

Linear Programming & Theory of Games.Theorems/Results on Simplex Method & DualityTheorem 1 - Improved Basic Feasible Solution.

Statement - Let \bar{x}_B be a basic feasible solution to the L.P.P. : Maximize $Z = \bar{c}\bar{x}$ subject to $A\bar{x} = \bar{b}$; $\bar{x} \geq \bar{0}$.

Let x_B be another basic feasible solution obtained by admitting a non-basis column vector \bar{a}_j in the basis, for which the net evaluation $z_j - c_j$ is negative. Then, \hat{x}_B is an improved basic feasible solution to the problem, that is,

$$\hat{c}_B \hat{x}_B > \bar{c}_B \bar{x}_B.$$

Proof - The L.P.P. is to determine \bar{x} , so as to Maximize $Z = \bar{c}\bar{x}$; $\bar{c}, \bar{x}^T \in \mathbb{R}^n$, subject to the constraints : $A\bar{x} = \bar{b}$ and $\bar{x} \geq \bar{0}$, $\bar{b} \in \mathbb{R}^m$, where A is an $m \times n$ real matrix.

We are given that, \bar{x}_B is a basic feasible solution
let $Z_0 = \bar{c}_B \bar{x}_B$.

Let \hat{a}_j be the column vector introduced in the basis, such that $z_j - c_j < 0$.

Let \bar{b}_2 be the vector removed from the basis & let \hat{x}_B be the new basic feasible solution.

Then,

$$\hat{x}_{B1} = x_{B1} - x_{B2} \frac{y_{1j}}{y_{rj}} \quad \& \quad \hat{x}_{B2} = \frac{x_{B2}}{y_{rj}}$$

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Thus, The new value of the objective function

$$\text{is, } \hat{z} = \sum_{i=1}^m \hat{c}_{Bi} \cdot \hat{x}_{Bi}$$
$$= \sum_{i=1}^m \bar{c}_{Bi} \left(\bar{x}_{Bi} - \bar{x}_{Br} \cdot \frac{y_{ij}}{y_{rj}} \right) + \hat{c}_{Br} \frac{x_{Br}}{y_{rj}}$$

$$= \sum_{i=1}^m \bar{c}_{Bi} \left(\bar{x}_{Bi} - \bar{x}_{Br} \frac{y_{ij}}{y_{rj}} \right) + c_j \cdot \frac{x_{Br}}{y_{rj}}$$

{Since $\hat{c}_{Br} = c_j$ }

$$= z_0 - (z_j - c_j) \cdot \frac{x_{Br}}{y_{rj}}$$

$$> z_0 \quad \{ \text{since } x_{Br}/y_{rj} > 0 \}$$

Hence, The new basic feasible solution \hat{x}_B gives an improved value of the objective function.

Exp. Theorem 2 - If a linear programming problem, $\text{Max. } Z = \bar{c}\bar{x}$ Page No.

Such that, $A\bar{x} = \bar{b}$, $\bar{x} \geq 0$, has an optimal feasible solution, then atleast one basic feasible solution must be optimal.

Proof - Let $z_0 = \bar{c}_B \bar{x}_B$ with $\bar{x}_B = B^{-1} \bar{b}$, be a basic feasible solution to the problem.

Now, to show that, The basic feasible solution is optimal, we are required to prove $z_0 \geq z^*$ where z^* is the optimal value, given by,

$$z^* = \sum_{j=1}^N c_j x_j.$$

The constraint equation $A\bar{x} = \bar{b}$ can be expressed as,

$$\sum_{j=1}^N a_j x_j = \bar{b} \quad \text{--- (i)}$$

Since, any vector $a_j \in A$ can be expressed as a linear combination of vectors in B , i.e.

$$a_j = \sum_{i=1}^m \gamma_{ij} \beta_j \quad \text{--- (ii)}$$

Substituting (ii) in (i); we get,

$$\sum_{j=1}^N \left(\sum_{i=1}^m \gamma_{ij} \beta_j \right) x_j = \bar{b}.$$

$$\text{or } \sum_{i=1}^m \left(\sum_{j=1}^N \gamma_{ij} x_j \right) \beta_j = \bar{b}$$

Since, $\bar{x}_B = B^{-1} \bar{b}$ is the basic feasible solution,

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$$\sum_{i=1}^m \alpha_{Bi} \cdot \beta_i = \bar{b}$$

Any vector can be uniquely expressed as a function of its basis vectors,
Therefore,

$$\bar{x}_B = \sum_{i=1}^m \alpha_{Bi} \cdot \beta_i$$

The optimality condition is $z_j - c_j \geq 0$ & $x_j \geq 0$.

Therefore,
$$\sum_{j=1}^N \alpha_j \cdot z_j \geq \sum_{j=1}^N c_j \cdot x_j = z^* \quad \text{--- (iii)}$$

But, we have,

$$\sum_{j=1}^N \alpha_j \cdot z_j \geq \sum_{j=1}^N \alpha_i \left(\sum_{l=1}^m c_{Bi} \cdot y_{ij} \right)$$

$$= \sum_{i=1}^m \left(\sum_{j=1}^N \alpha_i \cdot y_{ij} \right) c_{Bi}$$

$$= \sum_{i=1}^m \alpha_{Bi} \cdot c_{Bi} = z_0 \quad \text{--- (iv)}$$

So, from (iii) & (iv),

$$z_0 \geq z^*$$

Thus, The basic feasible solution is optimal.

Theorem 3 - If X is any feasible solution to the primal problem & W is any feasible solution to the dual problem, then $\bar{C}X \leq \bar{b}^T W$ i.e. $Z_x \leq Z_w$

Alternative Statement - Given problems are;

$$AX \leq \bar{b}, \quad X \geq \bar{0}; \quad \text{Max } Z_x = \bar{C}X \quad \text{--- (i)}$$

$$A^T W \geq \bar{C}^T, \quad W \geq \bar{0}; \quad \text{Min } Z_w = \bar{b}^T W \quad \text{--- (ii)}$$

Then, for any feasible solution X to (i) and any feasible solution W to (ii), $\bar{C}X \leq \bar{b}^T W$ i.e. $Z_x \leq Z_w$.

Proof - Let X & W be any feasible solutions to the primal and the dual problems respectively.

We have,

$$\begin{aligned} \bar{C}X &= \sum_{j=1}^n C_j \cdot x_j \leq \sum_{j=1}^n \left(\sum_{i=1}^m a_{ij} \cdot w_i \right) x_j \\ &= \sum_{i=1}^m w_i \cdot \left(\sum_{j=1}^n a_{ij} x_j \right) \leq \sum_{i=1}^m w_i \cdot b_i = \bar{b}^T W \end{aligned}$$

So, $\bar{C}X \leq \bar{b}^T W$ or $Z_x \leq Z_w$.

This completes the proof of the theorem.

Theorem 6.6 . *If the kth constraint of the primal is an equality, then the dual variable w_k is unrestricted in sign.* [Meerut 90]

Proof. As pointed out earlier, if the kth constraint is an equality, then we can write the primal in the form:

$$\left. \begin{aligned} c_1x_1 + c_2x_2 + \dots + c_nx_n &= z \text{ (max.)} \\ a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n &\leq b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n &\leq b_2 \\ &\vdots \\ a_{k1}x_1 + a_{k2}x_2 + \dots + a_{kn}x_n &\leq b_k \\ -a_{k1}x_1 - a_{k2}x_2 - \dots - a_{kn}x_n &\leq -b_k \\ &\vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n &\leq b_m \end{aligned} \right\} \dots(6.22)$$

so that its dual is given by

$$\left. \begin{aligned} b_1w_1 + b_2w_2 + \dots + b_k(w_k' - w_k'') + \dots + b_m w_m &= z_w \text{ (min.)} \\ a_{11}w_1 + a_{21}w_2 + \dots + a_{k1}(w_k' - w_k'') + \dots + a_{m1}w_m &\geq c_1 \\ a_{12}w_1 + a_{22}w_2 + \dots + a_{k2}(w_k' - w_k'') + \dots + a_{m2}w_m &\geq c_2 \\ &\vdots \\ a_{1n}w_1 + a_{2n}w_2 + \dots + a_{kn}(w_k' - w_k'') + \dots + a_{mn}w_m &\geq c_n \end{aligned} \right\} \dots(6.23)$$

If we replace $(w_k' - w_k'')$ by a new variable w_k , then we observe that the system (6.23) becomes equivalent to (6.2) except that w_k is unrestricted in sign.

6-5-3 Complementary Slackness Theorem

In this section, we shall make a further effort to relate the primal and dual formulations of linear programming problem. The following theorem (called the *complementary slackness theorem*) shows the intimate relation between primal and the dual.

Theorem 6-11. (Complementary Slackness Theorem) For the optimal feasible solutions of the primal and dual systems,

- (i) if the inequality occurs in the i th relation of either system (the corresponding slack or surplus variable x_{n+i} is +ive), then the i th variable w_i of its dual is zero, i.e. $w_i = 0$;
- (ii) if the j th variable x_j is +ive in either system, the j th dual constraint holds as a strict equality (i.e. the corresponding slack or surplus variable w_{m+j} is zero).

Proof. Since the primal objective function in explicit form can be written as

$$c_1x_1 + c_2x_2 + \dots + c_nx_n = z_x \text{ (max.)} \quad \dots(6-24)$$

and the dual objective function can be written as

$$b_1w_1 + b_2w_2 + \dots + b_mw_m = z_w \text{ (min.)} \quad \dots(6-25)$$

Also, the primal constraint-inequalities can be written as equations after introducing the non-negative slack variables $x_{n+1}, x_{n+2}, \dots, x_{n+m}$ as follows :

$$\left. \begin{array}{l} a_{11}x_1 + \dots + a_{1n}x_n + x_{n+1} = b_1 \\ \vdots \qquad \qquad \qquad \vdots \qquad \qquad \qquad \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n + x_{n+m} = b_m \end{array} \right\} \quad \dots(6-26)$$

Similarly, the dual-inequalities can be written as equations after introducing the non-negative surplus variables $w_{m+1}, w_{m+2}, \dots, w_{m+n}$ as follows :

$$\left. \begin{array}{l} a_{11}w_1 + \dots + a_{m1}w_m - w_{m+1} = c_1 \\ \vdots \qquad \qquad \qquad \vdots \qquad \qquad \qquad \vdots \\ a_{1n}w_1 + \dots + a_{mn}w_m - w_{m+n} = c_n \end{array} \right\} \quad \dots(6-27)$$

Now, multiplying the equations in (6-26) by w_1, w_2, \dots, w_m , respectively, and adding the resultant set, we get

$$x_1 \sum_{i=1}^m a_{i1}w_i + x_2 \sum_{i=1}^m a_{i2}w_i + \dots + x_n \sum_{i=1}^m a_{in}w_i + x_{n+1}w_1 + x_{n+2}w_2 + \dots + x_{n+m}w_m = \sum_{i=1}^m w_i b_i \quad \dots(6-28)$$

Next, we subtract (6-28) from (6-24) and obtain

$$(c_1 - \sum_{i=1}^m a_{i1}w_i)x_1 + (c_2 - \sum_{i=1}^m a_{i2}w_i)x_2 + \dots + (c_n - \sum_{i=1}^m a_{in}w_i)x_n - w_1x_{n+1} - w_2x_{n+2} - \dots - w_mx_{n+m}$$

$$= (z_x - \sum_{i=1}^m w_i b_i) \quad \dots(6-29)$$

But, from (6-25), we have

$$z_w = \sum_{i=1}^m w_i b_i \quad \dots(6-30)$$

and from (6-27), we have

$$-w_{m+j} = c_j - (a_{1j}w_1 + a_{2j}w_2 + \dots + a_{mj}w_m) \text{ or } -w_{m+j} = c_j - \sum_{i=1}^m a_{ij}w_i, j = 1, 2, \dots, n \quad \dots(6-31)$$

Now, using the results (6-30) and (6-31), eqn. (6-29) becomes

$$-(w_{m+1}x_1 + w_{m+2}x_2 + \dots + w_{m+n}x_n) - (w_1x_{n+1} + w_2x_{n+2} + \dots + w_mx_{n+m}) = z_x - z_w \quad \dots(6-32)$$

If $\mathbf{x}^* = (x_1^*, x_2^*, \dots, x_n^*)$ and $\mathbf{w}^* = (w_1^*, w_2^*, \dots, w_m^*)$ be the optimum solutions to the primal and dual problems respectively, then

$$z_x^* = z_w^* \quad \dots(6-33)$$

as claimed by the *Duality Theorem*.

Thus, for these optimum solutions and corresponding slack surplus variables

$$x_{n+i}^* \geq 0 \ (i = 1, 2, \dots, m) \text{ and } w_{m+j}^* \geq 0 \ (j = 1, 2, \dots, n),$$

the equation (6-32) becomes

$$-(w_{m+1}^*x_1^* + w_{m+2}^*x_2^* + \dots + w_{m+n}^*x_n^*) - (w_1^*x_{n+1}^* + \dots + w_m^*x_{n+m}^*) = 0 \text{ [from (6-33)]} \quad \dots(6-34)$$

Since all the variables in (6-34) are restricted to be non-negative, all the product terms of (6-34) are also non-negative, and for the validity of (6-34), each term must individually be equal to zero.

Thus, $w_{m+j}^*x_j^* = 0$, and $w_i^*x_{n+i}^* = 0$, for all i and j .

(a) Now, if $w_{m+j}^* > 0$, we must have $x_j^* = 0$; if $x_{n+i}^* > 0$, then $w_i^* = 0$, which proves the first part of the theorem.

(b) If $x_j^* > 0$, then $w_{m+j}^* = 0$, i.e., the j th constraint of dual problem is strict equality; if $w_i^* > 0$, then $x_{n+i}^* = 0$, i.e., the i th primal constraint is strict equality, which proves the second part of the theorem.

Hence the theorem is now completely established.

Alternative Statement