

Corollary f Let $K \subseteq \mathbb{R}^n$ be a proper convex cone, let $A \in \mathbb{R}^{m \times n}$ be a matrix and $c \in \mathbb{R}^m$ be a vector. Then exactly one of the following two alternatives holds true:

- (i) $\exists x \in K \setminus \{0\} : Ax = 0 \text{ and } c^T x \geq 0$
- (ii) $\exists y \in \mathbb{R}^m : A^T y - c \in \text{int } K^*$.

§ 5 Duality theory

$$(P) \quad p^* = \sup_{\substack{x \in K \\ Ax = b}} c^T x$$

$$(D) \quad d^* = \inf_{\substack{y \in \mathbb{R}^m \\ A^T y - c \in K^*}} b^T y$$

Def. 1 The difference $d^* - p^*$ (which is nonnegative) is called duality gap.

Theorem 2 (a) (weak duality)

Suppose x is feasible for (P) and y is feasible for (D), then $c^T x \leq b^T y$. In particular, $p^* \leq d^*$.

(b) (complementary slackness) $p^* = d^*$

Let x be optimal for (P) and let y be optimal for (D).

Then $(A^T y - c)^T x = 0$.

(c) (optimality condition)

Let x be feasible for (P) and let y be feasible for (D).

(x, y) is optimal if and only if $(A^T y - c)^T x = 0$.

(d) (strong duality)

Suppose $d^* > -\infty$ and (D) is strictly feasible. Then there is an optimal solution of (P), i.e. there is a feasible x with $c^T x = p^*$, and $p^* = d^*$ holds.

Suppose $p^* < \infty$ and (P) is strictly feasible. Then there is a feasible solution y of (D) with $d^* = b^T y$, and $p^* = d^*$ holds.

Remark (c): is related to the KKT-condition

(KKT = Karush-Kuhn-Tucker)

(d): if (P) & (D) are both strictly feasible, then we say that Slater's condition is fulfilled.

Proof

$$\underline{(a) - (c)} : 0 \leq (A^T y - c)^T x = y^T A x - c^T x = y^T b - c^T x.$$

(d) Suppose $d^* > -\infty$ and (D) is strictly feasible.

To show: $\exists x^* \in K : Ax^* = b, c^T x^* \geq d^*$.

1st case : $b = 0$.

Then $d^* = 0$. Set $x^* = 0$.

2nd case : $b \neq 0$.

Consider the non-empty convex set

$$M = \{ A^T y - c : y \in \mathbb{R}^m, b^T y \leq d^* \}.$$

We have

$$M \cap \text{int } K^* = \emptyset.$$

[Otherwise there would exist a $y \in \text{int } K^*$ with $y^T b = d^*$ and an $\varepsilon > 0$ such that $B(y, \varepsilon) = \{ z : \|z - y\| \leq \varepsilon \} \subseteq K^*$.]

For $y - \varepsilon \frac{b}{\|b\|} \in K^*$ we would get

$$b^T \left(y - \varepsilon \frac{b}{\|b\|} \right) = b^T y - \varepsilon \|b\| < d^*. \quad \leftarrow [$$

By the separation theorem, Theorem 4.6, there exists an $x \in \mathbb{R}^n \setminus \{0\}$ such that

$$\sup_{z \in M} x^T z \leq \inf_{z \in \text{int } K^*} x^T z = \inf_{z \in K^*} x^T z \quad (*)$$

hold.

Claim: 1) $x \in K$

2) $\exists \mu > 0 : Ax = \mu b$ and $c^T x \geq \mu d^*$

3) $x^* = \frac{x}{\mu}$ is optimal for (P)

3) follows from 1) and 2) and weak duality.

1) Show: $\inf_{z \in K^*} x^T z \geq 0$. Then $x \in (K^*)^* = K$.

Suppose $\exists z \in K^*$ with $x^T z < 0$. Then $x^T(tz) \rightarrow -\infty$ for $t \rightarrow +\infty$ in contradiction to (*) and $d^* > -\infty$ ($M \neq \emptyset$).

Since $0 \in K^*$, claim 1) implies $\inf_{z \in K^*} x^T z = 0$.

Hence, $\sup_{z \in M} x^T z = 0$. That means

$b^T y \leq d^*$ implies $x^T(A^T y - c) \leq 0 \iff x^T A^T y \leq x^T c$.

So: The half space $\{y : b^T y \leq d^*\}$ is contained in the half space $\{y : (Ax)^T y \leq x^T c\}$.

Hence, b and Ax are linearly dependent and point in the same direction: $\exists \mu \geq 0 : Ax = \mu b$ and $\mu d^* \leq x^T c$ (**)

Assume $\mu = 0$. (D) is strictly feasible: $\exists y' : A^T y' - c \in \text{int } K'$

Using Exercise 2.4 gives

$$0 < (A^T y' - c)^T x = (y')^T Ax - c^T x = (y')^T \mu b - c^T x \stackrel{(\mu=0)}{=} -c^T x,$$

and so $c^T x < 0$ in contradiction to (**). \square

The second statement of (d) follows from the first by taking the dual.

Some examples

Example 3 d^* not attained

$$d^* = \sup_{\substack{X \in S_{\geq 0}^2}} \langle E_{12}, X \rangle$$

$$X \in S_{\geq 0}^2$$

$$\langle E_{11}, X \rangle = 1, \langle E_{22}, X \rangle = 0$$

$$d^* = \inf_{y_1} y_1$$

$$\underbrace{y_1 E_{11} + y_2 E_{22} - E_{12}}_{= \begin{bmatrix} y_1 & -\frac{1}{2} \\ -\frac{1}{2} & y_2 \end{bmatrix}} \geq 0$$

We have $p^* = d^* = 0$, but d^* is not attained.

Problem: (P) is not strictly feasible.

Example 4 positive duality gap.

$$p^* = \sup_{X \in S_{\geq 0}^3} \langle -E_{11} - E_{22}, X \rangle$$

$$X \in S_{\geq 0}^3$$

$$\langle E_{11}, X \rangle = 0$$

$$\langle E_{22} + 2E_{13}, X \rangle = 1$$

$$d^* = \inf y_2$$

$$y_1 E_{11} + y_2 (E_{22} + 2E_{13})$$

$$+ E_{11} + E_{22} \geq 0$$

$$= \begin{bmatrix} y_1 + 1 & 0 & y_2 \\ 0 & y_2 + 1 & 0 \\ y_2 & 0 & 0 \end{bmatrix} \geq 0$$

Every feasible solution of (P) satisfies

$$X_{11} = 0 = X_{13}, X_{22} = 1.$$

$$\text{Hence, } p^* = -1.$$

Every feasible solution of (D) satisfies $y_2 = 0$. Hence, $d^* = 0$.