*Example1*: Solve the following linear program using the simplex method.

```
Maximize z = 5x_1 + 4x_2
subject to
6x_1 + 4x_2 \le 24x_1 + 2x_2 \le 6-x_1 + x_2 \le 1x_2 \le 2x_1, x_2 \ge 0
```

The inequality of each constraint should convert to equality; thus, the canonical form of the model is converted to standard form as follows:

```
Maximize z = 5x_1 + 4x_2 + 0s_1 + 0s_2 + 0s_3 + 0s_4
subject to
6x_1 + 4x_2 + s_1 = 24
x_1 + 2x_2 + s_2 = 6
-x_1 + x_2 + s_3 = 1
x_2 + s_4 = 2
x_1, x_2, s_1, s_2, s_3, s_4 \ge 0
```

The variables  $s_1$ ,  $s_2$ ,  $s_3$ , and  $s_4$  are the slacks associated with the respective constraints. Next, we write the objective equation as  $z - 5x_1 - 4x_2 = 0$ 

In this manner, the starting simplex tableau can be represented as follows:

Basic	$x_1$	$x_2$	<i>s</i> <sub>1</sub>	<i>s</i> <sub>2</sub>	<b>S</b> 3	<i>S</i> <sub>4</sub>	solution
Z.	-5	-4	0	0	0	0	0
<i>s</i> <sub>1</sub>	6	4	1	0	0	0	24
<b>s</b> <sub>2</sub>	1	2	0	1	0	0	6
<i>s</i> <sub>3</sub>	-1	1	0	0	1	0	1
<i>S</i> 4	0	1	0	0	0	1	2

The result can be seen by setting the non-basic variables  $(x_1, x_2)$  equal to zero in all the equations, and also by noting the special identity-matrix arrangement of the constraint coefficients of the basic variables (all diagonal elements are 1, and all off-diagonal elements are 0).

In the simplex tableau where the objective function is written as  $z - 5x_1 - 4x_2 = 0$ , the selected variable is the non-basic variable with the most negative coefficient in the objective equation. In the terminology of the simplex algorithm,  $x_1$  is known as the entering variable because it enters the basic solution.

If  $x_1$  is the entering variable, one of the current basic variables must leave; that is, it becomes non-basic at zero level (recall that the number of non-basic variable must always be n - m).

The mechanics for determining the *leaving variable* calls for computing the *ratios* of the right-hand side of the equations (Solution column) to the corresponding (strictly) positive constraint coefficients under the entering variable,  $x_1$ , as the following table shows.

Basic	$x_1$	solution	
<i>s</i> <sub>1</sub>	6	24	$x_1 = 24/6 = 4$
<i>s</i> <sub>2</sub>	1	6	$x_1 = 6/1 = 6$
<i>S</i> 3	-1	1	$x_1 = 1/-1 = -1$ (non-negative denominator, ignore)
<i>s</i> <sub>4</sub>	0	2	$x_1=2/0=\infty$ (zero denominator, ignore)

 $x_1$  enters (at level 4) and  $s_1$  leaves (at level zero).

The new solution is determined by "swapping" the entering variable  $x_1$  and the leaving variable  $s_1$  in the simplex tableau to yield non-basic variables which are  $s_1$ ,  $x_2$ . Basic variables are  $x_1$ ,  $s_2$ ,  $s_3$ ,  $s_4$ 

The swapping process is based on the Gauss-Jordan row operations. It identifies the entering variable column as the *pivot column* and the leaving variable row as the pivot row with their intersection being the *pivot element*. The following tableau is a restatement of the starting tableau with its pivot row and column highlighted.

	Enter						
Basic	$x_1$	$x_2$	<i>s</i> <sub>1</sub>	<i>s</i> <sub>2</sub>	S3	<i>S</i> <sub>4</sub>	solution
Z.	-5	-4	0	0	0	0	0
S <sub>1 Leave</sub>	6	4	1	0	0	0	24 Pivot row
<i>s</i> <sub>2</sub>	1	2	0	1	0	0	6
<i>S</i> <sub>3</sub>	-1	1	0	0	1	0	1
<i>s</i> <sub>4</sub>	0	1	0	0	0	1	2
	D' / 1						

Pivot column

Enton

The Gauss-Jordan computations needed to produce the new basic solution include two types.

## 1. Pivot row

a. Replace the leaving variable in the Basic column with the entering variable.

b. New pivot row = Current pivot row ÷ Pivot element.

2. All other rows, including z

*New row* = (*Current row*) – (*Pivot column coefficient*) × (*New pivot row*)

These computations are applied to the preceding tableau in the following manner:

1. Replace  $s_1$  in the Basic column with  $x_1$ : New  $x_1$ -row = Current  $s_1$ -row ÷ 6 = 1/6 (6 4 1 0 0 0 24) = (1 2/3 1/6 0 0 0 4)

2. New z-row = Current z-row – (-5) × New  $x_1$ -row = (-5 -4 0 0 0 0) – (-5) × (1 2/3 1/6 0 0 0) + = (0 -2/3 5/6 0 0 0 20) 3. New  $s_2$ -row = Current  $s_2$ -row – (1) × New  $x_1$ -row = (1 2 0 1 0 0 6) – (1) × (1 2/3 1/6 0 0 0 4) = (0 4/3 -1/6 1 0 0 2) 4. New  $s_3$ -row = Current  $s_3$ -row – (-1) × New  $x_1$ -row = (-1 1 0 0 1 0 1) - (-1) × (1 2/3 1/6 0 0 0 4) = (0 5/6 1/6 0 1 0 5)

5. New  $s_4$ -row = Current  $s_4$ -row - (0) × New  $x_1$ -row = (0 1 0 0 0 1 2) - (0) × (1 2/3 1/6 0 0 0 4) = (0 1 0 0 0 1 2)

The new basic solution is  $(x_1, s_2, s_3, s_4)$ , and the new tableau becomes

Enter

		Pivot colu	mn				
Basic	$x_1$	$x_2$	<i>s</i> <sub>1</sub>	<i>S</i> <sub>2</sub>	<i>S</i> <sub>3</sub>	<i>S</i> <sub>4</sub>	solution
Z.	0	-2/3	5/6	0	0	0	20
$x_1$	1	2/3	1/6	0	0	0	4
S <sub>2</sub> Leave	0	4/3	-1/6	1	0	0	2 Pivot row
<i>S</i> 3	0	5/3	1/6	0	1	0	5
<i>S</i> <sub>4</sub>	0	1	0	0	0	1	2

As a result, when we set the new non-basic variables  $x_2$  and  $s_1$  to zero, the Solution-column automatically yields the new basic solution ( $x_1 = 4$ ,  $s_2 = 2$ ,  $s_3 = 5$ ,  $s_4 = 2$ ) This "conditioning" of the tableau is the result of the application of the Gauss-Jordan row operations. The corresponding new objective value is z = 20.

In the last tableau, the optimality condition shows that  $x_2$  (with the most negative *z*-row coefficient) is the entering variable. The feasibility condition produces the following information:

Basic	$x_2$	solution	
$x_1$	2/3	4	$x_2 = 4 \div 2/3 = 6$
<i>s</i> <sub>2</sub>	4/3	2	$x_2 = 2 \div 4/3 = 1.5$ minimum
<i>S</i> 3	5/3	5	$x_2 = 5 \div 5/3 = 3$
<i>S</i> <sub>4</sub>	1	2	$x_2 = 2 \div 1 = 2$

Thus,  $s_2$  leaves the basic solution, and the new value of  $x_2$  is 1.5. The corresponding increase in z is  $2/3 x_2 = 2/3 \times 1.5 = 1$ , which yields new z = 20 + 1 = 21, as the tableau below confirms. Replacing  $s_2$  in the Basic column with entering  $x_2$ , the following Gauss-Jordan row operations are applied:

1. New pivot  $x_2$ -row = Current  $s_2$ -row ÷ 4/3

2. New z-row = Current z-row –  $(-2/3) \times \text{New } x_2$ -row

3. New  $x_1$ -row = Current  $x_1$ -row – (2/3) × New  $x_2$ -row

4. New  $s_3$ -row = Current  $s_3$ -row – (5/3) × New  $x_2$ -row

5. New  $s_4$ -row = Current  $s_4$ -row – (1) × New  $x_2$ -row

Basic	$x_1$	$x_2$	<i>s</i> <sub>1</sub>	<i>s</i> <sub>2</sub>	<i>S</i> <sub>3</sub>	<i>S</i> <sub>4</sub>	solution
Z.	0	0	3/4	1/2	0	0	21
$x_1$	1	0	1/4	-1/2	0	0	3
$x_2$	0	1	-1/8	3/4	0	0	3/2
<i>s</i> <sub>3</sub>	0	0	3/8	-5/4	1	0	5/2
<b>S</b> 4	0	0	1/8	-3/4	0	1	1/2

The operations above produce the following tableau:

Based on the optimality condition, none of the z-row coefficients are negative. Hence, the last tableau is optimal.

The optimal values of the variables in the *Basic* column are given in the right-hand-side Solution column and can be interpreted as

<b>Decision variables</b>	Optimum value	Recommendations
$x_1$	3	Produce 3 tons of product 1
$x_2$	3/2	Produce 1.5 tons of product 2
Z.	21	Maximum profit

The solution also gives the status of the resources. A resource is designated as scarce if its associated slack variable is zero; that is, the activities (variables) of the model have used the resource completely. Otherwise, if the slack is positive, then the resource is *abundant*. The following table classifies the constraints of the model:

Resource	Slack value	Status
Constraint 1	$s_{1=}0$	Scarce
Constraint 2	$s_{2=}0$	Scarce
Constraint 3 Constraint 4	$s_{3=5/2}$ $s_{4=1/2}$	Abundant Abundant