



## Tutorial 2 : Convex sets and functions

**Exercise 1** (Unproven proposition). If  $A$  is a convex set, then any convex combination of finite points of  $A$  is an element of  $A$ , i.e., for  $x_1, \dots, x_k \in A$ ,  $\alpha_1, \dots, \alpha_k \geq 0$ ,  $\sum_i \alpha_i = 1$ , we have:

$$\alpha_1 x_1 + \dots + \alpha_k x_k \in A.$$

**Exercise 2** (Convex sets). Decide whether the following sets are convex and justify your choice:

1. A slab, i.e., a set of the form:  $\{x \in \mathbb{R}^n \mid \alpha \leq x^\top a \leq \beta\}$  ( $\alpha, \beta, a$  are fixed scalars and vector).
2. A rectangle, i.e., a set of the form:  $\{x \in \mathbb{R}^d \mid \alpha_i \leq x_i \leq \beta_i\}$ .
3. A wedge, i.e.,  $\{x \in \mathbb{R}^d \mid a_1^\top x \leq b_1, a_2^\top x \leq b_2\}$ .
4. The set of points closer to a given point than a given set, i.e.,

$$\{x \in \mathbb{R}^d \mid \|x - x_0\|_2 \leq \|x - y\|_2, \forall y \in S\}$$

5. The set of points closer to one set than another, i.e.,

$$\{x \in \mathbb{R}^d \mid \text{dist}(x, S) \leq \text{dist}(x, T)\}$$

where  $\text{dist}(x, S) = \inf_{y \in S} \|x - y\|_2$ .

6. The set  $\{x \in \mathbb{R}^d \mid x + S_2 \subseteq S_1\}$  where  $S_1, S_2 \subseteq \mathbb{R}^d$  and  $S_1$  is convex.
7. The set of points whose distance to  $a$  does not exceed a fixed fraction  $\theta$  of the distance to  $b$ , i.e.,  $\{x \in \mathbb{R}^d \mid \|x - a\|_2 \leq \theta \|x - b\|_2\}$ . You can assume that  $a \neq b$  and  $\theta \in [0, 1]$ .

**Exercise 3** (Operations preserving convexity). Prove that the following operations preserve the convexity:

1. If  $f$  is convex, then  $\alpha f$  is also convex ( $\alpha > 0$ ).
2. If  $f$  and  $g$  are convex, then  $f + g$  is also convex.
3. If  $f_i, i = 1, \dots, n$  are convex and  $\alpha_i \geq 0, \forall i = 1, \dots, n$ , then  $\sum_i \alpha_i f_i$  is also convex.
4. If  $f, g$  are convex, then  $\max(f, g)$  is convex.
5. If  $f$  is convex, then  $g(x) = f(Ax + b)$  is also convex.

6. if  $f : \mathbb{R}^d \rightarrow \mathbb{R}$  is convex and  $g : \mathbb{R} \rightarrow \mathbb{R}$  is non-decreasing and convex, then  $g \circ f$  is convex.

*Proof.* To be done. □

**Exercise 4** (Sublevel sets of convex functions). Given a function  $f : \mathbb{R}^d \rightarrow \mathbb{R}$ , the  $\alpha$ , the sublevel sets of  $f$  is given by:

$$\{x \in \mathbb{R}^d \mid f(x) \leq c\}.$$

1. Prove that the sublevel set of a convex function is convex.
2. Is the converse correct, i.e., if all the sublevels of  $f$  are convex, is  $f$  necessarily a convex function?
3. Applying this result to show that the ellipsoid, i.e.,  $\{x \in \mathbb{R}^d \mid (x - y)^\top \mathbf{A}(x - y) \leq c\}$  for some  $\mathbf{A} \in \mathbb{R}^{d \times d}$ ,  $\mathbf{A} \succeq 0$ ,  $y \in \mathbb{R}^d$ ,  $c \in \mathbb{R}$ .

**Exercise 5** (Optimal condition for constrained convex optimization). Consider a constrained optimization problem given by:

$$\underset{x \in \mathcal{F}}{\text{Minimize}} \quad f(x)$$

where  $f$  is  $C^1$  and convex,  $\mathcal{F}$  is a convex set. A point  $x$  is a global solution of  $f(x)$  if and only if  $\nabla f(x)^\top (y - x) \geq 0, \forall y \in \mathcal{F}$ .

**Exercise 6** (Projection onto a convex set). In this exercise, we will prove for any non-empty closed convex set  $S \subseteq \mathbb{R}^d$ ,  $x \in \mathbb{R}^d$ , the set of projector of  $x$  onto  $S$  is a singleton, i.e.,  $\{y \in S \mid \|x - y\|_2 = \text{dist}(x, S)\}$ . To do so, answer the following questions:

1. Prove that the set of projectors of  $x$  onto  $S$  is non-empty.
2. Let  $y$  be a projector of  $x$  onto  $S$ , prove that:

$$(y - x)^\top (z - y) \geq 0, \forall z \in S.$$

Hint: Use the previous exercise.

3. Prove that the set of projectors has at most one element by contradiction.

*Proof.* Proof is left as exercise. □

**Exercise 7** (Separating plane of disjoint convex sets). In this exercise, we will prove the following results: For any two non-empty, convex sets  $C, D \subseteq \mathbb{R}^d$  such that  $C \cap D = \emptyset$ , there exists a separating hyperplane, i.e., a vector  $a \in \mathbb{R}^d, b \in \mathbb{R}$  satisfying:

$$\begin{aligned} x \in C &\implies x^\top a \leq b, \\ x \in D &\implies x^\top a \geq b. \end{aligned}$$

Prove a simpler version of this result where we assume that there exists  $c \in C, d \in D$  such that  $\|x - y\|_2 = \text{dist}(C, D)$  by following these steps:

1. Define  $a = d - c$ ,  $b = \frac{\|d\|^2 - \|c\|^2}{2}$ . Prove that:

$$f(x) = x^\top a - b = (d - c)^\top \left( x - \frac{1}{2}(d + c) \right) = \frac{1}{2} \|d - c\|^2 + (d - c)^\top (x - d).$$

2. Prove that for  $x \in D$ , we have:  $(d - c)^\top (x - d) \geq 0$ .
3. Conclude that for  $x \in D$ ,  $x^\top a \geq b$ . Conclude the proof by making the same argument can be used to prove that  $x \in C \implies x^\top a \leq b$ .