

Next: Gradient.

For  $f: U \rightarrow \mathbb{R}$  ( $U \subset \mathbb{R}^n$  open) differentiable, we have  $(Df)_i = (\nabla f)_i$ , where

$\nabla f = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{pmatrix}$  is called the gradient of  $f$ , or "nabla  $f$ ".

Note: Often we write  $\nabla = \begin{pmatrix} \frac{\partial}{\partial x_1} \\ \vdots \\ \frac{\partial}{\partial x_n} \end{pmatrix}$ , a differential operator.

Two results:

- If  $(\nabla f)(x) \neq 0$ , then  $f$  has greatest directional derivative in direction  $\frac{\nabla f(x)}{\|\nabla f(x)\|}$ .

( $D_u f = Df \cdot u = \nabla f \cdot u = \underbrace{\|\nabla f\|}_{=1} \|u\| \cos \varphi$  is maximal for  $\varphi = 0$ . )  
↑ angle between  $\nabla f$  and  $u$

- If  $f$  has a local extremum at  $x$ , then  $\nabla f(x) = 0$ .

(If  $\nabla f(x) \neq 0$ , then  $f$  increases in at least one direction and decreases in the opposite direction (from 1-dim. calculus), thus it cannot have a local extremum.)

## 2.2 Higher Order Derivatives

We showed:  $f: U \rightarrow \mathbb{R}^m$  ( $U \subset \mathbb{R}^n$  open) continuously differentiable.

$\Leftrightarrow$

All  $\frac{\partial f_i}{\partial x_j}$  exist and are continuous.

We call this class of functions  $C^1$ , or  $C^1(U)$ .

Second partial derivatives are defined as  $\frac{\partial}{\partial x_i} \frac{\partial f_e}{\partial x_j} = \frac{\partial^2 f_e}{\partial x_i \partial x_j}$ .

We say  $f$  is of class  $C^k$  (or  $C^k(U)$ ) if all  $k$ -th partial derivatives exist in all components and are continuous.

Generally,  $\frac{\partial f_e}{\partial x_i \partial x_j} \neq \frac{\partial f_e}{\partial x_j \partial x_i}$  is possible, see homework.

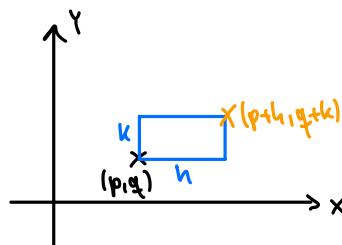
But, we have:

Theorem (Clairaut's thm., or Schwarz's thm.):

If  $f: U \rightarrow \mathbb{R}^m$  ( $U \subset \mathbb{R}^n$  open) is of class  $C^2$ , then  $\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i} \quad \forall i, j.$

This implies the more general result: If  $f$  is of class  $C^k$ , then all partial derivatives up to order  $k$  commute.  
can be interchanged

Sketch of proof:



We apply the mean-value thm. twice:  $\exists$  point  $(x, y)$  in  $\boxed{\quad}$  s.t.

$$\cdot \frac{f(p+h, q) - f(p, q)}{h} = \frac{\partial f}{\partial x}(x, y)$$

$$\cdot \frac{1}{k} \left[ \frac{f(p+h, q+k) - f(p, q+k)}{h} - \left( \frac{f(p+h, q) - f(p, q)}{h} \right) \right] = \frac{\partial}{\partial y} \frac{\partial f}{\partial x}(x, y)$$

$$\Rightarrow \frac{f(p+h, q+k) - f(p+h, q) - f(p, q+k) + f(p, q)}{hk} = \frac{\partial^2 f}{\partial x \partial y}(x, y)$$

$$\text{LHS} \rightarrow \frac{\frac{\partial f}{\partial y}(p+h) - \frac{\partial f}{\partial y}(p)}{h} \rightarrow \frac{\partial}{\partial x} \frac{\partial f}{\partial y}(p, q), \text{ RHS} \rightarrow \frac{\partial^2 f}{\partial x \partial y}(p, q) \text{ by continuity. } \square$$

(A more detailed proof is, e.g., in Kantorovitz: Theorem 2.2.2, or in Rudin: Theorem 9.41.)

Note: The matrix  $H$  with  $(H_f(x))_{ij} := \frac{\partial^2 f(x)}{\partial x_i \partial x_j}$  is called Hessian matrix of  $f$ .

Due to Schwarz,  $H_f$  is symmetric (for  $f \in C^2$ ) i.e.,  $(H_f)_{ij} = (H_f)_{ji}$ .