

Convex Optimization HW 1

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1. Assume that C is an affine set. By definition, we know that for any $x_1, x_2 \in C$, we have

$$\theta x_1 + (1 - \theta)x_2 \in C, \text{ for all } \theta \in \mathbb{R}.$$

Building upon the definition, show that if $x_i \in C$ for $i = 1, \dots, n$, then we have

$$\theta_1 x_1 + \dots + \theta_n x_n,$$

where $\sum_{i=1}^n \theta_i = 1$.

Proof. Given an affine set C , let $x_0 \in C$ be arbitrary, and recall that $V = C - x_0$ is a subspace. Then for all $x_i \in C$ and given $\sum_{i=1}^n \theta_i = 1$ for $i = 1, \dots, n$. Observe that

$$\begin{aligned} \sum_{i=1}^n \theta_i (x_i - x_0) &\in V \\ \sum_{i=1}^n \theta_i (x_i - x_0) + x_0 &\in C \\ \sum_{i=1}^n \theta_i x_i - \sum_{i=1}^n \theta_i x_0 + x_0 &\in C \\ \sum_{i=1}^n \theta_i x_i - x_0 \sum_{i=1}^n \theta_i + x_0 &\in C \\ \sum_{i=1}^n \theta_i x_i - x_0 \cdot 1 + x_0 &\in C \\ \sum_{i=1}^n \theta_i x_i &\in C. \end{aligned}$$

□

2. Answer the following questions:

- (a) What is the distance between two parallel hyperplanes. i.e., $\{x|a^T x = b\}$ and $\{x|a^T x = c\}$?
- (b) Let a, b be distinct points in \mathbb{R}^n . Show that the set of all points that are closer to a than b , i.e., $\{x| \|x - a\|_2 \leq \|x - b\|_2\}$, is a halfspace. Describe it explicitly as an inequality of the form $c^T x \leq d$. Draw a picture.

3. Which of the following sets are convex?

- (a) A slab $\{x \in \mathbb{R}^n | \alpha \leq a^T x \leq \beta\}$.
- (b) A rectangle $\{x \in \mathbb{R}^n | \alpha_i \leq x_i \leq \beta_i, i = 1, \dots, n\}$.
- (c) The set of points closer to a given point than a given set:

$$\{x | \|x - x_0\|_2 \leq x - y_2 \text{ for all } y \in S\}$$

where $S \subset \mathbb{R}^n$.

- (d) The set of points whose distance to a does not exceed a fixed fraction θ of the distance to b , i.e. the set $\{x | \|x - a\|_2 \leq \theta \|x - b\|_2\}$. You can assume $a \neq b$ and $\theta \leq 1$.

4. Show the following statements.

- (a) A polyhedron, i.e. $P = \{x | Ax \succeq b, Cx = d\}$ where $A \in \mathbb{R}^{m \times n}$ and $C \in \mathbb{R}^{p \times n}$ is a convex set.
- (b) Consider an ellipsoid $\epsilon = \{x | (x - x_c)^T P^{-1} (x - x_c) \leq 1\}$. Assume that the eigenvalues of $P \in \mathbb{R}^{n \times n}$ is $\lambda_1^2, \dots, \lambda_n^2$ in descending order. Show that the largest and smallest distances from any point on the boundary of the ellipsoid to x_c are λ_1 and λ_n respectively.

5. Show the following statements.

- (a) In machine learning, we are often given training samples in the form of (x_i, y_i) for $i = 1, \dots, n$ where $x_i \in \mathbb{R}^d$ is the feature vector and $y_i \in \mathbb{R}$ is the label of this example. The empirical risk of Euclidean distance based linear regression can be expressed as follows:

$$f(a) = \frac{1}{n} \sum_{i=1}^n (y_i - a^T x_i)^2.$$

Show the function $f(a)$ is convex in a .

- (b) Suppose $p < 1, p \neq 0$. Show that the function

$$f(x) = \left(\sum_{i=1}^n x_i^p \right)^{1/p}$$

with $\text{dom}(f) = \mathbb{R}_{++}^n$ is concave.

- (c) Show that $f(X) = \text{tr}(X^{-1})$ is convex on $\text{dom}(f) = \mathbb{S}_{++}^n$.

6. Show the conjugate of $f(X) = \text{tr}(X^{-1})$ with $\text{dom}(f) = \mathbb{S}_{++}^n$ is given by

$$f^*(Y) = -2\text{tr}(-Y)^{1/2}, \text{dom}(f) = -\mathbb{S}_{++}^n.$$

(Hint: for unconstrained and differentiable convex problems, \min / \max can be found by looking for where the function has zero gradient.)

7. Show that the following function is convex.

$$f(x) = x^T (A(x))^{-1} x, \text{dom}(f) = \{x | A(x) \succ 0\},$$

where $A(x) = A_0 + A_1 x_1 + \dots + A_n x_n \in \mathbb{S}^n$ and $A_i \in \mathbb{S}^n, i = 1, \dots, n$. Hint: you are allowed to use a special form of Schur complement, described as follows: Suppose $A \succ 0$. then

$$\begin{pmatrix} A & b \\ b^T & c \end{pmatrix} \succ 0 \Leftrightarrow c - b^T A^{-1} b \geq 0.$$

You will need to study "epigraph" from chapter 3 of the textbook to answer this question.