

### Simplex method in matrix form (revised simplex method)

- A LP with  $n$  decision variables and  $m$  constraints can be written as

$$\begin{aligned} \max \quad & Z = \mathbf{c}\mathbf{x} \\ \text{s.t.} \quad & \mathbf{A}\mathbf{x} = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned}$$

$$\text{where } \mathbf{c} = (c_1, c_2, c_3, \dots, c_n), \quad \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, \quad A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}, \quad \mathbf{b} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix}$$

- Alternatively, the LP can be written as

$$\begin{aligned} \max \quad & Z = \mathbf{c}_B \mathbf{x}_B + \mathbf{c}_N \mathbf{x}_N \\ \text{s.t.} \quad & \mathbf{B}\mathbf{x}_B + \mathbf{N}\mathbf{x}_N = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned} \tag{1}$$

where the subscripts “B” and “N” denote basic and nonbasic variables respectively.

- For example,

$$\left\{ \begin{array}{ll} \max & Z = 2x_1 + 3x_2 \\ \text{s.t.} & x_1 + x_2 \leq 50 \\ & 2x_1 + x_2 \leq 30 \\ & x_1, x_2 \geq 0 \end{array} \right. \Leftrightarrow \left\{ \begin{array}{ll} \max & Z = 2x_1 + 3x_2 \\ \text{s.t.} & x_1 + x_2 + S_1 = 50 \\ & 2x_1 + x_2 + S_2 = 30 \\ & x_1, x_2, S_1, S_2 \geq 0 \end{array} \right.$$

Then, at  $O(0,0)$ ,

$$\mathbf{x}_B = \begin{pmatrix} S_1 \\ S_2 \end{pmatrix}, \quad \mathbf{x}_N = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \quad \mathbf{c}_N = (2, 3), \quad \mathbf{c}_B = (0, 0), \quad \mathbf{b} = \begin{pmatrix} 50 \\ 30 \end{pmatrix}$$

$$\mathbf{B} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \quad N = \begin{pmatrix} 1 & 1 \\ 2 & 1 \end{pmatrix}$$

➤ Solving for  $\mathbf{x}_B$  in (1) gives

$$\mathbf{B}\mathbf{x}_B = \mathbf{b} - \mathbf{N}\mathbf{x}_N \Rightarrow \mathbf{x}_B = \mathbf{B}^{-1}(\mathbf{b} - \mathbf{N}\mathbf{x}_N) = \mathbf{B}^{-1}\mathbf{b} - \mathbf{B}^{-1}\mathbf{N}\mathbf{x}_N$$

➤ The LP can then be rewritten as

$$\begin{aligned} \max \quad & Z = \mathbf{c}_B \mathbf{B}^{-1} \mathbf{b} - (\mathbf{c}_B \mathbf{B}^{-1} \mathbf{N} - \mathbf{c}_N) \mathbf{x}_N \\ \text{s.t.} \quad & \mathbf{x}_B + \mathbf{B}^{-1} \mathbf{N} \mathbf{x}_N = \mathbf{B}^{-1} \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{aligned}$$

➤ Recall that each iteration of the simplex method allows a nonbasic variable (the entering variable) to increase from zero.

until one of the basic variables (the leaving variable) hits zero.

➤ Let  $N_j$  be the  $j^{\text{th}}$  column of  $N$ , and  $V_i$  be the  $i^{\text{th}}$  component of vector  $V$ . Then, an iteration of the simplex method, with  $x_j$  being the basic variable and  $x_i$  being basic variables, can be represented by the following equations:

$$\begin{aligned} Z &= \mathbf{c}_B \mathbf{B}^{-1} \mathbf{b} - (\mathbf{c}_B \mathbf{B}^{-1} \mathbf{N}_j - \mathbf{c}_j) x_j & \Leftrightarrow Z^{\text{new}} &= Z^0 - \mathbf{c}_j^0 x_j \\ x_i &= (\mathbf{B}^{-1} \mathbf{b})_i - (\mathbf{B}^{-1} \mathbf{N}_j)_i x_j & \Leftrightarrow x_i^{\text{new}} &= \mathbf{b}_i^0 - \mathbf{A}_{ij}^0 x_j \end{aligned}$$

where the superscripts “0” and “new” denote the current solution and the new solution generated by the next iteration.

- For a max problem, we choose the entering variable,  $x_k$ , s.t.
- $$z_k - c_k = \min\{z_j - c_j \mid j \text{ is nonbasic and } z_j - c_j < 0\}, \quad (2)$$

where,  $z_j \equiv \mathbf{c}_B \mathbf{B}^{-1} \mathbf{N}_j$ .<sup>1</sup>

- The leaving variable is  $x_r$  s.t.

$$\frac{(\mathbf{B}^{-1} \mathbf{b})_r}{(\mathbf{B}^{-1} \mathbf{N}_k)_r} = \min \left\{ \frac{(\mathbf{B}^{-1} \mathbf{b})_i}{(\mathbf{B}^{-1} \mathbf{N}_k)_i} \mid i \text{ is basic and } (\mathbf{B}^{-1} \mathbf{N}_k)_i > 0 \right\}. \quad (3)$$

- The above provides the rational for the revised simplex method which proceeds as follows

**Step 0.** Determine a starting basic feasible solution with basis  $\Omega$ .

**Step 1.** Evaluate  $\mathbf{B}^{-1}$ .

**Step 2.** Compute  $(z_j - c_j)$  for all nonbasic variables. If  $(z_j - c_j) \geq 0$  for a maximization problem ( $\leq 0$  for a minimization), then stop.

The optimal solution is  $\mathbf{x}_B^* = \mathbf{B}^{-1} \mathbf{b}$ ,  $Z^* = \mathbf{c}_B \mathbf{x}_B^*$ . Else, determine the entering variable,  $x_k$ , using (2), and go to Step 3.

**Step 3.** Compute  $\mathbf{B}^{-1} \mathbf{N}_k$ . If all elements of  $\mathbf{B}^{-1} \mathbf{N}_k \leq 0$ , then stop, the solution is unbounded. Else, compute  $\mathbf{B}^{-1} \mathbf{b}$  and determine the leaving variable,  $x_r$ , using (3).

**Step 4.** determine the new basis,  $\Omega_{\text{new}} = \Omega \cup \{x_k\} - \{x_r\}$ . Set  $\Omega = \Omega_{\text{new}}$ , and go to Step 1.

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<sup>1</sup> This rule for choosing the entering variable is just a rule of thumb. In some cases, there might be better ways to choose the entering variable.

- Example

$$\left\{ \begin{array}{ll} \max & Z = 2x_1 + 3x_2 \\ \text{s.t.} & x_1 + x_2 \leq 50 \\ & 2x_1 + x_2 \leq 30 \\ & x_1, x_2 \geq 0 \end{array} \right. \Leftrightarrow \left\{ \begin{array}{ll} \max & Z = 2x_1 + 3x_2 \\ \text{s.t.} & x_1 + x_2 + S_1 = 50 \\ & 2x_1 + x_2 + S_2 = 30 \\ & x_1, x_2, S_1, S_2 \geq 0 \end{array} \right.$$

➤ **Iteration 1.**

- **Step 0.** Starting basic feasible solution at  $O(0,0)$ ,

$$\Omega = \{S_1, S_2\}, \mathbf{x}_B = \begin{pmatrix} S_1 \\ S_2 \end{pmatrix}, \mathbf{x}_N = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \mathbf{c}_N = (2, 3), \mathbf{c}_B = (0, 0),$$

$$\mathbf{B} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \mathbf{N} = \begin{pmatrix} 1 & 1 \\ 2 & 1 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} 50 \\ 30 \end{pmatrix}$$

- **Step 1.**

$$\mathbf{B}^{-1} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

- **Step 2.**

$$z_{x_1} - c_{x_1} = \mathbf{c}_B \mathbf{B}^{-1} \mathbf{N}_{x_1} - c_{x_1} = (0, 0) \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} - 2 = -2$$

$$z_{x_2} - c_{x_2} = \mathbf{c}_B \mathbf{B}^{-1} \mathbf{N}_{x_2} - c_{x_2} = (0, 0) \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} - 3 = -3$$

The entering variable is  $x_2$ .

- **Step 3.**

$$\mathbf{B}^{-1} \mathbf{N}_{x_2} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \begin{matrix} (S_1) \\ (S_2) \end{matrix}$$

$$\mathbf{B}^{-1} \mathbf{b} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 50 \\ 30 \end{pmatrix} = \begin{pmatrix} 50 \\ 30 \end{pmatrix} \begin{matrix} (S_1) \\ (S_2) \end{matrix}$$

$$\text{Then, } \frac{(\mathbf{B}^{-1} \mathbf{b})_{S_1}}{(\mathbf{B}^{-1} \mathbf{N}_{x_2})_{S_1}} = \frac{50}{1} = 50, \quad \frac{(\mathbf{B}^{-1} \mathbf{b})_{S_2}}{(\mathbf{B}^{-1} \mathbf{N}_{x_2})_{S_2}} = \frac{30}{1} = 30.$$

The leaving variable is  $S_2$ .

- **Step 4.** The new basis is  $\Omega = \{S_1, S_2\} \cup \{x_2\} - \{S_2\} = \{S_1, x_2\}$ .

➤ **Iteration 2.**

○ **Step 1.**

$$\Omega = \{S_1, x_2\}, \mathbf{x}_B = \begin{pmatrix} S_1 \\ x_2 \end{pmatrix}, \mathbf{x}_N = \begin{pmatrix} x_1 \\ S_2 \end{pmatrix}, \mathbf{c}_B = (0, 3), \mathbf{c}_N = (2, 0),$$

$$\mathbf{B} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}, \mathbf{N} = \begin{pmatrix} 1 & 0 \\ 2 & 1 \end{pmatrix}$$

$$\mathbf{B} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \Rightarrow \mathbf{B}^{-1} = \frac{1}{1} \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix},^2$$

○ **Step 2.**

$$z_{x_1} - c_{x_1} = \mathbf{c}_B \mathbf{B}^{-1} \mathbf{N}_{x_1} - c_{x_1} = (0, 3) \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \end{pmatrix} - 2 = (0, 3) \begin{pmatrix} -1 \\ 2 \end{pmatrix} - 2 = 4$$

$$z_{S_2} - c_{S_2} = \mathbf{c}_B \mathbf{B}^{-1} \mathbf{N}_{S_2} - c_{S_2} = (0, 3) \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} - 0 = (0, 3) \begin{pmatrix} -1 \\ 1 \end{pmatrix} - 0 = 3$$

All  $z_j - c_j > 0$ , for all  $j$  nb. Stop. The optimal solution is reached. The optimal solution is

$$\mathbf{x}_B^* = \begin{pmatrix} S_1^* \\ x_2^* \end{pmatrix} = \mathbf{B}^{-1} \mathbf{b} = \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 50 \\ 30 \end{pmatrix} = \begin{pmatrix} 20 \\ 30 \end{pmatrix}$$

$$Z^* = \mathbf{c}_B \mathbf{x}_B^* = (0 \ 3) \begin{pmatrix} 20 \\ 30 \end{pmatrix} = 90$$

Therefore, the optimal solution is  $x_1^* = 0$ ,  $x_2^* = 30$ , and  $Z^* = 90$ .

**Remark.** The solution to the same problem in tabular form is presented on the next page. It is instructive to compare the two solution methods.

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<sup>2</sup> The inverse of  $\mathbf{B} = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$  is  $\mathbf{B}^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$ . (This inverse exists only if  $ad - bc \neq 0$ .)

### Iteration 1.

Entering variable

$z_j - c_j$

Leaving (blocking) variable

Basic	$x_1$	$x_2$	$S_1$	$S_2$	RHS	Ratio
Z	-2	-3	0	0	0	-
$S_1$	1	1	1	0	50	50/1=50
$S_2$	2	1	0	1	30	30/1=30

$B^{-1}$

$\frac{(B^{-1}b)_i}{(B^{-1}N_k)_i}$

### Iteration 2.

$z_j - c_j$

$c_B B^{-1}b$

$B^{-1}b$

$B^{-1}$

Basic	$x_1$	$x_2$	$S_1$	$S_2$	RHS
Z	4	0	0	3	90
$S_1$	-1	0	1	-1	20
$x_2$	2	1	0	1	30