

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

③ A more general notation for CP

CP: $\min_x c^T x$

\geq CD: $\max_{\lambda} \langle b, \lambda \rangle$

s.t. $Ax - b \in K, x \in \mathbb{R}^n$

s.t. $A^* \lambda = c, \lambda \in K^*$

$A: \mathbb{R}^n \rightarrow V$ is a linear map

$A^*: V \rightarrow \mathbb{R}^n$ is the adjoint

(Eg: V could be S^n
 $K \subseteq V$ could be S_+^n)

linear map of $A: \mathbb{R}^n \rightarrow V$

$\equiv \langle Au, v \rangle_V = \langle u, A^*v \rangle_{\mathbb{R}^n}$

$\forall u \in \mathbb{R}^n \text{ \& \; } \forall v \in V$

H/W: Could the implications we will derive hold if \mathbb{R}^n is further generalized

Eg: if A is matrix multiplication then $A^* = A^T$ (verify)

④

$$\min_{\lambda \in K^*} \langle \lambda, x \rangle = \begin{cases} 0 & \text{if } x \in K, \text{ since } 0 \in K^* \\ -\infty & \text{if } x \notin K \end{cases}$$

Using this result, we can get the conic program as the dual of its dual (assuming $K^{**} = K$)

⑤ Consider Conic dual

$$d^* = \max_{\lambda} \langle b, \lambda \rangle$$

$$A^* \lambda = c, \lambda \in K^*$$

$$d^* = \max_{\lambda} \min_{\substack{y \in \mathbb{R}^n \\ x \in K^{**} = K}} \langle b, \lambda \rangle + \langle y, (c - A^* \lambda) \rangle + \langle x, \lambda \rangle$$

s.t. $\lambda \in K^*$
 $A^* \lambda = c$

from pt ④, $\min_{x \in K^*} \langle a, \lambda \rangle = 0$
 if $\lambda \in K^*$

$$\leq \max_{\lambda} \min_{\substack{y \in \mathbb{R}^n \\ x \in K}} \langle b, \lambda \rangle + \langle y, (c - A^* \lambda) \rangle + \langle x, \lambda \rangle \quad \text{[Max min inequality]}$$

$\min_y f(x, y) \leq f(x, y)$
 $\Rightarrow \max_x \min_y f(x, y) \leq \max_x f(x, y) \quad (\forall y)$
 $\Rightarrow \max_x \min_y f(x, y) \leq \min_y \max_x f(x, y)$

$$\leq \min_{\substack{y \in \mathbb{R}^n \\ x \in K}} \max_{\lambda} \langle b, \lambda \rangle + \langle y, (c - A^* \lambda) \rangle + \langle x, \lambda \rangle$$

$$\max_{\lambda} \langle b - Ay + x, \lambda \rangle + \langle y, c \rangle$$

$$\because \langle y, A^* \lambda \rangle = \langle Ay, \lambda \rangle \quad \forall y \in \mathbb{R}^n \ \& \ \lambda \in V$$

$= \langle y, c \rangle$ if $b - Ay + x = 0$
 i.e. if $Ay - b \in K$

$= \infty$ o/w

$$= \min_{y \in \mathbb{R}^n} \langle y, c \rangle$$

s.t. $Ay - b \in K$

Dual cones and generalized inequalities

dual cone of a cone K :

$$K^* = \{y \mid y^T x \succeq 0 \text{ for all } x \in K\}$$

examples

- $K = \mathbf{R}_+^n$: $K^* = \mathbf{R}_+^n$
- $K = \mathbf{S}_+^n$: $K^* = \mathbf{S}_+^n$
- $K = \{(x, t) \mid \|x\|_2 \leq t\}$: $K^* = \{(x, t) \mid \|x\|_2 \leq t\}$
- $K = \{(x, t) \mid \|x\|_1 \leq t\}$: $K^* = \{(x, t) \mid \|x\|_\infty \leq t\}$

first three examples are **self-dual** cones

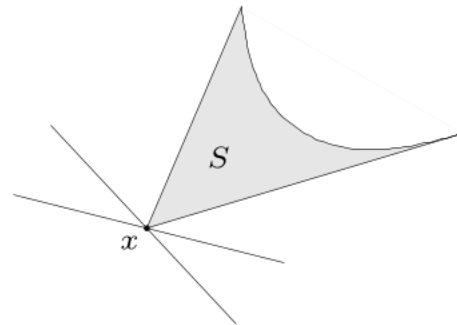
dual cones of proper cones are proper, hence define generalized inequalities:

$$y \succeq_{K^*} 0 \iff y^T x \geq 0 \text{ for all } x \succeq_K 0$$

Minimum and minimal elements via dual inequalities

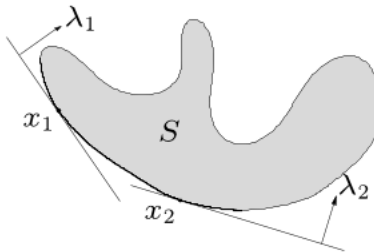
minimum element w.r.t. \preceq_K

x is minimum element of S iff for all $\lambda \succeq_{K^*} 0$, x is the unique minimizer of $\lambda^T z$ over S



minimal element w.r.t. \preceq_K

- if x minimizes $\lambda^T z$ over S for some $\lambda \succeq_{K^*} 0$, then x is minimal



- if x is a minimal element of a convex set S , then there exists a nonzero $\lambda \succeq_{K^*} 0$ such that x minimizes $\lambda^T z$ over S

lie + yes, x, y

FROM DUAL OF NORM CONE TO DUAL NORM

Let $\|\cdot\|$ be a norm on \mathbb{R}^n
The dual of $K = \{(x, t) \in \mathbb{R}^{n+1} \mid \|x\| \leq t\}$
is $K^* = \{(u, v) \in \mathbb{R}^{n+1} \mid \|u\|_* \leq v\}$

Where

$$\|u\|_* = \sup \{u^T x \mid \|x\| \leq 1\}$$

Proof: We need to show that

$$x^T u + tv \geq 0 \text{ whenever } \|x\| \leq t \iff \|u\|_* \leq v. \quad (2.20)$$

Let us start by showing that the righthand condition on (u, v) implies the lefthand condition. Suppose $\|u\|_* \leq v$, and $\|x\| \leq t$ for some $t > 0$. (If $t = 0$, x must be zero, so obviously $u^T x + vt \geq 0$.) Applying the definition of the dual norm, and the fact that $\| -x/t \| \leq 1$, we have

$$u^T (-x/t) \leq \|u\|_* \leq v,$$

and therefore $u^T x + vt \geq 0$.

Next we show that the lefthand condition in (2.20) implies the righthand condition in (2.20). Suppose $\|u\|_* > v$, i.e., that the righthand condition does not hold. Then by the definition of the dual norm, there exists an x with $\|x\| \leq 1$ and $x^T u > v$. Taking $t = 1$, we have

$$u^T (-x) + v < 0,$$

which contradicts the lefthand condition in (2.20).

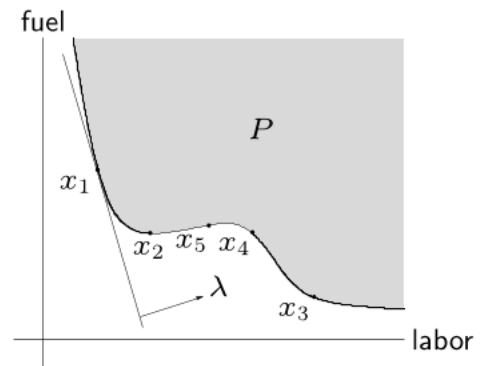
(proof from Boyd)

optimal production frontier

- different production methods use different amounts of resources $x \in \mathbf{R}^n$
- production set P : resource vectors x for all possible production methods
- efficient (Pareto optimal) methods correspond to resource vectors x that are minimal w.r.t. \mathbf{R}_+^n

example ($n = 2$)

x_1, x_2, x_3 are efficient; x_4, x_5 are not



① LP is only a special case of CP
with $K = \boxed{\mathbb{R}^n_+}$

② When will solution to CP = solution to CD?

CP: $\min \langle c, x \rangle$
 $x \in \mathbb{R}^n$
s.t. $Ax \geq_K b$

CD: $\max \langle b, \lambda \rangle$
 $\lambda \in K^*$
s.t. $A^* \lambda = c$

Since $K^{**} = K$, we saw that dual of CD is CP
While there exist multiple ways of writing CP & CD, hereafter we pick another standard format (to help you get used to various representations)

CP: $\min \langle c, x \rangle_V$
s.t. $Ax = b$
 $x \in K \subseteq V$
 $F_P \quad A: V \rightarrow \mathbb{R}^n$

CD: $\max \langle b, \lambda \rangle_{\mathbb{R}^n}$
s.t. $c - \lambda^T A \in K^*$
 $\lambda \in \mathbb{R}^n$
 $F_D \quad K^* \subseteq V$

Examples:

① LP

$$\text{LP: } \min_x \quad c^T x \\ \text{s.t. } Ax = b \\ x \geq 0$$

$$\text{Dual: } \max_{\lambda} \quad \lambda^T b \\ \text{s.t. } c - \lambda^T A \in \mathbb{R}_+^n \\ \lambda \in \mathbb{R}^m$$

② Semi-definite program

$$\text{SDP: } \inf_x \langle c, x \rangle_F \\ \text{(i=1..m) s.t. } \langle A_i, X \rangle = b_i \\ A_i \in S^n, X \in S_n^+$$

$$\text{SDP } \sup_{\lambda} \quad \lambda^T b \\ \text{s.t. } c - \sum_{i=1}^m \lambda^T A_i \in S_n^+ \\ \lambda \in \mathbb{R}^m$$

Weak duality theorem for CP

$$\langle c, x \rangle_V \geq \langle b, \lambda \rangle_{\mathbb{R}^n} \text{ where } x \in F_P \quad \lambda \in F_D$$

If $V = \mathbb{R}^m$ & $V_b = \mathbb{R}^m$ & $\langle \cdot, \cdot \rangle$ is dot product

$$c^T x - b^T \lambda = \underbrace{(c - A^T \lambda)^T}_{\text{duality gap}} x \geq 0$$

(we have proved this twice: from CP \rightarrow CD & from CD \rightarrow CP)

$\langle c, x \rangle_V - \langle b, \lambda \rangle_{\mathbb{R}^n}$ is called duality gap

Corollary ① If CP or CD is feasible but unbounded, then the other is infeasible or has no feasible soln

Eg: $\min x \text{ st } x \in (0, 10]$

② If a pair of feasible solutions can be found to the primal & dual problems with equal objective value then they are both optimal

Proof: let $\langle c, x^* \rangle_V = \langle b, \lambda^* \rangle \rightarrow \textcircled{\text{I}}$
st $x^* \in F_P$ & $\lambda^* \in F_D$

But we know: $\langle c, x^* \rangle \geq \min_{x \in F_P} \langle c, x \rangle \geq \max_{\lambda \in F_D} \langle b, \lambda \rangle \geq \langle b, \lambda^* \rangle \rightarrow \textcircled{\text{II}}$

Because of $\textcircled{\text{I}}$, all inequalities should become equalities in $\textcircled{\text{II}}$
 $\Rightarrow \min_{x \in F_P} \langle c, x \rangle = \max_{\lambda \in F_D} \langle b, \lambda \rangle$

Q: Does "STRONG DUALITY" hold for CP?
when?

$$\textcircled{a} \quad C = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad A_1 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad A_2 = \begin{bmatrix} 0 & -1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$
$$b = \begin{bmatrix} 0 \\ 2 \end{bmatrix} \quad K = S_+^3$$

in an SDP?

$$\textcircled{b} \quad C = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \quad A_1 = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \quad A_2 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad b = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$
$$K \in S_+^2$$

in an SDP?

STRONG DUALITY THM:

- ① Let CP or CD be infeasible & let other be feasible & have an interior. Then the other is unbounded
- ② Let CP and CD be both feasible, and let one of them have an interior. Then there is 0 duality gap
- ③ Let CP and CD be both feasible and have interior. Then both have optimal solutions with 0 duality gap

Proof:

We need the theorem of alternatives to prove strong conic duality [Also called Farkas' Lemma for Convex Cone]

Theorem of alternatives:

Consider $\{x \mid Ax=b, x \in K\}$ for a proper cone $K \subseteq V$
& $A: V \rightarrow \mathbb{R}^n$

Suppose $\exists \lambda$ s.t. $-A^* \lambda \in \text{int}(K^*)$. Then

- (a) $\left\{x \mid Ax=b, x \in K\right\}$ has a feasible soln x iff
- (b) $\left\{\lambda \mid -A^* \lambda \in K^*, \langle b, \lambda \rangle > 0\right\}$ has no feasible solution

PROOF:

(i) $C = \{y = Ax \in \mathbb{R}^m, x \in K\}$ is a closed convex set

(ii) Let $\bar{\lambda}$ be s.t. $-A^* \bar{\lambda} \in K^*$ and let $\{x \mid Ax=b, x \in K\}$ have a feasible solution \bar{x}

$$\Rightarrow -\langle \bar{\lambda}, b \rangle = -\langle \bar{\lambda}, A\bar{x} \rangle$$

.....
..... $\equiv \left\{\lambda \mid -A^* \lambda \in K^*, \langle b, \lambda \rangle > 0\right\}$ has no solution

(iii) Let $\{x \mid Ax=b, x \in K\}$ have no feasible solution
ie $b \notin C$

We will show that $\{\lambda \mid -A^*\lambda \in K, \langle \lambda, b \rangle > 0\}$ must be non-empty

Since C is a closed convex set, from the strict separating hyperplane theorem, $\exists \lambda \in \mathbb{R}^m$ s.t

$$\langle \lambda, b \rangle > \langle \lambda, y \rangle \quad \forall y \in C$$

Since $\exists x \in K$ s.t $Ax = y$ for any $y \in C$

$$\langle \lambda, b \rangle > \langle \lambda, Ax \rangle = \langle A^*\lambda, x \rangle \quad \forall x \in K$$

• Thus, $\langle A^*\lambda, x \rangle$ is bounded above $\forall x \in K$

• Since $0 \in K$, $\langle A^*\lambda, 0 \rangle > 0$

• Additionally, it must be that $\langle A^*\lambda, x \rangle \leq 0 \quad \forall x \in K$.

Otherwise if $\exists x \in K$ s.t $\langle A^*\lambda, x \rangle > 0$ then

if $\alpha \rightarrow +\infty$ then $\langle A^*\lambda, \alpha x \rangle \rightarrow \infty$ contradicting

that $\langle A^*\lambda, x \rangle$ is bounded above for all x

• Since $\langle A^*\lambda, x \rangle \leq 0 \quad \forall x$

$$\langle A^*\lambda, x \rangle \geq 0 \quad \forall x \Rightarrow \langle A^*\lambda, x \rangle = 0 \quad \forall x \in K^*$$

• Thus, λ is a (feasible) solution for \dots

\dots complete the proof \dots