Optimization Methods

Fall 2023

Homework 1

Instructor: Lijun Zhang

Name: Ji Qiankun, StudentId: 221300066

Problem 1: Norms

A function $f: \mathbb{R}^n \to \mathbb{R}$ with dom $f = \mathbb{R}^n$ is called a norm if

• f is nonnegative: $f(x) \geq 0$ for all $x \in \mathbb{R}^n$

• f is definite: f(x) = 0 only if x = 0

• f is homogeneous: f(tx) = |t| f(x), for all $x \in \mathbb{R}^n$ and $t \in \mathbb{R}$

• f satisfies the triangle inequality: $f(x+y) \leq f(x) + f(y)$, for all $x, y \in \mathbb{R}^n$

We use the notation f(x) = ||x||. Let $||\cdot||$ be a norm on \mathbb{R}^n . The associated dual norm, denoted $||\cdot||_*$, is defined as

$$||z||_* = \sup \{z^{\mathrm{T}}x \mid ||x|| \le 1\}$$

a) Prove that $\|\cdot\|_*$ is a valid norm.

b) Prove that the dual of the Euclidean norm (ℓ_2 -norm) is the Euclidean norm, i.e., prove that

$$||z||_{2*} = \sup \{z^T x \mid ||x||_2 \le 1\} = ||z||_2$$

(Hint: Use Cauchy-Schwarz inequality.)

Solution. a) To prove that $\|\cdot\|_*$ is a valid norm, need to show that it satisfies the four properties of a norm:

- 1. Non-negativity: By definition, $||z||_* = \sup_{||x|| \le 1} z^T x$. Since the supremum of a set of real numbers is always greater than or equal to any element of the set, we have $z^T x \le ||z||_*$ for all x with $||x|| \le 1$. In particular, setting x = 0, we have $z^T 0 = 0 \le ||z||_*$, so $||\cdot||_*$ is non-negative.
- 2. Definiteness: If z = 0, then clearly $\|\cdot\|_*(z) = 0$. Conversely, suppose that $\|\cdot\|_*(z) = 0$. Then by definition, $z^Tx \leq 0$ for all x with $\|x\| \leq 1$. In particular, setting $x = z/\|z\|$ (which has norm $\|x\| = 1$), we have $z^T(z/\|z\|) = \|z\|^2/\|z\| = \|z\|$ (since $\|\cdot\|$ is a norm), so $\|z\| = 0$, which implies that z = 0.
 - 3. Homogeneity: For any scalar α , we have

$$\|\alpha z\|_* = \sup_{\|x\| \le 1} (\alpha z)^{\mathrm{T}} x = \sup_{\|y\| \le |\alpha|} z^{\mathrm{T}}(y/\alpha) = |\alpha| \sup_{\|y/\alpha\| \le 1} z^{\mathrm{T}}(y/\alpha) = |\alpha| \| \cdot \|_*(z).$$

4. Triangle inequality: For any vectors z, w, we have

$$||z+w||_* = \sup_{\|x\| \le 1} (z+w)^{\mathrm{T}} x = \sup_{\substack{\|y_1+y_2\| \le 1 \\ y_1, y_2}} z^{\mathrm{T}} y_1 + w^{\mathrm{T}} y_2 \le \sup_{y_1} \left(z^{\mathrm{T}} y_1 + \sup_{y_2} \left(w^{\mathrm{T}} y_2 : ||y_2|| \le 1 - ||y_1|| \right) \right) = ||z||_* + ||w||_*.$$

Therefore, it is shown that $\|\cdot\|_*$ is therefore a valid norm.

b) To prove that the dual of the Euclidean norm is the Euclidean norm itself, need to show that

$$||z||_{2*} = \sup \{z^T x \mid ||x||_2 \le 1\} = ||z||_2.$$

By Cauchy-Schwarz inequality, we have

$$|z^T x| \le ||z||_2 ||x||_2 \le ||z||_2,$$

where the last inequality follows since ||x|| = 1. Therefore, we have

$$||z||_{2*} = \sup \{z^T x \mid ||x||_2 \le 1\}$$

Problem 2: Inequalities

Let $x \in \mathbb{R}^n$, $y \in \mathbb{R}^n$, where n is a positive integer. Let $\|\cdot\|$ denote the Euclidean norm.

- a) Prove the triangle inequality $||x + y|| \le ||x|| + ||y||$.
- b) Prove $||x+y||^2 \le (1+\epsilon)||x||^2 + (1+\frac{1}{\epsilon})||y||^2$ for any $\epsilon > 0$.

(*Hint*: You may need the Young's inequality for products, i.e. if a and b are nonnegative real numbers and p and q are real numbers greater than 1 such that 1/p + 1/q = 1, then $ab \le \frac{a^p}{p} + \frac{b^q}{q}$.)

Solution. a) First square both sides and then expand:

$$||x + y||^2 = (x + y)^T (x + y)$$
$$= x^T x + 2x^T y + y^T y$$
$$= ||x||^2 + 2x^T y + ||y||^2.$$

Since $x^T y \leq ||x|| ||y||$ by the Cauchy-Schwarz inequality, we have

$$||x + y||^2 \le ||x||^2 + 2||x||||y|| + ||y||^2 = (||x|| + ||y||)^2.$$

Taking the square root of both sides, then get the triangle inequality $||x+y|| \le ||x|| + ||y||$ b) Using Young's inequality. Let $a = \sqrt{\epsilon}||x||$ and $b = \frac{1}{\sqrt{\epsilon}}||y||$ in Young's inequality, then we have

$$ab \le \frac{a^2}{2} + \frac{b^2}{2},$$

which is equivalent to

$$\sqrt{\epsilon} \|x\| \frac{1}{\sqrt{\epsilon}} \|y\| \le \frac{\epsilon}{2} \|x\|^2 + \frac{1}{2} \|y\|^2.$$

Rearranging terms gives us

$$(1+\epsilon)\|x\|^2 + (1+\frac{1}{\epsilon})\|y\|^2 - 2\|x\|\|y\| \ge 0,$$

Since $||x + y||^2 = ||x||^2 + 2||x|| ||y|| + ||y||^2$, it is equivalent to the desired inequality $||x + y||^2 \le (1 + \epsilon)||x||^2 + (1 + \frac{1}{\epsilon})||y||^2$ for any $\epsilon > 0$.

Problem 3: Definition of convexity

Which of the following sets are convex? Please provide explanations for your choices.

- a) a set of the form $\{x \in \mathbf{R}^n \mid \alpha \leq a^T x \leq \beta\}$.
- b) a set of the form $\{x \in \mathbf{R}^n \mid \alpha_i \leq a^T x_i \leq \beta_i, i = 1, 2, ..., n\}$.
- c) a set of the form $\{x \in \mathbf{R}^n \mid a_1^T x \leq b_1, a_2^T x \leq b_2\}.$
- d) The set of points closer to a given point than a given set, i.e.,

$$\{x \mid ||x - x_0||_2 \le ||x - y||_2 \text{ for all } y \in S\}$$

where $S \subseteq \mathbf{R}^n$.

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

$$\{x \mid \mathbf{dist}(x, S) \leq \mathbf{dist}(x, T)\},\$$

where $S, T \subseteq \mathbf{R}^n$, and

$$dist(x, S) = \inf \{ ||x - z||_2 \mid z \in S \}.$$

f) The set of points whose distance to a does not exceed a fixed fraction θ of the distance to b, i.e., the set $\{x \mid \|x-a\|_2 \leq \theta \|x-b\|_2\}$. You can assume $a \neq b$ and $0 \leq \theta \leq 1$.

Solution. a) The set is convex. This is because it represents a hyperplane in \mathbb{R}^n , and all points between any two points in the set also belong to the set.

- b) The set is not necessarily convex. This is because each constraint applies to a different dimension of x, and there may exist two points in the set such that the line segment connecting them does not entirely belong to the set.
- c) The set is convex. This is because it represents the intersection of two half-spaces, which are both convex sets.
 - d) The set is convex. This is known as a Voronoi cell, which is always convex.
- e) The set is convex. This is because it represents the region of space where every point is closer to the set S than to the set T, which forms a convex region.
- f) The set is not convex when $\theta = 1$ and $a \neq b$. This is because it forms an ellipse with foci at a and b, but excludes the segment between a and b. However, for other values of θ , it forms a convex region.

Problem 4: Examples

Let $C \subseteq \mathbb{R}^n$ be the solution set of a quadratic inequality,

$$C = \{x \in \mathbb{R}^n | x^T A x + b^T x + c \leqslant 0\}$$

with $A \in \mathbb{S}^n$, $b \in \mathbb{R}^n$, and $c \in \mathbb{R}$.

- a) Show that C is convex if $A \succeq 0$.
- b) Is the following statement true? The intersection of C and the hyperplane defined by $g^Tx + h = 0$ is convex if $A + \lambda gg^T \succeq 0$ for some $\lambda \in \mathbb{R}$.

Solution. a) The function $f(x) = x^T A x + b^T x + c$ is a convex function when A is positive semi-definite (i.e., $A \succeq 0$). The sublevel set of a convex function, which is the set of points for which the function value is less than or equal to a certain value, is always a convex set. Therefore, the set C is convex if $A \succeq 0$.

b) It is true. This is because the intersection of two convex sets is always a convex set. The set C is convex as shown in a), and the hyperplane defined by $g^Tx + h = 0$ is also a convex set. Therefore, their intersection is convex. Furthermore, if $A + \lambda gg^T \succeq 0$, then the function defining the set C remains convex, so the intersection of C with any hyperplane (including the one defined by $g^Tx + h = 0$) remains convex.

Problem 5: Dual cones

Describe the dual cone for each of the following cones.

- a) $K = \{0\}.$
- b) $K = \mathbf{R}^2$.
- c) $K = \{(x_1, x_2) | |x_1| \le x_2 \}$.
- d) $K = \{(x_1, x_2) \mid x_1 + x_2 = 0\}$.

Solution. The dual cone of a cone K is defined as the set of all vectors y such that $y^T x \leq 0$ for all x in K.

- a) the dual cone is \mathbb{R}^n , where n is the dimension of the space.
- b) the dual cone is also R^2 .
- c) the dual cone is $\{(y1, y2)|y1>=0 \text{ and } y2>=0\}.$
- d) the dual cone is $\{(y1, y2)|y1 y2 >= 0 \text{ and } y1 + y2 >= 0\}$