

# Homework 1

1) Prove:

$f$  is convex  $\Leftrightarrow x_1, \dots, x_k \in D, \alpha_1, \dots, \alpha_k \in [0,1], \sum \alpha_i = 1: f\left(\sum \alpha_i x_i\right) \leq \sum \alpha_i f(x_i)$

Induction: for proving  $\Rightarrow$

$$1) n=2 \quad f(tx + (1-t)y) \leq tf(x) + (1-t)f(y) \quad (\text{from definition of convexity})$$

2) Assume it holds for  $n=k$ :

$$\boxed{f\left(\sum \alpha_i x_i\right) \leq \sum \alpha_i f(x_i)}$$

Prove it holds for  $n=k+1$ :

$$f\left(t \sum \alpha_i x_i + (1-t)x_{k+1}\right) \stackrel{\substack{\text{because} \\ f \text{ is convex}}}{\leq} t f\left(\sum \alpha_i x_i\right) + (1-t) f(x_{k+1}) \stackrel{\text{assumption}}{\leq}$$

$$\leq t \cdot \sum \alpha_i f(x_i) + (1-t) f(x_{k+1}) \leq \sum \alpha_i t f(x_i) + (1-t) f(x_{k+1})$$

Define coefficients:  $\beta_1 = \alpha_1 t, \dots, \beta_k = \alpha_k t, \beta_{k+1} = 1-t$

Check the condition:  $\sum \alpha_i t + (1-t) = t + 1 - t = 1$  ✓

$$\text{Finally: } \boxed{f\left(\sum \beta_i x_i\right) \leq \sum \beta_i f(x_i)}$$

$\Leftarrow$  if  $f\left(\sum \alpha_i x_i\right) \leq \sum \alpha_i f(x_i)$  we can set all  $\alpha_3, \alpha_4, \dots, \alpha_k = 0$  and  $f(\alpha_1 x_1 + \alpha_2 x_2) \leq \alpha_1 f(x_1) + \alpha_2 f(x_2)$ , since that is the convexity definition it follows that  $f$  is convex function.

2)  $f(x,y) = x^2 + e^x + y^2 - xy \quad f: [-2,2] \times [-2,2] \rightarrow \mathbb{R}$

Finding  $\nabla f(x,y) = \begin{bmatrix} 2x + e^x - y \\ 2y - x \end{bmatrix}$ ; by definition  $f$  is  $L$ -Lipschitz iff  $\|\nabla f\| \leq L$

$$\|\nabla f(x,y)\| = \sqrt{(2x + e^x - y)^2 + (2y - x)^2} \quad \text{because } \sqrt{x} \text{ is monotone increasing}$$

finding the max of  $\sqrt{x}$  and  $x$  is the same problem.

$$\frac{\|\nabla f\|}{2x} = 2(2x + e^x - y)(2 + e^x) - 2(2y - x) = 0 \quad \frac{\|\nabla f\|}{2y} = 2(2x + e^x - y) + 2(2y - x) \cdot 2 = 0$$

$$4x + 2e^x - 2y + 2xe^x + e^{2x} - ye^x - 2y + x = 0 \quad -2x - e^x + y + 4y - 2x = 0$$

$$5x + 2e^x + 2xe^x + e^{2x} - \frac{4}{5}(4x + e^x) - e^x \cdot \frac{1}{5}(4x + e^x) = 0 \quad 5y = 4x + e^x \quad y = \frac{1}{5}(4x + e^x)$$

$\Rightarrow$  no solution in  $\mathbb{R} \Rightarrow$  no max  $\|\nabla f\|$

Let's check on the boundaries:

$$\|\nabla f\|^2 = (2x + e^x - y)^2 + (2y - x)^2$$

$$x=2 \quad \|\nabla f\|^2 = (4 + e^2 - y)^2 + (2y - 2)^2$$

In this case we observe  $\|\nabla f\|^2$  is larger as  $y$  is larger, so

it is checked in  $y=2/2$

$$x=2 \quad \|\nabla f\|^2 = (-4 + \frac{1}{e^2} - y)^2 + (2y - 2)^2$$

Again when  $y \rightarrow -\infty, +\infty$   $\|\nabla f\|^2$  is the biggest so check for  $y=-2/2$

$$\|\nabla f(2,2)\|^2 = (2 + e^2)^2 + 4 = \boxed{1}$$

$$\|\nabla f(-2,2)\|^2 = (-6 + e^{-2})^2 + 4 = 38,39$$

$$\|\nabla f(2,-2)\|^2 = (6 + e^2)^2 + 36 = \boxed{215,26} \quad \|\nabla f(-2,-2)\|^2 = (-2 + e^{-2})^2 + 36 = 39,47$$

$$\sqrt{215,26} = 14,67 = L$$

Finding  $\beta$  and  $\alpha$

- Eigenvalues of  $\nabla^2 f$ -smooth funct. lie on  $[0, \beta]$

- Eigenvalues of  $\nabla^2 f$ , of  $\alpha$ -strongly convex funct. are all  $\geq \alpha$

$$\nabla^2 f = \begin{bmatrix} 2+e^x & -1 \\ -1 & 2 \end{bmatrix} \quad \det(\nabla^2 f - \lambda I_2) = \begin{vmatrix} 2+e^x-\lambda & -1 \\ -1 & 2-\lambda \end{vmatrix} = 4-2\lambda+2e^x-2\lambda-\lambda e^x+\lambda^2-1=0$$

$$= \lambda^2 - \lambda(4+e^x) + 2e^x + 3 = 0$$

$$\lambda_{1,2} = \frac{4+e^x \pm \sqrt{16+8e^x+e^{2x}-8e^x-12}}{2} = \frac{4+e^x \pm \sqrt{4+e^{2x}}}{2}$$

The biggest  $\lambda$  is at  $x=2$

$$\beta = \frac{4+e^2 + \sqrt{4+e^4}}{2} = \boxed{9,522 = \beta}$$

The smallest  $\lambda$  is at  $x=-2$

$$\alpha = \frac{4+e^{-2} - \sqrt{4+e^{-4}}}{2} = \boxed{1,07 = \alpha}$$

Proving it is convex:

$$\begin{bmatrix} 2+e^x & -1 \\ -1 & 2 \end{bmatrix}$$

Sylvester thm:

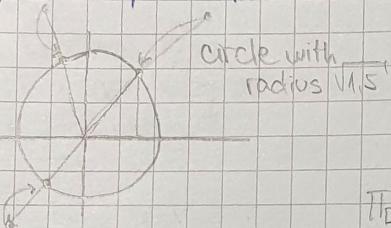
- check  $\det(1|2+e^x|) \geq 0 \quad \checkmark$

- check  $\det(\begin{vmatrix} 2+e^x & -1 \\ -1 & 2 \end{vmatrix}) = 4+2e^x-1 = 3+2e^x \geq 0 \quad \checkmark$

Therefore  $\nabla^2 f$  is PSD.  $e^x \geq 0$

3) Find  $\Pi_D : \mathbb{R}^2 \rightarrow K$  for the following convex sets:

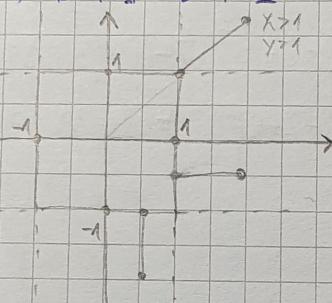
a)  $x^2 + y^2 \leq 1,5$



We can see that the newly projected coordinates  $x, y$  and the old should be lin. dependant!

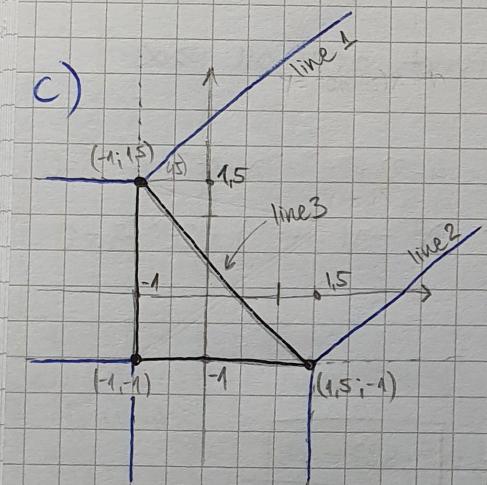
$$\Pi_D(\vec{x}) = \begin{cases} \vec{x} & \text{if } \vec{x} \in K \\ \frac{\vec{x}}{\|\vec{x}\|} \cdot \sqrt{1.5} & \text{else} \end{cases}$$

b)  $[-1, 1] \times [-1, 1]$



$$\Pi_D(x, y) = \begin{cases} (x, y) & \text{if } (x, y) \in K \\ (1, y) & x \geq 1, |y| \leq 1 \\ (-1, y) & x \leq -1, |y| \leq 1 \\ (x, 1) & y \geq 1, |x| \leq 1 \\ (x, -1) & y \leq -1, |x| \leq 1 \\ (1, 1) & x \geq 1, y \geq 1 \\ (-1, 1) & x \leq -1, y \geq 1 \\ (-1, -1) & x \leq -1, y \leq -1 \\ (1, -1) & x \geq 1, y \leq -1 \end{cases}$$

c)



$$\Pi_D(x, y) = \begin{cases} (x, y) & \text{if } (x, y) \in K \\ (-1, y) & -1 \leq y \leq 1.5, x < -1 \\ (x, -1) & -1 \leq x \leq 1.5, y < -1 \\ (-1, -1) & x < -1, y < -1 \\ (1, 1.5) & y > 1.5 \wedge y > x + 2.5 \\ (1.5, -1) & x > 1.5 \wedge y < x - 2.5 \\ \left(\frac{1}{2}(x+y)\right), \left(\frac{1}{2}(\frac{1}{2}+y-x)\right) & \text{otherwise} \end{cases}$$

line 1:  $1.5 = 1 \cdot (-1) + n$   
 $n = 2.5$        $y = x + 2.5$

line 2:  $1.5 = -1 + n$   
 $n = -2.5$        $y = x - 2.5$

line 3:  $1.5 = -1 - 1.5$   
 $n = -2.5$        $y = -x + 0.5$

$$y + 1 = (-1)(x - 1.5)$$

$$y = -x + 0.5$$

Slope of perpendicular will be:  $k = 1$

Say we have point  $(x_1, y_1)$        $\begin{cases} y_1 = x_1 + n \\ n = y_1 - x_1 \end{cases}$

intersection of two lines  $\Rightarrow$

$$\begin{cases} y = x + (y_1 - x_1) \\ y = -x + 0.5 \end{cases}$$

$$2y = 0.5 + (y_1 + x_1) \quad y = \frac{1}{2}(0.5 + y_1 - x_1)$$

$$X = \frac{1}{2} - \frac{1}{4} - \frac{1}{2}y_1 + \frac{1}{2}x_1 = +\frac{1}{4} - \frac{1}{2}y_1 + \frac{1}{2}x_1 = \frac{1}{2}(x_1 - y_1 + \frac{1}{2})$$

$$X = 0.5 - y$$

ii) Let  $f(x_1, y) = x_1^2 + 2y^2$ .  $x_1 = (1, 1)$

(a) What is the minimal value that can be achieved with one step of the GD i.e. find min of  $f(x_2)$

$$x_2 = x_1 - \gamma \nabla f(x_1) = x_1 - \gamma \cdot \begin{bmatrix} 2x_1 \\ 4y \end{bmatrix} = \begin{bmatrix} 1-\gamma 2x_1 \\ 1-\gamma 4y \end{bmatrix} = \begin{bmatrix} 1-2\gamma \\ 1-4\gamma \end{bmatrix}$$

$$f(x_2) = (1-2\gamma)^2 + 2(1-4\gamma)^2$$

$$\frac{\nabla f(x_2)}{2\gamma} = 2(1-2\gamma) \cdot (-2) + 4(1-4\gamma) \cdot (-4) = -4(1-2\gamma) - 16(1-4\gamma) = 0$$

$$-1+2\gamma = 4-16\gamma \quad 18\gamma = 5 \quad \gamma = \frac{5}{18}$$

$$f(x_2) \Big|_{\gamma=\frac{5}{18}} = \frac{2}{9}$$

b) How close can GD come to  $x^*$  in one step?

$$\nabla f = \begin{bmatrix} 2x \\ 4y \end{bmatrix} \quad \begin{array}{l} 2x=0 \\ 4y=0 \end{array} \quad (x, y) = (0, 0) = x^*$$

$$x_2 = \begin{bmatrix} 1-2\gamma \\ 1-4\gamma \end{bmatrix}$$

$$\|x_2 - x^*\| = \sqrt{(1-2\gamma)^2 + (1-4\gamma)^2}$$

Again, minimizing  $\|x_2 - x^*\|$  is equal to minimizing  $\|x_2 - x^*\|^2$

$$\|x_2 - x^*\|^2 = (1-2\gamma)^2 + (1-4\gamma)^2$$

$$\frac{2\|x_2 - x^*\|^2}{2\gamma} = 2(1-2\gamma) \cdot (-2) + 2(1-4\gamma) \cdot (-4) = -4 + 8\gamma - 8 + 32\gamma = 0$$

$$2\gamma$$

$$40\gamma - 12 = 0 \quad 40\gamma = 12 \quad \gamma = \frac{12}{40} = \frac{3}{10}$$

(given  $\|x_2 - x^*\|^2$  is quadratic, and leadin coeff. is  $> 0$  then  $\cup$  and  $\gamma = \frac{3}{10}$  is min)

$$\|x_2 - x^*\|^2 = \sqrt{(1-\frac{6}{10})^2 + (1-\frac{12}{10})^2} =$$

$$= \sqrt{\frac{16}{100} + \frac{4}{100}} = \sqrt{\frac{20}{100}} = \frac{2\sqrt{5}}{10} = \boxed{\frac{\sqrt{5}}{5}}$$