

Homework 1

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1. Consider the set defined by the following inequalities

$$S = \{x \in \mathbb{R}^2 : x_1 \geq x_2 - 1, x_2 \geq 0\} \cup \{x \in \mathbb{R}^2 : x_1 \leq x_2 - 1, x_2 \leq 0\}$$

- (a) Draw the set S . Is it convex?
 (b) Show that the set S can be described as a single quadratic inequality of the form

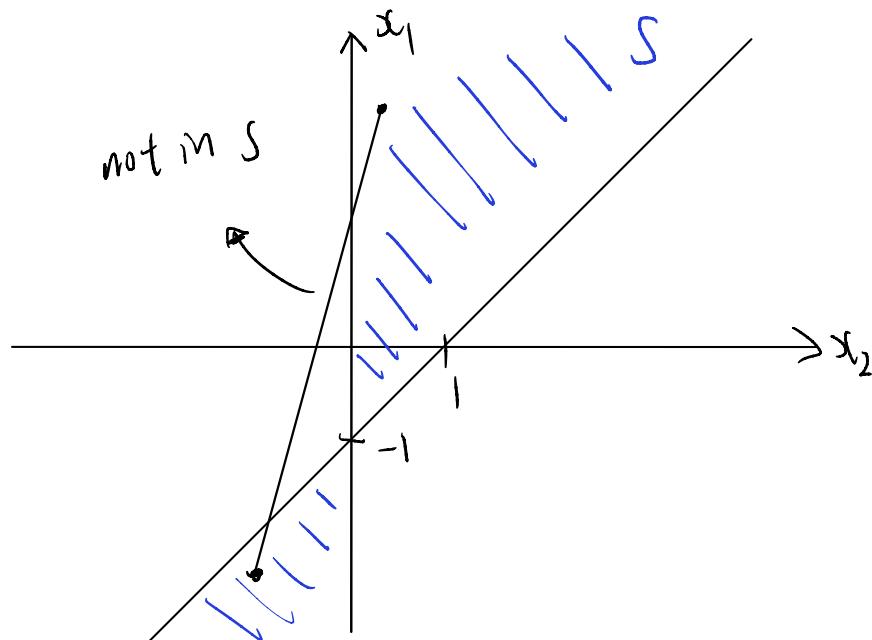
$$q(x) = x^T A x + 2b^T x + c \leq 0$$

for a matrix $A \in \mathbf{S}^2$ (2 by 2 symmetric matrix), $b \in \mathbb{R}^2$ and $c \in \mathbb{R}$ which you should determine.

Hint: Note that q is continuous, $q \leq 0$ on S and $q > 0$ on S^c , hence q must be 0 on the boundary of S .

- (c) What is the convex hull of the set S ?

(a)



Set S is drawn as the blue shaded region

Set S is not convex as the definition of convex

lets state that $S \subseteq \mathbb{R}^n$ is convex if the

line segment between any 2 points in S

lie in S

(not the case as seen in the diagram)

(b)

Consider the following 2 inequalities

$$\begin{array}{l} x_1 \geq x_2 - 1, \quad x_2 \geq 0 \\ x_1 \leq x_2 - 1, \quad x_2 \leq 0 \end{array} \quad \left. \begin{array}{l} \text{symmetric} \\ \text{about } x_2 \text{ axis} \end{array} \right\}$$

↓ rewritten as $\left\{ x \in \mathbb{R}^2 : \left(1 - \frac{x_1+1}{x_2}\right) \left(1 + \frac{x_1+1}{x_2}\right) \leq 0 \right\}$

$$\left(1 - \frac{x_1+1}{x_2}\right) \left(1 + \frac{x_1+1}{x_2}\right) \leq 0$$

$$1 - \frac{x_1^2 + 2x_1 + 1}{x_2^2} \leq 0$$

$$x_2^2 \leq x_1^2 + 2x_1 + 1$$

$$x_2^2 - x_1^2 - 2x_1 - 1 \leq 0$$

(written as)

$$\begin{aligned} -2 &= 2(f \ 0) \\ \Rightarrow f &= -1 \end{aligned}$$

$$\begin{aligned} x_2^2 - x_1^2 &= (x_1 x_2) \left(\frac{ax_1 + bx_2}{cx_1 + dx_2} \right) \\ &= ax_1^2 + bx_1 x_2 \\ &\quad + cx_1 x_2 + dx_2^2 \end{aligned}$$

$$\begin{aligned} b, c &= 0 \\ a &= -1, \quad d = 1 \end{aligned}$$

$$(x_1 x_2) \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + 2 \begin{pmatrix} -1 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} - 1 \leq 0$$

(in the form stated)

where $A = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}$, $b = \begin{pmatrix} -1 \\ 0 \end{pmatrix}$ and $c = -1$
symmetric

(c)

Convex hull : smallest convex set that contains the original set

$(0,0)$ belongs to both subsets \rightarrow part of convex hull

For $x_1 > 0$, convex hull includes $(t_1, 0)$ where $t \geq 0$

For $x_1 \leq 0$, convex hull includes $(t_1, 0)$ where $t \leq 0$

Hence convex hull is the entire \mathbb{R}^2 plane as
set S is non-convex so it has to include all
points in x_1 and extending in all directions

2. The following is a very important result in linear programming duality. Let $A \in \mathbb{R}^{m \times n}$ and $y \in \mathbb{R}^m$. Show that one and only one of the following two conditions is satisfied.

- the system of linear equations $Ax = y$ admits a non-negative solution $x \geq 0$; —(1)
- there exists $z \in \mathbb{R}^m$ such that $z^T A \geq 0$ and $z^T y < 0$. —(2)

You should use the *strict separating hyperplane theorem*.

Firstly, we show that we cannot have both (1) and (2)
(cannot hold at the same time)

Take $z^T Ax$,

$$\text{case 1: } z^T Ax = z^T(Ax) \\ = z^T(y) < 0$$

since by (1): $Ax = y$ and by (2): $z^T y < 0$

$$\text{case 2: } z^T Ax = (z^T A)x \quad (\text{take transpose})$$

$$\geq 0$$

since by (1): $x \geq 0$ and by (2): $z^T A \geq 0$

Next, WTS: if (1) does not hold then (2) must hold

Let $A = \begin{pmatrix} | & | & | \\ a_1, a_2, \dots, a_n \\ | & | & | \end{pmatrix}$ where a_1, a_2, \dots, a_n
are the columns of matrix A

we define $\Omega = \text{cone}(a_1, a_2, \dots, a_n)$

$$= \left\{ s \in \mathbb{R}^m : s = \sum_{i=1}^n \lambda_i a_i, \lambda_i \geq 0 \forall i \right\}$$

which is non-empty, closed and convex

Then $Ax = \sum_{i=1}^n x_i a_i$, $\exists x$ such that $As = y$ and $x \geq 0$
 iff $y \in Q$

Assume ① does not hold,

then $y \notin Q$

→ Strict Version

Recall separating hyperplane theorem,

$\Leftrightarrow \exists \alpha, \beta \in \mathbb{R}^m$, $\alpha \neq 0$ such that

$\alpha^T y > \beta$ and $\alpha^T s < \beta$, $\forall s \in Q$
 (and $y \notin Q$)

$0 \in Q$, $\beta > 0$, $\lambda a_i \in C \quad \forall \lambda > 0$

\Rightarrow since $\alpha^T s < \beta$, $\forall s \in Q$

we have $\alpha^T(\lambda a_i) \in Q$, $\forall \lambda > 0$

$\Leftrightarrow \alpha^T(\lambda a_i) < \beta \Rightarrow \alpha^T a_i < \frac{\beta}{\lambda}, \forall \lambda > 0$

since $\beta > 0$ as $\lambda \rightarrow \infty \Rightarrow \alpha^T a_i \leq 0$

Set $z = -\alpha$, we get $z^T y < 0$ and $z^T a_i \geq 0 \quad \forall i$

$\Rightarrow z^T A \geq 0$ (since a_i are columns of A)

Hence showing ② holds

3. The *polar* of an arbitrary set $C \subset \mathbb{R}^n$ is defined as the set

$$C^\circ := \{y \in \mathbb{R}^n : y^T x \leq 1 \text{ for all } x \in C\}$$

- (a) Let $C \subset \mathbb{R}^n$ be any set, not necessarily convex. Is C° convex? Justify your answer carefully.
- (b) Recall that a *cone* K is such that if $x \in K$ and $\theta \geq 0$, then $\theta x \in K$. Recall that the *dual cone* is defined as

$$K^* := \{y \in \mathbb{R}^n : y^T x \geq 0 \text{ for all } x \in K\}$$

It is known that the polar of the cone K , denoted as K° , and the dual cone, denoted as K^* , are related as follows:

$$K^\circ = -cK^*$$

for some *positive* number c . Find c .

- (c) Show carefully that the polar of the unit ball $B(0, 1) := \{x \in \mathbb{R}^n : \|x\|_2 \leq 1\}$ is the unit ball. You may find the Cauchy-Schwarz inequality (i.e., $x^T y \leq \|x\|_2 \|y\|_2$) useful.

(a) C° convex $\Rightarrow \forall y_1, y_2 \in C^\circ \text{ and } \theta \in [0, 1]$
 then $\theta y_1 + (1-\theta)y_2 \in C^\circ$

Since $y_1, y_2 \in C^\circ$,

$$y_1^T x \leq 1 \text{ and } y_2^T x \leq 1$$

we take linear combination of y_1 and y_2 with x

$$\begin{aligned} \Rightarrow (1-\theta) y_1^T x + \theta y_2^T x &\leq (1-\theta)(1) + (\theta)(1) \\ &= 1 - \theta + \theta \\ &= 1 \end{aligned}$$

This shows that linear combi of y_1 and y_2
 satisfies conditions for C°

$$\text{hence } \forall x \in C, (1-\theta)y_1 + \theta y_2 \in C^\circ$$

(b) K is a cone

$$\rightarrow \text{polar of cone } K^0 = \{y \in \mathbb{R}^n : y^T x \leq 0, \forall x \in K\}$$

since x is either all non-positive or all non-negative (but not both),

we are able to derive for $\forall c$:

$$y^T x \leq 0 \Leftrightarrow - (y^T x) \geq - (0)$$
$$- y^T x \geq 0$$

Hence proving $K^0 = -K^*$ where $c = -1$

(c) Polar of unit ball denoted a)

$$B^*(0,1) = \{y \in \mathbb{R}^n : y^T x \leq 1 \text{ for } \forall x \in B(0,1)\}$$

$$\text{WTS: } B^*(0,1) = B(0,1)$$

(\Rightarrow) prove $B^*(0,1) \subseteq B(0,1)$

consider $x \in B^*(0,1)$

For any $y \in$ unit ball $B(0,1)$, we have inner product

$$y^T x \leq 1 \quad (\text{definition of polar})$$

$$\leq \|y\|_2 \|x\|_2 \quad (\text{by Cauchy-Schwarz inequality})$$

since inequality holds for $\forall y \in B(0,1)$, we can choose y to be x itself

$$x^T x \leq \|x\|_2 \|x\|_2$$

$$\Rightarrow \|x\|_2^2 \leq \|x\|_2^2 \quad (\|x\|_2^2 \text{ less than equal to itself} \Rightarrow \|x\|_2^2 \text{ is non-negative})$$

since $\|x\|_2^2 \leq 1$ and $\|y\|_2$ is non-negative, $\Rightarrow \|x\|_2$ also non-negative
it follows that $\|x\|_2 \leq 1$ hence

$\forall x \in B^*(0,1)$ satisfies $\|x\|_2 \leq 1$ and belongs inside unit ball

(\Leftarrow) prove $B(0,1) \subseteq B^o(0,1)$

consider $y \in B(0,1)$,

WTS: for this given y , $y^T x \leq 1$ holds for

$\forall x \in B^o(0,1)$

let x be a point in $B^o(0,1)$, by the
definition of polar: $y^T x \leq 1$

This inequality holds for $\forall y$ in $B(0,1)$ so it
will be satisfied for any value of x we choose

Hence $B(0,1) \subseteq B^o(0,1)$

thus this proves equality

4. (a) For x, y both positive scalars, show that

$$ye^{x/y} = \sup_{\alpha > 0} \alpha(x+y) - y\alpha \log \alpha.$$

(log here refers to the natural logarithm.) Use the above result to prove that the function $f : \mathbb{R}_{++}^n \rightarrow \mathbb{R}$ defined as

$$f(x, y) = ye^{x/y}$$

is convex.

- (b) Show that the function $f : \mathbb{R}_+^2 \rightarrow \mathbb{R}$

$$f(x) = (\sqrt{x_1} + \sqrt{x_2})^2$$

is concave by examining the definiteness of its Hessian.

Differentiate both sides

$$\frac{\partial}{\partial \alpha} ye^{\frac{x}{y}} = \frac{\partial}{\partial \alpha} \left\{ \sup_{\alpha > 0} \alpha(x+y) - y\alpha \log \alpha \right\}$$

$$\Rightarrow 0(x+y) - y \left\{ \log \alpha + \alpha \left(\frac{1}{\alpha} \right) \right\} = 0$$

$$x+y - y \log \alpha - y = 0$$

$$x - y \log \alpha = 0 \Rightarrow \log \alpha = \frac{x}{y}$$

$$\alpha = e^{\frac{x}{y}} \quad (\text{maximizing initial expression})$$

hence

$$\begin{aligned} \sup_{\alpha > 0} \alpha(x+y) - y\alpha \log \alpha &= e^{\frac{x}{y}}(x+y) - y(e^{\frac{x}{y}})(\log e^{\frac{x}{y}}) \\ &= e^{\frac{x}{y}} \left\{ x+y - y \left(\frac{x}{y} \right) \right\} \\ &= ye^{\frac{x}{y}} \end{aligned}$$

(addition of linear functions)

Since linear functions are convex and addition preserves linearity, $f(x, y)$ is convex

$$(b) f(x_1, x_2) = (\sqrt{x_1} + \sqrt{x_2})^2$$

$$\nabla f(x_1, x_2) = \begin{bmatrix} 1 + 2\sqrt{x_2} \left(\frac{1}{2}\right) \frac{1}{\sqrt{x_1}} \\ 1 + 2\sqrt{x_1} \left(\frac{1}{2}\right) \frac{1}{\sqrt{x_2}} \end{bmatrix}$$

$$= \begin{bmatrix} 1 + \frac{\sqrt{x_2}}{\sqrt{x_1}} \\ 1 + \frac{\sqrt{x_1}}{\sqrt{x_2}} \end{bmatrix}$$

$$\nabla^2 f(x_1, x_2) = \begin{bmatrix} -\frac{\sqrt{x_2}}{2(x_1)^{\frac{3}{2}}} & \frac{1}{2\sqrt{x_1}\sqrt{x_2}} \\ \frac{1}{2\sqrt{x_1}\sqrt{x_2}} & -\frac{\sqrt{x_1}}{2(x_2)^{\frac{3}{2}}} \end{bmatrix}$$

principals
minor
if non-
positive 100

$$-\frac{\sqrt{x_2}}{2(x_1)^{\frac{3}{2}}} < 0 \quad \text{given } \sqrt{x_2} \geq 0 \text{ and } (x_1)^{\frac{3}{2}} > 0$$

$$\det \{ \nabla f(x_1, x_2) \} = \frac{-\sqrt{x_2}}{2(x_1)^{\frac{3}{2}}} \cdot \frac{-\sqrt{x_1}}{2(x_2)^{\frac{3}{2}}} - \left(\frac{1}{2\sqrt{x_1}\sqrt{x_2}} \right)^2$$

$$> 0$$

$\nabla^2 f(x_1, x_2) \leq 0$ hence f is concave