

# Tutorial 9 & Ex 8

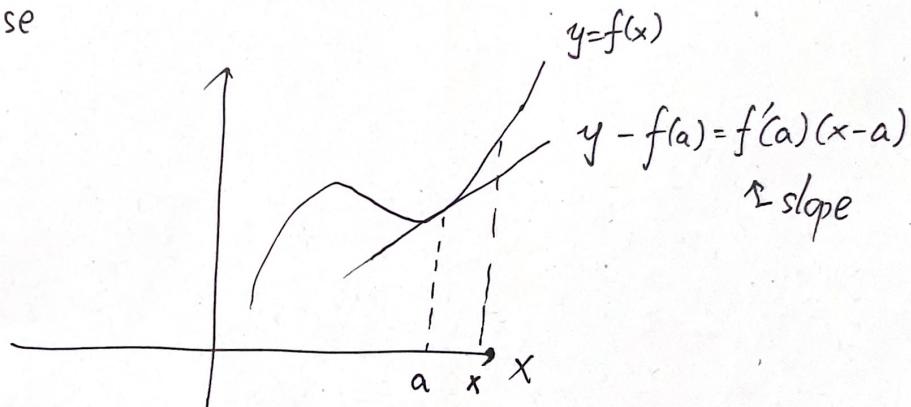
Today we work on differentials.

Questions for Ex 8?

~~1. 1-dim case~~

Slogan: differential  $\approx$  approximate fcts by linear fcts.

1. 1-dim case



$$f'(a) = \lim_{x \rightarrow a} \frac{f(x) - f(a)}{x - a} = \lim_{t \rightarrow 0} \frac{f(a+t) - f(a)}{t}$$

Ex. For  $f(x) = |x|$ , compute  $f'(x)$

Ex. For  $f(x) = \begin{cases} x \sin \frac{1}{x} & x \neq 0 \\ 0 & x = 0 \end{cases}$  compute  $f'(x)$ .

Ex. For  $f: [a, b] \rightarrow \mathbb{R}$  cont & differentiable in  $(a, b)$ ,

if  $f'(x) \equiv 0$ , then  $f(x) \equiv C$ .

Recall: mean value theorem

(Mittelwertsatz der Differentialrechnung)

We have two generalizations of 1-dim differential:

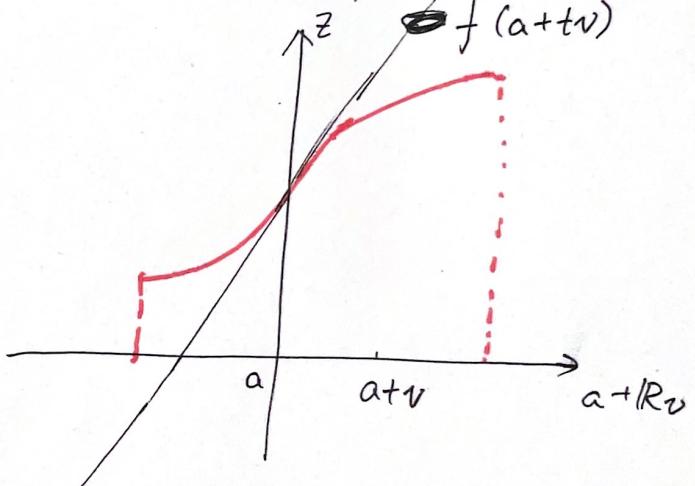
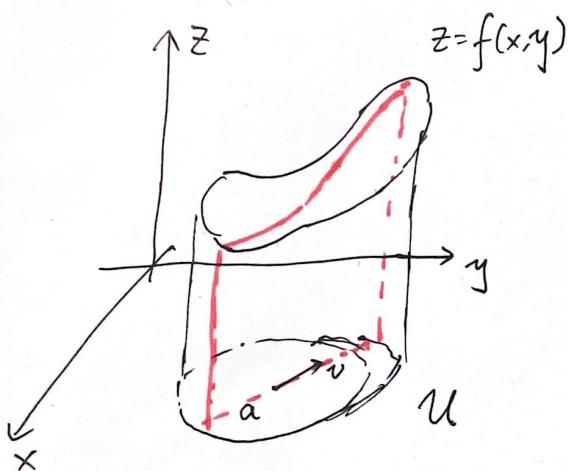
## 2. Directional derivative (reduced to 1-dim)

Def (Directional derivative, Partial derivative, Jacobi matrix/determinant)

For  $U \subseteq \mathbb{R}^n$  open,  $a \in U$ ,  $v \in \mathbb{R}^n$ ,  $f: U \rightarrow \mathbb{R}^m$ , ~~define~~ define

$$(\partial_v f)(a) := \lim_{t \rightarrow 0} \frac{f(a+tv) - f(a)}{t} \quad e \in \mathbb{R}^m \text{ or } \text{NaN}$$

slope:  $\partial_v f(a)$



E.p. one can define partial derivatives

$$\frac{\partial f}{\partial x_i} := \partial_i f := \partial_{e_i} f: U \rightarrow \mathbb{R}^m$$

$$e_i = \begin{pmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{pmatrix} \leftarrow \text{the } i\text{-th place}$$

Write  $f = (f_1, \dots, f_m)$ , denote

$$Df = \left( \frac{\partial f}{\partial x_1}, \dots, \frac{\partial f}{\partial x_n} \right) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \cdots & \frac{\partial f_m}{\partial x_n} \end{pmatrix} = \begin{pmatrix} Df_1 \\ \vdots \\ Df_m \end{pmatrix}$$

$Df$  is called the Jacobian matrix.

When  $m=n$ ,  $\det Df$  is called the Jacobian determinant of  $f$ .

Rmk. In this course, people use  $\nabla f$  as the transpose of  $Df$ , and call it the gradient.

$$\nabla f = (Df)^T = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{pmatrix} = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_m}{\partial x_1} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_1}{\partial x_n} & \cdots & \frac{\partial f_m}{\partial x_n} \end{pmatrix} = (\nabla f_1, \dots, \nabla f_m)$$

Task 1. For

$$f: \mathbb{R}^3 \longrightarrow \mathbb{R}^2 \quad f(x) = e_2 + Ax + (x^T B^T x) e_1$$

where  $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \quad B = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$ ,

compute  $Df$ .

$$\begin{aligned} A: \quad f(x) &= \begin{pmatrix} 0 \\ 1 \end{pmatrix} + \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} + \left( \begin{pmatrix} x_1 & x_2 & x_3 \end{pmatrix} \begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \right) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \\ &= \begin{pmatrix} 0 \\ 1 \end{pmatrix} + \begin{pmatrix} x_1 + 2x_2 + 3x_3 \\ 4x_1 + 5x_2 + 6x_3 \end{pmatrix} + \begin{pmatrix} x_1 x_2 + x_3 x_2 \\ 0 \end{pmatrix} \\ &= \begin{pmatrix} x_1 + 2x_2 + 3x_3 + x_2 x_1 + x_3 x_2 \\ 1 + 4x_1 + 5x_2 + 6x_3 \end{pmatrix} \end{aligned}$$

$$Df = \begin{pmatrix} 1+x_2 & 2+x_1+x_3 & 3+x_2 \\ 4 & 5 & 6 \end{pmatrix}$$

Task 3. For  $\Omega \subseteq \mathbb{R}^n$  open and connected,  $u \in C^1(\Omega; \mathbb{R})$ , show that

$$\nabla u(x) = 0 \quad \forall x \in \Omega \Rightarrow u \equiv C \text{ in } \Omega \setminus \bar{\Omega}$$

Hint: first show it for  $\Omega = (-1, 1)^n$

$$\frac{\partial u}{\partial x_1} \equiv 0 \Rightarrow u(x_1, 0, \dots, 0) = u(0, 0, \dots, 0) \stackrel{\Delta}{=} C$$

$$\frac{\partial u}{\partial x_2} \equiv 0 \Rightarrow u(x_1, x_2, \dots, 0) = u(x_1, 0, \dots, 0) = C$$

:

$$\frac{\partial u}{\partial x_n} \equiv 0 \Rightarrow u(x_1, x_2, \dots, x_n) = u(x_1, x_2, \dots, x_{n-1}, 0) = C$$

Then, show that, for a fixed  $x_0 \in \Omega$ ,

$\{x \in \Omega \mid u(x) = u(x_0)\}$  is both open and closed in  $\Omega$ .

$$\Rightarrow \{x \in \Omega \mid u(x) = u(x_0)\} = \Omega \setminus \bar{\Omega}$$

□

### 3. Total derivative (linear fct approximation)

Def For  $U \subseteq \mathbb{R}^n$  open, we say that  $f: U \rightarrow \mathbb{R}^m$  is (totally) differentiable at  $a \in U$ , if

$\exists A: \mathbb{R}^n \rightarrow \mathbb{R}^m$  linear s.t

$$\lim_{x \rightarrow a} \frac{\|f(x) - f(a) - A(x-a)\|}{\|x-a\|} = 0$$

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cont	$f(x) = f(a) + o(1)$
differentiable	$f(x) = f(a) + A(x-a) + o(\ x-a\ )$

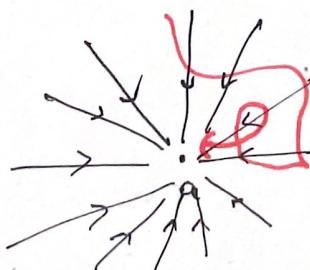
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Rmk. If  $f$  is differentiable, then  $A = (Df)(a)$ .

In this case,  $Df$  is called the (total) derivative or (total) differential.

Rmk. Total derivative is stronger than the directional derivative.  
When  $f$  is differentiable at  $a$ , then  $\forall v, w \in \mathbb{R}^n$ ,  $t, s \in \mathbb{R}$

$$D_{t+sw} f = t Dv f + s Dw f$$



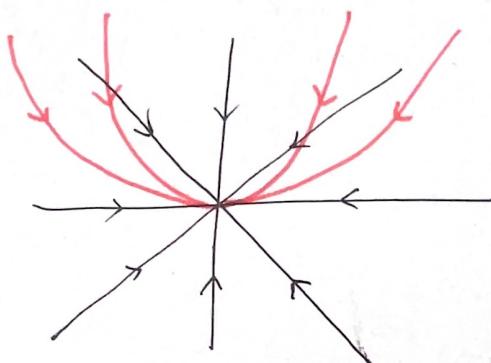
Task 2. Define

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

Show that  $g(x,y)$  is not cont at  $(0,0)$ , while

$$\lim_{t \rightarrow 0} g(tv) = 0$$

$$\forall v = (x_0, y_0) \in \mathbb{R}^2$$



→ goes to 0

→ does not go to 0

Similarly, define

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

Show that  $f(x,y)$  is cont.

$$\left[ \text{Hint: } \left| \frac{x^2y}{x^2+y^2} \right| \leq \frac{|x^2+y^2||y|}{|x^2+y^2|} = |y| \quad \text{or} \quad \left| \frac{x^2y}{x^2+y^2} \right| \leq \frac{|x| \cdot \frac{1}{2}|x^2+y^2|}{|x^2+y^2|} = \frac{1}{2}|x| \right]$$

Moreover, show that  $f(x,y)$  is not differentiable at  $(0,0)$ , while

$$Df = 0$$

$$D_v f = \cos^2 \theta \sin \theta$$

$$\forall v = (\cos \theta, \sin \theta) \in \mathbb{R}^2$$