

# ASSIGNMENT 3

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## Exercise 3.1)

**a)**

Let  $C_1$  and  $C_2$  be two convex sets. Let  $[P_1, P_2]$  be the line segment whose endpoints are  $P_1$  and  $P_2$ .

$$[P_1, P_2] = \{t \times P_1 + (1 - t) \times P_2 : 0 \leq t \leq 1\}.$$

If  $C_1 \cap C_2$  contains 0 or 1 element, such intersection is convex by definitions. Otherwise,  $C_1 \cap C_2$  is convex if

$$P_1, P_2 \in C_1 \cap C_2 \Rightarrow [P_1, P_2] \subseteq C_1 \cap C_2 :$$

$$P_1, P_2 \in C_1 \cap C_2 \Rightarrow P_1, P_2 \in C_1 \text{ and also } P_1, P_2 \in C_2 \Rightarrow [P_1, P_2] \subseteq C_1 \text{ and } [P_1, P_2] \subseteq C_2 \Rightarrow [P_1, P_2] \subseteq C_1 \cap C_2$$

**b)**

Let  $C_1 = \{1\}$  and  $C_2 = \{2\}$  (both convex by definition)

$$C_1 \cup C_2 = \{1, 2\}$$

Applying the condition for convexity for  $t = 0.5$ :

$$t \times 1 + (1 - t) \times 2 \Rightarrow t + 2 - 2t \Rightarrow 2 - t \Rightarrow 1.5$$

$\{1.5\} \notin C_1 \cup C_2 \Rightarrow$  The union of two convex sets is not necessarily convex.

**c)**

Let  $C_1 = \{x \mid x \in \mathbb{R}, 1 \leq x \leq 2\}$  and  $C_2 = \{1.5\}$  (both convex sets)

$$C_1 \setminus C_2 = \{x \mid x \in \mathbb{R}, 1 \leq x \leq 2\} \setminus \{1.5\}$$

Applying the condition for convexity for  $t = 0.5$ :

$$t \times 1 + (1 - t) \times 2 \Rightarrow t + 2 - 2t \Rightarrow 2 - t \Rightarrow 1.5$$

$\{1.5\} \notin C_1 \setminus C_2 \Rightarrow$  The difference between two convex sets is not necessarily convex.

**d)**

Let's pick two arbitrary point  $x, y \in S$ . We need to prove that  $[x, y] \subseteq S$ .

Considering an arbitrary  $\lambda \in [0, 1]$  :

As  $x \in S$ ,  $\langle c, x \rangle \leq b$  holds, so does  $\langle c, \lambda x \rangle = \lambda \langle c, x \rangle \leq \lambda b$

As  $y \in S$ ,  $\langle c, y \rangle \leq b$  holds, so does  $\langle c, (1 - \lambda)y \rangle = (1 - \lambda) \langle c, y \rangle \leq (1 - \lambda)b$

Adding then the two equations together we get :

$$\langle c, \lambda x \rangle + \langle c, (1 - \lambda)y \rangle \leq \lambda b + (1 - \lambda)b = b$$

This proves that  $\lambda x + (1 - \lambda)y \in S \quad \forall \lambda \in [0, 1] \Rightarrow$  The whole segment  $[x, y] \in S$ .

$[x, y] \subseteq S \Rightarrow S$  is convex.

e)

Let  $x, y$  be two vectors in the set,  $x, y \in S$ , and  $\lambda$  an arbitrary number,  $\lambda \in [0, 1]$ .

As  $x \in S$ ,  $Ax \leq b$  holds, so does  $A(\lambda x) = \lambda \cdot Ax \leq \lambda b$

As  $y \in S$ ,  $Ay \leq b$  holds, so does  $A[(1 - \lambda)y] = (1 - \lambda) \cdot Ay \leq (1 - \lambda)b$

Adding the two equations :

$$A(\lambda x) + A[(1 - \lambda)y] \leq \lambda b + (1 - \lambda)b = b$$

Similarly to the point  $d$ ), this proves that  $\lambda x + (1 - \lambda)y \in S \quad \forall \lambda \in [0, 1] \Rightarrow$  The whole segment  $[x, y] \in S$

The set  $S$  is a convex subset of  $\mathbb{R}^n$ .

## Exercise 3.2)

a)

Given  $k \in \{1, \dots, n\}$ ,  $\theta \in [0, 1]$ , and considering arbitrary  $x, y \in \mathbb{R}^n$ . It is possible to apply the definition of convexity for a function:

$$f(\theta x_k + (1 - \theta)y_k) \leq \theta f(x_k) + (1 - \theta)f(y_k)$$

$$\min_k[\theta x_k + (1 - \theta)y_k] \leq \theta \cdot \min_i(x_i) + (1 - \theta) \cdot \min_i(y_i)$$

This statement does not hold for all  $k$ , it is possible to easily prove that by an example:

Let  $x, y \in \mathbb{R}^3$  with  $x = \{25, 50, 5\}$  and  $y = \{10, 20, 12\}$

Applying the function for  $\theta = 0.5$ :

$$\min_k[\theta x_k + (1 - \theta)y_k] = \min[0.5 \cdot 5 + 0.5 \cdot 12] = 8.5$$

$$\theta \cdot \min_i(x_i) + (1 - \theta) \cdot \min_i(y_i) = 0.5 \cdot 5 + 0.5 \cdot 10 = 7.5$$

This proves that the function is not convex.

**b)**

Given  $k \in \{1, \dots, n\}$ ,  $\theta \in [0, 1]$ , and considering arbitrary  $x, y \in \mathfrak{R}^n$ . It is possible to apply the definition of convexity for a function:

$$f(\theta x_k + (1 - \theta)y_k) \leq \theta f(x_i) + (1 - \theta)f(y_i)$$

$$(1) \quad \max_k[\theta x_k + (1 - \theta)y_k] \leq \theta \cdot \max_i(x_i) + (1 - \theta) \cdot \max_i(y_i)$$

As  $x_k \leq \max_i(x_i)$  and  $y_k \leq \max_i(y_i) \quad \forall k, i$

it is possible to conclude that the equation (1) is true for all k, the function is thus convex.

**c)**

$\sqrt{x}$  is differentiable for  $x > 0$ , but not for  $x = 0$ , because the  $\lim_{x \rightarrow 0^-} \sqrt{x}$  does not exist as the function is not defined for  $x < 0$ .

$f'(x) = \frac{1}{2} x^{\left(\frac{1}{2}-1\right)} = \frac{1}{2\sqrt{x}}$ , whose domain is  $(0, +\infty)$ , again,  $f'(x)$  is not defined for  $x < 0$ , but in this case  $x \neq 0$  must hold as well.

$$f''(x) = \frac{1}{2} x^{\left(\frac{1}{2}-1\right)} \times \left(-\frac{1}{2}\right) = -\frac{1}{4\sqrt{x^3}}$$

The function is not convex as  $f''(x) < 0 \quad \forall x \in \text{dom}(f''(x))$ .

**d)**

The function  $f(x) = e^{ax}$  is differentiable  $\forall x \in \mathfrak{R}$ .

$$f'(x) = a \cdot e^{ax}$$

$$f''(x) = a^2 \cdot e^{ax}$$

$$f''(x) \geq 0 \quad \forall x \in \mathfrak{R} \Rightarrow f(x) \text{ is convex.}$$

**e)**

$f(x)$  is differentiable  $\forall x \in \mathfrak{R}$

$$f'(x) = x^2 - 1 \Rightarrow f'(x) = 0 \Rightarrow x = \pm 1$$

$$f''(x) = 2x \Rightarrow \text{positive only for } x \geq 0$$

$$f''(x) \geq 0 \quad \forall x \geq 0 \Rightarrow f(x) \text{ is convex only for } x \geq 0.$$

$f(x)$  is NOT a convex function.

## Exercise 3.3)

Definitions:

- $S$  is the set of scenarios
- $C$  is the set of crops
- $F$  is the set of fields
- $b_{f,c}$  is a binary variable indicating whether crop  $c \in C$  is planted in the field  $f \in F$ .
- $x_c$  indicates the amount of crop  $c \in C$  planted.
- $z_{f,c}$  is an auxiliary matrix used to linearize the problem, it contains the number of acres of each crop  $c \in C$  planted in a field  $f \in F$
- $c_c$  indicates the planting costs of crop  $c \in C$ .
- $F_{\max_f}$  indicates the maximum amount of acres available in field  $f \in F$ .
- $w^s_w$  and  $w^s_{corn}$  indicate respectively the amount of wheat and of corn sold,  $w, corn \in C$  in scenario  $s \in S$
- $e^s_H, e^s_L$  indicate respectively the amount of sugar beet sold at high/low price in scenario  $s \in S$
- $y^s_w$  and  $y^s_{corn}$  indicate the amounts of wheat/corn purchased in scenario  $s \in S$
- $p^s_c$  indicate the yields for each crop  $c \in C$  in each scenario

$$\max_{x,b,y,w,e} \sum_{i=1}^3 \left( \frac{(170w^i_W + 150w^i_c + 36e^i_H + 10e^i_L - 238y^i_w - 210y^i_c)}{3} \right) - \left( c^T \cdot \sum_{f \in F} z_{f,c} \right)$$

$s, t.$

$$\sum_{c \in C} b_{f,c} \leq 1 \quad \forall f \in F$$

$$F_i \leq F_{\max_i} \quad \forall i \in F$$

$$z_{i,j} \leq F_{\max_i} \cdot b_{i,j} \quad \forall i \in F, \forall j \in C$$

$$z_{i,j} \geq F_i - F_{\max_i} \cdot (1 - b_{i,j}) \quad \forall i \in F, \forall j \in C$$

$$p^s_w \cdot \sum_{i \in F} z_{i,w} + y^s_w - w^s_w = 200 \quad \forall s \in S$$

$$p^s_{corn} \cdot \sum_{i \in F} z_{i,corn} + y^s_{corn} - w^s_{corn} = 240 \quad \forall s \in S$$

$$p^s_{sb} \cdot \sum_{i \in F} z_{i,sb} - e^s_H - e^s_L = 0 \quad \forall s \in S$$

$$e^i_H \leq 6000 \quad \forall s \in S$$

$$w^i_c, y^i_c \geq 0 \quad \forall i \in S, \forall c \in C$$

$$e^i_L, e^i_H \geq 0 \quad \forall i \in S$$

$$x_c \geq 0 \quad \forall c \in C$$

$$F_f \geq 0 \quad \forall f \in F$$

$$z_{i,j} \geq 0 \quad \forall i \in F, \forall j \in C$$

### Exercise 3.4)

a)

The optimal solution for the problem depends on the value that the random variable  $\omega_1$  assumes.

In particular, in order to minimize the sum of  $x_1$  and  $x_2$  while always satisfying the constraint, we define a stepwise function depending on  $\omega_1$ . If  $\omega_1$  is between 0 and 1, it will be optimal to set  $x_2 = 4$  and  $x_1 = 0$ , in order to satisfy the constraint by equaling it and at the same time minimising the sum of  $x_1$  and  $x_2$ .

If  $1 \leq \omega_1 \leq 3$ , the optimal solution would be to divide 4 by the value of  $\omega_1$ , so that the constraint is still satisfied, by equaling it, and the objective function assumes values that decrease with increasing  $\omega_1$ .

The optimal solution is:

$$x_1^* = \begin{cases} 0 & \text{if } 0 \leq \omega_1 < 1 \\ \frac{4}{\omega_1} & \text{if } 1 \leq \omega_1 \leq 3 \end{cases}$$

$$x_2^* = \begin{cases} 4 & \text{if } 0 \leq \omega_1 < 1 \\ 0 & \text{if } 1 \leq \omega_1 \leq 3 \end{cases}$$

$$v^* = \begin{cases} 4 & \text{if } 0 \leq \omega_1 < 1 \\ \frac{4}{\omega_1} & \text{if } 1 \leq \omega_1 \leq 3 \end{cases}$$

b)

Calculate the CDF of  $v^*$ ,  $F_{v^*}$ :

The range of the function  $v^*$  is  $R_{v^*} = \left[\frac{4}{3}, 4\right]$

From  $R_{v^*}$  it is possible to say that:

$$F_{v^*}(v^*) = P(V^* \leq v^*) = 1 \quad \text{if } v^* \geq 4$$

$$F_{v^*}(v^*) = P(V^* \leq v^*) = 0 \quad \text{if } v^* < \frac{4}{3}$$

With values in the range of the function, the CDF can be derived as:

$$\begin{aligned}
F_{v^*}(v^*) &= P(V^* \leq v^*) \quad \text{if } \frac{4}{3} \leq v^* < 4 \\
&= P\left(\frac{4}{\omega_1} \leq v^*\right) = P\left(\omega_1 \geq \frac{4}{v^*}\right) = \\
&= 1 - P\left(\omega_1 \leq \frac{4}{v^*}\right) \\
&= 1 - F_{\omega_1}\left(\frac{4}{v^*}\right) \\
&= 1 - \int_0^{\frac{4}{v^*}} \frac{1}{3} dt = 1 - \left[\frac{x}{3}\right]_0^{\frac{4}{v^*}} = \\
&1 - \left(\frac{4}{3v^*} - 0\right) = 1 - \frac{4}{3v^*}
\end{aligned}$$

Then,  $F_{v^*}$  is equal to:

$$F_{v^*} = \begin{cases} 0 & \text{if } v^* < \frac{4}{3} \\ 1 - \frac{4}{3v^*} & \text{if } \frac{4}{3} \leq v^* < 4 \\ 1 & v^* \geq 4 \end{cases}$$

c)

The PDF of  $v^*$  is equal to:

$$\begin{aligned}
f_{v^*}(v^*) &= \frac{d}{dv^*} F_{v^*}(v^*) = \\
&= \begin{cases} 0 & \text{if } v^* < \frac{4}{3} \\ \frac{4}{3v^{*2}} & \text{if } \frac{4}{3} \leq v^* < 4 \\ 0 & \text{if } v^* \geq 4 \end{cases}
\end{aligned}$$

Calculate the expected value  $E(v^*)$  :

$$E(v^*) = \int_{-\infty}^{\infty} v^* f_{v^*}(v^*) \, dv^* =$$

$$\int_{\frac{4}{3}}^4 v^* \frac{4}{3v^{*2}} \, dv^* =$$

$$\frac{4}{3} \int_{\frac{4}{3}}^4 \frac{1}{v^*} \, dv^* =$$

$$\frac{4}{3} \ln(v^*) \Big|_{\frac{4}{3}}^4 =$$

$$\frac{4}{3} \left( \ln(4) - \ln\left(\frac{4}{3}\right) \right) =$$

$$\frac{4}{3} \ln(3) = 1.46482$$