

# Matematiske metoder (MM 529)

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Functions of  $n$  variables:  $f : D \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $z = f(x_1, \dots, x_n)$ .

Difference quotient for variable  $x_i$  ( $1 \leq i \leq n$ ):

$$\frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h}.$$

Partial derivatives:

$$\begin{aligned} f_{x_i}(x_1, \dots, x_n) &= \frac{\partial f}{\partial x_i}(x_1, \dots, x_n) = \\ &= \lim_{h \rightarrow 0} \frac{f(x_1, \dots, x_i + h, \dots, x_n) - f(x_1, \dots, x_i, \dots, x_n)}{h}. \end{aligned}$$

Note: For a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  the partial derivative w.r.t.  $x_i$  at  $(a_1, \dots, a_n)$  is the derivative at  $a_i$  of the function  $g : \mathbb{R} \rightarrow \mathbb{R}$   
 $g(x_i) = f(a_1, \dots, a_{i-1}, x_i, a_{i+1}, \dots, a_n)$ .

Partial derivatives w.r.t.  $x_i$ : Treat all variables except  $x_i$  as constants and form the derivative w.r.t. the variable  $x_i$ .

Vectors in  $\mathbb{R}^n$ : A vector  $\mathbf{v}$  is an object having a **length** and a **direction** (representable by an arrow).

Two vectors with the same length and same direction are considered **equal** (no matter where they are located in  $\mathbb{R}^n$ ).

Ordered pair  $(P, Q)$  of two points in  $\mathbb{R}^n$  determines a vector  $\vec{PQ} = \mathbf{v}$  whose tail is  $P$  and whose head is  $Q$ .

The **length**  $|\mathbf{v}|$  (or  $\|\mathbf{v}\|$ ) of  $\mathbf{v}$  is the Euclidean distance between  $P$  and  $Q$ .

Example:

$\mathbf{v}$  vector from  $(1, 1)$  to  $(2, 3)$  in  $\mathbb{R}^2$ .

$\mathbf{v}$  is also the vector from  $(2, 3)$  to  $(3, 5)$  and from  $(0, 0)$  to  $(1, 2)$ .

The length  $|\mathbf{v}| = \sqrt{(2-1)^2 + (3-1)^2} = \sqrt{5}$ .

Representation of a vector: Coordinates  $(a_1, a_2, \dots, a_n)$  of the head, if the tail is  $(0, 0, \dots, 0)$ .

**Scalar multiples** of  $\mathbf{v} = (a_1, a_2, \dots, a_n)$ , multiplying  $\mathbf{v}$  with the scalar  $t \in \mathbb{R}$ :

$$t \cdot \mathbf{v} = (t \cdot a_1, t \cdot a_2, \dots, t \cdot a_n)$$

Length  $|t \cdot \mathbf{v}| = |t| \cdot |\mathbf{v}|$ .

$(-1) \cdot \mathbf{v}$  is the vector with the same length as  $\mathbf{v}$  but the opposite direction.

**Unit vector**: Vector of length 1.

$\frac{\mathbf{v}}{|\mathbf{v}|} = \frac{1}{|\mathbf{v}|} \cdot \mathbf{v}$  is the unit vector in the direction of  $\mathbf{v}$ .

$\mathbf{e}_i$ : Unit vector in the direction of the  $i$ th coordinate axis.

**Addition of vectors**  $\mathbf{v} = (a_1, a_2, \dots, a_n)$  and  $\mathbf{w} = (b_1, b_2, \dots, b_n)$ :

$$\mathbf{v} + \mathbf{w} = (a_1 + b_1, a_2 + b_2, \dots, a_n + b_n).$$

**Subtraction**:  $\mathbf{v} - \mathbf{w} = \mathbf{v} + (-1) \cdot \mathbf{w}$ .

**Dot product** (also called scalar product) of a vector

$\mathbf{v} = (a_1, a_2, \dots, a_n)$  with a vector  $\mathbf{w} = (b_1, b_2, \dots, b_n)$ :

$$\mathbf{v} \cdot \mathbf{w} = a_1 \cdot b_1 + a_2 \cdot b_2 + \dots + a_n \cdot b_n = \sum_{i=1}^n a_i \cdot b_i \in \mathbb{R}.$$

Properties:

$\mathbf{v} \cdot \mathbf{w} = |\mathbf{v}| \cdot |\mathbf{w}| \cdot \cos \theta$  where  $\theta$  is the angle between the vectors  $\mathbf{v}$  and  $\mathbf{w}$ .

Rules:

$\mathbf{v} \cdot \mathbf{w} = \mathbf{w} \cdot \mathbf{v}$  (commutative law)

$\mathbf{u} \cdot (\mathbf{v} + \mathbf{w}) = \mathbf{u} \cdot \mathbf{v} + \mathbf{u} \cdot \mathbf{w}$  (distributive law)

$t \cdot (\mathbf{v} \cdot \mathbf{w}) = (t \cdot \mathbf{v}) \cdot \mathbf{w} = \mathbf{v} \cdot (t \cdot \mathbf{w})$  if  $t \in \mathbb{R}$

$\mathbf{v} \cdot \mathbf{v} = |\mathbf{v}|^2$ .

If  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is partially differentiable in the point  $\mathbf{a} = (a_1, a_2, \dots, a_n)$  then the **gradient** is the tuple (vector)

$$\nabla f(\mathbf{a}) = \left( \frac{\partial}{\partial x_1} f(\mathbf{a}), \frac{\partial}{\partial x_2} f(\mathbf{a}), \dots, \frac{\partial}{\partial x_n} f(\mathbf{a}) \right) \in \mathbb{R}^n.$$

$n = 2$ : Gradient at  $(a, b) \in \mathbb{R}^2$  is

$$\nabla f(a, b) = (f_x(a, b), f_y(a, b)).$$

More general:  $f : D \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$  partially differentiable on  $D$  then

$$\nabla f(x, y) = (f_x(x, y), f_y(x, y)).$$

Example:  $f(x, y) = x^2 - y^2$  partially differentiable on  $\mathbb{R}^2$ .

$$\nabla f(x, y) = (f_x(x, y), f_y(x, y)) = (2x, -2y).$$

$$\nabla f(-1, 3) = (-2, -6).$$

$f : D \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$  continuous and partially differentiable w.r.t.  $x$  and  $y$  and  $f_x$  and  $f_y$  continuous in  $(a, b) \in \mathbb{R}^2$ .

Equation of the tangent plane at  $(a, b) \in \mathbb{R}^2$ :

$$z = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b).$$

$f : D \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  partially differentiable w.r.t.  $x_i$  and  $f_{x_i}$  continuous for all  $i$  with  $1 \leq i \leq n$  in  $\mathbf{a} = (a_1, \dots, a_n) \in \mathbb{R}^n$ .

Equation of the tangent space (hyperplane) at  $\mathbf{a} \in \mathbb{R}^n$ :

$$z = f(\mathbf{a}) + f_{x_1}(\mathbf{a})(x_1 - a_1) + \dots + f_{x_n}(\mathbf{a})(x_n - a_n)$$

$$z = f(a_1, \dots, a_n) + \sum_{i=1}^n f_{x_i}(\mathbf{a})(x_i - a_i).$$

Sum is the dot product of vectors:

$$z = f(a_1, \dots, a_n) + \nabla f(\mathbf{a}) \cdot (x_1 - a_1, x_2 - a_2, \dots, x_n - a_n).$$

Linear approximation for functions  $f : \mathbb{R} \rightarrow \mathbb{R}$ : Approximate differentiable function  $f$  near  $a \in \mathbb{R}$  by tangent line through point  $(a, f(a))$ .

Linear approximation:  $f(x) \approx P(x) = f(a) + f'(a)(x - a)$  for  $x$  near  $a$ .

Linear approximation of functions  $f : D \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ : Approximate  $f(\mathbf{x})$  for  $\mathbf{x} = (x_1, x_2, \dots, x_n)$  near  $\mathbf{a} = (a_1, a_2, \dots, a_n) \in D$  by the corresponding point on the tangent space of  $f$  in  $\mathbf{a}$ .

Equation of the tangent plane at  $(a, b) \in D$ :

$$z = L(x, y) = f(a, b) + f_x(a, b)(x - a) + f_y(a, b)(y - b).$$

Example:  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $f(x, y) = x^2 - y^2$ ,  $(a, b) = (2, 1)$ .

Partial derivatives  $f_x(x, y) = 2x$  and  $f_y(x, y) = -2y$ ,

Tangent plane in  $(2, 1)$  consists of all points  $(x, y, z)$  satisfying the equation

$$\begin{aligned} z = L(x, y) &= f(2, 1) + f_x(2, 1)(x - 2) + f_y(2, 1)(y - 1) \\ &= 4 - 1 + 4(x - 2) - 2(y - 1) = 4x - 2y - 3. \end{aligned}$$



## Linear approximation (example)

Example:  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $f(x, y) = x^2 - y^2$ ,  $(a, b) = (2, 1)$ .

Tangent plane in  $(2, 1)$  consists of all points  $(x, y, z)$  satisfying the equation

$$\begin{aligned} z = L(x, y) &= f(2, 1) + f_x(2, 1)(x - 2) + f_y(2, 1)(y - 1) \\ &= 4 - 1 + 4(x - 2) - 2(y - 1) = 4x - 2y - 3. \end{aligned}$$

Linear approximation of  $f$  at  $(2.2, 0.9)$  near  $(2, 1)$ :

$$4.03 = 4.84 - 0.81 = f(2.2, 0.9) \approx L(2.2, 0.9) = 8.8 - 1.8 - 3 = 4.$$

The general case:

$f : D \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$  partially differentiable w.r.t.  $x_i$  and  $f_{x_i}$  continuous for all  $i$  with  $1 \leq i \leq n$  in  $(a_1, \dots, a_n) \in D$  ( $D$  open subset of  $\mathbb{R}^n$ ).

Linear approximation of  $f$  near  $\mathbf{a} = (a_1, \dots, a_n)$ :

$f(x_1, \dots, x_n) \approx L(x_1, \dots, x_n)$ , where

$$\begin{aligned} L(x_1, \dots, x_n) &= f(\mathbf{a}) + f_{x_1}(\mathbf{a})(x_1 - a_1) + \dots + f_{x_n}(\mathbf{a})(x_n - a_n) \\ &= f(\mathbf{a}) + \nabla f(\mathbf{a}) \cdot (x_1 - a_1, x_2 - a_2, \dots, x_n - a_n). \end{aligned}$$

Unit vector in the direction of  $\mathbf{v} \neq (0, 0, \dots, 0)$ :

$$\mathbf{v}_0 = \frac{1}{|\mathbf{v}|} \cdot \mathbf{v}.$$

**Directional derivative** in the direction of  $\mathbf{v} \neq (0, 0, \dots, 0)$  at  $\mathbf{a} = (a_1, \dots, a_n) \in \mathbb{R}^n$ :

$$D_{\mathbf{v}}f(\mathbf{a}) = \lim_{h \rightarrow 0+} \frac{f(\mathbf{a} + h\mathbf{v}_0) - f(\mathbf{a})}{h},$$

if the limit exists.

Case  $n = 2$ :  $\mathbf{a} = (a, b) \in \mathbb{R}^2$ ,  $\mathbf{v}_0 = (u, w) \in \mathbb{R}^2$ .

Directional derivative in direction of  $\mathbf{v}$  and  $\mathbf{a}$ :

$$D_{\mathbf{v}}f(a, b) = \lim_{h \rightarrow 0+} \frac{f(a + hu, b + hw) - f(a, b)}{h}.$$

## Directional derivatives and partial derivatives

Case  $n = 2$ :  $\mathbf{a} = (a, b) \in \mathbb{R}^2$ ,  $\mathbf{v}_0 = (u, w) \in \mathbb{R}^2$ .

Directional derivative in direction of  $\mathbf{v}$  at  $\mathbf{a}$ :

$$D_{\mathbf{v}}f(a, b) = \lim_{h \rightarrow 0+} \frac{f(a + hu, b + hw) - f(a, b)}{h}.$$

If  $\mathbf{v}_0 = \mathbf{e}_1 = (1, 0)$  (i.e.  $\mathbf{v}$  parallel to the  $x$ -axis) then

$$D_{\mathbf{v}}f(a, b) = \lim_{h \rightarrow 0+} \frac{f(a + h, b) - f(a, b)}{h} = f_x(a, b).$$

If  $\mathbf{v}_0 = \mathbf{e}_2 = (0, 1)$  (i.e.  $\mathbf{v}$  parallel to the  $y$ -axis) then

$$D_{\mathbf{v}}f(a, b) = \lim_{h \rightarrow 0+} \frac{f(a, b + h) - f(a, b)}{h} = f_y(a, b).$$

In general (directional derivatives in direction of a coordinate axis):

$$D_{\mathbf{e}_i}f(\mathbf{a}) = \lim_{h \rightarrow 0+} \frac{f(\mathbf{a} + h\mathbf{e}_i) - f(\mathbf{a})}{h} = f_{x_i}(\mathbf{a}).$$

The directional derivative in  $\mathbf{a}$  need not exist if all partial (first order) derivatives exist (the function need not be continuous in  $\mathbf{a}$ ). If  $D \subseteq \mathbb{R}^n$  is open, and all (first order) partial derivatives of  $f : D \rightarrow \mathbb{R}$  exist and are continuous, then the directional derivative exists for every vector  $\mathbf{v} \neq (0, \dots, 0)$ .

In that case

$$D_{\mathbf{v}}f(\mathbf{a}) = \lim_{h \rightarrow 0+} \frac{f(\mathbf{a} + h\mathbf{v}_0) - f(\mathbf{a})}{h} = \mathbf{v}_0 \cdot \nabla f(\mathbf{a}).$$

Example:  $f(x, y) = x^2 - y^2$ , partial derivatives  $f_x(x, y) = 2x$ ,  $f_y(x, y) = -2y$  are continuous. Therefore:

$$D_{(1,1)}f(2, 1) = \mathbf{v}_0 \cdot \nabla f(\mathbf{a}) = \left( \frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}} \right) \cdot (4, -2) = \sqrt{2}.$$

Indicates the slope of  $f$  in  $(2, 1)$  in direction of the angle bisector.