

CvxOpt Homework #1

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2.2 Show that a set is convex if and only if its intersection with any line is convex. Show that a set is affine if and only if its intersection with any line is affine.

First we proof *a set is convex iff. its intersection with any line is convex.*

\Rightarrow Proof: Let set C is convex, then if the intersection of a line l and C , $l \cap C$, only contains one point, then it is obviously convex; otherwise, assume $l \cap C$ has at least two points. Then for any two points A, B in $l \cap C$, because the convexity of C , for any θ with $0 \leq \theta \leq 1$ we have $\theta A + (1 - \theta)B \in C$. Also because point $\theta A + (1 - \theta)B$ belongs to l , so point $\theta A + (1 - \theta)B \in l \cap C$. So the intersection of the set and any line is convex.

\Leftarrow Proof: Let set C satisfies that for any line l , $l \cap C$ is convex, then if C contains only one point it is obviously convex; otherwise, assume C has at least two points. Then for any two points A, B in C , the intersection of line AB and C , let it be D , is convex, so for any θ with $0 \leq \theta \leq 1$, we have $\theta A + (1 - \theta)B \in D$. Plus we have $D \subseteq C$, so $\theta A + (1 - \theta)B \in C$, which indicates the convexity of C .

Then we proof *a set is affine iff. its intersection with any line is affine.*

\Rightarrow Proof: Let set C is affine, then if the intersection of a line l and C , $l \cap C$, only contains one point, then it is obviously affine; otherwise, assume $l \cap C$ has at least two points. Then for any two points A, B in $l \cap C$, because the affinity of C , for any $\theta \in \mathbf{R}$ we have $\theta A + (1 - \theta)B \in C$. Also because point $\theta A + (1 - \theta)B$ belongs to l , so point $\theta A + (1 - \theta)B \in l \cap C$. So the intersection of the set and any line is affine.

\Leftarrow Proof: Let set C satisfies that for any line l , $l \cap C$ is affine, then if C contains only one point it is obviously affine; otherwise, assume C has at least two points. Then for any two points A, B in C , the intersection of line AB and C , let it be D , is affine, so for any $\theta \in \mathbf{R}$ we have $\theta A + (1 - \theta)B \in D$. Plus we have $D \subseteq C$, so $\theta A + (1 - \theta)B \in C$, which indicates the affinity of C .

2.10 *Solution set of a quadratic inequality.* Let $C \subseteq \mathbf{R}^n$ be the solution set of a quadratic inequality,

$$C = \{x \in \mathbf{R}^n \mid x^T A x + b^T x + c \leq 0\},$$

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

(a) Show that C is convex if $A \succeq 0$.

Proof: If C only has one element then it is obviously convex; otherwise assume C has at least two elements. Then for any two elements $x_1, x_2 \in C$ and any θ with $0 \leq \theta \leq 1$, we proof $\theta x_1 + (1 - \theta)x_2 \in C$.

Because $x_1, x_2 \in C$ we have

$$\begin{cases} x_1^T A x_1 + b^T x_1 + c \leq 0 \\ x_2^T A x_2 + b^T x_2 + c \leq 0 \end{cases}$$

then we get

$$\theta(x_1^T A x_1 + b^T x_1 + c) + (1 - \theta)(x_2^T A x_2 + b^T x_2 + c) \leq 0$$

then

$$\begin{aligned}
& \theta(x_1^T A x_1 + b^T x_1 + c) + (1 - \theta)(x_2^T A x_2 + b^T x_2 + c) \\
& - ((\theta x_1 + (1 - \theta)x_2)^T A (\theta x_1 + (1 - \theta)x_2) + b^T (\theta x_1 + (1 - \theta)x_2) + c) \\
& = \theta x_1^T A x_1 + (1 - \theta)x_2^T A x_2 - (\theta x_1 + (1 - \theta)x_2)^T A (\theta x_1 + (1 - \theta)x_2) \\
& = \theta x_1^T A x_1 + (1 - \theta)x_2^T A x_2 - \theta^2 x_1^T A x_1 - (1 - \theta)\theta x_2^T A x_1 - \theta(1 - \theta)x_1^T A x_2 - (1 - \theta)^2 x_2^T A x_2 \\
& = \theta(1 - \theta)(x_1^T A x_1 + x_2^T A x_2 - x_2^T A x_1 - x_1^T A x_2) \\
& = \theta(1 - \theta)(x_1 - x_2)^T A (x_1 - x_2) \\
& \geq 0
\end{aligned}$$

The last step is because the semidefiniteness of A , so that $(x_1 - x_2)^T A (x_1 - x_2) \geq 0$.

So we have

$$\begin{aligned}
& ((\theta x_1 + (1 - \theta)x_2)^T A (\theta x_1 + (1 - \theta)x_2) + b^T (\theta x_1 + (1 - \theta)x_2) + c) \\
& \leq \theta(x_1^T A x_1 + b^T x_1 + c) + (1 - \theta)(x_2^T A x_2 + b^T x_2 + c) \\
& \leq 0
\end{aligned}$$

Which indicates $\theta x_1 + (1 - \theta)x_2 \in C$, so C is convex.

2.24 Supporting hyperplanes.

(a) Express the closed convex set $\{x \in \mathbf{R}_+^2 \mid x_1 x_2 \geq 1\}$ as an intersection of halfspaces.

Let $C = \{x \in \mathbf{R}_+^2 \mid x_1 x_2 \geq 1\}$, then $\mathbf{bd}C = \{x \in \mathbf{R}_+^2 \mid x_1 x_2 = 1\}$, the for any point $x = (t, \frac{1}{t})$ in $\mathbf{bd}C$, the supporting hyperplane is

$$\{x \in \mathbf{R}^2 \mid \frac{x_1}{t^2} + x_2 \geq \frac{2}{t}\}$$

then C can be expressed as

$$C = \bigcap_{t>0} \{x \in \mathbf{R}^2 \mid \frac{x_1}{t^2} + x_2 \geq \frac{2}{t}\}$$

2.32 Find the dual cone of $\{Ax \mid x \succeq 0\}$, where $A \in \mathbf{R}^{m \times n}$.

The dual cone can be expressed as

$$\{y \mid Ay \succeq 0\}$$

2.35 Copositive matrices. A matrix $X \in \mathbf{S}^n$ is called *copositive* if $z^T X z \geq 0$ for all $z \succeq 0$. Verify that the set of copositive matrices is a proper cone. Find its dual cone.

Let the set of all *copositive* matrices be K in \mathbf{S}^n . So K can be expressed as infinite halfspaces defined by

$$z^T X z = \sum_i \sum_j z_i z_j X_{ij} \geq 0,$$

so K is closed convex cone. K has nonempty interior because it includes \mathbf{S}_+^n . K is pointed because $z^T X z \geq 0$ and $z^T (-X) z \geq 0$ gives $z^T X z = 0$ for all $z \succeq 0$, hence $X = 0$. Hence K is a proper cone.

The dual cone of K is the set of normal vectors of all halfspaces supporting K . It can be expressed as

$$K = \{zz^T | z \succeq 0\}$$

3.6 Functions and epigraphs. When is the epigraph of a function a halfspace? When is the epigraph of a function a convex cone? When is the epigraph of a function a polyhedron?

a) When is the epigraph of a function a halfspace?

Let the epigraph of $f : \mathbf{R}^n \rightarrow \mathbf{R}$ be $E = \{s = (x, t) | f(x) \leq t\}$. Because E is a halfspace, so $\exists A = (a, a^*)$ and b , where $a \in \mathbf{R}^n$, $a^* \in \mathbf{R}$, and $b \in \mathbf{R}$, s. t.

$$\begin{aligned} E &= \{s | A^T s \leq b\} \\ &= \{(x, t) | a^T x + a^* t \leq b\} \\ &= \{(x, t) | a^T x - b \leq -a^* t\} \end{aligned}$$

Because E has the form of epigraph, so the coefficient of t must be positive, i. e., $-a^* > 0$.

So $f(x) = \frac{a^T}{-a^*}x + \frac{b}{a^*}$, which is a affine function.

b) When is the epigraph of a function a convex cone?

Let the epigraph of $f : \mathbf{R}^n \rightarrow \mathbf{R}$ be $E = \{s = (x, t) | f(x) \leq t\}$. By definition for any $x \in \mathbf{R}^n$ and any $\alpha \in \mathbf{R}^+$, $(x, f(x)) \in E$ and $(\alpha x, f(\alpha x)) \in E$. Then because E is a cone, $(\alpha x, \alpha f(x)) \in E$ and $(\frac{1}{\alpha}(\alpha x), \frac{1}{\alpha}f(\alpha x)) \in E$, which indicates

$$\begin{cases} f(\alpha x) \leq \alpha f(x) \\ f(\frac{1}{\alpha}(\alpha x)) \leq \frac{1}{\alpha}f(\alpha x) \end{cases}$$

which is $f(\alpha x) \leq \alpha f(x)$ and $f(\alpha x) \geq \alpha f(x)$. So $f(\alpha x) = \alpha f(x)$. In addition, because the epigraph E of $f(x)$ is convex, $f(x)$ is convex function.

So, $f(x)$ is a convex function which holds the property of positively homogeneous, i. e., for any $x \in \mathbf{R}^n$ and $\alpha \in \mathbf{R}^+$, $f(\alpha x) = \alpha f(x)$.

c) When is the epigraph of a function a polyhedron?

Let the epigraph of $f : \mathbf{R}^n \rightarrow \mathbf{R}$ be $E = \{s = (x, t) | f(x) \leq t\}$. Because E is a polyhedron, so $\exists A_j = (a_j, a_j^*)$ and $b_j (j = 1, \dots, m)$, where $a_j \in \mathbf{R}^n$, $a_j^* \in \mathbf{R}$, and $b \in \mathbf{R}$, s. t.

$$\begin{aligned} E &= \{s | A_j^T s \leq b_j, j = 1, \dots, m\} \\ &= \{(x, t) | a_j^T x + a_j^* t \leq b_j, j = 1, \dots, m\} \\ &= \{(x, t) | a_j^T x - b_j \leq -a_j^* t, j = 1, \dots, m\} \end{aligned}$$

Because E has the form of epigraph, so the coefficient of t must be positive, i. e., $-a_j^* > 0$ for $j = 1, \dots, m$. So

$$\begin{aligned} E &= \{(x, t) | \frac{a_j^T}{-a_j^*}x + \frac{b_j}{a_j^*} \leq t, j = 1, \dots, m\} \\ &= \{(x, t) | \max_{j=1, \dots, m} \left(\frac{a_j^T}{-a_j^*}x + \frac{b_j}{a_j^*} \right) \leq t\} \end{aligned}$$

So $f(x) = \max_{j=1,\dots,m} \left(\frac{a_j^T}{-a_j^*} x + \frac{-b_j}{-a_j^*} \right)$, which is a piecewise affine function.

3.20 Composition with an affine function. Show that the following functions $f : \mathbf{R}^n \rightarrow \mathbf{R}$ are convex.

- (a) $f(x) = \|Ax - b\|$, where $A \in \mathbf{R}^{m \times n}$, $b \in \mathbf{R}^m$, and $\|\cdot\|$ is a norm on \mathbf{R}^m .
- (b) $f(x) = -(\det(A_0 + x_1 A_1 + \dots + x_n A_n))^{1/m}$, on $\{x \mid A_0 + x_1 A_1 + \dots + x_n A_n \succ 0\}$, where $A_i \in \mathbf{S}^m$.
- (c) $f(X) = \text{tr}(A_0 + x_1 A_1 + \dots + x_n A_n)^{-1}$, on $\{x \mid A_0 + x_1 A_1 + \dots + x_n A_n \succ 0\}$, where $A_i \in \mathbf{S}^m$. (Use the fact that $\text{tr}(X^{-1})$ is convex on \mathbf{S}_{++}^m ; see exercise 3.18.)

Using the conclusion of:

3.18 Adapt the proof of concavity of the log-determinant function in §3.1.5 to show the following.

- (a) $f(X) = \text{tr}(X^{-1})$ is convex on $\text{dom } f = \mathbf{S}_{++}^n$.
- (b) $f(X) = (\det X)^{1/n}$ is concave on $\text{dom } f = \mathbf{S}_{++}^n$.

a) $f(x)$ is a composition of a norm and an affine function, hence convex;

b) $f(x)$ is a composition of a convex function in 3.18 b) and an affine transformation, hence convex;

c) $f(X)$ is a composition of a convex function in 3.18 a) and an affine transformation, hence convex.

3.21 Pointwise maximum and supremum. Show that the following functions $f : \mathbf{R}^n \rightarrow \mathbf{R}$ are convex.

- (a) $f(x) = \max_{i=1,\dots,k} \|A^{(i)}x - b^{(i)}\|$, where $A^{(i)} \in \mathbf{R}^{m \times n}$, $b^{(i)} \in \mathbf{R}^m$ and $\|\cdot\|$ is a norm on \mathbf{R}^m .
- (b) $f(x) = \sum_{i=1}^r |x|_{[i]}$ on \mathbf{R}^n , where $|x|$ denotes the vector with $|x|_i = |x_i|$ (i.e., $|x|$ is the absolute value of x , componentwise), and $|x|_{[i]}$ is the i th largest component of $|x|$. In other words, $|x|_{[1]}, |x|_{[2]}, \dots, |x|_{[n]}$ are the absolute values of the components of x , sorted in nonincreasing order.

a) Let $f(x) = \max_{i=1,\dots,k} f_i(x)$. By 3.20 a) we get $f_i(x) = \|A^{(i)}x - b^{(i)}\|$ is convex. Hence as pointwise maximum of k convex function, $f(x)$ is also convex.

b) Write $f(x)$ as

$$f(x) = \sum_{i=1}^r |x|^{[i]} = \max_{1 \leq i_1 < i_2 < \dots < i_r \leq n} |x_{i_1}| + |x_{i_2}| + \dots + |x_{i_r}|$$

which is the pointwise maximum of $n!/r!(n-r)!$ convex functions, hence convex.