# **Discounted Generalized Transportation Problem**

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Abstract- In generalized transportation problem(GTP), the cost of transportation  $c_{ii}$  per unit product from the  $i^{th}$  origin to the  $j^{th}$ destination is considered as independent of amount of transported commodity x<sub>ii</sub>. But in real life problems, there are many situations, e.g. quantity discount, price break etc., in which cost of transportation c<sub>ij</sub> depends upon the amount of transported commodity x<sub>ii</sub>. Based on these situations, in this paper, we consider a new type of discounted generalized transportation problem in which the cost of transportation c<sub>ii</sub> per unit product depends on the amount of transported commodity  $x_{ii}$ . Thereby, we develop a new algorithm for obtaining the optimum solution of this problem. Finally, a numerical example is illustrated to support the algorithm.

Index Terms- Generalized transportation problem, Step function, Discount, Discounted generalized transportation problem

## I. INTRODUCTION

Titchcock [15] was pioneer of the basic transportation **П**problem, Dantzig [13], Charnes and Cooper [11], Appa [1] developed further. Now a days, there are several procedures to solve the transportation problems. Arsham and Khan [2] considered simplex-type algorithm for general transportation problems. Basu et. al. [7,8,9] considered different types of transportation problems.

But in real life, there are many situations, e.g. quantity discount, price breaks etc. where the transportation cost may not be linear. Non linearity depends upon the character of the objective function as well as the character of the constraints. Cooper and Dredes [12] considered an approximate solution method for the fixed charge problem. Bhatia et. al. [10] considered time-cost trade-off in a transportation problem. Klingman and Russel [16] considered solving constrained transportation problems. Thirwani [17] considered fixed charge bi-criterion transportation problem with enhanced flow.

There are many business problems, industrial problems, machine assignment problems, routing problems, etc. that have the characteristics in common with generalized transportation problem that has been studied by several authors. Balas and Ivanescu [4] introduced on the generalized transportation problem. Balachandrana and Thompson [3] considered an operator theory of parametric programming for the generalized transportation Problem. In 1987, Hadley [14] gave the detailed solution procedure for solving generalized transportation problem. Basu and Acharya [5,6] considered different types of generalized transportation problem.

Day by day, the importance of discounted generalized transportation problem is increasing practically in a great deal, but the method for finding the optimum solution of this kind of generalized transportation problems, however, lacks of the desired attention.

There are several differences between classical transportation problem and generalized transportation problem which are given as follows:

1. The rank of the co-efficient matrix of  $[x_{ii}]_{mn}$  in generalized transportation problem is (m+n), where as in classical transportation problem it is (m+n-1)

2. In generalized transportation problem the value of  $x_{ii}$  may not be integer though it must be integer in classical transportation problem.

3. The activity vector in generalized transportation problem is  $p_{ij} = d_{ij} e_i + e_{m+j}$ , where  $d_{ij} = positive$  constants rather than unity.

Where as in classical transportation problem it is given by  $p_{ij} = e_i + e_{m+j}$ 

4. In generalized transportation problem, it need not be true that cells corresponding to basic solution form a tree.

5. In generalized transportation problem that  $\sum_{i=1}^{m} a_{i=1}$  $\sum_{j=1}^{n} b_j$  is not necessary.

In this paper we develop a new algorithm to find the solution of discounted generalized transportation problem where the cost function is taken as step function. Thereby we illustrate this problem numerically.

# **II. PROBLEM FORMULATION**

Let the discounted generalized transportation problem consists of 'm' origins and 'n' destinations, where

- $x_{ij}$  = the amount of product transported from the i<sup>th</sup> origin to the j<sup>th</sup> destination,  $c_{ij}$  = the cost involved in transporting per unit product from the i<sup>th</sup> origin to the j<sup>th</sup> destination,
- $a_i$  = the number of units available at the origin i,
- $b_i$  = the number of units required at the destination j,

 $d_{ii}$  = positive constants rather than unity.

Then the cost minimizing discounted generalized transportation problem can be stated as:

$$P_1: \qquad Min Z = \sum_{i=1}^m \sum_{j=1}^n c_{ij} x_{ij} \tag{1}$$

subject to,

$$\sum_{j=1}^{n} d_{ij} x_{ij} \leq a_i$$
; for i=1,2,3,....,m.

$$\sum_{i=1}^{m} x_{ij} = b_j, \quad \text{for } j = 1, 2, 3, \dots, n.$$
(3)

$$c_{ij} = \begin{cases} c_{ij}^{1} & if \ o \le x_{ij} \le x_{ij}^{1} \\ c_{ij}^{2} & if \ x_{ij}^{1} \le x_{ij} \le x_{ij}^{2} \\ c_{ij}^{2} & if \ x_{ij}^{2} \le x_{ij} \le x_{ij}^{3} \\ \dots \dots \dots \dots \dots \\ \dots \dots \dots \dots \dots \\ c_{ij}^{g} & if \ x_{ij}^{g-1} < x_{ij} \end{cases}$$
(4)

and

where  $c_{ij}^1 > c_{ij}^2 > c_{ij}^3 > \dots > c_{ij}^g$  and  $x_{ij} \ge 0$ .

## **Net Evaluation**

Introducing slack variables, the problem P1 can be written as

$$P_2: \qquad Min \ Z = \sum_{i=1}^{m} \sum_{j=1}^{n+1} c_{ij} x_{ij}$$
(5)

subject to,

$$a_{i} - \sum_{j=1}^{n} d_{ij} x_{ij} - x_{i,n+1} = 0; \quad \text{for } i = 1,2,3,\dots,m.$$
(6)  
$$b_{j} - \sum_{i=1}^{m} x_{ij} = 0; \quad \text{for } j = 1,2,3,\dots,n,n+1.$$
(7)

where  $d_{ij}$  positive constants rather than unity,  $x_{ij} \ge 0$  for all (i,j), and the values of  $c_{ij}$  are given in (4).

Let 
$$u_i (1 \le i \le m)$$
 and  $v_j (1 \le j \le n+1)$  be the dual variables.  
So that,  $d_{ij}u_i + v_j = c_{ij}$  for  $1 \le I \le m$ ,  $1 \le j \le n$ .  
 $u_i = 0$  for  $1 \le I \le m$ ,  $j = n + 1$ .  
(8)  
where  $u_i, v_j$  are unrestricted for all (i,j).

Now for any standard Primal L.P.P. with basis B and associated cost vector  $c_B$ , the associated solution to its dual problem  $W_B$ , is given by  $W_B = c_B B^{-1}$ . Thus, if  $p_j$  is the j<sup>th</sup> column of the primal constraint matrix, then an expression for evaluating the net-evaluation for minimization problem is given by

$$Z_{j} - c_{j} = c_{B} (B^{-1} p_{j}) - c_{j}$$

$$= W_{B} p_{j} - c_{j} \qquad \forall j.$$
(9)

But, in the present case of rectangular transportation problem, the dual solution can be represented by  $(u,v) = (u_1,u_2, \ldots, u_m,v_1,v_2, \ldots, v_n)$  and therefore the net evaluations are analogously obtained by simply replacing  $c_j \rightarrow c_{ij}$ ,  $W_B \rightarrow (u,v)$ ,  $p_j \rightarrow p_{ij}$  in the above formula. Thus we have the net evaluation as:

 $\begin{aligned} Z_{ij} - c_{ij} = &(u,v) \ p_{ij} - c_{ij} \\ = &(u_1, u_2, \dots, u_m, v_1, v_2, \dots, v_n) \ [ \ d_{ij}e_i + e_{m+j}] - c_{ij} \\ = &( \ d_{ij} \ u_i + v_j \ ) - c_{ij} \qquad \text{for} \quad i=1,2,\dots,m; \ j=1,2,\dots,n+1. \end{aligned}$ 

where  $p_{ij}(=d_{ij}e_i+e_{m+j})$  is the column vector of the constraint matrix associated with the rectangular variable  $x_{ij}$ . For simplicity, we shall denote the net evaluation  $Z_{ij} - c_{ij}$  by  $\Delta_{ij}$  in all our further discussion.

**Theorem:** A solution of the discounted generalized transportation problem will be feasible solution if for any cell (q,r),  $c_{qr}^k x_{qr} < c_{qr}^l$  ( $x_{qr} + y_{qr}$ ), where  $1 \le k < l \le g$  also  $y_{qr} > 0$  and  $c_{qr}^k > c_{qr}^l$  for  $1 \le q \le m$ ,  $1 \le r \le n$ . **Proof:** Let  $Z_1$  be the total cost and  $F_1$  be the total flow where (q,r) be one of its allocated cell and  $x_{qr}$  be its allocation where

**Proof:** Let  $Z_1$  be the total cost and  $F_1$  be the total flow where (q,r) be one of its allocated cell and  $x_{qr}$  be its allocation where  $(x_{qr}^{k-1} < x_{qr} < x_{qr}^k)$ .

$$Then \qquad Z_1 = \sum_{i=1, i \neq q}^m \sum_{j=1, j \neq r}^n c_{ij} x_{ij} + c_{qr}^k x_{qr} \qquad (11)$$

$$R_1 = \sum_{i=1, i \neq q}^m \sum_{j=1, j \neq r}^n x_{ij} + x_{qr} \qquad \text{where} \qquad x_{qr}^{k-1} < x_{qr} \le x_{qr}^k \qquad (12)$$

(2)

Also let  $Z_2$  be the total cost and  $F_2$  be the total flow if  $(x_{qr} + y_{qr})$  where  $x_{qr}^{l-1} < x_{qr} + y_{qr} \le x_{qr}^{l}$ , be allocated at the (q,r) cell and all other allocations are same as in flow  $F_1$ . Therefore

$$Z_{2} = \sum_{i=1, i \neq q}^{m} \sum_{j=1, j \neq r}^{n} c_{ij} x_{ij} + c_{qr}^{l} (x_{qr} + y_{qr})$$
(13)  
$$F_{2} = \sum_{i=1, i \neq q}^{m} \sum_{j=1, j \neq r}^{n} x_{ij} + (x_{qr} + y_{qr}) \quad \text{where} \quad x_{qr}^{l-1} < (x_{qr} + y_{qr}) \leq x_{qr}^{l}$$
(14)

Obviously  $F_1 + y_{qr} > F_1$  where  $x_{qr}^{l-1} < (x_{qr} + y_{qr}) \le x_{qr}^{l}$ 

Let us assume that the given condition  $c_{qr}^k x_{qr} > c_{qr}^l (x_{qr}+y_{qr})$  does not hold. Then  $Z_1 - Z_2 = c_{qr}^l x_{qr} - c_{qr} (x_{qr} + y_{qr}) \ge 0; \qquad => Z_1 > Z_2$ 

But  $F_2 > F_1$ ; So  $Z_1 > Z_2 \implies F_1 < F_2$ , which is a contradiction. So, our assumption is wrong. Hence the theorem.

#### III. ALGORITHM

Step 1. Find the initial solution  $X_{ij}^{E1}$  with the associated cost vector  $c_{B}^{1}$  and the corresponding cost  $Z^{1}$  by using North West Corner Rule.

Step 2. Set r=1.

and

Step 3. Write  $X_{ij}^{Br}$  with the associated vector  $c_{B}^{r}$  and corresponding cost Z<sup>r</sup>.

Step 4. Calculate dual variables  $u_i (1 \le i \le m)$  and  $v_j (1 \le j \le n)$ .

Step 5. Calculate net evaluation  $\Delta_{ij} \forall (i,j) \notin B$ . If  $\Delta_{ij} \leq 0 \forall (i,j) \notin B$ . Then go to step 10.

Step 6. Calculate  $\Delta_{st} = Max \{\Delta_{ij} : \Delta_{ij} > 0\}$ . Then (s,t) cell enters into the basis.

Step 7. Let  $\sum_{j \in B} d_{ij} y_{st}^{ij} = d_{st} e_{s}; \qquad 1 \le i \le m.$  $\sum_{i \in B} y_{st}^{ij} = e_{m+t}; \qquad 1 \le j \le n+1.$ where  $p_{st}=d_{st} e_s + e_{m+t}$ . After determining  $y_{st}^{ij}$  we calculate  $\theta = \min\left\{\frac{y_{tj}^{ij}}{y_{st}^{ij}}: y_{st}^{ij} > 0\right\} = \frac{x_{kl}^{ij}}{y_{st}^{kl}}$ 

Then (k,l) cell leaves the basis.

Step 8. The improved solution is  $\widehat{x_{ij}^{B}} = x_{ij}^{B} - y_{st}^{ij} \times \theta$  for  $(i,j) \in B$  and  $(i,j) \neq (s,t)$ =  $\theta$  for (i,j) = (s,t)

Step 9. Set r=r+1, goto step 3.

Step 10. Write the optimum solution is  $X^* = X_{ij}^{Br}$  with the associated cost vector  $c_B^* = c_B^r$  and the minimum cost is  $Z^* = Z^r$ . Step 11. Stop.

## IV. NUMERICAL EXAMPLE

We consider the following problem given in Table 1

Table 1								
	$D_1$	D <sub>2</sub>	D <sub>3</sub>	D <sub>4</sub>	SI	a <sub>i</sub>		
					ас			
					k			
<b>O</b> <sub>1</sub>	C <sub>11</sub>	C <sub>12</sub>	C <sub>13</sub>	C <sub>14</sub>	0	200		
	d <sub>11</sub> =	d <sub>12</sub> =	d <sub>13</sub> =0.	d <sub>14</sub> =0.5	1			
	0.35	0.5	35					
0 <sub>2</sub>	C <sub>21</sub>	C <sub>22</sub>	C <sub>23</sub>	C <sub>24</sub>	0	500		
	d <sub>21</sub> =	d <sub>22</sub> =	d <sub>23</sub> =0.	d <sub>24</sub> =0.4	1			
	0.9	0.84	3					
O <sub>3</sub>	C <sub>31</sub>	C <sub>32</sub>	C <sub>33</sub>	C <sub>34</sub>	0	400		
	d <sub>31</sub> =	d <sub>22</sub> =	d <sub>23</sub> =0.	d <sub>24</sub> =0.9	1			
	0.8	0.4	74					
b <sub>i</sub>	200	400	500	1000				
						1		

The variable costs are given below:	
$c_{11} = 203$ if $0 \le x_{11} \le 100$	$c_{12} = 401$ if $0 \le x_{12} \le 100$
$= 201 \text{ if } 100 < x_{11} \le 150$	$= 400$ if $100 < x_{12} \le 300$
$= 200$ if $x_{11} > 150$	$= 399$ if $x_{12} > 300$
$c_{13} = 400  \text{ if }  0 \le x_{13} \le \ 150$	$c_{14} = 751$ if $0 \le x_{14} \le 400$
$= 399 \text{ if } 150 < x_{13} \le 350$	$= 750$ if $400 < x_{14} \le 700$
$= 398$ if $x_{13} > 350$	$= 749$ if $x_{14} > 700$
$c_{21} = 502  \  \  if  0 \leq x_{21} \leq \ 100$	$c_{22} = 604  if  0 \le x_{22} \le 100$
$= 500 \text{ if } 100 < x_{21} \le 150$	$= 600$ if $100 < x_{22} \le 300$
$=498$ if $x_{21}>150$	$= 599$ if $x_{22} > 300$
$c_{23} = 602$ if $0 \le x_{23} \le 200$	$c_{24} = 752$ if $0 \le x_{24} \le 350$
$= 600 \text{ if } 200 < x_{23} \le 400$	$= 750$ if $350 < x_{24} \le 600$
$= 599$ if $x_{23} > 400$	$= 749$ if $x_{24} > 600$
$c_{31} = 401  \text{ if } \ 0 \le x_{31} \le \ 100$	$c_{32} = 502  \text{if}  0 \le x_{32} \le 150$
$= 400 \text{ if } 100 < x_{31} \le 150$	$= 500  if \ 150 < x_{32} \le 300$
$= 398$ if $x_{31} > 150$	$=499$ if $x_{32}>300$
$c_{33} = 602  if  0 \le x_{33} \le \ 200$	$c_{34} = 901  if  0 \le x_{34} \le 500$
$= 600 \text{ if } 200 < x_{33} \le 350$	$= 900$ if $500 < x_{34} \le 750$
$= 599$ if $x_{33} > 350$	$= 899$ if $x_{34} > 750$

Applying step 1, we get the following result in Table 2

Table – 2							
	D <sub>1</sub>	D <sub>2</sub>	D <sub>3</sub>	D <sub>4</sub>	Slack	a <sub>i</sub>	
01	200 200 0.35	400 260 0.5	400 0.35	751 0.5	0	200	
02	502 0.9	600 140 0.84	599 500 0.3	750 581 0.4	0	500	
O <sub>3</sub>	398 0.8	502 0.4	599 0.74	901 419 0.9	0 22.9 1	400	
bj	200	400	500	1000			

Applying step 2, set r=1.

Applying step 3, we get  $X_{ij}^{B_1} = \{x_{11} = 200, x_{12} = 260, x_{22} = 140, x_{23} = 500, x_{24} = 581, x_{34} = 419, x_{35} = 22.9\}$  with the associated cost vector  $C_{B}^{1} = \{c_{11} = 200, c_{12} = 400, c_{22} = 600, c_{23} = 599, c_{24} = 750, c_{34} = 901, c_{35} = 0\}$  and the corresponding cost  $Z^{1} = 1340769$ . Applying step 4, we get dual variables are  $u_{1} = -1034.2, u_{2} = -377.5, u_{3} = 0, v_{1} = 561.97, v_{2} = 917.1, v_{3} = 712.25$  and  $v_{4} = 901$ .

Applying step 5, we get the values of  $\Delta_{ij} \forall (i,j) \notin B$  are  $\Delta_{13} = -49.72$ ,  $\Delta_{14} = -367.1$ ,  $\Delta_{1S} = -1034.2$ ,  $\Delta_{21} = -275.78$ ,  $\Delta_{2S} = -377.5$ ,  $\Delta_{31} = 163.97$ ,  $\Delta_{32} = 415.1$  and  $\Delta_{33} = 113.25$ .

Applying step 6, we get  $\Delta_{st} = Max \{163.97, 415.1, 113.25\} = 415.1 \text{ at } (3,2) \text{ cell. Therefore } (3,2) \text{ cell enter into the basis.}$ Applying step 7, we get the values of  $\mathcal{Y}_{22}^{ij}$  are  $\mathcal{Y}_{32}^{11} = 0$ ,  $\mathcal{Y}_{32}^{22} = 0$ ,  $\mathcal{Y}_{32}^{22} = 0$ ,  $\mathcal{Y}_{32}^{22} = -2.1$ ,  $\mathcal{Y}_{32}^{24} = -2.1$ ,  $\mathcal{Y}_{32}^{24} = -1.49$ .

:. 
$$\theta = \min \{ 140, 199.5 \} = 140 = \overline{y_{32}^{22}}$$
. Therefore (2,2) cell leaves the basis.

Applying step 8, we get the values of  $\vec{x_{1j}}$  are  $\vec{x_{11}} = 200$ ,  $\vec{x_{12}} = 260$ ,  $\vec{x_{23}} = 500$ ,  $\vec{x_{24}} = 875$ ,  $\vec{x_{32}} = 140$ ,  $\vec{x_{34}} = 125$ ,  $\vec{x_{35}} = 231.5$  and tabulated in Table 3.

	$D_1$	D <sub>2</sub>	D <sub>3</sub>	D <sub>4</sub>	Slack	a <sub>i</sub>
<b>O</b> <sub>1</sub>	200	400	399	751	0	
	200	260				200
	0.35	0.5	0.35	0.5	1	
O <sub>2</sub>	502	600	599	750	0	
			500	875		500
	0.9	0.84			1	

Table – 3

[				0.3	0.4		
ſ	O <sub>3</sub>	398	502	602	901	0	
			140		125	231.5	400
		0.8	0.4	0.74	0.9	1	
l	b <sub>j</sub>	200	400	500	1000		

Applying step 9, set r = r + 1 and go to Step 3. The values of  $X_{ij}^{B^2}$  with the associated cost vector  $C_{B}^{2}$  give in Table 3 and the corresponding cost  $Z^2=1281780$ . Proceeding in this way the final table given by Table 4.

Table - 4								
	$D_1$	D <sub>2</sub>	D <sub>3</sub>	D <sub>4</sub>	Slack	a <sub>i</sub>		
<b>O</b> <sub>1</sub>	203	401	398	751	0			
	71.5		500			200		
	0.35	0.5	0.35	0.5	1			
02	500	600	602	749	0			
				1000	100	500		
	0.9	0.84	0.3	0.4	1			
O <sub>3</sub>	400	499	602	901	0			
	128.5	400			137.2	400		
	0.8	0.4	0.74	0.9	1			
b <sub>i</sub>	200	400	500	1000				

Applying step 10, the optimum solution is  $X^* = \{ x_{11} = 71.5, x_{13} = 500, x_{24} = 1000, x_{31} = 128.5, x_{32} = 400 \}$ , with the associated cost vector  $C_{a}^* = \{c_{11} = 203, c_{13} = 398, c_{24} = 749, c_{31} = 400, c_{32} = 499 \}$  and the optimum cost  $Z^* = 1213514.5$ 

## **REMARK:**

If we consider the problem

 $P^{1}: \quad Min \ Z = \sum_{i=1}^{m} \sum_{j=1}^{n} c_{ij} x_{ij}$ subject to,  $\sum_{i=1}^{m} d_{ij}^{2} x_{ij} = b_{j}; \qquad \text{for } i = 1,2,3,\dots,m.$ where the values of  $c_{ii}$  are given in equation (2.4), and  $x_{ii} \ge 0$  ( $1 \le i \le m, 1 \le j \le n$ ).

Then this problem (P<sup>1</sup>) can easily be converted to our proposed problem (P<sub>1</sub>) by transformation  $d_{ij}^2 x_{ij} = w_{ij}$  for  $(1 \le i \le m, 1 \le j \le n)$ ;

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