

SI251 - Convex Optimization, 2024 Spring

Homework 1

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1 Convex sets

1. Please prove that the following sets are convex:

- 1) $S = \{x \in \mathbf{R}^m \mid |p(t)| \leq 1 \text{ for } |t| \leq \pi/3\}$, where $p(t) = x_1 \cos t + x_2 \cos 2t + \cdots + x_m \cos mt$. (5 pts)
- 2) (Ellipsoids) $\left\{x \mid \sqrt{(x - x_c)^T P (x - x_c)} \leq r\right\}$ ($x_c \in \mathbb{R}^n, r \in \mathbb{R}, P \succeq 0$). (5 pts)
- 3) (Symmetric positive semidefinite matrices) $S_+^{n \times n} = \left\{P \in S^{n \times n} \mid P \succeq 0\right\}$. (5 pts)
- 4) The set of points closer to a given point than a given set, i.e.,

$$\left\{x \mid \|x - x_0\|_2 \leq \|x - y\|_2 \text{ for all } y \in S\right\},$$

where $S \in \mathbf{R}^n$. (5 pts)

(1)

(2)

(3)

(4)

2. (15 pts) For a given norm $\|\cdot\|$ on \mathbf{R}^n , the dual norm, denoted $\|\cdot\|_*$, is defined as

$$\|y\|_* = \sup_{x \in \mathbf{R}^n} \{y^T x \mid \|x\| \leq 1\}.$$

Show that the dual of Euclidean norm is the Euclidean norm, i.e., $\sup_{x \in \mathbf{R}^n} \{z^T x \mid \|x\|_2 \leq 1\} = \|z\|_2$.

3. (15 pts) Define a norm cone as

$$\mathcal{C} \equiv \{(x, t) : x \in \mathbb{R}^d, t \geq 0, \|x\| \leq t\} \subseteq \mathbb{R}^{d+1}$$

Show that the norm cone is convex by using the definition of convex sets.

2 Convex functions

4. (18 pts) Let $C \subset \mathbb{R}^n$ be convex and $f : C \rightarrow \mathbb{R}^*$. Show that the following statements are equivalent:

(a) $\text{epi}(f)$ is convex.

(b) For all points $x_i \in C$ and $\{\lambda_i | \lambda_i \geq 0, \sum_{i=1}^n \lambda_i = 1, i = 1, 2, \dots, n\}$, we have

$$f\left(\sum_{i=1}^n \lambda_i x_i\right) \leq \sum_{i=1}^n \lambda_i f(x_i).$$

(c) For $\forall x, y \in C$ and $\lambda \in [0, 1]$,

$$f\left((1 - \lambda)x + \lambda y\right) \leq (1 - \lambda)f(x) + \lambda f(y).$$

(a)

(b)

(c)

5. (14 pts) Monotone Mappings. A function $\psi : \mathbf{R}^n \rightarrow \mathbf{R}^n$ is called monotone if for all $x, y \in \text{dom}\psi$,

$$(\psi(x) - \psi(y))^T(x - y) \geq 0$$

Suppose $f : \mathbf{R}^n \rightarrow \mathbf{R}$ is a differentiable convex function. Show that its gradient ∇f is monotone. Is the converse true, i.e., is every monotone mapping the gradient of a convex function?

6. (18 pts) Please determine whether the following functions are convex, concave or none of those, and give a detailed explanation for your choice.

1)

$$f_1(x_1, x_2, \dots, x_n) = \begin{cases} -(x_1 x_2 \cdots x_n)^{\frac{1}{n}}, & \text{if } x_1, \dots, x_n > 0 \\ \infty & \text{otherwise;} \end{cases}$$

2) $f_2(x_1, x_2) = x_1^\alpha x_2^{1-\alpha}$, where $0 \leq \alpha \leq 1$, on \mathbb{R}_{++}^2 ;

3) $f_3(x, u, v) = -\log(uv - x^T x)$ on $\text{dom } f = \{(x, u, v) | uv > x^T x, \ u, v > 0\}$.

(1)

(2)

(3)