

## Simplex - first example

Suppose we are given the problem

$$\text{Minimize } z = -x_1 + 2x_2 - x_3$$

subject to

$$\begin{cases} x_1 - 2x_2 + x_3 + x_4 = 10 \\ 2x_1 - 3x_2 - x_3 + x_5 = 6 \\ x_1, x_2, x_3, x_4, x_5 \geq 0. \end{cases}$$

- ▶ Rewrite the objective function as  $0 = -z - x_1 + 2x_2 - x_3$ .
- ▶ For the constraints, swap the LHS and RHS.
- ▶ We have the following:

$$\begin{cases} 0 = -z - x_1 + 2x_2 - x_3 \\ 10 = x_1 - 2x_2 + x_3 + x_4 \\ 6 = 2x_1 - 3x_2 - x_3 + x_5 \\ x_1, x_2, x_3, x_4, x_5 \geq 0. \end{cases}$$

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Set  $x_0 = -z$ . We have the following initial tableau:

	$x_0$	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$
$x_0 = -z$	0	1	-1	2	-1	0
$x_4$	10	0	1	-2	1	0
$x_5$	6	0	2	-3	-1	0

The tableau is solved for the basis  $B = (4, 5)$  and the corresponding basic feasible solution is  $\mathbf{x} = (0, 0, 0, 10, 6)^T$ . (Note: in the future we will not include the  $x_0$  column in the tableau. We refer to the entries in the first column as,

$$\begin{bmatrix} a_{0,0} \\ \vdots \\ a_{m,0} \end{bmatrix}, \text{ the } x_1 \text{ column as } \begin{bmatrix} a_{0,1} \\ \vdots \\ a_{m,1} \end{bmatrix}, \text{ etc.}).$$

	$x_0$	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$
$x_0 = -z$	0	1	-1	2	-1	0
$x_4$	10	0	1	-2	1	0
$x_5$	6	0	2	-3	-1	0

1. Check for negative entries in the top row other than  $a_{0,0}$ .
2. Choose ANY such negative entry, for example  $a_{0,1} = -1$ . The corresponding column is called the *pivot column*.
3. Both entries in this column are positive, so we must compare ratios to determine the pivot column.
4. The ratio  $\frac{a_{1,0}}{a_{1,1}} = \frac{10}{1}$  is greater than the ratio  $\frac{a_{2,0}}{a_{2,1}} = \frac{6}{2}$ , so the *pivot row* is row 2. The entry  $a_{2,1} = 2$  is called the *pivot entry*.
5. Pivoting on the pivot entry will solve the tableau for the basis  $(4, 1)$ . We replace 5 with 1 in the basis, because the  $x_1$  column is the pivot column and the pivot row is solved for  $x_5$ .

	$x_0$	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	
$x_0 = -z$	3	1	0	$1/2$	$-3/2$	0	$1/2$
$x_4$	7	0	0	$-1/2$	<b><math>3/2</math></b>	1	$-1/2$
$x_1$	3	0	1	$-3/2$	$-1/2$	0	$1/2$

1. This tableau is solved for the basis  $B = (4, 1)$  and corresponding bfs is  $\mathbf{x} = (3, 0, 0, 7, 0)^T$ .
2. The tableau implies that  $z = -3 + \frac{1}{2} \cdot x_2 - \frac{3}{2} \cdot x_3 + \frac{1}{2} \cdot x_5$ , so  $z = -3$  for the current bfs, and this solution is better than the previous one.
3. The only negative entry in the 0th row is the coefficient at  $x_3$ ,  $a_{0,3} = -3/2$ , so the pivot column is the  $x_3$  column.
4. In this column, there is only one positive entry,  $a_{1,3} = 3/2$ , so we pivot on this entry to solve for the basis  $(3, 1)$ .

	$x_0$	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$
$x_0 = -z$	10	1	0	0	1	0
$x_3$	$14/3$	0	0	$-1/3$	$2/3$	$-1/3$
$x_1$	$16/3$	0	1	$-5/3$	$1/3$	$1/3$

1. This tableau corresponds to the basis  $(3, 1)$  and the correspond bfs is  $\mathbf{x} = (16/3, 0, 14/3, 0, 0)^T$ .
2. We have that  $z = -10 + x_4$  which implies that  $z \geq -10$ , because in any feasible solution  $x_4 \geq 0$ . Because the value of the object function at the current bfs is  $-10$ , the current bfs is an optimum and the algorithm terminates.
3. In general, *the simplex algorithm terminates when all of the entries in the top row, except possibly  $a_{0,0}$ , are non-negative.* (Note: If the LP is unbounded there is a different condition which causes the algorithm to terminate).

1. Recall our initial tableau  $\mathbf{T}$  - now written without the headings

$$\mathbf{T} = \left( \begin{array}{c|ccc|ccc} 0 & 1 & -1 & 2 & -1 & 0 & 0 \\ 10 & 0 & 1 & -2 & 1 & 1 & 0 \\ 6 & 0 & 2 & -3 & -1 & 0 & 1 \end{array} \right)$$

2. In the first pivot step, we multiplied row 2 by  $1/2$  and then added row 2 to row 0 and then subtracted row 2 from row 1.
3. These three steps can be thought for a left multiplication or pre-multiplication by the following three invertible matrices.

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \frac{1}{2} \end{pmatrix} = \begin{pmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & -\frac{1}{2} \\ 0 & 0 & \frac{1}{2} \end{pmatrix}$$

4. Similarly, the second pivot step is equivalent to pre-multiplication by the following three invertible matrices.

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & \frac{1}{2} & 1 \end{pmatrix} \begin{pmatrix} 1 & \frac{3}{2} & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & \frac{2}{3} & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & \frac{2}{3} & 0 \\ 0 & \frac{1}{3} & 1 \end{pmatrix}$$

1. So the final tableau  $\mathbf{T}'$  is the initial tableau  $\mathbf{T}$  pre-multiplied by the following invertible matrix  $\mathbf{X}$

$$\mathbf{X} := \begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & 0 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & -\frac{1}{2} \\ 0 & 0 & \frac{1}{2} \end{pmatrix} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & -\frac{1}{2} \\ 0 & 1 & \frac{1}{2} \end{pmatrix}$$

2. That is,

$$\begin{aligned} \mathbf{X}\mathbf{T} &= \begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & -\frac{1}{2} \\ 0 & 1 & \frac{1}{2} \end{pmatrix} \left( \begin{array}{c|ccc|ccc} 0 & 1 & -1 & 2 & -1 & 0 & 0 \\ \hline 10 & 0 & 1 & -2 & 1 & 1 & 0 \\ \hline 6 & 0 & 2 & -3 & -1 & 0 & 1 \end{array} \right) \\ &= \begin{pmatrix} 10 & 1 & 0 & 0 & 0 & 1 & 0 \\ \hline \frac{14}{3} & 0 & 0 & -\frac{1}{3} & 1 & \frac{2}{3} & -\frac{1}{3} \\ \hline \frac{16}{3} & 0 & 1 & -\frac{1}{3} & 0 & \frac{1}{3} & \frac{1}{3} \end{pmatrix} = \mathbf{T}' \end{aligned}$$

3. Question: Why is  $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \mathbf{e}_1$  the first column of  $\mathbf{X}$ ?

Because the  $x_0$  column of every tableau must be  $\mathbf{e}_1$ .

1. From now on, let's remove the  $x_0$  column from  $\mathbf{T}$  since it never changes, so

$$\mathbf{T} = \left( \begin{array}{c|ccc|ccc} 0 & -1 & 2 & -1 & 0 & 0 \\ \hline 10 & 1 & -2 & 1 & 1 & 0 \\ \hline 6 & 2 & -3 & -1 & 0 & 1 \end{array} \right) = \left( \begin{array}{c|c} f & \mathbf{c}^T \\ \hline \mathbf{b} & \mathbf{A} \end{array} \right)$$

2. Using block multiplication to compute  $\mathbf{T}' = \mathbf{X}\mathbf{T}$  we get

$$\begin{aligned} \mathbf{X}\mathbf{T} &= \begin{pmatrix} 1 & 1 & 0 \\ 0 & 2 & -\frac{1}{2} \\ 0 & 1 & \frac{1}{2} \end{pmatrix} \left( \begin{array}{c|c} f & \mathbf{c}^T \\ \hline \mathbf{b} & \mathbf{A} \end{array} \right) \\ &= \left( \begin{array}{c|c} (1)f + (1 \ 0)\mathbf{b} & (1)\mathbf{c}^T + (1 \ 0)\mathbf{A} \\ \hline \begin{pmatrix} 0 \\ 0 \end{pmatrix} f + \begin{pmatrix} 2 \\ 1 \end{pmatrix} \mathbf{b} & \begin{pmatrix} 0 \\ 0 \end{pmatrix} \mathbf{c}^T + \begin{pmatrix} 2 & -\frac{1}{2} \\ 1 & \frac{1}{2} \end{pmatrix} \mathbf{A} \end{array} \right) \\ &= \left( \begin{array}{c|c} f + (1 \ 0)\mathbf{b} & \mathbf{c}^T + (1 \ 0)\mathbf{A} \\ \hline \begin{pmatrix} 2 \\ 1 \end{pmatrix} \mathbf{b} & \begin{pmatrix} 2 & -\frac{1}{2} \\ 1 & \frac{1}{2} \end{pmatrix} \mathbf{A} \end{array} \right) \end{aligned}$$

1. In general, we can write  $\mathbf{X}$  as  $\begin{pmatrix} 1 & \mathbf{d}^T \\ 0 & \mathbf{U} \\ \vdots & \\ 0 & \end{pmatrix} = \left( \begin{array}{c|c} 1 & \mathbf{d}^T \\ \hline \mathbf{0} & \mathbf{U} \end{array} \right)$ .

2. So,

$$\begin{aligned} \mathbf{X}\mathbf{T} &= \begin{pmatrix} 1 & \mathbf{d}^T \\ 0 & \mathbf{U} \end{pmatrix} \left( \begin{array}{c|c} f & \mathbf{c}^T \\ \hline \mathbf{b} & \mathbf{A} \end{array} \right) = \left( \begin{array}{c|c} f + \mathbf{d}^T \mathbf{b} & \mathbf{c}^T + \mathbf{d}^T \mathbf{A} \\ \hline \mathbf{U}\mathbf{b} & \mathbf{U}\mathbf{A} \end{array} \right) \\ &= \left( \begin{array}{c|c|c|c|c} f + \mathbf{d}^T \mathbf{b} & c_1 + \mathbf{d}^T \mathbf{A}_1 & c_2 + \mathbf{d}^T \mathbf{A}_2 & \dots & c_n + \mathbf{d}^T \mathbf{A}_n \\ \hline \mathbf{U}\mathbf{b} & \mathbf{U}\mathbf{A}_1 & \mathbf{U}\mathbf{A}_2 & \dots & \mathbf{U}\mathbf{A}_n \end{array} \right) \end{aligned}$$

3. Suppose  $\mathbf{T}' = \mathbf{X}\mathbf{T}$  is solved for the basis  $B = (j_1, \dots, j_m)$ .

4. Question: What is  $\mathbf{U}\mathbf{A}_B = (\mathbf{U}\mathbf{A}_{j_1} \mid \mathbf{U}\mathbf{A}_{j_2} \mid \dots \mid \mathbf{U}\mathbf{A}_{j_m})$ ?

$\mathbf{U}\mathbf{A}_B = \mathbf{I}_m$ , because the last tableau is solved for  $B$ .

5. Question: What does that imply about  $\mathbf{U}$ ?  $\mathbf{U} = \mathbf{A}_B^{-1}$

6. Question: What is

$$\mathbf{c}_B^T + \mathbf{d}^T \mathbf{A}_B = (c_{j_1} + \mathbf{d}^T \mathbf{A}_{j_1} \mid \dots \mid c_{j_m} + \mathbf{d}^T \mathbf{A}_{j_m})^T$$

$$\mathbf{c}_B^T + \mathbf{d}^T \mathbf{A}_B = (0 \ 0 \ \dots \ 0) = \mathbf{0}^T.$$

7. Question: What does that imply about  $\mathbf{d}^T$ ?  $\mathbf{d}^T = -\mathbf{c}_B^T \mathbf{A}_B^{-1}$

### You should know this!

1. To solve a tableau  $\mathbf{T} = \left( \begin{array}{c|c} f & \mathbf{c}^T \\ \mathbf{b} & \mathbf{A} \end{array} \right)$  for the basis  $B$ ,  
pre-multiply by

$$\mathbf{X} = \left( \begin{array}{c|c} 1 & -\mathbf{c}_B^T \mathbf{A}_B^{-1} \\ \mathbf{0} & \mathbf{A}_B^{-1} \end{array} \right) = \left( \begin{array}{c|c} 1 & \mathbf{c}_B^T \\ \mathbf{0} & \mathbf{A}_B \end{array} \right)^{-1}.$$

2. The tableau solved for the basis  $B$  can be written as

$$\mathbf{X}\mathbf{T} = \left( \begin{array}{c|c} f - \mathbf{c}_B^T \mathbf{A}_B^{-1} \mathbf{b} & \mathbf{c}^T - \mathbf{c}_B^T \mathbf{A}_B^{-1} \mathbf{A} \\ \mathbf{A}_B^{-1} \mathbf{b} & \mathbf{A}_B^{-1} \mathbf{A} \end{array} \right).$$

3. We often define  $\pi^T := \mathbf{c}_B^T \mathbf{A}_B^{-1}$ , so  $\mathbf{X} = \left( \begin{array}{c|c} 1 & -\pi^T \\ \mathbf{0} & \mathbf{A}_B^{-1} \end{array} \right)$  and

$$\mathbf{X}\mathbf{T} = \left( \begin{array}{c|c} f - \pi^T \mathbf{b} & \mathbf{c}^T - \pi^T \mathbf{A} \\ \mathbf{A}_B^{-1} \mathbf{b} & \mathbf{A}_B^{-1} \mathbf{A} \end{array} \right).$$

4. Furthermore, we let  $\bar{\mathbf{c}}^T = \mathbf{c}^T - \pi^T \mathbf{A}$ . We call  $\bar{\mathbf{c}}^T$  the *relative cost vector* and  $\bar{c}_j$  the *relative cost* of  $x_j$  or the *relative cost* of Column  $j$ .