

23/11/2015.

LECTURE 18: CBI.11.

PROOF.

 $\Leftarrow$

If $y_i > 0$ then $(Ax)_i = b_i$ [i.e. CONSTRAINT IS TIGHT]. Then,

$$y^T Ax = \sum_{i=1}^m y_i (Ax)_i = \sum_{i: y_i > 0} y_i b_i = b^T y \quad \text{AS OTHER } y_i = 0$$

Similarly, if $x_j > 0$ then $(A^T y)_j = c_j$ [i.e. CONSTRAINT IS TIGHT]. Then,

$$x^T A^T y = \sum_{j=1}^n x_j (A^T y)_j = \sum_{j: x_j > 0} x_j c_j = c^T x \quad \text{AS OTHER } x_j = 0$$

Hence, for feasible x, y we have

$$b^T y = y^T Ax = x^T A^T y = c^T x.$$

Thus, by weak duality corollary 2, x is optimal for (P) and y optimal for (D).

 $\Rightarrow$

Suppose x optimal for (P), y optimal for (D) then $c^T x = b^T y$. Now, from (P), $Ax \leq b$ whilst from (D), $A^T y \geq c$. As $y \geq 0_m$, $x \geq 0_n$ we have:

$$y^T Ax \leq y^T b = c^T x \leq x^T A^T y = y^T Ax.$$

$$\text{i.e. } y^T Ax = y^T b \text{ AND } x^T A^T y = x^T c$$

$$\Rightarrow \sum_{i=1}^m y_i [(Ax)_i - b_i] = 0 \quad \text{AND} \quad \sum_{j=1}^n x_j [(A^T y)_j - c_j] = 0$$

as x, y feasible.

$$\text{i.e. } y_i [(Ax)_i - b_i] = 0 \quad \forall i = 1, \dots, m$$

$$x_j [(A^T y)_j - c_j] = 0 \quad \forall j = 1, \dots, n$$

□

(B)

EXAMPLE.

$$\begin{array}{ll}
 \text{maximum} & z = 4x_1 + 2x_2 + 3x_3 \\
 \text{subject to} & 2x_1 + 3x_2 + x_3 \leq 12 \quad (A_2)_1 \leq b_1 \\
 & x_1 + 4x_2 + 2x_3 \leq 10 \quad (A_2)_2 \leq b_2 \\
 & 3x_1 + x_2 + x_3 \leq 10 \quad (A_2)_3 \leq b_3 \\
 & x_1, x_2, x_3 \geq 0.
 \end{array}$$

The optimal solution is $x_1=2, x_2=0, x_3=4$. We can use the complementary slackness conditions to find the optimal solution of the dual problem.

$$\text{NB. } \begin{array}{c} \underline{A} \\ \left(\begin{array}{ccc|c} 2 & 3 & 1 & 12 \\ 1 & 4 & 2 & 10 \\ 3 & 1 & 1 & 10 \end{array} \right) \begin{array}{c} \underline{x} \\ \left(\begin{array}{c} 2 \\ 0 \\ 4 \end{array} \right) \end{array} = \begin{array}{c} \left(\begin{array}{c} 8 \\ 10 \\ 10 \end{array} \right) \end{array} \quad \begin{array}{l} \text{SLACK} \\ \text{TIGHT} \\ \text{TIGHT} \end{array}
 \end{array}$$

$$\text{So, } (A_2)_1 - b_1 \neq 0 \text{ but } (A_2)_2 = b_2 \quad \text{FIRST CONSTRAINT SLACK.} \quad (A_2)_3 = b_3$$

[IN CANONICAL FORM $s_1 \neq 0, s_2 = 0, s_3 = 0$ OPTIMAL BASIS $\{x_1, x_3, s_1\}$]

If x optimal then CS conditions imply $y_1 = 0$ in optimal solution to the dual problem

$$\begin{array}{ll}
 \text{minimum} & z' = 12y_1 + 10y_2 + 10y_3 \\
 \text{subject to} & 2y_1 + y_2 + 3y_3 \geq 4 \quad (A^T y)_1 \geq c_1 \\
 & 3y_1 + 4y_2 + y_3 \geq 2 \quad (A^T y)_2 \geq c_2 \\
 & y_1 + 2y_2 + y_3 \geq 3 \quad (A^T y)_3 \geq c_3 \\
 & y_1, y_2, y_3 \geq 0
 \end{array}$$

As $x_1 > 0$ and $x_3 > 0$ from the CS conditions,

$$\begin{array}{lcl}
 (A^T y)_1 - c_1 = 0 & 2y_1 + y_2 + 3y_3 = 4 & \} \quad y_2 + 3y_3 = 4 \\
 (A^T y)_3 - c_3 = 0 & y_1 + 2y_2 + y_3 = 3 & \} \quad 2y_2 + y_3 = 3 \\
 & y_1 = 0 & \text{i.e. } y_2 = y_3 = 1.
 \end{array}$$

We need to check that $y_1=0, y_2=1, y_3=1$ is feasible (if it wasn't then x would not be optimal). Checking the second constraint

$$3(0) + 4(1) + 1 = 5 \geq 2$$

As $y \geq \underline{0}_3$ and $\underline{A}^T y \geq \underline{c}$ this is feasible. Thus, it is optimal.

Note: $\underbrace{4(2) + 2(0) + 3(4)}_{\leq^T x} = 20 = \underbrace{12(0) + 10(1) + 10(1)}_{\geq^T y}$.

Symmetric CS conditions.

- If a variable in one problem is positive, corresponding constraint in the other must be tight.

[Example: $x_1, x_3 > 0$ so 1st, 3rd constraints of dual tight].

- If a constraint in one problem is not tight, that is it is **SLACK**, then corresponding variable in the other problem must be zero.

[Example: 1st constraint of primal slack so $y_1 = 0$].

As $x \geq \underline{0}_n, y \geq \underline{0}_m, \underline{A}x \leq \underline{b}, \underline{A}^T y \geq \underline{c}$ for feasible x, y then CS conditions are equivalent to:

Used these two constraints in the example.

$$\left\{ \begin{array}{l} y_i > 0 \Rightarrow (\underline{A}x)_i = b_i \\ x_j > 0 \Rightarrow (\underline{A}^T y)_j = c_j \end{array} \right\} \text{ or } \left\{ \begin{array}{l} (\underline{A}x)_i < b_i \Rightarrow y_i = 0 \\ (\underline{A}^T y)_j < c_j \Rightarrow x_j = 0 \end{array} \right\}$$

[Positivity $\Rightarrow$ Tight Constraint]

[Slackness $\Rightarrow$ variable zero].

Theorem (Asymptotic Complementary Slackness).

Consider the LP problem in canonical form:

(P) maximize $z = \underline{c}^T x$
subject to $\underline{A}x = \underline{b}$
 $x \geq \underline{0}_n$.

(D) maximize $z' = \underline{b}^T y$
subject to $\underline{A}^T y \geq \underline{c}$

(D).

Then $\underline{x} \in \mathbb{R}^n$, $\underline{y} \in \mathbb{R}^m$ optimal for (P) and (D) respectively if and only if they are feasible solutions and

$$x_j [(A^T \underline{y})_j - c_j] = 0 \quad \forall j = 1, \dots, n.$$

$$[(A \underline{x})_i - b_i] = 0 \quad \forall i = 1, \dots, m \text{ AS CANONICAL SO } A \underline{x} = \underline{b}.$$

which is equivalent to:

$$x_j > 0 \Rightarrow (A^T \underline{y})_j = c_j \quad \text{OR} \quad (A^T \underline{y})_j > c_j \Rightarrow x_j = 0.$$

HERE -

Proof.

$\Rightarrow$

Suppose $\underline{x}, \underline{y}$ optimal then $\underline{c}^T \underline{x} = \underline{b}^T \underline{y}$ [Shown results for symmetric dual; results for asymmetric dual identical: see question sheet eight]. Thus,

$$\underline{y}^T A \underline{x} = \underline{y}^T \underline{b} = \underline{c}^T \underline{x}$$

$$\text{i.e. } \underline{x}^T (A^T \underline{y} - \underline{c}) = 0 \Rightarrow \sum_{j=1}^n x_j ((A^T \underline{y})_j - c_j) = 0$$

As $\underline{x}, \underline{y}$ feasible $x_j \geq 0$, $(A^T \underline{y})_j - c_j \geq 0$ then $x_j ((A^T \underline{y})_j - c_j) = 0$ $\forall j = 1, \dots, n$.

$\Leftarrow$

Suppose $x_j [(A^T \underline{y})_j - c_j] = 0 \quad \forall j = 1, \dots, n$ then:

$$\underline{x}^T A^T \underline{y} = \underline{c}^T \underline{x}$$

However, as $\underline{x}$ feasible, $A \underline{x} = \underline{b} \Rightarrow \underline{b}^T = \underline{x}^T A^T$ so that $\underline{b}^T \underline{y} = \underline{c}^T \underline{x}$. $\square$

We'll make use of this theorem when studying the transportation algorithm.