

Old exercise sheet

Ex 7.1

a) $f(x,y) = y$, $p_0 = (3, \frac{1}{3})$, $v = (-2, -4)$

1. We want to determine the differential of f at point p_0 : $df(p_0)$

• We do this by determining the Jacobi matrix of f at point p_0 :

$$J_f(p_0) = \left(\frac{\partial f}{\partial x}(p_0), \frac{\partial f}{\partial y}(p_0) \right) = \underline{(0 \ 1)} \cdot \begin{matrix} (p_{0,1}) \\ (p_{0,2}) \end{matrix}$$

→ p_0 is the point at which the partial derivatives are evaluated!!

• The differential of f at point p_0 is therefore given by

$$df(p_0): \mathbb{R}^2 \rightarrow \mathbb{R}, \begin{pmatrix} x \\ y \end{pmatrix} \mapsto J_f(p_0) \cdot \begin{pmatrix} x \\ y \end{pmatrix}$$

→ the Jacobi matrix is the representation of the differential!

→ When asked for the differential, we sometimes just state the

Jacobi matrix! $df(p_0) \neq J_f \cdot \begin{pmatrix} p_{0,1} \\ p_{0,2} \end{pmatrix}$ (matrix product)

2. Calculate $df(p_0) \cdot v$ (which is essentially $df(p_0)(v)$)

linear map argument/point where to evaluate

• $df(p_0) \cdot v = (0 \ 1) \cdot \begin{pmatrix} -2 \\ -4 \end{pmatrix} = 0 \cdot (-2) + 1 \cdot (-4) = \underline{-4}$

Ex 7.2

b) $f(x,y,z) = \int_{\cos x + \sin y}^z e^{tz} dt$ $df(\frac{\pi}{2}, \frac{\pi}{3}, 0) = ?$

• Evaluate integral $\tilde{f}(x,y,z) = \frac{1}{z} e^{z^2} - \frac{1}{z} e^{z(\cos x + \sin y)}$ (z ≠ 0)

• $\partial_x \tilde{f}(\frac{\pi}{2}, \frac{\pi}{3}, 0)$, $\partial_y \tilde{f}(\frac{\pi}{2}, \frac{\pi}{3}, 0)$

• 1. $\partial_z f(\frac{\pi}{2}, \frac{\pi}{3}, z) = -\frac{\exp(z^2)}{z^2} + 2e^{z^2} + \frac{\exp(z \cdot a)}{z^2} - \frac{a \cdot \exp(z \cdot a)}{z}$, $a = \frac{\sqrt{3}}{2}$

$$\lim_{z \rightarrow 0} \partial_z f\left(\frac{\pi}{2}, \frac{\pi}{3}, z\right) = \dots = \frac{5}{8} \quad (\text{l'Hospital's rule})$$

$$2. \quad \partial_z f\left(\frac{\pi}{2}, \frac{\pi}{3}, 0\right) = \lim_{h \rightarrow 0} \frac{f\left(\frac{\pi}{2}, \frac{\pi}{3}, h\right) - f\left(\frac{\pi}{2}, \frac{\pi}{3}, 0\right)}{h} = \dots = \frac{5}{8}$$

Note that you have to use \otimes to evaluate $f\left(\frac{\pi}{2}, \frac{\pi}{3}, 0\right)$!

→ you can also use Leibniz's rule

$$c) \quad f(x, y) = |xy| \quad df(x, y) = ?$$

$$\bullet \quad xy > 0 \quad \rightarrow \quad df(x, y) = (y \quad x)$$

$$\bullet \quad xy < 0 \quad \rightarrow \quad df(x, y) = (-y \quad -x)$$

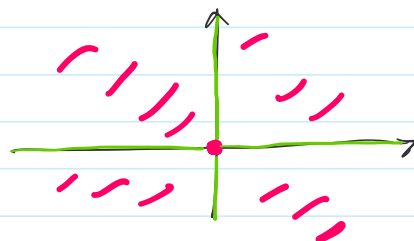
$$\bullet \quad xy = 0 \quad 1. \quad x \neq 0, \quad y = 0$$

$$\lim_{h \rightarrow 0} \frac{f(x, h) - f(x, 0)}{h} = \frac{|x \cdot h| - 0}{h} = \frac{|h|}{h} \cdot |x|$$

→ no limit!

$$2. \quad y \neq 0, \quad x = 0 \quad (\text{as part 1.})$$

$$3. \quad x = 0, \quad y = 0 : \quad \lim_{(x, y) \rightarrow (0, 0)} \frac{f(x, y) - f(0, 0)}{\sqrt{x^2 + y^2}} = 0 \quad \rightarrow \quad df(0, 0) = 0$$

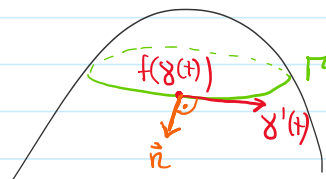


Ex 7.3

c) To show: $D_\sigma f(x, y)$ for $(x, y) \in \Gamma$ and $\|\sigma\| = 1$

assumes its maximum if

σ is orthogonal to Γ



Proof: Idea: we want to write every vector σ as a linear comb. of vectors pointing along Γ and orthogonal to Γ and show that the coefficient of the latter has to be maximised!

Let $\sigma \in \mathbb{R}^2$ arbitrary with $\|\sigma\| = 1$. Then we can write it as

$$\sigma = a \cdot \gamma'(t) + b \cdot n \quad \boxed{\square}$$

where $\gamma'(t)$ is the tangent of Γ at point $\gamma(t) = (x, y)$ and n is the

where $\gamma'(t)$ is the tangent of Γ at point $\gamma(t) = (x, y)$ and n is the vector orthogonal to $\gamma'(t)$ with $\|n\|=1$, meaning $n \cdot \gamma'(t) = 0$.

Then we have (using (a))

$$\begin{aligned} D_v f(\gamma(t)) &= \int_f(\gamma(t)) \cdot v \\ &= a \cdot \underbrace{\nabla f(\gamma(t)) \cdot \gamma'(t)}_{=0 \text{ (a)}} + b \cdot \nabla f(\gamma(t)) \cdot n \end{aligned}$$

\Rightarrow a has to be zero as its contribution vanishes!

b has to be ± 1 to maximise the expression.

\rightarrow whatever vector we take, its normal component to Γ maximises the gradient!

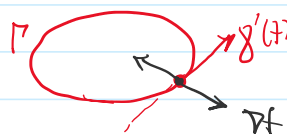
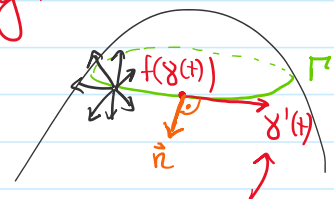
a) $\Gamma = f^{-1}(\{c\}) \quad \gamma((a,b)) = \Gamma$

$$\frac{d}{dt} f(\gamma(t)) = \frac{d}{dt} c = 0$$

$$\nabla f(\gamma(t)) \cdot \gamma'(t)$$

$\gamma'(t)$ is pointing along Γ

$$\begin{aligned} \gamma' \perp v \\ \gamma'(t) \cdot v = 0 \end{aligned}$$



Revision

• Taylor approximation

functions whose partial derivatives of order k exist & are continuous!

Def: $X \subset \mathbb{R}^n$ open, $x_0 \in X$, $k \geq 1$ $f \in C^k(X; \mathbb{R})$. Then the k -th Taylor polynomial

of f at point x_0 is a polynomial in n variables of degree $\leq k$ given by

$$\begin{aligned} T_k f(y; x_0) &= f(x_0) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(x_0) \cdot y_i + \frac{1}{2} \cdot \sum_{i,j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(x_0) \cdot y_i \cdot y_j + \dots \\ &= df(x_0) \cdot y \quad = \frac{1}{2} y^T \cdot \text{Hess}_f(x_0) \cdot y \end{aligned}$$

point to look at \uparrow point around which we are approximating

$$+ \sum_{m_1 + \dots + m_n = k} \frac{1}{m_1! \dots m_n!} \frac{\partial^k f(x_0)}{\partial x_1^{m_1} \dots \partial x_n^{m_n}} y_1^{m_1} y_2^{m_2} \dots y_n^{m_n}$$

Question: Can we use these polynomials to approximate f ?

Answer: Yes!

Theorem (Taylor approximation)

$X \subset \mathbb{R}^n$ open, $x_0 \in X$, $k \geq 1$, $f \in C^k(X; \mathbb{R})$. Then

$$f(x) = T_k f(x-x_0; x_0) + E_k(f, x, x_0)$$

↑
Errors of approximation

where

$$\lim_{\substack{x \rightarrow x_0 \\ x \neq x_0}} \frac{E_k(f, x, x_0)}{\|x-x_0\|^k} = 0.$$

Rmk: The second order Taylor approximation of a function $f: \mathbb{R}^2 \rightarrow \mathbb{R}$

is given by:

$$T_1 f \left[\begin{pmatrix} x-x_0 \\ y-y_0 \end{pmatrix}, \begin{pmatrix} x_0 \\ y_0 \end{pmatrix} \right] = f(x_0, y_0) + df(x_0, y_0) \cdot \begin{pmatrix} x-x_0 \\ y-y_0 \end{pmatrix}$$

$$T_2 f \left[\begin{pmatrix} x-x_0 \\ y-y_0 \end{pmatrix}, \begin{pmatrix} x_0 \\ y_0 \end{pmatrix} \right] = f(x_0, y_0) + df(x_0, y_0) \cdot \begin{pmatrix} x-x_0 \\ y-y_0 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} x-x_0 & y-y_0 \end{pmatrix} \cdot \text{Hess}_f(x_0, y_0) \cdot \begin{pmatrix} x-x_0 \\ y-y_0 \end{pmatrix}$$

where $\text{Hess}_f(x, y) = \begin{pmatrix} \partial_x^2 f & \partial_x \partial_y f \\ \partial_y \partial_x f & \partial_y^2 f \end{pmatrix}$ is the Hessian matrix of f

Examples:

(1) $f(x) = \frac{1}{1-x}$, $x_0 = 0$, $k = 1, 2$

• $\frac{df}{dx} = \frac{1}{(1-x)^2}$, $\frac{d^2f}{dx^2} = \frac{2}{(1-x)^3}$, $\frac{d^2f}{dx^2}(0) = \frac{2}{(1-0)^3} = 2 \otimes$

• $T_1 f(x; 0) = f(0) + \frac{df}{dx}(0) \cdot x = 1 + \frac{1}{(1-x)^2} \Big|_{x=0} \cdot x$ (affine linear approximation)

↑ point to look at
↑ point around which we are approximating
= 1

• $T_2 f(x; 0) = T_1 f(x; 0) + \frac{1}{2} \frac{d^2f}{dx^2}(0) \cdot x^2$

= 2 \otimes

$$= 1 + x + x^2$$

$$\Rightarrow \text{Geometric series: } \frac{1}{1-x} = \sum_{n=0}^{\infty} x^n \text{ for } |x| < 1$$

$$(2) \quad g(x, y) = x \cdot y, \quad (x_0, y_0) = (1, 1) \quad k=1, 2$$

$$\bullet \quad \frac{\partial g}{\partial x} = y, \quad \frac{\partial g}{\partial y} = x \quad \rightarrow \quad dg(x, y) = (y \ x)$$

$$\frac{\partial^2 g}{\partial x \partial y} = \frac{\partial^2 g}{\partial y \partial x} = 1, \quad \frac{\partial^2 g}{\partial x^2} = \frac{\partial^2 g}{\partial y^2} = 0 \quad \rightarrow \quad \text{Hess}_g(x, y) = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

$$\bullet \quad T_1 g \left[\begin{pmatrix} x \\ y \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right] = g(1, 1) + dg(1, 1) \cdot \begin{pmatrix} x-1 \\ y-1 \end{pmatrix}$$

(affine linear approximation)

$$= 1 + (1 \ 1) \cdot \begin{pmatrix} x-1 \\ y-1 \end{pmatrix}$$

$$= 1 + (x-1) + (y-1)$$

$$= \underline{-1 + x + y}$$

$$\bullet \quad T_2 g \left[\begin{pmatrix} x \\ y \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right] = g(1, 1) + dg(1, 1) \cdot \begin{pmatrix} x-1 \\ y-1 \end{pmatrix} + \frac{1}{2} \cdot (x-1, y-1) \cdot \text{Hess}_g(1, 1) \cdot \begin{pmatrix} x-1 \\ y-1 \end{pmatrix}$$

$$= -1 + x + y + \frac{1}{2} (x-1 \ y-1) \cdot \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \cdot \begin{pmatrix} x-1 \\ y-1 \end{pmatrix}$$

$$= -1 + x + y + \frac{1}{2} (x-1 \ y-1) \cdot \begin{pmatrix} y-1 \\ x-1 \end{pmatrix}$$

$$= -1 + x + y + 2 \cdot \frac{1}{2} \cdot (x-1) \cdot (y-1)$$

$$= \underline{-1} + \underline{x} + \underline{y} + \underline{x} \underline{y} - \underline{x} - \underline{y} + \underline{1}$$

$$= x \cdot y = g(x, y)! \quad \rightarrow \quad \text{2nd order polynomial!}$$

$$(3) \quad h(x_1, x_2, x_3) = x_1 x_2 x_3, \quad x_0 = (2, 2, 2)$$

$$\rightarrow \bullet \quad \partial_1 f = x_2 x_3, \quad \partial_2 f = x_1 x_3, \quad \partial_3 f = x_1 x_2 \quad \rightarrow \quad dh(x_1, x_2, x_3) = (x_2 x_3 \ x_1 x_3 \ x_1 x_2)$$

$$\partial_1^2 f = \partial_2^2 f = \partial_3^2 f = 0, \quad \partial_1 \partial_2 f = x_3, \quad \partial_1 \partial_3 f = x_2, \quad \partial_2 \partial_3 f = x_1$$

$$\rightarrow \quad \text{Hess}_h(x_1, x_2, x_3) = \begin{pmatrix} 0 & x_3 & x_2 \\ x_3 & 0 & x_1 \\ x_2 & x_1 & 0 \end{pmatrix}$$

$$\rightarrow \bullet \quad T_2 h \left[\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}, \begin{pmatrix} 2 \\ 2 \\ 2 \end{pmatrix} \right] = \underbrace{h(2, 2, 2) + dh(2, 2, 2) \cdot \begin{pmatrix} x_1-2 \\ x_2-2 \\ x_3-2 \end{pmatrix}}_{\text{affine linear approximation}}$$

$$+ \frac{1}{2} (x_1-2 \ x_2-2 \ x_3-2) \cdot \text{Hess}_f(2, 2, 2) \cdot \begin{pmatrix} x_1-2 \\ x_2-2 \\ x_3-2 \end{pmatrix}$$

$$= 2^3 + 2^2 \cdot [(x_1-2) + (x_2-2) + (x_3-2)]$$

$$+ 2 \cdot \sum_{i < j}^3 (x_i-2) \cdot (x_j-2)$$

Exercise sheet

MC 8.1 (Fct dis-/continuous, diff, C^1)

- differentiability \Rightarrow continuity (use differential quotient at $(x,y)=(0,0)$)
- Def: $X \subset \mathbb{R}^n$ open, $f: X \rightarrow \mathbb{R}^n$ differentiable. We say that f is of class C^1 (or $f \in C^1(X; \mathbb{R}^n)$) if it has all partial derivatives and they are continuous!
 \rightarrow are the partial derivatives continuous?

MC 8.2 (Critical point, tangent plane)

Have a look at the definition of the tangent space. What's the equation of the affine linear approximation?

Exc 8.1 (Fct \rightarrow tangent plane)

a) $G = \{(x, y, z) \in \mathbb{R}^3 \mid z = f(x, y)\}$

$E = \{(x, y, z) \in \mathbb{R}^3 \mid z = \tilde{z}\}$

\rightarrow affine linear approximation

b) Use mathematica (like this plot)

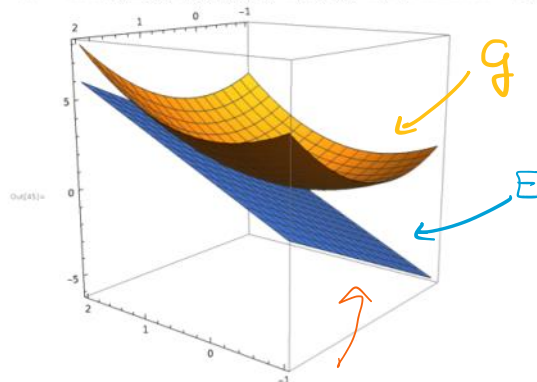
or other programs like pgfplots, octave, sagemath

c) What's the condition for $df(x,y)$ such that the tangent plane is

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In[20]:= g[x_, y_] = x^2 + y^2
Out[20]= x^2 + y^2
In[33]:= a = 1; b = 1;
dg[x_, y_] = g[a, b] + Derivative[1, 0][g][a, b] * (x - a) + Derivative[0, 1][g][a, b] * (y - b)
Out[33]= 2 + 2(-1 + x) + 2(-1 + y)
In[45]:= Plot3D[{g[x, y], dg[x, y]}, {x, -1, 2}, {y, -1, 2}, BoxRatios -> {1, 1, 1}]

```



c) What's the condition for $dt(x,y)$ such that the tangent plane is parallel to the x - y -plane?

Ex 8.2 (Secd order Taylor approx, error estimation)

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a) Use what we discussed!

b) It's all about estimation!

$$R_1 f(x,y) = \frac{1}{2} \underbrace{\frac{\partial^2 f}{\partial x^2}(x_0, y_0)}_{\leftarrow \dots} \underbrace{(x-x_0)^2}_{\leftarrow \dots} + \dots$$

Ex 8.3

a) $\sum_{n=0}^{\infty} b^n = \frac{1}{1-b}$ if $|b| < 1$ (Geometric series)

b) $\arctan(t) = \sum_{n=0}^{\infty} (-1)^n \cdot \frac{t^{2n+1}}{2n+1} = t + \dots$

c) $\log(1+t) = \sum_{k=1}^{\infty} \frac{(-1)^{k-1}}{k} t^k$ for $|t| < 1$

d) Consider the simplified example (3)