



Using the Simplex Method to Solve Linear Programming Maximization Problems

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A key problem faced by managers is how to allocate scarce resources among activities or projects. Linear programming, or LP, is a method of allocating resources in an optimal way. It is one of the most widely used operations research (OR) tools. It has been used successfully as a decision-making aid in almost all industries, and in financial and service organizations.

Programming refers to mathematical programming. In this context, it refers to a planning process that allocates resources—labor, materials, machines, and capital—in the best possible (optimal) way so that costs are minimized or profits are maximized. In LP, these resources are known as *decision variables*. The criterion for selecting the best values of the decision variables (e.g., to maximize profits or minimize costs) is known as the *objective function*. The limitations on resource availability form what is known as a *constraint set*.

For example, let's say a furniture manufacturer produces wooden tables and chairs. Unit profit for tables is \$6, and unit profit for chairs is \$8. To simplify our discussion, let's assume the only two resources the company uses to produce tables and chairs are wood (board feet) and labor (hours). It takes 30 bf and 5 hours to make a table, and 20 bf and 10 hours to make a chair.

About this series

According to the Operations Research Society of America, "Operations research [OR] is concerned with scientifically deciding how to best design and operate man-machine systems, usually under conditions requiring the allocation of scarce resources."

This publication, part of a series, should be useful for supervisors, lead people, middle managers, and anyone who has planning responsibility for either a single manufacturing facility or for corporate planning over multiple facilities. Although managers and planners in other industries can learn about OR techniques through this series, practical examples are geared toward the wood products industry.

Decision variables. . .

“The resources available.”

Constraint set. . .

“The limitations on resource availability.”

Objective function. . .

“The criterion for selecting the best values of the decision variables.”

There are 300 bf of wood available and 110 hours of labor available. The company wishes to maximize profit, so profit maximization becomes the *objective function*. The resources (wood and labor) are the *decision variables*. The limitations on resource availability (300 bf of wood and 110 hours of labor) form the *constraint set*, or operating rules that govern the process. Using LP, management can decide how to allocate the limited resources to maximize profits.

The “linear” part of the name refers to the following:

- The objective function (i.e., maximization or minimization) can be described by a linear function of the decision variables, that is, a mathematical function involving only the first powers of the variables with no cross products. For example, $23X_2$ and $4X_{16}$ are valid decision variable terms, while $23X_2^2$, $4X_{16}^3$, and $(4X_1 * 2X_1)$ are not. The entire problem can be expressed as straight lines, planes, or similar geometrical figures.
- The constraint set can be expressed as a set of linear equations.

In addition to the linear requirements, nonnegativity conditions state that the variables cannot assume negative values. It is not possible to have negative resources. Without these conditions, it would be mathematically possible to use more resources than are available.

In EM 8719, *Using the Graphical Method to Solve Linear Programs*, we use the graphical method to solve an LP problem involving resource allocation and profit maximization for a furniture manufacturer. In that example, there were only two variables (wood and labor), which made it possible to solve the problem graphically.

Problems with three variables also can be graphed, but three-dimensional graphs quickly become cumbersome. Problems with more than three variables cannot be graphed. Most real-world problems contain numerous objective criteria and resources, so they’re too complicated to represent with only two or three variables. Thus, for all practical purposes, the graphical method for solving LP problems is used only to help students better understand how other LP solution procedures work.

This publication will build on the example of the furniture company by introducing a way to solve a more complex LP problem. The method we will use is the *simplex* method.

Overview of the simplex method

The simplex method is the most common way to solve large LP problems. *Simplex* is a mathematical term. In one dimension, a simplex is a line segment connecting two points. In two dimensions, a simplex is a triangle formed by joining the points. A three-dimensional simplex is a four-sided pyramid having four corners. The underlying concepts are geometrical, but the solution algorithm, developed by George Dantzig in 1947, is an algebraic procedure.

As with the graphical method, the simplex method finds the most attractive corner of the feasible region to solve the LP problem. Remember, any LP problem having a solution must have an optimal solution that corresponds to a corner, although there may be multiple or alternative optimal solutions.

Simplex usually starts at the corner that represents doing nothing. It moves to the neighboring corner that best improves the solution. It does this over and over again, making the greatest possible improvement each time. When no more improvements can be made, the most attractive corner corresponding to the optimal solution has been found.

A moderately sized LP with 10 products and 10 resource constraints would involve nearly 200,000 corners. An LP problem 10 times this size would have more than a trillion corners. Fortunately, the search procedure for the simplex method is efficient enough that only about 20 of the 200,000 corners are searched to find the optimal solution.

In the real world, computer software is used to solve LP problems using the simplex method, but you will better understand the results if you understand how the simplex method works. The example in this publication will help you do so.

Review of the graphical method

First, let's quickly review the graphical procedure for solving an LP problem, which is presented in EM 8719, *Using the Graphical Method to Solve Linear Programs*. Let's say a furniture manufacturer wishes to maximize profit. Information about available resources (board feet of wood and hours of labor) and the objective criterion is presented in Table 1. For a complete, step-by-step review of the graphical method, see EM 8719 or one of the textbooks listed in the "For more information" section.



Tip . .

In our example, X_1 refers to tables, X_2 refers to chairs, and Z refers to profit.

Table 1.—Information for the wooden tables and chairs linear programming problem.

Resource	Table (X_1)	Chair (X_2)	Available
Wood (bf)	30	20	300
Labor (hr)	5	10	110
Unit profit	\$6	\$8	

Maximize: $Z = 6X_1 + 8X_2$ (objective function)

Subject to: $30X_1 + 20X_2 \leq 300$ (wood constraint: 300 bf available)

$5X_1 + 10X_2 \leq 110$ (labor constraint: 110 hours available)

$X_1, X_2 \geq 0$ (nonnegativity conditions)

Based on the above information, graphically solve the LP (Figure 1). Graph the two constraint equation lines. Then plot two objective function lines by arbitrarily setting $Z = 48$ and $Z = 72$ to find the direction to move to determine the most attractive corner. The coordinates for the most attractive corner (where the wood and labor constraint equations intersect) can be found by simultaneously solving the constraint equations with two unknowns.

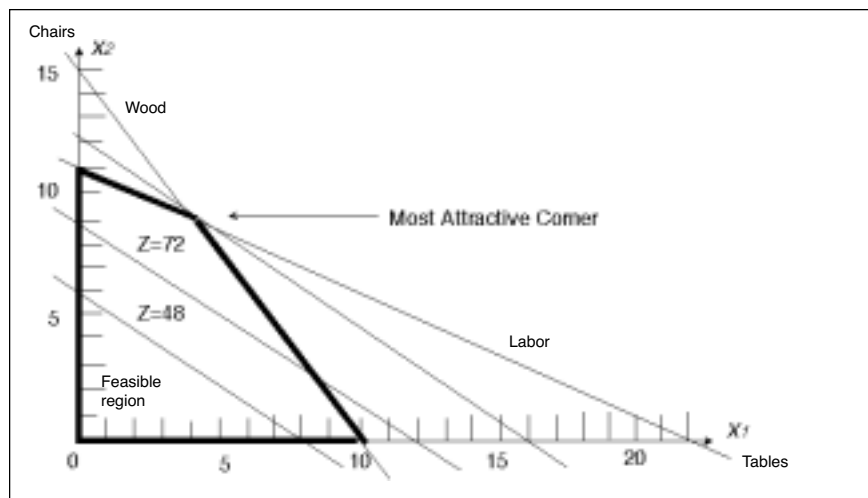


Figure 1.—Determining the most attractive corner corresponding to the optimal solution.

To simultaneously solve the two constraint equations, first multiply the labor equation by -2 , and add it to the wood equation:

$$\begin{array}{r} 30X_1 + 20X_2 = 300 \quad (\text{wood}) \\ -2(5X_1 + 10X_2 = 110) \quad (\text{labor}) \\ \hline 20X_1 + 0 = 80 \\ X_1 = 4 \text{ tables} \end{array}$$

Next, substitute into either of the constraint equations to find the number of chairs. We can substitute into both equations to illustrate that the same value is found.

Wood constraint

$$30(4) + 20X_2 = 300$$

$$120 + 20X_2 = 300$$

$$20X_2 = 300 - 120$$

$$20X_2 = 180$$

$$X_2 = 180/20$$

$$X_2 = 9 \text{ chairs}$$

Labor constraint

$$5(4) + 10X_2 = 110$$

$$20 + 10X_2 = 110$$

$$10X_2 = 110 - 20$$

$$10X_2 = 90$$

$$X_2 = 90/10$$

$$X_2 = 9 \text{ chairs}$$

Now, determine the value of the objective function for the optimal solution. Substitute into the equation the number of tables and chairs, and solve for Z .

$$Z = \$6(4) + \$8(9) = \$96$$

The optimal solution is to manufacture four tables and nine chairs for a profit of \$96.



Using the simplex method

By introducing the idea of *slack variables* (unused resources) to the tables and chairs problem, we can add two more variables to the problem. With four variables, we can't solve the LP problem graphically. We'll need to use the simplex method to solve this more complex problem.

We'll briefly present the steps involved in using the simplex method before working through an example. Table 2 shows an example of a simplex tableau. Although these steps will give you a general overview of the procedure, you'll probably find that they become much more understandable as you work through the example.

A list of shortcuts is found on page 23. You can refer to the six steps and shortcuts while working through the example.

Step 1. Formulate the LP and construct a simplex tableau. Add slack variables to represent unused resources, thus eliminating inequality constraints. Construct the simplex tableau—a table that allows you to evaluate various combinations of resources to determine which mix will most improve your solution. Use the slack variables in the starting basic variable mix.

Step 2. Find the sacrifice and improvement rows. Values in the sacrifice row indicate what will be lost in per-unit profit by making a change in the resource allocation mix. Values in the improvement row indicate what will be gained in per-unit profit by making a change.

Table 2.—Example of a simplex tableau.

Unit profit		6	8	0	0		
	Basic mix	X_1	X_2	S_w	S_L	Solution	
0	S_w	30	20	1	0	300	
0	S_L	5	10	0	1	110	
	Sacrifice	0	0	0	0	0	← Current profit
	Improvement	6	8	0	0	—	

Step 3. Apply the entry criteria. Find the entering variable and mark the top of its column with an arrow pointing down. The entering variable is defined as the current non-basic variable that will most improve the objective if its value is increased from 0. If ties occur, arbitrarily choose one as the entering variable. When no improvement can be found, the optimal solution is represented by the current tableau.

If no positive number appears in the entering variable's column, this indicates that one or more constraints are unbounded. Since it is impossible to have an unlimited supply of a resource, an unbounded solution indicates that the LP problem was formulated incorrectly.

Step 4. Apply the exit criteria. Using the current tableau's exchange coefficient from the entering variable column, calculate the following exchange ratio for each row as:

$$\text{Solution value/Exchange coefficient}$$

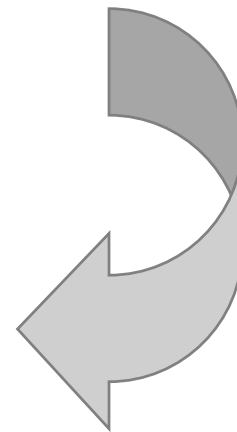
The exchange ratio tells you which variable is the *limiting resource*, i.e., the resource that would run out first.

Find the lowest nonzero and nonnegative value. This variable is the limiting resource. The basic variable in this row becomes the exiting variable. In case of identical alternatives, arbitrarily choose one. Mark the exiting variable row with an arrow pointing left.

Step 5. Construct a new simplex tableau. Constructing a new tableau is a way to evaluate a new corner. One variable will enter the basic mix (entering variable), and one variable will leave the basic mix and become a non-basic variable (exiting variable). The operation of an entering variable and an exiting variable is called a *pivot operation*. The simplex method is made up of a sequence of such pivots. The pivot identifies the next corner to be evaluated. The new basic mix always differs from the previous basic mix by one variable (exiting variable being replaced by the entering variable).

To construct the new tableau, replace the exiting variable in the basic mix column with the new entering variable. Other basic mix variables remain unchanged. Change the unit profit or unit loss column with the value for the new entering variable. Compute the new row values to obtain a new set of exchange coefficients applicable to each basic variable.

Step 6. Repeat steps 2 through 5 until you no longer can improve the solution.



Slack variables. . .

“A mathematical representation of surplus resources.” In real-life problems, it’s unlikely that all resources will be used completely, so there usually are unused resources.

A simplex method example: Production of wooden tables and chairs

Step 1. Formulate the LP and construct a simplex tableau.

From the information in Table 3, we can formulate the LP problem as before.

Table 3.—Information for the wooden tables and chairs linear programming problem.

Resource	Table (X_1)	Chair (X_2)	Available
Wood (bf)	30	20	300
Labor (hr)	5	10	110
Unit profit	\$6	\$8	

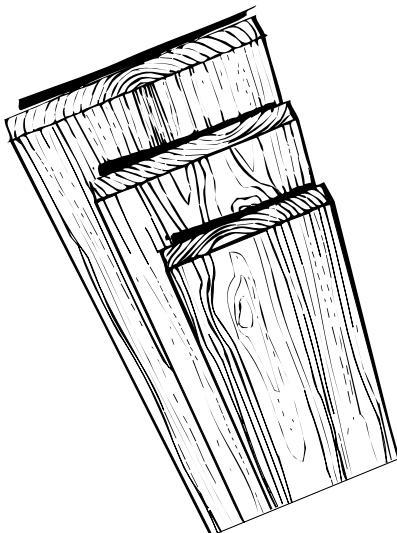
$$\begin{aligned} \text{Maximize: } & Z = 6X_1 + 8X_2 && \text{(objective function)} \\ \text{Subject to: } & 30X_1 + 20X_2 \leq 300 && \text{(wood constraint: 300 bf available)} \\ & 5X_1 + 10X_2 \leq 110 && \text{(labor constraint: 110 hours available)} \\ & X_1, X_2 \geq 0 && \text{(nonnegativity conditions)} \end{aligned}$$

Slack variables

Using the simplex method, the first step is to recognize surplus resources, represented in the problem as *slack variables*. In most real-life problems, it’s unlikely that all resources (usually a large mix of many different resources) will be used completely. While some might be used completely, others will have some unused capacity. Also, slack variables allow us to change the inequalities in the constraint equations to equalities, which are easier to solve algebraically. Slack variables represent the unused resources between the left-hand side and right-hand side of each inequality; in other words, they allow us to put the LP problem into the standard form so it can be solved using the simplex method.

The first step is to convert the inequalities into equalities by adding slack variables to the two constraint inequalities. With S_w representing surplus wood, and S_L representing surplus labor, the constraint equations can be written as:

$$\begin{aligned} 30X_1 + 20X_2 + S_w &= 300 && \text{(wood constraint: 300 bf)} \\ 5X_1 + 10X_2 + S_L &= 110 && \text{(labor constraint: 110 hours)} \end{aligned}$$



All variables need to be represented in all equations. Add slack variables to the other equations and give them coefficients of 0 in those equations. Rewrite the objective function and constraint equations as:

$$\begin{aligned} \text{Maximize: } Z &= 6X_1 + 8X_2 + 0S_w + 0S_L && \text{(objective function)} \\ \text{Subject to: } 30X_1 + 20X_2 + S_w + 0S_L &= 300 && \text{(wood constraint: 300 bf)} \\ 5X_1 + 10X_2 + 0S_w + S_L &= 110 && \text{(labor constraint: 110 hours)} \\ X_1, X_2, S_w, S_L &\geq 0 && \text{(nonnegativity conditions)} \end{aligned}$$

We can think of slack or surplus as unused resources that don't add any value to the objective function. Thus, their coefficients are 0 in the objective function equation.

Basic variable mix and non-basic variables

Since there are more unknown variables (four) than equations (two), we can't solve for unique values for the X and S variables using algebraic methods. Whenever the number of variables is greater than the number of equations, the values of the extra variables must be set arbitrarily, and then the other variables can be solved for algebraically.

First we'll choose which variables to solve for algebraically. These variables are defined to be in the *basic variable mix*. We can solve for these variables after we fix the other variables at some arbitrary level.

The fixed-value variables are identified as not being in the basic mix and are called *non-basic variables*. We'll arbitrarily give the non-basic variables the value of 0. The algebraic solution of the constraint equations, with non-basic variables set to 0, represents a corner.

For any given set of variables, there are several possible combinations of basic variables and non-basic variables. For illustration, Table 4 contains the six basic mix pairs and the corresponding non-basic variables for the tables and chairs LP problem. Figure 2 illustrates where each corner (A through F in Table 4) lies on a graph.

Basic variable mix. . .

“The variables that we choose to solve for algebraically.”

Non-basic variables. . .

“Variables that are arbitrarily given a value of 0 so that we can solve for the variables in the basic variable mix.”

Table 4.—Basic variable mix combinations and algebraic solutions.

Basic variable mix	Non-basic variables	Algebraic solution				Z(\$)	Corners (Figure 2)
		X_1	X_2	S_w	S_L		
$S_w S_L$	$X_1 X_2$	0	0	300	110	0	A
$S_w X_2$	$X_1 S_L$	0	11	80	0	88	B
$S_L X_1$	$X_2 S_w$	10	0	0	60	60	C
$X_1 X_2$	$S_w S_L$	4	9	0	0	96	D
$S_L X_2$	$S_w X_1$	0	15	0	-40	infeasible	E
$S_w X_1$	$S_L X_2$	22	0	-360	0	infeasible	F

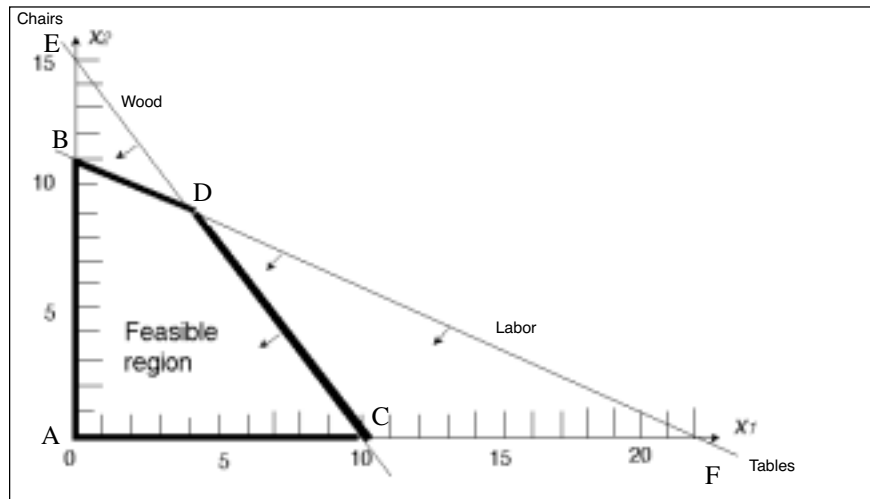


Figure 2.—Corners corresponding to Table 4 data.

We can evaluate each corner to find the values of the basic variables and of Z:

Corner A:

$$\text{Set } X_1 = 0 \text{ and } X_2 = 0$$

$$30(0) + 20(0) + S_w + 0S_L = 300$$

$$5(0) + 10(0) + 0S_w + S_L = 110$$

$$\text{Therefore, } S_w = 300$$

$$\text{and } S_L = 110$$

$$\text{Solution: } X_1 = 0, X_2 = 0, S_w = 300, S_L = 110$$

$$\text{Profit: } Z = \$6(0) + \$8(0) + 0(300) + 0(110) = 0$$

Corner B:

Set $X_1 = 0$ and $S_L = 0$

$$30(0) + 20X_2 + S_w + 0(0) = 300$$

$$5(0) + 10X_2 + 0S_w + 0 = 110$$

Therefore, $20X_2 + S_w = 300$

and $10X_2 = 110$

Solution: $X_1 = 0, X_2 = 11, S_w = 80, S_L = 0$

Profit: $Z = \$6(0) + \$8(11) + 0(80) + 0(0) = \$88$

Corner C:

Set $X_2 = 0$ and $S_w = 0$

$$30X_1 + 20(0) + 0 + 0S_L = 300$$

$$5X_1 + 10(0) + 0(0) + S_L = 110$$

Therefore, $30X_1 = 300$

and $5X_1 + S_L = 110$

Solution: $X_1 = 10, X_2 = 0, S_w = 0, S_L = 60$

Profit: $Z = \$6(10) + \$8(0) + 0(0) + 0(60) = \$60$

Corner D:

Set $S_w = 0$ and $S_L = 0$

$$30X_1 + 20X_2 + 0 + 0(0) = 300$$

$$5X_1 + 10X_2 + 0(0) + 0 = 110$$

Therefore, $30X_1 + 20X_2 = 300$

and $5X_1 + 10X_2 = 110$

Solution: $X_1 = 4, X_2 = 9, S_w = 0, S_L = 0$

Profit: $Z = \$6(4) + \$8(9) + 0(0) + 0(0) = \$96$

Point E is infeasible because it violates the labor constraint, and point F is infeasible because it violates the wood constraint. The simplex algorithm never evaluates infeasible corners.

Remember, with slack variables added, the tables and chairs LP is now four-dimensional and is not represented by Figure 2. Points on the constraint lines in Figure 2 represent 0 slack for both wood

The simplex algorithm never evaluates infeasible corners, i.e., those that violate one of the constraint equations.

Most real-world problems are too complex to solve graphically. They have too many corners to evaluate, and the algebraic solutions are lengthy. A simplex tableau is a way to systematically evaluate variable mixes in order to find the best one.

and labor resources. A feasible point off a constraint line represents positive slack and cannot be read off the two-dimensional graph.

In Table 4, all corners in the LP were identified, and all feasible corners were algebraically evaluated to find the optimum solution. You can see that a graph wasn't necessary to list all variable mixes and that each variable-mix pair corresponded to a corner solution.

One reason we can't use this procedure to solve most LP problems is that the number of corners for real-life LP problems usually is very large. Another reason is that each corner evaluation requires a lengthy algebraic solution. To obtain each corner solution for a 10-constraint linear program, 10 equations with 10 unknowns must be solved, which is not a simple arithmetic task. Many LP problems are formulated with many more than 10 constraints.

Simplex tableau

As you'll recall, we formulated the tables and chairs LP in standard form as:

$$\begin{aligned} \text{Maximize: } Z &= 6X_1 + 8X_2 + 0S_w + 0S_L && \text{(objective function)} \\ \text{Subject to: } 30X_1 + 20X_2 + S_w + 0S_L &= 300 && \text{(wood constraint: 300 bf)} \\ &5X_1 + 10X_2 + 0S_w + S_L = 110 && \text{(labor constraint: 110 hours)} \\ &X_1, X_2, S_w, S_L \geq 0 && \text{(nonnegativity conditions)} \end{aligned}$$

The information for the tables and chairs example can be incorporated into a simplex *tableau* (Table 5). A tableau is a table that allows you to evaluate the values of variables at a given corner to determine which variable should be changed to most improve the solution.

Table 5.—Tables and chairs simplex tableau.

Unit profit		6	8	0	0	
Basic mix		X_1	X_2	S_w	S_L	Solution
S_w		30	20	1	0	300
S_L		5	10	0	1	110

The top of the tableau lists the per-unit profit for the objective function. The rows in the body of the tableau indicate the basic variable mix for the corner point being evaluated. The first row in the body of the tableau lists the coefficients of the first constraint

equation (the wood constraint) in their original order. The second row lists the coefficients of the second constraint equation (the labor constraint) in their original order.

The basic mix column lists the slack variables. All variables not listed in this column are designated as non-basic variables and will be arbitrarily fixed at a value of 0 when we plug them into the constraint equations.

The solution column lists the values of the basic variables, $S_w = 300$ and $S_L = 110$. Thus, the solution mix shows that all of the resources (wood and labor) remain unused. The solutions to the two constraint equations after zeroing out X_1 and X_2 are as follows:

$$30(0) + 20(0) + S_w + 0S_L = 300 \quad \text{or} \quad S_w = 300$$

$$5(0) + 10(0) + 0S_w + S_L = 110 \quad \text{or} \quad S_L = 110$$

The original constraint coefficients, highlighted in the simplex tableau, are called *exchange coefficients*. They indicate how many units of the variable listed on the left (basic mix column) must be given up to achieve a unit increase in the variable listed at the top of the tableau. The 30 indicates that 30 board feet of unused wood can be exchanged for one table, and the 20 indicates that 20 board feet of wood can be exchanged for one chair. Likewise, 5 hours of labor can be exchanged for one table, and 10 hours of labor can be exchanged for one chair.

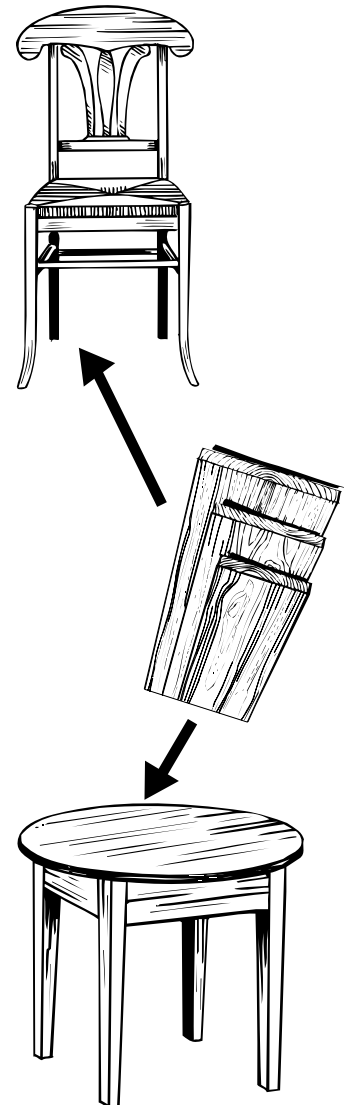
The exchange coefficients are 0 or 1 for the basic mix variables. These numbers are not very meaningful. For example, the 1 in the first row indicates that 1 board foot of wood can be exchanged for 1 board foot of wood. The 0 in row one indicates that no unused wood is required to accommodate more unused labor.

Step 2. Find the sacrifice and improvement rows.

The next step is to expand the simplex tableau as in Table 6.

Table 6.—Expanded simplex tableau.

Unit profit		6	8	0	0	
	Basic mix	X_1	X_2	S_w	S_L	Solution
0	S_w	30	20	1	0	300
0	S_L	5	10	0	1	110
	Sacrifice	0	0	0	0	0 ← Current profit
	Improvement	6	8	0	0	—



The other sacrifice row values are calculated in the same manner. In the case of our example, the unit profit values are 0, so the sacrifice row values are all 0.

The values in the solution column of the sacrifice row are calculated as:

$$\begin{array}{rclcl}
 \text{(Unit profit column value)} & * & \text{(Solution column value)} & & \\
 \hline
 & 0 & * & 300 & = 0 \\
 + & 0 & * & 110 & = 0 \text{ Add the products} \\
 & & & & 0
 \end{array}$$

The sum of these products represents the current profit (Z).

Improvement row

Improvement row values are calculated by subtracting each value in the sacrifice row from the value found above it in the unit profit row. Therefore:

$$\text{Unit improvement} = \text{Unit profit} - \text{Unit sacrifice}$$

For example, the improvement for X_1 is calculated as:

$$\begin{array}{r}
 \text{Unit profit} = 6 \\
 - \text{Unit sacrifice} = 0 \\
 \hline
 \$6 \text{ (first value in the improvement row)}
 \end{array}$$

Since all of the sacrifice values are 0 in this example, all of the improvement row values are the same as those found in the unit profit row.

Step 3. Apply the entry criterion.

The next step is to apply the entry criterion, that is, to determine the *entering variable*. The entering variable is defined as the current non-basic variable that will most improve the objective if it is increased from 0. It's called the entering variable because it will enter the basic mix when you construct your next tableau to evaluate a new corner. For profit maximization problems, you determine the entering variable by finding the largest value in the improvement row.

Entering variable. . .

“The current non-basic variable that will most improve the objective if its value is increased from 0.”



In our example, the largest value in the improvement row is 8. Thus, we can increase profit (improve the current solution) by \$8 per unit for each chair made. Increasing the value of X_2 from 0 to 8 is the best improvement that can be made. If we increase the value of X_1 , our solution improves by only \$6. Therefore, X_2 is the entering variable. The entering variable is marked by placing a downward facing arrow in the X_2 (chair) column (Table 7).

Table 7.—Entering variable, exchange ratios, exiting variable, and pivot element.

Unit profit		6	8	0	0		Exchange ratios:
	Basic mix	X_1	$X_2 \downarrow$	S_w	S_L	Solution	
0	S_w	30	20	1	0	300	$300/20 = 15$
0	$S_L \leftarrow$	5	10	0	1	110	$110/10 = 11$
	Sacrifice	0	0	0	0	0	← Current profit
	Improvement	6	8	0	0	—	

Step 4. Apply the exit criterion.

The next step is to determine the *exiting variable*. The exiting variable is the variable that will exit the basic mix when you construct your next simplex tableau.

We'll find the exiting variable by calculating the *exchange ratio* for each basic variable. The exchange ratio tells us how many tables or chairs can be made by using all of the resource for the current respective basic variable. To find the exchange ratio, divide the solution value by the corresponding exchange coefficient in the entering variable column. The exchange ratios are:

$$300/20 = 15 \quad (S_w \text{ basic mix row})$$

$$\text{and } 110/10 = 11 \quad (S_L \text{ basic mix row})$$

By using all 300 board feet of wood, we can make 15 chairs because it takes 20 board feet of wood to make a chair. By using all 110 hours of labor, we can make 11 chairs because it takes 10 hours of labor per chair. Thus, it's easy to see the plant can't manufacture 15 chairs. We have enough wood for 15 chairs but only enough labor for 11. In this case, labor is the *limiting resource*. If all the labor were used, there would be leftover wood.

Limiting resource. . .
 "The resource that would run out first."

Exchange ratio. . .
 "Tells you which variable is the limiting resource."

The lowest nonzero and nonnegative exchange ratio denotes the limiting resource. The basic variable in this row becomes the exiting variable.

The exit criterion requires that the limiting resource (the basic mix variable with the smallest exchange ratio) exit the basic mix. In this case, the exiting variable is S_L . Because of this, wood (S_w) remains in the basic mix. Indicate the exiting variable by placing a small arrow pointing toward the S_L (Table 7).

Next circle the *pivot element*—the value found at the intersection of the entering variable column and the exiting variable row. In this case, the value 10 (X_2 column and S_L row) is the pivot element. We'll use this value to evaluate the next corner point represented by exchanging X_2 and S_L .

Step 5. Construct a new simplex tableau.

The next step is to create a new simplex tableau. First, let's look at the old constraint equations that represented the X_1 and X_2 rows in our original tableau:

$$30X_1 + 20X_2 + S_w + 0S_L = 300 \quad (\text{wood constraint})$$

$$5X_1 + 10X_2 + 0S_w + S_L = 110 \quad (\text{labor constraint})$$

Since X_2 is to replace S_L , we need to transform the second equation so that X_2 will have a coefficient of 1. This requires some algebraic manipulation. Although the resulting equations will look different, they will be equivalent to the original constraints of the LP problem.

First, we'll multiply the labor constraint equation by $1/10$ (the same as dividing each of the variables by 10). We get the following equivalent equation:

$$\frac{1}{2}X_1 + X_2 + 0S_w + \frac{1}{10}S_L = 11$$

This now becomes the new X_2 row. By setting the non-basic variables, X_1 and S_L , both to 0, we get:

$$\frac{1}{2}(0) + X_2 + 0S_w + \frac{1}{10}(0) = 11$$

$$X_2 = 11$$

We want the solution, $X_2 = 11$, to satisfy both constraint equations. We can do this by zeroing the X_2 term in the first equation (the wood constraint). We'll do this by multiplying the second equation by -20 and adding it to the first equation:

Multiply times -20:

$$-20(\frac{1}{2}X_1 + X_2 + 0S_w + \frac{1}{10}S_L) = 11$$

$$-10X_1 - 20X_2 - 0S_w - 2S_L = -220$$

Constructing a new tableau is a way to evaluate a new corner point. One variable will enter the basic mix (entering variable), and one variable will leave the basic mix and become a non-basic variable (exiting variable). The operation of an entering variable and an exiting variable is called a pivot operation. The simplex method is made up of a sequence of such pivots.

Add to the first equation:

$$\begin{array}{r} 30X_1 + 20X_2 + S_w + 0S_L = 300 \\ -10X_1 - 20X_2 - 0S_w - 2S_L = -220 \\ \hline 20X_1 + 0X_2 + S_w - 2S_L = 80 \end{array}$$

This equation becomes the new S_w row. Thus, our new constraint equations are:

$$20X_1 + 0X_2 + S_w - S_L = 80 \quad (\text{wood})$$

$$\frac{1}{2}X_1 + X_2 + 0S_w + \frac{1}{10}S_L = 11 \quad (\text{labor})$$

When the non-basic variables X_1 and S_L are set to 0, the solution becomes:

$$20(0) + 0 + S_w - 2(0) = 80$$

$$S_w = 80$$

$$\text{and } \frac{1}{2}(0) + X_2 + 0 + \frac{1}{10}(0) = 11$$

$$X_2 = 11$$

These two new equations give us some information. Remember, S_w represents the amount of surplus or slack wood, that is, the amount of wood not used. When 11 chairs are manufactured, 80 board feet of surplus or slack wood will remain.

The new simplex tableau is shown in Table 8c.

An easier method

The above equations can be calculated much more easily directly from the original simplex tableau than by doing the algebra. Refer to Tables 8a through 8c as we work through the example.

1. Fill in the new X_2 row.

Referring to Table 8a, divide all values in the exiting variable row, S_L , by the pivot element, 10. The calculations are $5/10$, $10/10$ (pivot element divided by itself), $0/10$, and $1/10$. Place the new values in the same location in the new tableau (Table 8b).

It's easier to find a solution by using the simplex tableau than by doing the algebra.

Place the unit profit row value for X_2 , the new entering variable (8), into the unit profit column.

Table 8a.—Original simplex tableau.

Unit profit		6	8	0	0		
	Basic mix	X_1	$X_2 \downarrow$	S_w	S_L	Solution	Exchange ratios:
0	S_w	30	20	1	0	300	$300/20 = 15$
0	S_L ←	5	10	0	1	110	$110/10 = 11$
	Sacrifice	0	0	0	0	0 ←	Current profit
	Improvement	6	8	0	0	—	

Table 8b.—Second simplex tableau— X_2 row.

Unit profit		6	8	0	0	
	Basic mix	X_1	X_2	S_w	S_L	Solution
0	S_w					
8	X_2	0.5	1	0	0.1	11
	Sacrifice					
	Improvement					

2. Fill in the new S_w row.

Now we'll find the values for the S_w row. Referring to Table 8a, find the value in the S_w row in the old tableau in the pivot element column (20). Multiply it times the first value in the new X_2 row (0.5 from Table 8b). Subtract your answer from the value in the first position of the old S_w row.

Thus, for the first value (to replace the 30 in the first tableau):

$$(20 * 0.5) = 10$$

$$30 - 10 = 20$$

For the second value (to replace 20 in the first tableau):

$$(20 * 1) = 20$$

$$20 - 20 = 0$$

For the third value in this row:

$$20 * 0 = 0$$

$$1 - 0 = 1 \text{ (Stays the same in the new tableau.)}$$

For the fourth value in this row:

$$20 * 0.1 = 2$$

$$0 - 2 = -2$$

For the solution column value for this row:

$$20 * 11 = 220$$

$$300 - 220 = 80$$

The new S_w row is shown in Table 8c.

Table 8c.—Second simplex tableau— S_w row.

Unit profit		6	8	0	0	
	Basic mix	X_1	X_2	S_w	S_L	Solution
0	S_w	20	0	1	-2	80
8	X_2	0.5	1	0	0.1	11
	Sacrifice					
	Improvement					

3. Find the sacrifice and improvement rows.

Find the sacrifice and improvement rows using the same method as in the first tableau. See Table 8d.

Table 8d.—Second simplex tableau—sacrifice and improvement rows.

Unit profit		6	8	0	0	
	Basic mix	X_1	X_2	S_w	S_L	Solution
0	S_w	20	0	1	-2	80
8	X_2	0.5	1	0	0.1	11
	Sacrifice	4	8	0	0.8	88 ← Current profit
	Improvement	2	0	0	-0.8	—

We now see that profit has been improved from 0 to \$88.

4. Complete the pivot operation (entering and exiting variables).

Recall that the pivot operation results in new entering and exiting variables. The greatest per-unit improvement is

2 (X_1 column). The others offer no improvement (either 0 or a negative number). X_1 becomes the new entering variable. Mark the top of its column with an arrow (Table 8e). Remember, when no improvement can be found at this step, the current tableau represents the optimal solution.

Now determine the exiting variable. To do so, first determine the exchange ratios:

$$80/20 = 4$$

$$\text{and } 11/0.5 = 22$$

Now choose the smallest nonnegative exchange ratio (4 versus 22). S_w becomes the exiting variable. Mark that row with an arrow. Draw a circle around the pivot element, 20. (Table 8e).

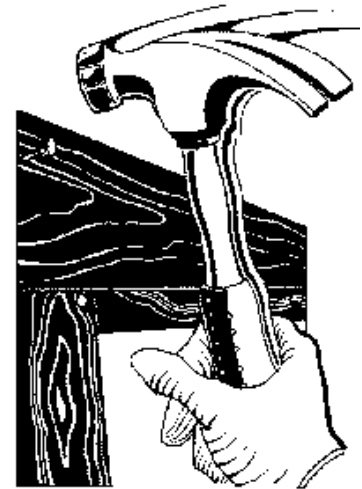


Table 8e.—Second simplex tableau—pivot operation.

Unit profit		6	8	0	0		Exchange ratios:
	Basic mix	$X_1 \downarrow$	X_2	S_w	S_L	Solution	
0	$S_w \leftarrow$	20	0	1	-2	80	$80/20 = 4$
8	X_2	0.5	1	0	0.1	11	$11/0.5 = 22$
	Sacrifice	4	8	0	0.8	88	
	Improvement	2	0	0	-0.8	—	

5. Construct the third tableau from the second tableau.

Replace the entering variable in the basic mix where the exiting variable left. Bring over the unit profit from the top row of the old table to the new table. Fill in the pivot element row by dividing through by the pivot element (Table 8f).

Table 8f.—Third simplex tableau— X_1 row.

Unit profit		6	8	0	0	
	Basic mix	X_1	X_2	S_w	S_L	Solution
6	X_1	1	0	0.05	-0.1	4
8	X_2					
	Sacrifice					
	Improvement					

Fill in the first value in the X_2 row as follows. First, multiply the previous tableau's X_2 pivot value (0.5) times the first value in the new tableau's X_1 row (1):

$$0.5 * 1 = 0.5$$

Now subtract this number from the first value in the previous tableau's X_2 row (0.5):

$$0.5 - 0.5 = 0$$

Place this value in the first position of the new tableau's X_2 row. Repeat this process to fill in the remaining values in the new X_2 row (Table 8g).

Table 8g.—Third simplex tableau— X_2 row.

		Unit profit				
		6	8	0	0	
	Basic mix	X_1	X_2	S_w	S_L	Solution
6	X_1	1	0	0.05	-0.1	4
8	X_2	0	1	-0.025	0.15	9
	Sacrifice					
	Improvement					

Fill in the sacrifice row (Table 8h). The first value is $(6 * 1) + (8 * 0) = 6$.

Fill in the improvement row. The first value is $6 - 6 = 0$.

Table 8h.—Third simplex tableau—sacrifice and improvement rows.

		Unit profit				
		6	8	0	0	
	Basic mix	X_1	X_2	S_w	S_L	Solution
6	X_1	1	0	0.05	-0.1	4
8	X_2	0	1	-0.025	0.15	9
	Sacrifice	6	8	0.1	0.6	96
	Improvement	0	0	-0.1	-0.6	—

There are no positive numbers in the new improvement row. Thus, we no longer can improve the solution to the problem. This simplex tableau represents the optimal solution to the LP problem and is interpreted as:

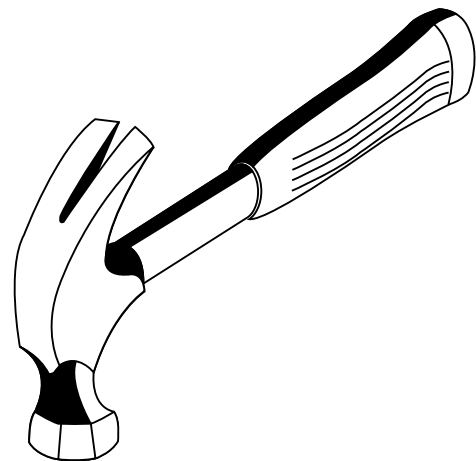
$$X_1 = 4, X_2 = 9, S_w = 0, S_L = 0, \text{ and profit or } Z = \$96$$

The optimal solution (maximum profit to be made) is to manufacture four tables and nine chairs for a profit of \$96.

Shortcuts

Several shortcuts can make the construction of simplex tableaus easier.

- In the new tableau, only the columns for the non-basic and exiting variables change. Move the values for all other basic variables directly into the new tableau.
- When 0 is found in the pivot column, that row always is the same in the new tableau. When 0 is found in the pivot row, that column always is the same in the new tableau.
- With two exceptions, the newly entered basic variable's column will contain a zero in all locations. A 1 will go in the same location as the pivot element in the preceding tableau, and the per-unit profit or loss for this variable will appear in the sacrifice row.



Summary

In this publication, the simplex method was used to solve a maximization problem with constraints of the form of \leq (less than or equal to). In the next publication in this series, we will discuss how to handle \geq (greater than or equal to) and $=$ (equal to) constraints. We'll also explore how to use LP as an economic tool with sensitivity analysis.



Quick review of the simplex method

Step 1. Formulate the LP and construct a simplex tableau.

Add slack variables to represent unused resources.

Step 2. Find the sacrifice and improvement rows. Values in the sacrifice row indicate what will be lost in per-unit profit by making a change in the resource allocation mix. Values in the improvement row indicate what will be gained in per-unit profit by making a change.

Step 3. Apply the entry criteria. The entering variable is defined as the current non-basic variable that will most improve the objective if its value is increased from 0.

Step 4. Apply the exit criteria. Using the current tableau's exchange coefficient from the entering variable column, calculate the exchange ratio for each row. Find the lowest nonzero and nonnegative value. The basic variable in this row becomes the exiting variable.

Step 5. Construct a new simplex tableau. Replace the exiting variable in the basic mix column with the new entering variable. Change the unit profit or unit loss column with the value for the new entering variable. Compute the new row values to obtain a new set of exchange coefficients applicable to each basic variable.

Step 6. Repeat steps 2 through 5 until you no longer can improve the solution.

For more information

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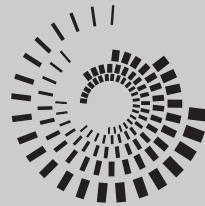
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