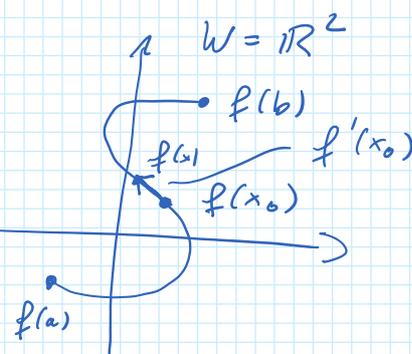
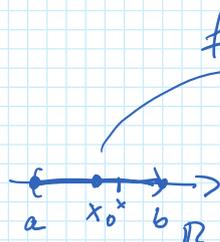
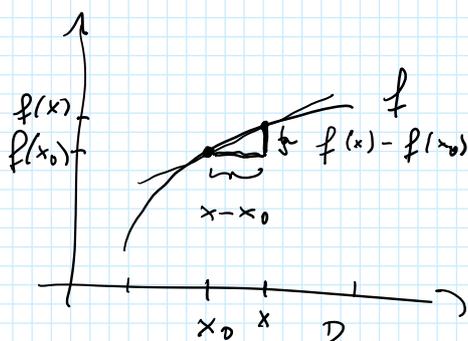


Recall that for  $f: \mathbb{R} \supset D \rightarrow \mathbb{R}^2 (W, \|\cdot\|)$

$$\begin{aligned} \underline{f'(x_0)} &:= \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0} \\ &= \lim_{h \rightarrow 0} \frac{f(x_0 + h) - f(x_0)}{h} \end{aligned}$$

$\forall (x_n) \text{ in } D \setminus \{x_0\} \text{ with } \lim_{n \rightarrow \infty} x_n = x_0$   
 $\lim_{n \rightarrow \infty} \frac{f(x_n) - f(x_0)}{x_n - x_0} \text{ exists.}$



$f'(x_0)$  = slope of tangent  
to the graph at  $(x_0, f(x_0))$

$f'(x_0) \in W$   
= velocity vector

Def.: Let  $n \in \mathbb{N}$ ,  $D \subset \mathbb{R}^n$  open,  $(W, \|\cdot\|)$  a normed space. For  $x \in D$  and  $j \in \{1, \dots, n\}$  a function  $f: D \rightarrow W$  is called partially differentiable in the  $j$ th coord. direction at  $x$ , iff the limit

$$\lim_{h \rightarrow 0} \frac{f(x + h e_j) - f(x)}{h} = \lim_{h \rightarrow 0} \frac{f(x_1, x_2, \dots, x_j + h, \dots, x_n) - f(x_1, \dots, x_n)}{h}$$

exists. One writes

$$\frac{\partial f}{\partial x_j}(x) = \partial_j f(x) := \lim_{h \rightarrow 0} \frac{f(x + h e_j) - f(x)}{h}$$

and calls the vector  $\partial_j f(x) \in W$  the  $j$ th partial

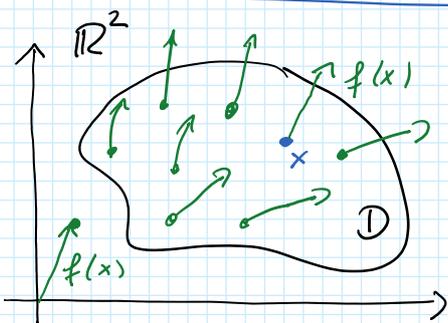
and calls the vector  $\partial_j f(x) \in W$  the  $j$ th partial derivative at  $x$ .

If  $f$  is partially diff. and the partial derivatives  $\partial_j f: D \rightarrow W$  are cont. fcts., then  $f$  is called cont. part. diff. The vector space of c.p.d. fcts on  $D \subset \mathbb{R}^n$  is denoted by  $C^1(D, W)$ .

The gradient of  $f$  at  $x$  is

$$\nabla f(x) := (\partial_1 f(x), \dots, \partial_n f(x)) \in W^n$$

Def.: A map  $f: \mathbb{R}^n \supset D \rightarrow \mathbb{R}^n$  is called a vector field.



Ex.: The gradient

$\nabla f: D \rightarrow \mathbb{R}^n$   
of a fct.  $f: D \rightarrow \mathbb{R}^n$

defines a vector field.

Def.: A fct.  $f: \mathbb{R}^n \supset D \rightarrow W^{\leftarrow (k-\text{VSp.})}$  is called  $r$ -times cont. part. diff., iff for all  $j = (j_1, \dots, j_r)$ ,  $j_i \in \{1, \dots, n\}$

- $f$  is c.p.d.
- $\partial_{j_1} f$  is c.p.d.
- $\partial_{j_2} \partial_{j_1} f$  is c.p.d.
- $\vdots$

- $\partial_{j_1} f$  is c.p.d.
- $\partial_{j_2} \partial_{j_1} f$  is c.p.d.
- $\vdots$
- $\partial_{j_r} \dots \partial_{j_1} f$  is continuous

The  $\mathbb{K}$ -vector space of  $r$ -times c.p.d. fcts is denoted by  $C^r(D, \mathbb{W})$ .

Def.: Let  $D \subset \mathbb{R}^n$ ,  $g \in C^1(D, \mathbb{R}^n)$ ,  $f \in C^2(D, \mathbb{R})$ .

Then  $\operatorname{div} g: D \rightarrow \mathbb{R}$ ,  $x \mapsto \operatorname{div} g(x) := \sum_{j=1}^n \frac{\partial g_j}{\partial x_j}(x)$

is called the divergence of  $g$ ,

$$\operatorname{curl} g: D \rightarrow \mathbb{R}^3, x \mapsto \operatorname{curl} g(x) := \begin{pmatrix} \partial_2 g_3(x) - \partial_3 g_2(x) \\ \partial_3 g_1(x) - \partial_1 g_3(x) \\ \partial_1 g_2(x) - \partial_2 g_1(x) \end{pmatrix}$$

for  $n=3$  is called the curl of  $g$ ,

and

$$\Delta f: D \rightarrow \mathbb{R}, x \mapsto \Delta f(x) := \left( \operatorname{div}(\nabla f) \right)(x) = \sum_{j=1}^n \frac{\partial^2 f}{\partial x_j^2}(x)$$

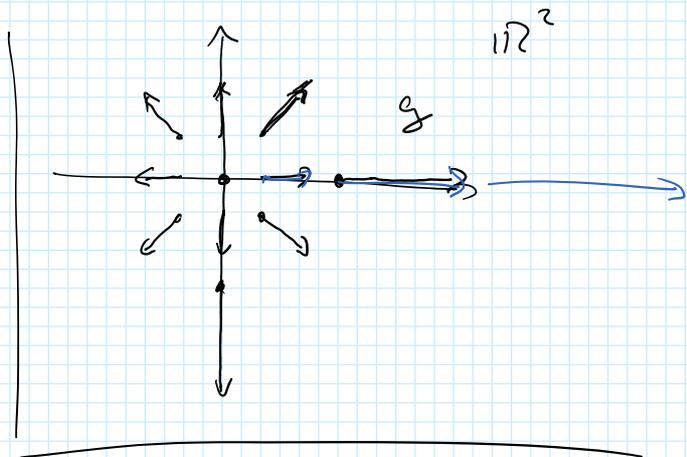
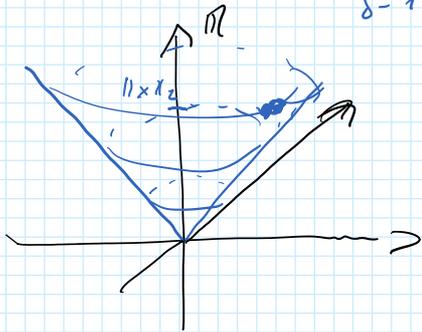
is called Laplace of  $f$ .

Ex. 1: Let  $g: \mathbb{R}^n \rightarrow \mathbb{R}^n$ ,  $x \mapsto g(x) = x$  ( $g = \operatorname{id}$ )

and  $f: \mathbb{R}^n \setminus \{0\} \rightarrow \mathbb{R}$ ,  $x \mapsto \|x\|_2$ . ( $n=1$ )  $f(x) = |x|$

Compute  $\operatorname{div}(g)$  and  $\Delta f$ .

Sol. 1:  $\operatorname{div}(g)(x) = \sum_{j=1}^n \frac{\partial g_j(x)}{\partial x_j} = \sum_{j=1}^n \frac{\partial x_j}{\partial x_j} = \sum_{j=1}^n 1 = n.$



$\Delta f(x) = \operatorname{div}(\nabla f)(x)$

$(\nabla f(x))_j = \partial_j f(x) = \partial_j \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$

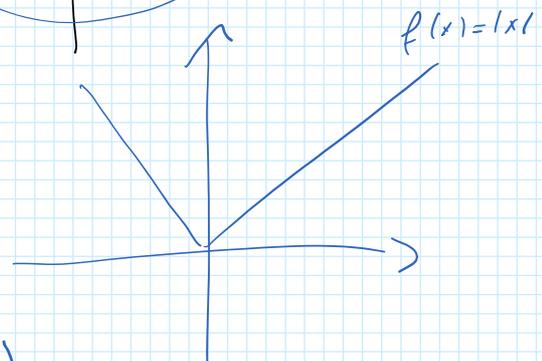
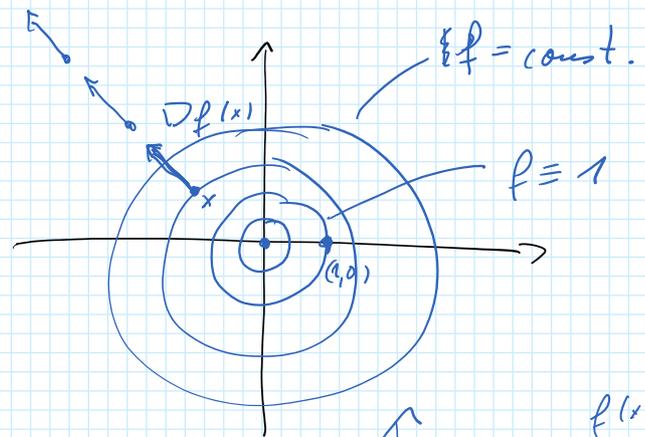
$= \frac{1}{2} \frac{2x_j}{\sqrt{x_1^2 + \dots + x_n^2}} = \frac{x_j}{\|x\|_2} \Rightarrow \nabla f(x) = \frac{x}{\|x\|_2}$

$\Delta f(x) = \operatorname{div}\left(\frac{x}{\|x\|_2}\right) =$

$= \sum_{j=1}^n \partial_j \left(\frac{x_j}{\|x\|_2}\right) =$

$= \sum_{j=1}^n \left( \frac{1}{\|x\|_2} - \frac{1}{\cancel{2}} \frac{x_j \cdot 2x_j}{\|x\|_2^3} \right)$

$= \frac{n}{\|x\|_2} - \frac{\|x\|_2^2}{\|x\|_2^3} = \frac{n-1}{\|x\|_2}$



Thm. (Schauder)

Let  $f \in C^2(D, \mathbb{R})$ ,  $D \subset \mathbb{R}^n$ .

Then  $\forall x \in D$ ,  $j, i \in \{1, \dots, n\}$

$\partial_j \partial_i f(x) = \partial_i \partial_j f(x)$

$$\partial_j \partial_i f(x) = \partial_i \partial_j f(x)$$

Corollary: Let  $D \subset \mathbb{R}^3$ ,  $f \in C^2(D, \mathbb{R})$  and  $g \in C^2(D, \mathbb{R}^3)$ .

Then  $\text{curl}(\nabla f) = 0$  and  $\text{div}(\text{curl } g) = 0$

Def.: Let  $V$  be a real normed space,  $W$  a normed space,

$D \subset V$  open and  $f: D \rightarrow W$ .

The directional derivative of  $f$  at  $x \in D$  in the direction  $v \in V$  is

$$\partial_v f(x) := \lim_{h \rightarrow 0} \frac{f(x + hv) - f(x)}{h} = \frac{d}{dh} f(x + hv) \Big|_{h=0}$$

if the limit exists.

Example: For  $f: \mathbb{R}^n \rightarrow W$  we have  $\partial_{e_j} f(x) = \partial_j f(x)$ .

Different viewpoint: Derivative as lin. approx.

For  $f: \mathbb{R} \rightarrow W$  differentiability at  $x_0 \in \mathbb{R}$  means

$$\textcircled{0} = \lim_{x \rightarrow x_0} \left( \frac{f(x) - f(x_0)}{x - x_0} - f'(x_0) \right) =: \lim_{x \rightarrow x_0} \frac{\varphi(x, x_0)}{x - x_0}$$

where  $\varphi(x, x_0) = f(x) - f(x_0) - f'(x_0)(x - x_0)$

or, after reshuffling

or, after reshuffling

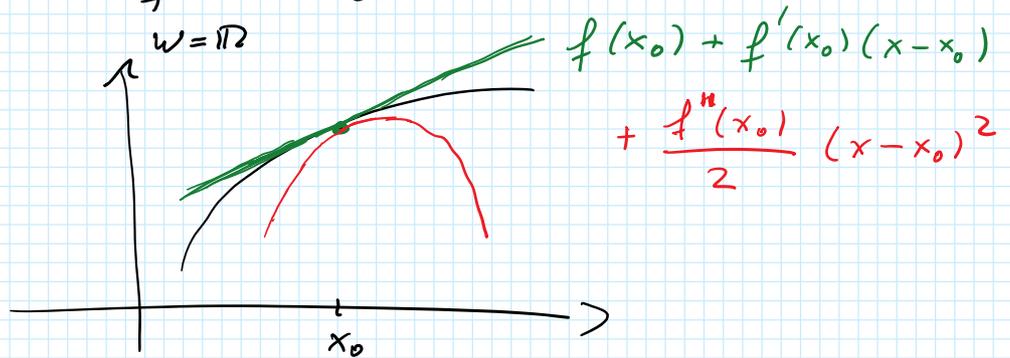
$$f(x) = \underbrace{f(x_0) + f'(x_0) \cdot (x - x_0)}_{(*)} + \varphi(x, x_0)$$

where

$$\varphi(x, x_0) = o(|x - x_0|) \Leftrightarrow \lim_{x \rightarrow x_0} \frac{|\varphi(x, x_0)|}{|x - x_0|} = 0$$

The map  $\mathbb{R} \rightarrow W$ ,  $x \mapsto f'(x_0) \cdot x$  is  $\mathbb{R}$ -linear  
and the map  $\mathbb{R} \rightarrow W$ ,  $x \mapsto f(x_0) + f'(x_0)(x - x_0)$   
is affine- $\mathbb{R}$ -linear.

Hence, we think of  $(*)$  as the (affine) linear  
approximation to  $f$  near  $x_0$ .



### Def.: Total derivative

Let  $V$  be a finite dim. real vector space,  
 $W$  a normed space,  $G \subset V$  open and  $f: G \rightarrow W$ .  
 $f$  is differentiable at  $x_0 \in G$ , iff there exist  
an  $\mathbb{R}$ -linear map  $A: V \rightarrow W$  such that

$$\lim_{x \rightarrow x_0} \frac{f(x) - (f(x_0) + A(x - x_0))}{\|x - x_0\|_V} = 0$$

(here  $\|\cdot\|_V$  is any  
norm on  $V$ )

Then  $A$  is uniquely determined (Ex. 2: prove this!),  
is denoted by  $Df|_x$ , and called the total derivative

is denoted by  $Df|_{x_0}$ , and called the total derivative or the differential of  $f$  at  $x_0$ .

If  $f: G \rightarrow W$  is diff. at all  $x \in G$ , then  $f$  is called differentiable on  $G$  and

$$Df: G \rightarrow \mathcal{L}(V, W), \quad x \mapsto Df|_x$$

is a fct. on  $G$  taking values in the linear maps from  $V$  to  $W$ .

Sol. 2: Let  $\tilde{A}: V \rightarrow W$  be linear and also satisfy

$$\lim_{x \rightarrow x_0} \frac{f(x) - f(x_0) - \tilde{A}(x - x_0)}{\|x - x_0\|_V} = 0$$

$$\Rightarrow \lim_{x \rightarrow x_0} \frac{A(x - x_0) - \tilde{A}(x - x_0)}{\|x - x_0\|_V} = \lim_{x \rightarrow x_0} \frac{\overbrace{(A - \tilde{A})}^B (x - x_0)}{\|x - x_0\|_V} = 0$$

$$\text{i.e. } \lim_{n \rightarrow \infty} \frac{B h_n}{\|h_n\|_V} = 0$$

$\forall (h_n)$  with  $\lim_{n \rightarrow \infty} h_n = 0$

$v \neq 0$

Let  $(\lambda_n)$  in  $\mathbb{R}$  with  $\lambda_n \xrightarrow{n \rightarrow \infty} 0$

$$Bv = \|v\| \cdot \frac{Bv}{\|v\|}$$

$$\lim_{n \rightarrow \infty} \frac{B(\lambda_n v)}{\|\lambda_n v\|} = 0 \Rightarrow Bv = 0$$

lin. of  $B = A - \tilde{A}$

Rem.:  $f: G \rightarrow W$  is diff. at  $x_0$

$$\Rightarrow f(x) = f(x_0) + Df|_{x_0}(x - x_0) + o(\|x - x_0\|_V)$$

$$\Rightarrow \left( f(x) = f(x_0) + \underbrace{Df|_{x_0}}_{\rightarrow 0} (x-x_0) + \underbrace{o(\|x-x_0\|_V)}_{\rightarrow 0 \text{ faster?}} \right)$$

Example: Let  $L: V \rightarrow W$  be an  $\mathbb{R}$ -lin. map.

$$\begin{aligned} \text{Then } L(x) &= L(x_0 + x - x_0) = L(x_0) + L(x - x_0) \\ &=: L(x_0) + DL|_{x_0} (x - x_0) \end{aligned}$$

and hence  $DL|_{x_0} = L$ .

$$\left( f(x) = x, f'(x) = 1 \quad ? \right)$$

Thm.: Let  $G \subset \mathbb{R}^n$  and let  $f: G \rightarrow \mathbb{K}^m$  be differentiable at  $x_0 \in G$ . Then

$$(Df|_{x_0})_{ij} = \frac{\partial f_i}{\partial x_j}(x_0)$$

or, more explicitly,

$$Df|_{x_0} = \left( \begin{array}{cccc} \partial_1 f_1(x_0) & \dots & \partial_n f_1(x_0) & \\ \vdots & & & \\ \partial_1 f_m(x_0) & \dots & \partial_n f_m(x_0) & \end{array} \right) \left. \vphantom{\begin{array}{c} \partial_1 f_1(x_0) \\ \vdots \\ \partial_1 f_m(x_0) \end{array}} \right\} \begin{array}{l} m \text{ rows} \\ n \text{ columns} \end{array}$$

$$= \left( \begin{array}{c} \nabla f_1(x_0) \\ \vdots \\ \nabla f_m(x_0) \end{array} \right)$$

"Jacobi matrix"

Thm.: Let  $G \subset \mathbb{R}^n$  open and  $f \in C^1(G, \mathbb{K}^m)$ .

Then  $f$  is differentiable.

$\uparrow$   
cont. part. diff.

Then  $f$  is differentiable.

cont. part. diff.

cont. part. diff.  $\Rightarrow$  differentiable  $\Rightarrow$  part. diff.

$\Downarrow$

continuous

None of the implications holds in the reversed direct.  $\nabla$

But

cont. part. diff.  $\Leftrightarrow$  diff. with cont. derivative

Thm.: Chain rule

Let  $U, V$  be finite dim. real vector spaces,

$W$  a normed space,  $G \subset U$ ,  $H \subset V$  open and

$g: G \rightarrow V$ ,  $f: H \rightarrow W$  maps with  $g(G) \subset H$ , i.e.

$$U \supset G \xrightarrow{g} H \subset V \xrightarrow{f} W.$$

If  $g$  is diff. at  $x \in G$  and  $f$  is diff. at  $g(x) \in H$ ,

then  $f \circ g: G \rightarrow W$  is diff. at  $x$  and

$$\underline{D(f \circ g)|_x} = \underbrace{Df|_{g(x)}}_{W \leftarrow V} \cdot \underbrace{Dg|_x}_{V \leftarrow U}$$

Corollary: For  $f \in C^1(G, W)$ ,  $G \subset V$ ,  $x_0 \in G$ , and

$v \in V$  we have

$$\partial_v f(x_0) = Df|_{x_0} v$$

$$\partial_v f(x_0) = Df|_{x_0} v$$

Ex. 3: prove this  $\forall$

$$g: \mathbb{R} \supset I \rightarrow G, h \mapsto x_0 + hv$$

$$\text{Sol. 3: } \partial_v f(x_0) = \left. \frac{d}{dh} f(\underbrace{x_0 + hv}_{=: g(h)}) \right|_{h=0} = \left. \frac{d}{dh} (f \circ g)(h) \right|_{h=0}$$

$$= Df|_{g(0)} g'(0) = Df|_{x_0} v$$

For  $f: G \rightarrow W$  the differential  $Df$  is a map

$$Df: G \rightarrow \mathcal{L}(V, W)$$

Thus the "second derivative"

$D(Df)$  is a map

$$g(v_1)(v_2) \in W$$

$$\cong g(v_1, v_2)$$

$$\begin{pmatrix} g_{ij} \dots \end{pmatrix}$$

$$D(Df): G \rightarrow \mathcal{L}(V, \mathcal{L}(V, W)) \cong \mathcal{L}_2(V \times V, W)$$

$$:= \text{bilinear maps } V \times V \rightarrow W$$

and the  $n$ th derivative

$$D^n f: G \rightarrow \mathcal{L}_n(\underbrace{V \times \dots \times V}_{n \text{ copies}}, W)$$

$$g(v_1, v_2) = \begin{pmatrix} g_{ij} v_1^i v_2^j \end{pmatrix} \omega$$

### Thm. Taylor

Let  $G \subset V$  open,  $x_0 \in G$ , and  $\delta > 0$  s.t.  $B_\delta(x_0) \subset G$ .

Then for any  $f \in C^2(G, W)$  and  $x \in B_\delta(x_0)$

$$f(y) - f(x) = Df|_x (y-x) + \frac{1}{2} D^2 f|_x (y-x, y-x) + \dots$$

$$f(x) = f(x_0) + \cancel{Df|_{x_0} \underbrace{(x-x_0)}_{=: h}} + \frac{1}{2} D^2 f|_{x_0}(h, h) + \dots$$

$$\dots + \frac{1}{n!} D^n f|_{x_0}(h, \dots, h) + o(\|h\|_V^2)$$

Def.: Let  $X$  be a <sup>top. space</sup> and  $f: X \rightarrow \mathbb{R}$ . A point  $x_0 \in X$  is called a (strict) local maximum of  $f$ , iff  $\exists U \subset \mathcal{U}(x_0)$  s.t.

$$\forall x \in U \setminus \{x_0\}: f(x) \leq f(x_0)$$

loc. minimum analogous.

Thm.: Let  $G \subset V$  and  $f \in C^1(G, \mathbb{R})$  have a local extremum at  $x_0 \in G$ . Then  $Df|_{x_0} = 0$ .

Thm.: Let  $G \subset V$  and  $f \in C^2(G, \mathbb{R})$  and  $x_0 \in G$  such that  $Df|_{x_0} = 0$ .

(a) If  $D^2 f|_{x_0}(h, h) > 0 \quad \forall h \in V \setminus \{0\}$ , then  $f$  has a strict local minimum at  $x_0$

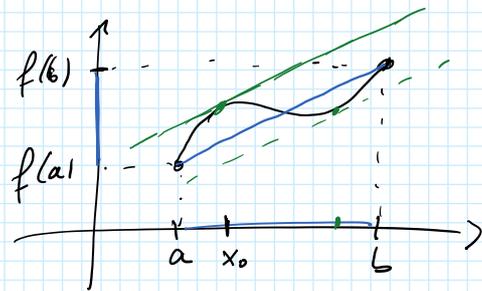
(b)  $D^2 f|_{x_0}(h, h) < 0 \quad \forall h \in V \setminus \{0\} \Rightarrow$  strict. loc. max. at  $x_0$

(c) If  $D^2 f|_{x_0}$  is indefinite, then  $f$  has no local extremum at  $x_0$

Mean value thm:  $f: [a, b] \rightarrow \mathbb{R}$  cont. and  
 cont. differentiable on  $(a, b)$ . Then  $\exists x_0 \in (a, b)$ :

$$f'(x_0) = \frac{f(b) - f(a)}{b - a}$$

$$\Rightarrow f(b) - f(a) = f'(x_0)(b - a)$$



Thm: Let  $G \subset \mathbb{R}^n$  and  $f: G \rightarrow \mathbb{K}^m$  cont. diff.

Let  $\gamma: [a, b] \rightarrow G$  cont. diff. Then

$$f(\gamma(b)) - f(\gamma(a)) = \int_a^b \underbrace{Df|_{\gamma(t)}}_{\mathbb{K}^m \times \mathbb{R}^n} \cdot \underbrace{\gamma'(t)}_{\in \mathbb{R}^n} dt$$

Thm (mean value thm)

$G \subset \mathbb{R}^n$ ,  $f \in C^1(G, \mathbb{K}^m)$ .

Let  $x \in G$  and  $h \in \mathbb{R}^n$  s.t.

$\{x + th \mid t \in [0, 1]\} \subset G$ . Then

$$f(x+h) - f(x) = \left( \int_0^1 Df|_{x+th} dt \right) \cdot h$$

$$\Rightarrow \|f(x) - f(y)\| \leq \left\| \int_0^1 Df|_{x+th} dt \right\|_{op} \cdot \|x - y\|$$

$$\sup_{z \in \overline{xy}} \|Df|_z\|$$

For  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  we obtain again  $\exists z \in \overline{xy}$

$$f(y) - f(x) = Df|_z \cdot (y - x)$$

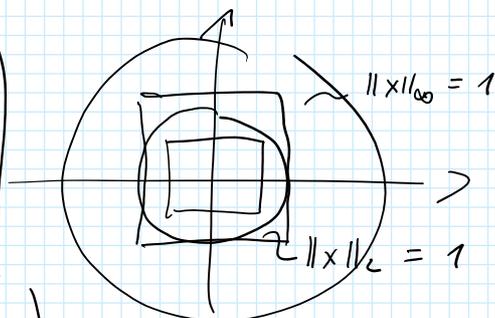
Def.: Equivalence of norms

Two norms  $\|\cdot\|_a$  and  $\|\cdot\|_b$  on a vector space  $V$  are equivalent, iff  $\exists c, C > 0 \forall x \in V$

$$c \|x\|_a \leq \|x\|_b \leq C \|x\|_a$$

(equivalence rel.)

Thm: On a finite dim. vector space all norms are equivalent.



Thm: All finite dim. normed spaces are complete.

Def.: Fréchet derivative

Let  $X$  and  $Y$  be Banach spaces and  $G \subset X$  open.

A map  $f: G \rightarrow Y$  is differentiable at  $x \in G$ , iff

$\exists$  continuous linear map  $A: X \rightarrow Y$  s.t.

$$f(x+h) = f(x) + Ah + o(\|h\|_X)$$

for  $h$  in a neighbourhood of  $0 \in X$ .  $A = Df|_x$

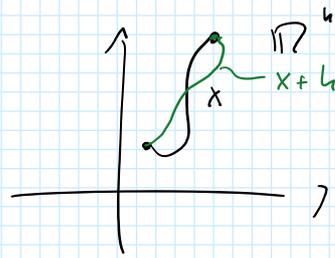
Example: Euler-Lagrange equations

$$X = C^2([0, T], \mathbb{R}^n)$$

$$x \in X, x: [0, T] \rightarrow \mathbb{R}^n$$

$$\|x\|_{\bar{X}} := \|x\|_\infty + \|\dot{x}\|_\infty + \|\ddot{x}\|_\infty$$

" $\int_0^T \|x(t)\|_2$ "



$$S: \bar{X} \rightarrow \mathbb{R} \quad (\text{action})$$

$$x \mapsto S(x) := \int_0^T L(x(t), \dot{x}(t)) dt$$

where  $L: \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $(q, v) \mapsto L(q, v) \in C^2$   
 (e.g.  $\underline{\underline{L(q, v) := \frac{1}{2} m \|v\|^2 - V(q)}}$ )

We compute the derivative  $DS|_x$  of  $S$ :  $x, h \in \bar{X}$

$$S(x+h) = \int_0^T L(x(t)+h(t), \dot{x}(t)+\dot{h}(t)) dt$$

$$= \int_0^T \left( \underline{\underline{L(x(t), \dot{x}(t))}} + \left\{ \left. \begin{aligned} &D_q L|_{(x(t), \dot{x}(t))} \cdot h(t) + \\ &D_v L|_{(x(t), \dot{x}(t))} \cdot \dot{h}(t) \end{aligned} \right\} \right) dt + \cancel{O(\|h\|_X)} + \underbrace{O(\|h\|_X^2)}$$

$$= S(x) + \int_0^T \left( \left. \begin{aligned} &D_q L|_{(x(t), \dot{x}(t))} h(t) + \\ &D_v L|_{(x(t), \dot{x}(t))} \dot{h}(t) \end{aligned} \right) dt \right. \\
\left. = DS|_x \cdot h + O(\|h\|_X^2) \right)$$

$\Rightarrow$

$$= S(x) + \underbrace{D_v L|_{(x(T), \dot{x}(T))} \cdot h(T)}_{=0} - \underbrace{D_v L|_{(x(0), \dot{x}(0))} \cdot h(0)}_{=0} \\
+ \int_0^T \left( \underbrace{D_q L|_{(x(t), \dot{x}(t))} - \left( \frac{d}{dt} D_v L|_{(x(t), \dot{x}(t))} \right)}_{=0} \right) \underline{\underline{h(t)}} dt \\
= DS|_x h + O(\|h\|_X^2)$$

for  $h \in X$  s.t.  $h(0) = h(T) = 0$

for  $h \in X$  s.t.  $h(0) = h(T) = 0$

$$DS|_X h = 0$$

$$\Leftrightarrow \left[ D_{\dot{q}} L \Big|_{(x(t), \dot{x}(t))} - \frac{d}{dt} D_{\dot{q}} L \Big|_{(x(t), \dot{x}(t))} \right] = 0$$

Euler-Lagrange equ.

$$-D_{\dot{q}} V(\dot{q}(t)) - m \ddot{x}(t) = 0$$

$$f: V_1 \times V_2 \rightarrow W \quad (v_1, v_2) \in V_1 \times V_2$$

$$f_{v_2}: V_1 \rightarrow W, \quad f_{v_2}(v_1) = f(v_1, v_2)$$

$$D_{v_1} f|_{(v_1, v_2)} := D f_{v_2}|_{v_1}$$

$$f: \mathbb{R}_x^n \times \mathbb{R}_y^m \rightarrow \mathbb{K}^l$$

$$Df = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} & \frac{\partial f_1}{\partial y_1} & \dots & \frac{\partial f_1}{\partial y_m} \\ \vdots & & \vdots & & & \vdots \\ \frac{\partial f_l}{\partial x_1} & \dots & \frac{\partial f_l}{\partial x_n} & \dots & \dots & \frac{\partial f_l}{\partial y_m} \end{pmatrix}$$

$$\underbrace{\hspace{10em}}_{D_x f} \quad \underbrace{\hspace{10em}}_{D_y f}$$

Def.: A subset  $A \subset X$  of a top. space  $(X, \mathcal{T})$  is called dense, iff  $\bar{A} = X$ .

Rem.: If  $A \subset X$  is dense, then  $A \cap O \neq \emptyset$  for any  $O \in \mathcal{T} \setminus \{\emptyset\}$ .

Prop.: Let  $f, g: X \rightarrow Y$  be cont. fcts.,  $Y$  Hausdorff, and  $A \subset X$  dense. Then

$$f|_A = g|_A \Rightarrow f = g.$$

Exc: prove this.