

## Sensitivity Analysis

### 1 Changes in the RHS (b)

Consider an optimal LP solution. Suppose that the original RHS ( $\mathbf{b}$ ) is changed from  $\mathbf{b}^0$  to  $\mathbf{b}^{\text{new}}$ . In the following, we study the affect of this change the optimal solution. First, note that changing  $\mathbf{b}$  will not affect optimality. Recall that optimality is based on the condition  $\mathbf{c}_B \mathbf{B}^{-1} \mathbf{A}_j - c_j = z_j - c_j \geq 0$ , for all variables  $x_j$  (for max problems). Changing  $\mathbf{b}$  will not affect  $z_j - c_j$ . Hence, changing  $\mathbf{b}$  will not affect optimality. However; changing  $\mathbf{b}$  may affect feasibility.

To check, whether the LP is still feasible after changing  $\mathbf{b}$  from  $\mathbf{b}^0$  to  $\mathbf{b}^{\text{new}}$  compute

$$\mathbf{x}^{\text{new}} = \mathbf{B}^{-1} \mathbf{b}^{\text{new}}.$$

- If  $\mathbf{x}^{\text{new}} \geq 0$  (i.e.,  $x_i^{\text{new}} \geq 0$ , for all  $i$ ), then the LP is still feasible and the optimal solution is  $\mathbf{x}^{\text{new}}$ .
- If  $x_i^{\text{new}} < 0$ , for some  $i$ , then the current basic solution is no longer feasible. Then, the dual simplex method is used to restore feasibility (leading to a new optimal basis).

Note that when the LP remains feasible the rate of change of the optimal objective function,  $Z^*$ , as a function of a RHS value  $b_i$  is  $y_i^*$  the corresponding dual variable.

This follows by noting that  $Z^* = \sum b_i y_i^*$  and  $\frac{\partial Z^*}{\partial b_i} = y_i^*$ . Therefore, changing the  $b_i$  by 1 unit will change  $Z^*$  by  $y_i^*$  (assuming that this change will not affect feasibility )

#### Example 1.

consider the LP

$$\begin{aligned} \max Z &= 3x_1 + 2x_2 + 5x_3 \\ \text{s.t.} \quad &x_1 + 2x_2 + x_3 \leq 430 \\ &3x_1 + x_2 + 2x_3 \leq 460 \\ &x_1 + 4x_2 \leq 420 \\ &x_1, x_2, x_3 \geq 0 \end{aligned}$$

Let  $S_1, S_2, S_3$  be the slack variables corresponding to the three constraints. Upon solving the LP, the optimal simplex tableau is as follows.

	$x_1$	$x_2$	$x_3$	$S_1$	$S_2$	$S_3$	RHS
$Z$	4	0	0	1	2	0	1350
$x_2$	-1/4	1	0	1/2	-1/4	0	100
$x_3$	3/2	0	1	0	1/2	0	230
$S_3$	2	0	0	-2	1	1	20

The optimal solution is  $x_1^* = 0$ ,  $x_2^* = 100$ ,  $x_3^* = 230$ , and  $Z^* = 1350$ .

- Suppose that  $\mathbf{b}$  is changed from  $\begin{pmatrix} 430 \\ 460 \\ 420 \end{pmatrix}$  to  $\begin{pmatrix} 500 \\ 460 \\ 600 \end{pmatrix}$ . What is the new optimal solution?

$$\mathbf{x}^{new} = \mathbf{B}^{-1}\mathbf{b}^{new} = \begin{pmatrix} 1/2 & -1/4 & 0 \\ 0 & 1/2 & 0 \\ -2 & 1 & 1 \end{pmatrix} \begin{pmatrix} 500 \\ 460 \\ 600 \end{pmatrix} = \begin{pmatrix} 135 \\ 230 \\ 60 \end{pmatrix} > 0.$$

Since  $\mathbf{x}^{new}$  is feasible. Then, the new optimal solution is  $\mathbf{x}^{*new} = \mathbf{x}^{new}$ . That is, the new optimal solution is  $x_1^{*new} = 0$ ,  $x_2^{*new} = 135$ ,  $x_3^{*new} = 230$ , and the new optimal

objective value is  $Z^{*new} = \mathbf{c}_B \mathbf{x}^{*new} = (2 \ 5 \ 0) \begin{pmatrix} 135 \\ 230 \\ 60 \end{pmatrix} = 1420$ . Note that  $Z^{*new}$  can

also be obtained using duality. From the simplex tableau, the optimal dual solution is  $y_1^* = 1$ ,  $y_2^* = 2$ ,  $y_3^* = 0$ . Then,

$$\begin{aligned} Z^{*new} &= Z^* + (b_1^{new} - b_1) y_1^* + (b_2^{new} - b_2) y_2^* + (b_3^{new} - b_3) y_3^* \\ &= 1350 + (500 - 430) \times 1 + (460 - 460) \times 2 + (600 - 420) \times 0 = 1420. \end{aligned}$$

- What if  $\mathbf{b}$  is changed from  $\begin{pmatrix} 430 \\ 460 \\ 420 \end{pmatrix}$  to  $\begin{pmatrix} 450 \\ 460 \\ 400 \end{pmatrix}$ ?

$$\mathbf{x}^{new} = \mathbf{B}^{-1}\mathbf{b} = \begin{pmatrix} 1/2 & -1/4 & 0 \\ 0 & 1/2 & 0 \\ -2 & 1 & 1 \end{pmatrix} \begin{pmatrix} 450 \\ 460 \\ 400 \end{pmatrix} = \begin{pmatrix} 110 \\ 230 \\ -40 \end{pmatrix}.$$

In this case,  $\mathbf{x}^{new}$  is not feasible. The new optimal solution is obtained by applying the dual simplex method. First, change the RHS of the optimal tableau to  $\mathbf{x}^{new}$  and the

objective value to  $Z(\mathbf{x}^{new}) = \mathbf{c}_B \mathbf{x}^{new} = (2 \ 5 \ 0) \begin{pmatrix} 110 \\ 230 \\ -40 \end{pmatrix} = 1370$ .

	$x_1$	$x_2$	$x_3$	$S_1$	$S_2$	$S_3$	RHS
Z	4	0	0	1	2	0	1370
$x_2$	-1/4	1	0	1/2	-1/4	0	110
$x_3$	3/2	0	1	0	1/2	0	230
$S_3$	2	0	0	-2	1	1	-40
Ratio	--	--	--	1/2	--	--	--

	$x_1$	$x_2$	$x_3$	$S_1$	$S_2$	$S_3$	RHS
Z	5	0	0	0	5/2	1/2	1350
$x_2$	1/4	1	0	0	0	1/4	100
$x_3$	3/2	0	1	0	1/2	0	230
$S_1$	-1	0	0	1	-1/2	-1/2	20

The last tableau is optimal (RHS > 0). The new optimal solution  $\mathbf{x}^{*new}$  is  $x_1^{*new} = 0$ ,  $x_2^{*new} = 100$ ,  $x_3^{*new} = 230$ , and  $Z^{*new} = 1350$ .

#### Range of $b_i$ where LP remains feasible

In Example 1, suppose only  $b_1$  is changed to  $b_1 + \delta$ . In the following, we find the range of  $\delta$  that keeps the LP feasible (and keeps the current basis ( $x_2, x_3, S_3$ ) optimal).

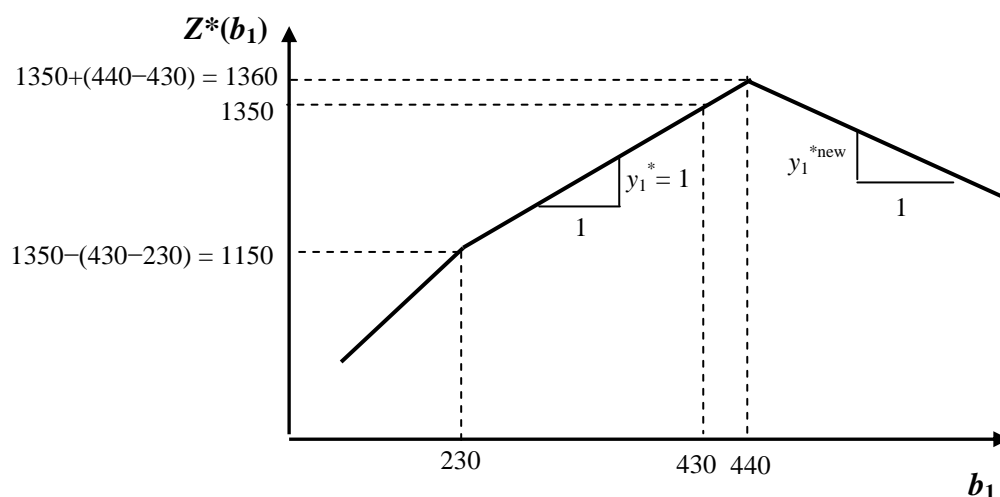
With this change

$$\mathbf{x}^{new} = \mathbf{B}^{-1} \mathbf{b}^{new} = \begin{pmatrix} 1/2 & -1/4 & 0 \\ 0 & 1/2 & 0 \\ -2 & 1 & 1 \end{pmatrix} \begin{pmatrix} 430 + \delta \\ 460 \\ 420 \end{pmatrix} = \begin{pmatrix} 100 + \delta/2 \\ 230 \\ 20 - 2\delta \end{pmatrix}.$$

$$\text{LP remains feasible if } \mathbf{x}^{new} > 0 \Rightarrow \begin{cases} 100 + \frac{\delta}{2} \geq 0 \Rightarrow \delta \geq -200 \\ 20 - 2\delta \geq 0 \Rightarrow \delta \leq 10 \end{cases}.$$

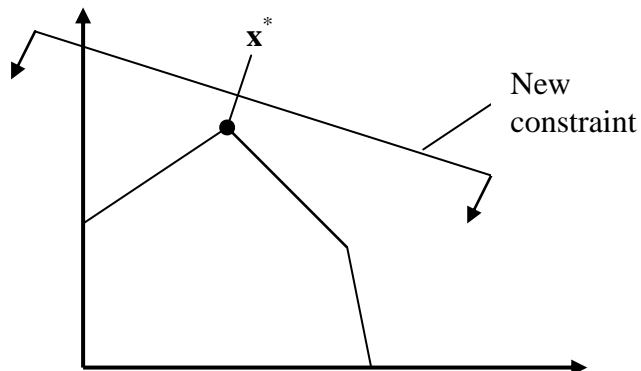
Then, LP remains feasible if  $-200 \leq \delta \leq 10$ , or equivalently  $230 \leq b_1 \leq 440$ .

Graphically,

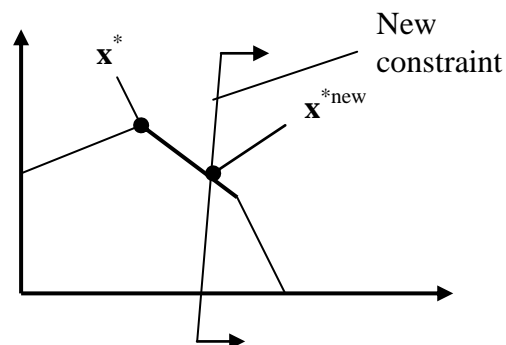


## 2 Adding a new Constraint

When a new constraint is added, the optimal solution does not improve. If the current optimal solution satisfies the new constraint, the optimal solution does not change. Graphically,



Otherwise, if the optimal solution does not satisfy the constraint, the current optimal solution becomes infeasible, and a new optimal solution could be obtained using the dual simplex method. Graphically,



### Example 2.

- Suppose the constraint  $3x_1 + x_2 + x_3 \leq 500$  is added to the LP in Example 1.

What is the new optimal solution?

The current optimal solution  $x_1^* = 0$ ,  $x_2^* = 100$ , and  $x_3^* = 230$ , satisfies the constraint (since  $3 \times 0 + 100 + 230 = 330 < 500$ ). The optimal solution does not change.

- Suppose the constraint  $3x_1 + 3x_2 + x_3 \leq 500$  is added to the LP in Example 1.

What is the new optimal solution?

The current optimal solution does not satisfy the constraint (since  $3 \times 0 + 3 \times 100 + 230 = 530 > 500$ ). Then, add the constraint to the simplex tableau, rearrange the tableau (to be in the right form) and proceed with the dual simplex method. Let  $S_4$  be the slack variable corresponding to the new constraint.

	$x_1$	$x_2$	$x_3$	$S_1$	$S_2$	$S_3$	$S_4$	RHS
Z	4	0	0	1	2	0	0	1350
$x_2$	-1/4	1	0	1/2	-1/4	0	0	100
$x_3$	3/2	0	1	0	1/2	0	0	230
$S_3$	2	0	0	-2	1	1	0	20
$S_4$	3	3	1	0	0	0	1	500

Rearrange the tableau so that the basic variables  $x_2$  and  $x_3$  have zero coefficients in the  $S_4$  row (note that  $S_4$  is now considered a basic variable). Then apply dual simplex.

↓

	$x_1$	$x_2$	$x_3$	$S_1$	$S_2$	$S_3$	$S_4$	RHS
Z	4	0	0	1	2	0	0	1350
$x_2$	-1/4	1	0	1/2	-1/4	0	0	100
$x_3$	3/2	0	1	0	1/2	0	0	230
$S_3$	2	0	0	-2	1	1	0	20
$S_4$	9/4	0	0	-3/2	1/4	0	1	-30
Ratio	--	--	--	2/3	--	--	--	

$l_4' = l_4 - l_2 - 3l_1$

←

	$x_1$	$x_2$	$x_3$	$S_1$	$S_2$	$S_3$	$S_4$	RHS
Z	11/2	0	0	0	13/6	0	2/3	1330
$x_2$	1/2	1	0	0	-1/6	0	1/3	90
$x_3$	3/2	0	1	0	1/2	0	0	230
$S_3$	-1	0	0	0	2/3	1	-4/3	60
$S_1$	-3/2	0	0	1	-1/6	0	-2/3	20

The last tableau is optimal. The new optimal solution is  $x_1^{*new} = 0$ ,  $x_2^{*new} = 90$ ,

$x_3^{*new} = 230$ , and  $Z^{*new} = 1330$ .

### 3 Changes in the objective function coefficients (c)

Consider an optimal LP tableau. Recall that the RHS is given by  $\mathbf{B}^{-1}\mathbf{b}$ . Therefore, changing  $\mathbf{c}$  will not affect feasibility. It will only affect the optimality conditions given by  $\mathbf{c}_B\mathbf{B}^{-1}\mathbf{A}_j - c_j = z_j - c_j \geq 0$ , (for max problems). When  $\mathbf{c}$  is changed from  $\mathbf{c}^0$  to  $\mathbf{c}^{\text{new}}$ , two situations could result:

- i) The optimality conditions continue to hold (if  $z_j^{\text{new}} - c_j \geq 0$ , for all  $j$ , for max problems). Then, the optimal solution does not change (but the optimal objective value  $Z^{\text{new}} = \mathbf{c}_B^{\text{new}} \mathbf{x}^*$  could change).
- ii) The optimality conditions do not hold anymore ( $z_j^{\text{new}} - c_j < 0$ , for some  $j$ , for max problems). Then, the optimal solution changes. The (usual “primal”) simplex method is used to obtain a new optimal solution.

#### Example 3.

- In the LP of Example 1, suppose the objective function is changed from  $Z = 3x_1 + 2x_2 + 5x_3$  to  $Z = 2x_1 + 3x_2 + 4x_3$ . That is,  $\mathbf{c}$  is changed from  $\mathbf{c}^0 = (3 \ 2 \ 5)$  to  $\mathbf{c}^{\text{new}} = (2 \ 3 \ 4)$ . What is the new optimal solution?

In terms of the current optimal solution basis ( $x_2, x_3, S_3$ ),  $\mathbf{c}_B$  is changed from  $\mathbf{c}_B^0 = (2 \ 5 \ 0)$  to  $\mathbf{c}_B^{\text{new}} = (3 \ 4 \ 0)$ . Then  $z_j^{\text{new}} - c_j^{\text{new}}$  for nonbasic variables are

$$\begin{aligned} \mathbf{c}_B^{\text{new}}\mathbf{B}^{-1}\mathbf{N} - \mathbf{c}_N^{\text{new}} &= (3 \ 4 \ 0) \begin{pmatrix} 1/2 & -1/4 & 0 \\ 0 & 1/2 & 0 \\ -2 & 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 & S_1 & S_2 \\ 1 & 1 & 0 \\ 3 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} - (2 \ 0 \ 0) \\ &= (3/2 \ 5/4 \ 0) \begin{pmatrix} 1 & 1 & 0 \\ 3 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} - (2 \ 0 \ 0) \\ &= (21/4 \ 3/2 \ 5/4) - (2 \ 0 \ 0) \\ &= \begin{pmatrix} 13/4 & 3/2 & 5/4 \end{pmatrix} \geq \mathbf{0} . \end{aligned}$$

Therefore, the optimal solution remains the same,  $x_1^* = 0$ ,  $x_2^* = 100$ ,  $x_3^* = 230$ , and

$$Z^{*new} = \mathbf{c}_B^{new} \mathbf{x}^* = (3 \ 4 \ 0) \begin{pmatrix} 100 \\ 230 \\ 20 \end{pmatrix} = 1220.$$

- Suppose now that  $\mathbf{c}$  is changed from  $\mathbf{c}^0 = (3 \ 2 \ 5)$  to  $\mathbf{c}^{new} = (6 \ 3 \ 4)$ . What is the new optimal solution?

In this case,  $\mathbf{c}_B$  is changed from  $\mathbf{c}_B^0 = (2 \ 5 \ 0)$  to  $\mathbf{c}_B^{new} = (3 \ 4 \ 0)$ , and  $z_j^{new} - c_j^{new}$  ( $j$  nonbasic) is given by

$$\mathbf{c}_B^{new} \mathbf{B}^{-1} \mathbf{N} - \mathbf{c}_N^{new} = (21/4 \ 3/2 \ 5/4) - (6 \ 0 \ 0) = \begin{pmatrix} \overbrace{-3/4 \quad 3/2 \quad 5/4}^{x_1 \quad S_1 \quad S_2} \end{pmatrix}. \text{ The}$$

current solution is no longer optimal. Modify the Z-row of the current simplex tableau is as follow.

$Z^{new} = \mathbf{c}_B^{new} \mathbf{x}^0 = 1220$

	$x_1$	$x_2$	$x_3$	$S_1$	$S_2$	$S_3$	RHS	Ratio
Z	-3/4	0	0	3/2	5/4	0	1220	--
$x_2$	-1/4	1	0	1/2	-1/4	0	100	--
$x_3$	3/2	0	1	0	1/2	0	230	460/3
$S_3$	2	0	0	-2	1	1	20	10

Then, proceed with the simplex method.

	$x_1$	$x_2$	$x_3$	$S_1$	$S_2$	$S_3$	RHS
Z	0	0	0	3/4	13/8	3/8	2455/2
$x_2$	0	1	0	1/4	-1/8	1/8	205/2
$x_3$	0	0	1	3/2	-1/4	-3/4	215
$x_1$	1	0	0	-1	1/2	1/2	10

The last tableau is optimal. The new optimal solution is  $x_1^{*new} = 10$ ,  $x_2^{*new} = 205/2$ ,  $x_3^{*new} = 215$ , and  $Z^{*new} = 2425/2$ .

### Optimality Range of $c_j$

Suppose, in the LP of Example 1,  $c_1$  is changed to  $c_1 + \delta$ . In the following, we find the range of  $\delta$  values that keep the LP optimal with the current basis. In this case,

$\mathbf{c}^{new} = (3 + \delta \ 2 \ 5)$ , and  $\mathbf{c}_B^{new} = (2 \ 5 \ 0)$ . Then,  $z_j - c_j$  ( $j$  nonbasic) is given by

$$\begin{aligned} \mathbf{c}_B^{\text{new}} \mathbf{B}^{-1} \mathbf{N} - \mathbf{c}_N^{\text{new}} &= (2 \quad 5 \quad 0) \begin{pmatrix} 1/2 & -1/4 & 0 \\ 0 & 1/2 & 0 \\ -2 & 1 & 1 \end{pmatrix} \begin{pmatrix} \overbrace{x_1 \quad s_1 \quad s_2} \\ 1 & 1 & 0 \\ 3 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} - (3 + \delta \quad 0 \quad 0) \\ &= (4 - \delta \quad 1 \quad 2) \end{aligned}$$

Then, LP is optimal if  $4 - \delta \geq 0 \Leftrightarrow \delta \leq 4 \Leftrightarrow c_1 \leq 7$ .

**Remark.** In cases like the above where  $c_j$  is changed for  $j$  nonbasic. The optimality range for  $c_j$  can be deduced directly from the current optimal tableau. The corresponding  $\delta$  is the coefficient of  $x_j$  in the Z-row of the current optimal tableau. This coefficient is called **reduced cost** of  $x_j$ . It represents the minimum amount by which the objective function coefficient of  $x_j$  should be improved in order for  $x_j$  to become basic. E.g., in the above, increasing  $c_1$  by 4 will make  $x_1$  basic.

#### 4 Addition of a new variable $x_j$

The new variable,  $x_j$ , can be thought of as a nonbasic variable. Consider an optimal tableau. To investigate the effect of adding the new variable  $x_j$  find

$$z_j - c_j = \mathbf{c}_B \mathbf{B}^{-1} \mathbf{A}_j - c_j = \mathbf{y}^* \mathbf{A}_j - c_j,$$

where  $\mathbf{y}^*$  is the current dual variables vector ( $\mathbf{y}^*$  is given by the Z-row coefficients of the starting basic variables).

- If  $z_j - c_j \geq 0$  (for max problem), the optimal solution is unchanged with  $x_j^* = 0$  (since  $x_j$  is nonbasic).
- Otherwise, if  $z_j - c_j < 0$ , then use the simplex method to find a new optimal solution with possibly  $x_j^* > 0$  (since  $x_j$  enters the basis).

#### Example 4.

- In the LP of Example 1, suppose the variable  $x_4$  with  $c_4 = 2$ , and  $\mathbf{A}_4 = \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}$  is

added. Then,

$$z_4 - c_4 = \mathbf{y}^* \mathbf{A}_4 - c_4 = (1 \quad 2 \quad 0) \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix} - 2 = 3 - 2 = 1 > 0.$$



Therefore, the optimal solution is unchanged with  $x_1^* = 0$ ,  $x_2^* = 0$ ,  $x_3^* = 230$ ,  $x_4^* = 0$ , and  $Z^* = 1350$ .

- Suppose now that  $x_4$  with  $c_4 = 4$  and  $\mathbf{A}_4 = \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}$  is added. Then,

$$z_4 - c_4 = \mathbf{y}^* \mathbf{A}_4 - c_4 = (1 \ 2 \ 0) \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix} - 4 = 3 - 4 = -1 < 0.$$

Therefore, the current solution is no longer optimal. Add  $x_4$  to the simplex tableau and proceed with the simplex method. The coefficient of  $x_4$  in the Z-row is  $z_4 - c_4 = -1$ .

The constraint coefficients corresponding to  $x_4$  are given by

$$B^{-1} \mathbf{A}_4 = \begin{pmatrix} 1/2 & -1/4 & 0 \\ 0 & 1/2 & 0 \\ -2 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 1/4 \\ 1/2 \\ 1 \end{pmatrix}.$$

The new simplex tableau is given below with subsequent simplex iterations.

↓

	$x_1$	$x_2$	$x_3$	$x_4$	$S_1$	$S_2$	$S_3$	RHS	Ratio
Z	4	0	0	-1	1	2	0	1350	--
$x_2$	-1/4	1	0	1/4	1/2	-1/4	0	100	400
$x_3$	3/2	0	1	1/2	0	1/2	0	230	460
$S_3$	2	0	0	1	-2	1	1	20	20

→

↓

	$x_1$	$x_2$	$x_3$	$x_4$	$S_1$	$S_2$	$S_3$	RHS	Ratio
Z	6	0	0	0	-1	3	1	1370	--
$x_2$	-3/4	1	0	0	1	-1/2	-1/4	95	95
$x_3$	1/2	0	1	0	1	0	-1/2	220	220
$x_4$	2	0	0	1	-2	1	1	20	--

→

	$x_1$	$x_2$	$x_3$	$x_4$	$S_1$	$S_2$	$S_3$	RHS
Z	21/4	1	0	0	0	5/2	3/4	1465
$S_1$	-3/4	1	0	0	1	-1/2	-1/4	95
$x_3$	5/4	-1	1	0	0	1/2	-1/4	125
$x_4$	1/2	2	0	1	0	0	1/2	210

The last tableau is optimal. The new optimal solution (with  $x_4$  basic) is

$x_1 = 0$ ,  $x_2^* = 0$ ,  $x_3^* = 125$ ,  $x_4^* = 210$ , and  $Z^* = 1465$ .

**Remarks.**

- Determining the effect of changing the constraint coefficients of a nonbasic variable is similar to the above.
- Determining the effect of changing the constraint coefficients of a basic variable is a bit more involved. But it can be done by applying similar principles (see, for example, Bazaraa et al., *Linear Programming and Network Flows*).