

Conic Programming and Duality (Lecture 18) ①

Separation theorem. Let $K \subseteq V$ be a closed convex cone and let $b \in V \setminus K$. Then there exists $y \in V$ s.t.

$$(y, x) \geq 0 \text{ for all } x \in K \text{ and } (y, b) < 0.$$

The separation theorem when applied to

$$K = \{Ax : x \in \mathbb{R}_+^n\} \subseteq V = \mathbb{R}^m \quad (\text{here } A \text{ is an } m \times n \text{ matrix})$$

implies the Farkas lemma in the LP setting.

For any $b \in \mathbb{R}^m$, exactly one of the following holds:

- either $Ax = b$, $x \geq 0$ has a solution $x \in \mathbb{R}_+^n$
- or $\underbrace{A^T y \geq 0}_{\text{or}} \text{, } b^T y < 0$ has a solution $y \in \mathbb{R}^m$.

The separation theorem implies that if the first system has no solution then there exists $y \in \mathbb{R}^m$ s.t.

$$y^T Ax \geq 0 + x \in \mathbb{R}_+^n \text{ and } y^T b < 0.$$

$$\underbrace{y^T Ax \geq 0}_{\text{This means }} + x \in \mathbb{R}_+^n \text{ and } y^T b < 0 \Rightarrow A^T y \in (\mathbb{R}_+^n)^* = \mathbb{R}_+^n, \text{ i.e., } A^T y \geq 0.$$

Our goal now is to generalize the Farkas lemma to deal with systems of the form $Ax = b$, $x \in K$. We would like to be able to claim the following:

Let $K \subseteq V$ be a closed convex cone, and

let $b \in W$. Then

- either $A(x) = b$, $x \in K$ has a solution $x \in V$
- or the system $A^T(y) \in K^*$, $(b, y) < 0$ has a solution, but not both.

This is a straightforward generalization of the Farkas lemma in the LP setting. However the above claim is not true.

Consider the following example: $K = \text{toppled ice cream cone in } \mathbb{R}^3$,
 and let $A = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ and $b = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \in \mathbb{R}^2$. ②

The first system asks for a point $(x, y, z) \in K$ such that $y=0$ and $z=1$. There is no such point since $xy \geq z^2$ is one of the constraints for K .

The second system asks for a point $(\alpha, \beta) \in \mathbb{R}^2$ such that $(\alpha \ \beta) \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \underbrace{(0 \ \alpha \ \beta)}_{\text{so } \beta^2 \leq 0 \Rightarrow \beta = 0} \in K^*$

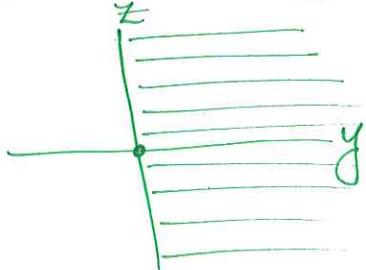
$$\text{and } ((0, 1), (\alpha, \beta)) = \underbrace{\beta}_{< 0}.$$

Thus both the systems are unsolvable.

So what is going wrong here? To apply the separation theorem, the cone $C = A(K) = \{A(x) : x \in K\}$ needs to be closed. Observe that this is not true in our example. Here C is the projection of K onto the $y-z$ plane. That is,

$$C = \{(y, z) \in \mathbb{R}^2 : (x, y, z) \in K\} = \{\vec{0}\} \cup \{(y \in \mathbb{R} : y > 0) \times \mathbb{R}\}$$

Thus C is



where the only point on \mathbb{Z} -axis that belongs to C is the origin. So C is not closed.

Hence we cannot apply the separation theorem on C .

To save the situation, we work with the closure of C .

The closure \bar{C} of C is the set of all limit points of C . Formally, $b \in \bar{C}$ if and only if there exists a sequence $(y_k)_{k \in \mathbb{N}}$ s.t. $y_k \in C$ for all k and $\lim_{k \rightarrow \infty} y_k = b$.

When C is a convex cone, \overline{C} is a closed convex cone.

Definition. Let $K \subseteq V$ be a closed convex cone.

The system $A(x) = b$, $x \in K$ is called limit-feasible if there exists a sequence $(x_k)_{k \in \mathbb{N}}$ such that $x_k \in K$ for all $k \in \mathbb{N}$ and $\lim_{k \rightarrow \infty} A(x_k) = b$.

So if $A(x) = b$, $x \in K$ is limit-feasible then $b \in \overline{C}$ where $C = \{A(x) : x \in K\}$. Moreover, if

$b \in \overline{C}$ then the system $A(x) = b$, $x \in K$ is limit-feasible.

In more detail, if $(y_k)_{k \in \mathbb{N}}$ is a sequence in C converging to b then any sequence $(x_k)_{k \in \mathbb{N}}$ s.t. $y_k = A(x_k)$ for all k proves the limit-feasibility of $A(x) = b$, $x \in K$.

The correct Farkas lemma (cone version)

Let $K \subseteq V$ be a closed convex cone and let $b \in W$.

- Either the system $A(x) = b$, $x \in K$ is limit-feasible
- Or the system $A^T(y) \in K^*$, $(b, y) \leq 0$ has a solution but not both.

Proof. If $A(x) = b$, $x \in K$ is limit-feasible

then we choose any sequence $(x_k)_{k \in \mathbb{N}}$ that proves its limit feasibility. For $y \in W$, we compute $(y, b) = (y, \lim_{k \rightarrow \infty} A(x_k)) = \lim_{k \rightarrow \infty} (y, A(x_k)) = \lim_{k \rightarrow \infty} (A^T(y), x_k)$.

(4)

If $A^T(y) \in K^*$ then $x_k \in K$ implies that
 $(A^T(y), x_k) \geq 0$ for all $k \in \mathbb{N}$. Thus we get
 $(y, b) \geq 0$. In other words, the second system
has no solution.

Now suppose $A(x) = b$, $x \in K$ is not limit-feasible.
This is the same as saying $b \notin \overline{C}$ where
 $C = \{A(x) : x \in K\}$. Since \overline{C} is a closed convex
cone, we can apply the separation theorem and
obtain a hyperplane that strictly separates b
from \overline{C} (and thus from C).

This means there exists $y \in W$ s.t. $(y, b) < 0$
and for all $x \in K$, $(y, A(x)) = \underbrace{(A^T(y), x)}_{\text{Equivalently, } A^T(y) \in K^*} \geq 0$.

Cone Programs

A cone program is an optimization problem of
the form $\max (c, x)$
s.t.

$$A(x) = b$$

$x \in K$ (K is a closed convex cone)

Note that we can be more general and say it
is an optimization problem of the form

$$\max (c, x)$$

s.t. $b - A(x) \in L$ (both L and K are
 $x \in K$ closed convex cones)

For the sake of simplicity, let us take $L = \{\vec{0}\}$.

(5)

So our primal program is the following:

$$\max (c, x)$$

s.t.

$$A(x) = b$$

$$x \in K.$$

Given a feasible sequence $(x_k)_{k \in \mathbb{N}}$ of the above cone program, we define its value as

$$(c, (x_k)_{k \in \mathbb{N}}) = \limsup_{k \rightarrow \infty} (c, x_k)$$

The limit value of the above cone program is defined as $\sup \{(c, (x_k)_{k \in \mathbb{N}}) : (x_k)_{k \in \mathbb{N}} \text{ is a feasible sequence of the cone program}\}$.

What is the dual program of the above primal program? Taking a cue from duality in linear programming, let us write it as:

$$\begin{aligned} & \min (b, y) \\ \text{s.t. } & A^T(y) = c + s \\ & s \in K^* \end{aligned} \quad \left. \begin{array}{l} \text{We can write this more succinctly as} \\ A^T(y) - c \in K^* \end{array} \right\}$$

Let us start with the weak duality theorem.

Theorem. If the dual program is feasible and if the primal program is limit-feasible then the limit value of the primal program \leq the value of the dual program.

Proof. Let y be any feasible solution of the dual program and let $(x_k)_{k \in \mathbb{N}}$ be any feasible

sequence of the primal program. We have:

$$0 \leq (\underbrace{A^T(y) - c}_{\in K^*}, \underbrace{x_k}_{\in K}) = (A^T(y), x_k) - (c, x_k)$$

$$= (y, A(x_k)) - (c, x_k) \text{ for } k \in \mathbb{N}$$

Hence $\limsup_{k \rightarrow \infty} (c, x_k) \leq \limsup_{k \rightarrow \infty} (y, A(x_k))$

$$= \lim_{k \rightarrow \infty} (y, A(x_k)) = (y, b).$$

Since the feasible sequence $(x_k)_{k \in \mathbb{N}}$ was arbitrary, this means that the limit value of the primal program is at most (y, b) . Since y was an arbitrary feasible solution of the dual program, the theorem follows. \square

Regular Duality

Theorem. The dual program is feasible and has a finite value β if and only if the primal program is limit-feasible and has a finite limit value r . Moreover, $\beta = r$.

Before we prove the above theorem, let us introduce some notation. Let V and W be real and finite-dimensional vector spaces.

$V \oplus W$, the direct sum of V and W , is the set $V \times W$, turned into a vector space with scalar product via $(x, y) + (x', y') = (x+x', y+y')$, $\lambda(x, y) = (\lambda x, \lambda y)$, and $((x, y), (x', y')) = \underbrace{(x, x')}_{\substack{\text{scalar} \\ \text{product}}} + \underbrace{(y, y')}_{\substack{\text{scalar} \\ \text{product}}}$.

(7)

Fact. Let $K \subseteq V$ and $L \subseteq W$ be closed convex cones.

Then $K \oplus L = \{(x, y) \in V \oplus W : x \in K, y \in L\}$
is again a closed convex cone.

Claim. Let $K \subseteq V$ and $L \subseteq W$ be closed convex cones.

Then $(K \oplus L)^* = K^* \oplus L^*$.

Proof of Regular Duality. If the dual program
is feasible and has value β , then we have:

$$A^T(y) - c \in K^* \Rightarrow \underbrace{(b, y)}_{\text{the inner product}} \geq \beta. \quad (1)$$

We also have $A^T(y) \in K^* \Rightarrow (b, y) \geq 0. \quad (2)$

Otherwise we could add a large positive multiple
of y to any dual feasible solution and obtain
a dual feasible solution of value smaller than β .

We can now merge (1) and (2) into the single
implication: $A^T(y) - zc \in K^*, z \geq 0 \Rightarrow (b, y) \geq z\beta.$

For $z > 0$, we obtain the above implication from (1)
by multiplying all terms with z and calling " zy ".

For $z=0$, this is simply (2).

In order to apply Farkas lemma, let us
rewrite the above implication as:

$$\left(\begin{array}{c|c} A^T & -c \\ \hline 0 & 1 \end{array} \right) (y, z) \in K^* \oplus \mathbb{R}_+ \Rightarrow \langle (b, -\beta), (y, z) \rangle \geq 0. \quad (3)$$

We are now ready to apply the Farkas lemma. We are
precisely in the situation where the second system has
no solution. So implication (3) holds if and only if

$$\left(\begin{array}{c|c} A & 0 \\ \hline -c^T & 1 \end{array} \right) (x, z) = (b, -\beta), (x, z) \in (K^* \oplus \mathbb{R}_+)^*$$

$$= K \oplus \mathbb{R}_+$$

is limit-feasible. This system is limit-feasible if and only if there are sequences $(x_k)_{k \in \mathbb{N}}, (z_k)_{k \in \mathbb{N}}$ with $x_k \in K$, $z_k \geq 0$ for all k such that

$$\lim_{k \rightarrow \infty} A(x_k) = b \quad \text{and} \quad \lim_{k \rightarrow \infty} \langle c, x_k \rangle - z_k = \beta.$$

This shows that the primal program is limit-feasible

This shows that the limit value of the primal program $\geq \beta$.

Moreover, weak duality shows that the limit value of the primal program $\leq \beta$. This proves the "only if" direction.

The "if" direction. Let the primal program be limit-feasible with finite limit value r and suppose the dual program is infeasible. This implies:

$$A^T(y) - z \in K^* \Rightarrow z \leq 0,$$

since if (y, z) violates it, then $\frac{1}{z} \cdot y$ would be a dual feasible solution. Let us write the above implication as:

$$(A^T | -c)(y, z) \in K^* \Rightarrow ((0, -1), (y, z)) \geq 0.$$

This means that the system $\left(\begin{array}{c|c} A & 0 \\ \hline -c^T & 1 \end{array} \right) (x) = (0, -1), x \in (K^*)^* = K$

is limit-feasible. This means that there are sequences $(x_k)_{k \in \mathbb{N}}$ with $x_k \in K$ for all k such that

$$\lim_{k \rightarrow \infty} A(x_k) = 0 \quad \text{and} \quad \lim_{k \rightarrow \infty} \langle c, x_k \rangle = 1.$$

But this is a contradiction as elementwise addition of $(x_k)_{k \in \mathbb{N}}$ to any feasible sequence of the primal program that attains the limit value r would result in a feasible sequence that witness limit value $\geq r + 1$. Thus the dual program must have been feasible. Easy to show $\beta = r$. \square