

## 1. Exercise on Convex Optimization

### Problem 1: (Properties of Convex Sets)

Show that:

- a) The intersection  $\cap_{i \in I} C_i$  of any collection  $\{C_i \mid i \in I\}$  of convex sets is convex.
- b) The vector sum  $C_1 + C_2$  of two convex sets  $C_1$  and  $C_2$  is convex.
- c) The set  $\lambda C$  is convex for any convex set  $C$  and scalar  $\lambda$ . Furthermore, if  $C$  is a convex set and  $\lambda_1, \lambda_2$  are positive scalars,

$$(\lambda_1 + \lambda_2)C = \lambda_1 C + \lambda_2 C.$$

Show by example that this need not be true when  $C$  is not convex.

- d) The closure and the interior of a convex set are convex.
- e) The image and the inverse image of a convex set under an affine function are convex.

### Problem 2: (Properties of Cones)

Show that:

- a) The intersection  $\cap_{i \in I} C_i$  of a collection  $\{C_i \mid i \in I\}$  of cones is a cone.
- b) The Cartesian product  $C_1 \times C_2$  of two cones  $C_1$  and  $C_2$  is a cone.
- c) The vector sum  $C_1 + C_2$  of two cones  $C_1$  and  $C_2$  is a cone.
- d) The closure of a cone is a cone.
- e) The image and the inverse image of a cone under a linear transformation is a cone.

### Problem 3: (Convexity under Composition)

Let  $C \subset \mathbb{R}^n$  nonempty convex. Let  $f = (f_1, \dots, f_m)$ , where each  $f_i : C \rightarrow \mathbb{R}$  is a convex function, and let  $g : \mathbb{R}^m \rightarrow \mathbb{R}$  be a function that is convex and monotonically non-decreasing over a convex set  $D \supset \{f(x) \mid x \in C\}$ . Show that the function  $h$  defined by  $h(x) = g[f(x)]$  is convex over  $C \times \dots \times C$ .

### Problem 4: (Examples of Convex and Concave Functions)

Show that the following functions from  $\mathbb{R}^n$  to  $(-\infty, \infty]$  are convex:

- a)  $f_1(x) := \|x\|^p$  with  $p \geq 1$ .
- b)  $f_2(x) := 1/f(x)$ , where  $f$  is concave and  $f(x)$  is a positive number for all  $x$ .
- c)  $f_3(x) := \alpha f(x) + \beta$ , where  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is a convex function, and  $\alpha, \beta$  are scalars such that  $\alpha \geq 0$ .
- d)  $f_4(x) := e^{\beta x' A x}$ , where  $A \in \mathbb{R}^{n \times n}$  is symmetric and positive semidefinite and  $\beta > 0$ .
- e)  $f_5(x) := \|Ax - b\|_2^2$ , where  $A \in \mathbb{R}^{m \times n}$  and  $b \in \mathbb{R}^m$ .
- f)  $f_6^*(y) := \sup_{x \in \text{dom } f} [y'x - f(x)]$ , where  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ .

Moreover show that the following function from  $\mathbb{R}^{n \times n}$  to  $(-\infty, \infty]$  is concave:

- g)  $f_7(X) := \log \det(X)$ , where  $X \in \mathbb{R}^{n \times n}$  is symmetric and positive definite.