

# Mathematics II

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- Matrix calculus

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- Functions of several variables

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- Antiderivative and the Riemann Integral

V.1.  $\mathbb{R}^n$  as a linear and metric space

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## Definition

The set  $\mathbb{R}^n$ ,  $n \in \mathbb{N}$ , is the set of all ordered  $n$ -tuples of real numbers, i.e.

$$\mathbb{R}^n = \{[x_1, \dots, x_n] : x_1, \dots, x_n \in \mathbb{R}\}.$$

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For  $\mathbf{x} = [x_1, \dots, x_n] \in \mathbb{R}^n$ ,  $\mathbf{y} = [y_1, \dots, y_n] \in \mathbb{R}^n$  and  $\alpha \in \mathbb{R}$  we set

$$\mathbf{x} + \mathbf{y} = [x_1 + y_1, \dots, x_n + y_n], \quad \alpha \mathbf{x} = [\alpha x_1, \dots, \alpha x_n].$$

Further, we denote  $\mathbf{o} = [0, \dots, 0]$  – the **origin**.

## Definition

The **Euclidean metric (distance)** on  $\mathbb{R}^n$  is the function  $\rho: \mathbb{R}^n \times \mathbb{R}^n \rightarrow [0, +\infty)$  defined by

$$\rho(\mathbf{x}, \mathbf{y}) = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}.$$

The number  $\rho(\mathbf{x}, \mathbf{y})$  is called the **distance of the point  $\mathbf{x}$  from the point  $\mathbf{y}$** .



## Theorem 1 (properties of the Euclidean metric)

*The Euclidean metric  $\rho$  has the following properties:*

(i)  $\forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n: \rho(\mathbf{x}, \mathbf{y}) = 0 \Leftrightarrow \mathbf{x} = \mathbf{y},$

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- (iii)  $\forall \mathbf{x}, \mathbf{y}, \mathbf{z} \in \mathbb{R}^n: \rho(\mathbf{x}, \mathbf{y}) \leq \rho(\mathbf{x}, \mathbf{z}) + \rho(\mathbf{z}, \mathbf{y})$ ,  
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(triangle inequality)
- (iv)  $\forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n, \forall \lambda \in \mathbb{R}: \rho(\lambda \mathbf{x}, \lambda \mathbf{y}) = |\lambda| \rho(\mathbf{x}, \mathbf{y})$ ,  
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(homogeneity)
- (v)  $\forall \mathbf{x}, \mathbf{y}, \mathbf{z} \in \mathbb{R}^n: \rho(\mathbf{x} + \mathbf{z}, \mathbf{y} + \mathbf{z}) = \rho(\mathbf{x}, \mathbf{y})$ .  
(translation invariance)

## Definition

Let  $\mathbf{x} \in \mathbb{R}^n$ ,  $r \in \mathbb{R}$ ,  $r > 0$ . The set  $B(\mathbf{x}, r)$  defined by

$$B(\mathbf{x}, r) = \{\mathbf{y} \in \mathbb{R}^n; \rho(\mathbf{x}, \mathbf{y}) < r\}$$

is called an **open ball with radius  $r$  centred at  $\mathbf{x}$**  or the **neighbourhood of  $\mathbf{x}$** .

## Definition

Let  $M \subset \mathbb{R}^n$ . We say that  $\mathbf{x} \in \mathbb{R}^n$  is an **interior point of  $M$** , if there exists  $r > 0$  such that  $B(\mathbf{x}, r) \subset M$ .

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The set  $M \subset \mathbb{R}^n$  is **open in  $\mathbb{R}^n$** , if each point of  $M$  is an interior point of  $M$ , i.e. if  $M = \text{Int } M$ .

## Theorem 2 (properties of open sets)

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- (iii) *Let  $G_i \subset \mathbb{R}^n$ ,  $i = 1, \dots, m$ , be open in  $\mathbb{R}^n$ . Then  $\bigcap_{i=1}^m G_i$  is open in  $\mathbb{R}^n$ .*

## Remark

- (ii) *A union of an arbitrary system of open sets is an open set.*
- (iii) *An intersection of a finitely many open sets is an open set.*

## Definition

Let  $M \subset \mathbb{R}^n$  and  $\mathbf{x} \in \mathbb{R}^n$ . We say that  $\mathbf{x}$  is a **boundary point of  $M$**  if for each  $r > 0$

$$B(\mathbf{x}, r) \cap M \neq \emptyset \quad \text{and} \quad B(\mathbf{x}, r) \cap (\mathbb{R}^n \setminus M) \neq \emptyset.$$

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A set  $M \subset \mathbb{R}^n$  is said to be **closed in  $\mathbb{R}^n$**  if it contains all its boundary points, i.e. if  $\text{bd } M \subset M$ , or in other words if  $\overline{M} = M$ .



# VI. Matrix calculus

# VI.1. Basic operations with matrices

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## Definition

A table of numbers

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

where  $a_{ij} \in \mathbb{R}$ ,  $i = 1, \dots, m$ ,  $j = 1, \dots, n$ , is called a **matrix of type  $m \times n$**  (shortly, an  **$m$ -by- $n$  matrix**). We also write  $(a_{ij})_{\substack{i=1..m \\ j=1..n}}$  for short.

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The set of all  $m$ -by- $n$  matrices is denoted by  **$M(m \times n)$** .

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Let

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

The  $n$ -tuple  $(a_{i1}, a_{i2}, \dots, a_{in})$ , where  $i \in \{1, 2, \dots, m\}$ , is called the  **$i$ th row** of the matrix  $\mathbf{A}$ .

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## Definition

We say that two matrices are equal, if they are of the same type and the corresponding elements are equal, i.e. if  $\mathbf{A} = (a_{ij})_{\substack{i=1..m \\ j=1..n}}$  and  $\mathbf{B} = (b_{uv})_{\substack{u=1..r \\ v=1..s}}$ , then  $\mathbf{A} = \mathbf{B}$  if and only if  $m = r$ ,  $n = s$  and  $a_{ij} = b_{ij} \forall i \in \{1, \dots, m\}, \forall j \in \{1, \dots, n\}$ .



## Definition

Let  $\mathbf{A}, \mathbf{B} \in M(m \times n)$ ,  $\mathbf{A} = (a_{ij})_{\substack{i=1..m \\ j=1..n}}$ ,  $\mathbf{B} = (b_{ij})_{\substack{i=1..m \\ j=1..n}}$ ,  $\lambda \in \mathbb{R}$ .

The **sum of the matrices  $\mathbf{A}$  and  $\mathbf{B}$**  is the matrix defined by

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

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The **product of the real number  $\lambda$  and the matrix  $\mathbf{A}$**  (or the  $\lambda$ -multiple of the matrix  $\mathbf{A}$ ) is the matrix defined by

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

## Proposition 3 (basic properties of the sum of matrices and of a multiplication by a scalar)

*The following holds:*

- $\forall \mathbf{A}, \mathbf{B}, \mathbf{C} \in M(m \times n): \mathbf{A} + (\mathbf{B} + \mathbf{C}) = (\mathbf{A} + \mathbf{B}) + \mathbf{C},$   
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- $\forall \mathbf{A} \in M(m \times n) \exists \mathbf{C}_A \in M(m \times n): \mathbf{A} + \mathbf{C}_A = \mathbf{O}$ ,  
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- $\forall \mathbf{A}, \mathbf{B}, \mathbf{C} \in M(m \times n): \mathbf{A} + (\mathbf{B} + \mathbf{C}) = (\mathbf{A} + \mathbf{B}) + \mathbf{C}$ ,  
(associativity)
- $\forall \mathbf{A}, \mathbf{B} \in M(m \times n): \mathbf{A} + \mathbf{B} = \mathbf{B} + \mathbf{A}$ , (commutativity)
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(existence of a zero element)
- $\forall \mathbf{A} \in M(m \times n) \exists \mathbf{C}_A \in M(m \times n): \mathbf{A} + \mathbf{C}_A = \mathbf{O}$ ,  
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- The matrix  $\mathbf{O}$  from the previous proposition is called a **zero matrix** and all its elements are all zeros.
- The matrix  $\mathbf{C}_A$  from the previous proposition is called a **matrix opposite to  $\mathbf{A}$** . It is determined uniquely, it is denoted by  $-\mathbf{A}$ , and it satisfies  $-\mathbf{A} = (-a_{ij})_{\substack{i=1..m \\ j=1..n}}$  and  $-\mathbf{A} = -1 \cdot \mathbf{A}$ .

## Definition

Let  $\mathbf{A} \in M(m \times n)$ ,  $\mathbf{A} = (a_{is})_{\substack{i=1..m, \\ s=1..n}}$ ,  $\mathbf{B} \in M(n \times k)$ ,  
 $\mathbf{B} = (b_{sj})_{\substack{s=1..n, \\ j=1..k}}$ . Then the **product of matrices**  $\mathbf{A}$  and  $\mathbf{B}$  is  
 defined as a matrix  $\mathbf{AB} \in M(m \times k)$ ,  $\mathbf{AB} = (c_{ij})_{\substack{i=1..m, \\ j=1..k}}$ ,  
 where

$$c_{ij} = \sum_{s=1}^n a_{is}b_{sj}.$$

# Matrix multiplication

$$\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \\ a_{41} & a_{42} \end{pmatrix} \cdot \begin{pmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \end{pmatrix}$$

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Let  $m, n, k, l \in \mathbb{N}$ . Then:

- (i)  $\forall \mathbf{A} \in M(m \times n) \forall \mathbf{B} \in M(n \times k) \forall \mathbf{C} \in M(k \times l)$ :  
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### Remark

Warning! The matrix multiplication is not commutative.



## Definition

A **transpose** of a matrix

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is the matrix

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i.e. if  $\mathbf{A} = (a_{ij})_{\substack{i=1..m \\ j=1..n}}$ , then  $\mathbf{A}^T = (b_{uv})_{\substack{u=1..n \\ v=1..m}}$ , where  $b_{uv} = a_{vu}$   
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## Definition

We say that the matrix  $\mathbf{A} \in M(n \times n)$  is **symmetric** if  $\mathbf{A} = \mathbf{A}^T$ .

# VI.2. Invertible matrices

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## Remark

A matrix  $\mathbf{A} \in M(n \times n)$  is invertible if and only if it has an inverse.

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Let  $\mathbf{A}, \mathbf{B} \in M(n \times n)$  be invertible matrices. Then

- (i)  $\mathbf{A}^{-1}$  is invertible and  $(\mathbf{A}^{-1})^{-1} = \mathbf{A}$ ,

## Remark

- If  $\mathbf{A} \in M(n \times n)$  is invertible, then it has exactly one inverse, which is denoted by  $\mathbf{A}^{-1}$ .
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- $\mathbf{AB}$  is invertible and  $(\mathbf{AB})^{-1} = \mathbf{B}^{-1}\mathbf{A}^{-1}$ .

## Definition

Let  $k, n \in \mathbb{N}$  and  $\mathbf{v}^1, \dots, \mathbf{v}^k \in \mathbb{R}^n$ . We say that a vector  $\mathbf{u} \in \mathbb{R}^n$  is a **linear combination of the vectors  $\mathbf{v}^1, \dots, \mathbf{v}^k$  with coefficients  $\lambda_1, \dots, \lambda_k \in \mathbb{R}$**  if

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By a **trivial linear combination** of vectors  $\mathbf{v}^1, \dots, \mathbf{v}^k$  we mean the linear combination  $0 \cdot \mathbf{v}^1 + \dots + 0 \cdot \mathbf{v}^k$ .

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By a **trivial linear combination** of vectors  $\mathbf{v}^1, \dots, \mathbf{v}^k$  we mean the linear combination  $0 \cdot \mathbf{v}^1 + \dots + 0 \cdot \mathbf{v}^k$ . Linear combination which is not trivial is called **non-trivial**.

## Definition

We say that vectors  $\mathbf{v}^1, \dots, \mathbf{v}^k \in \mathbb{R}^n$  are **linearly dependent** if there exists their non-trivial linear combination which is equal to the zero vector.



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## Remark

Vectors  $\mathbf{v}^1, \dots, \mathbf{v}^k$  are linearly dependent if and only if one of them can be expressed as a linear combination of the others.

## Definition

Let  $\mathbf{A} \in M(m \times n)$ . The **rank** of the matrix  $\mathbf{A}$  is the maximal number of linearly independent row vectors of  $\mathbf{A}$ , i.e. the rank is equal to  $k \in \mathbb{N}$  if

- (i) there is  $k$  linearly independent row vectors of  $\mathbf{A}$  and
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The rank of the zero matrix is zero. Rank of  $\mathbf{A}$  is denoted by  $\text{rank}(\mathbf{A})$ .

## Definition

We say that a matrix  $\mathbf{A} \in M(m \times n)$  is in a **row echelon form** if for each  $i \in \{2, \dots, m\}$  the  $i$ th row of  $\mathbf{A}$  is either a zero vector or it has more zeros at the beginning than the  $(i - 1)$ th row.

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## Remark

The rank of a row echelon matrix is equal to the number of its non-zero rows.

### Definition

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## Definition

A matrix **transformation** is a finite sequence of elementary row operations. If a matrix  $\mathbf{B} \in M(m \times n)$  results from the matrix  $\mathbf{A} \in M(m \times n)$  by applying a transformation  $T$  on the matrix  $\mathbf{A}$ , then this fact is denoted by  $\mathbf{A} \xrightarrow{T} \mathbf{B}$ .

## Theorem 7 (properties of matrix transformations)

- (i) *Let  $\mathbf{A} \in M(m \times n)$ . Then there exists a transformation transforming  $\mathbf{A}$  to a row echelon matrix.*

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- (iii) *Let  $\mathbf{A}, \mathbf{B} \in M(m \times n)$  and there exist a transformation  $T$  such that  $\mathbf{A} \xrightarrow{T} \mathbf{B}$ . Then  $\text{rank}(\mathbf{A}) = \text{rank}(\mathbf{B})$ .*

# Transformation to a row echelon form

$$\begin{pmatrix} \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \end{pmatrix}$$

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$$\begin{pmatrix} \bullet & \bullet & \bullet & \bullet & \bullet & \bullet \\ 0 & \bullet & \bullet & \bullet & \bullet & \bullet \\ 0 & \bullet & \bullet & \bullet & \bullet & \bullet \\ 0 & \bullet & \bullet & \bullet & \bullet & \bullet \\ 0 & \bullet & \bullet & \bullet & \bullet & \bullet \end{pmatrix}$$

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### Remark

Similarly as the elementary row operations one can define also elementary column operations. It can be shown that the elementary column operations do not change the rank of the matrix.

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## Remark

It can be shown that  $\text{rank}(\mathbf{A}) = \text{rank}(\mathbf{A}^T)$  for any  $\mathbf{A} \in M(m \times n)$ .

### Theorem 8 (representation of a transformation)

*Let  $T$  be a transformation on  $m \times n$  matrices. Then there exists an invertible matrix  $\mathbf{C}_T \in M(m \times m)$  satisfying: whenever we apply the transformation  $T$  to a matrix  $\mathbf{A} \in M(m \times n)$ , we obtain the matrix  $\mathbf{C}_T \mathbf{A}$ .*

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### Remark

Also the converse is true: For every invertible matrix  $\mathbf{C}$  the mapping  $\mathbf{A} \mapsto \mathbf{C}\mathbf{A}$  is a transformation.

## Lemma 9

*Let  $\mathbf{A} \in M(n \times n)$  and  $\text{rank}(\mathbf{A}) = n$ . Then there exists a transformation transforming  $\mathbf{A}$  to  $\mathbf{I}$ .*

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### Theorem 10

*Let  $\mathbf{A} \in M(n \times n)$ . Then  $\mathbf{A}$  is invertible if and only if  $\text{rank}(\mathbf{A}) = n$ .*

# VI.3. Determinants

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## Definition

Let  $\mathbf{A} \in M(n \times n)$ . The symbol  $\mathbf{A}_{ij}$  denotes the  $(n - 1)$ -by- $(n - 1)$  matrix which is created from  $\mathbf{A}$  by omitting the  $i$ th row and the  $j$ th column.



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$$\mathbf{A} = \begin{pmatrix} a_{1,1} & \cdots & a_{1,j-1} & a_{1,j} & a_{1,j+1} & \cdots & a_{1,n} \\ \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{i-1,1} & \cdots & a_{i-1,j-1} & a_{i-1,j} & a_{i-1,j+1} & \cdots & a_{i-1,n} \\ a_{i,1} & \cdots & a_{i,j-1} & a_{i,j} & a_{i,j+1} & \cdots & a_{i,n} \\ a_{i+1,1} & \cdots & a_{i+1,j-1} & a_{i+1,j} & a_{i+1,j+1} & \cdots & a_{i+1,n} \\ \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n,1} & \cdots & a_{n,j-1} & a_{n,j} & a_{n,j+1} & \cdots & a_{n,n} \end{pmatrix}$$

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## Definition

Let  $\mathbf{A} = (a_{ij})_{i,j=1..n}$ . The **determinant** of the matrix  $\mathbf{A}$  is defined by

$$\det \mathbf{A} = \begin{cases} a_{11} & \text{if } n = 1, \\ \sum_{i=1}^n (-1)^{i+1} a_{i1} \det \mathbf{A}_{i1} & \text{if } n > 1. \end{cases}$$

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For  $\det \mathbf{A}$  we will also use the symbol

$$\begin{vmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \ddots & \vdots & \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{vmatrix}.$$

## Theorem 11 (cofactor expansion)

Let  $\mathbf{A} = (a_{ij})_{i,j=1..n}$ ,  $k \in \{1, \dots, n\}$ . Then

$$\det \mathbf{A} = \sum_{i=1}^n (-1)^{i+k} a_{ik} \det \mathbf{A}_{ik} \quad (\text{expansion along } k\text{th column}),$$

$$\det \mathbf{A} = \sum_{j=1}^n (-1)^{k+j} a_{kj} \det \mathbf{A}_{kj} \quad (\text{expansion along } k\text{th row}).$$

## Lemma 12

*Let  $j, n \in \mathbb{N}$ ,  $j \leq n$ , and the matrices  $\mathbf{A}, \mathbf{B}, \mathbf{C} \in M(n \times n)$  coincide at each row except for the  $j$ th row. Let the  $j$ th row of  $\mathbf{A}$  be equal to the sum of the  $j$ th rows of  $\mathbf{B}$  and  $\mathbf{C}$ . Then  $\det \mathbf{A} = \det \mathbf{B} + \det \mathbf{C}$ .*



## Lemma 12

Let  $j, n \in \mathbb{N}$ ,  $j \leq n$ , and the matrices  $\mathbf{A}, \mathbf{B}, \mathbf{C} \in M(n \times n)$  coincide at each row except for the  $j$ th row. Let the  $j$ th row of  $\mathbf{A}$  be equal to the sum of the  $j$ th rows of  $\mathbf{B}$  and  $\mathbf{C}$ . Then  $\det \mathbf{A} = \det \mathbf{B} + \det \mathbf{C}$ .

$$\begin{vmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{j-1,1} & \dots & a_{j-1,n} \\ u_1+v_1 & \dots & u_n+v_n \\ a_{j+1,1} & \dots & a_{j+1,n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{vmatrix} = \begin{vmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{j-1,1} & \dots & a_{j-1,n} \\ u_1 & \dots & u_n \\ a_{j+1,1} & \dots & a_{j+1,n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{vmatrix} + \begin{vmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{j-1,1} & \dots & a_{j-1,n} \\ v_1 & \dots & v_n \\ a_{j+1,1} & \dots & a_{j+1,n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{nn} \end{vmatrix}$$

## Theorem 13 (determinant and transformations)

Let  $\mathbf{A}, \mathbf{A}' \in M(n \times n)$ .

- (i) *If the matrix  $\mathbf{A}'$  is created from the matrix  $\mathbf{A}$  by multiplying one row in  $\mathbf{A}$  by a real number  $\mu$ , then  $\det \mathbf{A}' = \mu \det \mathbf{A}$ .*

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- (ii) *If the matrix  $\mathbf{A}'$  is created from  $\mathbf{A}$  by interchanging two rows in  $\mathbf{A}$  (i.e. by applying the elementary row operation of the first type), then  $\det \mathbf{A}' = -\det \mathbf{A}$ .*

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- (iii) *If the matrix  $\mathbf{A}'$  is created from  $\mathbf{A}$  by adding a  $\mu$ -multiple of a row in  $\mathbf{A}$  to another row in  $\mathbf{A}$  (i.e. by applying the elementary row operation of the third type), then  $\det \mathbf{A}' = \det \mathbf{A}$ .*

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- (iv) *If  $\mathbf{A}'$  is created from  $\mathbf{A}$  by applying a transformation, then  $\det \mathbf{A} \neq 0$  if and only if  $\det \mathbf{A}' \neq 0$ .*

### Remark

The determinant of a matrix with a zero row is equal to zero.

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The determinant of a matrix with a zero row is equal to zero. The determinant of a matrix with two identical rows is also equal to zero.

## Definition

Let  $\mathbf{A} = (a_{ij})_{i,j=1..n}$ . We say that  $\mathbf{A}$  is an **upper triangular matrix** if  $a_{ij} = 0$  for  $i > j$ ,  $i, j \in \{1, \dots, n\}$ .



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## Theorem 14 (determinant of a triangular matrix)

Let  $\mathbf{A} = (a_{ij})_{i,j=1..n}$  be an upper or lower triangular matrix.  
Then

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

## Theorem 15 (determinant and invertibility)

*Let  $\mathbf{A} \in M(n \times n)$ . Then  $\mathbf{A}$  is invertible if and only if  $\det \mathbf{A} \neq 0$ .*

## Theorem 16 (determinant of a product)

Let  $\mathbf{A}, \mathbf{B} \in M(n \times n)$ . Then  $\det \mathbf{AB} = \det \mathbf{A} \cdot \det \mathbf{B}$ .

### Theorem 16 (determinant of a product)

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### Theorem 17 (determinant of a transpose)

Let  $\mathbf{A} \in M(n \times n)$ . Then  $\det \mathbf{A}^T = \det \mathbf{A}$ .

# VI.4. Systems of linear equations

A system of  $m$  equations in  $n$  unknowns  $x_1, \dots, x_n$ :

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

$$a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2,$$

$$\vdots$$

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

(S)

where  $a_{ij} \in \mathbb{R}$ ,  $b_i \in \mathbb{R}$ ,  $i = 1, \dots, m$ ,  $j = 1, \dots, n$ .

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where  $a_{ij} \in \mathbb{R}$ ,  $b_i \in \mathbb{R}$ ,  $i = 1, \dots, m$ ,  $j = 1, \dots, n$ . The matrix form is

$$\mathbf{Ax} = \mathbf{b},$$

where  $\mathbf{A} = \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix} \in M(m \times n)$ , is called the **coefficient matrix**,  $\mathbf{b} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix} \in M(m \times 1)$  is called the

**vector of the right-hand side** and  $\mathbf{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in M(n \times 1)$  is the **vector of unknowns**.



## Definition

The matrix

$$(\mathbf{A}|\mathbf{b}) = \left( \begin{array}{ccc|c} a_{11} & \dots & a_{1n} & b_1 \\ \vdots & \ddots & \vdots & \vdots \\ a_{m1} & \dots & a_{mn} & b_m \end{array} \right)$$

is called the **augmented matrix of the system** (S).

## Proposition 18 (solutions of a transformed system)

Let  $\mathbf{A} \in M(m \times n)$ ,  $\mathbf{b} \in M(m \times 1)$  and let  $T$  be a transformation of matrices with  $m$  rows. Denote  $\mathbf{A} \xrightarrow{T} \mathbf{A}'$ ,  $\mathbf{b} \xrightarrow{T} \mathbf{b}'$ . Then for any  $\mathbf{y} \in M(n \times 1)$  we have  $\mathbf{A}\mathbf{y} = \mathbf{b}$  if and only if  $\mathbf{A}'\mathbf{y} = \mathbf{b}'$ , i.e. the systems  $\mathbf{A}\mathbf{x} = \mathbf{b}$  and  $\mathbf{A}'\mathbf{x} = \mathbf{b}'$  have the same set of solutions.

### Theorem 19 (Rouché-Fontené)

*The system (S) has a solution if and only if its coefficient matrix has the same rank as its augmented matrix.*

## **Systems of $n$ equations in $n$ variables**

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#### Theorem 20 (solvability of an $n \times n$ system)

Let  $\mathbf{A} \in M(n \times n)$ . Then the following statements are equivalent:

- (i) the matrix  $\mathbf{A}$  is invertible,

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- (iv)  $\det \mathbf{A} \neq 0$ .



### Theorem 21 (Cramer's rule)

Let  $\mathbf{A} \in M(n \times n)$  be an invertible matrix,  $\mathbf{b} \in M(n \times 1)$ ,  $\mathbf{x} \in M(n \times 1)$ , and  $\mathbf{Ax} = \mathbf{b}$ . Then

$$x_j = \frac{\begin{vmatrix} a_{11} & \dots & a_{1,j-1} & b_1 & a_{1,j+1} & \dots & a_{1n} \\ \vdots & & & \vdots & & & \vdots \\ a_{n1} & \dots & a_{n,j-1} & b_n & a_{n,j+1} & \dots & a_{nn} \end{vmatrix}}{\det \mathbf{A}}$$

for  $j = 1, \dots, n$ .

# VI.5. Definiteness of matrices

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## Definition

We say that a symmetric matrix  $\mathbf{A} \in M(n \times n)$  is

- **positive definite** (PD), if  $\mathbf{u}^T \mathbf{A} \mathbf{u} > 0$  for all  $\mathbf{u} \in \mathbb{R}^n$ ,  
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## Proposition 22 (definiteness of diagonal matrices)

Let  $\mathbf{A} \in M(n \times n)$  be diagonal (i.e.  $a_{ij} = 0$  whenever  $i \neq j$ ).

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- $\mathbf{A}$  is PD if and only if  $a_{ii} > 0$  for all  $i = 1, 2, \dots, n$ ,



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## Proposition 23 (necessary conditions for definiteness)

Let  $\mathbf{A} \in M(n \times n)$  be a symmetric matrix. Then

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## Theorem 24 (Sylvester's criterion)

Let  $\mathbf{A} = (a_{ij}) \in M(n \times n)$  be a symmetric matrix. Then  $\mathbf{A}$  is

- *positive definite if and only if*

$$\begin{vmatrix} a_{11} & \dots & a_{1k} \\ \vdots & & \vdots \\ a_{k1} & \dots & a_{kk} \end{vmatrix} > 0 \quad \text{for all } k = 1, \dots, n,$$

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## Definition

Let  $\mathbf{x}^j \in \mathbb{R}^n$  for each  $j \in \mathbb{N}$  and  $\mathbf{x} \in \mathbb{R}^n$ . We say that a sequence  $\{\mathbf{x}^j\}_{j=1}^{\infty}$  **converges to  $\mathbf{x}$** , if

$$\lim_{j \rightarrow \infty} \rho(\mathbf{x}, \mathbf{x}^j) = 0.$$

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The sequence  $\{\mathbf{y}^j\}_{j=1}^{\infty}$  of points in  $\mathbb{R}^n$  is called **convergent** if there exists  $\mathbf{y} \in \mathbb{R}^n$  such that  $\{\mathbf{y}^j\}_{j=1}^{\infty}$  converges to  $\mathbf{y}$ .

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## Remark

The sequence  $\{\mathbf{x}^j\}_{j=1}^{\infty}$  converges to  $\mathbf{x} \in \mathbb{R}^n$  if and only if

$$\forall \varepsilon \in \mathbb{R}, \varepsilon > 0 \exists j_0 \in \mathbb{N} \forall j \in \mathbb{N}, j \geq j_0 : \mathbf{x}^j \in B(\mathbf{x}, \varepsilon).$$



## Theorem 25 (convergence is coordinatewise)

*Let  $\mathbf{x}^j \in \mathbb{R}^n$  for each  $j \in \mathbb{N}$  and let  $\mathbf{x} \in \mathbb{R}^n$ . The sequence  $\{\mathbf{x}^j\}_{j=1}^{\infty}$  converges to  $\mathbf{x}$  if and only if for each  $i \in \{1, \dots, n\}$  the sequence of real numbers  $\{x_i^j\}_{j=1}^{\infty}$  converges to the real number  $x_i$ .*

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Theorem 25 says that the convergence in the space  $\mathbb{R}^n$  is the same as the “coordinatewise” convergence.

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### Remark

Theorem 25 says that the convergence in the space  $\mathbb{R}^n$  is the same as the “coordinatewise” convergence. It follows that a sequence  $\{\mathbf{x}^j\}_{j=1}^{\infty}$  has at most one limit. If it exists, then we denote it by  $\lim_{j \rightarrow \infty} \mathbf{x}^j$ . Sometimes we also write simply  $\mathbf{x}^j \rightarrow \mathbf{x}$  instead of  $\lim_{j \rightarrow \infty} \mathbf{x}^j = \mathbf{x}$ .

## Theorem 26 (characterisation of closed sets)

*Let  $M \subset \mathbb{R}^n$ . Then the following statements are equivalent:*

- (i)  $M$  is closed in  $\mathbb{R}^n$ .*
- (ii)  $\mathbb{R}^n \setminus M$  is open in  $\mathbb{R}^n$ .*
- (iii) Any  $\mathbf{x} \in \mathbb{R}^n$  which is a limit of a sequence from  $M$  belongs to  $M$ .*

## Theorem 27 (properties of closed sets)

- (i) *The empty set and the whole space  $\mathbb{R}^n$  are closed in  $\mathbb{R}^n$ .*

Remark

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- (ii) *Let  $F_\alpha \subset \mathbb{R}^n$ ,  $\alpha \in A \neq \emptyset$ , be closed in  $\mathbb{R}^n$ . Then  $\bigcap_{\alpha \in A} F_\alpha$  is closed in  $\mathbb{R}^n$ .*

## Remark

- (ii) *An intersection of an arbitrary system of closed sets is closed.*

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- (iii) *Let  $F_i \subset \mathbb{R}^n$ ,  $i = 1, \dots, m$ , be closed in  $\mathbb{R}^n$ . Then  $\bigcup_{i=1}^m F_i$  is closed in  $\mathbb{R}^n$ .*

### Remark

- (ii) *An intersection of an arbitrary system of closed sets is closed.*
- (iii) *A union of finitely many closed sets is closed.*

## Theorem 28

Let  $M \subset \mathbb{R}^n$ . Then the following holds:

- (i) The set  $\overline{M}$  is closed in  $\mathbb{R}^n$ .
- (ii) The set  $\text{Int } M$  is open in  $\mathbb{R}^n$ .
- (iii) The set  $M$  is open in  $\mathbb{R}^n$  if and only if  $M = \text{Int } M$ .



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## Remark

The set  $\text{Int } M$  is the largest open set contained in  $M$  in the following sense: If  $G$  is a set open in  $\mathbb{R}^n$  and satisfying  $G \subset M$ , then  $G \subset \text{Int } M$ .

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The set  $\text{Int } M$  is the largest open set contained in  $M$  in the following sense: If  $G$  is a set open in  $\mathbb{R}^n$  and satisfying  $G \subset M$ , then  $G \subset \text{Int } M$ . Similarly  $\overline{M}$  is the smallest closed set containing  $M$ .

## Definition

We say that the set  $M \subset \mathbb{R}^n$  is **bounded** if there exists  $r > 0$  such that  $M \subset B(\mathbf{o}, r)$ .

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## Theorem 29

*A set  $M \subset \mathbb{R}^n$  is bounded if and only if its closure  $\overline{M}$  is bounded.*

# VII.2. Continuous functions of several variables

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## Definition

Let  $M \subset \mathbb{R}^n$ ,  $\mathbf{x} \in M$ , and  $f: M \rightarrow \mathbb{R}$ . We say that  $f$  is **continuous at  $\mathbf{x}$  with respect to  $M$** , if we

$$\forall \varepsilon \in \mathbb{R}, \varepsilon > 0 \exists \delta \in \mathbb{R}, \delta > 0 \forall \mathbf{y} \in B(\mathbf{x}, \delta) \cap M: f(\mathbf{y}) \in B(f(\mathbf{x}), \varepsilon).$$

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We say that  $f$  is **continuous at the point  $\mathbf{x}$**  if it is continuous at  $\mathbf{x}$  with respect to a neighbourhood of  $\mathbf{x}$ , i.e.

$$\forall \varepsilon \in \mathbb{R}, \varepsilon > 0 \exists \delta \in \mathbb{R}, \delta > 0 \forall \mathbf{y} \in B(\mathbf{x}, \delta): f(\mathbf{y}) \in B(f(\mathbf{x}), \varepsilon).$$



### Theorem 30

*Let  $M \subset \mathbb{R}^n$ ,  $\mathbf{x} \in M$ ,  $f: M \rightarrow \mathbb{R}$ ,  $g: M \rightarrow \mathbb{R}$ , and  $c \in \mathbb{R}$ . If  $f$  and  $g$  are continuous at the point  $\mathbf{x}$  with respect to  $M$ , then the functions  $cf$ ,  $f + g$  and  $fg$  are continuous at  $\mathbf{x}$  with respect to  $M$ . If the function  $g$  is nonzero at  $\mathbf{x}$ , then also the function  $f/g$  is continuous at  $\mathbf{x}$  with respect to  $M$ .*

### Theorem 31

*Let  $r, s \in \mathbb{N}$ ,  $M \subset \mathbb{R}^s$ ,  $L \subset \mathbb{R}^r$ , and  $\mathbf{y} \in M$ . Let  $\varphi_1, \dots, \varphi_r$  be functions defined on  $M$ , which are continuous at  $\mathbf{y}$  with respect to  $M$  and  $[\varphi_1(\mathbf{x}), \dots, \varphi_r(\mathbf{x})] \in L$  for each  $\mathbf{x} \in M$ . Let  $f: L \rightarrow \mathbb{R}$  be continuous at the point  $[\varphi_1(\mathbf{y}), \dots, \varphi_r(\mathbf{y})]$  with respect to  $L$ . Then the compound function  $F: M \rightarrow \mathbb{R}$  defined by*

$$F(\mathbf{x}) = f(\varphi_1(\mathbf{x}), \dots, \varphi_r(\mathbf{x})), \quad \mathbf{x} \in M,$$

*is continuous at  $\mathbf{y}$  with respect to  $M$ .*

## Theorem 32 (Heine)

Let  $M \subset \mathbb{R}^n$ ,  $\mathbf{x} \in M$ , and  $f: M \rightarrow \mathbb{R}$ . Then the following are equivalent.

- (i) The function  $f$  is continuous at  $\mathbf{x}$  with respect to  $M$ .
- (ii)  $\lim_{j \rightarrow \infty} f(\mathbf{x}^j) = f(\mathbf{x})$  for each sequence  $\{\mathbf{x}^j\}_{j=1}^{\infty}$  such that  $\mathbf{x}^j \in M$  for  $j \in \mathbb{N}$  and  $\lim_{j \rightarrow \infty} \mathbf{x}^j = \mathbf{x}$ .

## Definition

Let  $M \subset \mathbb{R}^n$  and  $f: M \rightarrow \mathbb{R}$ . We say that  $f$  is **continuous on  $M$**  if it is continuous at each point  $\mathbf{x} \in M$  with respect to  $M$ .

## Definition

Let  $M \subset \mathbb{R}^n$  and  $f: M \rightarrow \mathbb{R}$ . We say that  $f$  is **continuous on  $M$**  if it is continuous at each point  $\mathbf{x} \in M$  with respect to  $M$ .

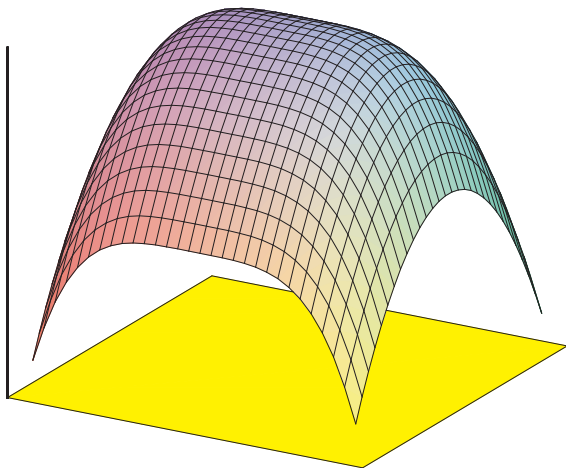
## Remark

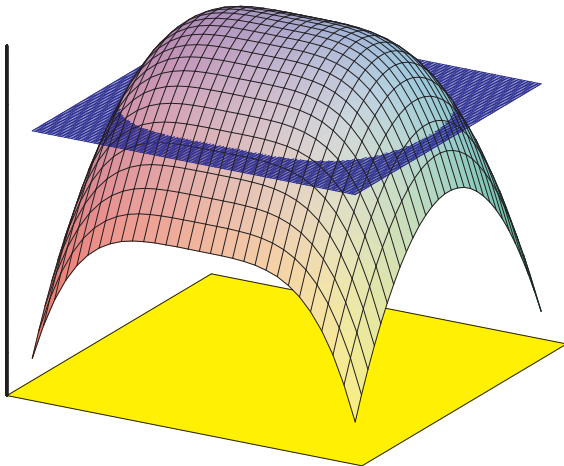
The functions  $\pi_j: \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $\pi_j(\mathbf{x}) = x_j$ ,  $1 \leq j \leq n$ , are continuous on  $\mathbb{R}^n$ . They are called **coordinate projections**.

### Theorem 33

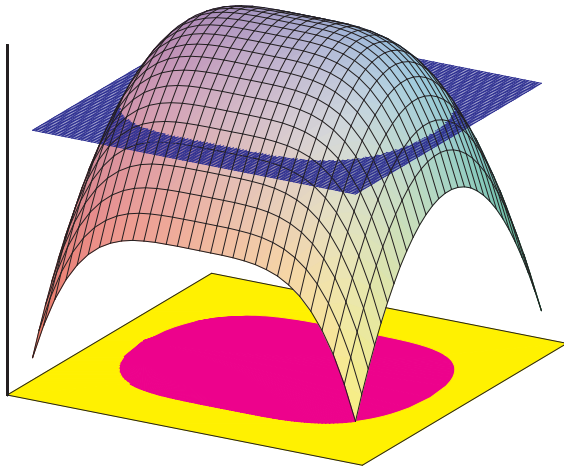
*Let  $f$  be a continuous function on  $\mathbb{R}^n$  and  $c \in \mathbb{R}$ . Then the following holds:*

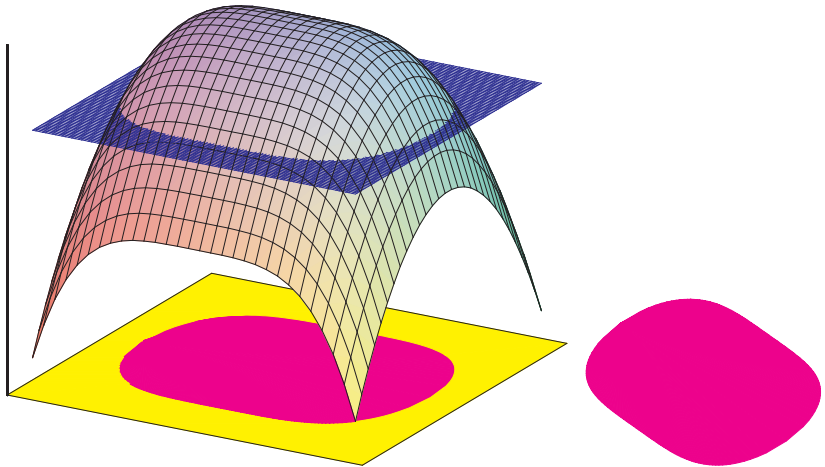
- (i) *The set  $\{\mathbf{x} \in \mathbb{R}^n; f(\mathbf{x}) < c\}$  is open in  $\mathbb{R}^n$ .*
- (ii) *The set  $\{\mathbf{x} \in \mathbb{R}^n; f(\mathbf{x}) > c\}$  is open in  $\mathbb{R}^n$ .*
- (iii) *The set  $\{\mathbf{x} \in \mathbb{R}^n; f(\mathbf{x}) \leq c\}$  is closed in  $\mathbb{R}^n$ .*
- (iv) *The set  $\{\mathbf{x} \in \mathbb{R}^n; f(\mathbf{x}) \geq c\}$  is closed in  $\mathbb{R}^n$ .*
- (v) *The set  $\{\mathbf{x} \in \mathbb{R}^n; f(\mathbf{x}) = c\}$  is closed in  $\mathbb{R}^n$ .*

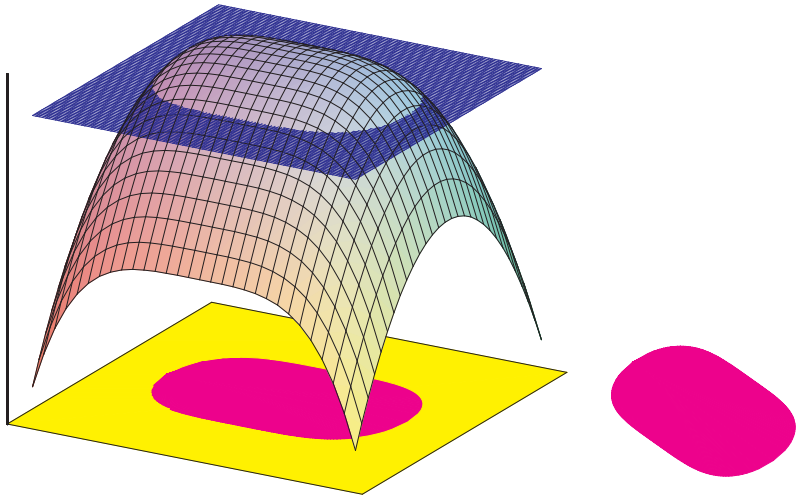


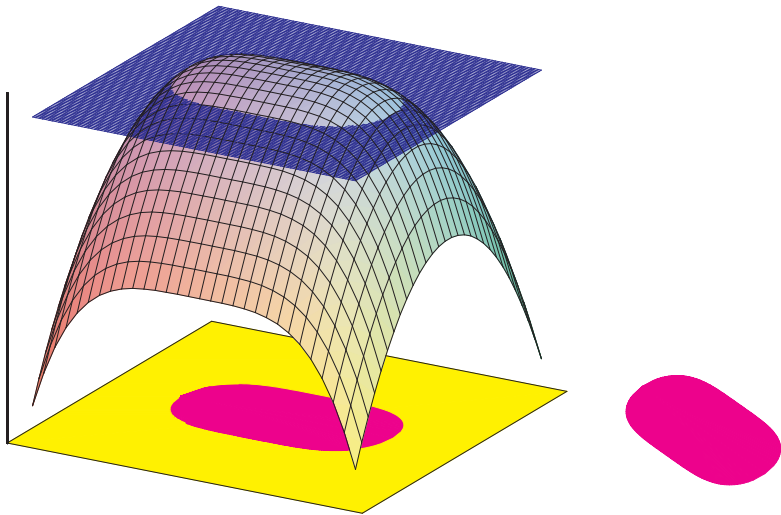












## Definition

We say that a set  $M \subset \mathbb{R}^n$  is **compact** if for each sequence of elements of  $M$  there exists a convergent subsequence with a limit in  $M$ .

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## Theorem 34 (characterisation of compact subsets of $\mathbb{R}^n$ )

*The set  $M \subset \mathbb{R}^n$  is compact if and only if  $M$  is bounded and closed.*

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### Theorem 34 (characterisation of compact subsets of $\mathbb{R}^n$ )

*The set  $M \subset \mathbb{R}^n$  is compact if and only if  $M$  is bounded and closed.*

### Lemma 35

*Omitted.*

## Definition

Let  $M \subset \mathbb{R}^n$ ,  $\mathbf{x} \in M$ , and let  $f$  be a function defined at least on  $M$  (i.e.  $M \subset D_f$ ). We say that  $f$  attains at the point  $\mathbf{x}$  its

- **maximum on  $M$**  if  $f(\mathbf{y}) \leq f(\mathbf{x})$  for every  $\mathbf{y} \in M$ ,



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- **strict local maximum with respect to  $M$**  if there exists  $\delta > 0$  such that  $f(\mathbf{y}) < f(\mathbf{x})$  for every  $\mathbf{y} \in (B(\mathbf{x}, \delta) \setminus \{\mathbf{x}\}) \cap M$ .

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The notions of a **minimum**, a **local minimum**, and a **strict local minimum** with respect to  $M$  are defined in analogous way.

## Definition

We say that a function  $f$  attains a **local maximum** at a point  $\mathbf{x} \in \mathbb{R}^n$  if  $\mathbf{x}$  is a local maximum with respect to some neighbourhood of  $\mathbf{x}$ .

## Definition

We say that a function  $f$  attains a **local maximum** at a point  $\mathbf{x} \in \mathbb{R}^n$  if  $\mathbf{x}$  is a local maximum with respect to some neighbourhood of  $\mathbf{x}$ .

Similarly we define **local minimum**, **strict local maximum** and **strict local minimum**.

### Theorem 36 (attaining extrema)

*Let  $M \subset \mathbb{R}^n$  be a non-empty compact set and  $f: M \rightarrow \mathbb{R}$  a function continuous on  $M$ . Then  $f$  attains its maximum and minimum on  $M$ .*

### Theorem 36 (attaining extrema)

*Let  $M \subset \mathbb{R}^n$  be a non-empty compact set and  $f: M \rightarrow \mathbb{R}$  a function continuous on  $M$ . Then  $f$  attains its maximum and minimum on  $M$ .*

### Corollary

*Let  $M \subset \mathbb{R}^n$  be a non-empty compact set and  $f: M \rightarrow \mathbb{R}$  a continuous function on  $M$ . Then  $f$  is bounded on  $M$ .*

## Definition

We say that a function  $f$  of  $n$  variables has a limit at a point  $\mathbf{a} \in \mathbb{R}^n$  equal to  $A \in \mathbb{R}^*$  if

$$\forall \varepsilon \in \mathbb{R}, \varepsilon > 0 \exists \delta \in \mathbb{R}, \delta > 0 \forall \mathbf{x} \in B(\mathbf{a}, \delta) \setminus \{\mathbf{a}\}: f(\mathbf{x}) \in B(A, \varepsilon).$$



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## Remark

- Each function has at a given point at most one limit.  
We write  $\lim_{\mathbf{x} \rightarrow \mathbf{a}} f(\mathbf{x}) = A$ .

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## Remark

- Each function has at a given point at most one limit.  
We write  $\lim_{\mathbf{x} \rightarrow \mathbf{a}} f(\mathbf{x}) = A$ .
- The function  $f$  is continuous at  $\mathbf{a}$  if and only if  $\lim_{\mathbf{x} \rightarrow \mathbf{a}} f(\mathbf{x}) = f(\mathbf{a})$ .

## Definition

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## Remark

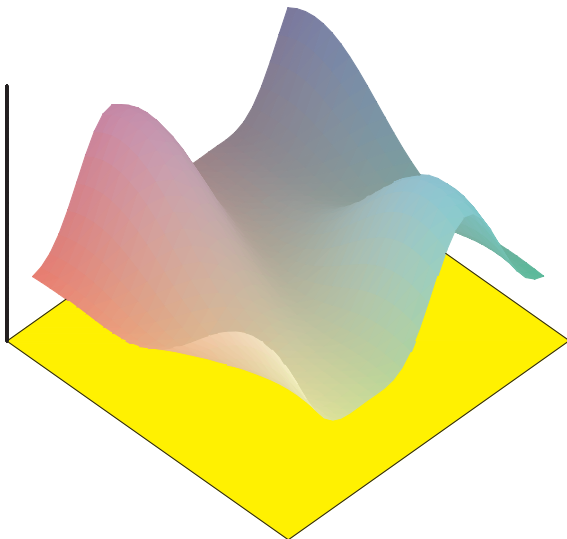
- Each function has at a given point at most one limit.  
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- The function  $f$  is continuous at  $\mathbf{a}$  if and only if  $\lim_{\mathbf{x} \rightarrow \mathbf{a}} f(\mathbf{x}) = f(\mathbf{a})$ .
- For limits of functions of several variables one can prove similar theorems as for limits of functions of one variable (arithmetics, the sandwich theorem, ...).

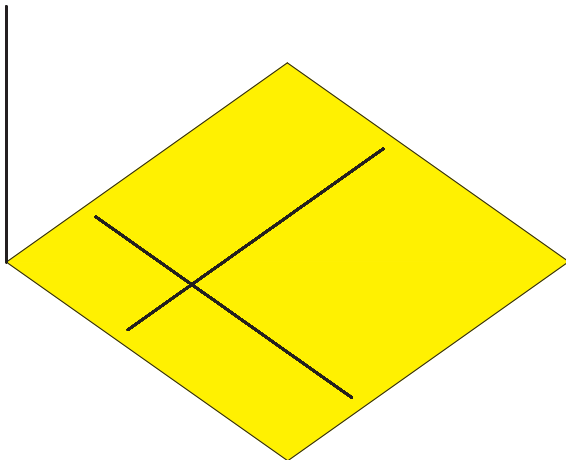
### Theorem 37

Let  $r, s \in \mathbb{N}$ ,  $\mathbf{a} \in \mathbb{R}^s$ , and let  $\varphi_1, \dots, \varphi_r$  be functions of  $s$  variables such that  $\lim_{\mathbf{x} \rightarrow \mathbf{a}} \varphi_j(\mathbf{x}) = b_j$ ,  $j = 1, \dots, r$ . Set  $\mathbf{b} = [b_1, \dots, b_r]$ . Let  $f$  be a function of  $r$  variables which is continuous at the point  $\mathbf{b}$ . If we define a compound function  $F$  of  $s$  variables by

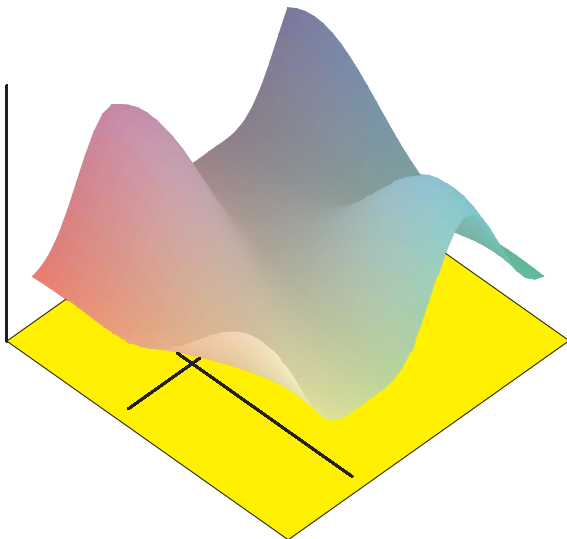
$$F(\mathbf{x}) = f(\varphi_1(\mathbf{x}), \varphi_2(\mathbf{x}), \dots, \varphi_r(\mathbf{x})),$$

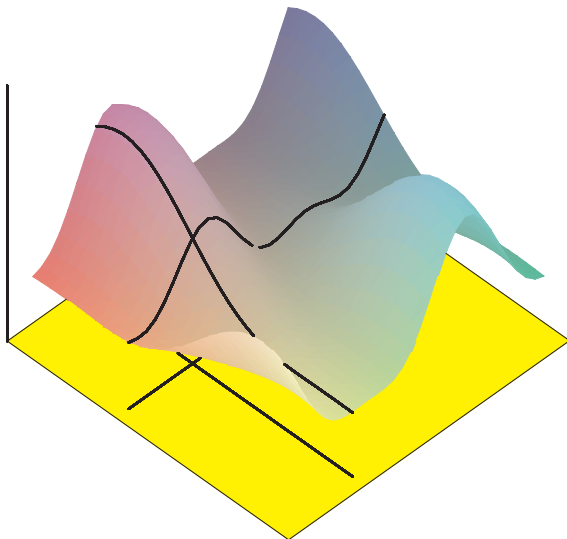
then  $\lim_{\mathbf{x} \rightarrow \mathbf{a}} F(\mathbf{x}) = f(\mathbf{b})$ .



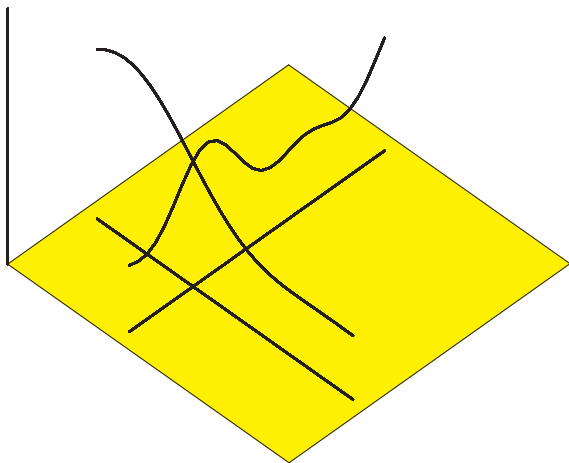


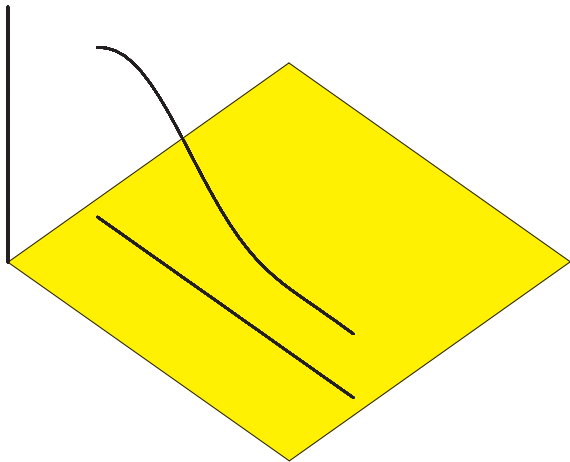
## VII.3. Partial derivatives and tangent hyperplane



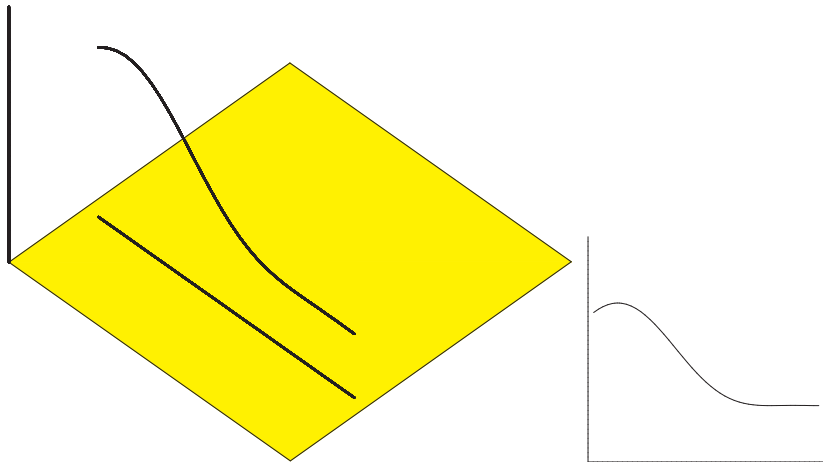


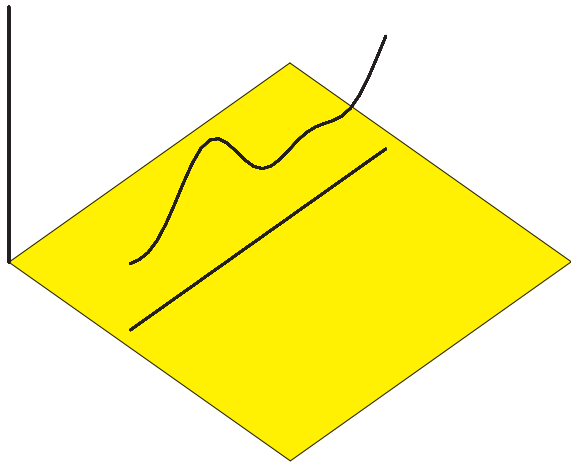




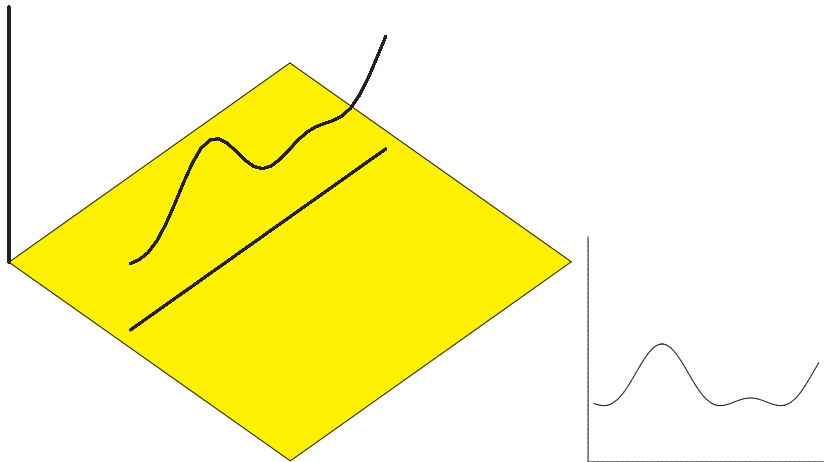


## VII.3. Partial derivatives and tangent hyperplane





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Set  $\mathbf{e}^j = [0, \dots, 0, \underset{j\text{th coordinate}}{1}, 0, \dots, 0]$ .

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### Definition

Let  $f$  be a function of  $n$  variables,  $j \in \{1, \dots, n\}$ ,  $\mathbf{a} \in \mathbb{R}^n$ .  
Then the number

$$\frac{\partial f}{\partial x_j}(\mathbf{a}) = \lim_{t \rightarrow 0} \frac{f(\mathbf{a} + t\mathbf{e}^j) - f(\mathbf{a})}{t}$$

is called the **partial derivative (of first order) of function  $f$  according to  $j$ th variable at the point  $\mathbf{a}$**  (if the limit exists).

Set  $\mathbf{e}^j = [0, \dots, 0, \underset{j\text{th coordinate}}{1}, 0, \dots, 0]$ .

### Definition

Let  $f$  be a function of  $n$  variables,  $j \in \{1, \dots, n\}$ ,  $\mathbf{a} \in \mathbb{R}^n$ .  
Then the number

$$\begin{aligned} \frac{\partial f}{\partial x_j}(\mathbf{a}) &= \lim_{t \rightarrow 0} \frac{f(\mathbf{a} + t\mathbf{e}^j) - f(\mathbf{a})}{t} \\ &= \lim_{t \rightarrow 0} \frac{f(a_1, \dots, a_{j-1}, a_j + t, a_{j+1}, \dots, a_n) - f(a_1, \dots, a_n)}{t} \end{aligned}$$

is called the **partial derivative (of first order) of function  $f$  according to  $j$ th variable at the point  $\mathbf{a}$**  (if the limit exists).

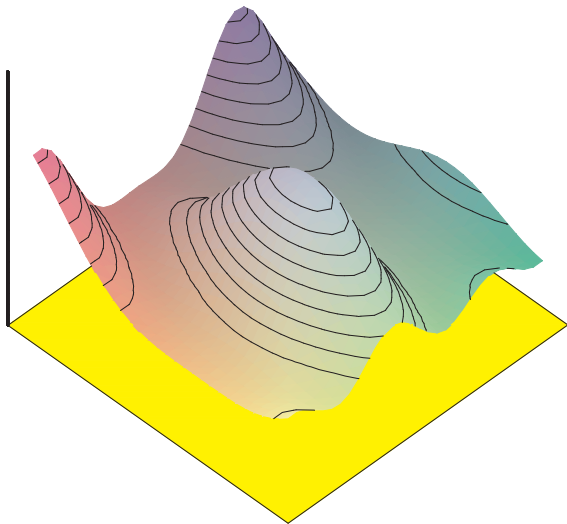


### Theorem 38 (necessary condition of the existence of local extremum)

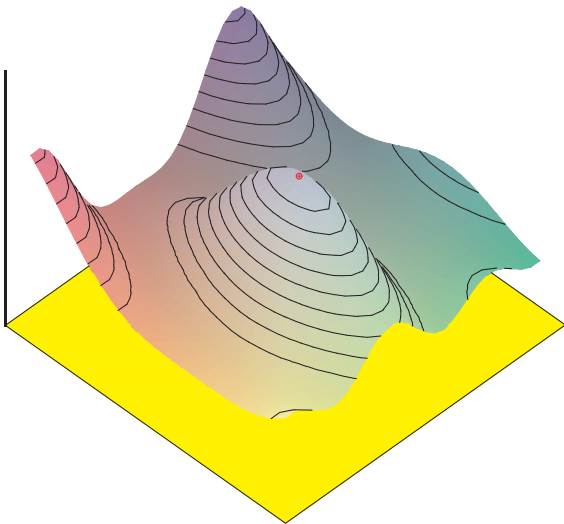
*Let  $G \subset \mathbb{R}^n$  be an open set,  $\mathbf{a} \in G$ , and suppose that a function  $f: G \rightarrow \mathbb{R}$  has a local extremum (i.e. a local maximum or a local minimum) at the point  $\mathbf{a}$ . Then for each  $j \in \{1, \dots, n\}$  the following holds:*

*The partial derivative  $\frac{\partial f}{\partial x_j}(\mathbf{a})$  either does not exist or it is equal to zero.*

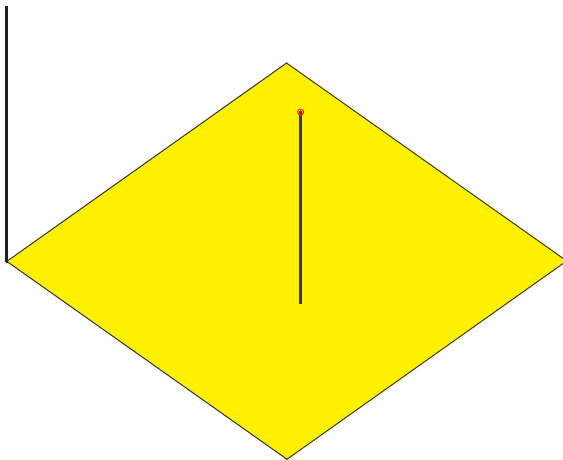
## VII.3. Partial derivatives and tangent hyperplane

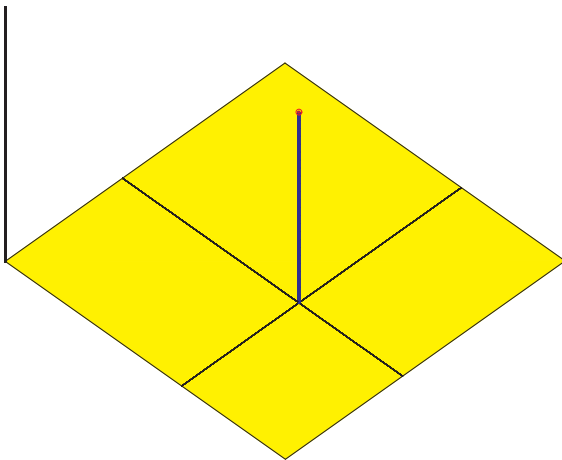


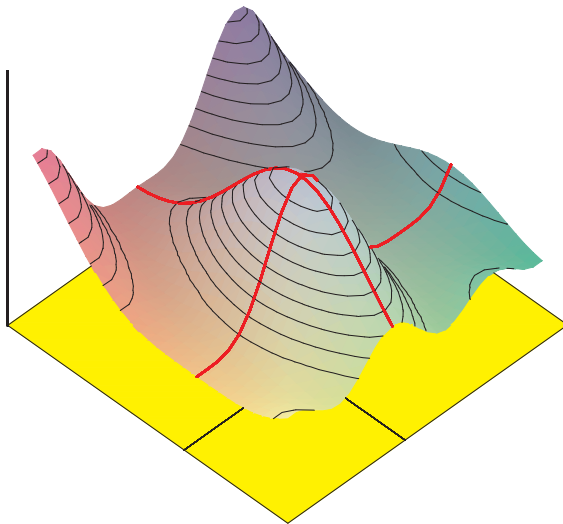
## VII.3. Partial derivatives and tangent hyperplane

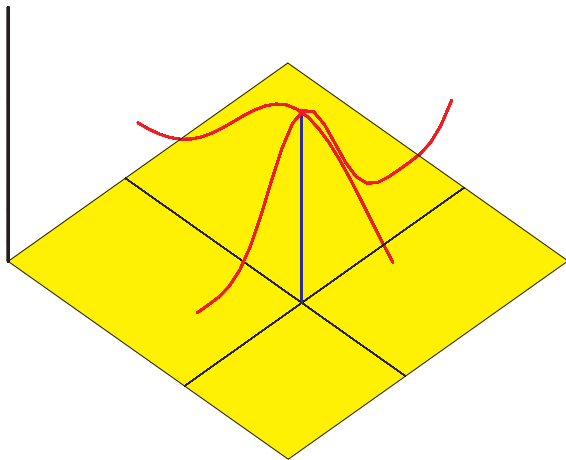


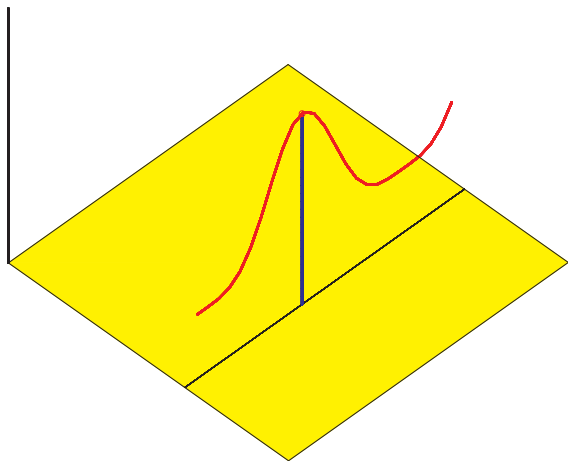
## VII.3. Partial derivatives and tangent hyperplane





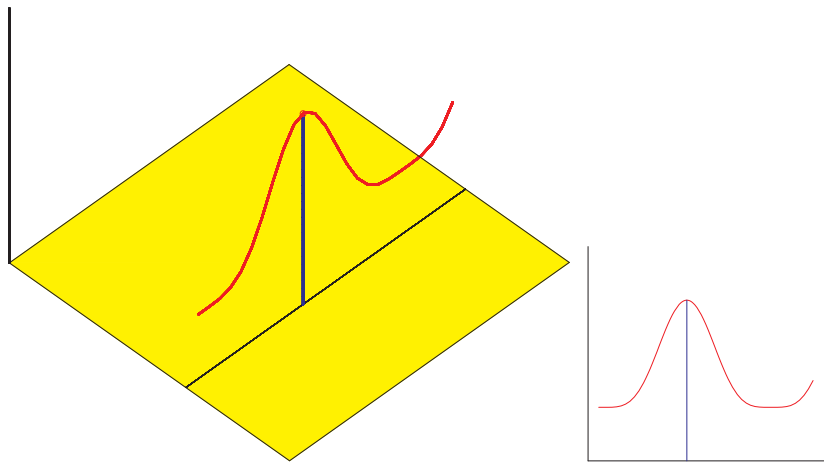








## VII.3. Partial derivatives and tangent hyperplane



## Definition

Let  $G \subset \mathbb{R}^n$  be a non-empty open set. If a function  $f: G \rightarrow \mathbb{R}$  has all partial derivatives continuous at each point of the set  $G$  (i.e. the function  $\mathbf{x} \mapsto \frac{\partial f}{\partial x_j}(\mathbf{x})$  is continuous on  $G$  for each  $j \in \{1, \dots, n\}$ ), then we say that  $f$  is of the **class  $C^1$  on  $G$** . The set of all of these functions is denoted by  $C^1(G)$ .

## Definition

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## Remark

If  $G \subset \mathbb{R}^n$  is a non-empty open set and  $f, g \in C^1(G)$ , then  $f + g \in C^1(G)$ ,  $f - g \in C^1(G)$ , and  $fg \in C^1(G)$ . If moreover  $g(\mathbf{x}) \neq 0$  for each  $\mathbf{x} \in G$ , then  $f/g \in C^1(G)$ .

## Proposition 39 (weak Lagrange theorem)

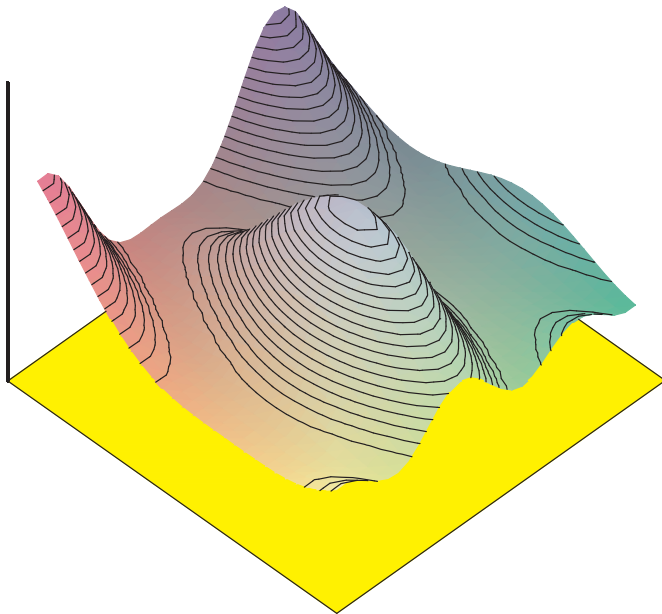
*Omitted.*

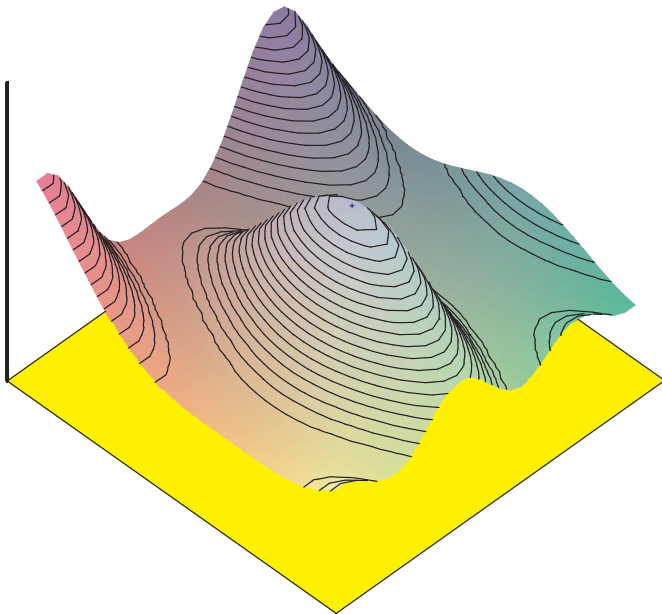
## Definition

Let  $G \subset \mathbb{R}^n$  be an open set,  $\mathbf{a} \in G$ , and  $f \in C^1(G)$ . Then the graph of the function

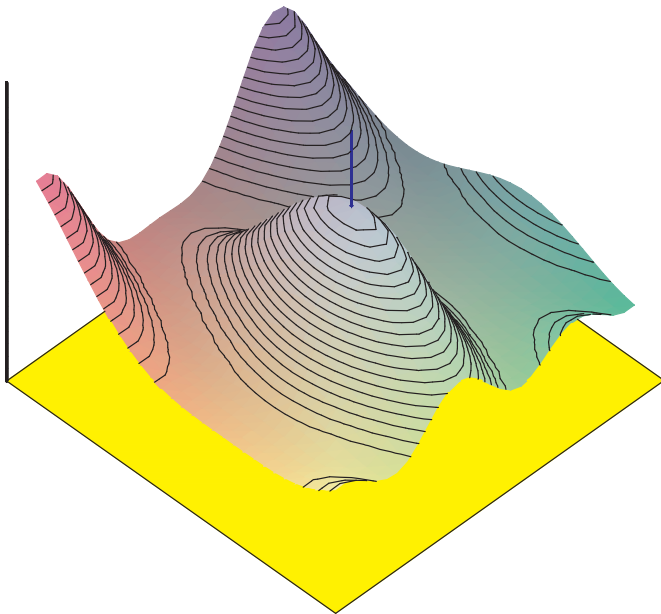
$$T: \mathbf{x} \mapsto f(\mathbf{a}) + \frac{\partial f}{\partial x_1}(\mathbf{a})(x_1 - a_1) + \frac{\partial f}{\partial x_2}(\mathbf{a})(x_2 - a_2) + \cdots + \frac{\partial f}{\partial x_n}(\mathbf{a})(x_n - a_n), \quad \mathbf{x} \in \mathbb{R}^n,$$

is called the **tangent hyperplane** to the graph of the function  $f$  at the point  $[\mathbf{a}, f(\mathbf{a})]$ .



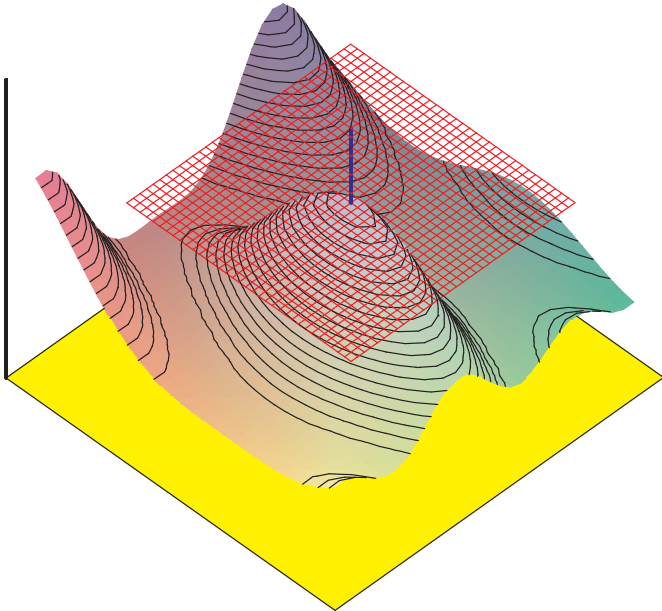


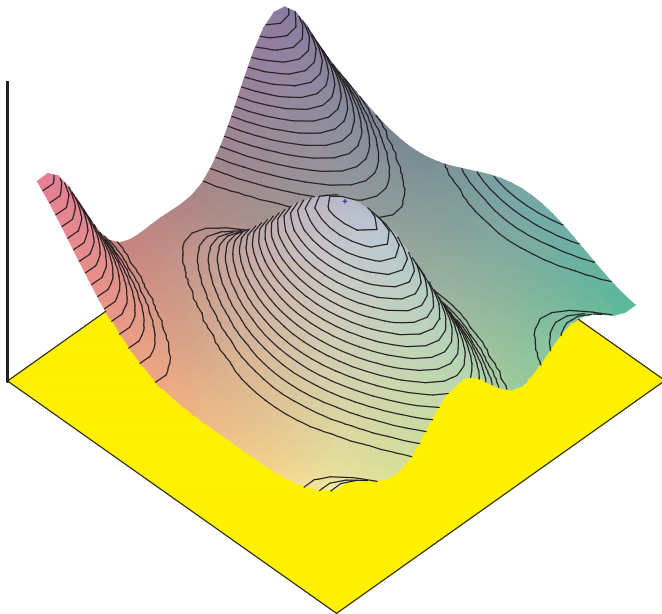
## VII.3. Partial derivatives and tangent hyperplane



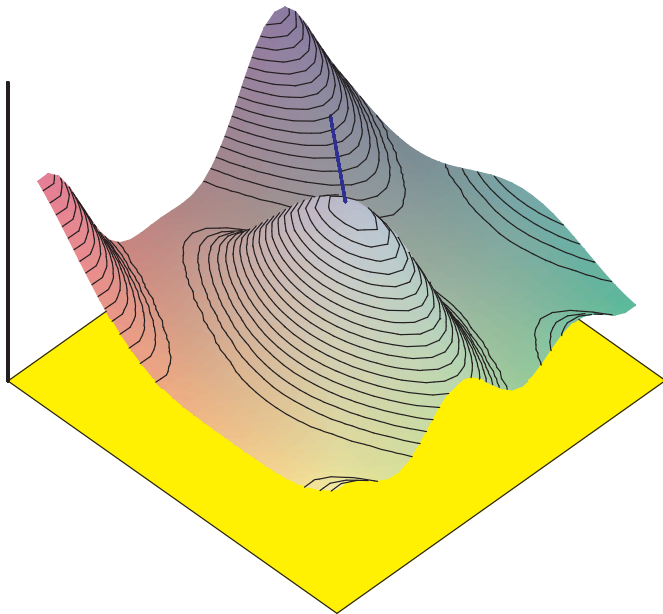


## VII.3. Partial derivatives and tangent hyperplane

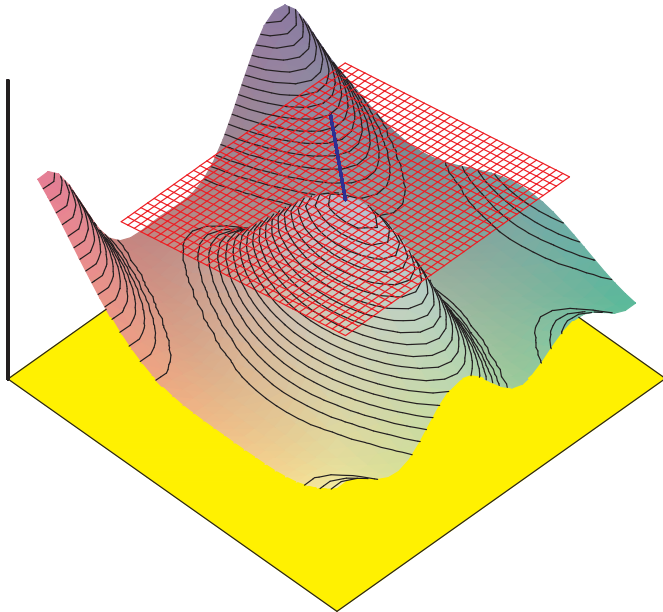




## VII.3. Partial derivatives and tangent hyperplane



## VII.3. Partial derivatives and tangent hyperplane



## Theorem 40 (tangent hyperplane)

Let  $G \subset \mathbb{R}^n$  be an open set,  $\mathbf{a} \in G$ ,  $f \in C^1(G)$ , and let  $T$  be a function whose graph is the tangent hyperplane of the function  $f$  at the point  $[\mathbf{a}, f(\mathbf{a})]$ . Then

$$\lim_{\mathbf{x} \rightarrow \mathbf{a}} \frac{f(\mathbf{x}) - T(\mathbf{x})}{\rho(\mathbf{x}, \mathbf{a})} = 0.$$

## Theorem 40 (tangent hyperplane)

Let  $G \subset \mathbb{R}^n$  be an open set,  $\mathbf{a} \in G$ ,  $f \in C^1(G)$ , and let  $T$  be a function whose graph is the tangent hyperplane of the function  $f$  at the point  $[\mathbf{a}, f(\mathbf{a})]$ . Then

$$\lim_{\mathbf{x} \rightarrow \mathbf{a}} \frac{f(\mathbf{x}) - T(\mathbf{x})}{\rho(\mathbf{x}, \mathbf{a})} = 0.$$

## Theorem 41

Let  $G \subset \mathbb{R}^n$  be an open non-empty set and  $f \in C^1(G)$ . Then  $f$  is continuous on  $G$ .

## Theorem 40 (tangent hyperplane)

Let  $G \subset \mathbb{R}^n$  be an open set,  $\mathbf{a} \in G$ ,  $f \in C^1(G)$ , and let  $T$  be a function whose graph is the tangent hyperplane of the function  $f$  at the point  $[\mathbf{a}, f(\mathbf{a})]$ . Then

$$\lim_{\mathbf{x} \rightarrow \mathbf{a}} \frac{f(\mathbf{x}) - T(\mathbf{x})}{\rho(\mathbf{x}, \mathbf{a})} = 0.$$

## Theorem 41

Let  $G \subset \mathbb{R}^n$  be an open non-empty set and  $f \in C^1(G)$ . Then  $f$  is continuous on  $G$ .

## Remark

Existence of partial derivatives at  $\mathbf{a}$  **does not** imply continuity at  $\mathbf{a}$ .

## Theorem 42 (derivative of a composite function; chain rule)

Let  $r, s \in \mathbb{N}$  and let  $G \subset \mathbb{R}^s$ ,  $H \subset \mathbb{R}^r$  be open sets. Let  $\varphi_1, \dots, \varphi_r \in C^1(G)$ ,  $f \in C^1(H)$  and  $[\varphi_1(\mathbf{x}), \dots, \varphi_r(\mathbf{x})] \in H$  for each  $\mathbf{x} \in G$ . Then the compound function  $F: G \rightarrow \mathbb{R}$  defined by

$$F(\mathbf{x}) = f(\varphi_1(\mathbf{x}), \varphi_2(\mathbf{x}), \dots, \varphi_r(\mathbf{x})), \quad \mathbf{x} \in G,$$

is of the class  $C^1$  on  $G$ . Let  $\mathbf{a} \in G$  and  $\mathbf{b} = [\varphi_1(\mathbf{a}), \dots, \varphi_r(\mathbf{a})]$ . Then for each  $j \in \{1, \dots, s\}$  we have

$$\frac{\partial F}{\partial x_j}(\mathbf{a}) = \sum_{i=1}^r \frac{\partial f}{\partial y_i}(\mathbf{b}) \frac{\partial \varphi_i}{\partial x_j}(\mathbf{a}).$$



## Definition

Let  $G \subset \mathbb{R}^n$  be an open set,  $\mathbf{a} \in G$ , and  $f \in C^1(G)$ . The **gradient of  $f$  at the point  $\mathbf{a}$**  is the vector

$$\nabla f(\mathbf{a}) = \left[ \frac{\partial f}{\partial x_1}(\mathbf{a}), \frac{\partial f}{\partial x_2}(\mathbf{a}), \dots, \frac{\partial f}{\partial x_n}(\mathbf{a}) \right].$$

## Definition

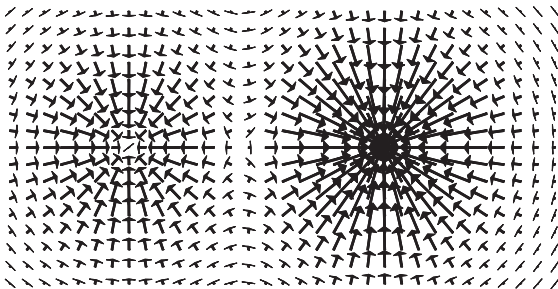
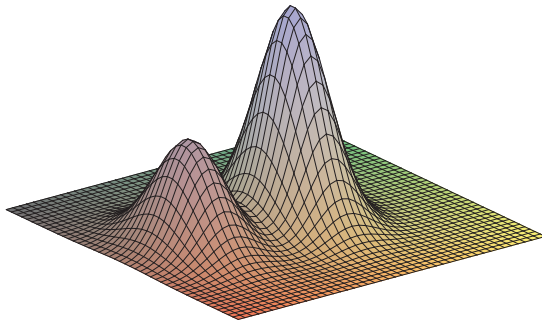
Let  $G \subset \mathbb{R}^n$  be an open set,  $\mathbf{a} \in G$ , and  $f \in C^1(G)$ . The **gradient of  $f$  at the point  $\mathbf{a}$**  is the vector

$$\nabla f(\mathbf{a}) = \left[ \frac{\partial f}{\partial x_1}(\mathbf{a}), \frac{\partial f}{\partial x_2}(\mathbf{a}), \dots, \frac{\partial f}{\partial x_n}(\mathbf{a}) \right].$$

## Remark

The gradient of  $f$  at  $\mathbf{a}$  points in the direction of steepest growth of  $f$  at  $\mathbf{a}$ . At every point, the gradient is perpendicular to the contour of  $f$ .

## VII.3. Partial derivatives and tangent hyperplane



## Definition

Let  $G \subset \mathbb{R}^n$  be an open set,  $\mathbf{a} \in G$ ,  $f \in C^1(G)$ , and  $\nabla f(\mathbf{a}) = \mathbf{0}$ . Then the point  $\mathbf{a}$  is called a **stationary** (or **critical**) **point** of the function  $f$ .

## Definition

Let  $G \subset \mathbb{R}^n$  be an open set,  $f: G \rightarrow \mathbb{R}$ ,  $i, j \in \{1, \dots, n\}$ , and suppose that  $\frac{\partial f}{\partial x_i}(\mathbf{x})$  exists finite for each  $\mathbf{x} \in G$ . Then the **partial derivative of the second order** of the function  $f$  according to  $i$ th and  $j$ th variable at a point  $\mathbf{a} \in G$  is defined by

$$\frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{a}) = \frac{\partial \left( \frac{\partial f}{\partial x_i} \right)}{\partial x_j}(\mathbf{a})$$

If  $i = j$  then we use the notation  $\frac{\partial^2 f}{\partial x_i^2}(\mathbf{a})$ .

## Definition

Let  $G \subset \mathbb{R}^n$  be an open set,  $f: G \rightarrow \mathbb{R}$ ,  $i, j \in \{1, \dots, n\}$ , and suppose that  $\frac{\partial f}{\partial x_i}(\mathbf{x})$  exists finite for each  $\mathbf{x} \in G$ . Then the **partial derivative of the second order** of the function  $f$  according to  $i$ th and  $j$ th variable at a point  $\mathbf{a} \in G$  is defined by

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If  $i = j$  then we use the notation  $\frac{\partial^2 f}{\partial x_i^2}(\mathbf{a})$ .

Similarly we define higher order partial derivatives.

## Remark

In general it is not true that  $\frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{a}) = \frac{\partial^2 f}{\partial x_j \partial x_i}(\mathbf{a})$ .

## Remark

In general it is not true that  $\frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{a}) = \frac{\partial^2 f}{\partial x_j \partial x_i}(\mathbf{a})$ .

## Theorem 43 (interchanging of partial derivatives)

*Let  $i, j \in \{1, \dots, n\}$  and suppose that a function  $f$  has both partial derivatives  $\frac{\partial^2 f}{\partial x_i \partial x_j}$  and  $\frac{\partial^2 f}{\partial x_j \partial x_i}$  on a neighbourhood of a point  $\mathbf{a} \in \mathbb{R}^n$  and that these functions are continuous at  $\mathbf{a}$ . Then*

$$\frac{\partial^2 f}{\partial x_i \partial x_j}(\mathbf{a}) = \frac{\partial^2 f}{\partial x_j \partial x_i}(\mathbf{a}).$$



## Definition

Let  $G \subset \mathbb{R}^n$  be an open set and  $k \in \mathbb{N}$ . We say that a function  $f$  is of the class  $C^k$  on  $G$ , if all partial derivatives of  $f$  of all orders up to  $k$  are continuous on  $G$ . The set of all of these functions is denoted by  $C^k(G)$ .

## Definition

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We say that a function  $f$  is of the **class  $C^\infty$  on  $G$** , if all partial derivatives of all orders of  $f$  are continuous on  $G$ . The set of all of these functions is denoted by  $C^\infty(G)$ .

# VII.4. Implicit function theorem and Lagrange multiplier theorem

## Theorem 44 (implicit function)

*Let  $G \subset \mathbb{R}^{n+1}$  be an open set,  $F: G \rightarrow \mathbb{R}$ , and  $\tilde{\mathbf{x}} \in \mathbb{R}^n$ ,  $\tilde{y} \in \mathbb{R}$  such that  $[\tilde{\mathbf{x}}, \tilde{y}] \in G$ . Suppose that*

## Theorem 44 (implicit function)

Let  $G \subset \mathbb{R}^{n+1}$  be an open set,  $F: G \rightarrow \mathbb{R}$ , and  $\tilde{\mathbf{x}} \in \mathbb{R}^n$ ,  $\tilde{y} \in \mathbb{R}$  such that  $[\tilde{\mathbf{x}}, \tilde{y}] \in G$ . Suppose that

- (i)  $F \in C^1(G)$ ,

## Theorem 44 (implicit function)

Let  $G \subset \mathbb{R}^{n+1}$  be an open set,  $F: G \rightarrow \mathbb{R}$ , and  $\tilde{\mathbf{x}} \in \mathbb{R}^n$ ,  $\tilde{y} \in \mathbb{R}$  such that  $[\tilde{\mathbf{x}}, \tilde{y}] \in G$ . Suppose that

- (i)  $F \in C^1(G)$ ,
- (ii)  $F(\tilde{\mathbf{x}}, \tilde{y}) = 0$ ,

## Theorem 44 (implicit function)

Let  $G \subset \mathbb{R}^{n+1}$  be an open set,  $F: G \rightarrow \mathbb{R}$ , and  $\tilde{\mathbf{x}} \in \mathbb{R}^n$ ,  $\tilde{y} \in \mathbb{R}$  such that  $[\tilde{\mathbf{x}}, \tilde{y}] \in G$ . Suppose that

- (i)  $F \in C^1(G)$ ,
- (ii)  $F(\tilde{\mathbf{x}}, \tilde{y}) = 0$ ,
- (iii)  $\frac{\partial F}{\partial y}(\tilde{\mathbf{x}}, \tilde{y}) \neq 0$ .

### Theorem 44 (implicit function)

Let  $G \subset \mathbb{R}^{n+1}$  be an open set,  $F: G \rightarrow \mathbb{R}$ , and  $\tilde{\mathbf{x}} \in \mathbb{R}^n$ ,  $\tilde{y} \in \mathbb{R}$  such that  $[\tilde{\mathbf{x}}, \tilde{y}] \in G$ . Suppose that

- (i)  $F \in C^1(G)$ ,
- (ii)  $F(\tilde{\mathbf{x}}, \tilde{y}) = 0$ ,
- (iii)  $\frac{\partial F}{\partial y}(\tilde{\mathbf{x}}, \tilde{y}) \neq 0$ .

Then there exist a neighbourhood  $U \subset \mathbb{R}^n$  of the point  $\tilde{\mathbf{x}}$  and a neighbourhood  $V \subset \mathbb{R}$  of the point  $\tilde{y}$  such that for each  $\mathbf{x} \in U$  there exists a unique  $y \in V$  satisfying  $F(\mathbf{x}, y) = 0$ .



## Theorem 44 (implicit function)

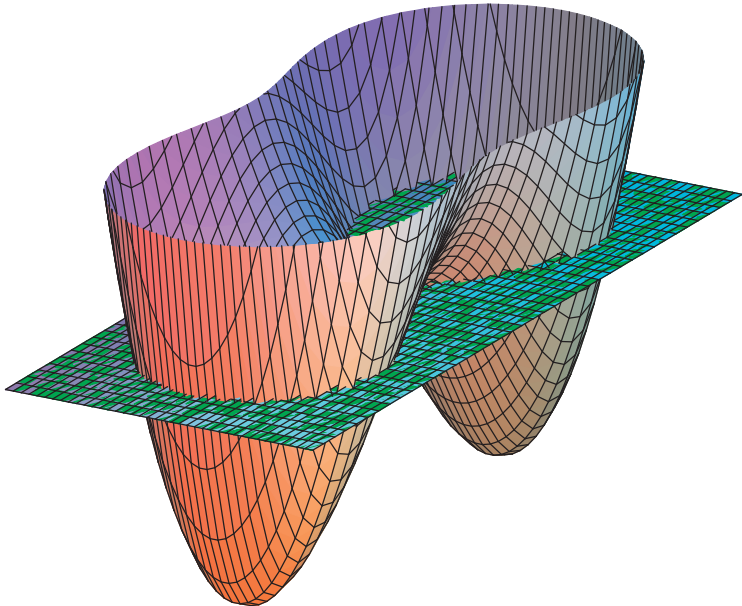
Let  $G \subset \mathbb{R}^{n+1}$  be an open set,  $F: G \rightarrow \mathbb{R}$ , and  $\tilde{\mathbf{x}} \in \mathbb{R}^n$ ,  $\tilde{y} \in \mathbb{R}$  such that  $[\tilde{\mathbf{x}}, \tilde{y}] \in G$ . Suppose that

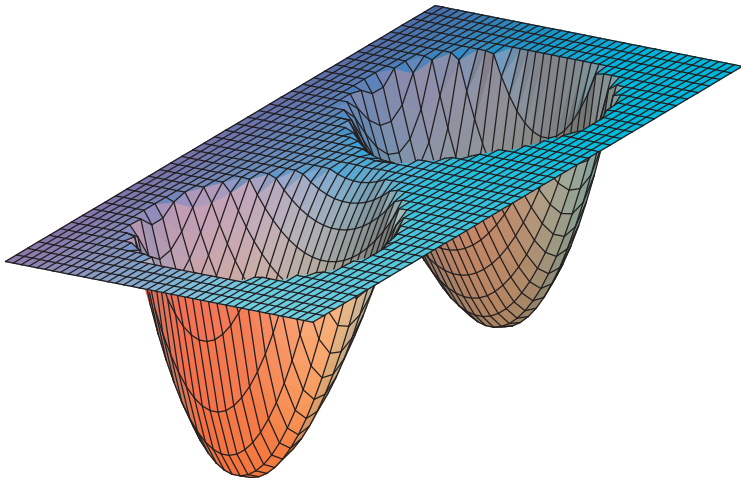
- (i)  $F \in C^1(G)$ ,
- (ii)  $F(\tilde{\mathbf{x}}, \tilde{y}) = 0$ ,
- (iii)  $\frac{\partial F}{\partial y}(\tilde{\mathbf{x}}, \tilde{y}) \neq 0$ .

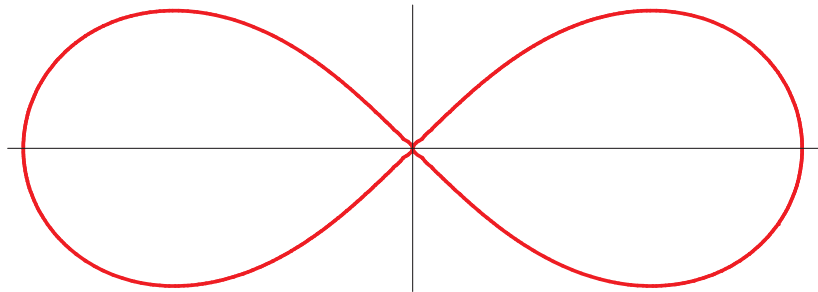
Then there exist a neighbourhood  $U \subset \mathbb{R}^n$  of the point  $\tilde{\mathbf{x}}$  and a neighbourhood  $V \subset \mathbb{R}$  of the point  $\tilde{y}$  such that for each  $\mathbf{x} \in U$  there exists a unique  $y \in V$  satisfying  $F(\mathbf{x}, y) = 0$ . If we denote this  $y$  by  $\varphi(\mathbf{x})$ , then the resulting function  $\varphi$  is in  $C^1(U)$  and

$$\frac{\partial \varphi}{\partial x_j}(\mathbf{x}) = - \frac{\frac{\partial F}{\partial x_j}(\mathbf{x}, \varphi(\mathbf{x}))}{\frac{\partial F}{\partial y}(\mathbf{x}, \varphi(\mathbf{x}))} \quad \text{for } \mathbf{x} \in U, j \in \{1, \dots, n\}.$$

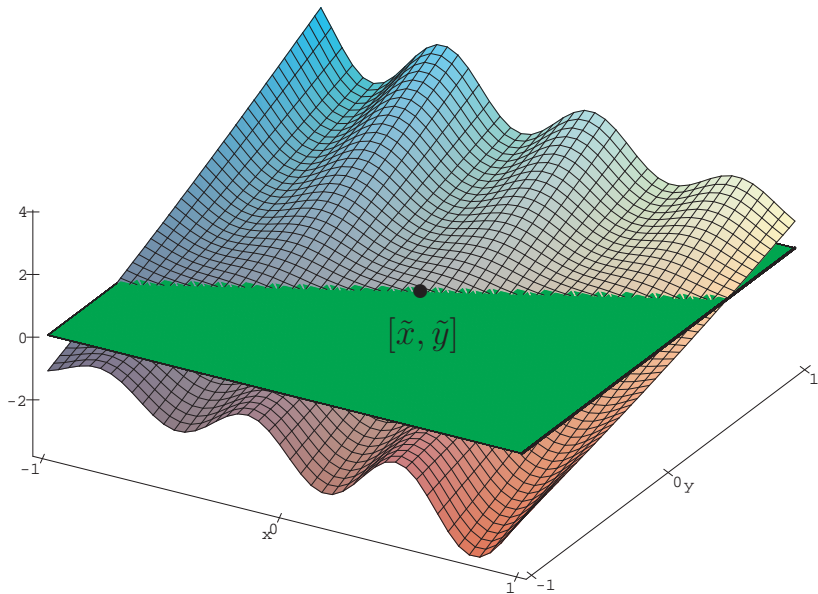
## VII.4. Implicit function theorem and Lagrange multiplier theorem



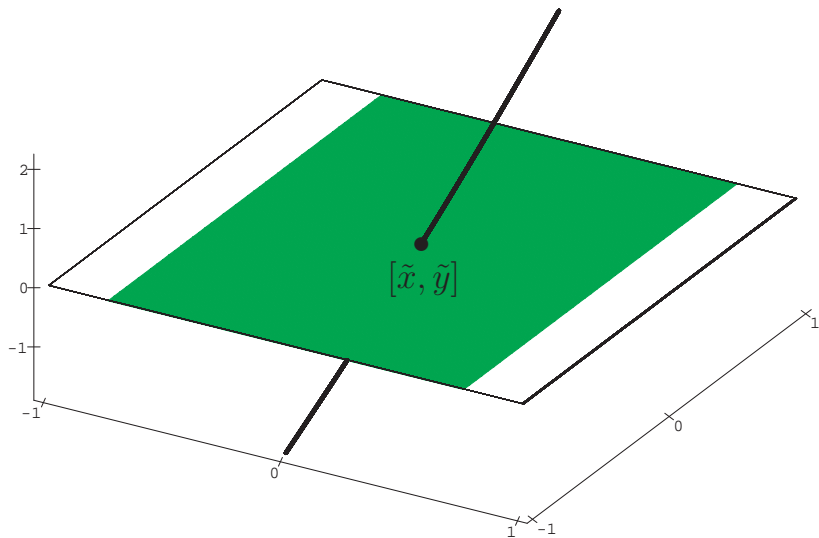




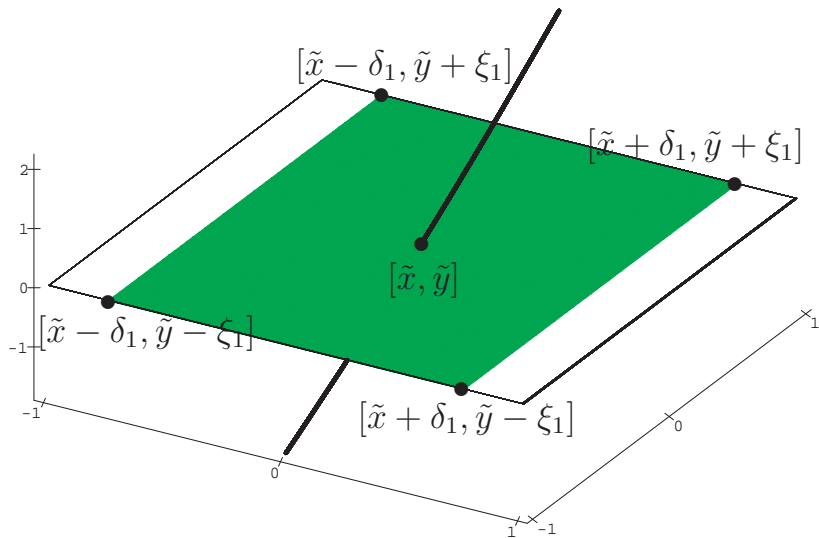
## VII.4. Implicit function theorem and Lagrange multiplier theorem



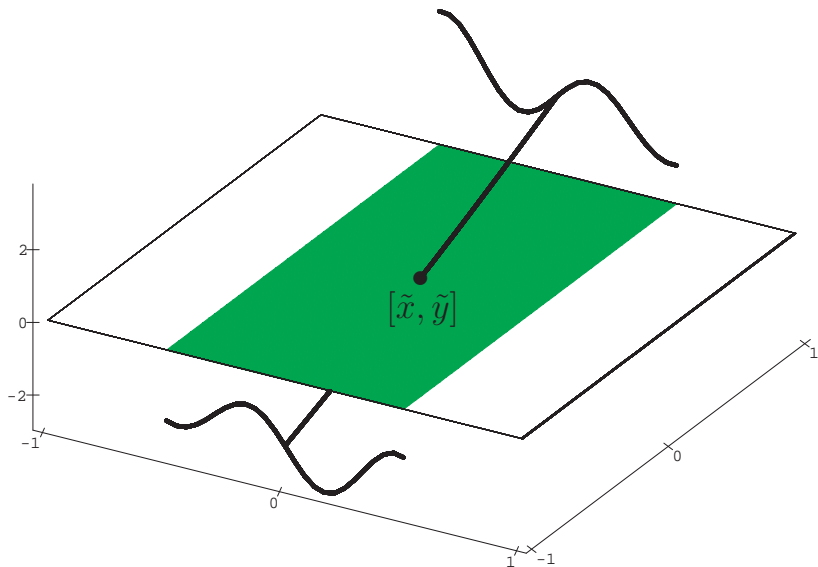
## VII.4. Implicit function theorem and Lagrange multiplier theorem



## VII.4. Implicit function theorem and Lagrange multiplier theorem

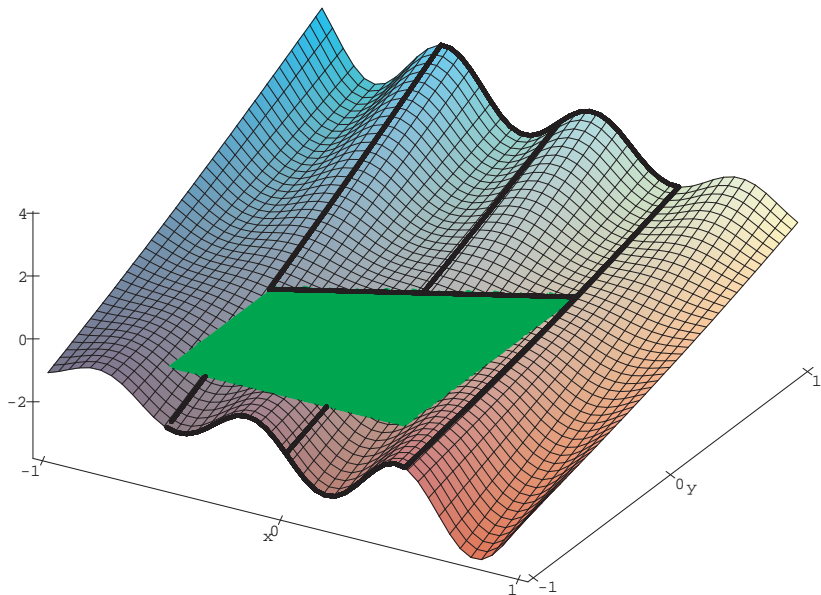


## VII.4. Implicit function theorem and Lagrange multiplier theorem





## VII.4. Implicit function theorem and Lagrange multiplier theorem



## Theorem 45 (Lagrange multiplier theorem)

*Let  $G \subset \mathbb{R}^2$  be an open set,  $f, g \in C^1(G)$ ,  
 $M = \{[x, y] \in G; g(x, y) = 0\}$  and let  $[\tilde{x}, \tilde{y}] \in M$  be a point  
of local extremum of  $f$  with respect to  $M$ . Then at least  
one of the following conditions holds:*

## Theorem 45 (Lagrange multiplier theorem)

*Let  $G \subset \mathbb{R}^2$  be an open set,  $f, g \in C^1(G)$ ,  
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of local extremum of  $f$  with respect to  $M$ . Then at least  
one of the following conditions holds:*

(I)  $\nabla g(\tilde{x}, \tilde{y}) = \mathbf{o}$ ,

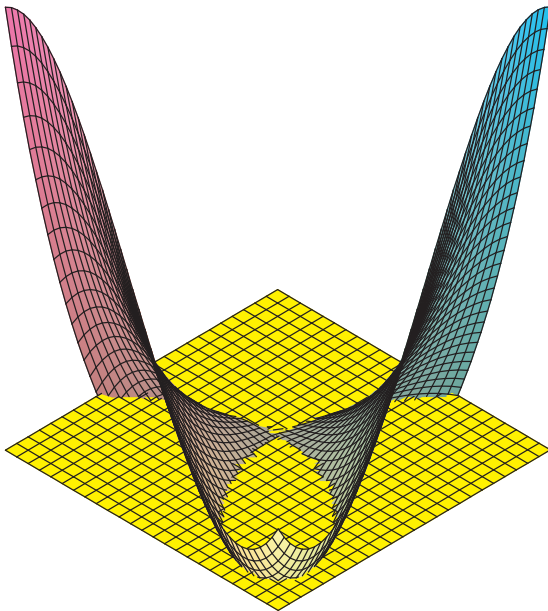
## Theorem 45 (Lagrange multiplier theorem)

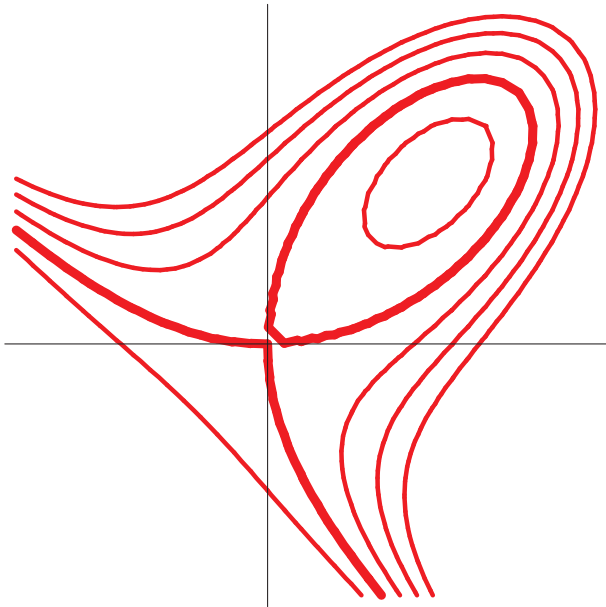
Let  $G \subset \mathbb{R}^2$  be an open set,  $f, g \in C^1(G)$ ,  
 $M = \{[x, y] \in G; g(x, y) = 0\}$  and let  $[\tilde{x}, \tilde{y}] \in M$  be a point  
of local extremum of  $f$  with respect to  $M$ . Then at least  
one of the following conditions holds:

- (I)  $\nabla g(\tilde{x}, \tilde{y}) = \mathbf{o}$ ,
- (II) there exists  $\lambda \in \mathbb{R}$  satisfying

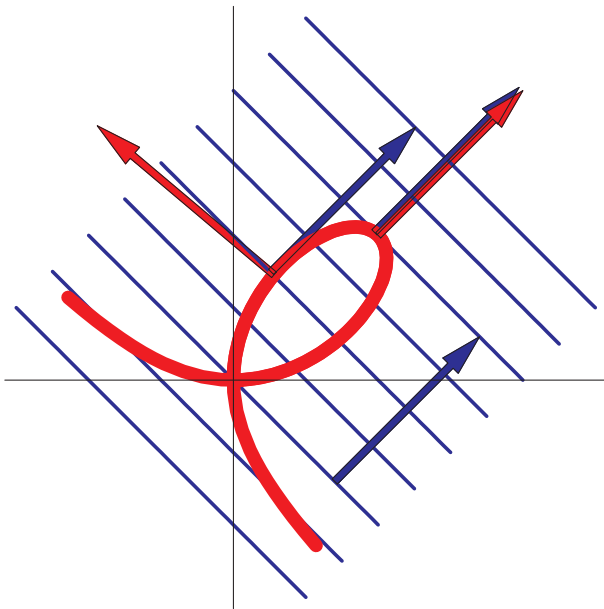
$$\frac{\partial f}{\partial x}(\tilde{x}, \tilde{y}) + \lambda \frac{\partial g}{\partial x}(\tilde{x}, \tilde{y}) = 0,$$
$$\frac{\partial f}{\partial y}(\tilde{x}, \tilde{y}) + \lambda \frac{\partial g}{\partial y}(\tilde{x}, \tilde{y}) = 0.$$

## VII.4. Implicit function theorem and Lagrange multiplier theorem





## VII.4. Implicit function theorem and Lagrange multiplier theorem



## Theorem 46 (implicit functions)

Let  $m, n \in \mathbb{N}$ ,  $k \in \mathbb{N} \cup \{\infty\}$ ,  $G \subset \mathbb{R}^{n+m}$  an open set,  
 $F_j: G \rightarrow \mathbb{R}$  for  $j = 1, \dots, m$ ,  $\tilde{\mathbf{x}} \in \mathbb{R}^n$ ,  $\tilde{\mathbf{y}} \in \mathbb{R}^m$ ,  $[\tilde{\mathbf{x}}, \tilde{\mathbf{y}}] \in G$ .  
Suppose that



## Theorem 46 (implicit functions)

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 $F_j: G \rightarrow \mathbb{R}$  for  $j = 1, \dots, m$ ,  $\tilde{\mathbf{x}} \in \mathbb{R}^n$ ,  $\tilde{\mathbf{y}} \in \mathbb{R}^m$ ,  $[\tilde{\mathbf{x}}, \tilde{\mathbf{y}}] \in G$ .  
Suppose that

- (i)  $F_j \in C^k(G)$  for all  $j \in \{1, \dots, m\}$ ,

## Theorem 46 (implicit functions)

Let  $m, n \in \mathbb{N}$ ,  $k \in \mathbb{N} \cup \{\infty\}$ ,  $G \subset \mathbb{R}^{n+m}$  an open set,  
 $F_j: G \rightarrow \mathbb{R}$  for  $j = 1, \dots, m$ ,  $\tilde{\mathbf{x}} \in \mathbb{R}^n$ ,  $\tilde{\mathbf{y}} \in \mathbb{R}^m$ ,  $[\tilde{\mathbf{x}}, \tilde{\mathbf{y}}] \in G$ .  
Suppose that

- (i)  $F_j \in C^k(G)$  for all  $j \in \{1, \dots, m\}$ ,
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- (iii) 
$$\begin{vmatrix} \frac{\partial F_1}{\partial y_1}(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) & \dots & \frac{\partial F_1}{\partial y_m}(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) \\ \vdots & \ddots & \vdots \\ \frac{\partial F_m}{\partial y_1}(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) & \dots & \frac{\partial F_m}{\partial y_m}(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) \end{vmatrix} \neq 0.$$

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Then there are a neighbourhood  $U \subset \mathbb{R}^n$  of  $\tilde{\mathbf{x}}$  and a neighbourhood  $V \subset \mathbb{R}^m$  of  $\tilde{\mathbf{y}}$  such that for each  $\mathbf{x} \in U$  there exists a unique  $\mathbf{y} \in V$  satisfying  $F_j(\mathbf{x}, \mathbf{y}) = 0$  for each  $j \in \{1, \dots, m\}$ .

## Theorem 46 (implicit functions)

Let  $m, n \in \mathbb{N}$ ,  $k \in \mathbb{N} \cup \{\infty\}$ ,  $G \subset \mathbb{R}^{n+m}$  an open set,  $F_j: G \rightarrow \mathbb{R}$  for  $j = 1, \dots, m$ ,  $\tilde{\mathbf{x}} \in \mathbb{R}^n$ ,  $\tilde{\mathbf{y}} \in \mathbb{R}^m$ ,  $[\tilde{\mathbf{x}}, \tilde{\mathbf{y}}] \in G$ . Suppose that

- (i)  $F_j \in C^k(G)$  for all  $j \in \{1, \dots, m\}$ ,
- (ii)  $F_j(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) = 0$  for all  $j \in \{1, \dots, m\}$ ,
- (iii) 
$$\begin{vmatrix} \frac{\partial F_1}{\partial y_1}(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) & \dots & \frac{\partial F_1}{\partial y_m}(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) \\ \vdots & \ddots & \vdots \\ \frac{\partial F_m}{\partial y_1}(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) & \dots & \frac{\partial F_m}{\partial y_m}(\tilde{\mathbf{x}}, \tilde{\mathbf{y}}) \end{vmatrix} \neq 0.$$

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### Remark

The symbol in the condition (iii) of Theorem 46 is called a **determinant**. The general definition will be given later.

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For  $m = 1$  we have  $|a| = a$ ,  $a \in \mathbb{R}$ . In particular, in this case the condition (iii) in Theorem 46 is the same as the condition (iii) in Theorem 44.

For  $m = 2$  we have  $\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$ ,  $a, b, c, d \in \mathbb{R}$ .

## Theorem 47 (Lagrange multipliers theorem)

Let  $m, n \in \mathbb{N}$ ,  $m < n$ ,  $G \subset \mathbb{R}^n$  an open set,  
 $f, g_1, \dots, g_m \in C^1(G)$ ,

$$M = \{\mathbf{z} \in G; g_1(\mathbf{z}) = 0, g_2(\mathbf{z}) = 0, \dots, g_m(\mathbf{z}) = 0\}$$

and let  $\tilde{\mathbf{z}} \in M$  be a point of local extremum of  $f$  with respect to the set  $M$ . Then at least one of the following conditions holds:



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and let  $\tilde{\mathbf{z}} \in M$  be a point of local extremum of  $f$  with respect to the set  $M$ . Then at least one of the following conditions holds:

(I) the vectors

$\nabla g_1(\tilde{\mathbf{z}}), \nabla g_2(\tilde{\mathbf{z}}), \dots, \nabla g_m(\tilde{\mathbf{z}})$   
are linearly dependent,

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and let  $\tilde{\mathbf{z}} \in M$  be a point of local extremum of  $f$  with respect to the set  $M$ . Then at least one of the following conditions holds:

(I) the vectors

$$\nabla g_1(\tilde{\mathbf{z}}), \nabla g_2(\tilde{\mathbf{z}}), \dots, \nabla g_m(\tilde{\mathbf{z}})$$

are linearly dependent,

(II) there exist numbers  $\lambda_1, \lambda_2, \dots, \lambda_m \in \mathbb{R}$  satisfying

$$\nabla f(\tilde{\mathbf{z}}) + \lambda_1 \nabla g_1(\tilde{\mathbf{z}}) + \lambda_2 \nabla g_2(\tilde{\mathbf{z}}) + \dots + \lambda_m \nabla g_m(\tilde{\mathbf{z}}) = \mathbf{o}.$$

### Remark

- The notion of **linearly dependent vectors** will be defined later.

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For  $m = 1$ : One vector is linearly dependent if it is the zero vector.

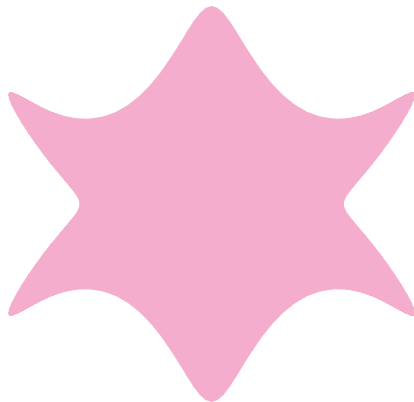
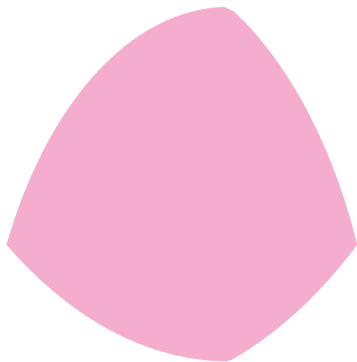
For  $m = 2$ : Two vectors are linearly dependent if one of them is a multiple of the other one.

## Remark

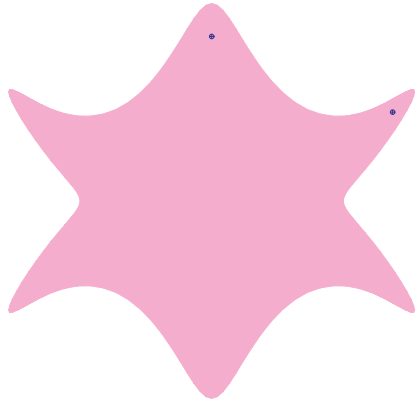
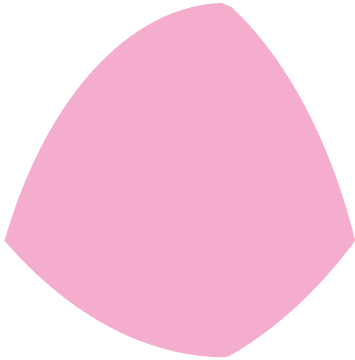
- The notion of **linearly dependent vectors** will be defined later.  
For  $m = 1$ : One vector is linearly dependent if it is the zero vector.  
For  $m = 2$ : Two vectors are linearly dependent if one of them is a multiple of the other one.
- The numbers  $\lambda_1, \dots, \lambda_m$  are called the **Lagrange multipliers**.

# VII.5. Concave and quasiconcave functions

## VII.5. Concave and quasiconcave functions

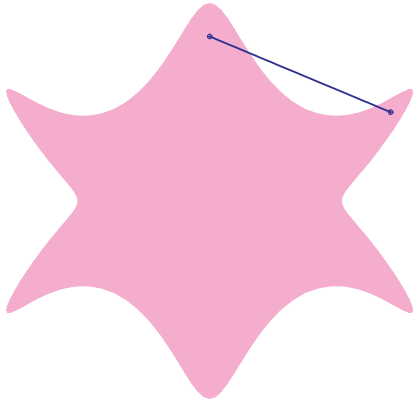
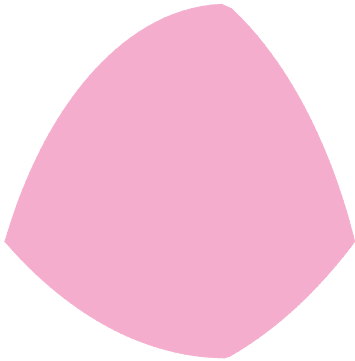


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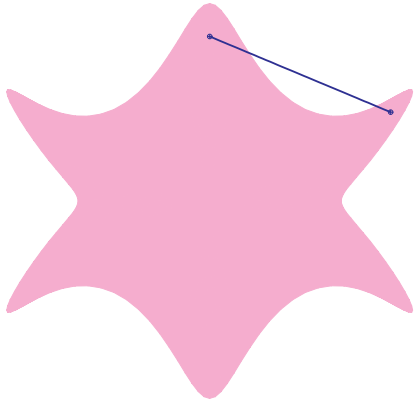
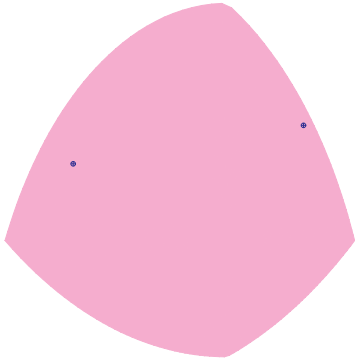




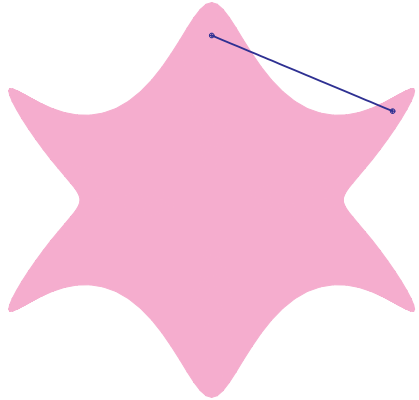
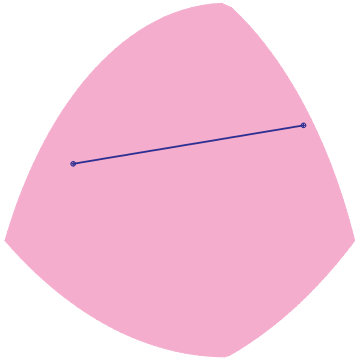
## VII.5. Concave and quasiconcave functions



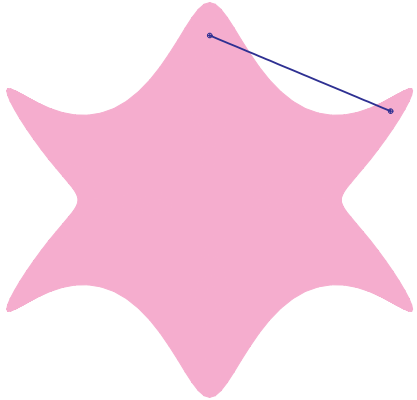
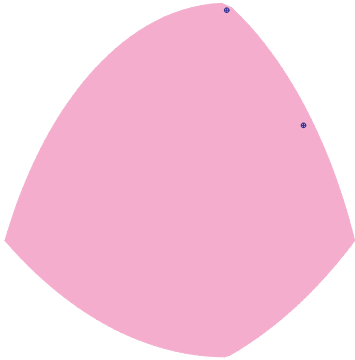
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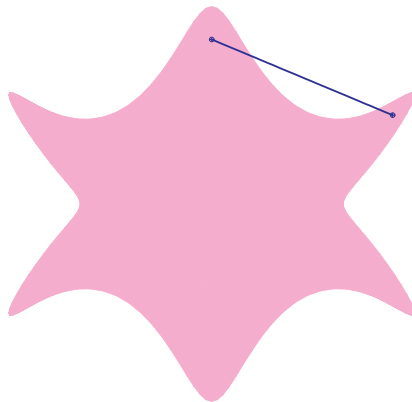
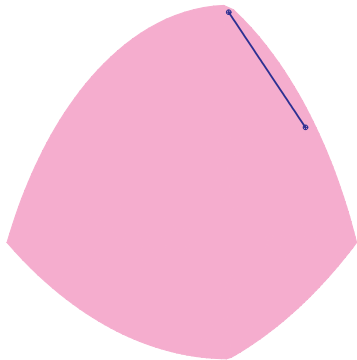
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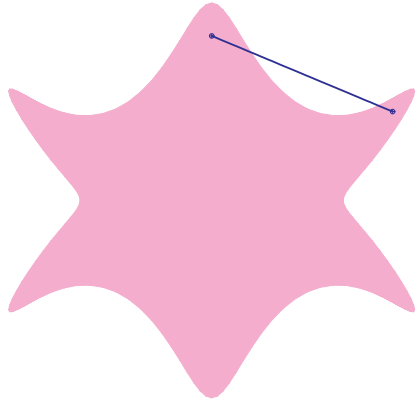
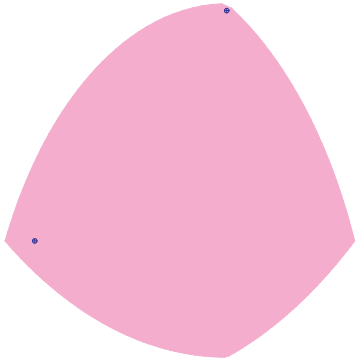
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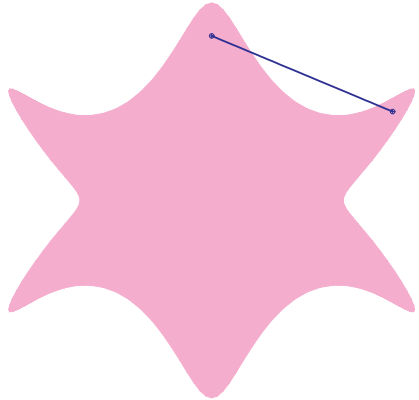
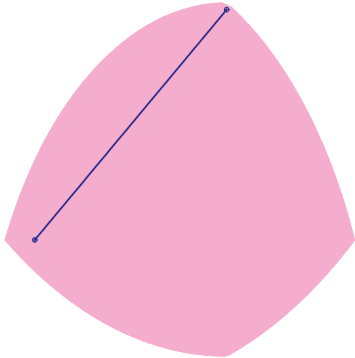
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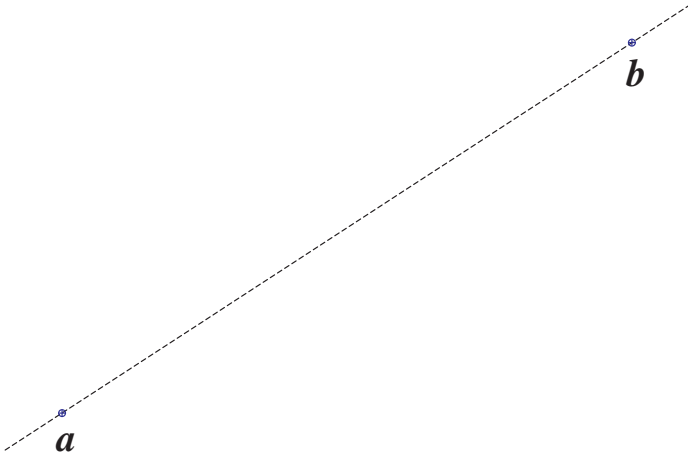


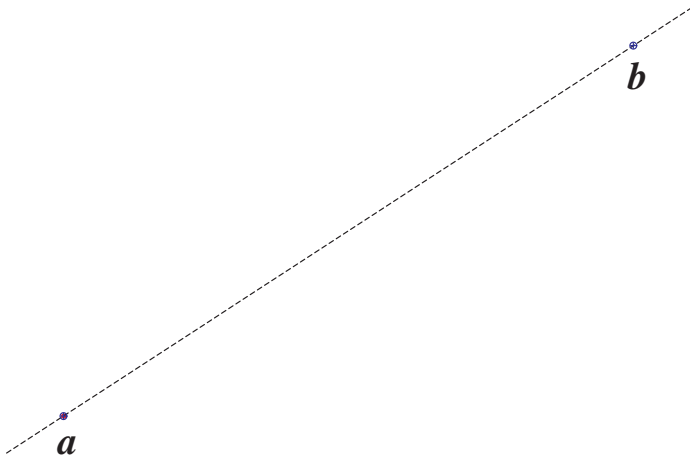
$\oplus$   
***b***

$\oplus$   
***a***

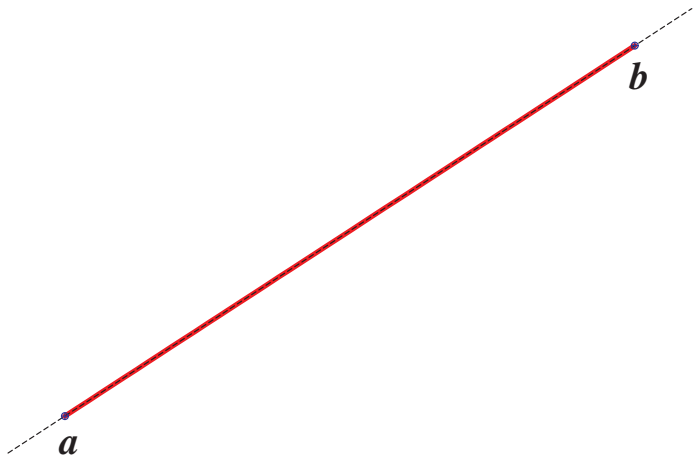


## VII.5. Concave and quasiconcave functions



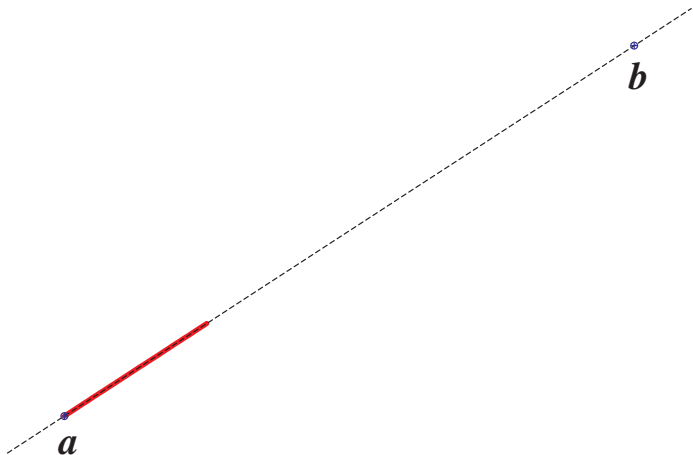


$$\mathbf{a} = 1 \cdot \mathbf{a} + 0 \cdot \mathbf{b} = \mathbf{a} + 0 \cdot (\mathbf{b} - \mathbf{a})$$

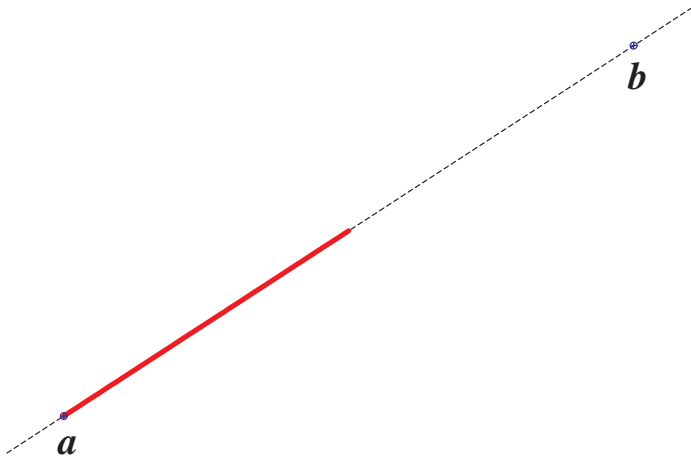


$$\mathbf{b} = 0 \cdot \mathbf{a} + 1 \cdot \mathbf{b} = \mathbf{a} + 1 \cdot (\mathbf{b} - \mathbf{a})$$

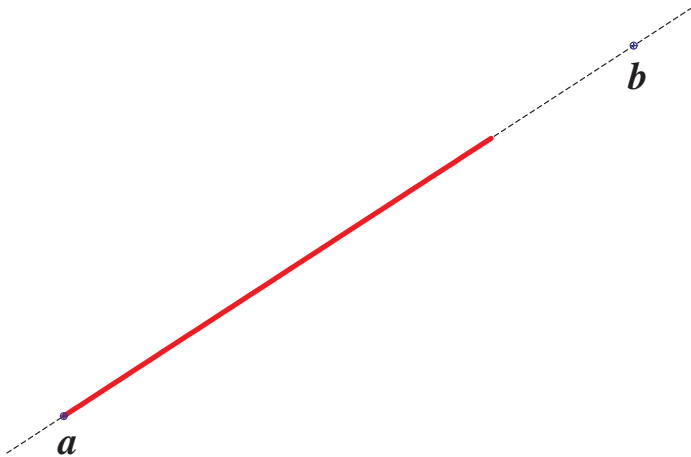
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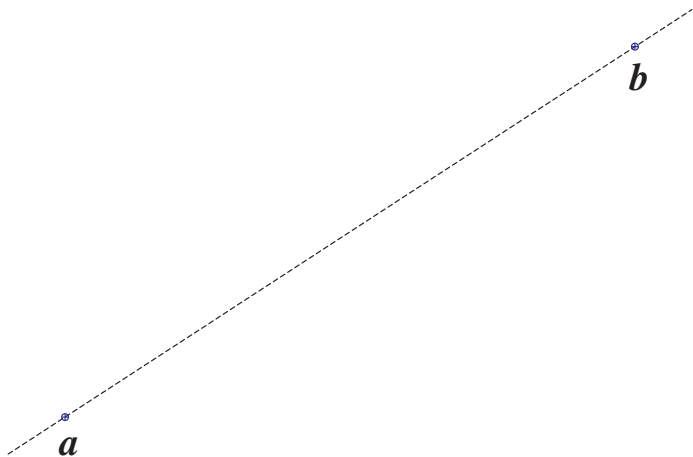
$$\frac{3}{4} \cdot \mathbf{a} + \frac{1}{4} \cdot \mathbf{b} = \mathbf{a} + \frac{1}{4} \cdot (\mathbf{b} - \mathbf{a})$$



$$\frac{1}{2} \cdot \mathbf{a} + \frac{1}{2} \cdot \mathbf{b} = \mathbf{a} + \frac{1}{2} \cdot (\mathbf{b} - \mathbf{a})$$



$$\frac{1}{4} \cdot \mathbf{a} + \frac{3}{4} \cdot \mathbf{b} = \mathbf{a} + \frac{3}{4} \cdot (\mathbf{b} - \mathbf{a})$$



$$t \cdot \mathbf{a} + (1 - t) \cdot \mathbf{b} = \mathbf{a} + (1 - t) \cdot (\mathbf{b} - \mathbf{a})$$

## Definition

Let  $M \subset \mathbb{R}^n$ . We say that  $M$  is **convex** if

$$\forall \mathbf{x}, \mathbf{y} \in M \forall t \in [0, 1]: t\mathbf{x} + (1 - t)\mathbf{y} \in M.$$



## Definition

Let  $M \subset \mathbb{R}^n$  be a convex set and  $f$  a function defined on  $M$ . We say that  $f$  is

- **concave on  $M$**  if

$$\forall \mathbf{a}, \mathbf{b} \in M \forall t \in [0, 1]: f(t\mathbf{a} + (1-t)\mathbf{b}) \geq tf(\mathbf{a}) + (1-t)f(\mathbf{b}),$$

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- **strictly concave on  $M$**  if

$$\forall \mathbf{a}, \mathbf{b} \in M, \mathbf{a} \neq \mathbf{b} \forall t \in (0, 1): \\ f(t\mathbf{a} + (1-t)\mathbf{b}) > tf(\mathbf{a}) + (1-t)f(\mathbf{b}).$$

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## Remark

By changing the inequalities to the opposite we obtain a definition of a *convex* and a *strictly convex* function.

### Remark

A function  $f$  is convex (strictly convex) if and only if the function  $-f$  is concave (strictly concave).

All the theorems in this section are formulated for concave and strictly concave functions. They have obvious analogies that hold for convex and strictly convex functions.

### Remark

- If a function  $f$  is strictly concave on  $M$ , then it is concave on  $M$ .

## Remark

- If a function  $f$  is strictly concave on  $M$ , then it is concave on  $M$ .
- Let  $f$  be a concave function on  $M$ . Then  $f$  is strictly concave on  $M$  if and only if the graph of  $f$  “does not contain a segment”, i.e.

$$\neg(\exists \mathbf{a}, \mathbf{b} \in M, \mathbf{a} \neq \mathbf{b}, \forall t \in [0, 1]:$$

$$f(t\mathbf{a} + (1 - t)\mathbf{b}) = tf(\mathbf{a}) + (1 - t)f(\mathbf{b}))$$

### Theorem 48

*Let  $f$  be a function concave on an open convex set  $G \subset \mathbb{R}^n$ . Then  $f$  is continuous on  $G$ .*

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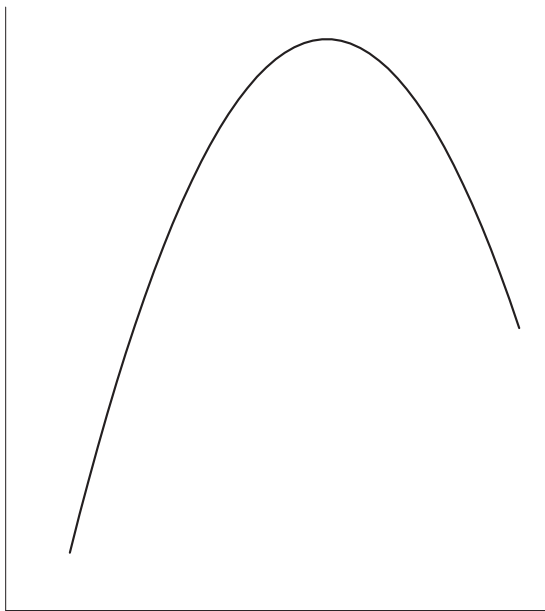
## Theorem 49 (characterisation of strictly concave functions of the class $C^1$ )

*Let  $G \subset \mathbb{R}^n$  be a convex open set and  $f \in C^1(G)$ . Then the function  $f$  is strictly concave on  $G$  if and only if*

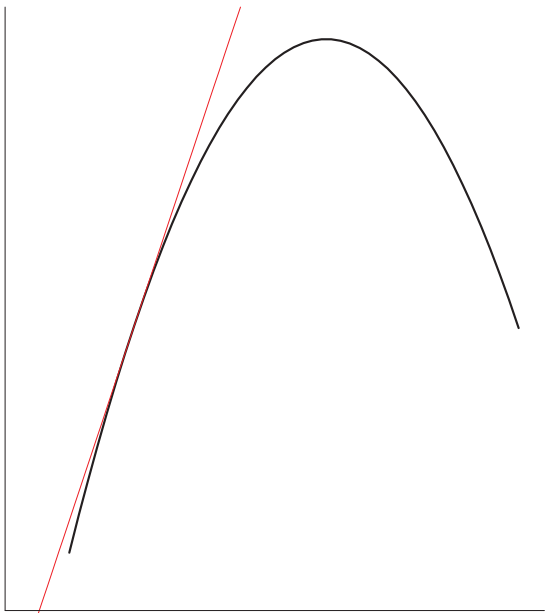
$$\forall \mathbf{x}, \mathbf{y} \in G, \mathbf{x} \neq \mathbf{y}: f(\mathbf{y}) < f(\mathbf{x}) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{x})(y_i - x_i).$$



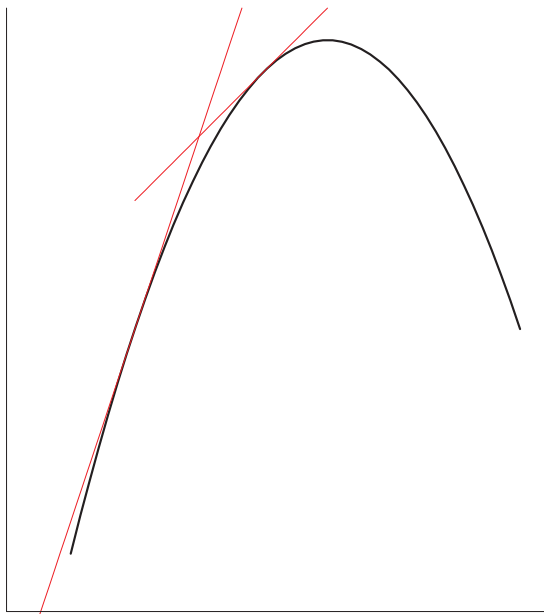
## VII.5. Concave and quasiconcave functions



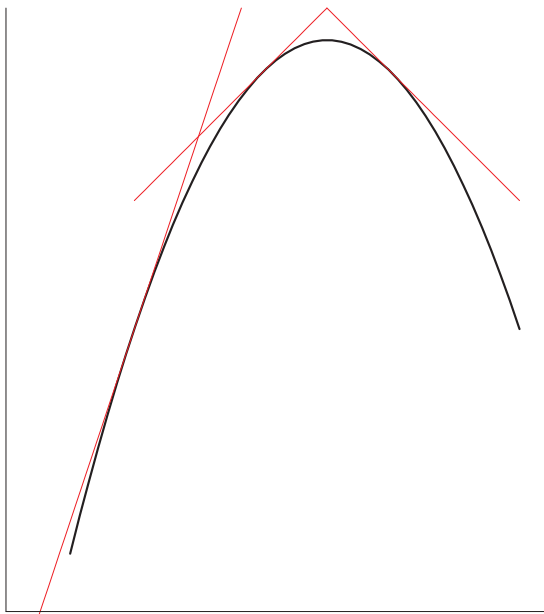
## VII.5. Concave and quasiconcave functions



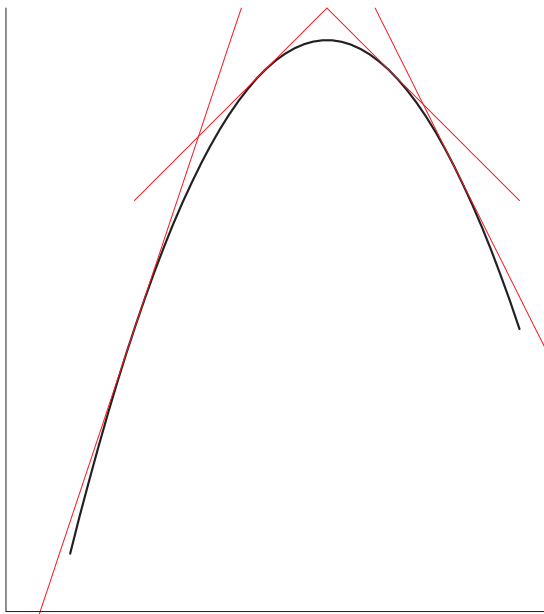
## VII.5. Concave and quasiconcave functions



## VII.5. Concave and quasiconcave functions



## VII.5. Concave and quasiconcave functions



## Theorem 50 (characterisation of concave functions of the class $C^1$ )

*Let  $G \subset \mathbb{R}^n$  be a convex open set and  $f \in C^1(G)$ . Then the function  $f$  is concave on  $G$  if and only if*

$$\forall \mathbf{x}, \mathbf{y} \in G: f(\mathbf{y}) \leq f(\mathbf{x}) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{x})(y_i - x_i).$$

## Theorem 50 (characterisation of concave functions of the class $C^1$ )

*Let  $G \subset \mathbb{R}^n$  be a convex open set and  $f \in C^1(G)$ . Then the function  $f$  is concave on  $G$  if and only if*

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## Corollary 51

*Let  $G \subset \mathbb{R}^n$  be a convex open set,  $f \in C^1(G)$ , and let  $\mathbf{a} \in G$  be a critical point of  $f$  (i.e.  $\nabla f(\mathbf{a}) = \mathbf{0}$ ). If  $f$  is concave on  $G$ , then  $\mathbf{a}$  is a maximum point of  $f$  on  $G$ .*

## Theorem 50 (characterisation of concave functions of the class $C^1$ )

*Let  $G \subset \mathbb{R}^n$  be a convex open set and  $f \in C^1(G)$ . Then the function  $f$  is concave on  $G$  if and only if*

$$\forall \mathbf{x}, \mathbf{y} \in G: f(\mathbf{y}) \leq f(\mathbf{x}) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{x})(y_i - x_i).$$

## Corollary 51

*Let  $G \subset \mathbb{R}^n$  be a convex open set,  $f \in C^1(G)$ , and let  $\mathbf{a} \in G$  be a critical point of  $f$  (i.e.  $\nabla f(\mathbf{a}) = \mathbf{0}$ ). If  $f$  is concave on  $G$ , then  $\mathbf{a}$  is a maximum point of  $f$  on  $G$ . If  $f$  is strictly concave on  $G$ , then  $\mathbf{a}$  is a strict maximum point of  $f$  on  $G$ .*



### Theorem 52 (level sets of concave functions)

*Let  $f$  be a function concave on a convex set  $M \subset \mathbb{R}^n$ . Then for each  $\alpha \in \mathbb{R}$  the set  $Q_\alpha = \{\mathbf{x} \in M; f(\mathbf{x}) \geq \alpha\}$  is convex.*

## Definition

Let  $M \subset \mathbb{R}^n$  be a convex set and let  $f$  be a function defined on  $M$ . We say that  $f$  is

- **quasiconcave on  $M$**  if

$$\forall \mathbf{a}, \mathbf{b} \in M \forall t \in [0, 1]: f(t\mathbf{a} + (1-t)\mathbf{b}) \geq \min\{f(\mathbf{a}), f(\mathbf{b})\},$$

## Definition

Let  $M \subset \mathbb{R}^n$  be a convex set and let  $f$  be a function defined on  $M$ . We say that  $f$  is

- **quasiconcave on  $M$**  if

$$\forall \mathbf{a}, \mathbf{b} \in M \forall t \in [0, 1]: f(t\mathbf{a} + (1-t)\mathbf{b}) \geq \min\{f(\mathbf{a}), f(\mathbf{b})\},$$

- **strictly quasiconcave on  $M$**  if

$$\forall \mathbf{a}, \mathbf{b} \in M, \mathbf{a} \neq \mathbf{b}, \forall t \in (0, 1): \\ f(t\mathbf{a} + (1-t)\mathbf{b}) > \min\{f(\mathbf{a}), f(\mathbf{b})\}.$$

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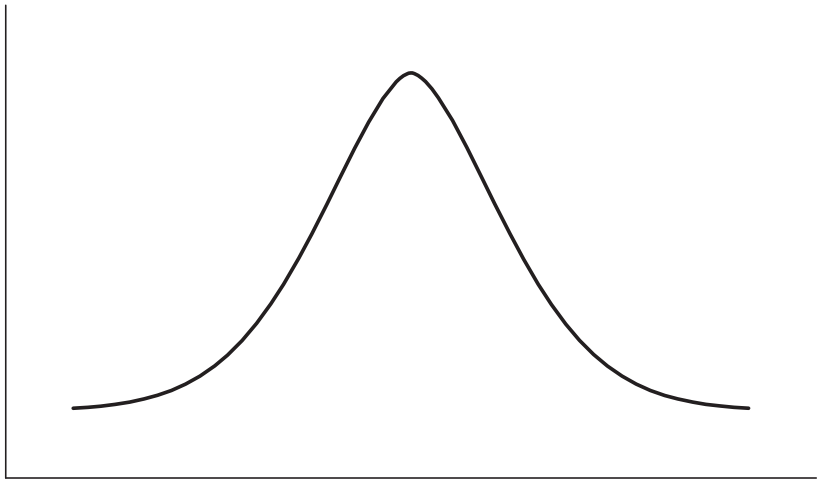
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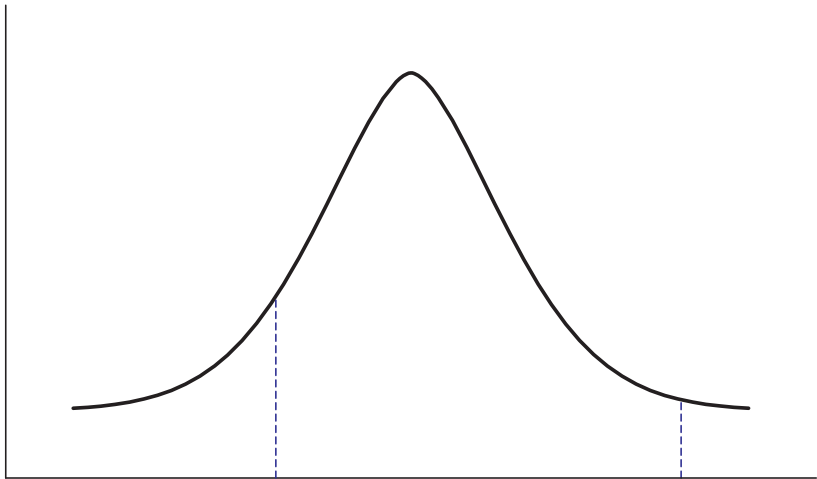
## Remark

By changing the inequalities to the opposite and changing the minimum to a maximum we obtain a definition of a *quasiconvex* and a *strictly quasiconvex* function.

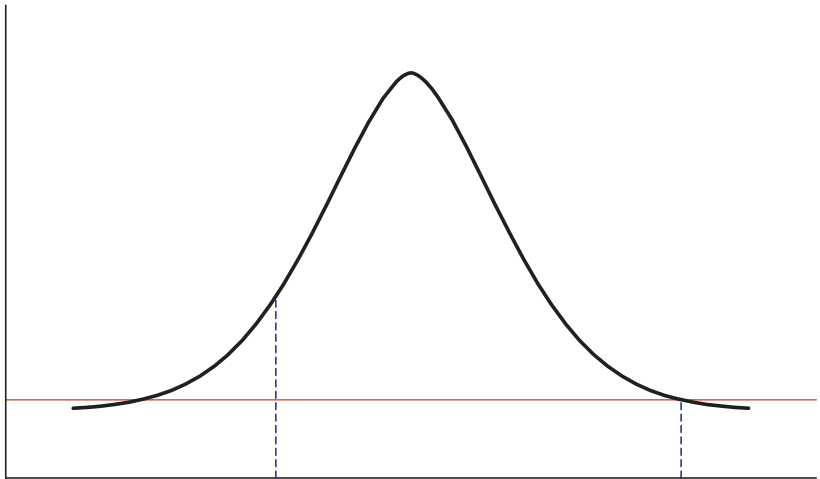
## VII.5. Concave and quasiconcave functions



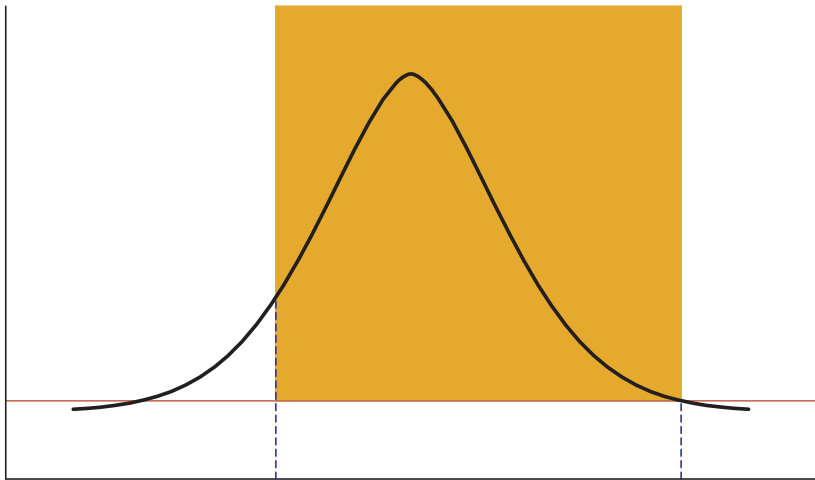
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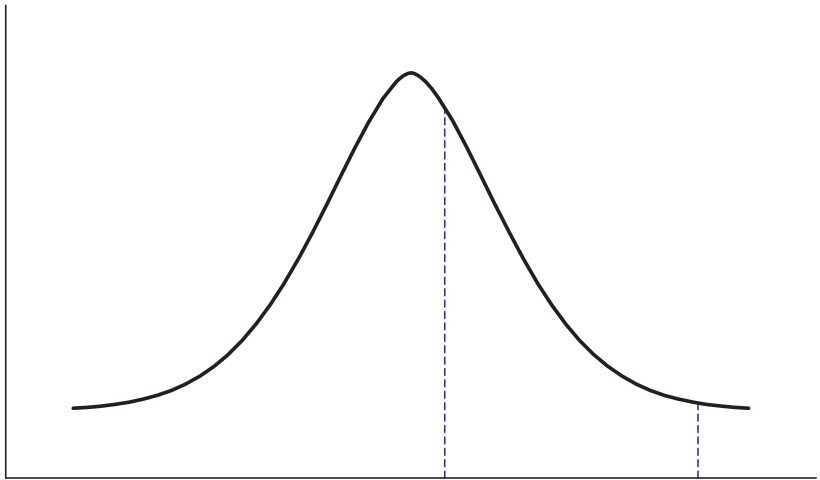


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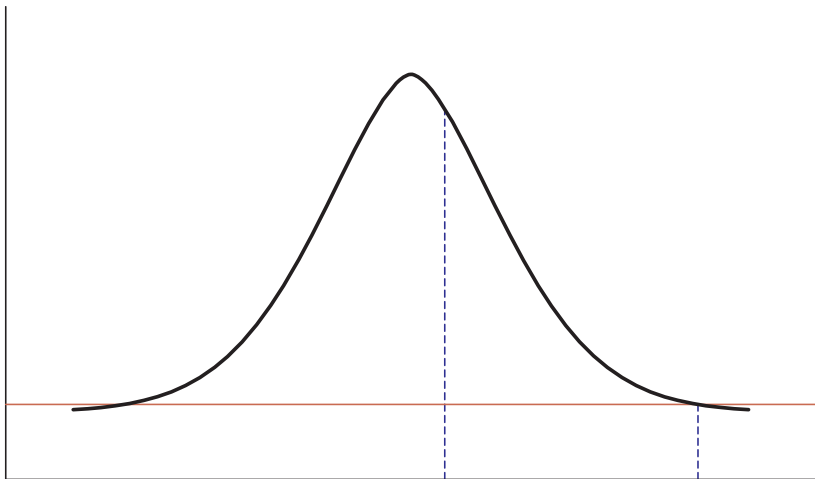




## VII.5. Concave and quasiconcave functions



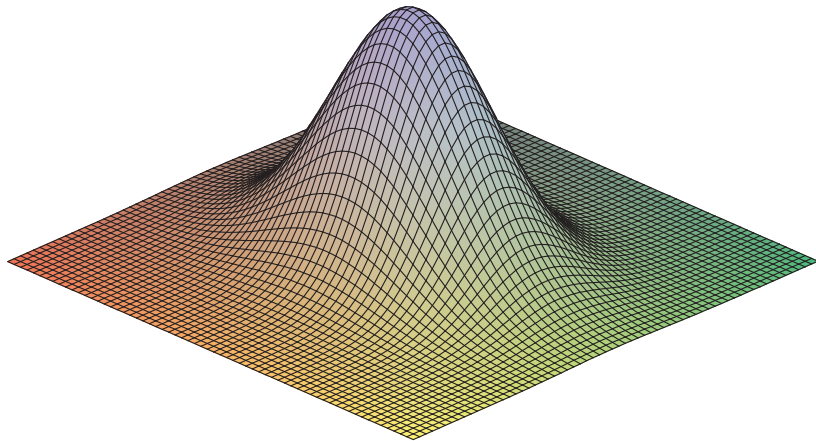
## VII.5. Concave and quasiconcave functions



## VII.5. Concave and quasiconcave functions



## VII.5. Concave and quasiconcave functions



### Remark

A function  $f$  is quasiconvex (strictly quasiconvex) if and only if the function  $-f$  is quasiconcave (strictly quasiconcave).

All the theorems in this section are formulated for quasiconcave and strictly quasiconcave functions. They have obvious analogies that hold for quasiconvex and strictly quasiconvex functions.

### Remark

- If a function  $f$  is strictly quasiconcave on  $M$ , then it is quasiconcave on  $M$ .

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- If a function  $f$  is strictly quasiconcave on  $M$ , then it is quasiconcave on  $M$ .
- Let  $f$  be a quasiconcave function on  $M$ . Then  $f$  is strictly quasiconcave on  $M$  if and only if the graph of  $f$  “does not contain a horizontal segment”, i.e.

$$\neg(\exists \mathbf{a}, \mathbf{b} \in M, \mathbf{a} \neq \mathbf{b}, \forall t \in [0, 1]: f(t\mathbf{a} + (1-t)\mathbf{b}) = f(\mathbf{a})).$$

### Remark

Let  $M \subset \mathbb{R}^n$  be a convex set and  $f$  a function defined on  $M$ .



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Let  $M \subset \mathbb{R}^n$  be a convex set and  $f$  a function defined on  $M$ .

- If  $f$  is concave on  $M$ , then  $f$  is quasiconcave on  $M$ .

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Let  $M \subset \mathbb{R}^n$  be a convex set and  $f$  a function defined on  $M$ .

- If  $f$  is concave on  $M$ , then  $f$  is quasiconcave on  $M$ .
- If  $f$  is strictly concave on  $M$ , then  $f$  is strictly quasiconcave on  $M$ .

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## Theorem 53 (characterization of quasiconcave functions using level sets)

*Let  $M \subset \mathbb{R}^n$  be a convex set and  $f$  a function defined on  $M$ . Then  $f$  is quasiconcave on  $M$  if and only if for each  $\alpha \in \mathbb{R}$  the set  $Q_\alpha = \{\mathbf{x} \in M; f(\mathbf{x}) \geq \alpha\}$  is convex.*

## Theorem 54 (a uniqueness of an extremum)

Let  $f$  be a *strictly* quasiconcave function on a convex set  $M \subset \mathbb{R}^n$ . Then there exists *at most one* point of maximum of  $f$ .

## Theorem 54 (a uniqueness of an extremum)

Let  $f$  be a *strictly* quasiconcave function on a convex set  $M \subset \mathbb{R}^n$ . Then there exists *at most one* point of maximum of  $f$ .

### Corollary

Let  $M \subset \mathbb{R}^n$  be a convex, closed, bounded and nonempty set and  $f$  a continuous and strictly quasiconcave function on  $M$ . Then  $f$  attains its maximum at exactly one point.

## Theorem 55 (sufficient condition for concave and convex functions in $\mathbb{R}^2$ )

Let  $G \subset \mathbb{R}^2$  be convex and  $f \in C^2(G)$ .

If  $\frac{\partial^2 f}{\partial x^2} \leq 0$ ,  $\frac{\partial^2 f}{\partial y^2} \leq 0$ , and  $\frac{\partial^2 f}{\partial x^2} \frac{\partial^2 f}{\partial y^2} - \left(\frac{\partial^2 f}{\partial x \partial y}\right)^2 \geq 0$  hold on  $G$ , then  $f$  is concave on  $G$ .

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Let  $f \in C^2(G)$ . Then the matrix

$$H_f(\mathbf{x}) = \begin{pmatrix} \frac{\partial^2 f}{\partial x_1^2}(\mathbf{x}) & \frac{\partial^2 f}{\partial x_1 \partial x_2}(\mathbf{x}) & \cdots & \frac{\partial^2 f}{\partial x_1 \partial x_n}(\mathbf{x}) \\ \frac{\partial^2 f}{\partial x_2 \partial x_1}(\mathbf{x}) & \frac{\partial^2 f}{\partial x_2^2}(\mathbf{x}) & \cdots & \frac{\partial^2 f}{\partial x_2 \partial x_n}(\mathbf{x}) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1}(\mathbf{x}) & \frac{\partial^2 f}{\partial x_n \partial x_2}(\mathbf{x}) & \cdots & \frac{\partial^2 f}{\partial x_n^2}(\mathbf{x}) \end{pmatrix}$$

is called **Hessian matrix** of  $f$ .

Let  $f \in C^2(G)$ . Then the matrix

$$H_f(x) = \begin{pmatrix} \frac{\partial^2 f}{\partial x_1^2}(x) & \frac{\partial^2 f}{\partial x_1 \partial x_2}(x) & \cdots & \frac{\partial^2 f}{\partial x_1 \partial x_n}(x) \\ \frac{\partial^2 f}{\partial x_2 \partial x_1}(x) & \frac{\partial^2 f}{\partial x_2^2}(x) & \cdots & \frac{\partial^2 f}{\partial x_2 \partial x_n}(x) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_n \partial x_1}(x) & \frac{\partial^2 f}{\partial x_n \partial x_2}(x) & \cdots & \frac{\partial^2 f}{\partial x_n^2}(x) \end{pmatrix}$$

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### Theorem 56

*Let  $G \subset \mathbb{R}^n$  be convex and  $f \in C^2(G)$ . If the Hessian matrix of  $f$  is positive semidefinite for every  $x \in G$ , then  $f$  is convex on  $G$ .*

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### Theorem 56

*Let  $G \subset \mathbb{R}^n$  be convex and  $f \in C^2(G)$ . If the Hessian matrix of  $f$  is positive semidefinite for every  $x \in G$ , then  $f$  is convex on  $G$ . If the Hessian matrix of  $f$  is positive definite for every  $x \in G$ , then  $f$  is strictly convex on  $G$ .*

# VIII. Antiderivatives and Riemann integral

# VIII.1. Antiderivatives

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## Definition

Let  $f$  be a function defined on an open interval  $I$ . We say that a function  $F: I \rightarrow \mathbb{R}$  is an **antiderivative of  $f$  on  $I$**  if for each  $x \in I$  the derivative  $F'(x)$  exists and  $F'(x) = f(x)$ .

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An antiderivative of  $f$  is sometimes called a function primitive to  $f$ .

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If  $F$  is an antiderivative of  $f$  on  $I$ , then  $F$  is continuous on  $I$ .



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An antiderivative of  $f$  is sometimes called a function primitive to  $f$ .

If  $F$  is an antiderivative of  $f$  on  $I$ , then  $F$  is continuous on  $I$ .

## Theorem 57 (Uniqueness of an antiderivative)

*Let  $F$  and  $G$  be antiderivatives of  $f$  on an open interval  $I$ . Then there exists  $c \in \mathbb{R}$  such that  $F(x) = G(x) + c$  for each  $x \in I$ .*

## Remark

The set of all antiderivatives of  $f$  on an open interval  $I$  is denoted by

$$\int f(x) dx.$$

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The fact that  $F$  is an antiderivative of  $f$  on  $I$  is expressed by

$$\int f(x) dx \stackrel{c}{=} F(x), \quad x \in I.$$

*Table of basic antiderivatives*

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- $\int x^n dx \stackrel{c}{=} \frac{x^{n+1}}{n+1}$  on  $\mathbb{R}$  for  $n \in \mathbb{N} \cup \{0\}$ ; on  $(-\infty, 0)$  and on  $(0, \infty)$  for  $n \in \mathbb{Z}, n < -1$ ,

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- $\int \frac{1}{\cos^2 x} dx \stackrel{c}{=} \operatorname{tg} x$  on each of the intervals  
 $(-\frac{\pi}{2} + k\pi, \frac{\pi}{2} + k\pi), k \in \mathbb{Z},$

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- $\int \frac{1}{\sqrt{1-x^2}} dx \stackrel{c}{=} \operatorname{arcsin} x$  on  $(-1, 1)$ ,

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- $\int \frac{1}{\sqrt{1-x^2}} dx \stackrel{c}{=} \operatorname{arcsin} x$  on  $(-1, 1)$ ,
- $\int -\frac{1}{\sqrt{1-x^2}} dx \stackrel{c}{=} \operatorname{arccos} x$  on  $(-1, 1)$ .

### Theorem 58 (Existence of an antiderivative)

*Let  $f$  be a continuous function on an open interval  $I$ . Then  $f$  has an antiderivative on  $I$ .*



## Theorem 59 (Linearity of antiderivatives)

*Suppose that  $f$  has an antiderivative  $F$  on an open interval  $I$ ,  $g$  has an antiderivative  $G$  on  $I$ , and let  $\alpha, \beta \in \mathbb{R}$ . Then the function  $\alpha F + \beta G$  is an antiderivative of  $\alpha f + \beta g$  on  $I$ .*

## Theorem 60 (substitution)

- (i) *Let  $F$  be an antiderivative of  $f$  on  $(a, b)$ . Let  $\varphi: (\alpha, \beta) \rightarrow (a, b)$  have a finite derivative at each point of  $(\alpha, \beta)$ . Then*

$$\int f(\varphi(x))\varphi'(x) dx \stackrel{c}{=} F(\varphi(x)) \quad \text{on } (\alpha, \beta).$$

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$$\int f(\varphi(x))\varphi'(x) dx \stackrel{c}{=} F(\varphi(x)) \quad \text{on } (\alpha, \beta).$$

- (ii) Let  $\varphi$  be a function with a finite derivative in each point of  $(\alpha, \beta)$  such that the derivative is either everywhere positive or everywhere negative, and such that  $\varphi((\alpha, \beta)) = (a, b)$ . Let  $f$  be a function defined on  $(a, b)$  and suppose that

$$\int f(\varphi(t))\varphi'(t) dt \stackrel{c}{=} G(t) \quad \text{on } (\alpha, \beta).$$

Then

$$\int f(x) dx \stackrel{c}{=} G(\varphi^{-1}(x)) \quad \text{on } (a, b).$$

## Theorem 61 (integration by parts)

*Let  $I$  be an open interval and let the functions  $f$  and  $g$  be continuous on  $I$ . Let  $F$  be an antiderivative of  $f$  on  $I$  and  $G$  an antiderivative of  $g$  on  $I$ . Then*

$$\int f(x)G(x) dx = F(x)G(x) - \int F(x)g(x) dx \quad \text{on } I.$$

### Example

Denote  $I_n = \int \frac{1}{(1+x^2)^n} dx$ ,  $n \in \mathbb{N}$ . Then

$$I_{n+1} = \frac{x}{2n(1+x^2)^n} + \frac{2n-1}{2n} I_n, \quad x \in \mathbb{R}, \quad n \in \mathbb{N},$$

$$I_1 \stackrel{c}{=} \arctg x, \quad x \in \mathbb{R}.$$

### Definition

A **rational function** is a ratio of two polynomials, where the polynomial in the denominator is not a zero polynomial.

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## Theorem (“fundamental theorem of algebra”)

Let  $n \in \mathbb{N}$ ,  $a_0, \dots, a_n \in \mathbb{C}$ ,  $a_n \neq 0$ . Then the equation

$$a_n z^n + a_{n-1} z^{n-1} + \dots + a_1 z + a_0 = 0$$

has at least one solution  $z \in \mathbb{C}$ .

## Lemma 62 (polynomial division)

*Let  $P$  and  $Q$  be polynomials (with complex coefficients) such that  $Q$  is not a zero polynomial. Then there are uniquely determined polynomials  $S$  and  $R$  satisfying:*

- $\deg R < \deg Q$ ,
- $P(x) = S(x)Q(x) + R(x)$  for all  $x \in \mathbb{C}$ .

*If  $P$  and  $Q$  have real coefficients then so have  $S$  and  $R$ .*



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## Corollary

*If  $P$  is a polynomials and  $\lambda \in \mathbb{C}$  its **root** (i.e.  $P(\lambda) = 0$ ), then there is a polynomial  $S$  satisfying  $P(x) = (x - \lambda)S(x)$  for all  $x \in \mathbb{C}$ .*

### Theorem 63 (factorisation into monomials)

*Let  $P(x) = a_n x^n + \dots + a_1 x + a_0$  be a polynomial of degree  $n \in \mathbb{N}$ . Then there are numbers  $x_1, \dots, x_n \in \mathbb{C}$  such that*

$$P(x) = a_n(x - x_1) \cdots (x - x_n), \quad x \in \mathbb{C}.$$

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## Definition

Let  $P$  be a polynomial that is not zero,  $\lambda \in \mathbb{C}$ , and  $k \in \mathbb{N}$ . We say that  $\lambda$  is a **root of multiplicity  $k$**  of the polynomial  $P$  if there is a polynomial  $S$  satisfying  $S(\lambda) \neq 0$  and  $P(x) = (x - \lambda)^k S(x)$  for all  $x \in \mathbb{C}$ .

## Theorem 64 (roots of a polynomial with real coefficients)

*Let  $P$  be a polynomial with real coefficients and  $\lambda \in \mathbb{C}$  a root of  $P$  of multiplicity  $k \in \mathbb{N}$ . Then the also the conjugate number  $\bar{\lambda}$  is a root of  $P$  of multiplicity  $k$ .*

## Theorem 65 (factorisation of a polynomial with real coefficients)

*Let  $P(x) = a_n x^n + \dots + a_1 x + a_0$  be a polynomial of degree  $n$  with real coefficients. Then there exist real numbers  $x_1, \dots, x_k, \alpha_1, \dots, \alpha_l, \beta_1, \dots, \beta_l$  and natural numbers  $p_1, \dots, p_k, q_1, \dots, q_l$  such that*

- $$P(x) = a_n (x - x_1)^{p_1} \dots (x - x_k)^{p_k} (x^2 + \alpha_1 x + \beta_1)^{q_1} \dots (x^2 + \alpha_l x + \beta_l)^{q_l},$$

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- no two polynomials from  $x - x_1, x - x_2, \dots, x - x_k, x^2 + \alpha_1 x + \beta_1, \dots, x^2 + \alpha_l x + \beta_l$  have a common root,

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- $P(x) = a_n (x - x_1)^{p_1} \cdots (x - x_k)^{p_k} (x^2 + \alpha_1 x + \beta_1)^{q_1} \cdots (x^2 + \alpha_l x + \beta_l)^{q_l}$ ,
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- the polynomials  $x^2 + \alpha_1 x + \beta_1, \dots, x^2 + \alpha_l x + \beta_l$  have no real root.

## Theorem 66 (decomposition to partial fractions)

Let  $P, Q$  be polynomials with real coefficients such that  $\deg P < \deg Q$  and let

$$Q(x) = a_n(x-x_1)^{p_1} \cdots (x-x_k)^{p_k} (x^2+\alpha_1x+\beta_1)^{q_1} \cdots (x^2+\alpha_lx+\beta_l)^{q_l}$$

be a factorisation of  $Q$  from Theorem 65. Then there exist unique real numbers  $A_1^1, \dots, A_{p_1}^1, \dots, A_1^k, \dots, A_{p_k}^k, B_1^1, C_1^1, \dots, B_{q_1}^1, C_{q_1}^1, \dots, B_1^l, C_1^l, \dots, B_{q_l}^l, C_{q_l}^l$  such that

$$\frac{P(x)}{Q(x)} =$$



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$$\begin{aligned} \frac{P(x)}{Q(x)} &= \frac{A_1^1}{(x-x_1)} + \cdots + \frac{A_{p_1}^1}{(x-x_1)^{p_1}} + \cdots + \frac{A_1^k}{(x-x_k)} + \cdots + \frac{A_{p_k}^k}{(x-x_k)^{p_k}} + \\ &+ \frac{B_1^1x+C_1^1}{(x^2+\alpha_1x+\beta_1)} + \cdots + \frac{B_{q_1}^1x+C_{q_1}^1}{(x^2+\alpha_1x+\beta_1)^{q_1}} + \cdots + \\ &+ \end{aligned}$$

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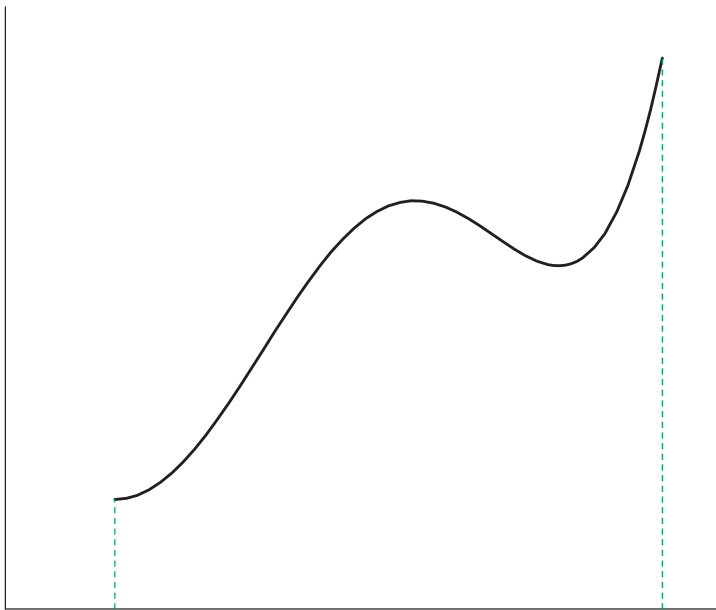
$$Q(x) = a_n(x-x_1)^{p_1} \cdots (x-x_k)^{p_k} (x^2+\alpha_1x+\beta_1)^{q_1} \cdots (x^2+\alpha_lx+\beta_l)^{q_l}$$

be a factorisation of  $Q$  from Theorem 65. Then there exist unique real numbers  $A_1^1, \dots, A_{p_1}^1, \dots, A_1^k, \dots, A_{p_k}^k, B_1^1, C_1^1, \dots, B_{q_1}^1, C_{q_1}^1, \dots, B_1^l, C_1^l, \dots, B_{q_l}^l, C_{q_l}^l$  such that

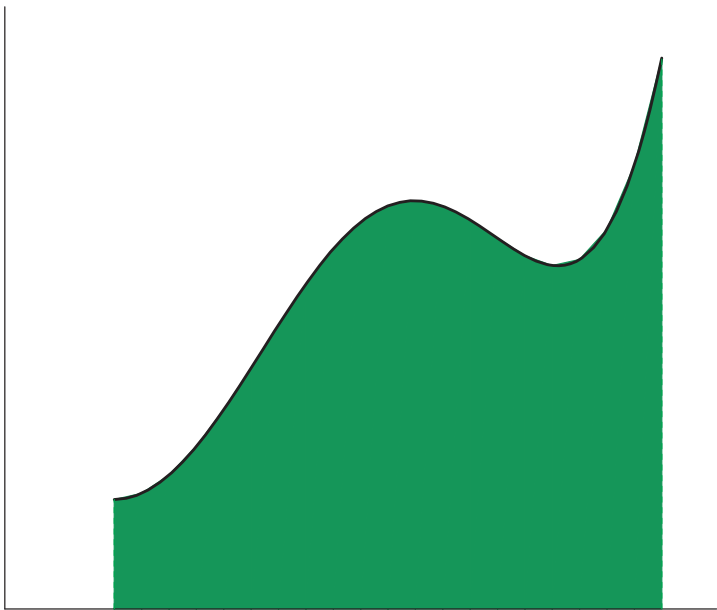
$$\begin{aligned} \frac{P(x)}{Q(x)} = & \frac{A_1^1}{(x-x_1)} + \cdots + \frac{A_{p_1}^1}{(x-x_1)^{p_1}} + \cdots + \frac{A_1^k}{(x-x_k)} + \cdots + \frac{A_{p_k}^k}{(x-x_k)^{p_k}} + \\ & + \frac{B_1^1x+C_1^1}{(x^2+\alpha_1x+\beta_1)} + \cdots + \frac{B_{q_1}^1x+C_{q_1}^1}{(x^2+\alpha_1x+\beta_1)^{q_1}} + \cdots + \\ & + \frac{B_1^lx+C_1^l}{(x^2+\alpha_lx+\beta_l)} + \cdots + \frac{B_{q_l}^lx+C_{q_l}^l}{(x^2+\alpha_lx+\beta_l)^{q_l}}, \quad x \in \mathbb{R} \setminus \{x_1, \dots, x_k\}. \end{aligned}$$

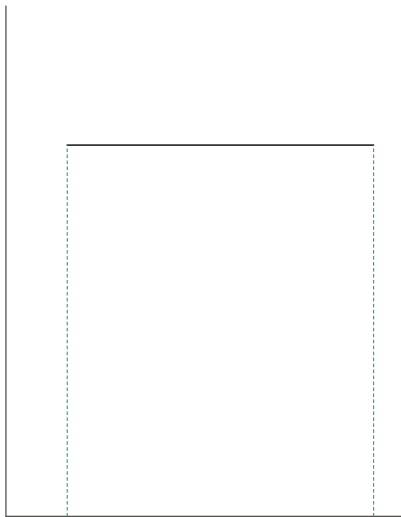
# VIII.2. Riemann integral

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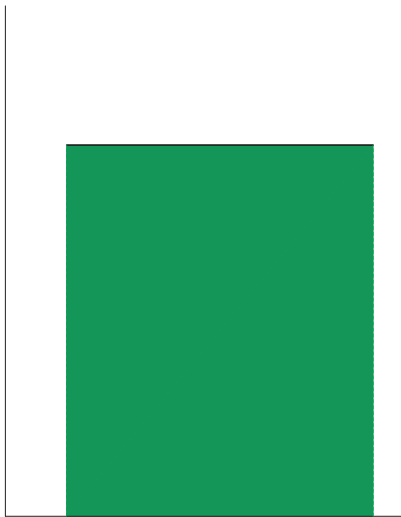


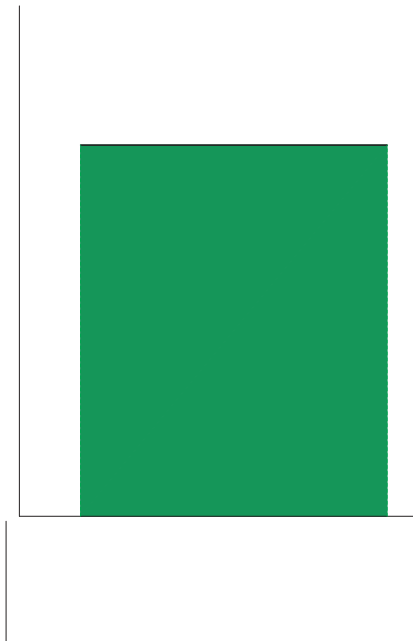
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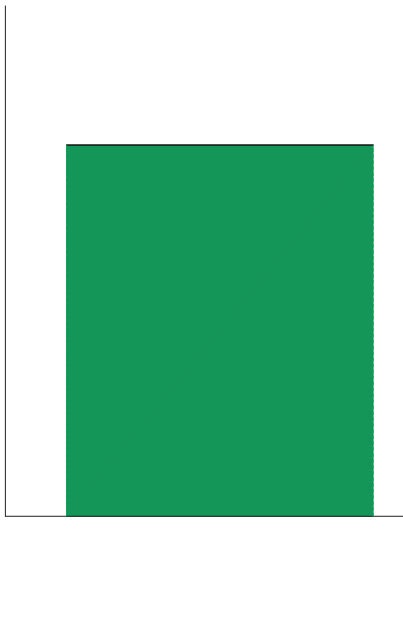




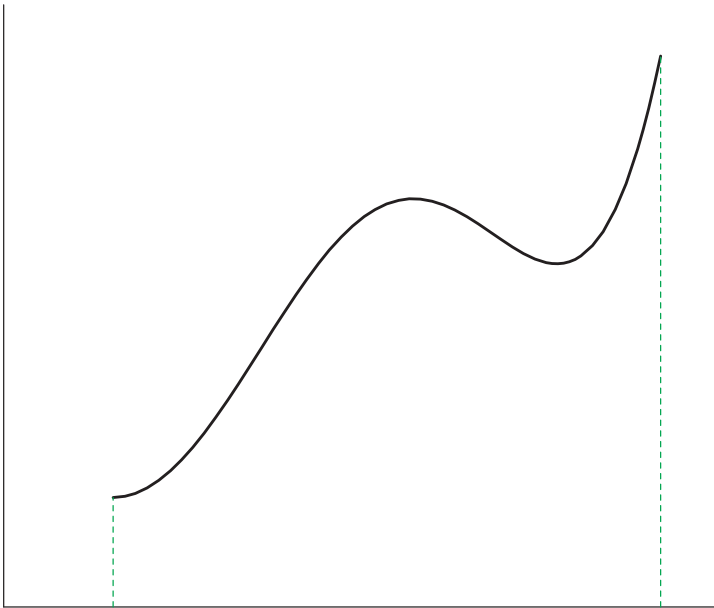




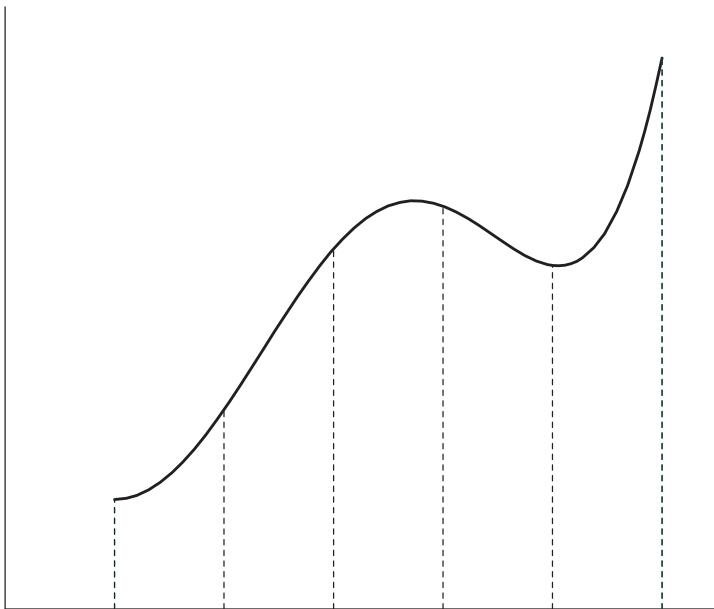




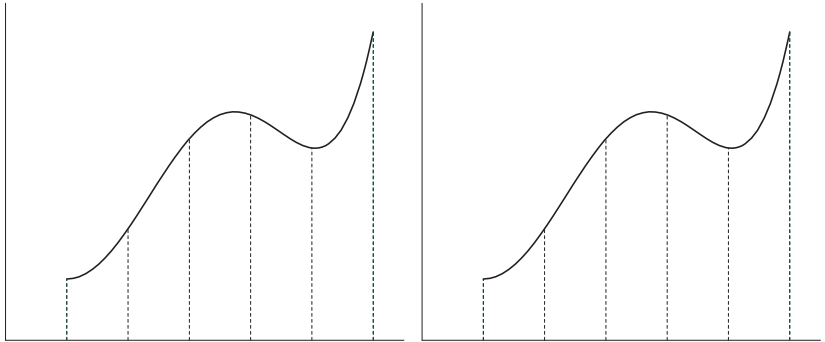
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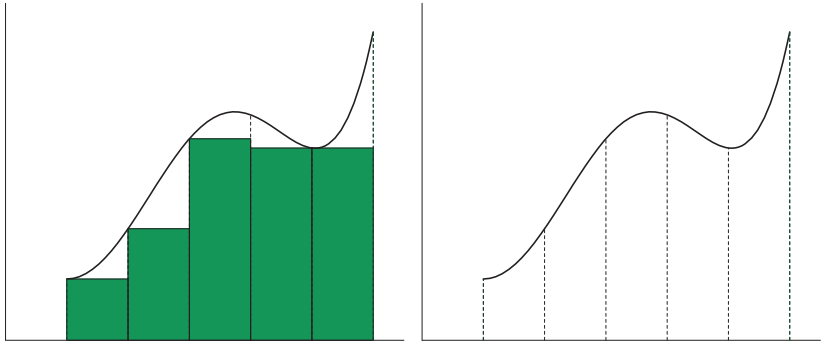
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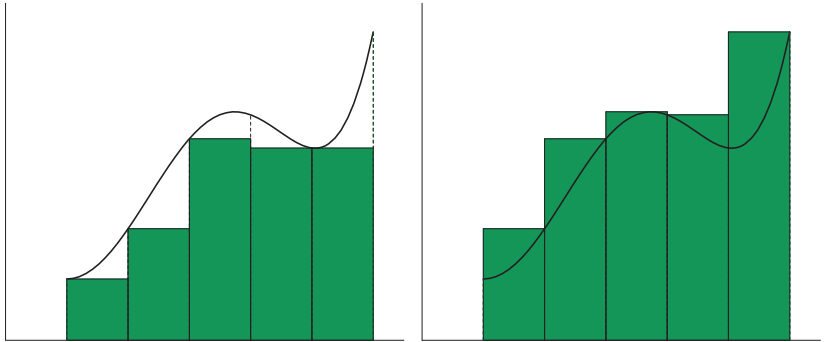
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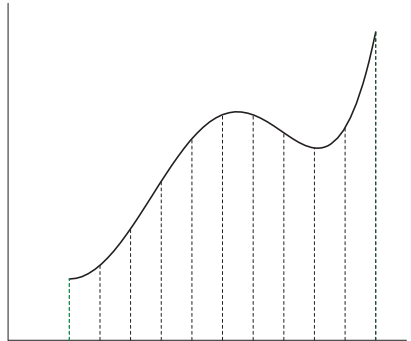
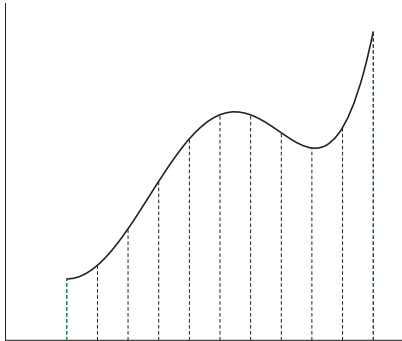


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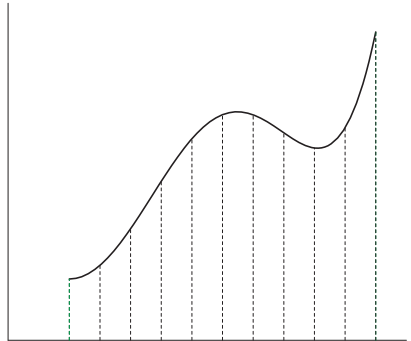
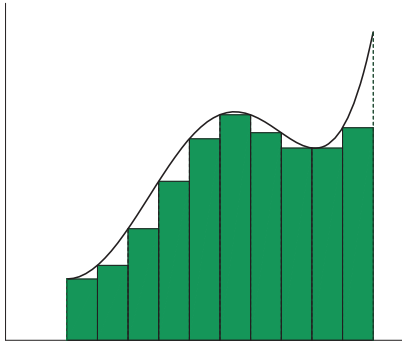




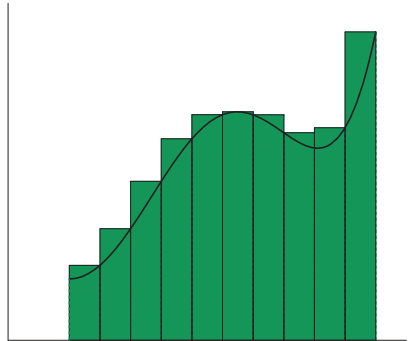
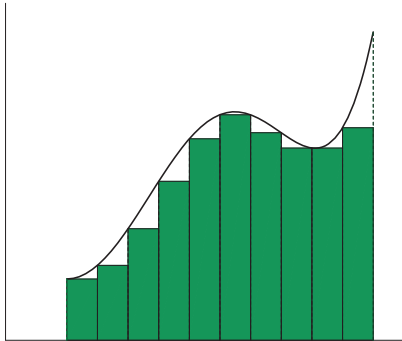
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## Definition

A finite sequence  $\{x_j\}_{j=0}^n$  is called a **partition of the interval**  $[a, b]$  if

$$a = x_0 < x_1 < \cdots < x_n = b.$$

The points  $x_0, \dots, x_n$  are called the **partition points**.

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The points  $x_0, \dots, x_n$  are called the **partition points**. We say that a partition  $D'$  of an interval  $[a, b]$  is a **refinement of the partition**  $D$  of  $[a, b]$  if each partition point of  $D$  is also a partition point of  $D'$ .

## Definition

Suppose that  $a, b \in \mathbb{R}$ ,  $a < b$ , the function  $f$  is bounded on  $[a, b]$ , and  $D = \{x_j\}_{j=0}^n$  is a partition of  $[a, b]$ . Denote

$$\overline{S}(f, D) = \sum_{j=1}^n M_j(x_j - x_{j-1}), \text{ where } M_j = \sup\{f(x); x \in [x_{j-1}, x_j]\},$$

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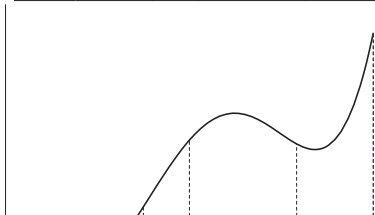
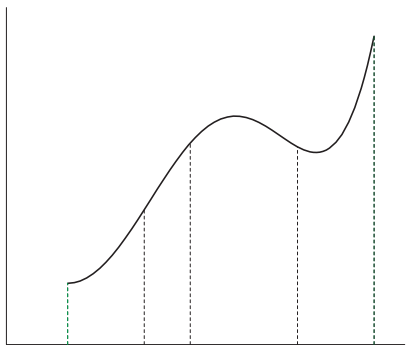
$\int_a^b f = -\int_b^a f$ , and in case that  $a = b$  we put  $\int_a^b f = 0$ .

### Remark

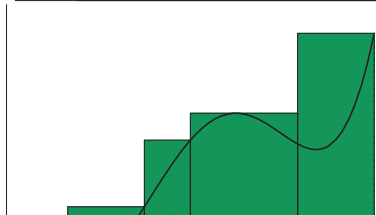
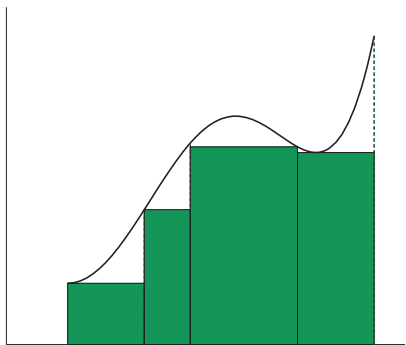
Let  $D, D'$  be partitions of  $[a, b]$ ,  $D'$  refines  $D$ , and let  $f$  be a bounded function on  $[a, b]$ . Then

$$\underline{S}(f, D) \leq \underline{S}(f, D') \leq \overline{S}(f, D') \leq \overline{S}(f, D).$$

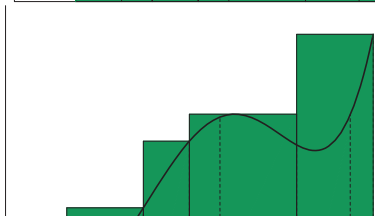
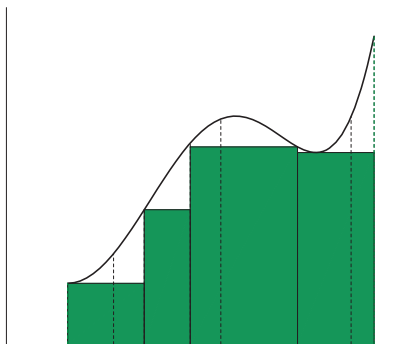
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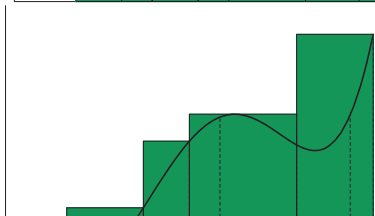
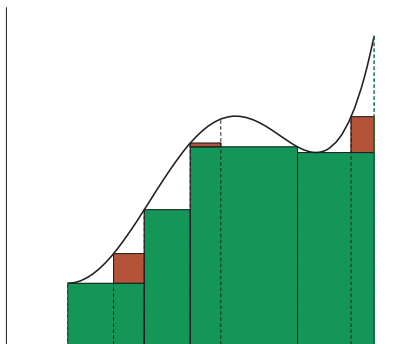


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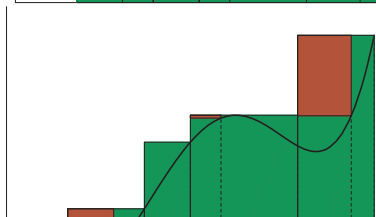
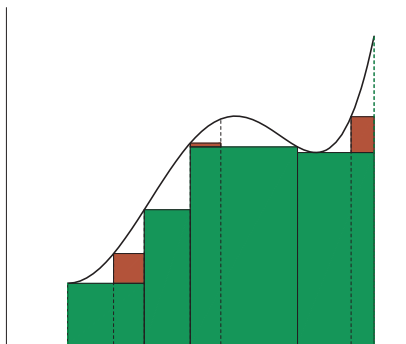




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Suppose that  $D_1, D_2$  are partitions of  $[a, b]$  and a partition  $D'$  refines both  $D_1$  and  $D_2$ . Then

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It easily follows that  $\underline{\int_a^b} f \leq \overline{\int_a^b} f$ .

## Theorem 67

- (i) *Suppose that  $f$  has the Riemann integral over  $[a, b]$  and let  $[c, d] \subset [a, b]$ . Then  $f$  has the Riemann integral also over  $[c, d]$ .*

## Theorem 67

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- (ii) *Suppose that  $c \in (a, b)$  and  $f$  has the Riemann integral over the intervals  $[a, c]$  and  $[c, b]$ . Then  $f$  has the Riemann integral over  $[a, b]$  and*

$$\int_a^b f = \int_a^c f + \int_c^b f. \quad (1)$$

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$$\int_a^b f = \int_a^c f + \int_c^b f. \quad (1)$$

## Remark

The formula (1) holds for all  $a, b, c \in \mathbb{R}$  if the integral of  $f$  exists over the interval  $[\min\{a, b, c\}, \max\{a, b, c\}]$ .



## Theorem 68 (linearity of the Riemann integral)

Let  $f$  and  $g$  be functions with Riemann integral over  $[a, b]$  and let  $\alpha \in \mathbb{R}$ . Then

- (i) the function  $\alpha f$  has the Riemann integral over  $[a, b]$  and

$$\int_a^b \alpha f = \alpha \int_a^b f,$$

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- (i) the function  $\alpha f$  has the Riemann integral over  $[a, b]$  and

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- (ii) the function  $f + g$  has the Riemann integral over  $[a, b]$  and

$$\int_a^b f + g = \int_a^b f + \int_a^b g.$$

## Theorem 69

Let  $a, b \in \mathbb{R}$ ,  $a < b$ , and let  $f$  and  $g$  be functions with Riemann integral over  $[a, b]$ . Then:

- (i) If  $f(x) \leq g(x)$  for each  $x \in [a, b]$ , then

$$\int_a^b f \leq \int_a^b g.$$

## Theorem 69

Let  $a, b \in \mathbb{R}$ ,  $a < b$ , and let  $f$  and  $g$  be functions with Riemann integral over  $[a, b]$ . Then:

(i) If  $f(x) \leq g(x)$  for each  $x \in [a, b]$ , then

$$\int_a^b f \leq \int_a^b g.$$

(ii) The function  $|f|$  has the Riemann integral over  $[a, b]$  and

$$\left| \int_a^b f \right| \leq \int_a^b |f|.$$

### Theorem 70

*Let  $f$  be a function continuous on an interval  $[a, b]$ ,  $a, b \in \mathbb{R}$ . Then  $f$  has the Riemann integral on  $[a, b]$ .*

## Theorem 71

*Let  $f$  be a function continuous on an interval  $(a, b)$  and let  $c \in (a, b)$ . If we denote  $F(x) = \int_c^x f(t) dt$  for  $x \in (a, b)$ , then  $F'(x) = f(x)$  for each  $x \in (a, b)$ . In other words,  $F$  is an antiderivative of  $f$  on  $(a, b)$ .*

## Theorem 72 (Newton-Leibniz formula)

*Let  $f$  be a function continuous on an interval  $(a - \varepsilon, b + \varepsilon)$ ,  $a, b \in \mathbb{R}$ ,  $a < b$ ,  $\varepsilon > 0$  and let  $F$  be an antiderivative of  $f$  on  $(a - \varepsilon, b + \varepsilon)$ . Then*

$$\int_a^b f(x) \, dx = F(b) - F(a). \quad (2)$$

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### Remark

The Newton-Leibniz formula (2) holds even if  $b < a$  (if  $F' = f$  on  $(b - \varepsilon, a + \varepsilon)$ ).



## Theorem 72 (Newton-Leibniz formula)

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### Remark

The Newton-Leibniz formula (2) holds even if  $b < a$  (if  $F' = f$  on  $(b - \varepsilon, a + \varepsilon)$ ). Let us denote

$$[F]_a^b = F(b) - F(a).$$

## Theorem 73 (integration by parts)

*Suppose that the functions  $f$ ,  $g$ ,  $f'$  and  $g'$  are continuous on an interval  $[a, b]$ . Then*

$$\int_a^b f'g = [fg]_a^b - \int_a^b fg'.$$

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Suppose that the functions  $f$ ,  $g$ ,  $f'$  and  $g'$  are continuous on an interval  $[a, b]$ . Then

$$\int_a^b f'g = [fg]_a^b - \int_a^b fg'.$$

## Theorem 74 (substitution)

Let the function  $f$  be continuous on an interval  $[a, b]$ . Suppose that the function  $\varphi$  has a continuous derivative on  $[\alpha, \beta]$  and  $\varphi$  maps  $[\alpha, \beta]$  into the interval  $[a, b]$ . Then

$$\int_{\alpha}^{\beta} f(\varphi(x))\varphi'(x) dx = \int_{\varphi(\alpha)}^{\varphi(\beta)} f(t) dt.$$