

- Lecture 13. TuTh

Proof of Strong Duality and Complementary Slackness.

(Lecture 14 Economic significance of dual variables .)

- Proof of Strong duality & CS.

Recall Can write the initial dictionary

$$z = \sum_{j=1}^n c_j x_j$$

$$x_{n+i} = b_i - \sum_{j=1}^n a_{ij} x_j, \quad i=1, \dots, m$$

Slack
variables

$$\vec{x}_{\text{Slack}} = \vec{b} - A \vec{x}$$

Question: Choose a wrong statement:

~~A) Consider a standard form LP problem with n decision variables x_1, x_2, \dots, x_n , and m slack variables $x_{n+1}, x_{n+2}, \dots, x_{n+m}$. After applying the simplex method starting from a feasible dictionary, we found its optimal solution, moreover, in the final dictionary, the objective function is written in the form $z = z^* + \sum_{k=1}^{n+m} c_k^* x_k$. Then, $c_k^* \leq 0$ for all k , and z^* is the optimal value.~~

B) Let $\vec{c} \in \mathbb{R}^n$ and $d \in \mathbb{R}$ are given. If $\vec{c} \cdot \vec{x} = d$ for all $\vec{x} \geq \vec{0}$, then $\vec{c} = \vec{0}$ and $d = 0$.

True

C) Let $p_1, \dots, p_m, q_1, \dots, q_n$ are given and $p_1, \dots, p_m \geq 0$ and $q_1, \dots, q_n \leq 0$.

True

If $\sum_{i=1}^m p_i = \sum_{j=1}^n q_j$, then $p_1 = \dots = p_m = q_1 = \dots = q_n = 0$.

D)
$$\sum_{j=1}^n c_j x_j + \sum_{i=1}^m \left[y_i \left(b_i - \sum_{j=1}^n a_{ij} x_j \right) \right]$$

$$= \sum_{i=1}^m b_i y_i + \sum_{j=1}^n \left[x_j \left(c_j - \sum_{i=1}^m a_{ij} y_i \right) \right]$$

True

WRONG

E) One of B, C, D is wrong.

~~$$\sum_{i=1}^m p_i = \sum_{j=1}^n q_j$$~~

If $p \geq 0, q \leq 0,$

and $p = q$

then $p = 0$ and $q = 0.$

$$\begin{aligned}
 D). \quad & \vec{c}^T \vec{x} + \vec{y}^T (\vec{b} - A\vec{x}) \\
 \text{True.} \quad & = \vec{c}^T \vec{x} + \vec{y}^T \vec{b} - \vec{y}^T A \vec{x} \\
 & = \vec{y}^T \vec{b} + \underbrace{\vec{c}^T \vec{x} - \vec{y}^T A \vec{x}} \\
 & = \vec{y}^T \vec{b} + (\vec{c}^T - \vec{y}^T A) \vec{x} \\
 & = \vec{b}^T \vec{y} + (\vec{c} - A^T \vec{y})^T \vec{x}
 \end{aligned}$$

$$\begin{aligned}
 & \vec{c}^T - \vec{y}^T A \\
 & = (\vec{c} - A^T \vec{y})^T
 \end{aligned}$$

$$\vec{c} - A^T \vec{y} = \begin{pmatrix} c_1 - \sum_{i=1}^m a_{i1} y_i \\ c_2 - \sum_{i=1}^m a_{i2} y_i \\ \vdots \end{pmatrix}$$

✓

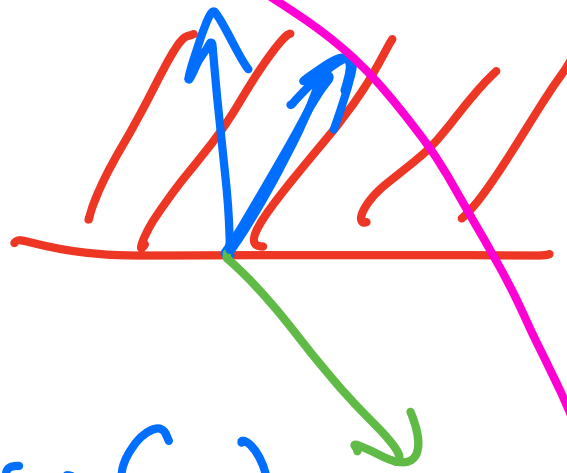
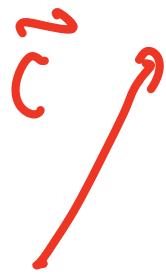
B)

$d=0$ because we can take

$$\vec{x} = \vec{0} \Rightarrow \vec{0}$$

$$\vec{c} \cdot \vec{0} = 0.$$

True



$\vec{x} \geq \vec{0}$

Suppose $\vec{c} = (c_1, \dots, c_n)$.

If $\vec{c} \neq \vec{0}$, then $c_i \neq 0$ for some i .

So we can choose $\vec{x} = (0, \dots, 0, \underbrace{|c_i|}_{i\text{-th}}, 0, \dots, 0)$

For this \vec{x} , $\vec{x} \geq \vec{0}$,

and $\vec{c} \cdot \vec{x} = c_i |c_i| \neq 0$.

But $\vec{c} \cdot \vec{x} = d = 0$. Therefore $\vec{c} = \vec{0}$.

Strong duality

Thm If an LP problem has an optimal solution,
so does its dual, and the respective optimal **objective values** are equal.

That is, if the primal problem

$$\begin{aligned} &\text{maximize } \vec{c} \cdot \vec{x} \\ &\text{subject to } A\vec{x} \leq \vec{b}, \vec{x} \geq \vec{0}. \end{aligned}$$

has an optimal solution $(x_1^*, x_2^*, \dots, x_n^*)$

then • the dual has an optimal solution (y_1^*, \dots, y_m^*)

and

$$\bullet \sum_{j=1}^n c_j x_j^* = \sum_{i=1}^m b_i y_i^*$$

where there is no entering variable.

Moreover, from the **z** row of the optimal **final** dictionary of the primal problem

$$z = z^* + \sum_{k=1}^{n+m} C_k^* x_k \quad \left(\begin{array}{l} \text{the optimal value} \\ \text{coefficient, } C_k^* \leq 0 \text{ for optimal} \\ \text{final dictionary} \end{array} \right) \quad \left(\begin{array}{l} \text{note} \\ C_k^* = 0 \text{ if } x_k \text{ basic} \end{array} \right)$$

optimal dictionary of the primal
can also determine
optimal solution for dual.

We can find

$$y_i^* = -C_{n+i}^*, \quad i=1, \dots, m, \quad \text{the coefficients for slack variable } x_{n+i}$$

for a dual optimal basic solution \vec{y}^* .

from these $(C_1^*, C_2^*, \dots, C_n^*, C_{n+1}^*, \dots, C_{n+m}^*)$
in the final dictionary,
a dual optimal basic solution \vec{y}^* is determined
as $y_i^* = -C_{n+i}^*, \quad i=1, \dots, m$
 $y_{n+j}^* = -C_j^*, \quad j=1, \dots, n$

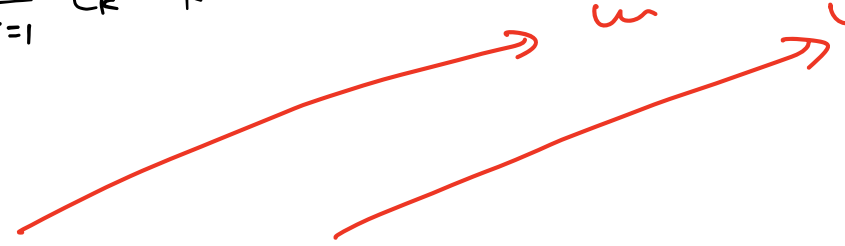
proof of strong duality

• Let $\vec{x} = (x_1, \dots, x_n)$ ← the decision variable.

$$\vec{x}_{\text{slack}} = (x_{n+1}, \dots, x_{n+m}) = \vec{b} - A\vec{x}$$

• Let $\vec{x}^* = (x_1^*, \dots, x_n^*)$ be an optimal solution.

• From the final ^{optimal} dictionary, get expression of $z = \vec{c} \cdot \vec{x}$

$$\text{as } z = z^* + \sum_{k=1}^{n+m} c_k^* x_k = z^* + \underbrace{\vec{c}^* \cdot \vec{x}} + \underbrace{\vec{c}_{\text{slack}}^* \cdot \vec{x}_{\text{slack}}}$$


• Note $\cdot \vec{c}^* \leq \vec{0}$, $\vec{c}_{\text{slack}}^* \leq \vec{0}$ as from the final optimal dictionary.

• $z^* = \vec{c} \cdot \vec{x}^*$ the optimal value.

So, $\vec{c} \cdot \vec{x} = \vec{c} \cdot \vec{x}^* + \vec{c}^* \cdot \vec{x} + \vec{c}_{\text{slack}}^* \cdot \vec{x}_{\text{slack}}$

$= \vec{c} \cdot \vec{x}^* + \vec{c}^* \cdot \vec{x} + \vec{c}_{\text{slack}}^* \cdot (\vec{b} - A\vec{x})$

$= \vec{c} \cdot \vec{x}^* + \vec{c}^* \cdot \vec{x} + \vec{c}_{\text{slack}}^* \cdot \vec{b} - \vec{c}_{\text{slack}}^* \cdot A\vec{x}$

$= \vec{c} \cdot \vec{x}^* + \vec{c}^* \cdot \vec{x} + \vec{c}_{\text{slack}}^* \cdot \vec{b} - (A^T \vec{c}_{\text{slack}}^*) \cdot \vec{x}$

Rearrange

$(\vec{c} - \vec{c}^* + A^T \vec{c}_{\text{slack}}^*) \cdot \vec{x} = \vec{c} \cdot \vec{x}^* + \vec{c}_{\text{slack}}^* \cdot \vec{b}$

This implies

$\vec{c} - \vec{c}^* + A^T \vec{c}_{\text{slack}}^* = \vec{0}$

$\vec{c} \cdot \vec{x}^* + \vec{c}_{\text{slack}}^* \cdot \vec{b} = 0$

Define $\vec{y}^* = -\vec{c}_{\text{slack}}^*$ (i.e. $y_i^* = -c_{n+i}^*$, $i=1, \dots, m$)

Note $\vec{y}^* \geq \vec{0}$ (as $\vec{c}_{\text{slack}}^* \leq \vec{0}$) Also, $\vec{c} - A^T \vec{y}^* = \vec{c}^* \leq \vec{0}$

So, \vec{y}^* is dual feasible.

Moreover, $\vec{c} \cdot \vec{x}^* = \vec{b} \cdot \vec{y}^*$

So, from weak duality, \vec{y}^* is (dual) optimal.

$(-c_{n+1}^*, -c_{n+2}^*, \dots, -c_{n+m}^*)$



They are equal as functions. (i.e. for all \vec{x}).

$\vec{x}_{\text{slack}} = \vec{b} - A\vec{x}$

$(\vec{c}_{\text{slack}}^*)^T A\vec{x} = (A^T \vec{c}_{\text{slack}}^*)^T \vec{x}$

$\vec{a} \cdot \vec{b} = \vec{a}^T \vec{b} = \vec{b}^T \vec{a}$

So $\vec{c} - A^T \vec{y}^* \leq \vec{0}$
i.e. $A^T \vec{y}^* \geq \vec{c}$

Complementary Slackness.

$\begin{cases} \vec{x}^* \text{ is optimal for primal.} \\ \vec{y}^* \text{ is optimal for dual.} \end{cases}$

$$\iff \left[\begin{array}{l} \vec{x}^* = (x_1^*, \dots, x_n^*) \text{ feasible for primal} \\ \vec{y}^* = (y_1^*, \dots, y_m^*) \text{ feasible for dual.} \\ \text{and } \textcircled{CS} \dots \begin{cases} y_i^* \cdot x_{n+i}^* = 0 & i=1, \dots, m \\ x_j^* \cdot y_{m+j}^* = 0 & j=1, \dots, n. \end{cases} \end{array} \right.$$

Here $x_{n+i}^* = i\text{-th slack of } \vec{x}^* = b_i - \sum_{j=1}^n a_{ij} x_j^*$
 $y_{m+j}^* = j\text{-th slack of } \vec{y}^* = \sum_{i=1}^m a_{ij} y_i^* - c_j$

Another (equivalent) form
of complementary slackness.

Complementary slackness

Thm Suppose • $\vec{x}^* = (x_1^*, \dots, x_n^*)$ be feasible for primal
 $\vec{y}^* = (y_1^*, \dots, y_m^*)$ be feasible for dual

Then,

• (\vec{x}^*, \vec{y}^*) optimal for (primal, dual)

\Leftrightarrow

(CS)

u.e.v.

$$y_i^* (b_i - \sum_{j=1}^n a_{ij} x_j^*) = 0 \text{ for each } i$$
$$\& x_j^* (c_j - \sum_{i=1}^m a_{ij} y_i^*) = 0 \text{ for each } j$$

Proof of complementary slackness

$$(*) \quad \bar{c} \cdot \bar{x} + \bar{y} \cdot (\bar{b} - A\bar{x}) = \bar{c} \cdot \bar{x} + \bar{y} \cdot \bar{b} - \bar{y} \cdot A\bar{x} = \bar{c} \cdot \bar{x} + \bar{y} \cdot \bar{b} - (A^T \bar{y}) \cdot \bar{x}$$

$$= \bar{b} \cdot \bar{y} + \bar{x} \cdot (\bar{c} - A^T \bar{y})$$

• optimality of \bar{x}^*, \bar{y}^* for primal/dual, respectively.

by strong and weak duality.

$$\Leftrightarrow \begin{cases} \bullet \text{ feasibility of } \bar{x}^*, \bar{y}^*, \text{ i.e. feasibility of } \bar{x}^*, \text{ and feasibility of } \bar{y}^* \\ \bullet \bar{c} \cdot \bar{x}^* = \bar{b} \cdot \bar{y}^* \end{cases} \quad \left(\begin{array}{l} \Leftarrow \text{ by weak duality} \\ \Rightarrow \text{ by strong duality} \end{array} \right)$$

$$\Leftrightarrow \begin{cases} \bullet \text{ feasibility of } \bar{x}^*, \bar{y}^* \\ \bullet \underbrace{\bar{y}^* \cdot (\bar{b} - A\bar{x}^*)}_{\geq 0} = \underbrace{\bar{x}^* \cdot (\bar{c} - A^T \bar{y}^*)}_{\leq 0} \quad (\text{by } (*)) \end{cases}$$

$\bar{x}_{\text{slack}}^* \geq 0, \quad \bar{y}_{\text{slack}}^* \leq 0$

$j=1, \dots, n$
 $i=1, \dots, m$

$\bar{y}_{\text{slack}}^* = A^T \bar{y}^* - \bar{c}$

$$\Leftrightarrow \begin{cases} \bullet \text{ feasibility of } \bar{x}^*, \bar{y}^* \\ \bullet \bar{y}^* \cdot (\bar{b} - A\bar{x}^*) = 0 \text{ \& } \bar{x}^* \cdot (\bar{c} - A^T \bar{y}^*) = 0 \end{cases} \Leftrightarrow \begin{cases} \bullet \text{ feasibility of } \bar{x}^*, \bar{y}^* \\ \bullet y_i^* \cdot x_{n+i}^* = 0, \quad i=1, \dots, m \\ x_j^* \cdot y_{m+j}^* = 0, \quad j=1, \dots, n \end{cases}$$

i.e. $\sum_{i=1}^m y_i^* \underbrace{x_{n+i}^*}_{\geq 0} = 0 \text{ \& } \sum_{j=1}^n x_j^* \underbrace{y_{m+j}^*}_{\geq 0} = 0$

We used:

For $p_i \geq 0, q_j \leq 0$

$$\sum_{i=1}^m p_i = \sum_{j=1}^n q_j \Rightarrow p_i = 0 = q_j \quad \forall i=1, \dots, m, j=1, \dots, n$$



