JKU: Mathematics for Al 1

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$\S 1.5$: An introduction to proof

Goldbach Conjecture:

An even natural number is equal to either 2 or the sum of two prime numbers.

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• First steps to understand the statement:

1. What are the **definitions** of the concepts that are used?

A natural number is an element of $\mathbb{N} := \{1, 2, 3, \ldots\}$.

The even natural numbers are $\{2n : n \in \mathbb{N}\}$.

A prime number in $\mathbb{N} \setminus \{1\}$ is not a product of two elements in $\mathbb{N} \setminus \{1\}$.

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2. Write down some **examples** in order to obtain intuition.

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$$8 = 3 + 5$$

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• The conjecture has been computationally verified for $n \le 4 \cdot 10^{18}$. This is however *evidence* for the conjecture but not a *proof*!

Some common structure in proofs

• Recall that $A \iff B$ is defined as

$$(A \Longrightarrow B) \land (A \Longleftarrow B).$$

 Recall that A ⇒ B is logically equivalent to its contrapositive statement:

$$\neg B \Longrightarrow \neg A$$
.

 A Lemma is usually an intermediate result for some main result called Theorem.

- We call $n \in \mathbb{N}$ even if $n \in \{2k : k \in \mathbb{N}\}$ and odd otherwise.
- **Lemma.** Let $n \in \mathbb{N}$. Then n is even $\iff n^2$ is even.

- We call $n \in \mathbb{N}$ even if $n \in \{2k : k \in \mathbb{N}\}$ and odd otherwise.
- **Lemma.** Let $n \in \mathbb{N}$. Then n is even $\iff n^2$ is even.

Proof.

First, we prove the \Longrightarrow direction:

$$n$$
 is even $\implies n = 2k$ for some $k \in \mathbb{N}$
 $\implies n^2 = (2k)^2 = 4k^2 = 2(2k)^2 \implies n^2$ is even.

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Next, we prove the \Leftarrow direction by proving its equivalent contrapositive[†] statement: n is odd $\Longrightarrow n^2$ is odd.

$$n ext{ is odd} \implies n = 2k + 1 ext{ for some } k \in \mathbb{N}_0$$

$$\implies n^2 = (2k + 1)^2 = 4k^2 + 4k + 1 = 2(2k^2 + 2k) + 1 \implies n^2 ext{ is odd.}$$

Recall.[†] $(A \Longrightarrow B) \iff (\neg B \Longrightarrow \neg A).$

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Recall.[†]
$$(A \Longrightarrow B) \iff (\neg B \Longrightarrow \neg A)$$
.

We showed both the \Longrightarrow and \Longleftarrow directions and therefore concluded the proof of the lemma.

Proof by definition: Exercise

• Exercise.

Prove that the sum of two odd natural numbers is even.

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- First recall the definitions! We call f invertible iff there exists $f^{-1} \colon N \to M$ such that $\forall x \in M \colon f^{-1}(f(x)) = x$ and $\forall y \in N \colon f(f^{-1}(y)) = y$.

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• Try easy but complicated enough examples for understanding.

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Try easy but complicated enough examples for understanding. f: {1,2} → {a,b}, f(1) := a and f(2) := b.
(f is bijective, f⁻¹(a) = 1 and f⁻¹(b) = 2).
g: {1,2} → {a,b,c}, g(1) := a and g(2) := b.
(g is not surjective and g⁻¹(c) is not defined).
h: {1,2} → {a,b} and h(1) := h(2) := a.
(h is not injective and h⁻¹(b) is not defined).

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- Definitions:

$$\forall x \in M \colon f^{-1}(f(x)) = x \text{ and } \forall y \in N \colon f(f^{-1}(y)) = y.$$

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 invertible iff there exists $f^{-1}: N \to M$ such that $\forall x \in M: f^{-1}(f(x)) = x$ and $\forall y \in N: f(f^{-1}(y)) = y$.

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Proof. Let us first prove the ⇒ direction:

$$f \text{ is invertible } \Longrightarrow \forall y \in N \colon f(f^{-1}(y)) = y$$

$$\Longrightarrow \forall y \in N, \exists x \in M \colon f(x) = y \Longrightarrow f \text{ is surjective.}$$

$$f \text{ is invertible } \Longrightarrow$$

$$(\forall x_1, x_2 \in M \colon f(x_1) = f(x_2) \Rightarrow x_1 = f^{-1}(f(x_1)) = f^{-1}(f(x_2)) = x_2)$$

$$\Longrightarrow f \text{ is injective.}$$

Since f is surjective and injective, we conclude that f is bijective.

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The function f is bijective iff $\forall y \in N, \exists ! x \in M : f(x) = y$.

Proof continued. Next, we prove the ← direction and thus we assume that the function f is bijective.

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Proof continued. Next, we prove the ← direction and thus we assume that the function f is bijective.

Let $g: N \to M$ be the function that sends $y \in N$ to the unique element $x \in M$ such that f(x) = y.

We observe that

$$\forall x \in M \colon g(f(x)) = g(y) = x \text{ and }$$

$$\forall y \in N \colon f(g(y)) = f(x) = y.$$

We conclude that f is invertible with $f^{-1} = g$.

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We conclude that f is invertible with $f^{-1} = g$.

As we showed both the \Longrightarrow and \Longleftrightarrow directions, we concluded the proof of the theorem. 9/13

More exercises for proof by definition.

• Exercise. Let $f: M \to N$ be a function. Let $\mathrm{id}_M \colon M \to M$ and $\mathrm{id}_N \colon N \to N$ be the identity functions. Prove that $f \circ \mathrm{id}_M = f$ and $\mathrm{id}_N \circ f = f$.

More exercises for proof by definition.

- Exercise. Let $f: M \to N$ be a function. Let $\mathrm{id}_M \colon M \to M$ and $\mathrm{id}_N \colon N \to N$ be the identity functions. Prove that $f \circ \mathrm{id}_M = f$ and $\mathrm{id}_N \circ f = f$.
- Exercise. Proof that the inverse f^{-1} of a function $f: M \to N$ is unique.

Some common proof techniques

• When proving the proposition

$$A \iff B \iff C$$

it is sufficient to prove

$$A \Longrightarrow B \Longrightarrow C \Longrightarrow A$$
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Example. See Theorem 1.26 in lecture notes.

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Example. See Theorem 1.26 in lecture notes.

• When proving that two sets X and Y are equal:

Step 1. Take an arbitrary (but fixed) element $x \in X$ and show that $x \in Y$. Step 2. Take an arbitrary (but fixed) element $y \in Y$ and show that $y \in X$.

Exercise. Let $f: M \to N$ be an invertible function and $S \subset N$. Prove the following proposition:

$$\{f^{-1}(y): y \in S\} = \{x \in M: f(x) \in S\}.$$

Hint.
$$x \in \{f^{-1}(y) : y \in S\} \Longrightarrow \exists y \in S : x = f^{-1}(y) \Longrightarrow \dots$$

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Assume by contradiction that there exists $x \in \mathbb{Q}$ such that $x^2 = 2$. This implies that $x = \frac{m}{n}$ with either m or n being odd (otherwise we divide both the numerator and denominator by two).

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Assume by contradiction that there exists $x \in \mathbb{Q}$ such that $x^2 = 2$. This implies that $x = \frac{m}{n}$ with either m or n being odd (otherwise we divide both the numerator and denominator by two). This leads to the following sequence of implications:

$$2 = x^2 = \frac{m^2}{n^2} \Longrightarrow m^2 = 2n^2 \Longrightarrow m^2$$
 is even
previous lemma m is even
 $\Longrightarrow \exists k \in \mathbb{Z} : 2n^2 = m^2 = (2k)^2 = 4k^2 \Longrightarrow n^2$ is even
previous lemma n is even.

Theorem.

There does not exist $x \in \mathbb{Q}$ such that $x^2 = 2$.

• Proof.

Assume by contradiction that there exists $x \in \mathbb{Q}$ such that $x^2 = 2$. This implies that $x = \frac{m}{n}$ with either m or n being odd (otherwise we divide both the numerator and denominator by two). This leads to the following sequence of implications:

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 is even
$$\Longrightarrow^{\text{previous lemma}} m \text{ is even}$$
$$\Longrightarrow \exists k \in \mathbb{Z} \colon 2n^2 = m^2 = (2k)^2 = 4k^2 \Longrightarrow n^2 \text{ is even}$$

$$\Longrightarrow^{\text{previous lemma}} n \text{ is even}.$$

We arrived at a contradiction as n must be odd.

Hence, we conclude the proof of the theorem.

Proof by induction

- Recall that a predicate P(n) is a proposition depending on a variable n.
- Induction basis:

P(1) is true.

Induction step:

If P(n) is true for some $n \in \mathbb{N}$, then P(n+1) is true.

• To prove $(\forall n \in \mathbb{N}: P(n) \text{ is true})$ it is sufficient to prove both the induction basis and induction step:

$$P(1)$$
 is true $\Longrightarrow P(2)$ is true $\Longrightarrow P(3)$ is true $\Longrightarrow \dots$

• Induction hypothesis:

P(n) is true for some $n \in \mathbb{N}$.

Proof by induction

- Recall. Induction basis: P(1) is true.
 Induction step: If P(n) is true for some n∈ N, then P(n+1) is true.
 Induction hypothesis: P(n) is true for some n∈ N.
- **Theorem**. For all $n \in \mathbb{N}$, we have

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

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- **Theorem.** For all $n \in \mathbb{N}$, we have

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Proof. Induction basis:

$$\sum_{k=1}^{1} k = 1 = \frac{1(1+1)}{2}.$$

Induction step: We observe that

$$\sum_{k=1}^{n+1} k = \left(\sum_{k=1}^{n} k\right) + n + 1.$$

By the induction hypothesis, we have $\sum_{k=1}^{n} k = \frac{n(n+1)}{2}$ and thus

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

We concluded the proof as we showed both the induction basis and induction step. (Recall previous lecture.)

14/134

Let k ≥ 1 be an integer.
 A predictate P(n) is true for all n ∈ N such that n ≥ k if the following holds:
 Induction basis: P(k) is true, and
 Induction step: If P(i) is true for all k ≤ i ≤ n, then P(n + 1) is true.

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 Induction basis: P(k) is true, and
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- **Theorem.** For all $n \in \mathbb{N}$ such that $n \geq 2$ there exists $i \in \mathbb{N}$ and prime numbers p_1, \ldots, p_i such that $n = p_1 \cdot p_2 \cdots p_i$.

• Let $k \ge 1$ be an integer.

A predictate P(n) is true for all $n \in \mathbb{N}$ such that $n \ge k$ if the following holds: **Induction basis:** P(k) is true, and **Induction step:** If P(i) is true for all $k \le i \le n$, then P(n+1) is true.

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Proof. Induction basis:

n=2 is a prime number and thus the proposition is true with i=1.

Induction step:

We make a case distinction on whether n+1 is prime.

case 1: If n + 1 is prime, then the proposition is true with i = 1.

• Let $k \ge 1$ be an integer.

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• **Theorem.** For all $n \in \mathbb{N}$ such that $n \geq 2$ there exists $i \in \mathbb{N}$ and prime numbers p_1, \ldots, p_i such that $n = p_1 \cdot p_2 \cdots p_i$.

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Induction step:

We make a case distinction on whether n+1 is prime.

case 1: If n + 1 is prime, then the proposition is true with i = 1.

case 2: Suppose that n+1 is not prime. In this case, $n+1=a\cdot b$ for some $a,b\in\mathbb{N}$ such that $2\leq a,b\leq n$. By the induction hypothesis there exists finitely many prime numbers p_1,\ldots,p_i and q_1,\ldots,q_j such that

$$a = p_1 \cdots p_i$$
 and $b = q_1 \cdots q_i$.

Thus $n+1=a\cdot b$ is indeed a product of finitely many prime numbers. We conclude the proof as we considered both cases for the induction step.

15/134

Proof by strong induction: exercise.

Exercise.

Prove that $2^n > 2n + 1$ for all natural numbers $n \ge 3$. **Hint.** use induction.

 $\S 1.6$: Bounded sets, infimum and supremum

• Suppose that $A \subset \mathbb{R}$.

- Suppose that $A \subset \mathbb{R}$.
- A upper bound for A is (if it exists) defined as $c \in \mathbb{R}$ such that $a \le c$ for all $a \in A$.

Examples. The closed interval [-1,3] has 3 and 4.1 as upper bounds. We say that [-1,3] is bounded from above.

Similarly, the half open interval [-2,4) is bounded from above.

The set $\mathbb{N} = \{1, 2, \ldots\}$ is not bounded from above.

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The set $\mathbb{N} = \{1, 2, \ldots\}$ is not bounded from above.

• The maximum max A for A is (if it exists) defined as the element $c \in A$ such that $a \le c$ for all $a \in A$.

Examples. $\max\{1,2,5\} = 5$, $\max[-1,3] = 3$ and $\max(-1,5] = 5$. However, the maximum of the open interval (-1,5) does not exist.

- Suppose that $A \subset \mathbb{R}$.
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The set $\mathbb{N}=\{1,2,\ldots\}$ is not bounded from above.

- The maximum max A for A is (if it exists) defined as the element c ∈ A such that a ≤ c for all a ∈ A.
 Examples. max{1,2,5} = 5, max[-1,3] = 3 and max(-1,5] = 5.
- However, the maximum of the open interval (-1,5) does not exist.
- The supremum sup A for A is (if it exists) defined as $c \in \mathbb{R}$ such that $a \le c$ for all $a \in A$ and for all $\epsilon > 0$ there exists $b \in A$ such that $b > c \epsilon$.

Examples. $\sup(-1,5) = 5$ and $\sup\{1,2,5\} = \max\{1,2,5\} = 5$. The supremum is a least upper bound. **18/**1

• Suppose that $A \subset \mathbb{R}$.

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- A lower bound for A is (if it exists) defined as $c \in \mathbb{R}$ such that $c \le a$ for all $a \in A$.

Examples.

The closed interval [-1,3] has -1, -2 and -1.1 as lower bounds.

We say that [-1,3] is bounded from below.

The set $\mathbb{Z} = \{\ldots, -2, -1, 0, 1, 2, \ldots\}$ is not bounded from below.

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The set $\mathbb{Z} = \{\ldots, -2, -1, 0, 1, 2, \ldots\}$ is not bounded from below.

• The minimum min A for A is (if it exists) defined as the element $c \in A$ such that c < a for all $a \in A$.

Examples. $\min\{1,2,5\} = 1$, $\min[-1,3] = -1$ and $\min[-1,5) = -1$. However, the minimum of the open interval (-1,5) does not exist.

- Suppose that $A \subset \mathbb{R}$.
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The closed interval [-1,3] has -1, -2 and -1.1 as lower bounds. We say that [-1,3] is bounded from below.

The set $\mathbb{Z} = \{\ldots, -2, -1, 0, 1, 2, \ldots\}$ is not bounded from below.

- The minimum min A for A is (if it exists) defined as the element $c \in A$ such that $c \le a$ for all $a \in A$.
 - **Examples.** $\min\{1,2,5\}=1$, $\min[-1,3]=-1$ and $\min[-1,5)=-1$. However, the minimum of the open interval (-1,5) does not exist.
- The infimum inf A for A is (if it exists) defined as c∈ ℝ such that c ≤ a for all a ∈ A and for all ε > 0 there exists b ∈ A such that c + ε > b.
 Examples. inf(-1,5) = -1 and inf{1,2,5} = min{1,2,5} = 1. The infimum is a greatest lower bound.

Completeness Axiom

- Completeness Axiom for the real numbers ℝ.
 If A ⊂ ℝ is a non-empty set that is bounded from above, then sup A exists.
- This axiom does not hold for the rational numbers Q. For example,

$$\sup\left\{x\in\mathbb{Q}:x\leq\sqrt{2}\right\}=\sqrt{2}\notin\mathbb{Q}.$$

Indeed, for all $q\in \mathbb{Q}$ such that $\sqrt{2}< q$, there exists $r\in \mathbb{Q}$ such that $\sqrt{2}< r< q.$

Completeness Axiom

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 If A ⊂ ℝ is a non-empty set that is bounded from above, then sup A exists.
- ullet This axiom does not hold for the rational numbers $\mathbb Q.$ For example,

$$\sup\left\{x\in\mathbb{Q}:x\leq\sqrt{2}\right\}=\sqrt{2}\notin\mathbb{Q}.$$

Indeed, for all $q \in \mathbb{Q}$ such that $\sqrt{2} < q$, there exists $r \in \mathbb{Q}$ such that

$$\sqrt{2} < r < q.$$

• Exercise. Let $A \subset \mathbb{R}$ and $-A := \{-a : a \in A\}$. Prove that if A is a non-empty set that is bounded from above, then $\sup A$ exists and $\sup A = -\inf(-A)$.

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then inf A exists.

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- Exercise. Let A ⊂ R and -A := {-a : a ∈ A}.
 Prove that if A is a non-empty set that is bounded from above, then sup A exists and sup A = -inf(-A).
- Exercise. Prove that if $A \subset \mathbb{R}$ is a non-empty set that is bounded from below,
 - Hint. Use the completeness axiom and the previous exercise 20/1

• Theorem.

For all $\epsilon > 0$ there exists $n \in \mathbb{N}$ such that $\frac{1}{n} < \epsilon$.

Proof. See lecture notes.

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Proof. We observe that 0 is a lower bound.

Now suppose by contradiction that there exists $\epsilon>0$ such that

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• Proposition. $\max\left\{\frac{1}{n}:n\in\mathbb{N}\right\}=1.$

Proof. Follows from the fact that $1 \in \left\{ \frac{1}{n} : n \in \mathbb{N} \right\}$ and $\frac{1}{n} \leq 1$ for all $n \in \mathbb{N}$.

Exercise

• Exercise. Let

$$A = \left\{ \frac{1}{n^2 - n - 3} : n \in \mathbb{N} \right\}.$$

Compute, if it exists, the following quantities:

 $\inf A$, $\sup A$, $\min A$, $\max A$.

 $\S 1.7$: Some basic combinatorial objects, identities and inequalities

Factorial

• The factorial of $n \in \mathbb{N}$ is defined as

$$n! := 1 \cdot 2 \cdot \cdot \cdot n$$
.

We define 0! := 1.

- **Examples.** $3! = 1 \cdot 2 \cdot 3 = 6$ and $4! = 1 \cdot 2 \cdot 3 \cdot 4 = 24$.
- The factorial n! is equal to the number of ways to put n distinct balls into n distinct boxes.
- **Problem.** In how many ways can we put 3 distinct balls into 3 distinct boxes?

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• Exercise. In how many ways can you write a 9-digit number that contains each digit in $\{1, \ldots, 9\}$?

Binomial coefficient

• For $n, k \in \mathbb{N}_0$ the binomial coefficient is defined as

$$\binom{n}{k} := \frac{n!}{k! \cdot (n-k)!}.$$

- The binomial coefficient $\binom{n}{k}$ shows in how many ways we can put k identical balls into n distinct boxes such that each box receives at most one ball.
- With a k-element set we mean a set that consists of k elements.
- **Problem.**In how many ways can we choose a 2-element subset from {1, 2, 3}?

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In how many ways can we choose a 2-element subset from $\{1,2,3\}$?

Answer. There are three boxes (labeled 1, 2 or 3) and there are two identical balls: $\binom{3}{2} = \frac{3!}{2! \cdot 1!} = 3$.

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• More generally, the number of k-element subsets of an n-element set is equal to $\binom{n}{k}$.

• **Lemma.** For all $n, k \in \mathbb{N}$ such that $k \leq n-1$, we have

$$\binom{n+1}{k+1} = \binom{n}{k} + \binom{n}{k+1}.$$

• Proof. Let

A:= set of (k+1)-element subsets of $S:=\{1,\ldots,n+1\}$,

 $B := \text{set of } k\text{-element subsets of } \{1, \dots, n\},$ $C := \text{set of } (k+1)\text{-element subsets of } \{1, \dots, n\},$

and recall that

$$|A| = {n+1 \choose k+1}, \qquad |B| = {n \choose k}, \qquad |C| = {n \choose k+1}.$$

We observe that

$$|A| = |U| + |V|,$$

where

U := set of (k+1)-element subsets of S containing n+1,

V := set of (k+1)-element subsets of S not containing n+1.

- The function $B \to U$ that sends $\{b_1, \ldots, b_k\}$ to $\{b_1, \ldots, b_k, n+1\}$ is a bijection and thus |B| = |U|. (See exercise on next slide.)
- We observe that C = V and thus |C| = |V|.
- We concluded the proof since |A| = |B| + |C|.

Exercises related to the lemma

- **Exercise.** Suppose that $n, k \in \mathbb{N}$.
 - Let B be the set of k-element subsets of $\{1, \ldots, n\}$. Let U be the set of (k+1)-element subsets of $\{1, \ldots, n+1\}$ containing the element n+1.
 - Determine the sets B and U and construct a bijection $B \to U$ under the assumption that n = 4 and k = 2.
 - Next, construct a bijection $B \to U$ for all $n, k \in \mathbb{N}$.
 - Express the cardinality |U| in terms of n and k.

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- Next, construct a bijection $B \to U$ for all $n, k \in \mathbb{N}$.
- Express the cardinality |U| in terms of n and k.
- **Exercise.** Determine the following numbers for all $0 \le k \le 6$:
 - The number of k-element subsets of $\{1, \ldots, 6\}$.
 - The number of all subsets of $\{1, \ldots, 6\}$.

Conclude from these countings the following numbers for all $0 \le k \le 7$:

- The number of k-element subsets of $\{1, \ldots, 7\}$ that contain 7.
- The number of k-element subsets of $\{1, \ldots, 7\}$ that do not contain 7.
- The number of all subsets of $\{1, \ldots, 7\}$.

Exercises related to the lemma

• **Exercise.** Prove the following identity for all $r, m, n \in \mathbb{N}_0$ such that $r \leq m + n$:

$$\binom{m+n}{r} = \sum_{k=0}^{r} \binom{m}{k} \binom{n}{r-k}.$$

Hint. Let A and B be sets such that |A| = m, |B| = n and $|A \cup B| = m + n$. For all r-element subsets $U \subset A \cup B$ we have that $A \cap U$ is an k-element subset of A and $B \cap U$ is an (r - k)-element subset of B where $k = |A \cap U|$.

• **Proposition.** For all $n \in \mathbb{N}$, we have

$$2^n = \sum_{k=0}^n \binom{n}{k}.$$

• **Exercise.** Prove this proposition by induction. **Hint:** Use the previous lemma for the induction step.

• **Proposition.** For all $n \in \mathbb{N}$, we have

$$2^n = \sum_{k=0}^n \binom{n}{k}.$$

- Exercise. Prove this proposition by induction.

 Hint: Use the previous lemma for the induction step.
- Alternative Proof.

Let P be the set of subsets of an n-element set S.

For each subset $U \in P$ and element $e \in S$ the are two choices: either $e \in U$ or $e \notin U$ and thus

$$|P| = 2^{|S|} = 2^n$$
.

On the other hand, P contains k-element subsets of S for all $0 \le k \le n$ and thus

$$|P| = \binom{n}{0} + \binom{n}{1} + \dots + \binom{n}{n} = \sum_{k=0}^{n} \binom{n}{k}.$$

Binomial Theorem

• **Binomial Theorem.** For all $x, y \in \mathbb{R}$ and $n \in \mathbb{N}$, we have

$$(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}.$$

Example.

$$(x+y)^4 = (x+y) \cdot (x+y) \cdot (x+y) \cdot (x+y)$$

$$= {4 \choose 0} y^4 + {4 \choose 1} xy^3 + {4 \choose 2} x^2 y^2 + {4 \choose 3} x^3 y + {4 \choose 4} x^4$$

$$= y^4 + 4xy^3 + 6y^2 x^2 + 4x^3 y + x^4$$

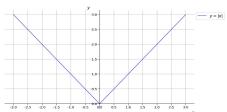
$\S 1.8$: Some important functions

Absolute value

• The absolute value of $x \in \mathbb{R}$ is defined as

$$|x| := \begin{cases} x & \text{if } x \ge 0, \\ -x & \text{if } x < 0. \end{cases}$$

The graph of this function for $x \in [-3, 3]$ is as follows:



• **Problem.** Determine the set

$$L := \{x \in \mathbb{R} : |x - 1| < 2\}.$$

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• First, suppose that $x - 1 \ge 0$ so that $1 \le x$. In this case |x - 1| = x - 1 and thus

$$L_1 := \{ x \in \mathbb{R} : |x - 1| < 2 \& 1 \le x \} = \{ x \in \mathbb{R} : x - 1 < 2 \& 1 \le x \}$$
$$= \{ x \in \mathbb{R} : 1 \le x < 3 \}.$$

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• Next, suppose that x - 1 < 0 so that x < 1. In this case |x - 1| = -(x - 1) = 1 - x and thus

$$L_2 := \{ x \in \mathbb{R} : |x - 1| < 2 \& x < 1 \} = \{ x \in \mathbb{R} : 1 - x < 2 \& x < 1 \}$$
$$= \{ x \in \mathbb{R} : -1 < x < 1 \}.$$

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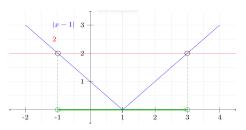
$$L_2 := \{ x \in \mathbb{R} : |x - 1| < 2 \& x < 1 \} = \{ x \in \mathbb{R} : 1 - x < 2 \& x < 1 \}$$
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We conclude that

$$L = L_1 \cup L_2 = \{x \in \mathbb{R} : -1 < x < 3\} = (-1, 3).$$

• The graphs of the functions f(x) := |x - 1| and g(x) := 2 illustrate the solution set

$${x \in \mathbb{R} : |x-1| < 2} = (-1,3).$$



• **Remark.** Notice that the inequality changes when multiplying both sides with a negative real number.

For example $-6x \ge 8$ implies that $x \le -\frac{8}{6} = -\frac{4}{3}$.

Problem. Determine the set

$$L := \{x \in \mathbb{R} : 2|x+3|-4|x-1| \ge 8x-2\}.$$

Solution. We make a case distinction:

Case 1. $x + 3 \ge 0$ and $x - 1 \ge 0$:

$$L_1 := \{x \in \mathbb{R} : 2(x+3) - 4(x-1) \ge 8x - 2, \ x+3 \ge 0, \ x-1 \ge 0\} = [1, \frac{6}{5}].$$

Case 2. $x + 3 \ge 0$ and x - 1 < 0.

$$L_2 := \{x \in \mathbb{R} : 2(x+3) - 4(-x+1) \ge 8x - 2, \ x+3 \ge 0, \ x-1 < 0\} = [-3,1).$$

Case 3. x + 3 < 0 and $x - 1 \ge 0$.

$$L_3 := \{ x \in \mathbb{R} : 2(-x-3) - 4(x-1) \ge 8x - 2, \ x+3 < 0, \ x-1 \ge 0 \}$$

= \{ x \in \mathbb{R} : x < -3 & x \ge 1 \} = \varnothing.

Case 4. x + 3 < 0 and x - 1 < 0.

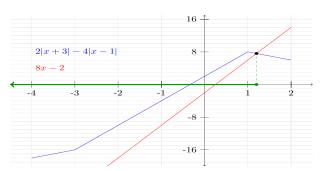
$$L_4:=\{x\in\mathbb{R}:2(-x-3)-4(-x+1)\geq 8x-2,\ x+3<0,\ x-1<0\}=(-\infty,-3).$$

We conclude that

$$L := L_1 \cup L_2 \cup L_3 \cup L_4 = (-\infty, -3) \cup [-3, 1) \cup [1, \frac{6}{5}] = (-\infty, \frac{6}{5}]$$

• The graphs of the functions f(x) := 2|x+3|-4|x-1| and g(x) := 8x-2 illustrating the solution set

$$\{x \in \mathbb{R} : 2|x+3|-4|x-1| \ge 8x-2\} = (-\infty, \frac{6}{5}].$$



Exercises for solving inequalities

• Exercise. Determine the set

$${x \in \mathbb{R} : 3|x+2|-4x+3 \le 5|x-1|}.$$

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- **Exercise.** Prove that the following three identities hold for all $x \in \mathbb{R}$ and $z \in [0, \infty)$:
 - |-x| = |x|.
 - $|x| \le z \iff -z \le x \le z$.
 - $|x| < z \iff -z < x < z$.

Hint. Use case distinctions.

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• **Proof.** Since $z \leq |z|$ and $-z \leq |z|$ for all $z \in \mathbb{R}$, we have

$$x + y \le |x| + |y|$$
 and $-x - y \le |x| + |y|$.

Since |x + y| is equal to either x + y or -(x + y) = -x - y, we concluded the proof.

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• **Exercise.** Let $x_i \in \mathbb{R}$ for all $1 \le i \le n$. Prove the generalized triangle inequality

$$|\sum_{i=1}^{n} x_i| \leq \sum_{i=1}^{n} |x_i|.$$

Hint. Use induction.

Examples of elementary functions

Examples and non-examples of real elementary functions $D \subset \mathbb{R} \to \mathbb{R}$:

function name	examples	non-example
affine linear	$2x+3, \qquad -3x+\pi$	x^2
constant		X
linear	$ \begin{vmatrix} 2x, & -\pi x \\ x^2 + \sqrt{2}x + 1, & (x+1)^4 \\ x^2, & x ^{\pi}, & x ^{-\frac{1}{2}} \\ 2^x, & e^x \end{vmatrix} $	x + 1
polynomial	$x^2 + \sqrt{2}x + 1$, $(x+1)^4$	x^{π}
power	$ x^2, x ^{\pi}, x ^{-\frac{1}{2}}$	$x^{\frac{1}{2}}$ when $x < 0$
exponential	2^x , e^x	
logarithmic	$ \ln x , \log_2 x $	$\ln x$ when $x < 0$
trigoniometric	$\sin x$, $\cos x$, $\tan x$	

Exponential and logarithmic function

The exponential function exp: $\mathbb{R} \to (0, \infty)$ is defined as

$$\exp(x) := e^x := \sum_{k=0}^{\infty} \frac{x^k}{k!} = 1 + \frac{x}{1} + \frac{x^2}{1 \cdot 2} + \frac{x^3}{1 \cdot 2 \cdot 3} + \dots$$

Euler's number is defined as $e:=e^1\approx 2.7$. For all $x_1,x_2\in\mathbb{R}$ we have $e^{x_1+x_2}=e^{x_1}\cdot e^{x_2}$.

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The logarithm function In:
$$(0,\infty)\to\mathbb{R}$$
 is defined as the inverse of exp:

$$\ln y := \exp^{-1}(y).$$

For all
$$y_1,y_2,y\in (0,\infty)$$
 and $x\in \mathbb{R}$ we have

$$\ln(y_1 \cdot y_2) = \ln(y_1) + \ln(y_2), \qquad e^{\ln y} = y, \qquad \ln(e^x) = x.$$

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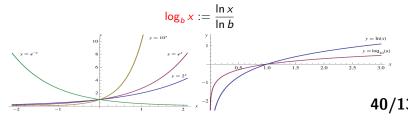
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$$\ln(y_1 \cdot y_2) = \ln(y_1) + \ln(y_2), \qquad e^{\ln y} = y, \qquad \ln(e^x) = x.$$

We define for all $b \in (0, \infty)$ such that $b \neq 1$:



Exercises for exp and In functions

• **Recall.** For all $x, y \in (0, \infty)$ we have

$$ln(x \cdot y) = ln(x) + ln(y), \qquad e^{ln \cdot x} = x, \qquad ln(e^x) = x.$$

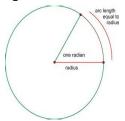
In particular, for all t>0 and constant $c\in(0,\infty)$ we have

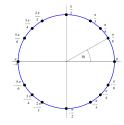
$$c^t = e^{\ln(c^t)} = e^{t \ln c}.$$

- Exercise. Determine the following sets:
 - $\{x \in (4,\infty) : 4 a^{2x-1} b^{4x+3} = 3 c^{x-4} \}$ where a,b,c>0 are constants.
 - $\{x \in (0, \infty) : \log_3 x = 5\}.$
 - $\{x \in (-1,\infty) : 2\ln(x+3) 3\ln(x+2) + \ln(x+1)\}.$

Radians and unit circle

 One radian defines the angle that subtends an arc on a circle, so that the length of the arc is equal to the radius of the circle:





An angle of α degrees corresponds to $\frac{\alpha}{180}\pi$ radians.

degrees
$$\begin{vmatrix} 0^{\circ} & 30^{\circ} & 45^{\circ} & 60^{\circ} & 90^{\circ} & 180^{\circ} & 270^{\circ} & 360^{\circ} \end{vmatrix}$$

radians $\begin{vmatrix} 0 & \frac{\pi}{6} & \frac{\pi}{4} & \frac{\pi}{3} & \frac{\pi}{2} & \pi & \frac{3\pi}{2} & 2\pi \end{vmatrix}$

Cos, sin and tan functions

The unit circle is defined as

$$\{(x,y)\in\mathbb{R}^2: x^2+y^2-1=0\}.$$

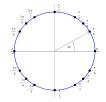
The cosine and sine are the functions $\mathbb{R} \to \mathbb{R}$ such that

$$(\cos \alpha, \sin \alpha)$$

is the end point on the unit-circle of an arc with length α :





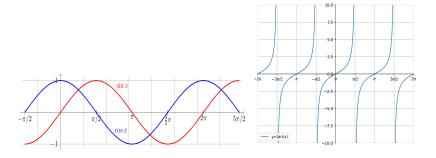


The tangent is the function $\mathbb{R}\setminus\left\{\frac{\pi}{2}+k\pi:k\in\mathbb{Z}\right\}\to\mathbb{R}$ such that

$$\tan\alpha:=\frac{\sin\alpha}{\cos\alpha}.$$

Graphs of cos, sin and tan functions

Graphs of trigonometric functions:



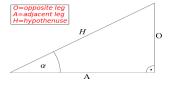
The arccos, arcsin and arctan are defined by the inverses of the following bijective functions respectively

$$\cos\colon [0,\pi]\to [-1,1], \quad \sin\colon [-\tfrac{\pi}{2},\tfrac{\pi}{2}]\to [-1,1], \quad \tan\colon (-\tfrac{\pi}{2},\tfrac{\pi}{2})\to \mathbb{R}.$$

Formulas for right triangles

Trigoniometric formulas for angles in a right triangle:

$$\sin \alpha = \frac{O}{H}, \quad \cos \alpha = \frac{A}{H}, \quad \tan \alpha = \frac{O}{A}.$$



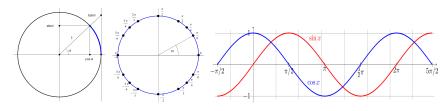
Exercise. Compute without a calculator the points $(\cos \alpha, \sin \alpha)$ for all

$$\alpha \in \{0, \frac{\pi}{6}, \frac{\pi}{4}, \frac{\pi}{3}, \frac{\pi}{2}\}.$$

Hint. Consider the triangles with side lengths (1,1,1) and $(1,1,\sqrt{2})$ and use the fact that the angles in a triangle add up to π .

Some trigoniometric identities for cos and sin

Recall that $(\cos \alpha, \sin \alpha)$ is the end point on the unit-circle of an arc with length α :



As a direct consequence, we observe the following identities for all $x \in \mathbb{R}$:

$$\cos^2 x + \sin^2 x = 1$$

$$\cos(x + 2\pi) = \cos x \qquad \sin(x + 2\pi) = \sin x$$

$$\cos(x + \pi) = -\cos x \qquad \sin(x + \pi) = -\sin x$$

$$\cos(x + \frac{\pi}{2}) = -\sin x \qquad \sin(x + \frac{\pi}{2}) = \cos x$$

$$\cos(-x) = \cos x \qquad \sin(-x) = -\sin x$$

Applying the trigoniometric identities

• Exercise. Compute without calculator the values $\cos \alpha$ and $\sin \beta$ for all

$$\alpha \in \left\{ \tfrac{\pi}{6}, \tfrac{\pi}{4}, \tfrac{\pi}{3} \right\} \quad \text{and} \quad \beta \in \left\{ \tfrac{2\pi}{3}, \tfrac{3\pi}{4}, \tfrac{5\pi}{6} \right\},$$

by using the following facts for all $x \in \mathbb{R}$:

- $\sin \frac{\pi}{6} = \frac{1}{2}$, $\sin \frac{\pi}{4} = \frac{1}{\sqrt{2}}$, $\sin \frac{\pi}{3} = \frac{\sqrt{3}}{2}$, and
- $\sin(-x) = -\sin x$, $\sin(x + \frac{\pi}{2}) = \cos x$, $\cos(x + \frac{\pi}{2}) = -\sin x$.

Applying the trigoniometric identities

• Exercise. Compute without calculator the values $\cos \alpha$ and $\sin \beta$ for all

$$\alpha \in \left\{ \frac{\pi}{6}, \frac{\pi}{4}, \frac{\pi}{3} \right\} \quad \text{and} \quad \beta \in \left\{ \frac{2\pi}{3}, \frac{3\pi}{4}, \frac{5\pi}{6} \right\},$$

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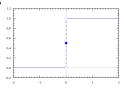
- $\sin \frac{\pi}{6} = \frac{1}{2}$, $\sin \frac{\pi}{4} = \frac{1}{\sqrt{2}}$, $\sin \frac{\pi}{3} = \frac{\sqrt{3}}{2}$, and
- $\sin(-x) = -\sin x$, $\sin(x + \frac{\pi}{2}) = \cos x$, $\cos(x + \frac{\pi}{2}) = -\sin x$.
- Exercise. Compute

$$\left\{ x \in [-\pi, \pi] : \cos x = \frac{\sqrt{3}}{2} \right\}$$
 and $\left\{ x \in [-\pi, \pi] : \sin^2 x = \frac{1}{2} \right\}$.

The Heaviside step and ReLu functions

The Heaviside step function $H: \mathbb{R} \to \mathbb{R}$ is defined as

$$H(x) := \begin{cases} 1 & \text{if } x \ge 0, \\ 0 & \text{if } x < 0. \end{cases}$$



The Heaviside step and ReLu functions

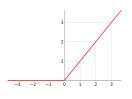
The Heaviside step function $H: \mathbb{R} \to \mathbb{R}$ is defined as

$$H(x) := \begin{cases} 1 & \text{if } x \ge 0, \\ 0 & \text{if } x < 0. \end{cases}$$



The ramp function $r: \mathbb{R} \to \mathbb{R}$ is defined as

$$r(x) := \begin{cases} x & \text{if } x \ge 0, \\ 0 & \text{if } x < 0. \end{cases}$$



In the context of artificial neural networks the ramp function is called a ReLU activation function, where ReLu stands for "rectified linear unit".

$\S 1.9$: Complex numbers

Complex numbers

The set of complex numbers with imaginary unit $\mathfrak i$ is defined as

$$\mathbb{C} := \{ x + \mathfrak{i} \, y : x, y \in \mathbb{R} \} \quad \text{where} \quad \mathfrak{i}^2 = -1.$$

Let z = x + i y be a complex number.

$$Re(z) := x$$
 (real part)
 $Im(z) := y$ (imaginary part)

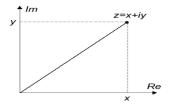
Addition and multiplication of the complex numbers z = x + i y and w = u + i v:

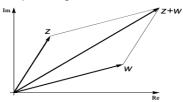
$$z + w = (x + iy) + (u + iv) = (x + u) + i(y + v).$$

$$z \cdot w = (x + iy) \cdot (u + iv) = (xu + i^2yv) + i(xv + yu) = (xu - yv) + i(xv + yu).$$

We can identify the complex numbers with the complex plane.

Geometrically, the sum z + w is the corner of a parallelogram:





Question. What does multiplication geometrically mean?

Complex numbers form a field

The complex conjugate of a complex number z = x + iy is defined as

$$\overline{z} := x - i y.$$

We observe that

$$z \cdot \overline{z} = (x + i y) \cdot (x - i y) = x^2 + y^2$$
.

We can use this to compute the multiplicative inverse of z:

$$z^{-1} = \frac{1}{z} = \frac{1}{z} \cdot \frac{\overline{z}}{\overline{z}} = \frac{x - iy}{x^2 + y^2} = \frac{x}{x^2 + y^2} - i\frac{y}{x^2 + y^2}.$$

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Exercise. Show that \mathbb{C} is a field.

Hint.

Notice that -z = -x - iy and 0 are the additive inverse and identity, resp. Moreover, z^{-1} and 1 are the multiplicative inverse and identity, resp.

It remains to show that

complex multiplication and addition are commutative and associative, and that multiplication distributes over addition.

51/134

Complex numbers

Problem. Compute the real and imaginary part of

$$\frac{2-3\mathfrak{i}}{1+5\mathfrak{i}}.$$

Complex numbers

Problem. Compute the real and imaginary part of

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$$\frac{2-\mathfrak{i}\,3}{1+\mathfrak{i}\,5} = \frac{(2-\mathfrak{i}\,3)\cdot(1-\mathfrak{i}\,5)}{(1+\mathfrak{i}\,5)\cdot(1-\mathfrak{i}\,5)} = \frac{(2-\mathfrak{i}\,3)\cdot(1-\mathfrak{i}\,5)}{1^2+5^2} = \frac{-13-\mathfrak{i}\,13}{26} = -\frac{1}{2}-\mathfrak{i}\,\frac{1}{2}.$$

Problem. Compute the real and imaginary part of

$$\sqrt{i}$$
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Complex numbers

Problem. Compute the real and imaginary part of

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Problem. Compute the real and imaginary part of

$$\sqrt{i}$$
.

Answer. Suppose that z = x + iy with $x, y \in \mathbb{R}$ such that $z^2 = i$. By comparing the real and complex part of

$$z^2 = x^2 - y^2 + i 2xy = 0 + i 1 = i$$

we find that

$$x^{2} - y^{2} = (x + y)(x - y) = 0$$
 and $2xy = 1$.

As xy > 0 we deduce that $x + y \neq 0$ and thus x - y = 0. It follows $2x^2 = 1$ so that $x = y = \pm \frac{1}{\sqrt{2}}$ and thus

$$z = \pm \left(\frac{1}{\sqrt{2}} + i \, \frac{1}{\sqrt{2}}\right).$$

Complex numbers

Problem. Compute the real and imaginary part of

$$\frac{2-3i}{1+5i}.$$

Answer.

$$\frac{2-\mathfrak{i}\,3}{1+\mathfrak{i}\,5} = \frac{(2-\mathfrak{i}\,3)\cdot(1-\mathfrak{i}\,5)}{(1+\mathfrak{i}\,5)\cdot(1-\mathfrak{i}\,5)} = \frac{(2-\mathfrak{i}\,3)\cdot(1-\mathfrak{i}\,5)}{1^2+5^2} = \frac{-13-\mathfrak{i}\,13}{26} = -\frac{1}{2}-\mathfrak{i}\,\frac{1}{2}.$$

Problem. Compute the real and imaginary part of

$$\sqrt{i}$$
.

Answer. Suppose that $z=x+\mathfrak{i}\,y$ with $x,y\in\mathbb{R}$ such that $z^2=\mathfrak{i}.$ By comparing the real and complex part of

$$z^2 = x^2 - y^2 + i \frac{2xy}{2} = 0 + i \frac{1}{2} = i$$

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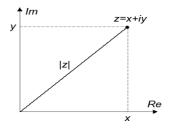
Exercise. What are the possible values of i^n for $n \in \mathbb{N}_0$?

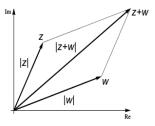
The absolute value of complex numbers

The absolute value of the complex number z = x + i y is defined as

$$|z| := \sqrt{x^2 + y^2}.$$

The absolute value is equal to the distance between 0 and z in the complex plane:





We see in these examples that

$$Im(z) = y \le |z|, \qquad Re(z) = x \le |z|, \qquad |z + w| \le |z| + |w|.$$

The latter is called the triangle inequality for complex numbers.

Triangle inequality

Lemma. For all $a, b \in \mathbb{C}$, we have

$$\left|a\right|^2=\overline{a}\cdot a,\quad \overline{a+b}=\overline{a}+\overline{b},\quad a+\overline{a}=2\operatorname{Re}(a),\quad \operatorname{Re}(a)\leq |a|,\quad |a\cdot b|=|a|\cdot |b|,\quad |a|=|\overline{a}|.$$

Proof. Exercise.

Theorem (triangle inequality). For all $z, w \in \mathbb{C}$ we have

$$|z+w|\leq |z|+|w|.$$

Proof. As both sides are positive real numbers it suffices to prove that

$$|z+w|^2 \leq (|z|+|w|)^2$$
.

We use the above lemma to show this inequality:

$$|z + w|^{2} = (z + w) \cdot (\overline{z + w}) = (z + w) \cdot (\overline{z} + \overline{w})$$

$$= |z|^{2} + (z\overline{w} + \overline{z}w) + |w|^{2} = |z|^{2} + 2\operatorname{Re}(z\overline{w}) + |w|^{2}$$

$$\leq$$

$$|z|^{2} + 2|z \cdot \overline{w}| + |w|^{2} = |z|^{2} + 2|z| \cdot |\overline{w}| + |w|^{2}$$

$$= |z|^{2} + 2|z| \cdot |w| + |w|^{2} = (|z| + |w|)^{2}.$$

This completes the proof.

Representations of complex numbers

We consider two equivalent representations of z

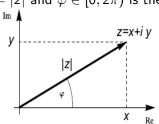
canonical representation:

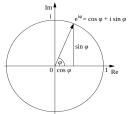
$$z = x + i y$$
.

Euler's formula:

$$z = r \cdot e^{i\varphi} = r \cdot (\cos \varphi + i \sin \varphi),$$

where r = |z| and $\varphi \in [0, 2\pi)$ is the argument of z.





We call (r, φ) the polar coordinate of z.

Exercise. Show for all $z, w \in \mathbb{C}$ that |z + w| = |z| + |w| if and only if z and w have the same argument.

55/134

Convert from polar to canonical representations

Problem. Suppose that z has polar coordinates $(r, \phi) = (2, \frac{\pi}{4})$. Determine the canonical representation of z.

Convert from polar to canonical representations

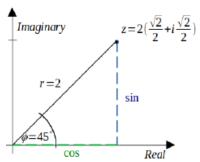
Problem. Suppose that z has polar coordinates $(r, \phi) = (2, \frac{\pi}{4})$. Determine the canonical representation of z.

Answer. We know that

$$\cos \frac{\pi}{4} = \sin \frac{\pi}{4} = \frac{1}{\sqrt{2}} = \frac{\sqrt{2}}{2}$$

and thus the canonical representation of z is

$$z = 2 \cdot \left(\cos \frac{\pi}{4} + i \sin \frac{\pi}{4}\right) = \sqrt{2} + i \sqrt{2}.$$



Real and imaginary parts using Euler's formula

Problem. Compute the real and imaginary part of

$$\mathrm{e}^{\mathrm{i}\,arphi} \qquad ext{for all} \qquad arphi \in \left\{ rac{\pi}{2}, \pi, rac{3\pi}{2}, 2\pi
ight\}.$$

Real and imaginary parts using Euler's formula

Problem. Compute the real and imaginary part of

$$e^{i\,\varphi}$$
 for all $\varphi\in\left\{\frac{\pi}{2},\pi,\frac{3\pi}{2},2\pi\right\}$.

Answer. We can read the real and imaginary parts from their canonical representations:

$$\begin{split} &e^{i\frac{\pi}{2}} = \cos\frac{\pi}{2} + i\sin\frac{\pi}{2} = i, \\ &e^{i\pi} = \cos\pi + i\sin\pi = -1, \\ &e^{i\frac{3\pi}{2}} = \cos\frac{3\pi}{2} + i\sin\frac{3\pi}{2} = -i, \\ &e^{i2\pi} = \cos2\pi + i\sin2\pi = 1. \end{split}$$

Euler's identity is considered to be an exemplar of mathematical beauty as it shows a profound connection between the most fundamental numbers in mathematics:

$$e^{\mathfrak{i}\,\pi}+1=0$$

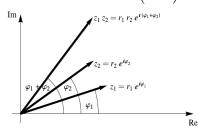
Multiplication and division using exponential representation

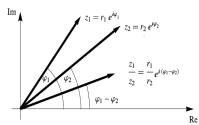
We can use Euler's formula to compute the product, quotient and powers of the complex numbers $z_1=r_1e^{\mathrm{i}\,\varphi_1}$, $z_2=r_2e^{\mathrm{i}\,\varphi_2}$ and $z=r\,e^{\mathrm{i}\,\varphi}$:

$$z_1 \cdot z_2 = (r_1 e^{i\varphi_1}) \cdot (r_2 e^{i\varphi_2}) = r_1 r_2 e^{i(\varphi_1 + \varphi_2)}.$$

$$\frac{z_1}{z_2} = \frac{r_1 e^{i\varphi_1}}{r_2 e^{i\varphi_2}} = \frac{r_1}{r_2} e^{i(\varphi_1 - \varphi_2)}.$$

$$z^n = (r e^{i\varphi})^n = r^n e^{in\varphi}.$$





de Moivre's formula: $z^n = r^n e^{in\varphi} = r^n (cos(n\varphi) + i sin(n\varphi))$. 58/

Problem. Compute the arithmetic representations of

$$(1+\mathfrak{i})^{12}$$
 and $\frac{1+\mathfrak{i}}{3e^{-\mathfrak{i}\frac{\pi}{4}}}.$

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Answer.

$$(1+i)^{12} = (\sqrt{2} e^{i\frac{\pi}{4}})^{12} = 2^6 e^{i3\pi} = 64 e^{i\pi} = -64.$$

Problem. Compute the arithmetic representations of

$$(1+\mathfrak{i})^{12}$$
 and $\frac{1+\mathfrak{i}}{3e^{-\mathfrak{i}\frac{\pi}{4}}}$.

Answer.

$$(1+i)^{12} = (\sqrt{2} e^{i\frac{\pi}{4}})^{12} = 2^6 e^{i3\pi} = 64 e^{i\pi} = -64.$$

$$\frac{1+\mathfrak{i}}{3\,e^{-\mathfrak{i}\frac{\pi}{4}}} = \frac{\sqrt{2}\,e^{\mathfrak{i}\frac{\pi}{4}}}{3\,e^{-\mathfrak{i}\frac{\pi}{4}}} = \frac{\sqrt{2}}{3}\,e^{\mathfrak{i}\frac{\pi}{2}} = \frac{\sqrt{2}}{3}\,\mathfrak{i}.$$

Problem. Compute the arithmetic representations of

$$(1+\mathfrak{i})^{12} \qquad \text{and} \qquad \frac{1+\mathfrak{i}}{3e^{-\mathfrak{i}\frac{\pi}{4}}}.$$

Answer.

$$(1+\mathfrak{i})^{12} = \left(\sqrt{2}\,e^{\mathfrak{i}\frac{\pi}{4}}\right)^{12} = 2^6\,e^{\mathfrak{i}3\pi} = 64\,e^{\mathfrak{i}\pi} = -64.$$

$$\frac{1+\mathfrak{i}}{3\,e^{-\mathfrak{i}\frac{\pi}{4}}} = \frac{\sqrt{2}\,e^{\mathfrak{i}\frac{\pi}{4}}}{3\,e^{-\mathfrak{i}\frac{\pi}{4}}} = \frac{\sqrt{2}}{3}\,e^{\mathfrak{i}\frac{\pi}{2}} = \frac{\sqrt{2}}{3}\,\mathfrak{i}.$$

Alternative.

$$\frac{1+i}{3e^{-i\frac{\pi}{4}}} = \frac{1}{3} \cdot \frac{1+i}{\frac{1}{\sqrt{2}} - \frac{1}{\sqrt{2}}i} \cdot \frac{\frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}}i}{\frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}}i}i$$

$$=$$

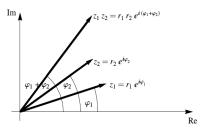
$$\frac{1}{3} \cdot \frac{\frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}} i + \frac{1}{\sqrt{2}} i - \frac{1}{\sqrt{2}}}{\frac{1}{2} + \frac{1}{2}} = \frac{\sqrt{2}}{3} i.$$

Multiplication and absolute value

Recall that multiplication of the complex numbers

$$z_1 = r_1 e^{i\varphi_1}$$
 and $z_2 = r_2 e^{i\varphi_2}$

has the following geometric interpretation:



Since $r_1 = |z_1|$ and $r_2 = |z_2|$, we observe that the following identity holds

$$|z_1\cdot z_2|=|z_1|\cdot |z_2|.$$

Quadratic formula

Quadratic-formula. Let $a, b, c \in \mathbb{C}$ and $a \neq 0$.

$$q(x) := ax^2 + bx + c = 0$$
 \iff $x = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}.$

If $a, b, c \in \mathbb{R}$, then the following holds:

- If $b^2 4ac > 0$, then q(x) has two real zeros.
- If $b^2 4ac = 0$, then q(x) has one real zero of multiplicity 2.
- If $b^2 4ac < 0$, then q(x) has two nonreal zeros.

Example.

$$x^{2} - 3x + 1 = 0 \quad \iff \quad x = \frac{3 \pm \sqrt{5}}{2}.$$

$$x^{2} - 2x + 1 = (x - 1)^{2} = 0 \quad \iff \quad x = 1 \text{ with multiplicity 2.}$$

$$x^2 - x + 1 = 0 \quad \iff \quad x = \frac{1 \pm i\sqrt{3}}{2}.$$

Remark. For all $u \in R$, we have

$$\sqrt{-u} = i\sqrt{u}$$
.

Roots of complex polynomials

Problem. Determine the following set of solutions

$$\left\{x\in\mathbb{C}:x^2+(1-\mathfrak{i})x-\mathfrak{i}=0\right\}.$$

Answer. We apply the quadratic formula:

$$x=\frac{\mathfrak{i}-1\pm\sqrt{(1-\mathfrak{i})^2+4\mathfrak{i}}}{2}=\frac{\mathfrak{i}-1\pm\sqrt{2}\mathfrak{i}}{2}=\frac{\mathfrak{i}-1\pm\sqrt{2}\sqrt{\mathfrak{i}}}{2}.$$

We recall that

$$\sqrt{i} = \frac{1}{\sqrt{2}} + i \frac{1}{\sqrt{2}},$$

and conclude that the solutions to our equation are

$$x = \frac{i - 1 + \sqrt{2}\left(\frac{1}{\sqrt{2}} + i\frac{1}{\sqrt{2}}\right)}{2} = i$$

and

$$x = \frac{i - 1 + \sqrt{2}\left(-\frac{1}{\sqrt{2}} - i\frac{1}{\sqrt{2}}\right)}{2} = -1.$$

This implies that $\{x \in \mathbb{C} : x^2 + (1 - i)x - i = 0\} = \{i, -1\}.$ **62/1**3

$\S 1.10$: Vectors and norms

Addition and scalar multiplication of vectors

A vector **v** and the zero vector **0** both in \mathbb{C}^d and of dimension $d \in \mathbb{N}$:

$$\mathbf{v} = (v_i)_{i=1}^d = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_d \end{pmatrix} \quad \text{and} \quad \mathbf{0} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}.$$

The addition and scalar multiplication of vectors is for all scalars $\lambda \in \mathbb{C}$ defined as

$$\mathbf{u} + \mathbf{v} = \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_d \end{pmatrix} + \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_d \end{pmatrix} = \begin{pmatrix} u_1 + v_1 \\ u_2 + v_2 \\ \vdots \\ u_d + v_d \end{pmatrix} \quad \text{and} \quad \lambda \cdot \mathbf{u} = \begin{pmatrix} \lambda \cdot u_1 \\ \lambda \cdot u_2 \\ \vdots \\ \lambda \cdot u_d \end{pmatrix}.$$

Euclidean norm and inner product

The Euclidean norm of $\mathbf{v}(v_i)_{i=1}^d \in \mathbb{C}^d$ is a real number:

$$\|\mathbf{v}\|_2 := \sqrt{\sum_{i=1}^d |v_i|^2}.$$

The inner product of the vectors $\mathbf{u} = (u_i)_{i=1}^d$ and $\mathbf{v} = (v_i)_{i=1}^d \in \mathbb{C}^d$:

$$\langle \mathbf{u}, \mathbf{v} \rangle := \sum_{i=1}^d u_i \bar{v}_i.$$

We observe the following equality:

$$\|\mathbf{v}\|_2 = \sqrt{\langle \mathbf{v}, \mathbf{v} \rangle}.$$

Remark. If d = 1, then the Euclidean norm is the absolute value.

Inner product

Lemma. The inner product is formally a mapping

$$\langle \cdot, \cdot \rangle \colon \mathbb{C}^d \times \mathbb{C}^d \to \mathbb{C}$$

that satisfies the following properties:

Positive definiteness:

For all
$$\mathbf{u} \in \mathbb{C}^d$$
, where $\mathbb{R}_{\geq 0} := \{x \in \mathbb{R} : x > 0\}$:

$$\langle \mathbf{u}, \mathbf{u} \rangle \in \mathbb{R}_{\geq 0}$$
 and $\langle \mathbf{u}, \mathbf{u} \rangle = 0 \iff \mathbf{u} = \mathbf{0}.$

Linearity in the first argument:

For all $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathbb{C}^d$ and $\lambda \in \mathbb{C}$,

$$\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{w} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle$$
 and $\langle \lambda \cdot \mathbf{u}, \mathbf{v} \rangle = \lambda \cdot \langle \mathbf{u}, \mathbf{v} \rangle$.

Conjugate symmetry:

For all $\mathbf{u}, \mathbf{v} \in \mathbb{C}^d$.

$$\langle \mathbf{u}, \mathbf{v}
angle = \overline{\langle \mathbf{v}, \mathbf{u}
angle}$$

Proof. Exercise.

Linearity of inner product

Lemma. For all $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathbb{C}^d$ and $\lambda, \mu \in \mathbb{C}$ we have

$$\langle \lambda \cdot \mathbf{u} + \mu \cdot \mathbf{v}, \mathbf{w} \rangle = \lambda \cdot \langle \mathbf{u}, \mathbf{w} \rangle + \mu \cdot \langle \mathbf{v}, \mathbf{w} \rangle \text{ and } \langle \mathbf{u}, \lambda \cdot \mathbf{v} + \mu \cdot \mathbf{w} \rangle = \overline{\lambda} \cdot \langle \mathbf{u}, \mathbf{v} \rangle + \overline{\mu} \cdot \langle \mathbf{v}, \mathbf{w} \rangle.$$

Proof. By linearity in the first argument,

$$\langle \lambda \cdot \mathbf{u} + \mu \cdot \mathbf{v}, \mathbf{w} \rangle = \langle \lambda \cdot \mathbf{u}, \mathbf{w} \rangle + \langle \mu \cdot \mathbf{v}, \mathbf{w} \rangle = \lambda \cdot \langle \mathbf{u}, \mathbf{w} \rangle + \mu \cdot \langle \mathbf{v}, \mathbf{w} \rangle.$$

We apply conjugate symmetry and properties of the complex conjugate:

$$\langle \mathbf{u}, \lambda \cdot \mathbf{v} + \mu \cdot \mathbf{w} \rangle = \overline{\langle \lambda \cdot \mathbf{v} + \mu \cdot \mathbf{w}, \mathbf{u} \rangle} = \overline{\lambda \cdot \langle \mathbf{v}, \mathbf{u} \rangle} + \mu \cdot \overline{\langle \mathbf{w}, \mathbf{u} \rangle} = \overline{\lambda} \cdot \overline{\langle \mathbf{v}, \mathbf{u} \rangle} + \overline{\mu} \cdot \overline{\langle \mathbf{w}, \mathbf{u} \rangle}.$$

We again apply conjugate symmetry and the fact that $\overline{(\bar{z})} = z$:

$$\overline{\lambda} \cdot \overline{\langle \mathbf{v}, \mathbf{u} \rangle} + \overline{\mu} \cdot \overline{\langle \mathbf{w}, \mathbf{u} \rangle} = \overline{\lambda} \cdot \langle \mathbf{u}, \mathbf{v} \rangle + \overline{\mu} \cdot \langle \mathbf{u}, \mathbf{w} \rangle.$$

Exercises for inner product and Euclidean norm

Lemma (semilinearity in the second argument).

For all $\lambda \in \mathbb{C}$ and $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathbb{C}^d$, we have

$$\langle \mathbf{u}, \mathbf{v} + \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{v} \rangle + \langle \mathbf{u}, \mathbf{w} \rangle$$
 and $\langle \mathbf{u}, \lambda \cdot \mathbf{v} \rangle = \overline{\lambda} \cdot \langle \mathbf{u}, \mathbf{v} \rangle$.

Proof. Exercise.

Exercise. Prove that for all $\lambda \in \mathbb{C}$ and $\mathbf{v} \in \mathbb{C}^d$, we have

$$\|\lambda \cdot \mathbf{v}\|_2 = |\lambda| \cdot |\mathbf{v}\|_2.$$

Cauchy-Schwarz inequality

Theorem (Cauchy-Schwarz inequality). For all $\mathbf{u}, \mathbf{v} \in \mathbb{C}^d$,

$$|\langle \mathbf{u}, \mathbf{v} \rangle| \leq \|\mathbf{u}\|_2 \|\mathbf{v}\|_2.$$

Proof. We observe that

$$|\langle \mathbf{u}, \mathbf{v} \rangle| = \left| \sum_{i=1}^d u_i \bar{v}_i \right| \le \sum_{i=1}^d |u_i \bar{v}_i| = \sum_{i=1}^d |u_i| |v_i| = \|\mathbf{u}\|_2 \|\mathbf{v}\|_2 \sum_{i=1}^d \frac{|u_i|}{\|\mathbf{u}\|_2} \frac{|v_i|}{\|\mathbf{v}\|_2}.$$

Since $(a - b)^2 = a^2 + b^2 - 2ab \ge 0$ we have $a \cdot b \le \frac{1}{2}(a^2 + b^2)$ and thus

$$\|\mathbf{u}\|_2 \|\mathbf{v}\|_2 \sum_{i=1}^d \frac{|u_i|}{\|\mathbf{u}\|_2} \frac{|v_i|}{\|\mathbf{v}\|_2} \le \|\mathbf{u}\|_2 \|\mathbf{v}\|_2 \sum_{i=1}^d \frac{1}{2} \left(\frac{|u_i|^2}{\|\mathbf{u}\|_2^2} + \frac{|v_i|^2}{\|\mathbf{v}\|_2^2} \right).$$

The proof is concluded as $\|\mathbf{w}\|_2 = \sqrt{\sum_{i=1}^d |w_i|^2}$ for all $\mathbf{w} \in \mathbb{C}^d$ and thus

$$\sum_{i=1}^{d} \frac{1}{2} \left(\frac{|u_i|^2}{\|\mathbf{u}\|_2^2} + \frac{|v_i|^2}{\|\mathbf{v}\|_2^2} \right) = \frac{1}{2} \left(\frac{1}{\|\mathbf{u}\|_2^2} \sum_{i=1}^{d} |u_i|^2 + \frac{1}{\|\mathbf{v}\|_2^2} \sum_{i=1}^{d} |v_i|^2 \right) = 1.$$

Triangle inequality

Theorem (Triangle inequality). For all $\mathbf{u}, \mathbf{v} \in \mathbb{C}^d$,

$$\| \textbf{u} + \textbf{v} \|_2 \leq \| \textbf{u} \|_2 + \| \textbf{v} \|_2.$$

Proof. See lecture notes: uses Cauchy-Schwarz inequality.

Unit vectors and standard basis

The unit vectors are defined as

$$\mathbf{e}_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \ \mathbf{e}_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}, \ \dots \ , \mathbf{e}_d = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}.$$

Any vector $\mathbf{v} = (v_i)_{i=1}^d \in \mathbb{C}^d$ can be uniquely represented in terms of unit vectors:

$$\mathbf{v} = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_d \end{pmatrix} = \sum_{i=1}^d v_i \mathbf{e}_i.$$

The **standard basis** for \mathbb{C}^d is defined as the set

$$\{\mathbf{e}_1,\ldots,\mathbf{e}_d\}.$$

Notice that $\|\mathbf{e}_i\| = 1$ for all $i = 1, \dots, d$.

$\S 2.1$: Matrices

Matrices

A real $m \times n$ matrix A is an array of entries in \mathbb{R} arranged in m rows and n columns:

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$

Notation. We write $A = (a_{ij})_{i,j=1}^{m,n}$.

Names for matrices:

$$D := \begin{pmatrix} a_{11} & 0 & 0 \\ 0 & a_{22} & 0 \\ 0 & 0 & a_{33} \end{pmatrix} \qquad \begin{pmatrix} a_{11} \\ a_{21} \\ a_{31} \end{pmatrix} \qquad \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} & a_{23} \\ 0 & 0 & a_{33} \end{pmatrix}$$

$$\begin{array}{c} \text{diagonal} \\ \text{matrix} \end{array} \qquad \begin{array}{c} \text{column} \\ \text{vector} \end{array} \qquad \begin{array}{c} \text{row} \\ \text{upper triangular} \\ \text{matrix} \end{array}$$

Notation. We write $D = \operatorname{diag}(a_{11}, a_{22}, a_{33})$.

Scalar multiplication of matrices

Scalar multiplication of a matrix $A := (a_{ij})_{i,j=1}^{m,n}$ with scalar $\lambda \in \mathbb{R}$:

$$\lambda \cdot A = \begin{pmatrix} \lambda \cdot a_{11} & \lambda \cdot a_{12} & \dots & \lambda \cdot a_{1n} \\ \lambda \cdot a_{21} & \lambda \cdot a_{22} & \dots & \lambda \cdot a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \lambda \cdot a_{m1} & \lambda \cdot a_{m2} & \dots & \lambda \cdot a_{mn} \end{pmatrix}.$$

Example.

Matrix sum

The matrix sum of $m \times n$ matrices $A := (a_{ij})_{i,j=1}^{m,n}$ and $B := (b_{ij})_{i,j=1}^{m,n}$ is defined as the matrix

$$A + B = (\mathbf{a}_{ij} + \mathbf{b}_{ij})_{i,j=1}^{m,n}.$$

Example.

$$\begin{pmatrix} 3 & -6 & 9 \\ 0 & 6 & -12 \end{pmatrix} + \begin{pmatrix} 2 & -2 & 0 \\ 2 & 6 & 2 \end{pmatrix} = \begin{pmatrix} 3+2 & -6-2 & 9+0 \\ 0+2 & 6+6 & -12+2 \end{pmatrix} = \begin{pmatrix} 5 & -8 & 9 \\ 2 & 12 & -10 \end{pmatrix}$$

Matrix product

The matrix product of an $m \times p$ matrix $A := (a_{ij})_{i,j=1}^{m,p}$ with an $p \times n$ matrix $B := (b_{ij})_{i,j=1}^{p,n}$ is defined as

$$A \cdot B = \left(\sum_{k=1}^{p} \mathbf{a}_{ik} \cdot \mathbf{b}_{kj}\right)_{i,j=1}^{m,n}$$

Example. m = 3, p = 4 and n = 2:

$$\begin{pmatrix} 2 & 1 & -3 & 0 \\ 1 & 0 & 1 & 2 \\ -1 & 1 & 3 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 \\ -1 & 2 \\ 2 & 1 \\ 3 & 1 \end{pmatrix} =$$

$$\begin{pmatrix} 2 \cdot 1 + 1 \cdot -1 + (-3) \cdot 2 + 0 \cdot 3 & 2 \cdot 0 + 1 \cdot 2 + (-3) \cdot 1 + 0 \cdot 1 \\ 1 \cdot 1 + 0 \cdot -1 + & 1 \cdot 2 + 2 \cdot 3 & 1 \cdot 0 + 0 \cdot 2 + & 1 \cdot 1 + 2 \cdot 1 \\ -1 \cdot 1 + 1 \cdot -1 + & 3 \cdot 2 + 1 \cdot 3 & -1 \cdot 0 + 1 \cdot 2 + & 3 \cdot 1 + 1 \cdot 1 \end{pmatrix} = \begin{pmatrix} -5 & -1 \\ 9 & 3 \\ 7 & 6 \end{pmatrix}.$$

Remarks. If the number of columns of A is not equal to the number of rows of B, then $A \cdot B$ is not defined!

Matrix multiplication is associative:

$$(A \cdot B) \cdot C = A \cdot (B \cdot C).$$

However, matrix multiplication is not commutative:

$$A \cdot B \neq B \cdot A$$
.

Matrix-vector product

Let \mathbf{a}_i denote the *i*th row of a matrix $A \in \mathbb{R}^{m \times n}$ and let $\mathbf{x} \in \mathbb{R}^n$ be a vector:

$$A = \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_2 \\ \vdots \\ \mathbf{a}_m \end{pmatrix} \quad \text{and} \quad \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \cdots \\ x_n \end{pmatrix}$$

The matrix-vector product $\langle \mathbf{a}_i, \mathbf{x} \rangle$ is defined as the *i*-th row of the following matrix product

$$A \cdot \mathbf{x} := egin{pmatrix} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n \ dots \ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \end{pmatrix} = egin{pmatrix} \langle \mathbf{a}_1, \mathbf{x}
angle \ \langle \mathbf{a}_2, \mathbf{x}
angle \ dots \ \langle \mathbf{a}_2, \mathbf{x}
angle \ dots \ \langle \mathbf{a}_m, \mathbf{x}
angle \end{pmatrix}.$$

Example.

$$A = \begin{pmatrix} 1 & 6 \\ 2 & 5 \\ 3 & 4 \end{pmatrix} \quad \text{and} \quad \mathbf{x} = \begin{pmatrix} 3 \\ 4 \end{pmatrix} \quad \Longrightarrow \quad A \cdot \mathbf{x} = \begin{pmatrix} 1 \cdot 3 + 6 \cdot 4 \\ 2 \cdot 3 + 5 \cdot 4 \\ 3 \cdot 3 + 4 \cdot 4 \end{pmatrix} = \begin{pmatrix} 27 \\ 26 \\ 25 \end{pmatrix}.$$

Calculation rules for matrices

Theorem.

• For all $A, B \in \mathbb{R}^{m \times n}$ and $\lambda \in \mathbb{R}$,

$$\lambda \cdot (A+B) = \lambda \cdot A + \lambda \cdot B.$$

• For all $A \in \mathbb{R}^{m \times p}$, $B \in \mathbb{R}^{p \times n}$ and $\lambda \in \mathbb{R}$,

$$A \cdot (\lambda \cdot B) = \lambda \cdot A \cdot B.$$

• For all $A \in \mathbb{R}^{m \times p}$ and $B, C \in \mathbb{R}^{p \times n}$,

$$A \cdot (B + C) = A \cdot B + A \cdot C.$$

• For all $A \in \mathbb{R}^{m \times n}$ and $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$,

$$A \cdot (\mathbf{x} + \mathbf{y}) = A \cdot \mathbf{x} + A \cdot \mathbf{y}.$$

• For all $A \in \mathbb{R}^{m \times n}$, $\mathbf{x} \in \mathbb{R}^n$ and $\lambda \in \mathbb{R}$,

$$A \cdot (\lambda \cdot \mathbf{x}) = \lambda \cdot A \cdot \mathbf{x}.$$

Proof. Exercise.

Additive and multiplicative identities

The zero matrix $0_{mn} \in \mathbb{R}^{m \times n}$ and identity matrix $I_n \in \mathbb{R}^{n \times n}$ are defined as

$$\mathbf{0}_{\textit{mn}} := \begin{pmatrix} 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & 0 \end{pmatrix} \quad \text{and} \quad \textit{I}_{\textit{n}} := \begin{pmatrix} 1 & 0 & 0 & \cdots & 0 \\ 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \end{pmatrix}.$$

Exercise. Prove that for all $A \in \mathbb{R}^{m \times n}$ we have

$$A + 0_{mn} = 0_{mn} + A = A$$
, $A \cdot I_n = A$ and $I_m \cdot A = A$.

Remark.

A square matrix $A \in \mathbb{R}^{n \times n}$ is invertible if there exists a matrix $A^{-1} \in \mathbb{R}^{n \times n}$ such that

$$A \cdot A^{-1} = A^{-1} \cdot A = I_n$$

In this case, we call A^{-1} the inverse of A.

Matrix transpose

The matrix transpose of $A := (a_{ij})_{ij}$ is obtained by reflecting the entries along the diagonal:

$$A^{\top} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}^{\top} = \begin{pmatrix} a_{11} & a_{21} & \dots & a_{m1} \\ a_{12} & a_{22} & \dots & a_{m2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \dots & a_{mn} \end{pmatrix}$$

Example.

$$\begin{pmatrix} 1 & 2 & -1 \\ 3 & -5 & 6 \end{pmatrix}^{\mathsf{T}} = \begin{pmatrix} 1 & 3 \\ 2 & -5 \\ -1 & 6 \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} 2 & -1 & 3 \\ 0 & 4 & -5 \\ 1 & 2 & 8 \end{pmatrix}^{\mathsf{T}} = \begin{pmatrix} 2 & 0 & 1 \\ -1 & 4 & 2 \\ 3 & -5 & 8 \end{pmatrix}$$

Transpose and multiplication

Let $(M)_{ij}$ denote the entry in the *i*th row and *j*th column of a matrix $M \in \mathbb{R}^{n \times m}$. We observe that for all $1 \leq i \leq n$ and $1 \leq j \leq m$,

$$(M^{\top})_{ij}=(M)_{ji}.$$

Lemma. If $A \in \mathbb{R}^{m \times p}$ and $B \in \mathbb{R}^{p \times n}$, then

$$(A\cdot B)^{\top}=B^{\top}\cdot A^{\top}.$$

Proof. Notice that $B^{\top} \cdot A^{\top}$ exists as $B^{\top} \in \mathbb{R}^{n \times p}$ and $A^{\top} \in \mathbb{R}^{p \times m}$. Hence, for all $1 \leq i \leq n$ and $1 \leq j \leq m$,

$$((A \cdot B)^{\top})_{ij} = (A \cdot B)_{ji} = \sum_{k=1}^{p} (A)_{jk} (B)_{ki}$$

and

$$(B^{\top} \cdot A^{\top})_{ij} = \sum_{k=1}^{p} (B^{\top})_{ik} (A^{\top})_{kj} = \sum_{k=1}^{p} (B)_{ki} (A)_{jk}.$$

We concluded the proof, since for all $1 \leq k \leq p$,

$$(A)_{ik}(B)_{ki} = (B)_{ki}(A)_{jk}$$

and thus for all $1 \le i \le n$ and $1 \le j \le m$,

$$((A \cdot B)^{\top})_{ij} = (B^{\top} \cdot A^{\top})_{ij}.$$

Transpose and addition

Exercise. Prove that for all $A, B \in \mathbb{R}^{m \times n}$,

$$(A+B)^{\top} = A^{\top} + B^{\top}.$$

§2.2: Systems of linear equations

Systems of linear equations

Problem. Determine all solutions $(x_1, x_2) \in \mathbb{R}$ that satisfy the following system of linear equations, for all $(\alpha, \beta) \in \{(6, 2), (6, 3), (4, 2)\}$:

$$2x_1 + x_2 = 1$$
$$\alpha x_1 + \beta x_2 = 2.$$

Answer.

If
$$(\alpha, \beta) = (6, 2)$$
, then there exists 1 solution: $(x_1, x_2) = (0, 1)$.

If
$$(\alpha, \beta) = (6, 3)$$
, then there exists no solutions for (x_1, x_2) .

If
$$(\alpha, \beta) = (4, 2)$$
, then there exists ∞ many solutions for (x_1, x_2) :

$$\{(x_1, x_2) \in \mathbb{R}^2 : 2x_1 + x_2 = 1 \text{ and } 4x_1 + 2x_2 = 2\}$$

$$\left\{ \begin{pmatrix} 1-\lambda \\ 2\lambda \end{pmatrix} : \lambda \in \mathbb{R} \right\} = \left\{ \begin{pmatrix} 0 \\ 1 \end{pmatrix} + \lambda \cdot \begin{pmatrix} 1 \\ -2 \end{pmatrix} : \lambda \in \mathbb{R} \right\}.$$

Coefficient matrix and RHS

Let $m, n \in \mathbb{N}$ and $a_{ii}, b_i \in \mathbb{R}$ for all $1 \le i \le m$ and $1 \le j \le n$.

A system of linear equations in n variables is given by

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

 $a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$
 \vdots

 $a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m$

The coefficient matrix A and right hand side (RHS) b are defined as

$$A = (a_{ij})_{i,j=1}^{m,n} \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \quad \text{and} \quad \mathbf{b} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix}$$

The set of solutions is $L(A, \mathbf{b}) := \{ \mathbf{x} \in \mathbb{R}^n : A \cdot \mathbf{x} = \mathbf{b} \}$, since

$$A \cdot \mathbf{x} = \begin{pmatrix} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix} = \mathbf{b}.$$

$$\mathbf{85/134}$$

Set of solutions

Problem. Determine all solutions $(x_1, x_2) \in \mathbb{R}$ that satisfy the following system of linear equations, for all $(\alpha, \beta) \in \{(6, 2), (6, 3), (4, 2)\}$:

$$2x_1 + x_2 = 1$$
$$\alpha x_1 + \beta x_2 = 2.$$

Answer.

If $(\alpha, \beta) = (6, 2)$, then there exists 1 solution:

$$L\left(\begin{pmatrix}2&1\\6&2\end{pmatrix},\begin{pmatrix}1\\2\end{pmatrix}\right)=\left\{\mathbf{x}\in\mathbb{R}^2:\begin{pmatrix}2&1\\6&2\end{pmatrix}\cdot\begin{pmatrix}x_1\\x_2\end{pmatrix}=\begin{pmatrix}1\\2\end{pmatrix}\right\}=\left\{\begin{pmatrix}0\\1\end{pmatrix}\right\}.$$

If $(\alpha, \beta) = (6,3)$, then there exists no solutions:

$$L\left(\begin{pmatrix}2&1\\6&3\end{pmatrix},\begin{pmatrix}1\\2\end{pmatrix}\right)=\varnothing.$$

If $(\alpha, \beta) = (4, 2)$, then there exists ∞ many solutions:

$$L\left(\begin{pmatrix}2&1\\4&2\end{pmatrix},\begin{pmatrix}1\\2\end{pmatrix}\right)=\left\{\begin{pmatrix}0\\1\end{pmatrix}+\lambda\cdot\begin{pmatrix}1\\-2\end{pmatrix}:\lambda\in\mathbb{R}\right\}.$$

Homogeneous system

Let us consider the following system of equations:

$$2x_1 + x_2 = 1$$
$$4x_1 + 2x_2 = 2.$$

Recall that its coefficient matrix A and RHS b are

$$A = \begin{pmatrix} 2 & 1 \\ 4 & 2 \end{pmatrix}$$
 and $\mathbf{b} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$.

Recall that its solution set is a follows

$$L\left(\begin{pmatrix}2&1\\4&2\end{pmatrix},\begin{pmatrix}1\\2\end{pmatrix}\right) = \left\{\begin{pmatrix}0\\1\end{pmatrix} + \lambda \cdot \begin{pmatrix}1\\-2\end{pmatrix} : \lambda \in \mathbb{R}\right\}.$$

The solution set of its corresponding homogeneous system is $A \cdot \mathbf{x} = \mathbf{0}$:

$$L\left(\begin{pmatrix}2&1\\4&2\end{pmatrix},\begin{pmatrix}0\\0\end{pmatrix}\right) = \left\{\lambda \cdot \begin{pmatrix}1\\-2\end{pmatrix} : \lambda \in \mathbb{R}\right\}.$$

Existence of infinitely many solutions

Suppose that $A \in \mathbb{R}^{m \times n}$ and $\mathbf{b} \in \mathbb{R}^m$.

Lemma. If $L(A, \mathbf{b}) \neq \emptyset$ and $L(A, \mathbf{0}) \neq \{\mathbf{0}\}$, then

$$|L(A, \mathbf{b})| = \infty.$$

Proof. By assumption, there exist non-zero vectors $\mathbf{y} \in L(A, \mathbf{b})$ and $\mathbf{z} \in L(A, \mathbf{0})$ such that $A \cdot \mathbf{y} = \mathbf{b}$ and $A \cdot \mathbf{z} = \mathbf{0}$.

This implies that for all $\lambda \in \mathbb{R}$,

$$A\cdot (\mathbf{y}+\lambda\cdot \mathbf{z})=A\cdot \mathbf{y}+\lambda\cdot A\cdot \mathbf{z}=\mathbf{b}+\lambda\cdot \mathbf{0}=\mathbf{b}.$$

This concludes the proof as $\mathbf{y} + \lambda \cdot \mathbf{z}$ is a solution for each $\lambda \in \mathbb{R}$.

Lemma. If $L(A, \mathbf{0}) = \{\mathbf{0}\}$, then $|L(A, \mathbf{b})| \le 1$.

Proof. Exercise.

$\S 2.3$: Gaussian elimination

Row echelon form

The leading coefficient of a row is the first nonzero coefficient from the left.

A matrix is in row echelon form if

- nonzero rows are above zero rows, and
- the leading coefficient of a row is strictly right of the leading coefficient of the row above.

A matrix is in reduced row echelon form if it is in row echelon form, and

• the leading coefficients are 1 and the only nonzero entry in their column.

Example.

The leading coefficients of the left matrix are 1, 2 and 1.

Quizzing row echelon forms

Problem. Which of the following matrices are in row echelon form or reduced row echelon form?

$$A = \begin{pmatrix} 1 & 0 & 2 & 5 \\ 0 & 0 & -3 & \frac{1}{2} \\ 0 & 0 & 0 & 1 \end{pmatrix}, \qquad B = \begin{pmatrix} -2 & 0 & 2 & 3 & -1 \\ 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 8 & \pi \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix},$$

$$C = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \qquad D = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix},$$

$$E = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \qquad F = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}.$$

Quizzing row echelon forms

Problem. Which of the following matrices are in row echelon form or reduced row echelon form?

$$A = \begin{pmatrix} 1 & 0 & 2 & 5 \\ 0 & 0 & -3 & \frac{1}{2} \\ 0 & 0 & 0 & 1 \end{pmatrix}, \qquad B = \begin{pmatrix} -2 & 0 & 2 & 3 & -1 \\ 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 8 & \pi \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix},$$

$$C = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \qquad D = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix},$$

$$E = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \qquad F = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}.$$

Answer.

Row echelon form: *A, B, C, D*. Reduced row echelon form: *C, D*. NOT in row echelon form: *E, F*.

Problem. Determine the solution set

$$L(A, \mathbf{b}),$$

where

$$A = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \quad \text{and} \quad \mathbf{b} = \begin{pmatrix} 2 \\ 1 \\ 0 \\ 1 \end{pmatrix}$$

Problem. Determine the solution set

$$L(A, \mathbf{b}),$$

where

$$A = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \quad \text{and} \quad \mathbf{b} = \begin{pmatrix} 2 \\ 1 \\ 0 \\ 1 \end{pmatrix}$$

Answer. We convert to a system of linear equations:

$$x_1 + 2x_2 = 2$$

 $x_3 = 1$
 $0 = 0$
 $0 = 1$.

Since the last equation is never valid, there are no solutions to this system:

$$L(A, \mathbf{b}) = \varnothing.$$

Problem. Determine the solution set

$$L(A, \mathbf{b})$$
 where $A = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$ and $\mathbf{b} = \begin{pmatrix} 2 \\ 1 \\ 0 \\ 0 \end{pmatrix}$.

Problem. Determine the solution set

$$L(A, \mathbf{b})$$
 where $A = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$ and $\mathbf{b} = \begin{pmatrix} 2 \\ 1 \\ 0 \\ 0 \end{pmatrix}$.

Answer. We convert to a system of linear equations in three variables:

$$x_1 + 2x_2 = 2$$

 $x_3 = 1$
 $0 = 0$
 $0 = 0$.

We find that $x_3 = 1$ and treat either x_2 or x_1 as a *free variable*:

$$\textit{L}(\textit{A}, \mathbf{b}) = \left\{ \begin{pmatrix} 2 \\ 0 \\ 1 \end{pmatrix} + \underbrace{\textbf{x}_2}_{2} \cdot \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix} : \underline{\textbf{x}_2} \in \mathbb{R} \right\} = \left\{ \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} + \underline{\textbf{x}_1}_{1} \cdot \begin{pmatrix} 2 \\ -1 \\ 0 \end{pmatrix} : \underline{\textbf{x}_1} \in \mathbb{R} \right\}.$$

Exercise. Show that both sets are really equal.

Gaussian elimination and rank

The row operations on a matrix are defined as follows

- Interchanging two rows.
- **2** Multiplication of a row with a non-zero scalar $\lambda \in \mathbb{R}$.
- 3 Addition of multiple of row to another row.

Gaussian elimination is a sequence of elementary row operations applied to a matrix A until the output matrix C is in row echelon form.

We define $\operatorname{rank} A$ as the number of non-zero rows of C.

Exercise. Prove that the number of non-zero rows of \boldsymbol{E} does not depend on the sequence of row operations.

Exercise. Show that rank $A \leq \min\{m, n\}$.

Computing the rank

Problem. Compute rank A, where

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}.$$

Answer.

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \xrightarrow{R_2 = R_2 - 4R_1} \begin{pmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \\ 7 & 8 & 9 \end{pmatrix}$$

$$\xrightarrow{R_3 = R_3 - 7R_1} \begin{pmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \\ 0 & -6 & -12 \end{pmatrix}$$

$$\xrightarrow{R_3 = R_3 - 2R_2} \begin{pmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \\ 0 & 0 & 0 \end{pmatrix} := C$$

We find that rank A=2, since the output matrix ${\it C}$ in row echelon form and has two non-zero rows.

Equivalent matrices

Recall that a relation $R \subset M \times M$ for some non-empty set M is called an equivalence relation if for all $x, y, z \in M$ the following holds:

- reflexive: $(x, x) \in R$,
- symmetric: $(x,y) \in R \Longrightarrow (y,x) \in R$, and
- transitive: $(x, y), (y, z) \in R \Longrightarrow (x, z) \in R$.

Exercise. Let $M := \mathbb{R}^{m \times n}$ and suppose that $R \subset M \times M$ is the relation such that $(A, B) \in R$ if and only if the matrices A and B are related by a sequence of row operations. Show that R is an equivalence relation.

Matrix concatenation

The matrix concatenation (A|B) of an $m \times n$ matrix A and $m \times p$ matrix B is obtained by appending the columns of B to the columns of A. **Example.**

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 1 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix}, \quad I_3 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad \mathbf{a} := \begin{pmatrix} 2 \\ 3 \\ 5 \end{pmatrix}, \quad \mathbf{b} := \begin{pmatrix} 8 \\ 9 \\ 10 \end{pmatrix} \quad \mathbf{c} := \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix},$$
$$(A|\mathbf{b}) = \begin{pmatrix} 1 & 2 & 3 & 8 \\ 1 & 1 & 1 & 9 \\ 1 & 0 & 1 & 10 \end{pmatrix},$$
$$(A|I) = \begin{pmatrix} 1 & 2 & 3 & 1 & 0 & 0 \\ 1 & 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{pmatrix},$$

$$(\mathbf{a} \,|\, \mathbf{b} \,|\, \mathbf{c}) = \begin{pmatrix} 2 & 8 & 0 \\ 3 & 9 & 1 \\ 5 & 10 & 2 \end{pmatrix}.$$

Solving linear systems using Gaussian elimination

Theorem. Let $A \in \mathbb{R}^{m \times n}$ and $\mathbf{b} \in \mathbb{R}^n$. If the matrix $(C|\mathbf{d})$ is obtained from $(A|\mathbf{b})$ using row operations, then

$$L(A, \mathbf{b}) = L(C, \mathbf{d}).$$

Proof. Recall that the row operations are:

- Interchanging two rows.
- Multiplication of a row with a non-zero scalar $\lambda \in \mathbb{R}$.
- Addition of multiple of row to another row.

We observe that the row operations applied the matrix $(A|\mathbf{b})$ correspond to the following operations applied to the system of linear equations defined by $A \cdot \mathbf{b} = \mathbf{x}$:

- Interchanging two equations.
- Multiplying both sides of an equation with a non-zero scalar $\lambda \in \mathbb{R}$.
- Adding a multiple of an equation to another equation.

We conclude the proof as the latter three operations do not change the solution set.

98/134

Example Gaussian elimination

Problem. Determine the solution set

$$L(A, \mathbf{b})$$
 where $A = \begin{pmatrix} 3 & 5 \\ 1 & -1 \end{pmatrix}$ and $\mathbf{b} = \begin{pmatrix} 42 \\ 6 \end{pmatrix}$.

Answer.

$$\begin{pmatrix}
3 & 5 & | & 42 \\
1 & -1 & | & 6
\end{pmatrix}
\xrightarrow{R_1 \leftrightarrow R_2}
\begin{pmatrix}
1 & -1 & | & 6 \\
3 & 5 & | & 42
\end{pmatrix}$$

$$\xrightarrow{R_2 = R_2 - 3R_1}
\begin{pmatrix}
1 & -1 & | & 6 \\
0 & 8 & | & 24
\end{pmatrix}$$

$$\xrightarrow{R_2 = \frac{1}{8}R_2}
\begin{pmatrix}
1 & -1 & | & 6 \\
0 & 1 & | & 3
\end{pmatrix}$$

$$\xrightarrow{R_1 = R_1 + R_2}
\begin{pmatrix}
1 & 0 & | & 9 \\
0 & 1 & | & 3
\end{pmatrix} =: (C|\mathbf{d}).$$

The matrix $(C|\mathbf{d})$ corresponds to the system $C \cdot \mathbf{x} = \mathbf{d}$:

$$x_1 + 0x_2 = 9$$
$$0x_1 + x_2 = 3.$$

It follows from the previous theorem that $L(A, \mathbf{b}) = L(C, \mathbf{d}) = \left\{ \begin{pmatrix} 9 \\ 3 \end{pmatrix} \right\}$.

Problem. Determine the solution set

$$L(A, \mathbf{b})$$
 where $A = \begin{pmatrix} 2 & 1 & -2 \\ 0 & 3 & 6 \\ 2 & 0 & -4 \end{pmatrix}$ and $\mathbf{b} = \begin{pmatrix} 5 \\ 3 \\ 4 \end{pmatrix}$.

Answer.

$$\begin{pmatrix}
2 & 1 & -2 & | & 5 \\
0 & 3 & 6 & | & 3 \\
2 & 0 & -4 & | & 4
\end{pmatrix}
\overrightarrow{R_3} = \overrightarrow{R_3} - \overrightarrow{R_1}
\begin{pmatrix}
2 & 1 & -2 & | & 5 \\
0 & 3 & 6 & | & 3 \\
0 & -1 & -2 & | & -1
\end{pmatrix}$$

$$\overrightarrow{R_3} = -\overrightarrow{R_3}
\begin{pmatrix}
2 & 1 & -2 & | & 5 \\
0 & 3 & 6 & | & 3 \\
0 & 1 & 2 & | & 1
\end{pmatrix}$$

$$\overrightarrow{R_2} \leftrightarrow \overrightarrow{R_3}$$

$$\overrightarrow{R_2} \leftrightarrow \overrightarrow{R_3}$$

$$\begin{pmatrix}
2 & 1 & -2 & | & 5 \\
0 & 1 & 2 & | & 1 \\
0 & 3 & 6 & | & 3
\end{pmatrix}$$

$$\overrightarrow{R_3} = \overrightarrow{R_3} - 3\overrightarrow{R_2}
\begin{pmatrix}
2 & 1 & -2 & | & 5 \\
0 & 1 & 2 & | & 1 \\
0 & 0 & 0 & | & 0
\end{pmatrix}$$

$$\overrightarrow{R_1} = \overrightarrow{R_1} - \overrightarrow{R_2}
\begin{pmatrix}
2 & 0 & -4 & | & 4 \\
0 & 1 & 2 & | & 1 \\
0 & 0 & 0 & | & 0
\end{pmatrix}$$

$$\overrightarrow{R_1} = \frac{1}{2}\overrightarrow{R_1}$$

$$\begin{pmatrix}
1 & 0 & -2 & | & 2 \\
0 & 1 & 2 & | & 1 \\
0 & 0 & 0 & | & 0
\end{pmatrix}$$

$$= (C, d).$$

$$x_1 + 2x_3 = 2$$

 $x_2 + 2x_3 = 1$

$$\textit{L}(\textit{A},\textbf{b}) = \textit{L}(\textit{C},\textbf{d}) = \left\{ \begin{pmatrix} 2 + 2x_3 \\ 1 - 2x_3 \\ x_3 \end{pmatrix} : x_3 \in \mathbb{R} \right\} = \left\{ \begin{pmatrix} 2 \\ 1 \\ 0 \end{pmatrix} + \lambda \cdot \begin{pmatrix} 2 \\ -2 \\ 1 \end{pmatrix} : \lambda \in \mathbb{R} \right\}.$$

100/134

Example. Let $A \in \mathbb{R}^{4 \times 3}$ and $\mathbf{b} \in \mathbb{R}^4$.

Suppose that (C, \mathbf{d}) is the reduced row echelon form of (A, \mathbf{b}) for some $\alpha \in \mathbb{R}$, where

$$C = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \quad \text{and} \quad \mathbf{d} = \begin{pmatrix} 2 \\ 1 \\ 0 \\ \alpha \end{pmatrix}$$

Let consider the system of linear equations $C \cdot \mathbf{x} = \mathbf{d}$:

$$x_1 + 2x_2 = 2$$

$$x_3 = 1$$

$$0 = 0$$

$$0 = \alpha$$

Notice that

rank
$$A = 2$$
 and $L(A, \mathbf{b}) = L(C, \mathbf{d})$.

If $\alpha = 0$, then

$$|L(A, \mathbf{b})| = \infty.$$

If $\alpha \neq 0$, then

$$L(A, \mathbf{b}) = \emptyset.$$

Rank and reduced echelon form

Theorem. Suppose that $A \in \mathbb{R}^{m \times n}$. Then:

rank
$$A = n = m \iff |L(A, \mathbf{b})| = 1$$
 for all $\mathbf{b} \in \mathbb{R}^n$.

Proof. Let (C, \mathbf{d}) be the reduced row echelon form of $(A|\mathbf{b})$ so that

$$L(A, \mathbf{b}) = L(C, \mathbf{d}).$$

 (\Longrightarrow) Notice that $C = I_n$ as a direct consequence of the definitions of "rank" and "reduced row echelon form".

Since $|L(I_n, \mathbf{d})| = 1$ for all $\mathbf{d} \in \mathbb{R}^n$ and $L(A, \mathbf{b}) = L(C, \mathbf{d})$, we concluded the proof.

(\iff) We find that $|L(A,\mathbf{d})|=|L(C,\mathbf{d})|=1$ for all $\mathbf{d}\in\mathbb{R}^n$. Suppose by contradiction that $C\neq I_n$. In this case one of the variables for the system $C\cdot\mathbf{x}=\mathbf{d}$ is a free variable and thus either

$$|L(C, \mathbf{d})| = \infty$$
 or $L(C, \mathbf{d}) = \emptyset$ (see example previous slide).

We arrived at a contradiction and thus $C = I_n$.

Hence, rank A = n = m as was to be shown.

Elementary matrices

An elementary matrix is a matrix that can be obtained from the identity matrix by a single row operation.

Example. Elementary 3×3 matrices for the row operations

$$R_2\leftrightarrow R_1$$
, $R_2=3R_2$ and $R_3=R_3+4R_1$:

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \qquad \begin{pmatrix} 1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \qquad \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 4 & 0 & 1 \end{pmatrix}.$$

The row operation $R_2 = R_2 + 3R_1$ on a matrix can be realized by multiplying with an elementary matrix:

$$\begin{pmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 & -2 \\ -3 & 1 & 4 \\ 2 & -3 & 4 \end{pmatrix} = \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & -2 \\ 2 & -3 & 4 \end{pmatrix}.$$

Elementary matrices

An elementary matrix is a matrix that can be obtained from the identity matrix by a single row operation.

Example. Elementary 3×3 matrices for the row operations $R_2 \leftrightarrow R_1$, $R_2 = 3R_2$ and $R_3 = R_3 + 4R_1$:

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \qquad \begin{pmatrix} 1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \qquad \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 4 & 0 & 1 \end{pmatrix}.$$

The row operation $R_2 = R_2 + 3R_1$ on a matrix can be realized by multiplying with an elementary matrix:

$$\begin{pmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 & -2 \\ -3 & 1 & 4 \\ 2 & -3 & 4 \end{pmatrix} = \begin{pmatrix} 1 & 0 & -2 \\ 0 & 1 & -2 \\ 2 & -3 & 4 \end{pmatrix}.$$

Remark. If $A \in \mathbb{R}^{m \times n}$ is a matrix and C its reduced row echelon form, then by definition there exists elementary matrices E_1, \ldots, E_k such that

$$C = E_k \cdots E_1 \cdot A$$
.

Exercise for elementary matrices

Exercise. Show that elementary matrices $E \in \mathbb{R}^{n \times n}$ are invertible and that the inverse E^{-1} is also an elementary matrix.

$\S 2.4$: Matrices as linear transformations

Matrix transformation

The matrix transformation $T_A \colon \mathbb{R}^n \to \mathbb{R}^m$ of a matrix $A \in \mathbb{R}^{m \times n}$ is defined as

$$T_A(\mathbf{x}) = A \cdot \mathbf{x}.$$

Example.

if
$$A = \begin{pmatrix} 1 & -3 \\ 3 & 5 \\ -1 & 7 \end{pmatrix} \in \mathbb{R}^{3 \times 2}$$
 and $\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \in \mathbb{R}^2$

then

$$T_A(\mathbf{x}) = A \cdot \mathbf{x} = \begin{pmatrix} 1 & -3 \\ 3 & 5 \\ -1 & 7 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 - 3x_2 \\ 3x_1 + 5x_2 \\ -x_1 + 7x_2 \end{pmatrix}.$$

106/134

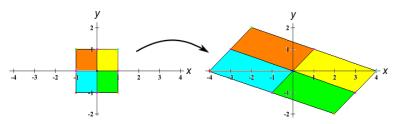
Linear maps and matrices

Example. Let the function $T: \mathbb{R}^2 \to \mathbb{R}^2$ be defined as

$$T\left(\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}\right) = \begin{pmatrix} 3x_1 + x_2 \\ -x_1 + x_2 \end{pmatrix}.$$

The function T is a matrix transformation:

$$\begin{pmatrix} 3 & 1 \\ -1 & 1 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 3x_1 + x_2 \\ -x_1 + x_2 \end{pmatrix}.$$



Composition of linear maps

Recall that if $f: \mathbb{R}^n \to \mathbb{R}^m$ and $g: \mathbb{R}^m \to \mathbb{R}^k$ are functions, then their composition is defined as the function

$$g \circ f : \mathbb{R}^n \to \mathbb{R}^k, \quad x \mapsto g(f(x)).$$

Problem. Suppose that A is an $m \times n$ matrix and B and $k \times m$ matrix with matrix transformations

$$T_A \colon \mathbb{R}^n \to \mathbb{R}^m$$
 and $T_B \colon \mathbb{R}^m \to \mathbb{R}^k$.

Determine the matrix M such that :

$$T_M = T_B \circ T_A \colon \mathbb{R}^n \to \mathbb{R}^k.$$

Composition of linear maps

Recall that if $f: \mathbb{R}^n \to \mathbb{R}^m$ and $g: \mathbb{R}^m \to \mathbb{R}^k$ are functions, then their composition is defined as the function

$$g \circ f : \mathbb{R}^n \to \mathbb{R}^k, \quad x \mapsto g(f(x)).$$

Problem. Suppose that A is an $m \times n$ matrix and B and $k \times m$ matrix with matrix transformations

$$T_A \colon \mathbb{R}^n \to \mathbb{R}^m$$
 and $T_B \colon \mathbb{R}^m \to \mathbb{R}^k$.

Determine the matrix M such that :

$$T_M = T_B \circ T_A \colon \mathbb{R}^n \to \mathbb{R}^k.$$

Answer. The $k \times n$ matrix

$$M = B \cdot A$$
.

Linear transformation

A function $T: \mathbb{R}^n \to \mathbb{R}^m$ is a linear transformation if the following holds:

• For all $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$,

$$T(\mathbf{x} + \mathbf{y}) = T(\mathbf{x}) + T(\mathbf{y}).$$

• For all $\mathbf{x} \in \mathbb{R}^n$ and all $\lambda \in \mathbb{R}$,

$$T(\lambda \cdot \mathbf{x}) = \lambda \cdot T(\mathbf{x}).$$

Theorem. A function $T: \mathbb{R}^n \to \mathbb{R}^m$ is a linear transformation if and only if there exists a matrix $A \in \mathbb{R}^{m \times n}$ with $T = T_A$.

Proof. Exercise.

Hint. For the ← direction show that

$$A = (T(\mathbf{e}_1) \cdots T(\mathbf{e}_n))$$

for some linear transformation $T: \mathbb{R}^n \to \mathbb{R}^m$.

Lemma for invertible linear transformations

Lemma. If $A \in \mathbb{R}^{n \times n}$ is an invertible matrix, then T_A is an invertible function.

Proof. By definition

$$A \cdot A^{-1} = I_n = A^{-1} \cdot A$$

and thus

$$T_A \circ T_{A^{-1}} = T_{A \cdot A^{-1}} = T_{I_n} = \mathrm{id} \quad \text{and} \quad T_{A^{-1}} \circ T_A = T_{A^{-1} \cdot A} = T_{I_n} = \mathrm{id} \ .$$

Lemma. If $A \in \mathbb{R}^{n \times n}$ is a matrix such that T_A is invertible, then its inverse $T := T_A^{-1}$ is a linear transformation.

Proof. Let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ and $\lambda \in \mathbb{R}$ be arbitrary.

By definition of the inverse,

$$(T_A \circ T)(\mathbf{x}) = \mathbf{x} = (T \circ T_A)(\mathbf{x})$$
 and $(T_A \circ T)(\mathbf{y}) = \mathbf{y} = (T \circ T_A)(\mathbf{y}).$

Therefore,

$$T(\mathbf{x} + \mathbf{y}) = T((T_A \circ T)(\mathbf{x}) + (T_A \circ T)(\mathbf{y}))$$

$$= T(A \cdot T(\mathbf{x}) + A \cdot T(\mathbf{y}))$$

$$= T(A \cdot (T(\mathbf{x}) + T(\mathbf{y})))$$

$$= (T \circ T_A)(T(\mathbf{x}) + T(\mathbf{y}))$$

$$= T(\mathbf{x}) + T(\mathbf{y}).$$

Similarly,

$$T(\lambda \cdot \mathbf{x}) = T(\lambda \cdot (T_A \circ T)(\mathbf{x}))$$

$$= T(\lambda \cdot A \cdot T(\mathbf{x}))$$

$$= T(A \cdot \lambda \cdot T(\mathbf{x}))$$

$$= (T \circ T_A)(\lambda \cdot T(\mathbf{x}))$$

$$= \lambda \cdot T(\mathbf{x}).$$

Hence, \mathcal{T} is a linear transformation and thus we concluded the proof.

Lemma. If $A \in \mathbb{R}^{n \times n}$ is a matrix such that $T_A \colon \mathbb{R}^n \to \mathbb{R}^n$ is invertible, then A is an invertible matrix.

Proof. By definition of the inverse, there exists a function $T: \mathbb{R}^n \to \mathbb{R}^n$ such that

$$T_A \circ T = id = T \circ T_A$$
.

It follows from the previous lemma that T is a linear transformation. By the previous theorem, there exists a matrix $B \in \mathbb{R}^{n \times n}$ such that

$$T=T_B$$
.

Therefore,

$$id = T_A \circ T_B = T_{A \cdot B} = T_{I_n}$$
 and $id = T_B \circ T_A = T_{B \cdot A} = T_{I_n}$.

Hence,

$$A \cdot B = I_n = B \cdot A$$

and thus

$$A^{-1}=B.$$

Invertible linear transformations

The following theorem summarizes the previous three lemma's:

Theorem. Let $A \in \mathbb{R}^{n \times n}$ be a square matrix. Then

 T_A is invertible \iff A is invertible.

§2.5: Determinants

Matrix determinant

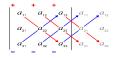
The determinant det M of a $n \times n$ matrix M is defined as follows if $n \le 3$,

$$det(a) = a$$

$$\det\begin{pmatrix} a & b \\ c & d \end{pmatrix} = a \cdot d - b \cdot c$$

$$\det\begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} = a_{11} \cdot a_{22} \cdot a_{33} + a_{12} \cdot a_{23} \cdot a_{31} + a_{13} \cdot a_{21} \cdot a_{32} - a_{31} \cdot a_{22} \cdot a_{13} - a_{32} \cdot a_{23} \cdot a_{11} - a_{33} \cdot a_{21} \cdot a_{12}.$$

The determinant for 3×3 matrices can be remembered using Sarrus rule:



Proposition. The absolute value of the determinant is equal to the volume of the area of a parallelogram (left) or volume of a parallelepiped (right):





 $|\det(r_1|r_2|r_3)|$

Computing determinant

Problem. Compute the following determinant using expansion along the first row

$$A = \begin{pmatrix} 3 & 2 & 1 \\ 6 & 7 & 8 \\ 1 & -1 & -1 \end{pmatrix}.$$

Answer.

$$\det\begin{pmatrix} 3 & 2 & 1 \\ 6 & 7 & 8 \\ 1 & -1 & -1 \end{pmatrix} = 3 \cdot \det\begin{pmatrix} 7 & 8 \\ -1 & -1 \end{pmatrix} - 2 \cdot \det\begin{pmatrix} 6 & 8 \\ 1 & -1 \end{pmatrix} + 1 \cdot \det\begin{pmatrix} 6 & 7 \\ 1 & -1 \end{pmatrix}$$

$$=$$

$$3 \cdot [7 \cdot (-1) - 8 \cdot (-1)] - 2 \cdot [6 \cdot (-1) - 8 \cdot 1] + 1 \cdot [6 \cdot (-1) - 7 \cdot 1]$$

$$=$$

$$3 \cdot 1 - 2 \cdot (-14) + 1 \cdot (-13) = 18.$$

Computing determinant

We may compute the determinant by expanding along any column or row:

Theorem. If $A \in \mathbb{R}^{n \times n}$ with $n \ge 2$, then for any $1 \le i, j \le n$,

$$\det A = \sum_{j=1}^n (-1)^{i+j} \cdot a_{ij} \cdot \det A_{ij} \quad \text{and} \quad \det A = \sum_{i=1}^n (-1)^{i+j} \cdot a_{ij} \cdot \det A_{ij},$$

where the matrix A_{ij} is the matrix A with the ith row and jth column removed.

Example.

$$\det \begin{pmatrix} 1 & 2 & 100 \\ 3 & 5 & \sqrt{2} \\ 0 & 0 & 1 \end{pmatrix} = \frac{0}{0} \cdot \det A_{31} - \frac{0}{0} \cdot \det A_{32} + 1 \cdot \det A_{33}$$
$$= 1 \cdot (1 \cdot 5 - 2 \cdot 3) = -1.$$

Exercise. Show that for all $n \times n$ matrices A,

$$\det A^{\top} = A$$
.

Determinant of a matrix

Example. Computing determinant of 4×4 matrix by expanding along the first column:

$$\det \begin{pmatrix} a & b & c & d \\ e & f & g & h \\ i & j & k & l \\ m & n & o & p \end{pmatrix} = a \cdot \det A_{11} - e \cdot \det A_{21} + i \cdot \det A_{31} - m \cdot \det A_{41}$$

$$a \cdot \det \begin{pmatrix} f & g & h \\ j & k & l \\ n & o & p \end{pmatrix} - e \cdot \det \begin{pmatrix} b & c & d \\ j & k & l \\ n & o & p \end{pmatrix} + i \cdot \det \begin{pmatrix} b & c & d \\ f & g & h \\ n & o & p \end{pmatrix} - m \cdot \det \begin{pmatrix} b & c & d \\ f & g & h \\ j & k & l \end{pmatrix}$$

$$\det A_{11} = \det \begin{pmatrix} f & g & h \\ j & k & l \\ n & o & p \end{pmatrix} = f \cdot \det \begin{pmatrix} k & l \\ o & p \end{pmatrix} - j \cdot \det \begin{pmatrix} g & h \\ o & p \end{pmatrix} + n \cdot \det \begin{pmatrix} g & h \\ k & l \end{pmatrix}$$

$$\det \begin{pmatrix} k & l \\ o & p \end{pmatrix} = k \cdot \det (p) - o \cdot \det (l)$$

$$\det(p) = p$$

Computing determinant

Problem. Compute the determinant of the following matrix

$$A := \begin{pmatrix} 3 & -7 & 8 & 9 & -6 \\ 0 & 2 & -5 & 7 & 3 \\ 0 & 0 & 1 & 5 & 0 \\ 0 & 0 & 2 & 4 & -1 \\ 0 & 0 & 0 & -2 & 0 \end{pmatrix}.$$

Computing determinant

Problem. Compute the determinant of the following matrix

$$A := \begin{pmatrix} 3 & -7 & 8 & 9 & -6 \\ 0 & 2 & -5 & 7 & 3 \\ 0 & 0 & 1 & 5 & 0 \\ 0 & 0 & 2 & 4 & -1 \\ 0 & 0 & 0 & -2 & 0 \end{pmatrix}.$$

Answer.

$$\det\begin{pmatrix} 3 & -7 & 8 & 9 & -6 \\ 0 & 2 & -5 & 7 & 3 \\ 0 & 0 & 1 & 5 & 0 \\ 0 & 0 & 2 & 4 & -1 \\ 0 & 0 & 0 & -2 & 0 \end{pmatrix} = 3 \cdot \det\begin{pmatrix} 2 & -5 & 7 & 3 \\ 0 & 1 & 5 & 0 \\ 0 & 2 & 4 & -1 \\ 0 & 0 & -2 & 0 \end{pmatrix}$$
$$= 3 \cdot 2 \cdot \det\begin{pmatrix} 1 & 5 & 0 \\ 2 & 4 & -1 \\ 0 & -2 & 0 \end{pmatrix} = 3 \cdot 2 \cdot (-1 \cdot -1) \cdot \det\begin{pmatrix} 1 & 5 \\ 0 & -2 \end{pmatrix}$$

$$= 3 \cdot 2 \cdot (-1 \cdot -1) \cdot (-2) = -12.$$
 119/134

Upper and lower triangular matrix

An $m \times n$ matrix is triangular if all the entries either below or above the main diagonal are zero.

Example. Determinant of an upper triangular matrix:

$$\det \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} & a_{23} \\ 0 & 0 & a_{33} \end{pmatrix} = a_{11} \cdot a_{22} \cdot a_{33}.$$

Lemma. If $A = (a_{ij})$ is an $n \times n$ triangular matrix, then

$$\det A = a_{11} \cdot a_{22} \cdots a_{nn}.$$

Proof. Exercise.

Determinant and multiplying row by scalar

Lemma. If the matrix B is obtained from $A \in \mathbb{R}^{n \times n}$ by multiplying a row with $\lambda \in \mathbb{R}^n$ such that $n \geq 2$, then

$$\det B = \lambda \cdot \det A$$
.

Proof. Suppose the *i*th row of *A* is multiplied with λ so that for all $1 \le j \le n$:

$$b_{ij} = \lambda \cdot a_{ij}$$
 and $\det B_{ij} = \det A_{ij}$.

By expanding along the *i*th row, we obtain,

$$\det B = \sum_{j=1}^{n} (-1)^{i+j} \cdot b_{ij} \cdot \det B_{ij}$$

$$= \sum_{j=1}^{n} (-1)^{i+j} \cdot \lambda \cdot a_{ij} \cdot \det B_{ij}$$

$$= \sum_{j=1}^{n} (-1)^{i+j} \cdot \lambda \cdot a_{ij} \cdot \det A_{ij}$$

$$= \lambda \cdot \sum_{j=1}^{n} (-1)^{i+j} \cdot a_{ij} \cdot \det A_{ij}$$

$$= \lambda \cdot \det A.$$

Determinant and interchanging rows

Lemma. If B is obtained by interchanging two rows of a matrix $A \in \mathbb{R}^{n \times n}$ such that $n \ge 2$, then

$$\det B = -\det A$$
.

Proof. We apply induction on n.

Induction basis: If n = 2, then det $A = -\det B$ (Exercise).

Induction step: Let $n \ge 3$. There exists $1 \le i \le n$ such that the *i*th row of A is equal to the *i*th row of B. This implies that for all $1 \le j \le n$,

$$b_{ij}=a_{ij}$$
.

Moreover, $B_{ij} \in \mathbb{R}^{(n-1)\times (n-1)}$ is for all $1 \leq j \leq n$ obtained from the matrix A_{ij} by interchanging two rows. Hence, by the induction hypothesis,

$$\det B_{ij} = -\det A_{ij}.$$

By expanding along the ith row, we obtain,

$$\det B = \sum_{j=1}^{n} (-1)^{i+j} \cdot b_{ij} \cdot \det B_{ij} = \sum_{j=1}^{n} (-1)^{i+j} \cdot a_{ij} \cdot \det B_{ij}$$

$$= \sum_{j=1}^{n} (-1)^{i+j} \cdot a_{ij} \cdot (-\det A_{ij}) = -\sum_{j=1}^{n} (-1)^{i+j} \cdot a_{ij} \cdot \det A_{ij} = -\det A.$$

Determinant and row operations

Lemma.

• If the matrix B is obtained from $A \in \mathbb{R}^{n \times n}$ by multiplying a row with $\lambda \in \mathbb{R}^n$ such that $n \geq 2$, then

$$\det B = \lambda \cdot \det A$$
.

• If *B* is obtained by interchanging two rows of a matrix $A \in \mathbb{R}^{n \times n}$ such that $n \ge 2$, then

$$\det B = -\det A$$
.

 If the matrix B is obtained from A by adding a multiple of one row of A to another row, then

$$\det B = \det A$$
.

Proof for 3th item. Exercise. Hint: use induction.

Corollary. If the matrix B is obtained from $A \in \mathbb{R}^{n \times n}$ using row operations, then

$$\det A = 0 \iff \det B = 0.$$
 123/134

Computing determinant using row operations

Example. Computing the determinant using Gaussian elimination:

$$\det\begin{pmatrix} 1 & -4 & 2 \\ -2 & 8 & -9 \\ -1 & 7 & 0 \end{pmatrix}$$

$$= \det\begin{pmatrix} 1 & -4 & 2 \\ 0 & 0 & -5 \\ 0 & 3 & 2 \end{pmatrix}$$

$$= -\det\begin{pmatrix} 1 & -4 & 2 \\ 0 & 3 & 2 \\ 0 & 0 & -5 \end{pmatrix}$$

$$= (-1) \cdot 1 \cdot 3 \cdot (-5) = 15.$$

Notice that we applied the row operations

$$R_2 = R_2 + 2R_1,$$
 $R_3 = R_3 + R_1$ and $R_2 \leftrightarrow R_3.$

Determinant and invertible matrices

Theorem. If $A \in \mathbb{R}^{n \times n}$ is a square matrix, then the following are equivalent

- A is invertible.
- $\det A \neq 0$.
- rank A = n.
- The linear system $A \cdot \mathbf{x} = \mathbf{b}$ has a unique solution for all $\mathbf{b} \in \mathbb{R}^n$.
- A is equivalent via row operations to the identity matrix I_n .

Proof. See lecture notes.

Determining whether a matrix is invertible

Problem. Determine whether the following matrix is invertible:

$$A := \begin{pmatrix} 3 & -7 & 8 & 9 & -6 \\ 0 & 2 & -5 & 7 & 3 \\ 0 & 0 & 1 & 5 & 0 \\ 0 & 0 & 2 & 4 & -1 \\ 0 & 0 & 0 & -2 & 0 \end{pmatrix}$$

Determining whether a matrix is invertible

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Answer. We already computed $\det A = -12$ and thus by the previous theorem this matrix must be invertible.

Multiplying with elementary matrices

Lemma. If $E \in \mathbb{R}^{n \times n}$ is an elementary matrix and $B \in \mathbb{R}^{n \times n}$ any matrix, then

$$\det(E \cdot B) = \det(E) \cdot \det(B)$$

Proof. By definition, E is obtained from the identity matrix I_n via a single row operation. Moreover, if we apply this row operation to B, then we obtain the matrix $E \cdot B$.

We make a case distinction on the type of row operation.

• If E is obtained from I_n by interchanging two rows, then by a previous lemma

$$\det(E\cdot B)=-\det B$$
 and $\det E=-\det I_n=-1.$

• If E is obtained from I_n by a row dilation λ , then by a previous lemma

$$\det(E \cdot B) = \lambda \cdot \det B$$
 and $\det(E) = \lambda \cdot \det I_n = \lambda$.

 If E is obtained from I_n by adding a multiple of a row to another row, then by a previous lemma

$$det(E \cdot B) = det B$$
 and $det E = det I_n = 1$.

This concludes the proof as we showed the equality in each of the three cases.

127/134

The determinant is compatible with multiplication

Theorem. For all square matrices $A, B \in \mathbb{R}^{n \times n}$,

$$\det(A \cdot B) = \det(A) \cdot \det(B).$$

Proof. By properties of Gaussian elimination, there exists elementary matrices $E_j, \ldots, E_1, F_k \cdots F_1$ and unique reduced echelon forms C_A and C_B such that

$$A = E_j \cdots E_1 \cdot C_A$$
 and $B = F_k \cdots F_1 \cdot C_B$.

By the repeated applications of the previous lemma, we have

$$\det A = \det(E_j) \cdot \det(E_{j-1} \cdots E_1 \cdot C_A) = \dots = \det(E_j) \cdots \det(E_1) \cdot \det(C_A),$$

$$\det B = \det(F_k) \cdot \det(F_{k-1} \cdots F_1 \cdot C_B) = \dots = \det(F_k) \cdots \det(F_1) \cdot \det(C_B),$$

and thus

$$\det(A \cdot B)$$

$$= \det(E_j \cdots E_1 \cdot C_A \cdot F_k \cdots F_1 \cdot C_B)$$

$$= \det(E_j) \cdots \det(E_1) \cdot \det(C_A) \cdot \det(F_k) \cdots \det(F_1) \cdot \det(C_B)$$

$$= \det(A) \cdot \det(B).$$

Remark. Notice that $det(A) = det(C_A) = 0$ or $det(B) = det(C_B) = 0$ if and only if $det(A \cdot B) = 0$.

Computing the determinant of a matrix product

Problem. Compute det *A*, where

$$A = \begin{pmatrix} 15 & 10 & 24 \\ 15 & 22 & 12 \\ 12 & 4 & 33 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 4 \\ 3 & 4 & 0 \\ 2 & 0 & 5 \end{pmatrix} \cdot \begin{pmatrix} 1 & 2 & 4 \\ 3 & 4 & 0 \\ 2 & 0 & 5 \end{pmatrix}.$$

Computing the determinant of a matrix product

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$$A = \begin{pmatrix} 15 & 10 & 24 \\ 15 & 22 & 12 \\ 12 & 4 & 33 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 4 \\ 3 & 4 & 0 \\ 2 & 0 & 5 \end{pmatrix} \cdot \begin{pmatrix} 1 & 2 & 4 \\ 3 & 4 & 0 \\ 2 & 0 & 5 \end{pmatrix}.$$

Answer. By the previous theorem,

$$\det A = \left(\det \begin{pmatrix} 1 & 2 & 4 \\ 3 & 4 & 0 \\ 2 & 0 & 5 \end{pmatrix}\right)^2$$

By expansion along the third row, we find that

$$\det \begin{pmatrix} 1 & 2 & 4 \\ 3 & 4 & 0 \\ 2 & 0 & 5 \end{pmatrix} = 2 \cdot \det \begin{pmatrix} 2 & 4 \\ 4 & 0 \end{pmatrix} + 5 \cdot \det \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} = 2 \cdot (-16) + 5 \cdot (-2) = -42.$$

Therefore,

$$\det A = (-42)^2 = 1764.$$

§2.6: Inverse matrices

Computing the inverse via Gaussian elimination

Theorem. If $A \in \mathbb{R}^{n \times n}$ is a matrix with inverse A^{-1} , then the reduced row echelon form of $(A|I_n)$ is equal to $(I_n|A^{-1})$.

Proof. Let \mathbf{c}_i denote the *i*th column of A^{-1} , so that

$$A^{-1} = (\mathbf{c}_1 | \cdots | \mathbf{c}_n)$$
 and $\mathbf{c}_i = A^{-1} \cdot \mathbf{e}_i$.

Hence,

$$A \cdot \mathbf{c}_i = \mathbf{e}_i$$

and thus c_i is the unique solution to the system

$$A \cdot \mathbf{x} = \mathbf{e}_i$$
.

This implies that $(I_n|\mathbf{c}_i)$ is the reduced row echelon form of

$$(A|\mathbf{e}_i).$$

Instead of computing the reduced row echelon form of $(A|\mathbf{e}_i)$ separately for all $1 \le i \le n$, we compute in one go the reduced row echelon form of the matrix

$$(A|\mathbf{e}_1|\cdots|\mathbf{e}_n)=(A|I_n),$$

which is

$$(I_n|\mathbf{c}_1|\cdots|\mathbf{c}_n)=(I_n|A^{-1}).$$

Problem. Compute the inverse of

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$$
.

Problem. Compute the inverse of

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$$
.

Answer. By Gaussian elimination

$$\begin{pmatrix}
1 & 2 & 1 & 0 \\
3 & 4 & 0 & 1
\end{pmatrix}
\overrightarrow{R_2 = R_2 - 3R_1}
\begin{pmatrix}
1 & 2 & 1 & 0 \\
0 & -2 & -3 & 1
\end{pmatrix}$$

$$\overrightarrow{R_1 = R_1 + R_2}
\begin{pmatrix}
1 & 0 & -2 & 1 \\
0 & -2 & -3 & 1
\end{pmatrix}$$

$$\overrightarrow{R_2 = (-1/2)R_2}
\begin{pmatrix}
1 & 0 & -2 & 1 \\
0 & 1 & 3/2 & -1/2
\end{pmatrix}.$$

It therefore follows from the previous theorem that

$$A^{-1} = \begin{pmatrix} -2 & 1 \\ 3/2 & -1/2 \end{pmatrix}.$$

Problem. Compute the inverse of

$$\begin{pmatrix} 1 & 0 & 2 \\ 4 & 1 & 8 \\ 0 & 1 & 1 \end{pmatrix}.$$

Problem. Compute the inverse of

$$\begin{pmatrix} 1 & 0 & 2 \\ 4 & 1 & 8 \\ 0 & 1 & 1 \end{pmatrix}.$$

Answer. By Gaussian elimination

$$\begin{pmatrix} 1 & 0 & 2 & 1 & 0 & 0 \\ 4 & 1 & 8 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 1 \end{pmatrix} \xrightarrow{R_2 = R_2 - 4R_1} \begin{pmatrix} 1 & 0 & 2 & 1 & 0 & 0 \\ 0 & 1 & 0 & -4 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 1 \end{pmatrix}$$

$$\xrightarrow{R_3 = R_3 - R_2} \begin{pmatrix} 1 & 0 & 2 & 1 & 0 & 0 \\ 0 & 1 & 0 & -4 & 1 & 0 \\ 0 & 0 & 1 & 4 & -1 & 1 \end{pmatrix}$$

$$\xrightarrow{R_1 = R_1 - 2R_3} \begin{pmatrix} 1 & 0 & 0 & -7 & 2 & -2 \\ 0 & 1 & 0 & -4 & 1 & 0 \\ 0 & 0 & 1 & 4 & -1 & 1 \end{pmatrix}.$$

It therefore follows from the previous theorem that

$$A^{-1} = \begin{pmatrix} -7 & 2 & -2 \\ -4 & 1 & 0 \\ 4 & -1 & 1 \end{pmatrix}.$$

Last slide