

$$\left. \begin{aligned} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n &\leq b_1, \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n &\leq b_2, \\ \dots &\dots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n &\leq b_m, \end{aligned} \right\} \quad (5)$$

and maximize a linear functional,

$$y = c_1x_1 + c_2x_2 + \dots + c_nx_n. \quad (6)$$

If the inequality signs of (5) are replaced with the equality signs by introducing slack variables, $x_{n+1}, x_{n+2}, \dots, x_{n+m}$, we may rewrite (5) as:

$$\left. \begin{aligned} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n + x_{n+1} &= b_1, \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n + x_{n+2} &= b_2, \\ \dots &\dots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n + x_{n+m} &= b_m. \end{aligned} \right\} \quad (7)$$

For the sake of abbreviation of the expressions, we put the column vectors as

$$P_1 = \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix}, P_2 = \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix}, \dots, P_n = \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix}, P_{n+1} = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \dots, P_{n+m} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix},$$

the row vector as $\mathbf{c} = (c_1, c_2, \dots, c_n, 0, \dots, 0)$ and the solution vector as $\mathbf{x} = (x_1, x_2, \dots, x_m, x_{n+1}, \dots, x_{n+m})$. So, we can rewrite our linear programming problem in the following compact form: "find $\mathbf{x} = (x_1, x_2, \dots, x_{n+m})$, $x_i \geq 0$, such that

$$L(\mathbf{x}) = \sum_{j=1}^{n+m} x_j P_j = P_0 \quad (8)$$

and

$$y = f(\mathbf{x}) = \mathbf{c} \cdot \mathbf{x} \quad (9)$$

is maximum."

A linear functional $y = f(\mathbf{x})$ defined on a convex polyhedron K takes its maximum value at an extreme point of K . Now, suppose that in any way we have found a basic solution, yielding an extreme point. Further we assume that, without loss of generality, the first m components of the extreme point \mathbf{x} are positive.

Then, (8) and (9) reduce to the following expressions;

$$\sum_{i=1}^m x_i P_i = P_0, \quad (x_1, x_2, \dots, x_m > 0). \quad (10)$$

and

$$\sum_{i=1}^m c_i x_i = f(\mathbf{x}) = y_0. \quad (11)$$

As the vectors P_1, P_2, \dots, P_m form a basis of the m -dimensional space R_m ,

every other vector is expressed in terms of this basis; or formally,

$$\sum_{i=1}^m x_{ij} \mathbf{P}_i = \mathbf{P}_j \quad (j = 1, 2, \dots, n + m). \quad (12)$$

Using (12), we may rewrite the solution (10), giving

$$\mathbf{P}_0 = \sum_{i=1}^m x_i \mathbf{P}_i - \theta \mathbf{P}_k + \theta \mathbf{P}_k \quad (13)$$

$$\begin{aligned} &= \sum_{i=1}^m x_i \mathbf{P}_i - \theta \sum_{i=1}^m x_{ik} \mathbf{P}_i + \theta \mathbf{P}_k \\ &= \sum_{i=1}^m (x_i - \theta x_{ik}) \mathbf{P}_i + \theta \mathbf{P}_k. \end{aligned} \quad (14)$$

Choosing the value of θ such that,

$$\theta = \min_i \left(\frac{x_i}{x_{ik}} \right) \quad (15)$$

for all i 's where $x_{ik} > 0$, the non-negative conditions,

- (a) $\theta \geq 0$,
- (b) $x_i - \theta x_{ik} \geq 0$

are obvious.

Then, these values are really the solution in the case where one of the basic vectors, \mathbf{P}_r , is replaced by new vector \mathbf{P}_k . It must be noticed, however, that for any other vector than \mathbf{P}_r , the relation $x_i - \theta x_{ik} = 0$ is also possible.

Referring (11) and (13), the corresponding value of y'_0 is

$$\begin{aligned} y'_0 &= \sum_{i=1}^m c_i (x_i - \theta x_{ik}) + \theta c_k \\ &= \sum_{i=1}^m c_i x_i + \theta (c_k - \sum_{i=1}^m c_i x_{ik}). \end{aligned} \quad (16)$$

Let,

$$y_j = \sum_{i=1}^m c_i x_{ij}, \quad (j = 1, 2, \dots, n + m), \quad (17)$$

then, (16) becomes

$$y'_0 = y_0 + \theta (c_k - y_k). \quad (18)$$

If $c_k - y_k > 0$ and $y_k > 0$, then $y'_0 > y_0$.

There exist three possibilities in (12);

(a) For some j , $y_j - c_j < 0$, and $x_{ij} \leq 0$ for every i . The case where the value of linear functional y_0 is unlimited.

(b) For some j , $y_j - c_j < 0$, and $x_{ij} > 0$ for some i . The case where a larger value of y_0 is obtained.

(c) $y_j - c_j \geq 0$, for all $j = 1, 2, \dots, n + m$. The case where the solution of (10) is maximum.

Substitution of (22) to P_j in (23) leads to

$$\begin{aligned}
 P_0 &= z_1 \left(\sum_{i \in I} x_{i1} P_i + \sum_{i \in I'} x_{i1} P_i \right) + z_2 \left(\sum_{i \in I} x_{i2} P_i + \sum_{i \in I'} x_{i2} P_i \right) + \dots \\
 &\quad \dots + z_n \left(\sum_{i \in I} x_{in} P_i + \sum_{i \in I'} x_{in} P_i \right) \\
 &= \sum_{i \in I} (z_1 x_{i1} + z_2 x_{i2} + \dots + z_n x_{in}) P_i + \sum_{i \in I'} (z_1 x_{i1} + z_2 x_{i2} + \dots + z_n x_{in}) P_i. \tag{24}
 \end{aligned}$$

Now using the fact that P_0 is uniquely expressible by P_i , then by comparing (21) and (24), the following identity is obtained ;

$$x_i = z_1 x_{i1} + z_2 x_{i2} + \dots + z_n x_{in}. \tag{25}$$

It can be supposed without loss of generality that the m basic vectors P_i ($i = 1, 2, \dots, m$) consist of q structural vectors, P_1, P_2, \dots, P_q and p artificial vectors, P_{n+1}, \dots, P_{n+p} . Restricting the attention to the second term on the right side of (24), and further considering the fact that vectors P_1, P_2, \dots, P_q are included in the basis, it is obvious that

$$x_{i1} = x_{i2} = \dots = x_{iq} = 0 \quad (i \in I'),$$

then

$$\left. \begin{aligned}
 x_{n+1} &= z_1 x_{n+1, q+1} + z_2 x_{n+1, q+2} + \dots + z_n x_{n+1, n} \geq 0, \\
 x_{n+2} &= z_1 x_{n+2, q+1} + z_2 x_{n+2, q+2} + \dots + z_n x_{n+2, n} \geq 0, \\
 &\dots \\
 x_{n+p} &= z_1 x_{n+p, q+1} + z_2 x_{n+p, q+2} + \dots + z_n x_{n+p, n} \geq 0,
 \end{aligned} \right\} \tag{26}$$

where

$$z_i \geq 0 \quad (i = 1, 2, \dots, n).$$

Thus, we have :

$$\begin{aligned}
 &x_{n+1} + x_{n+2} + \dots + x_{n+p} \\
 &= z_1 (x_{n+1, q+1} + x_{n+2, q+1} + \dots + x_{n+p, q+1}) \\
 &+ z_2 (x_{n+1, q+2} + x_{n+2, q+2} + \dots + x_{n+p, q+2}) \\
 &+ z_n (x_{n+1, n} + x_{n+2, n} + \dots + x_{n+p, n}) \\
 &z_i \geq 0 \quad (i = 1, 2, \dots, n).
 \end{aligned} \tag{27}$$

Then, it follows that for at least one suffix, say k , of column suffixes, $q+1, q+2, \dots, n$, non-negativeness,

$$x_{n+1, k} + x_{n+2, k} + \dots + x_{n+p, k} > 0 \tag{28}$$

holds.

Providing M be a sufficiently large positive value, the shadow cost for vector P_k is

$$y_k - c_k = c_1 x_{1k} + c_2 x_{2k} + \dots + c_q x_{qk} + (-M)(x_{n+1, k} + x_{n+2, k} + \dots + x_{n+p, k}) - c_k. \tag{29}$$

Then referring (28), the inequalities,

$$y_k - c_k < 0 \quad (30)$$

and, for at least one suffix, say r , which belongs to I ,

$$x_{rk} > 0 \quad (31)$$

are ascertained.

IV. Conclusion

When the artificial vectors are in a basis with positive coefficients at a certain stage of simplex table, we obtain (30) and (31) as a result of the study of mathematical property of the table. These findings are conceived as an assertion of methodological reasonability of introducing the parameter M into the linear programming problem, where the constraints are given by a set of equalities. As the results given above just belong to the case (b) of three possibilities (a), (b), (c) in II, then we can introduce the new vector P_k of the structural one to the basis in place of one of the basic vector, P_r , thus obtaining a new solution and a larger y_0 . Therefore, as far as some of the artificial vectors with positive coefficient appear in the basis, the simplex table is not optimum and further improvement is possible. Thus the final conclusion is: even if the linear programming problem is given under the constraints of a set of equalities, it is mathematically plausible to carry out the general simplex calculation by introducing the parameter M .

References

- 1) Charnes, W. W., Cooper, A. A. Henderson: *An Introduction to Linear Programming*, John Wiley and Sons, 1953.