

$$\lim_{h \rightarrow 0} \frac{f(x + hv) - f(x) - \nabla f(x) \cdot hv}{h} = 0,$$

which can be rearranged as

$$Df_v(x) = \lim_{h \rightarrow 0} \frac{f(x + hv) - f(x)}{h} = \nabla f(x) \cdot v.$$

□

**Theorem 10.3** (Fermat). Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable at  $x \in A$ . If  $x$  is a local extremum, then it is a critical point, i.e.  $\nabla f(x) = 0$ .

*Proof.*  $x$  is an extremum in every direction, thus  $0 = Df_v(x) = \nabla f(x) \cdot v$  for every  $v \in \mathbb{R}^n$  (including the canonical vectors  $e_i$ ). This gives that  $\nabla f(x) = 0$ . □

**Proposition 10.4** (Direction of steepest ascent/descent). Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable at  $x \in A$  with  $\nabla f(x) \neq 0$ . Then

- $\nabla f(x)$  gives the direction of fastest increase (steepest ascent).
- $-\nabla f(x)$  gives the direction of fastest decrease (steepest descent).

*Proof.* Since  $Df_v(x) = \nabla f(x) \cdot v$ , by the Cauchy-Schwarz inequality we have that

$$-\|\nabla f(x)\|\|v\| \leq Df_v(x) \leq \|\nabla f(x)\|\|v\|,$$

with the maximum obtained for  $v = \alpha \nabla f(x)$ , the minimum for  $v = -\alpha \nabla f(x)$ ,  $\alpha > 0$ . □

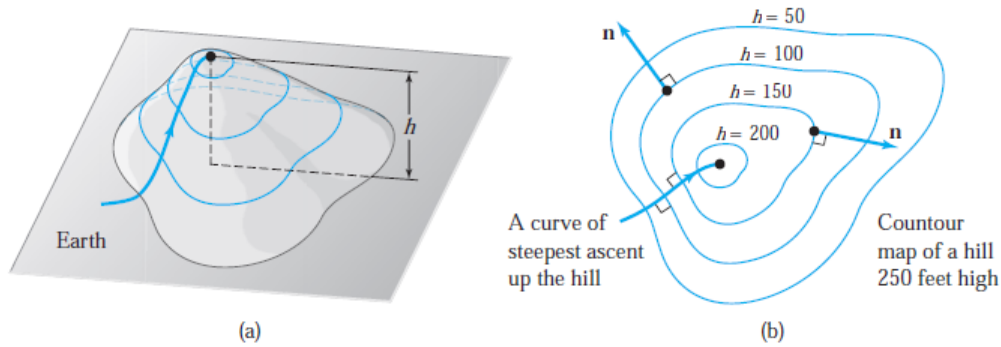


Figure 6: The gradient is the steepest ascent direction and is perpendicular to the level curves. Source: [4, page 140].

**Proposition 10.5** (Gradient perpendicular to the level set). Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable and let  $x_0 \in A$  be on the level set  $S$  defined by  $f(x) = f(x_0)$ . Then the gradient  $\nabla f(x)$  is perpendicular to the level set  $S$ , meaning that for a tangent vector  $v$  to the level set  $S$  it holds that  $Df_v(x) = \nabla f(x) \cdot v = 0$ .

*Proof.* Let  $c(t)$ ,  $t \geq 0$  be a path on the level set, i.e.  $f(c(t)) = \text{constant}$ , that starts from  $x_0 = c(0)$ . Let  $v = c'(0)$  be the tangent vector to the path at  $t = 0$ . By the chain rule

$$0 = \left. \frac{d}{dt} f(c(t)) \right|_{t=0} = \nabla f(c(0)) \cdot c'(0) = \nabla f(x_0) \cdot v.$$

□

**Example 10.6** (Tangent line to a level curve). Consider a level curve  $L = \{(x, y) \in \mathbb{R}^2 \mid f(x, y) = c\}$  and a point  $(x_0, y_0)$  on it. Take the tangent line at that point. If  $(x, y)$  is a point on the tangent line, then the gradient is perpendicular to the vector  $(x - x_0, y - y_0)$ ,

$$\nabla f(x_0, y_0) \cdot (x - x_0, y - y_0) = 0,$$

hence the equation of the tangent line is given by

$$\frac{\partial f}{\partial x}(x_0, y_0) \cdot (x - x_0) + \frac{\partial f}{\partial y}(x_0, y_0) \cdot (y - y_0) = 0.$$

**Example 10.7** (Tangent plane to a level surface). Consider a level curve  $L = \{(x, y, z) \in \mathbb{R}^3 \mid f(x, y, z) = c\}$  and a point  $(x_0, y_0, z_0)$  on it. Take the tangent plane at that point. If  $(x, y, z)$  is a point on the tangent plane, then

$$\nabla f(x_0, y_0, z_0) \cdot (x - x_0, y - y_0, z - z_0) = 0,$$

hence the equation of the tangent plane is given by

$$\frac{\partial f}{\partial x}(x_0, y_0, z_0) \cdot (x - x_0) + \frac{\partial f}{\partial y}(x_0, y_0, z_0) \cdot (y - y_0) + \frac{\partial f}{\partial z}(x_0, y_0, z_0) \cdot (z - z_0) = 0.$$

**Gradient descent: how neural networks learn (Youtube video 3Blue1Brown).** One of the most important optimization methods is based on the fundamental idea that the direction of steepest descent on the graph of a function  $f$  is given by  $-\nabla f$  (see [Theorem 10.4](#)). If we want to minimize a function  $f$ , this naturally suggests an iterative method in which from a current position  $x_k$  we move in the direction of  $-\nabla f(x_k)$  with a step size  $s_k > 0$  in order to get to the next position  $x_{k+1}$  such that  $f(x_{k+1}) < f(x_k)$ . The gradient descent method starts from an initial value  $x_0 \in \mathbb{R}^n$  and then for  $k \geq 0$  takes

$$x_{k+1} = x_k - s_k \nabla f(x_k).$$

The step size (also called learning rate)  $s_k$  has to be chosen at every iteration. One way of doing this is called *exact line search*: we look for the optimal step size  $s_k$  that minimizes the function  $\varphi(s) = f(x_{k+1}) = f(x_k - s \nabla f(x_k))$ . By the chain rule we have that

$$\varphi'(s) = \nabla f(x_{k+1}) \cdot \frac{d}{ds} x_{k+1} = \nabla f(x_{k+1}) \cdot (-\nabla f(x_k)).$$

Since  $\varphi'(s_k) = 0$  for the optimal step size  $s_k$ , we see that  $\nabla f(x_{k+1}) \cdot \nabla f(x_k) = 0$ , which means that two consecutive search directions are perpendicular, i.e.  $(x_{k+2} - x_{k+1}) \perp (x_{k+1} - x_k)$ .

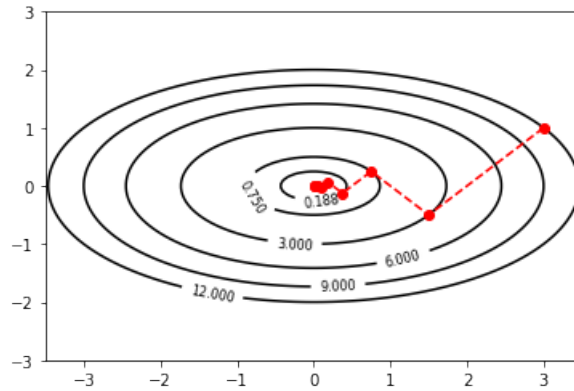


Figure 7: Level curves and gradient descent iterations for [Theorem 10.8](#).

**Example 10.8.** Consider the quadratic function  $f(x, y) = x^2 + 3y^2$  which has a unique global minimum at the origin  $(0, 0)$ . The gradient is given by  $\nabla f(x, y) = (2x, 6y)$ .

Gradient descent: starting from an initial value  $(x_0, y_0)$  consider the sequence

$$(x_{k+1}, y_{k+1}) = (x_k, y_k) - s \nabla f(x_k, y_k),$$

that is

$$x_{k+1} = (1 - 2s)x_k, \quad y_{k+1} = (1 - 6s)y_k.$$

The step size is determined using exact line search. We look for the optimal step size (learning rate)  $s > 0$  by minimizing the function

$$\varphi(s) = f(x_{k+1}, y_{k+1}) = (1 - 2s)^2 x_k^2 + 3(1 - 6s)^2 y_k^2 \rightarrow \min.$$

We want

$$\varphi'(s) = 0 \text{ with } \varphi'(s) = -4(1 - 2s)x_k^2 - 36(1 - 6s)y_k^2,$$

hence we obtain the optimal step size  $s = \frac{x_k^2 + 9y_k^2}{2x_k^2 + 54y_k^2}$ . In Fig. 7 we have the gradient descent iterations starting from the initial value  $(3, 1)$  and converging towards the solution  $(0, 0)$ .  $\square$

## ※ Higher order derivatives. Taylor expansion. Local extremum

The second order partial derivative with respect to  $x_i$  is simply

$$\frac{\partial}{\partial x_i} \left( \frac{\partial f}{\partial x_i} \right) =: \frac{\partial^2 f}{\partial x_i^2} = \partial_i^2 f$$

and the mixed second order partial derivative is

$$\frac{\partial}{\partial x_i} \left( \frac{\partial f}{\partial x_j} \right) =: \frac{\partial^2 f}{\partial x_i \partial x_j} = \partial_{i,j}^2 f.$$

**Example 11.1.** For  $f(x, y) = x^2y + (x + 2y)^3$  we have that

$$\begin{aligned} \frac{\partial f}{\partial x} &= 2xy + 3(x + 2y)^2, & \frac{\partial^2 f}{\partial x^2} &= 2y + 6(x + 2y), & \frac{\partial^2 f}{\partial y \partial x} &= 2x + 12(x + 2y) \\ \frac{\partial f}{\partial y} &= x^2 + 6(x + 2y)^2, & \frac{\partial^2 f}{\partial y^2} &= 24(x + 2y), & \frac{\partial^2 f}{\partial x \partial y} &= 2x + 12(x + 2y) \end{aligned}$$

**Theorem 11.2** (Schwarz). If  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  has continuous second order partial derivatives, then if  $i \neq j$

$$\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i}.$$

*Proof.* (Optional) See [4][page 151]. The proof is based on the mean value theorem.  $\square$

**Definition 11.3.** For  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  the *Hessian matrix* is defined by

$$H(x) = D^2f(x) = D(\nabla f)(x) = \begin{bmatrix} \nabla \left( \frac{\partial f}{\partial x_1} \right) \\ \nabla \left( \frac{\partial f}{\partial x_2} \right) \\ \vdots \\ \nabla \left( \frac{\partial f}{\partial x_n} \right) \end{bmatrix} = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_2 \partial x_1} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_1} \\ \frac{\partial^2 f}{\partial x_1 \partial x_2} & \frac{\partial^2 f}{\partial x_2^2} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_2} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_1 \partial x_n} & \frac{\partial^2 f}{\partial x_2 \partial x_n} & \cdots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}_{n \times n}.$$

If the second order derivatives are continuous, then the Hessian matrix  $H(x)$  is symmetric.

**Theorem 11.4** (Taylor). Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  be of class  $C^2$  (twice differentiable, with continuous second order partial derivatives) and  $x_0 \in A$ . Then we have that

$$f(x) = f(x_0) + \nabla f(x_0) \cdot (x - x_0) + \frac{1}{2}(x - x_0)^T H(x_0)(x - x_0) + R(x - x_0),$$

with a remainder  $R(x - x_0)$  s.t.  $R(x - x_0)/\|x - x_0\|^2 \rightarrow 0$  as  $x \rightarrow x_0$ . Taylor's expansion can also be written as

$$f(x_0 + h) = f(x_0) + \nabla f(x_0) \cdot h + \frac{1}{2}h^T H(x_0)h + R(h).$$

Note that  $\nabla f(x_0) \cdot h$  is linear in  $h$ ,  $\frac{1}{2}h^T H(x_0)h$  is quadratic in  $h$  and  $R(h)/\|h\|^2 \rightarrow 0$ .

*Proof.* Consider the function  $g : [0, 1] \rightarrow \mathbb{R}$ ,  $g(t) = f(x_0 + th)$ . Using the classical 1d Taylor's expansion ([Theorem 5.4](#))

$$g(1) = g(0) + g'(0) + \frac{1}{2}g''(0) + R_2, \quad (1)$$

where  $R_2$  is a remainder. Note that  $g(1) = f(x_0 + h)$  and  $g(0) = f(x_0)$ . Using the chain rule we have that

$$g'(t) = \nabla f(x_0 + th) \cdot h,$$

hence

$$g'(0) = \nabla f(x_0) \cdot h$$

and

$$\begin{aligned} g''(t) &= \frac{d}{dt} \left( \nabla f(x_0 + th) \cdot h \right) = \frac{d}{dt} \left( \sum_{i=1}^n \frac{\partial f}{\partial x_i}(x_0 + th) h_i \right) = \sum_{i=1}^n h_i \frac{d}{dt} \left( \frac{\partial f}{\partial x_i}(x_0 + th) \right) \\ &= \sum_{i=1}^n h_i \underbrace{\left( \nabla \left( \frac{\partial f}{\partial x_i}(x_0 + th) \right) \cdot h \right)}_{(H(x_0 + th)h)_i} = h \cdot (H(x_0 + th)h) = h^T H(x_0 + th)h. \end{aligned}$$

This gives

$$g''(0) = h^T H(x_0)h$$

and from (1) we finally get that

$$f(x_0 + h) = f(x_0) + \nabla f(x_0) \cdot h + \frac{1}{2}h^T H(x_0)h + R(h).$$

□

**Proposition 11.5.** Let  $H$  be an  $n \times n$  matrix such that  $x^T H x > 0$ ,  $\forall x \in \mathbb{R}^n$ . Then there exists a constant  $M > 0$  such that  $x^T H x \geq M\|x\|^2$ ,  $\forall x \in \mathbb{R}^n$ .

*Proof.* Define  $g : \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $g(x) = x^T H x$ . Notice that  $g(x) = g(\|x\| \frac{x}{\|x\|}) = \|x\|^2 g(\frac{x}{\|x\|})$  and that  $\frac{x}{\|x\|}$  has norm 1. Since the quadratic function  $g$  is positive and continuous on the unit sphere (vectors with norm 1), by the Weierstrass theorem it attains its minimum on the unit sphere, hence there exists  $M > 0$  such that  $g(\frac{x}{\|x\|}) \geq M$ . Thus  $g(x) \geq M\|x\|^2$ .  $\square$

**Corollary 11.6.** Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  be of class  $C^2$  with  $\nabla f(x_0) = 0$ . Then

- If  $h^T H(x_0) h > 0$ ,  $\forall h \in \mathbb{R}^n$ , then  $x_0$  is a local minimum.
- If  $h^T H(x_0) h < 0$ ,  $\forall h \in \mathbb{R}^n$ , then  $x_0$  is a local maximum.

*Proof.* Using that  $\nabla f(x_0) = 0$ , we have from Taylor's theorem that

$$f(x_0 + h) = f(x_0) + \frac{1}{2} h^T H(x_0) h + R(h), \quad \text{with } R(h)/\|h\|^2 \rightarrow 0.$$

Since  $h^T H(x_0) h > 0$ , by [Theorem 11.5](#) there exists  $M > 0$  such that  $\frac{1}{2} h^T H(x_0) h \geq M\|h\|^2$  then  $f(x_0 + h) \geq f(x_0) + M\|h\|^2 + R(h)$ . Since  $R(h)/\|h\|^2 \rightarrow 0$ , we have that  $M\|h\|^2 + R(h) > 0$  for small enough  $\|h\|$ . It follows that  $f(x_0 + h) \geq f(x_0)$  for small enough  $\|h\|$ , so  $x_0$  is a local minimum. The other case is similar.  $\square$

We see that Hessian matrices  $H$  for which the quadratic expression  $h^T H h$  is either positive or negative play a crucial role in determining if a critical point is a local minimum or maximum. We now recall the following related results from linear algebra.

**Definition 11.7.** An  $n \times n$  matrix  $A$  is called:

- *positive-definite* if  $x^T A x > 0$ ,  $\forall x \in \mathbb{R}^n$ .
- *negative-definite* if  $x^T A x < 0$ ,  $\forall x \in \mathbb{R}^n$ .
- *indefinite* if there exist  $x_1, x_2 \in \mathbb{R}^n$  s.t.  $x_1^T A x_1 > 0 > x_2^T A x_2$ .

**Proposition 11.8.** Let  $A$  be a symmetric  $n \times n$  matrix. Then

- $A$  is positive definite if and only if its eigenvalues are positive.
- $A$  is negative definite if and only if its eigenvalues are negative.
- $A$  is indefinite if and only if it has both positive and negative eigenvalues.

*Proof.* Can be found in any linear algebra textbook, see e.g. [\[6\]](#)[Section I.7].  $\square$

**Theorem 11.9** (Local extremum conditions). Let  $f : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  be of class  $C^2$  and  $x \in A$  a critical point with  $\nabla f(x) = 0$ .

- If  $H(x)$  is positive definite (positive eigenvalues), then  $x$  is a local minimum.
- If  $H(x)$  is negative definite (negative eigenvalues), then  $x$  is a local maximum.
- If  $H(x)$  is indefinite, then  $x$  is called a saddle point.

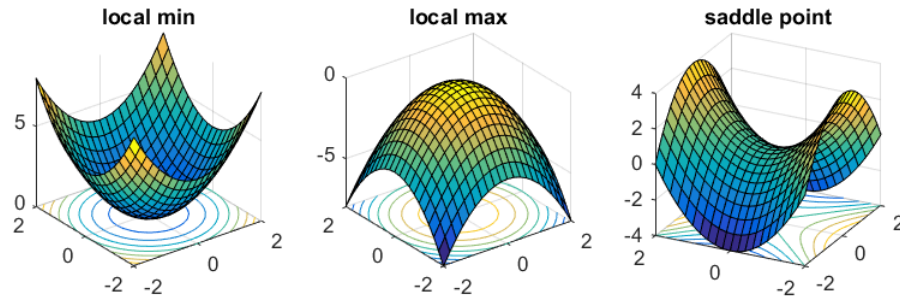


Figure 8: Examples of critical points: local minimum, local maximum and saddle point.

**Example 11.10.** The function  $f(x, y) = x^2 - y^2$  has a unique critical point because  $\nabla f(x, y) = (2x, -2y) = (0, 0)$  only at  $(0, 0)$ . This critical point is a saddle point since the Hessian matrix  $H(0, 0) = \begin{bmatrix} 2 & 0 \\ 0 & -2 \end{bmatrix}$  has eigenvalues 2 and  $-2$ . Note that a saddle point is a minimum in some directions, and a maximum in others.

**Applications. Linear regression.** Consider  $n$  data points  $(x_i, y_i)$ ,  $i = \overline{1, n}$ , with distinct  $x_i$ 's for which we want to find the line of best fit  $y = f(x) = ax + b$  that minimizes the least squares error

$$E = \frac{1}{2} \sum_{i=1}^n (y_i - f(x_i))^2 \rightarrow \min.$$

In other words, we are looking to find the optimal values  $a, b$  that minimize the function

$$E = E(a, b) = \frac{1}{2} \sum_{i=1}^n (ax_i + b - y_i)^2 \rightarrow \min,$$

which means we have to look through the critical points

$$\begin{aligned} 0 = \frac{\partial E}{\partial a} &= \sum_{i=1}^n x_i(ax_i + b - y_i) \implies a \sum_{i=1}^n x_i^2 + b \sum_{i=1}^n x_i = \sum_{i=1}^n x_i y_i \\ 0 = \frac{\partial E}{\partial b} &= \sum_{i=1}^n (ax_i + b - y_i) \implies a \sum_{i=1}^n x_i + bn = \sum_{i=1}^n y_i. \end{aligned}$$

These two simultaneous equations can be written as the linear system

$$\begin{bmatrix} \sum_{i=1}^n x_i^2 & \sum_{i=1}^n x_i \\ \sum_{i=1}^n x_i & n \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^n x_i y_i \\ \sum_{i=1}^n y_i \end{bmatrix},$$

which has a unique solution  $a, b$  since  $\det = n \sum_{i=1}^n x_i^2 - (\sum_{i=1}^n x_i)^2 > 0$  by the Cauchy-Schwarz inequality. Notice that the matrix of the linear systems is actually the Hessian matrix of  $E(a, b)$  and that it is positive definite (since its trace and determinant are positive, the eigenvalues must be positive), which means that the critical point is indeed a minimum point.

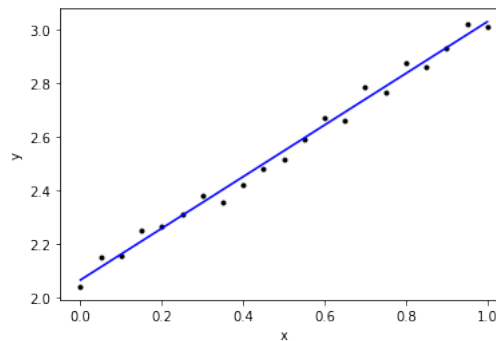


Figure 9: Example of linear regression.

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### Exercises. Seminar 11

1. Find the second-order Taylor polynomial for the following functions at the given points:
  - (a)  $f(x, y) = \sin(x + 2y)$  at  $(0, 0)$ .
  - (b)  $f(x, y) = e^{x+y}$  at  $(0, 0)$  and  $(1, -1)$ .
  - (c)  $f(x, y) = \sin(x) \sin(y)$  at  $(\pi/2, \pi/2)$ .
  - (d)  $f(x, y) = e^{-(x^2+y^2)}$  at  $(0, 0)$ .
2. Compute the Hessian matrix and its eigenvalues for the following:
  - (a)  $f(x, y) = (y - 1)e^x + (x - 1)e^y$  at  $(0, 0)$ .
  - (b)  $f(x, y) = \sin(x) \cos(y)$  at  $(\pi/2, 0)$ .
3. Find and classify the critical points for each of the following functions:
  - (a)  $f(x, y) = x^3 - 3x + y^2$ .
  - (b)  $f(x, y) = x^3 + y^3 - 6xy$ .
  - (c)  $f(x, y) = x^4 + y^4 - 4(x - y)^2$ .
  - (d)  $f(x, y, z) = x^2 + y^2 + z^2 - xy + x - 2z$ .
4. Let  $A$  be a symmetric  $n \times n$  matrix and the quadratic function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $f(x) = \frac{1}{2}x^T A x$ . Prove that  $\nabla f(x) = Ax$  and  $H(x) = A$ . *Hint: use the Taylor expansion.*



5. Let  $A$  be an  $m \times n$  matrix,  $b$  a vector in  $\mathbb{R}^m$  and the least squares minimization problem

$$\min_{x \in \mathbb{R}^n} \|Ax - b\|^2.$$

Prove that the solution  $x^*$  of this problem satisfies (the so-called normal equations)

$$A^T A x^* = A^T b.$$

### Exercises. Homework 11

1. Let  $A$  be a  $2 \times 2$  matrix and let the quadratic function  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $f(x) = \frac{1}{2}x^T A x$ .

- (a) Give a matrix  $A$  such that  $f$  has a unique minimum.
- (b) Give a matrix  $A$  such that  $f$  has a unique maximum.
- (c) Give a matrix  $A$  such that  $f$  has a unique saddle point.

In each case plot the 3d surface, three contour lines and the gradient at three different points.

## ※ Optimization with constraints. Lagrange multipliers

We have seen how to find the extremum points of a differentiable function by using [Theorem 11.9](#) – we take the critical points (where the gradient vanishes) and find if they are local minima/maxima by checking if the Hessian matrix is positive/negative definite.

Let us now consider the problem in which we want to optimize (minimize/maximize) a function and there is a constraint that must be satisfied, i.e.

$$\begin{aligned} &\text{optimize } f(x) \\ &\text{subject to } g(x) = c. \end{aligned}$$

We are thus looking for the minimum/maximum of  $f$  on the level set  $S = \{x \mid g(x) = c\}$ .

**Theorem 12.1.** Let  $f, g : A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable functions. Let  $x_0 \in A$  be a solution to the problem

$$\begin{aligned} &\text{optimize } f(x) \\ &\text{subject to } g(x) = c. \end{aligned}$$

Then  $x_0$  is a critical point of the so-called Lagrangian function

$$L(x, \lambda) = f(x) - \lambda(g(x) - c),$$

meaning that there exists  $\lambda \in \mathbb{R}$  (Lagrange multiplier) s.t.  $\nabla f(x_0) = \lambda \nabla g(x_0)$ .

*Proof.* Consider the level set  $S = \{x \in A \mid g(x) = c\}$  and an arbitrary path  $c(t)$  in  $S$  with  $c(0) = x_0$  and  $c'(0) = v$ . Since the gradient is perpendicular to the level set (see [Theorem 10.5](#)), we have that  $\nabla g(x_0) \cdot v = 0$ . Since  $x_0$  is a local extremum of  $f$  along a path  $c(t)$ , we have by the chain rule that

$$0 = \left. \frac{d}{dt} f(c(t)) \right|_{t=0} = \nabla f(x_0) \cdot v,$$

hence  $\nabla f(x_0) \cdot v = 0 = \nabla g(x_0) \cdot v$  for an arbitrary direction  $v$ , which means that there exists  $\lambda \in \mathbb{R}$  s.t.  $\nabla f(x_0) = \lambda \nabla g(x_0)$ .  $\square$

The idea for constrained optimization is thus to consider the Lagrangian function

$$L(x, \lambda) = f(x) - \lambda(g(x) - c)$$

and its critical points for which  $\nabla_x L = 0$  and  $\frac{\partial L}{\partial \lambda} = 0$ . This gives a system  $(n + 1)$  equations and  $(n + 1)$  unknowns.

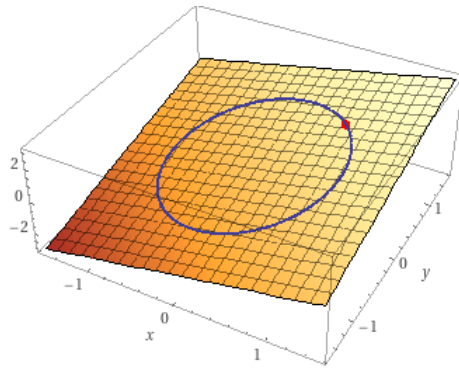


Figure 10: Sketch for [Theorem 12.2](#).

**Example 12.2.** Maximize  $f(x, y) = x + y$  subject to the constraint  $x^2 + y^2 = 1$ . This means finding the highest point on the plane  $z = x + y$  subject to  $(x, y)$  being on the unit circle. Consider the Lagrangian  $L(x, y, \lambda) = f(x, y) + \lambda(x^2 + y^2 - 1) = x + y + \lambda(x^2 + y^2 - 1)$  and look for the critical points:

$$\begin{aligned} 0 &= \frac{\partial L}{\partial x} = 1 + 2\lambda x, \\ 0 &= \frac{\partial L}{\partial y} = 1 + 2\lambda y, \\ 0 &= \frac{\partial L}{\partial \lambda} = x^2 + y^2 - 1. \end{aligned}$$

From the first two equations we get that  $x = y = -\frac{1}{2\lambda}$  and from the third one that  $\lambda^2 = \frac{1}{2}$ . Hence  $\lambda = \pm \frac{1}{\sqrt{2}}$  and the critical points of  $L$  are  $(\frac{\sqrt{2}}{2}, \frac{\sqrt{2}}{2}, -\frac{1}{\sqrt{2}})$  and  $(-\frac{\sqrt{2}}{2}, -\frac{\sqrt{2}}{2}, \frac{1}{\sqrt{2}})$ . By

evaluating the function  $f$  at the  $xy$ -coordinates of the two critical points, we see that the first one is a maximum and the second one is a minimum.  $\square$

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**Exercises. Seminar 12**

1. Use Lagrange multipliers to find the extrema of the following functions subject to constraints:

(a)  $x^2 + y^2$  subject to  $x - y + 1 = 0$ .

(d)  $x + 2y + 3z$  subject to  $x^2 + y^2 + z^2 = 1$ .

(b)  $(x + y)^2$  subject to  $x^2 + y^2 = 1$ .

(e)  $2x^2 + y^2 + 3z^2$  subject to  $x^2 + y^2 + z^2 = 1$ .

(c)  $x^2 - y^2$  subject to  $x^2 + y^2 = 1$ .

(f)  $x^3 + y^3 + z^3$  subject to  $x^2 + y^2 + z^2 = 1$ .

2. Find the minimum value of  $\frac{1}{2}(x_1^2 + x_2^2 + x_3^2)$  subject to the following constraints:

(a)  $x_1 + x_2 + x_3 = 3$ .

(b)  $x_1 + x_2 + x_3 = 3$  and  $x_1 + 2x_2 + 3x_3 = 12$ .

## ※ Double integrals. Triple integrals

**Rectangular domains.** We start by defining double integrals on very simple domains – rectangles.

Let  $A = [a, b] \times [c, d]$  be a rectangle and consider a partition of it into smaller rectangles

$$\mathcal{P} = \{A_{ij} \mid A_{ij} = [x_{i-1}, x_i] \times [y_{j-1}, y_j], i = \overline{1, m}, j = \overline{1, n}\},$$

where  $a = x_0 < x_1 < \dots < x_m = b$  and  $c = y_0 < y_1 < \dots < y_n = d$ . The norm of the partition is given by  $\|\mathcal{P}\| := \max \left\{ \max_{i=\overline{1, m}} \{x_i - x_{i-1}\}, \max_{j=\overline{1, n}} \{y_j - y_{j-1}\} \right\}$ . We also consider a set of intermediate points  $(x_{ij}^*, y_{ij}^*) \in A_{ij}$  attached to the partition  $\mathcal{P}$ .

**Definition 13.1.** For a function  $f : A \rightarrow \mathbb{R}$  and a partition  $\mathcal{P}$ , the Riemann sum is given by

$$\sigma(f, P) := \sum_{i=1}^m \sum_{j=1}^n f(x_{ij}^*, y_{ij}^*) (x_i - x_{i-1}) (y_j - y_{j-1}).$$

**Remark 13.2.** The Riemann sum collects the volumes of the parallelepipeds defined by the partition  $\mathcal{P}$  (and the intermediate points). In the limit one obtains the volume of the solid below the surface  $z = f(x, y)$ .

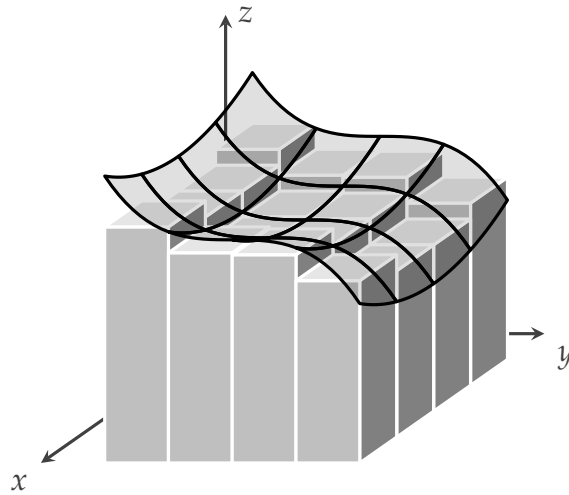


Figure 11: Volume of a solid approximated through parallelepipeds. Riemann sum in 2d.

**Definition 13.3.** Let  $A = [a, b] \times [c, d]$  and  $f : A \rightarrow \mathbb{R}$ . We say that  $f$  is Riemann integrable if there exists  $I \in \mathbb{R}$  s.t. for any partition  $\mathcal{P}$  of  $A$  the Riemann sum  $\sigma(f, P)$  converges to  $I$  as  $\|\mathcal{P}\| \rightarrow 0$ , i.e.

$$\lim_{\|\mathcal{P}\| \rightarrow 0} \sigma(f, P) = I =: \iint_A f(x, y) \, dx \, dy.$$

**Proposition 13.4.** Let  $f, g : A \rightarrow \mathbb{R}$  be Riemann integrable and  $\alpha \in \mathbb{R}$ . Then

- $\iint_A \alpha f(x, y) \, dx \, dy = \alpha \iint_A f(x, y) \, dx \, dy$ .
- $f + g$  is Riemann integrable and

$$\iint_A (f(x, y) + g(x, y)) \, dx \, dy = \iint_A f(x, y) \, dx \, dy + \iint_A g(x, y) \, dx \, dy.$$

- If  $f(x, y) \leq g(x, y)$ ,  $\forall (x, y) \in A$ , then  $\iint_A f(x, y) \, dx \, dy \leq \iint_A g(x, y) \, dx \, dy$ .

**Proposition 13.5.** Let  $A_1, A_2 \subset \mathbb{R}^2$  s.t.  $A = A_1 \cup A_2$  and  $\text{int}A_1 \cap \text{int}A_2 = \emptyset$ . Then

$$\iint_A f(x, y) \, dx \, dy = \iint_{A_1} f(x, y) \, dx \, dy + \iint_{A_2} f(x, y) \, dx \, dy.$$

The next fundamental result indicates *how to compute* double integrals: by iterated integrals. Note that the order of integration does not matter.

**Theorem 13.6** (Fubini). Let  $A = [a, b] \times [c, d]$  and  $f : A \rightarrow \mathbb{R}$  be Riemann integrable. Then

$$\iint_A f(x, y) \, dx \, dy = \int_a^b \int_c^d f(x, y) \, dy \, dx = \int_c^d \int_a^b f(x, y) \, dx \, dy.$$

*Proof.* We will only give a simplified proof, for which we assume that  $f$  is continuous. We consider the iterated integral and we aim to prove that it equals the double integral. Integrating first with respect to  $y$ , we define

$$F(x) := \int_c^d f(x, y) \, dy = \sum_{j=1}^n \int_{y_{j-1}}^{y_j} f(x, y) \, dy,$$

where  $c = y_0 < y_1 < \dots < y_n = d$ . Since  $f$  is continuous we can apply the mean value theorem for each integral

$$\int_{y_{j-1}}^{y_j} f(x, y) \, dy = f(x, y_j^*)(y_j - y_{j-1}), \text{ with } y_j^* \in [y_{j-1}, y_j]$$

and obtain that

$$F(x) = \sum_{j=1}^n f(x, y_j^*)(y_j - y_{j-1}).$$

Considering now the whole iterated integral and using the usual 1d Riemann sum

$$\int_a^b \int_c^d f(x, y) dy dx = \int_a^b F(x) dx = \lim_{n \rightarrow \infty} \sum_{i=1}^n F(x_i^*)(x_i - x_{i-1}),$$

where  $a = x_0 < x_1 < \dots < x_n = b$  and  $x_i^* \in [x_i - x_{i-1}]$ . We have that

$$\begin{aligned} \int_a^b \int_c^d f(x, y) dy dx &= \int_a^b F(x) dx = \lim_{n \rightarrow \infty} \sum_{i=1}^n F(x_i^*)(x_i - x_{i-1}) \\ &= \lim_{n \rightarrow \infty} \sum_{i=1}^n \sum_{j=1}^n f(x_i^*, y_j^*)(y_j - y_{j-1})(x_i - x_{i-1}) \\ &= \iint_A f(x, y) dx dy, \end{aligned}$$

where in the end we used the 2d Riemann sum that converges to the double integral. By a similar reasoning we also have that  $\int_c^d \int_a^b f(x, y) dx dy = \iint_A f(x, y) dx dy$ .  $\square$

**Example 13.7.** Let  $R = [-1, 1] \times [0, 1]$  and consider  $\iint_R (x^2 + y^2) dx dy$ . By Fubini's theorem

$$\iint_R (x^2 + y^2) dx dy = \int_{-1}^1 \int_0^1 (x^2 + y^2) dy dx = \int_{-1}^1 (x^2 + \frac{1}{3}) dx = \frac{4}{3}. \quad \square$$

There is a particular class of functions for which the double integral is very easy to compute, namely separable functions  $f(x, y) = g(x)h(y)$ . In this case the double integral is simply the product of two separate simple integrals.

**Corollary 13.8.** Let  $f : A = [a, b] \times [c, d] \rightarrow \mathbb{R}$  be Riemann integrable. If  $f(x, y) = g(x)h(y)$ , then

$$\iint_A f(x, y) dx dy = \int_a^b g(x) dx \int_c^d h(y) dy.$$

*Proof.* By Fubini's theorem

$$\iint_A g(x)h(y) dx dy = \int_a^b g(x) \underbrace{\int_c^d h(y) dy}_{\text{constant}} dx = \left( \int_a^b g(x) dx \right) \left( \int_c^d h(y) dy \right).$$

$\square$

**More general domains.** Based on the definition of the double integral on rectangles, let us now define the double integral on more general domains. For this let  $D \subset \mathbb{R}^2$  be a bounded set.

**Definition 13.9.** We say that  $f : D \rightarrow \mathbb{R}$  is Riemann integrable on  $D$  if there exists a rectangle  $A \subset \mathbb{R}^2$  s.t.  $D \subseteq A$  and the extension function  $\bar{f} : A \rightarrow \mathbb{R} = \begin{cases} f(x), & \text{if } x \in D \\ 0, & \text{if } x \in A \setminus D \end{cases}$  is Riemann integrable on  $A$ . Then  $\iint_D f(x, y) dx dy = \iint_A \bar{f}(x, y) dx dy$ .

An important class of domains consists of domains that have four sides – two straight and two curves – which can be described by letting one variable run in an interval and bounding the other variable by two functions. Such domains are called *simple* and the double integral can be computed through iterated integrals using Fubini's theorem.

**Definition 13.10.** A set  $D \subset \mathbb{R}^2$  is called

- *simple with respect to the  $y$ -axis* if there exist continuous functions  $\varphi_1, \varphi_2 : [a, b] \rightarrow \mathbb{R}$

$$D = \{(x, y) \in \mathbb{R}^2 \mid a \leq x \leq b, \varphi_1(x) \leq y \leq \varphi_2(x)\}.$$

- *simple with respect to the  $x$ -axis* if there exist continuous functions  $\psi_1, \psi_2 : [c, d] \rightarrow \mathbb{R}$

$$D = \{(x, y) \in \mathbb{R}^2 \mid c \leq y \leq d, \psi_1(y) \leq x \leq \psi_2(y)\}.$$

- *simple* if it is simple with respect to both  $x$ -axis and  $y$ -axis.

Domains that are simple w.r.t to the  $y$ -axis or  $x$ -axis are also called  $y$ -simple or  $x$ -simple.

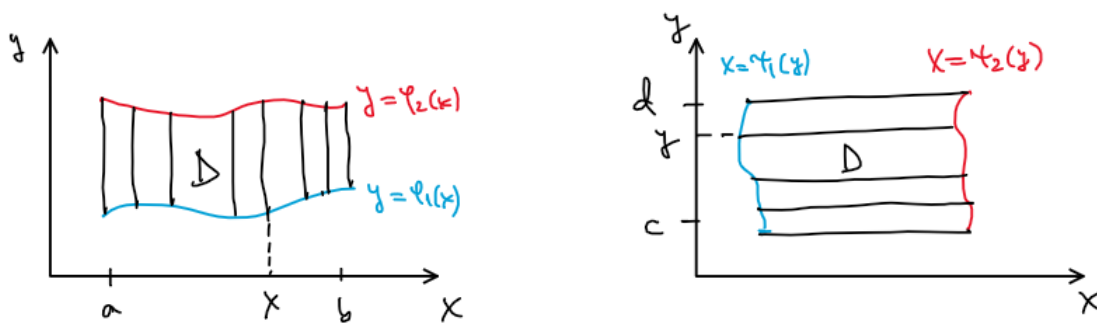


Figure 12: Simple domains.

**Theorem 13.11.** Let  $D \subset \mathbb{R}^2$  be a bounded set and  $f : D \rightarrow \mathbb{R}$  Riemann integrable on  $D$ .

- If  $D$  is  $y$ -simple, then

$$\iint_D f(x, y) \, dx \, dy = \int_a^b \int_{\varphi_1(x)}^{\varphi_2(x)} f(x, y) \, dy \, dx.$$

- If  $D$  is  $x$ -simple, then

$$\iint_D f(x, y) \, dx \, dy = \int_c^d \int_{\psi_1(y)}^{\psi_2(y)} f(x, y) \, dx \, dy.$$

- If  $D$  is simple, then

$$\iint_D f(x, y) \, dx \, dy = \int_a^b \int_{\varphi_1(x)}^{\varphi_2(x)} f(x, y) \, dy \, dx = \int_c^d \int_{\psi_1(y)}^{\psi_2(y)} f(x, y) \, dx \, dy.$$

Note that simple domains allow changing the order of integration.

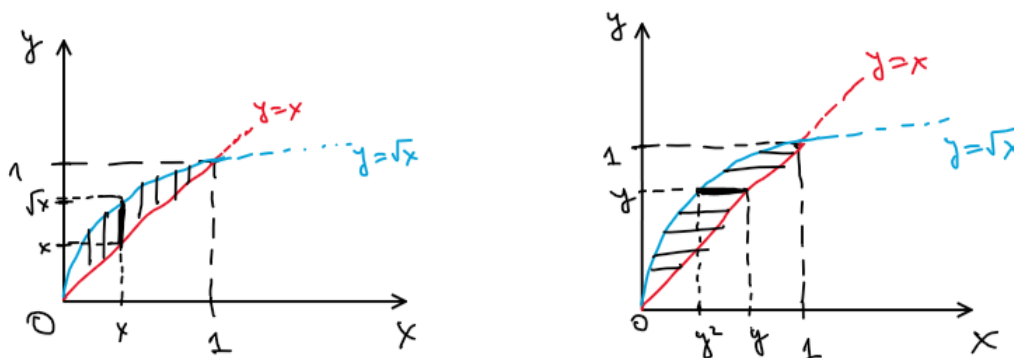


Figure 13: Simple domain in Theorem 13.12. Changing the order of integration.

**Example 13.12.** Let  $D = \{(x, y) \in \mathbb{R}^2 \mid 0 \leq x \leq 1, x \leq y \leq \sqrt{x}\}$ . Compute  $\iint_D \frac{e^y}{y} \, dx \, dy$ .

Notice that the domain  $D$  is  $y$ -simple and we could try to compute the integral in an iterated way (Fubini) – first w.r.t to  $y$ , then w.r.t  $x$ . However, this order of integration puts us in the situation of finding a primitive of  $\frac{e^y}{y}$  and we get stuck – there is no *elementary* primitive for this function. So let us try to change the order of integration. For this, we write the domain  $D$  as  $x$ -simple:  $D = \{(x, y) \in \mathbb{R}^2 \mid 0 \leq y \leq 1, y^2 \leq x \leq y\}$ . Then

$$\iint_D \frac{e^y}{y} \, dx \, dy = \int_0^1 \int_{y^2}^y \frac{e^y}{y} \, dx \, dy = \int_0^1 \frac{e^y}{y} (y - y^2) \, dy = \int_0^1 (e^y - ye^y) \, dy = e - 2.$$

Changing the order of integration worked – the new integral is much easier to compute!  $\square$



### Change of variables in double integrals

**Linear algebra recap.** Let us start by recalling some basic things from linear algebra and the geometry of linear transformations. This will provide the fundamental tools for doing a change of variables (or change of coordinates).

Let  $A$  be a  $2 \times 2$  matrix and consider a linear transformation  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ ,  $T(x) = Ax$ . A linear transformation maps parallelograms into parallelograms. Indeed, consider a parallelogram with edges  $u$  and  $v$ . This is given by linear combinations  $\alpha u + \beta v$  with  $\alpha, \beta \in [0, 1]$ . Applying the linear transformation  $T$ , we have that  $A(\alpha u + \beta v) = \alpha Au + \beta Av$ . This represents a parallelogram with edges  $Au$  and  $Av$ . So parallelograms  $\mapsto$  parallelograms.

How does a linear transformation change area (Youtube video 3Blue1Brown)? It scales it with the factor  $|\det(A)|$ .

The unit square with area 1 is mapped into a parallelogram with area  $|\det(A)|$ . For a domain  $D^*$  mapped into  $D = T(D^*)$ ,  $\text{Area}(T(D^*)) = |\det(A)| \text{Area}(D^*)$ .

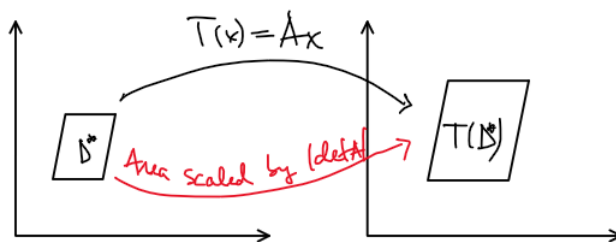


Figure 14: Linear transformation. The determinant gives the area scaling factor.

**Change of variables.** Let  $D \subseteq \mathbb{R}^2$  be a domain in the  $xy$  plane. We want to make a change of variables from  $xy$  to  $uv$ , writing  $x = x(u, v)$  and  $y = y(u, v)$ . Consider a domain  $D^*$  in the  $uv$  plane and a map  $T : D^* \rightarrow D$  bijective and of class  $C^1$  (differentiable and with continuous derivatives) such that  $(x, y) = T(u, v)$ . The question is: how are  $\text{Area}(D)$  and  $\text{Area}(D^*)$  related as we change coordinates from  $xy$  to  $uv$ ?

$$\iint_D 1 \, dx \, dy = \iint_{D^*} \dots \, du \, dv.$$

As discussed above, if  $T$  is a linear map with  $T(x) = Ax$  then  $\text{Area}(D) = |\det(A)| \text{Area}(D^*)$ , which can be written as

$$\iint_D 1 \, dx \, dy = \iint_{D^*} |\det(A)| \, du \, dv.$$

If  $T$  is not linear, we can take its linear approximation (the differential) around a point

and for  $u = u_0 + \Delta u$  and  $v = v_0 + \Delta v$  use that

$$T(u, v) \approx T(u_0, v_0) + T'(u_0, v_0) \begin{bmatrix} \Delta u \\ \Delta v \end{bmatrix},$$

where  $T'$  is the differential of  $T$ , namely the Jacobian matrix

$$T' = J = \begin{bmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{bmatrix}.$$

Consider a partition of the domain  $D^*$  in the  $uv$  plane into small rectangles  $R^*$  with sides  $\Delta u, \Delta v$ .  $\text{Area}(R^*) = \Delta u \Delta v$ . The image  $R = T(R^*)$  of a rectangle  $R^*$  can be a complicated domain with curved boundaries as shown in the figure below – we will aim to *approximate* its area. By taking the linear approximation of  $T$ , we approximate the region  $R = T(R^*)$  with a parallelogram  $T'(R^*)$  given by the linear transformation  $T'$ . The sides of the parallelogram are  $T' \begin{bmatrix} \Delta u \\ 0 \end{bmatrix} = \Delta u \begin{bmatrix} \frac{\partial x}{\partial u} \\ \frac{\partial y}{\partial u} \end{bmatrix}$  and  $T' \begin{bmatrix} 0 \\ \Delta v \end{bmatrix} = \Delta v \begin{bmatrix} \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial v} \end{bmatrix}$ . More importantly, the area of the parallelogram  $T'(R^*)$  approximates the area of the region  $R = T(R^*)$ , namely

$$\text{Area}(R) \approx \text{Area}(T'(R^*)) = |\det(J)| \text{Area}(R^*) = |\det(J)| \Delta u \Delta v,$$

where  $\det(J)$  is the determinant of the Jacobian matrix  $J$ . As the size of the rectangle  $R^*$  goes to zero, in the limit we have that

$$\text{Area}(D) = \iint_D 1 \, dx \, dy = \iint_{D^*} |\det(J)| \, du \, dv. \quad (2)$$

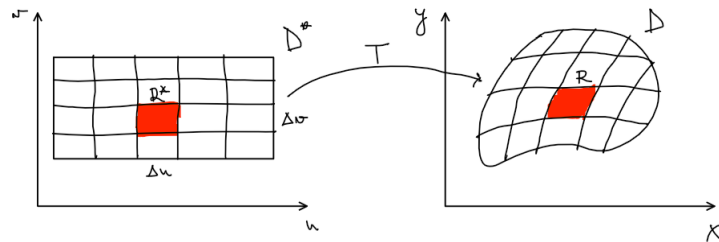


Figure 15: Change of variables.

**Theorem 13.13** (Change of variables). Let  $D, D^* \subseteq \mathbb{R}^2$  and  $T : D^* \rightarrow D$  bijective and of class  $C^1$  with the Jacobian  $J$ . Then for any Riemann integrable  $f : D \rightarrow \mathbb{R}$ , we have that

$$\iint_D f(x, y) \, dx \, dy = \iint_{D^*} f(x(u, v), y(u, v)) |\det(J)| \, du \, dv.$$

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**Exercises. Seminar 12**

1. Compute the following double integrals:

(a)  $\iint_R \cos x \sin y \, dx \, dy$ , where  $R = [0, \pi/2] \times [0, \pi/2]$ .

(b)  $\iint_R \frac{1}{(x+y)^2} \, dx \, dy$  and  $\iint_R y e^{xy} \, dx \, dy$ , where  $R = [1, 2] \times [0, 1]$ .

2. Let  $D \subseteq \mathbb{R}^2$  be the subset bounded by the parabola  $y = x^2$  and the lines  $x = 2$  and  $y = 0$ .

(a) Express  $D$  as a simple set, first w.r.t. the  $y$ -axis and then w.r.t. the  $x$ -axis.

(b) Compute  $\iint_D xy \, dx \, dy$  in two ways, by changing the order of integration.

3. By changing the order of integration, compute the following double integrals:

(a)  $\int_0^1 \int_y^1 \sin(x^2) \, dx \, dy$ .      (b)  $\int_0^1 \int_x^1 e^{y^2} \, dy \, dx$ .

4. Compute the following integrals by a change of variables:

(a)  $\iint_D e^{\frac{x-y}{x+y}} \, dx \, dy$ ,  $D = \{(x, y) \in \mathbb{R}^2 \mid x \geq 0, y \geq 0, x+y \leq 1\}$ . Use  $u = x-y$ ,  $v = x+y$ .

(b)  $\iint_D \left( \frac{x-y}{x+y+2} \right)^2 \, dx \, dy$ ,  $D = \{(x, y) \in \mathbb{R}^2 \mid |x| + |y| \leq 1\}$ . Use  $u = x-y$ ,  $v = x+y$ .

(c)  $\iint_D \frac{y^2}{x} \, dx \, dy$ ,  $D$  is bounded by  $x = 1-y^2$  and  $x = 3(1-y^2)$ . Use  $y = u$ ,  $x = (1-u^2)v$ .

**Polar coordinates.** One of the most important changes of coordinates is given by

$$\begin{aligned}x &= r \cos \theta, \\y &= r \sin \theta,\end{aligned}$$

having the Jacobian matrix

$$J = \begin{bmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{bmatrix} = \begin{bmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{bmatrix}$$

and the Jacobian determinant  $|\det(J)| = |r \cos^2 \theta + r \sin^2 \theta| = r$ , that is

$$|\det(J)| = r.$$

Changing from a domain  $D$  in the  $xy$ plane to a domain  $D^*$  in the  $r\theta$  plane, we have that

$$\iint_D f(x, y) \, dx dy = \iint_{D^*} f(r \cos \theta, r \sin \theta) \, r \, dr d\theta. \quad (3)$$

**Example 13.14.** Let  $D = \{(x, y) \in \mathbb{R}^2 \mid x^2 + y^2 \leq R^2\}$ . Compute  $\iint_D e^{x^2+y^2} \, dx dy$ .

Using the polar coordinates  $(r, \theta)$  we write  $x = r \cos \theta, y = r \sin \theta$  and  $(x, y) \in D$  for  $r \in [0, R], \theta \in [0, 2\pi]$ . This gives a rectangle  $D^* = [0, R] \times [0, 2\pi]$  in the  $r\theta$  plane.

$$\iint_D e^{x^2+y^2} \, dx dy = \int_0^R \int_0^{2\pi} e^{r^2} r \, d\theta dr = 2\pi \int_0^R r e^{r^2} \, dr = \pi(e^{R^2} - 1).$$

**Example 13.15** (Gaussian integral  $\int_{-\infty}^{\infty} e^{-x^2} \, dx$ ). To compute this integral it is easier to consider a double integral. Let  $D = \{(x, y) \in \mathbb{R}^2 \mid x^2 + y^2 \leq R^2\}$ .

$$\iint_D e^{-(x^2+y^2)} \, dx dy = \int_0^R \int_0^{2\pi} e^{-r^2} r \, d\theta dr = 2\pi \int_0^R r e^{-r^2} \, dr = \pi(-e^{-R^2} + 1).$$

Letting now  $R \rightarrow \infty$  we obtain that

$$\pi = \iint_{\mathbb{R}^2} e^{-(x^2+y^2)} \, dx dy = \int_{-\infty}^{\infty} e^{-x^2} \, dx \int_{-\infty}^{\infty} e^{-y^2} \, dy = \left( \int_{-\infty}^{\infty} e^{-x^2} \, dx \right)^2,$$

hence we obtain the Gaussian integral

$$\int_{-\infty}^{\infty} e^{-x^2} \, dx = \sqrt{\pi}.$$

**Triple integrals** The triple Riemann integral can be defined using Riemann sums, in a similar way to the double integral (with parallelepipeds instead of rectangles for partitioning the domain). Everything that we have seen about double integrals can be generalized to triple integrals. We will simply state the basic results.

Let  $A = [a_1, a_2] \times [b_1, b_2] \times [c_1, c_2]$  and  $f : A \rightarrow \mathbb{R}$  be Riemann integrable. Then

$$\begin{aligned} \iiint_A f(x, y, z) \, dx \, dy \, dz &= \int_{a_1}^{a_2} \int_{b_1}^{b_2} \int_{c_1}^{c_2} f(x, y, z) \, dz \, dy \, dx \\ &= \int_{c_1}^{c_2} \int_{b_1}^{b_2} \int_{a_1}^{a_2} f(x, y, z) \, dx \, dy \, dz \\ &= \dots \text{(all possible orders)}. \end{aligned}$$

If  $D = \{(x, y, z) \in \mathbb{R}^3 \mid a \leq x \leq b, \varphi_1(x) \leq y \leq \varphi_2(x), \psi_1(x, y) \leq z \leq \psi_2(x, y)\}$ , then

$$\iiint_D f(x, y, z) \, dx \, dy \, dz = \int_a^b \int_{\varphi_1(x)}^{\varphi_2(x)} \int_{\psi_1(x, y)}^{\psi_2(x, y)} f(x, y, z) \, dz \, dy \, dx.$$

In symmetric elementary regions the order of integration can be changed.

When changing variables in triple integrals, we should now have in mind that  $3 \times 3$  matrices map parallelepipeds to parallelepipeds with the *volume* scaling with the determinant (absolute value). Let  $D, D^* \subseteq \mathbb{R}^3$  and  $T : D^* \rightarrow D$  bijective and of class  $C^1$  with the Jacobian  $J$ . Then for any Riemann integrable  $f : D \rightarrow \mathbb{R}$ , we have that

$$\iiint_D f(x, y, z) \, dx \, dy \, dz = \iiint_{D^*} f(u, v, w) |\det(J)| \, du \, dv \, dw,$$

with the Jacobian matrix

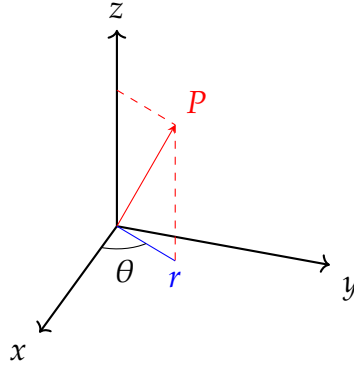
$$J = \begin{bmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} & \frac{\partial x}{\partial w} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} & \frac{\partial y}{\partial w} \\ \frac{\partial z}{\partial u} & \frac{\partial z}{\partial v} & \frac{\partial z}{\partial w} \end{bmatrix}$$

**Cylindrical coordinates.**

$$x = r \cos \theta,$$

$$y = r \sin \theta,$$

$$z = z,$$

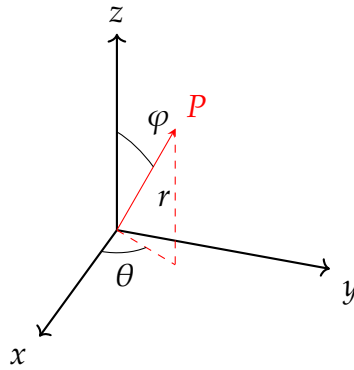


with radius  $r > 0$  and angle  $\theta$  (as in polar coordinates), together with the vertical component  $z$ . In this case the Jacobian matrix is given by

$$J = \begin{bmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} & \frac{\partial x}{\partial z} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} & \frac{\partial y}{\partial z} \\ \frac{\partial z}{\partial r} & \frac{\partial z}{\partial \theta} & \frac{\partial z}{\partial w} \end{bmatrix} = \begin{bmatrix} \cos \theta & -r \sin \theta & 0 \\ \sin \theta & r \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

and the Jacobian determinant is  $|\det(J)| = |r \cos^2 \theta + r \sin^2 \theta| = r$ .

**Spherical coordinates.**



$$x = (r \sin \varphi) \cos \theta,$$

$$y = (r \sin \varphi) \sin \theta,$$

$$z = r \cos \varphi,$$

with radius  $r$ , azimuthal angle  $\theta$ , and polar angle  $\varphi$ . In this case the Jacobian matrix is given by

$$J = \begin{bmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} & \frac{\partial x}{\partial \varphi} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} & \frac{\partial y}{\partial \varphi} \\ \frac{\partial z}{\partial r} & \frac{\partial z}{\partial \theta} & \frac{\partial z}{\partial \varphi} \end{bmatrix} = \begin{bmatrix} \sin \varphi \cos \theta & -r \sin \varphi \sin \theta & r \cos \varphi \cos \theta \\ \sin \varphi \sin \theta & r \sin \varphi \cos \theta & r \cos \varphi \sin \theta \\ \cos \varphi & 0 & -r \sin \varphi \end{bmatrix}$$

and the Jacobian determinant is

$$|\det(J)| = r^2 \sin \varphi.$$

The ball  $B(0, R)$  is given by  $r \in [0, R]$ ,  $\theta \in [0, 2\pi]$ ,  $\varphi \in [0, \pi]$  in spherical coordinates.

**Example 13.16.** The volume of the ball of radius  $R$  is given by

$$\iiint_{B(0,R)} 1 \, dx \, dy \, dz = \int_0^R \int_0^{2\pi} \int_0^\pi r^2 \sin \varphi \, d\varphi \, d\theta \, dr = \frac{R^3}{3} 2\pi \underbrace{\int_0^\pi \sin \varphi \, d\varphi}_{=2} = \frac{4\pi R^3}{3}.$$

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