

# Matrix Notation and Simplex (6.2)

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Consider the LP in standard form,

$$\begin{aligned} \max z &= \mathbf{c}^T \mathbf{x} \\ \text{st } A\mathbf{x} &= \mathbf{b} \\ \mathbf{x} &\geq 0 \end{aligned}$$

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Let a subscript  $B$  denote the set of basic variables, and a subscript  $N$  denote the set of remaining variables.

$$\mathbf{x} = \begin{bmatrix} \mathbf{x}_B \\ \mathbf{x}_N \end{bmatrix} \quad \mathbf{c} = \begin{bmatrix} \mathbf{c}_B \\ \mathbf{c}_N \end{bmatrix} \quad A = [B \ N] \quad \mathbf{b} = \mathbf{b}$$

Re-write the LP as:

$$\begin{aligned} \max z - & \mathbf{c}_B^T \mathbf{x}_B - \mathbf{c}_N^T \mathbf{x}_N = 0 \\ \text{st} & B\mathbf{x}_B + N\mathbf{x}_N = \mathbf{b} \\ & \mathbf{x} \geq 0 \end{aligned}$$

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NOTE ABOUT TEXT: They take  $\mathbf{c}$  as a *row* instead of the more usual column, so they do NOT use transposes.

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The coefficients that from which we compute the max (or min) are in the vector  $\mathbf{c}_B$ . With  $A$  and Row 0 in the correct order:

$$\begin{array}{c|c} ?? & \mathbf{c}_B^T B^{-1}\mathbf{b} \\ \hline B^{-1}A & B^{-1}\mathbf{b} \end{array} \Rightarrow \begin{array}{c|c} -\mathbf{c}^T + \mathbf{c}_B^T B^{-1}A & \mathbf{c}_B^T B^{-1}\mathbf{b} \\ \hline B^{-1}A & B^{-1}\mathbf{b} \end{array}$$

We need to be careful that the variable indices line up for the remaining row zero coefficients (We'll see an example below).

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$$B\vec{e}_k = k^{\text{th}} \text{ col of } B \doteq B_k$$

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- If  $x_j$  is an artificial variable, then  $c_j = M$  and

$$\hat{c}_j = \mathbf{c}_B^T (B^{-1})_k - M$$

## Numerical Example 1:

“Dakota Furniture” (max):

$x_1$	$x_2$	$x_3$	$s_1$	$s_2$	$s_3$	rhs		$x_1$	$x_2$	$x_3$	$s_1$	$s_2$	$s_3$	RHS
-60	-30	-20	0	0	0	0		0	5	0	0	10	10	280
8	6	1	1	0	0	48	$\Rightarrow$	0	-2	0	1	2	-8	24
4	2	$\frac{3}{2}$	0	1	0	20		0	-2	1	0	2	-4	8
2	$\frac{3}{2}$	$\frac{1}{2}$	0	0	1	8		1	$\frac{5}{4}$	0	0	$-\frac{1}{2}$	$\frac{3}{2}$	2

If the set of basic variables (the basis) is  $s_1, x_3, x_1$ , then verify the previous computations.

SOLUTION: The matrix  $B$  is given by the columns of  $A$  corresponding to  $s_1, x_3, x_1$ .

$$B = \begin{bmatrix} 1 & 1 & 8 \\ 0 & 3/2 & 4 \\ 0 & 1/2 & 2 \end{bmatrix} \Rightarrow B^{-1} = \begin{bmatrix} 1 & 2 & -8 \\ 0 & 2 & -4 \\ 0 & -1/2 & 3/2 \end{bmatrix}$$

Further, (note the order):

$$\mathbf{c}_B^T B^{-1} \mathbf{b} = 0 \cdot 24 + 20 \cdot 8 + 60 \cdot 2 = 280$$

The matrix  $B$  is made up of Columns 4, 3 and 1 of matrix  $A$ . Let  $N$  be made from columns 2, 5, 6 of matrix  $A$  (in that order). Then  $A = [B, N]$  and

$$[0, 20, 60] \cdot B^{-1} [B, N] - [0, 20, 60, 30, 0, 0] = [0, 0, 0, 5, 10, 10]$$

Order was important in the last example. That is, if we use the original order, then

$$\mathbf{c}_B^T B^{-1} A - \mathbf{c}^T$$

would be meaningless-

$$[0, 20, 60] B^{-1} A - [60, 30, 20, 0, 0, 0] = [-58, -67.5, 0, 0, 39, -77]$$

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The coefficient is:

$$\mathbf{c}_B^T B^{-1} \mathbf{a}_2 - 30 = [0, 20, 60] \begin{bmatrix} -2 \\ -2 \\ 1.25 \end{bmatrix} - 30 = -40 + 75 - 30 = 5$$

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The coefficient is:

$$\mathbf{c}_B^T B^{-1} \vec{e}_2 - 0 = [0, 20, 60] \begin{bmatrix} 2 \\ 2 \\ -1/2 \end{bmatrix} - 30 = 40 - 30 - 0 = 10$$

## Numerical Example 2:

This one is small enough to do by hand. Given that  $x_1, x_2$  are the basic variables for the optimal tableau, find that tableau directly.

$$\begin{aligned}\max z &= 3x_1 + x_2 \\ \text{st} \quad 2x_1 - x_2 &\leq 2 \\ -x_1 + x_2 &\leq 4 \\ x_1, x_2 &\geq 0\end{aligned}$$

SOLUTION: Write the LP in standard form

$$\begin{array}{cccc|c}
 x_1 & x_2 & s_1 & s_2 & \text{rhs} \\
 -3 & -1 & 0 & 0 & 0 \\
 \hline
 2 & -1 & 1 & 0 & 2 \\
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So that the matrix  $B$  and  $B^{-1}$  are:

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So that the matrix  $B$  and  $B^{-1}$  are:

$$B = \begin{bmatrix} 2 & -1 \\ -1 & 1 \end{bmatrix} \Rightarrow B^{-1} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}$$

- The optimal tableau is:

$$\begin{array}{c|c} & \\ \hline B^{-1}A & B^{-1}\mathbf{b} \end{array} = \begin{array}{cccc|c} 1 & 0 & 1 & 1 & 6 \\ 0 & 1 & 1 & 2 & 10 \end{array}$$

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$$\begin{aligned} -[3, 1, 0, 0] + [3, 1] \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 2 & -1 & 1 & 0 \\ -1 & 1 & 0 & 1 \end{bmatrix} &= \\ -[3, 1, 0, 0] + [3, 1, 4, 5] &= [0, 0, 4, 5] \end{aligned}$$

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