

$$\text{LP: } \min c^T x \quad \text{s.t. } Ax \geq b \quad x \in \mathbb{R}^n$$

$$\text{LD: } \max \gamma^T b$$

$$\text{s.t. } A^T \gamma = c \quad \gamma \in \mathbb{R}_+^m$$

Generalizations could be

by



Generalizing the
objective \dots to non-linear
(convex) objectives

$$\|x - c\|^2 \text{ or } x^T Q x + b$$

Generalize
inequalities
itself
the notion of

Generalizing constraints
to be nonlinear constraints

$$\|x - c\|^2 \leq t$$

This generali-
zation is
as powerful
as others

Consider linear programs (LP), dual of LP, conic programs & their duals

[Ref page 5 of

<http://www2.isye.gatech.edu/~nemirovs/ICMNemirovski.pdf>

LP affine objective

$$\min_{\mathbf{x} \in \mathbb{R}^n} \mathbf{c}^T \mathbf{x}$$

subject to $-\mathbf{A}\mathbf{x} + \mathbf{b} \leq 0$

Conic Program (CP)

$$\min_{\mathbf{x} \in \mathbb{R}^n} \mathbf{c}^T \mathbf{x}$$

subject to $-\mathbf{A}\mathbf{x} + \mathbf{b} \leq \mathbf{k}$

Let $\lambda \geq 0$ (i.e. $\lambda \in \mathbb{R}_+^n$)

Then $\lambda^T (-\mathbf{A}\mathbf{x} + \mathbf{b}) \leq 0$

$$\begin{aligned} \Rightarrow \bar{\mathbf{c}}^T \mathbf{x} &\geq \bar{\mathbf{c}}^T \mathbf{x} + \lambda^T (-\mathbf{A}\mathbf{x} + \mathbf{b}) \\ &= \bar{\mathbf{c}}^T \mathbf{x} + (\mathbf{c} - \bar{\mathbf{A}}^T \lambda)^T \mathbf{x} \\ &\geq \min \bar{\mathbf{c}}^T \mathbf{x} + (\mathbf{c} - \bar{\mathbf{A}}^T \lambda)^T \mathbf{x} \\ &= \begin{cases} \bar{\mathbf{c}}^T \mathbf{x} & \text{if } \bar{\mathbf{A}}^T \lambda = \mathbf{c} \\ -\infty & \text{if } \bar{\mathbf{A}}^T \lambda \neq \mathbf{c} \end{cases} \end{aligned}$$

independent of \mathbf{x}

$$\begin{aligned} \min_{\mathbf{x} \in \mathbb{R}^n} \bar{\mathbf{c}}^T \mathbf{x} \\ \text{s.t. } \mathbf{A}\mathbf{x} \geq \mathbf{b} \end{aligned} \quad \geq \quad \begin{aligned} \max_{\lambda \geq 0} \mathbf{b}^T \lambda \\ \text{s.t. } \bar{\mathbf{A}}^T \lambda = \mathbf{c} \end{aligned}$$

Primal LP
(lower bounded)

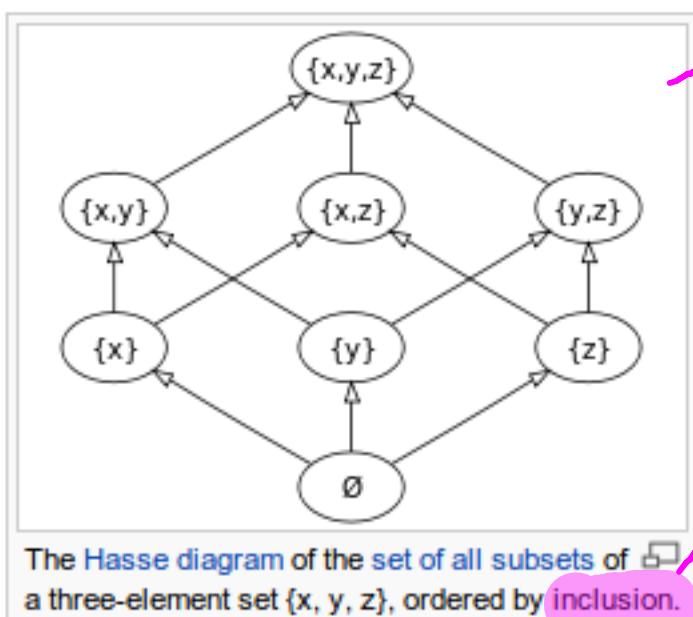
Dual LP
(upper bounded)

- Q: Had to generalise $-\mathbf{A}\mathbf{x} + \mathbf{b} \leq 0$ to $-\mathbf{A}\mathbf{x} + \mathbf{b} \leq \mathbf{k}$ s.t. \leq is a generalised inequality & \mathbf{k} some set?

- what properties should \mathbf{k} satisfy so that \mathbf{k} satisfies properties of generalized inequalities?

To prove that \geq_K being convex cone & pointed are necessary & sufficient conditions for \geq_K to be a valid inequality, recall that any partial order \geq_K should satisfy the following properties (refer page 51 of www2.isye.gatech.edu/~nemirovs/Lect_ModConvOpt.pdf i.e. Section 1.4.1)

1. Reflexivity: $a \geq a$;
2. Anti-symmetry: if both $a \geq b$ and $b \geq a$, then $a = b$;
3. Transitivity: if both $a \geq b$ and $b \geq c$, then $a \geq c$;
4. Compatibility with linear operations:
 - (a) Homogeneity: if $a \geq b$ and λ is a nonnegative real, then $\lambda a \geq \lambda b$
("One can multiply both sides of an inequality by a nonnegative real")
 - (b) Additivity: if both $a \geq b$ and $c \geq d$, then $a + c \geq b + d$
("One can add two inequalities of the same sign").



→ example partial order \subseteq over sets
(source: http://en.wikipedia.org/wiki/Partially_ordered_set)

that is, the \subseteq partial order

~~Proof:~~

a) K being pointed convex cone \Rightarrow

\geq_K is a partial order

① $a \geq_K a$ since
reflexivity

$a - a = 0 \in K$ ($\because K$ is cone)

② If $a \geq_K b$ & $b \geq_K a$ then $a = b$ {anti-symmetry}
since

$a - b \in K$ and $b - a \in K \Rightarrow b - a = 0$
($\because K$ is pointed)

③ If both $a \geq_K b$ and $b \geq_K c$ then $a \geq_K c$
since Transitivity

$a - b \in K$ and $b - c \in K \Rightarrow (a - b) + (b - c) \in K$
 $\underline{\underline{ie}} a - c \in K$
($\because K$ is a convex cone)

④ a) If $a \geq_K b$ and $\lambda \geq 0$ then $\lambda a \geq_K \lambda b$
since Homogeneity

If $a - b \in K$ & $\lambda > 0$ then $\lambda(a - b) \in K$
($\because K$ is a cone)

b) If both $a \geq_K b$ & $c \geq_K d$ then $a + c \geq_K b + d$
Additivity

since

If both $a-b \in K$ & $c-d \in K$ then

$$(a-b) + (c-d) = (a+c) - (b+d) \in K$$

($\because K$ is a convex cone)

b) \geq_K is a partial order $\Rightarrow K$ is a pointed convex cone

① If $x, y \in K$ then $\theta_1 x + \theta_2 y \in K$

$$\forall \theta_1, \theta_2 \geq 0 \quad (K \text{ is a convex cone})$$

since

If $x \geq_K 0$ & $y \geq_K 0$ then $\theta_1 x \geq_K 0 \quad \forall \theta_1 \geq 0$

and $\theta_2 y \geq_K 0 \quad \forall \theta_2 \geq 0$ (Homogeneity of \geq_K)

and thus $\theta_1 x + \theta_2 y \geq_K 0$ (Additivity of \geq_K)

② If $x \in K$ & $-x \in K$ then $x = 0$

(K is pointed)

since

If $x \geq_K 0$ & $-x \geq_K 0$ then

$0 \geq_K x$ ($x \geq_K x$ by reflexivity and adding
 $x \geq_K x$ & $-x \geq_K 0$ by additivity)

and $-x \geq_K x$ ($-x \geq_K 0$ & $0 \geq_K x$ just derived
and adding -.. by additivity)

and similarly $x \geq_K -x$ (similar use of
additivity & reflectivity)

and $-x \geq_K x$ & $x \geq_K -x \Rightarrow x = -x$ (by
anti-symmetry)

which is $x + x = 2x = 0$ ie $x = 0$

Questions: (Additional properties over & above K being
pointed convex cone)

① Suppose $a^i \geq_K b^i \forall i$ & $a^i \rightarrow a$ & $b^i \rightarrow b$

Then for $a \geq_K b$ what more is reqd
of K ?

↳ Ans: Necessary condition is that

$$a^i - b^i \rightarrow a - b \in K$$

i.e. K is closed (Also happens to be a
sufficient condition)

② What is reqd so that $\exists a \geq_K b$ (ie $b \geq_K a$)?

(H/W) ↳ Ans: Sufficient condition is that $a - b \in \text{int}(K)$
i.e. $\text{int}(K) \neq \emptyset$ OR K has non-empty interior

We will motivate through linear programming (LP) the concept of generalised inequalities:

$$\min_{x \in \mathbb{R}^n} c^T x$$

$$\text{subject to } -Ax + b \leq 0$$

LINEAR PROGRAM

can be rewritten as

$$Ax \geq b \text{ or } Ax - b \in \mathbb{R}_+^n$$

Note: \mathbb{R}_+^n is a CONE. How abt defining generalised inequality for a cone K as: $c \geq_d d$ iff $c-d \in K$ and a general conic program as:

$$\min_{x \in \mathbb{R}^n} c^T x$$

$$\text{subject to } -Ax + b \leq 0$$

CONIC PROGRAM

That is, $Ax - b \in K$
 K is a proper cone

Generalized inequalities

a convex cone $K \subseteq \mathbb{R}^n$ is a proper cone if

Also referred to as a regular cone
 } Some restrictions
 } on K that we will require. H/w: WHY?

- K is closed (contains its boundary)
- K is solid (has nonempty interior)
- K is pointed (contains no line)

examples

- nonnegative orthant $K = \mathbb{R}_+^n \subseteq \{x \in \mathbb{R}^n \mid x_i \geq 0, i = 1, \dots, n\}$
- positive semidefinite cone $K = \mathbb{S}_+^n$
- nonnegative polynomials on $[0, 1]$:

$$K = \{x \in \mathbb{R}^n \mid x_1 + x_2 t + x_3 t^2 + \dots + x_n t^{n-1} \geq 0 \text{ for } t \in [0, 1]\}$$

Q: What if $n \rightarrow \infty$... can you get proper cones under additional constraints?

Consider linear programs (LP), dual of LP, conic programs & their duals

[Ref page 5 of

<http://www2.isye.gatech.edu/~nemirovs/ICMNemirovski.pdf>

LP affine objective

$$\begin{aligned} \min_{\mathbf{x} \in \mathbb{R}^n} \quad & \mathbf{c}^T \mathbf{x} \\ \text{subject to} \quad & -A\mathbf{x} + \mathbf{b} \leq 0 \end{aligned}$$

Conic Program (CP)

$$\begin{aligned} \min_{\mathbf{x} \in \mathbb{R}^n} \quad & \mathbf{c}^T \mathbf{x} \\ \text{subject to} \quad & -A\mathbf{x} + \mathbf{b} \leq \mathbf{k} \end{aligned}$$

Let $\lambda \geq 0$ (i.e. $\lambda \in \mathbb{R}_+^n$)

Then $\lambda^T(-A\mathbf{x} + \mathbf{b}) \leq 0$

$$\begin{aligned} \Rightarrow \mathbf{c}^T \mathbf{x} &\geq \mathbf{c}^T \mathbf{x} + \lambda^T(-A\mathbf{x} + \mathbf{b}) \\ &= \lambda^T \mathbf{b} + (\mathbf{c} - A^T \lambda)^T \mathbf{x} \\ &\geq \min \lambda^T \mathbf{b} + (\mathbf{c} - K^* \lambda)^T \mathbf{x} \end{aligned}$$

$$\begin{aligned} &= \begin{cases} \lambda^T \mathbf{b} & \text{if } A^T \lambda = \mathbf{c} \\ -\infty & \text{if } A^T \lambda \neq \mathbf{c} \end{cases} \\ &\text{independent of } \mathbf{x} \end{aligned}$$

$$\begin{aligned} \min_{\mathbf{x} \in \mathbb{R}^n} \quad & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} \quad & A\mathbf{x} \geq \mathbf{b} \end{aligned} \geq \max_{\lambda \geq 0} \quad b^T \lambda$$

s.t. $A^T \lambda = \mathbf{c}$

Primal LP
(lower bounded)
Dual LP
(upper bounded)

K is a regular [proper cone]
Generalised cone program

$$\begin{aligned} \min_{\mathbf{x} \in V} \quad & \langle \mathbf{c}, \mathbf{x} \rangle_V \\ \text{subject to} \quad & A\mathbf{x} - \mathbf{b} \in K \end{aligned}$$

We need an equivalent
 $\lambda \in D \subseteq K^*$ s.t.

$$\langle \lambda, A\mathbf{x} - \mathbf{b} \rangle \geq 0$$

This K^* s.t.

$$D = \{ \lambda \mid \langle \lambda, A\mathbf{x} - \mathbf{b} \rangle \geq 0, \lambda \in V \wedge A\mathbf{x} - \mathbf{b} \in K \}$$

& $D \subseteq K^*$ is DUAL CONE
of K !

by dual) by primal)

Called the weak duality theorem for Linear Program

$K^* = \{\lambda : \lambda^T \xi \geq 0 \forall \xi \in K\}$ is the cone dual to K
{defn on page 7 of <http://www2.isye.gatech.edu/~nemirovs/ICMNemirovski.pdf>}

With this, follows weak duality theorem for CONIC PROGRAM

$$\min_{x \in V} \{c, x\}$$
$$\text{s.t. } Ax \geq b$$

$$\max_{\lambda \in K^*} \langle b, \lambda \rangle$$
$$\text{s.t. } A^T \lambda = c$$

Primal CP
(lower bounded by dual)

Dual CP
(upper bounded by primal)

- Notes:
- ① Both LP & CP dealt with affine objective
 - ② CP dealt with the generalised conic inequalities
 - ③ Later, in convex programs, we will deal with the more general convex functions in the objective

Notes:

① If $K = \mathbb{R}_+^n$, the CP is an LP

If $K = S_+^n$, the CP is an SDP

Set of all $n \times n$ symmetric positive semi-definite matrices

② Any generic convex program can be expressed as a cone program (CP)

① If K is a closed convex cone then $K^{**} = K$
 More generally $K^{**} = \text{closure}(K)$ (abbreviated as
 if K is just a convex cone $\text{cl}(K)$)

Proof: We will prove that if K is closed then

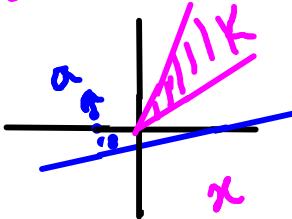
$$K^{**} = K$$

a) $K \subseteq K^{**}$ since $x \in K \Rightarrow \langle x, y \rangle \geq 0 \forall y \in K$
 $\Rightarrow x \in K^{**}$

b) $K^{**} \subseteq K$... We will prove by contradiction
 Suppose $x \in K^{**}$ but $x \notin K$

$\hookrightarrow K^{**}$ is closed since any dual cone
 is intersection of half spaces that
 are closed

$\hookrightarrow \{x\}$ is a singleton set



claim: $b=0$ if
 V is a closed convex cone

\Rightarrow By "strict separating hyperplane theorem"

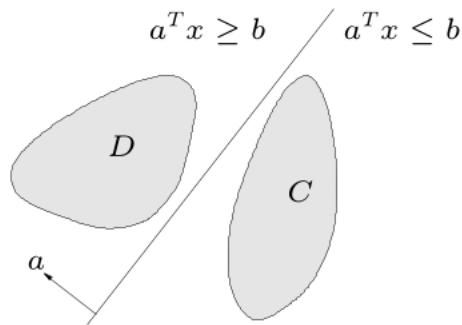
(on next page and proved later)

$\exists a \in V \wedge b \in \mathbb{R}$ s.t. $\langle a, x \rangle < b \wedge \langle a, y \rangle \geq b \forall y \in K$
 (since $y = 0 \in K^{**}$) $\Rightarrow \langle a, x \rangle < 0 \leq \langle a, y \rangle \forall y \in K$
 $\Rightarrow a \in K^{**} \wedge x \notin K^{**}$ [contradiction]

Separating hyperplane theorem

if C and D are disjoint convex sets, then there exists $a \neq 0, b$ such that

$$a^T x \leq b \text{ for } x \in C, \quad a^T x \geq b \text{ for } x \in D$$



the hyperplane $\{x \mid a^T x = b\}$ separates C and D

strict separation requires additional assumptions (e.g., C is closed, D is a singleton)

Convex sets

2-19

Consequence

Supporting hyperplane theorem

supporting hyperplane to set C at boundary point x_0 :

$$\{x \mid a^T x = a^T x_0\}$$

where $a \neq 0$ and $a^T x \leq a^T x_0$ for all $x \in C$



supporting hyperplane theorem: if C is convex, then there exists a supporting hyperplane at every boundary point of C

② In fact, if K is a proper cone then
 K^* is also proper