

The Transportation Problem: LP Formulations

An LP Formulation

Suppose a company has m warehouses and n retail outlets. A single product is to be shipped from the warehouses to the outlets. Each warehouse has a given level of supply, and each outlet has a given level of demand. We are also given the transportation costs between every pair of warehouse and outlet, and these costs are assumed to be linear. More explicitly, the assumptions are:

- The total supply of the product from warehouse i is a_i , where $i = 1, 2, \dots, m$.
- The total demand for the product at outlet j is b_j , where $j = 1, 2, \dots, n$.
- The cost of sending one unit of the product from warehouse i to outlet j is equal to c_{ij} , where $i = 1, 2, \dots, m$ and $j = 1, 2, \dots, n$. The total cost of a shipment is linear in the size of the shipment.

The problem of interest is to determine an optimal transportation scheme between the warehouses and the outlets, subject to the specified supply and demand constraints.

Graphically, a transportation problem is often visualized as a network with m source nodes, n sink nodes, and a set of $m \times n$ “directed arcs.” This is depicted in Figure TP-1.

We now proceed with a linear-programming formulation of this problem.

The Decision Variables

A transportation scheme is a complete specification of how many units of the product should be shipped from each warehouse to each outlet. Therefore, the decision variables are:

x_{ij} = the size of the shipment from warehouse i to outlet j , where $i = 1, 2, \dots, m$ and $j = 1, 2, \dots, n$.

This is a set of $m \times n$ variables.

The Objective Function

Consider the shipment from warehouse i to outlet j . For any i and any j , the transportation cost per unit is c_{ij} ; and the size of the shipment is x_{ij} . Since we assume that the cost function is linear, the total cost of this shipment is given by $c_{ij}x_{ij}$. Summing over all i and all j now

yields the overall transportation cost for all warehouse-outlet combinations. That is, our objective function is:

$$\text{Minimize } \sum_{i=1}^m \sum_{j=1}^n c_{ij} x_{ij}.$$

The Constraints

Consider warehouse i . The total outgoing shipment from this warehouse is the sum $x_{i1} + x_{i2} + \cdots + x_{in}$. In summation notation, this is written as $\sum_{j=1}^n x_{ij}$. Since the total supply from warehouse i is a_i , the total outgoing shipment cannot exceed a_i . That is, we must require

$$\sum_{j=1}^n x_{ij} \leq a_i, \quad \text{for } i = 1, 2, \dots, m.$$

Consider outlet j . The total incoming shipment at this outlet is the sum $x_{1j} + x_{2j} + \cdots + x_{mj}$. In summation notation, this is written as $\sum_{i=1}^m x_{ij}$. Since the demand at outlet j is b_j , the total incoming shipment should not be less than b_j . That is, we must require

$$\sum_{i=1}^m x_{ij} \geq b_j, \quad \text{for } j = 1, 2, \dots, n.$$

This results in a set of $m + n$ functional constraints. Of course, as physical shipments, the x_{ij} 's should be nonnegative.

LP Formulation

In summary, we have arrived at the following formulation:

$$\begin{aligned} &\text{Minimize } \sum_{i=1}^m \sum_{j=1}^n c_{ij} x_{ij} \\ &\text{Subject to:} \\ &\quad \sum_{j=1}^n x_{ij} \leq a_i \quad \text{for } i = 1, 2, \dots, m \\ &\quad \sum_{i=1}^m x_{ij} \geq b_j \quad \text{for } j = 1, 2, \dots, n \\ &\quad x_{ij} \geq 0 \quad \text{for } i = 1, 2, \dots, m \text{ and } j = 1, 2, \dots, n. \end{aligned}$$

This is a linear program with $m \times n$ decision variables, $m + n$ functional constraints, and $m \times n$ nonnegativity constraints.

A Numerical Example

In an actual instance of the transportation problem, we need to specify m and n , and replace the a_i 's, the b_j 's, and the c_{ij} 's with explicit numerical values.

As a simple example, suppose we are given: $m = 3$ and $n = 2$; $a_1 = 45$, $a_2 = 60$, and $a_3 = 35$; $b_1 = 50$ and $b_2 = 60$; and finally, $c_{11} = 3$, $c_{12} = 2$, $c_{21} = 1$, $c_{22} = 5$, $c_{31} = 5$, and $c_{32} = 4$. Then, substitution of these values into the above formulation leads to the following explicit problem:

$$\begin{array}{ll}
 \text{Minimize} & 3x_{11} + 2x_{12} + x_{21} + 5x_{22} + 5x_{31} + 4x_{32} \\
 \text{Subject to:} & \\
 & x_{11} + x_{12} \leq 45 \quad (1) \\
 & \quad \quad \quad x_{21} + x_{22} \leq 60 \quad (2) \\
 & \quad \quad \quad \quad \quad x_{31} + x_{32} \leq 35 \quad (3) \\
 & x_{11} \quad \quad \quad + x_{21} \quad \quad \quad + x_{31} \geq 50 \quad (4) \\
 & \quad \quad \quad x_{12} \quad \quad \quad + x_{22} \quad \quad \quad + x_{32} \geq 60 \quad (5) \\
 & x_{ij} \geq 0 \quad \text{for } i = 1, 2, 3 \text{ and } j = 1, 2.
 \end{array}$$

For an example of an explicit transportation scheme, let $x_{11} = 20$, $x_{12} = 20$, $x_{21} = 20$, $x_{22} = 20$, $x_{31} = 10$, and $x_{32} = 20$. It is easily seen that this proposed solution satisfies all of the constraints, and hence it is feasible. In words, the solution calls for shipping 20 units from warehouse 1 to outlet 1, 20 units from warehouse 1 to outlet 2, 20 units from warehouse 2 to outlet 1, ..., and finally 20 units from warehouse 3 to outlet 2. The total transportation cost associated with this transportation scheme can be computed as:

$$3 \times 20 + 2 \times 20 + 1 \times 20 + 5 \times 20 + 5 \times 10 + 4 \times 20 = 350.$$

An Equivalent Formulation

To derive an optimal transportation scheme for the above example, we can of course apply the standard Simplex algorithm. This means that we will first introduce 3 slack variables for constraints (1), (2), and (3) and 2 surplus variables for constraints (4) and (5), to convert these inequalities into equalities. On top of these, at the onset of Phase I, we also need to introduce 2 more artificial variables to serve as starting basic variables for constraints (4) and (5). We would then go through Phase I and Phase II to arrive at an optimal solution. This routine is standard fare, but the introduction of so many new variables seems tedious. Can we do better? It turns out that we can.

The first observation is that for a given problem to have any feasible solution, the total supply must not be less than the total demand. In this numerical example, we have a total

supply of 140 and a total demand of 110. Hence, feasible solutions exist; and indeed, we constructed one with ease. The second observation is that if the total supply happens to be equal to the total demand, then *any* feasible solution must satisfy all of the inequality constraints as equalities. (For example, this would be the case if a_1 , a_2 , and a_3 had been 40, 40, and 30, respectively.) As a consequence, whenever the given total supply and total demand are the same, we can replace all (functional) inequality constraints by equality constraints. Our third, and final, observation is that after such replacements, there is no longer any need for introducing slack or surplus variables.

Certainly, we cannot expect every problem to come with identical total supply and total demand. However, notice that if the total supply is strictly greater than the total demand, then one can artificially create a “dummy sink” to absorb the difference between the two. Of course, in order to preserve the original cost structure, the transportation cost for units sent to the dummy sink should be set to zero. Thus, for this specific example, we can introduce a third outlet to serve as the dummy sink; and let $b_3 = 30$ and $c_{13} = c_{23} = c_{33} = 0$. This yields the following new linear program:

$$\begin{array}{rllllllll}
 \text{Minimize} & 3x_{11} & +2x_{12} & & +x_{21} & +5x_{22} & & +5x_{31} & +4x_{32} & & \\
 \text{Subject to:} & & & & & & & & & & \\
 & x_{11} & +x_{12} & +x_{13} & & & & & & & = 45 \\
 & & & & x_{21} & +x_{22} & +x_{23} & & & & = 60 \\
 & & & & & & & x_{31} & +x_{32} & +x_{33} & = 35 \\
 & x_{11} & & & +x_{21} & & & +x_{31} & & & = 50 \\
 & & x_{12} & & & +x_{22} & & & +x_{32} & & = 60 \\
 & & & x_{13} & & & +x_{23} & & & +x_{33} & = 30 \\
 & x_{ij} \geq 0 & & & & & & & & & \text{for } i = 1, 2, 3 \text{ and } j = 1, 2, 3.
 \end{array}$$

It should be clear that by construction, this problem is equivalent to the original one.

With the above discussion, we can now assume without loss of generality that *every* transportation problem comes with identical total supply and total demand. This gives rise to what is called the *standard form* of the transportation problem. Formally, under the assumption that

$$\sum_{i=1}^m a_i = \sum_{j=1}^n b_j$$

holds, the standard form of the transportation problem is:

$$\begin{aligned} \text{Minimize} \quad & \sum_{i=1}^m \sum_{j=1}^n c_{ij} x_{ij} \\ \text{Subject to:} \quad & \sum_{j=1}^n x_{ij} = a_i \quad \text{for } i = 1, 2, \dots, m \\ & \sum_{i=1}^m x_{ij} = b_j \quad \text{for } j = 1, 2, \dots, n \\ & x_{ij} \geq 0 \quad \text{for } i = 1, 2, \dots, m \text{ and } j = 1, 2, \dots, n. \end{aligned}$$

Can we also do away with the need for introducing artificial variables? The answer again is a yes. However, we will delay this discussion a little bit.

The Transportation Tableau

The Simplex tableau serves as a very compact format for representing and manipulating linear programs. In the same spirit, we now introduce a tableau representation for transportation problems that are in the standard form.

For a problem with m sources and n sinks, the tableau will be a table with m rows and n columns. Specifically, each source will have a corresponding row; and each sink, a corresponding column. For ease of reference, we shall refer to the cell that is located at the intersection of the i th row and the j th column as ‘‘cell (i, j) ’’. Parameters of the problem will be entered into various parts of the table in the format below.

		Sink j				
		\vdots				
		\vdots				
		\vdots	c_{ij}			
Source i	\cdots	\cdots	x_{ij}	\cdots	\cdots	a_i
		\vdots				
		\vdots				
		\vdots				
			b_j			

That is, each row is labelled with its corresponding source name at the left margin; each column is labelled with its corresponding sink name at the top margin; the supply from source i is listed at the right margin of the i th row; the demand at sink j is listed at the bottom margin of the j th column; the transportation cost c_{ij} is listed in a subcell located

at the upper-left corner of cell (i, j) ; and finally, the value of x_{ij} is to be entered at the lower-right corner of cell (i, j) .

Again, we will use the numerical example above to illustrate the just-described format. With the introduction of a dummy sink to balance the total supply and total demand, this problem has three sources and three sinks. Therefore, the parameters of the problem can be specified in the explicit 3×3 tableau below.

		Sinks			
		1	2	3	
Sources	1	3	2	0	45
	2	1	5	0	60
	3	5	4	0	35
		50	60	30	

Notice that we have left the spaces for the x_{ij} 's blank; these will be filled in later, when we begin the solution process. As a specific example, the feasible solution $x_{11} = 20$, $x_{12} = 20$, $x_{13} = 5$, $x_{21} = 20$, $x_{22} = 20$, $x_{23} = 20$, $x_{31} = 10$, $x_{32} = 20$, and $x_{33} = 5$ would be entered as in the tableau below.

		Sinks			
		1	2	3	
Sources	1	3 20	2 20	0 5	45
	2	1 20	5 20	0 20	60
	3	5 10	4 20	0 5	35
		50	60	30	

All transportation problems in the standard form will henceforth be specified in this compact tableau format.

Discussion

In most applications, the shipments are integer-valued. We have ignored this requirement in our formulation. However, it will turn out that if the specified a_i 's and b_j 's are integers, then the optimal solution of a transportation problem produced by the Simplex algorithm also is integer-valued.

In applications where an oversupply exists, it is often desirable to assume that there is an inventory-holding cost for units that are not shipped out. If this is the case, then we can easily accommodate the inventory-holding costs as part of the objective function by reinterpreting them as transportation costs to the dummy sink.

When the total supply is *less* than the total demand, it is still possible to create a balanced transportation problem. The idea is to create a dummy source and let it have a supply that equals the difference between the total demand and the total supply. In doing so, it may be reasonable to assess a shortage penalty cost for units that are sent (fictitiously) from the dummy source to any sink.