Optimization Methods

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Homework 1

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Notice

• The submission email is: njuoptfall2019@163.com.

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Problem 1: Norms

A function $f: \mathbb{R}^n \leftarrow \mathbb{R}$ with $\text{dom} f = \mathbb{R}^n$ 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 $x, y \in v$

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

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

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

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

(Hint: Use CauchySchwarz inequality.)

Solution.

a) ::
$$||z||_* = \sup\{z^T x | ||x|| \le 1\} = \sup\{|z^T x| |||x|| \le 1\}\}$$

$$|z^T x| \ge 0$$

$$|z||_* \ge |z^T x| \ge 0$$

$$\forall x \in \mathbb{R}^n$$
, if $z = 0$, then $z^T x = 0$

$$\therefore \|z\|_* = 0 \text{ only if } z = 0$$

$$\therefore ||tz||_* = |t|||z||_*$$

$$\therefore \|z_1 + z_2\|_* \le \|z_1\|_* + \|z_2\|_*$$

 $\|\cdot\|$ is a valid norm.

b) : From Cauchy-Schwarz inequality, we have
$$|z^Tx| \leq \sqrt{\sum_{i=1}^n (z^T)^2 * \sum_{i=1}^n (x)^2} = ||z||_{2*} ||x||_{2*}$$

 \therefore For nonzero z, the value of x that maximize z^Tx over $||x||_2 \le 1$ is $\frac{z}{||z||_2}$

$$\therefore$$
 Let $x = \frac{z}{\|z\|_2}$, We get $\sup\{|z^T x| | \|x\|_{2*} \le 1\} = \|z\|_2 \|\frac{z}{\|z\|_2} \|_2 = \|z\|_2$

$$|z|_{2*} = ||z||_2$$

Problem 2: Affine and Convex Sets

Affine sets C_a and convex C_c sets are the sets satisfying the constraints below:

$$\theta x_1 + (1 - \theta)x_2 \in C_a$$
s.t. $x_1, x_2 \in C_a$ (1)

$$\theta x_1 + (1 - \theta)x_2 \in C_c$$

s.t. $x_1, x_2 \in C_c, 0 \ge \theta \le 1$ (2)

- a) Is the set $\{\alpha \in \mathbb{R}^k | p(0) = 1, |p(t)| \le 1 \text{ for } \alpha \le t \le \beta\}$, where $p(t) = \alpha_1 + \alpha_2 t + \dots + \alpha_k t^{k-1}$, affine?
- b) Determine if each set below is convex.
 - 1) $\{(x,y) \in \mathbf{R}^2_{++} | x/y \le 1\}.$
 - 2) $\{(x,y) \in \mathbf{R}_{++}^2 | x/y \ge 1\}.$
 - 3) $\{(x,y) \in \mathbf{R}^2_+ | xy \le 1\}.$
 - 4) $\{(x,y) \in \mathbf{R}^2_+ | xy \ge 1\}.$
 - 5) $\{(x,y) \in \mathbf{R}^2 | y = \tanh(x) = \frac{e^x e^{-x}}{e^x + e^{-x}} \}.$

Solution.

a) Assume $C = \{\alpha \in \mathbb{R}^k | p(0) = 1, |p(t)| \le 1 \text{ for } \alpha \le t \le \beta\}$, where $p(t) = \alpha_1 + \alpha_2 t + \dots + \alpha_k t^{k-1}$ Let $x1, x2 \in C$. For $\forall \theta$, we have:

$$\begin{aligned} & |[\theta x_1 + (1-\theta)x_2] + [\theta x_1 + (1-\theta)x_2]t + \dots + [\theta x_1 + (1-\theta)x_2]t^{k-1}| \\ &= |(\theta x_1 + \theta x_1 t + \dots + \theta x_1 t^{k-1}) + [(1-\theta)x_2 + (1-\theta)x_2 t + \dots + (1-\theta)x_2 t^{k-1}]| \\ &\leq |(\theta x_1 + \theta x_1 t + \dots + \theta x_1 t^{k-1})| + |[(1-\theta)x_2 + (1-\theta)x_2 t + \dots + (1-\theta)x_2 t^{k-1}]| \\ &= |\theta + (1-\theta)| \\ &= 1 \end{aligned}$$

$$\therefore \forall (\theta x_1 + (1 - \theta)x_2) \in C$$

 $\therefore \text{ Set } \{\alpha \in \mathbb{R}^k | p(0) = 1, |p(t)| \le 1 \text{ for } \alpha \le t \le \beta\}, \text{ where } p(t) = \alpha_1 + \alpha_2 t + \dots + \alpha_k t^{k-1}, \text{ is affine.}$

b)

1) Assume $(x_1, y_1), (x_2, y_2) \in \{(x, y) \in \mathbf{R}_{++}^2 | x/y \le 1\}$

$$\because \frac{x}{y} \le 1$$

$$\therefore x \leq y$$

 \therefore For $\forall \theta$ that satisfies $0 \le \theta \le 1$, we have:

$$\theta(x_1, y_1) + (1 - \theta)(x_2, y_2)$$

$$= (\theta x_1 + (1 - \theta)x_2, \theta y_1 + (1 - \theta)y_2)$$

$$= \frac{\theta x_1 + (1 - \theta)x_2}{\theta y_1 + (1 - \theta)y_2}$$

$$\leq \frac{\theta y_1 + (1 - \theta)y_2}{\theta y_1 + (1 - \theta)y_2} = 1$$

 $\therefore \{(x,y) \in \mathbf{R}^2_{++} | x/y \le 1\}$ is convex.

2) In a similar way with 1), Assume $(x_1, y_1), (x_2, y_2) \in \{(x, y) \in \mathbb{R}^2_{++} | x/y \ge 1\}$

 \therefore For $\forall \theta$ that satisfies $0 \le \theta \le 1$, we have:

$$\theta(x_1, y_1) + (1 - \theta)(x_2, y_2) = \frac{\theta x_1 + (1 - \theta)x_2}{\theta y_1 + (1 - \theta)y_2} \ge \frac{\theta y_1 + (1 - \theta)y_2}{\theta y_1 + (1 - \theta)y_2} = 1$$

 $\therefore \{(x,y) \in \mathbf{R}_{++}^2 | x/y \ge 1\} \text{ is convex.}$

3) Let
$$(2, \frac{1}{2}), (\frac{1}{2}, 2) \in \{(x, y) \in \mathbf{R}^2_+ | xy \le 1\}$$

: For
$$\forall \theta$$
 that satisfies $0 \le \theta \le 1$, assume $(x,y) = \theta(2,\frac{1}{2}) + (1-\theta)(\frac{1}{2},2) = (\frac{3}{2}\theta + \frac{1}{2}, -\frac{3}{2}\theta + 2)$

$$\therefore (\frac{3}{2}\theta + \frac{1}{2})(-\frac{3}{2}\theta + 2) = -\frac{9}{4}\theta^2 + \frac{9}{4}\theta + 1 = -\frac{9}{4}(\theta - \frac{1}{2})^2 + \frac{25}{16}$$

$$\therefore \exists \theta = \frac{1}{2} \in [0, 1], xy = \frac{25}{16} \notin \{(x, y) \in \mathbf{R}^2_+ | xy \le 1\}$$

 $\therefore \{(x,y) \in \mathbf{R}^2_+ | xy \le 1\}$ isn't convex.

4) Let
$$(1,1), (-1,-1) \in \{(x,y) \in \mathbf{R}^2_+ | xy \ge 1\}$$

$$(For) \forall \theta$$
 that satisfies $0 \le \theta \le 1$, assume $(x,y) = \theta(1,1) + (1-\theta)(-1,-1) = (2\theta-1,2\theta-1)$

$$\therefore \text{ Let } \theta = \frac{1}{2} \in [0,1], \text{ we have } xy = (2 \times \frac{1}{2} - 1)(2 \times \frac{1}{2} - 1) = 0 \notin \{(x,y) \in \mathbf{R}^2_+ | xy \ge 1\}$$

 $\therefore \{(x,y) \in \mathbf{R}^2_+ | xy \ge 1\}$ isn't convex.

5) Assume
$$(x_1, y_1), (x_2, y_2) \in \{(x, y) \in \mathbf{R}^2 | y = \tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}} \}$$

$$\therefore \quad \theta y_1 + (1 - \theta) y_2
= \theta \frac{e^{x_1} - e^{-x_1}}{e^{x_1} + e^{-x_1}} + (1 - \theta) \frac{e^{x_2} - e^{-x_2}}{e^{x_2} + e^{-x_2}}
= \theta \frac{e^{x_1} (1 - \frac{1}{e})}{e^{x_1} (1 + \frac{1}{e})} + (1 - \theta) \frac{e^{x_2} (1 - \frac{1}{e})}{e^{x_2} (1 + \frac{1}{e})}
= (\theta + 1 - \theta) \frac{1 - \frac{1}{e}}{1 + \frac{1}{e}}
= \frac{1 - \frac{1}{e}}{1 + \frac{1}{e}}
\therefore \quad \tanh(\theta x_1 + (1 - \theta) x_2)$$

$$\tanh(\theta x_1 + (1 - \theta) x_2)
= \frac{e^{\theta x_1 + (1 - \theta) x_2} - e^{-(\theta x_1 + (1 - \theta) x_2)}}{e^{\theta x_1 + (1 - \theta) x_2} + e^{-(\theta x_1 + (1 - \theta) x_2)}}
= \frac{e^{\theta x_1 + (1 - \theta) x_2} (1 - \frac{1}{e})}{e^{\theta x_1 + (1 - \theta) x_2} (1 + \frac{1}{e})}
= \frac{1 - \frac{1}{e}}{1 + \frac{1}{e}}$$

 \therefore For $\forall \theta$ that satisfies $0 \le \theta \le 1, \theta y_1 + (1 - \theta)y_2 = \tanh(\theta x_1 + (1 - \theta)x_2)$

:.
$$\{(x,y) \in \mathbf{R}^2 | y = \tanh(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}} \}$$
 is convex.

Problem 3: Examples

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

$$C = \{ x \in \mathbb{R}^n | x^{\top} A x + b^{\top} x + c \le 0 \},$$
 (3)

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

- 1) Show that C is convex if $A \succeq 0$.
- 2) Is the following statement true? The intersection of C and the hyperplane defined by $g^{\top}x + h = 0$ is convex if $A + \lambda g g^{\top} \succeq 0$ for some $\lambda \in \mathbb{R}$.
- b) The polar of $C \subseteq \mathbb{R}^n$ is defined as the set

$$C^{\circ} = \{ y \in \mathbb{R}^n | y^{\top} x \le 1 \text{ for all } x \in C \}$$

.

- 1) Show that C° is affine.
- 2) What is a polar of a polyhedra?
- 3) What is the polar of the unit ball for a norm $||\cdot||$?
- 4) Show that if C is closed and convex, with $0 \in C$, then $(C^{\circ})^{\circ} = C$

Solution.

a) 1) Assume $x_1, x_2 \in C, \theta \in [0, 1]$

$$\begin{array}{l} :: \theta x_{1}^{T}Ax_{1} + \theta b^{T}x_{1} + \theta c \leq 0, \ (1-\theta)x_{2}^{T}Ax_{2} + (1-\theta)b^{T}x_{2} + (1-\theta)c \leq 0 \\ :: \ (\theta x_{1} + (1-\theta)x_{2})^{T}A(\theta x_{1} + (1-\theta)x_{2}) + b^{T}(\theta x_{1} + (1-\theta)x_{2}) + c \\ &= \theta^{2}x_{1}^{T}Ax_{1} + \theta(1-\theta)x_{1}^{T}Ax_{2} + \theta(1-\theta)x_{2}^{T}Ax_{1} + (1-\theta)^{2}x_{2}^{T}Ax_{2} + \theta b^{T}x_{1} + (1-\theta)b^{T}x_{2} + \theta c + (1-\theta)c \\ &\leq \theta^{2}x_{1}^{T}Ax_{1} + \theta(1-\theta)x_{1}^{T}Ax_{2} + \theta(1-\theta)x_{2}^{T}Ax_{1} + (1-\theta)^{2}x_{2}^{T}Ax_{2} - \theta x_{1}^{T}Ax_{1} - (1-\theta)x_{2}^{T}Ax_{2} \\ &= -\theta(1-\theta)x_{1}^{T}Ax_{1} + \theta(1-\theta)x_{1}^{T}Ax_{2} + \theta(1-\theta)x_{2}^{T}Ax_{1} - \theta(1-\theta)x_{2}^{T}Ax_{2} \\ &= \theta(1-\theta)(x_{1}^{T}Ax_{2} + x_{2}^{T}Ax_{1} - x_{1}^{T}Ax_{1} - x_{2}^{T}Ax_{2}) \\ &= -\theta(1-\theta)(x_{1}^{T}-x_{2}^{T})A(x_{1} - x_{2}) \\ & \therefore A \geq 0 \\ & \therefore (x_{1}^{T}-x_{2}^{T})A(x_{1} - x_{2}) \geq 0 \\ & \therefore \theta \in [0,1] \\ & \therefore (\theta x_{1} + (1-\theta)x_{2})^{T}A(\theta x_{1} + (1-\theta)x_{2}) + b^{T}(\theta x_{1} + (1-\theta)x_{2}) + c \leq 0 \\ & \therefore C \text{ is convex if } A \geq 0 \end{array}$$

2) (Red color means I have referred to solutions here.)

Assume $x_1, x_2 \in \mathbb{C}, \theta \in [0,1]$

- \therefore A set is convex if and only if its intersection with an arbitrary line $\{\hat{x} + tv | t \in R\}$ is convex.
- \therefore Let $H = \{x | g^T x + h = 0\}$, We have

$$(\hat{x} + tv)^T A(\hat{x} + tv) + b^T (\hat{x} + tv) + c = \alpha t^2 + \beta t + \gamma$$

where $\alpha = v^T A v, \beta = b^T v + 2 \hat{x}^T A v, \gamma = c + b^T \hat{x} + \hat{x}^T A \hat{x}$ In addition, define $\delta = g^T v, \epsilon = g^T \hat{x} + h$

 \therefore Without loss of generality we can assume that $\hat{x} \in H, i.e., \epsilon = 0$. The intersection of $C \cap H$ with the line defined by \hat{x} and v is

$$\{\hat{x} + tv | \alpha t^2 + \beta t + \gamma \le 0, \delta t = 0\}.$$

... If $\delta = g^T v \neq 0$, the intersection is the singleton $\{\hat{x}\}\$, if $\gamma \leq 0$ or it is empty. In either case it is a convex set. If $\delta = g^T v = 0$, the set reduces to

$$\{\hat{x} + tv | \alpha t^2 + \beta t + \gamma \le 0\}$$

which is convex if $\alpha \geq 0$. Therefore $C \cap H$ is convex if

$$g^T v = 0 \Rightarrow v^T A v \ge 0.$$

... This is true if there exists λ such that $A + \lambda gg^T \succeq 0$; then $g^T v = 0 \Rightarrow v^T Av \geq 0$ holds. because then

$$v^T A v = v^T A v + \lambda (v^T g)(g^T v) = v^T (A + \lambda g g^T) v \ge 0$$

for all v satisfying $g^T v = 0$

 \therefore The intersection of C and the hyperplane defined by $g^{\top}x + h = 0$ is convex if $A + \lambda gg^{\top} \succeq 0$ for some $\lambda \in \mathbb{R}$ is true.

b) 1) Assume $y1, y2 \in C^{\circ}$, we have

$$(\theta y_1 + (1 - \theta)y_2)^T x = \theta y_1^T x + (1 - \theta)y_2^T x \le \theta + (1 - \theta) = 1$$

for all $\theta \in [0, 1]$.

 $\therefore \theta y_1 + (1 - \theta) y_2 \in C^{\circ}$

 $\therefore C^{\circ}$ is affine. 2) The polar of a polyhedra is

Problem 4: Operations That Preserve Convexity

Suppose $\phi: \mathbb{R}^n \to \mathbb{R}^m$ and $\psi: \mathbb{R}^m \to \mathbb{R}^p$ are the linear-fractional functions

$$\phi(x) = \frac{Ax+b}{c^{\top}x+d}, \psi(x) = \frac{Ey+f}{g^{\top}y+h}, \tag{4}$$

with domains **dom** $\phi = \{x | c^{\top}x + d > 0\}$, **dom** $\psi = \{y | g^{\top}x + h > 0\}$. We associate with ϕ and ψ the matrices

$$\begin{bmatrix} A & b \\ c^{\top} & d \end{bmatrix}, \begin{bmatrix} A & b \\ g^{\top} & h \end{bmatrix}, \tag{5}$$

respectively.

Now, consider the composition Γ of ϕ and ψ , i.e., $\Gamma(x) = \psi(\phi(x))$, with domain

$$\mathbf{dom}\Gamma = \{x \in \mathbf{dom} \ \phi | \phi(x) \in \mathbf{dom} \ \psi\}. \tag{6}$$

Show that Γ is linear-fractional, and that the matrix associate with it is the product

$$\begin{bmatrix} A & b \\ c^{\top} & d \end{bmatrix} \begin{bmatrix} A & b \\ g^{\top} & h \end{bmatrix} . \tag{7}$$

Solution. Write your solution here.

Problem 5: Generalized Inequalities

Let K^* be the dual cone of a convex cone K. Prove the following

- 1) K^* is indeed a convex cone.
- 2) $K_1 \subseteq K_2$ implies $K_1^* \subseteq K_2^*$.
- 3) K^{**} is the closure of K.

Solution. Write your solution here.