

Simplex Method:

It is an iterative procedure and is one of the most powerful methods for solving LPP. This can be applied for solving LPP in 2 or more variables (without any restriction on number of constraints).

If we apply simplex method to LPP with 2 variables we will see that value of variables at each iteration are extreme points of feasible region (in case feasible solution to LPP exists). Hence, we move from one extreme point to another in such a way that the value of objective function is improved. Since, number of extreme points in feasible region are finite, optimal solution is ~~reached~~ achieved in finite number of steps.

Feasible solution: A solution to LPP is feasible if all variables in it assume non-negative values i.e., $x_i \geq 0 \quad 1 \leq i \leq n$.

Consider LPP :
$$\left. \begin{array}{l} \max/\min z = Cx \\ \text{subject to } AX (\leq \text{ or } \geq) b \\ X \geq 0 \end{array} \right\} \rightarrow \text{Canonical Form}$$

Standard form of LPP:
$$\begin{array}{l} \max/\min z = Cx \\ \text{subject to } AX = b \\ X \geq 0 \end{array}$$

$C_{1 \times n}$, $A_{m \times n}$, $C_{1 \times n}$, $b_{m \times 1}$ are matrices of given order (subscript).

To apply analytical methods for solving LPP we need to convert it first into standard form. For this we need slack/surplus variables. Let's see the definition of these two:

Slack variable: Consider the i^{th} constraint:

$$a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n \leq b_i \quad 1 \leq i \leq m$$

\because L.H.S. \leq R.H.S., so we add a variable s_i (≥ 0) such that

$$a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n + s_i = b_i$$

$s_i \rightarrow$ is called slack variable. It represents ~~used~~ unused capacity in LPP.

Surplus variable: Consider the i^{th} constraint

$$a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n \geq b_i \quad 1 \leq i \leq m$$

\because L.H.S. \geq R.H.S., so we subtract a variable s_i (≥ 0) such that

$$a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n - s_i = b_i$$

$s_i \rightarrow$ is called surplus variable. As the name suggests, it represents surplus capacity in LPP

Important: Cost/Profit associated with slack/surplus variables is zero (as unused or surplus capacity do not contribute towards profit).

Example: Convert the given LPP in standard form

$$\max z = 2x + 3y$$

$$\text{subject to } x + 5y \leq 6$$

$$2x - y \geq 3$$

$$x, y \geq 0$$

Soln Add slack variable to 1st constraint
 & a surplus variable to 2nd constraint

$$\max z = 2x + 3y$$

subject to

$$x + 5y + s_1 = 6$$

$$2x - y - s_2 = 3$$

$$x, y, s_1, s_2 \geq 0$$

standard form

Consider $Ax = b$ ($m < n$) where A is $m \times n$ matrix

Let $B_{m \times m}$ be submatrix of A with linearly independent columns of A . Then solution obtained by setting $(n-m)$ -variables not associated with columns of B , equal to zero, and solving the resultant system, is called Basic solution to given system
 i.e., $X_B = B^{-1}b$. $B \rightarrow$ Basis matrix

Refer to case where $\text{rank}(Ab) = \text{rank}(A) < \text{no. of unknowns}$
 in case of system of equations

Basic feasible solution (BFS) A solution which is basic as well as feasible is called BFS.

Note: If less than m -variables are positive then the solution is called degenerate BFS. i.e., if, say, $x_i > 0$ for $i=1, 2, \dots, k; k+1, \dots, m$ and $x_{k+1} = 0$ then $(x_1, x_2, \dots, x_m, \underbrace{0, 0, \dots, 0}_{n-m})$ form degenerate BFS.

Optimal Solution: Feasible solution which maximizes or minimizes, the objective function of LPP.

Unbounded Solution: A solution is called unbounded if the value of objective function increase or decrease indefinitely.

Ex Obtain all BFS of the following system of linear equations.

$$2x_1 + 6x_2 + 2x_3 = 3$$

$$6x_1 + 4x_2 + 4x_3 = 2$$

Solution Given system of equations $Ax = b$ is

$$\begin{pmatrix} 2 & 6 & 2 \\ 6 & 4 & 4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 3 \\ 2 \end{pmatrix}$$

A X b

Since $\rho(A) = 2$, the maximum number of l.e. columns of A is 2. Thus we can take any of the following

2x2 basis matrix B

$$\begin{pmatrix} 2 & 6 \\ 6 & 4 \end{pmatrix}, \begin{pmatrix} 2 & 2 \\ 6 & 4 \end{pmatrix}, \begin{pmatrix} 6 & 2 \\ 4 & 4 \end{pmatrix}$$

Consider $B_1 = \begin{pmatrix} 2 & 6 \\ 6 & 4 \end{pmatrix}$ Then $x_3 = 0$. $x_{B_1} = B_1^{-1}b = -\frac{1}{28} \begin{pmatrix} 4 & -6 \\ -6 & 2 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \end{pmatrix}$

$$= \begin{pmatrix} 0 \\ 1/2 \end{pmatrix}$$

\therefore one BFS is $(0, 1/2, 0)$

Now let $B = \begin{pmatrix} 2 & 2 \\ 6 & 4 \end{pmatrix}$ $\therefore x_2 = 0$ $x_{B_2} = B^{-1}b = -\frac{1}{4} \begin{pmatrix} 4 & -2 \\ -6 & 2 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \end{pmatrix}$

$$= \begin{pmatrix} -2 \\ 7/2 \end{pmatrix}$$

$\therefore (-2, 0, 7/2) \rightarrow$ Not a BFS (as $x_1 = -2 < 0$)

Now, let $B = \begin{pmatrix} 6 & 2 \\ 4 & 4 \end{pmatrix}$ $\therefore x_1 = 0$ $x_{B_3} = B^{-1}b = \frac{1}{16} \begin{pmatrix} 4 & -2 \\ -4 & 6 \end{pmatrix} \begin{pmatrix} 3 \\ 2 \end{pmatrix}$

$$= \begin{pmatrix} 1/2 \\ 0 \end{pmatrix}$$

\therefore BFS is $(0, 1/2, 0)$

Hence, this system has only one BFS $(0, 1/2, 0)$

Note This is a degenerate BFS as one of the basic variable is zero ($x_1 = 0$ in x_{B_1} & $x_3 = 0$ in x_{B_3})

Ex Obtain all BFS Basic solutions to the system

$$x_1 + 2x_2 + x_3 = 4$$

$$2x_1 + x_2 + 5x_3 = 5$$

Solution $B = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}$ $XB = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$ $x_3 = 0$

$B = \begin{pmatrix} 1 & 1 \\ 2 & 5 \end{pmatrix}$ $XB = \begin{pmatrix} 5 \\ -1 \end{pmatrix}$ $x_2 = 0$

$B = \begin{pmatrix} 2 & 1 \\ 1 & 5 \end{pmatrix}$ $XB = \begin{pmatrix} 5/3 \\ 2/3 \end{pmatrix}$ $x_1 = 0$

Note! All above three basic solutions are non-degenerate.

Theorem! (Reduction of FS to BFS) If an LPP has a feasible solution, then it also has a basic feasible solution.

Proof! Consider the LPP $\max z = CX$
subject to $AX = b, X \geq 0$

$C_{1 \times n}, X_{n \times 1}, b_{m \times 1}, A_{m \times n}$ real matrices.
Let $f(A) = m$. Since, feasible solution exists, so $f(A, b) = f(A), m < n$

Let $X = (x_1, x_2, \dots, x_n)$ be feasible soln. with $x_i \geq 0 \forall i$

Let X have p -positive components and remaining $(n-p)$ components are zero. Relabel the components so that positive components are first p and accordingly relabel columns of A . Then

$$a_1 x_1 + a_2 x_2 + a_3 x_3 + \dots + a_p x_p = b$$

where a_1, a_2, \dots, a_p are first p columns of A .

Now, 2 cases arise:

Case I: Vectors a_1, a_2, \dots, a_p form l.i. set. Then $p = m$

If $p = m$, the given solution is non-degenerate BFS with x_1, x_2, \dots, x_p as basic variables.

If $p < m$, then the set $\{a_1, a_2, \dots, a_p\}$ can be extended to $\{a_1, \dots, a_p, a_{p+1}, \dots, a_m\}$ to form basis for columns of A . Then we have

$$a_1 x_1 + a_2 x_2 + \dots + a_m x_m = b$$

where $x_j = 0, j = p+1, p+2, \dots, m$.

Thus, we have a degenerate BFS with $m-p$ of the basic variables zero.

Case II: Set $\{a_1, a_2, \dots, a_p\}$ is l.d. Obviously $p > m$
 $\Rightarrow \alpha_1 a_1 + \alpha_2 a_2 + \dots + \alpha_p a_p = 0$ α_i 's are constants (defn of l.d. vectors)

$$\Rightarrow \exists r, s.t. \alpha_r \neq 0. \text{ Then } a_r = -\sum_{j \neq r} \frac{\alpha_j}{\alpha_r} a_j$$

$$\therefore \sum_{j \neq r} \alpha_j x_j + \left(-\sum_{j \neq r} \frac{\alpha_j}{\alpha_r} a_j\right) x_r = b.$$

$$\text{or, } \sum_{j \neq r} \left(x_j - x_r \frac{\alpha_j}{\alpha_r}\right) a_j = b$$

This is a solution in which not more than $(p-1)$ non-zero components are there.

To ~~ensure~~ ^{ensure} that these are positive, choose a_r , in such a way that $x_j - x_r \frac{\alpha_j}{\alpha_r} \geq 0 \quad \forall j \neq r$.

$$\Rightarrow \text{either } \alpha_j = 0 \text{ or } \frac{x_j}{\alpha_j} > \frac{x_r}{\alpha_r}, \text{ if } \alpha_j > 0.$$

$$\Rightarrow \frac{x_j}{\alpha_j} < \frac{x_r}{\alpha_r} \text{ if } \alpha_j < 0.$$

\therefore select a_r such that $\frac{x_r}{\alpha_r} = \min_j \left\{ \frac{x_j}{\alpha_j}, \alpha_j > 0 \right\}$

Then all $(p-1)$ variables are $\left(x_j - x_r \frac{\alpha_j}{\alpha_r}\right)$ are non-negative so, we have a F.S. with not more than $(p-1)$ non-zero components.

If the corresponding set of $(p-1)$ columns of A is l.i., case II applies and we have BFS. If not, we again have l.d. set. So repeat the process till we arrive at F.S. with associated set of columns of A l.i. Hence, from case I, we have BFS.

Corollary: There exists only finite number of BFS. \square
 a LPP