

Near Optimal Solution for the Step Fixed Charge Transportation Problem

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Received: 7 Oct. 2012, Revised: 22 Dec. 2012, Accepted: 25 Dec. 2012

Published online: 1 Jun. 2013

Abstract: Step Fixed-charge transportation problem (SFCTP) is considered to be one of the versions of Fixed-charge transportation problem (FCTP) where the fixed cost is incurred for every route that is used in the solution. This is considered to be an NP-hard problem since the cost structure causes the value of the objective function to behave like a step function. In this paper three formulae are proposed to construct intermediate coefficient matrix as a base for finding an initial solution for SFCTP. The proposed formulae overcome the drawbacks of one of the earlier proposed formulae, which fails to address the cases when load units become equal to or greater than the minimum of the supplies and demand for particular route. In addition, the achieved initial solution for the SFCTP is considered to be the best as compared to the initial solution achieved by the earlier proposed formulae in the literature. In order to confirm the superiority of the proposed formulae, forty problems with different sizes have been solved to evaluate and demonstrate the performance of the proposed formulae and to compare their performance with the earlier proposed formulae.

Keywords: Fixed charge transportation, Step fixed charge transportation, Heuristic algorithm.

1 Introduction

One of the versions of fixed charge transportation problem (FCTP) is the Step Fixed-charge transportation problem (SFCTP) where the fixed cost is incurred for every route that is used in the solution. In SFCTP, the fixed cost is proportional to the amount shipped. This cost structure causes the value of the objective function to behave like a step function. Unfortunately, not much research has been carried out in this area.

The FCTP is considered to be an NP-hard problem. After the fixed-charge problem was first formulated by Hirsch & Dantzig in 1954 [1]. During 1961 Balinski [2] showed that fixed-charge transportation problem is a special case of fixed-charge problem and an approximate solution was presented. Since then, considerable research has been carried out on this topic. In 1988 Sandrock [3] analyzed the source induced fixed-charge transportation problem. FCTP is generally formulated and solved as a mixed integer network programming problem. Theoretically, the FCTP can be solved by using any mixed integer programming solving technique. However,

these methods are not employed because of their inefficient and expensive computation.

Most of the solution methods of FCTP can be considered as either exact or heuristic. Exact methods include the cutting planes method [4], the vertex ranking method [5], and the branch-and-bound method [6] amongst others. These methods are generally not very useful when a problem reaches a certain level, because they do not make the most use of the special network structure of the FCTP. Therefore, heuristic methods have been proposed, such as the adjacent extreme point search method [2, 7], the Lagrangian relaxation method [8, 9] and such other heuristic methods [10, 11, 12, 13]. Although these methods are usually computationally efficient, the major disadvantage of heuristic methods is the possibility of terminating at a local optimum that is far distant from the global optimum.

Heuristic techniques for solving FCTP have been proposed by Balinski [2]. These techniques start with constructing a coefficient matrix and finding the optimal solution based on it. After that Kowalski & Lev [14] considered two more formulae in addition to Balinski's

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and suggested a heuristic approach for improving the derived local optimal solution found based on the coefficient matrices which were arrived at using those formulae.

In order to improve the solution quality of the SFCTP, this paper critically analyses the erstwhile proposed heuristic formulae by Balinski [2] and Kowalski & Lev [14] for constructing the coefficient matrix as a base for finding a good initial solution for SFCTPs. Further, three superior formulae have been proposed, which will also overcome the drawbacks of one of the earlier proposed formulae. In addition to that, forty problems with different dimensions have been solved to evaluate and demonstrate the performance of the proposed formulae and to compare their performance with the earlier proposed ones.

The rest of the paper is organized as follows: in section 2, SFCTP is described and its mathematical model is presented. The proposed formulae are described in section 3, followed by two illustrative examples in section 4. In section 5 the parametric analysis is carried out. Finally, the section 6 presents the conclusion and scope for future work.

2 Description and modeling of SFCTP

In this section the description and mathematical model of FCTP together with the modifications required to formulate SFCTP are presented.

The FCTP can be described as a distribution problem in which there are m suppliers and n customers. The suppliers denote warehouses, plants or factories while customers denote destinations or any demand points. Each of the m suppliers can ship to any of the n customers at a shipping cost per unit c_{ij} (unit cost for shipping from supplier i to customer j) plus a fixed cost f_{ij} , assumed for opening this route. Each supplier $i = 1, 2, \dots, m$ has s_i units of supply and each customer $j = 1, 2, \dots, n$ demands d_j units. x_{ij} is the unknown quantity to be transported on the route (i, j) from plant i to customer j . The objective is to determine which routes are to be opened and the size of the shipment, so that the total cost of meeting demand, given the supply constraints, is minimized. The mathematical model of FCTP can be represented as in (1) to (4).

$$\text{Min } z = \sum_{i=1}^m \sum_{j=1}^n (c_{ij}x_{ij} + f_{ij}y_{ij}) \quad (1)$$

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

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

$$\begin{aligned} x_{ij} &\geq 0 \quad \forall i, j \\ y_{ij} &= 0 \quad \text{if } x_{ij} = 0 \\ y_{ij} &= 1 \quad \text{if } x_{ij} > 0 \end{aligned} \quad (4)$$

In the SFCTP the fixed cost f_{ij} for route (i, j) is proportional to the transported amount through this route. So, an additional cost is added when the transported units exceeds a certain amount A_{ij} . The fixed cost f_{ij} can be divided into two parts. The first part is $f_{ij,1}$ which is the fixed cost to open a route (i, j) as long as the transported quantity x_{ij} is less than or equal to a certain amount A_{ij} . The second part $f_{ij,2}$ which is the additional fixed cost applied when the transported quantity x_{ij} exceeds this amount A_{ij} . Therefore the fixed cost f_{ij} can be calculated by (5).

$$f_{ij} = y_{ij,1}f_{ij,1} + y_{ij,2}f_{ij,2} \quad (5)$$

where

$$\begin{aligned} y_{ij,1} &= \begin{cases} 1 & \text{if } x_{ij} > 0 \\ 0 & \text{otherwise} \end{cases} \\ y_{ij,2} &= \begin{cases} 1 & \text{if } x_{ij} > A_{ij} \\ 0 & \text{otherwise} \end{cases} \end{aligned}$$

Incorporating (7) in the FCTP mathematical model, the standard mathematical model of the SFCTP can be represented as follows:

$$\text{Min } z = \sum_{i=1}^m \sum_{j=1}^n (c_{ij}x_{ij} + y_{ij,1}f_{ij,1} + y_{ij,2}f_{ij,2}) \quad (6)$$

$$\text{s.t } \sum_{i=1}^m x_{ij} = d_j \quad \text{for } j = 1, \dots, n \quad (7)$$

$$\sum_{j=1}^n x_{ij} = s_i \quad \text{for } i = 1, \dots, m \quad (8)$$

$$\begin{aligned} x_{ij} &\geq 0 \quad \forall i, j \\ y_{ij} &= 0 \quad \text{if } x_{ij} = 0 \\ y_{ij} &= 1 \quad \text{if } x_{ij} > 0 \\ y_{ij,1} &= \begin{cases} 1 & \text{if } x_{ij} > 0 \\ 0 & \text{otherwise} \end{cases} \\ y_{ij,2} &= \begin{cases} 1 & \text{if } x_{ij} > A_{ij} \\ 0 & \text{otherwise} \end{cases} \end{aligned}$$

$$\text{and } \sum_{i=1}^m s_i = \sum_{j=1}^n d_j$$

It can be observed that, if the shipment values A_{ij} is greater than or equal to the $Min(s_i, d_j) \forall i, j$ the optimal solution of SFCTP is an optimal solution of FCTP. i.e. If the $A_{ij} \geq Min(s_i, d_j) \forall i, j$, the optimal solution of SFCTP is an optimal solution of FCTP. This solution will be the lower bound of all solutions of SFCTP.

3 The Proposed Formula

Balinski [2] has provided a heuristic solution for FCTP. Assuming the fixed cost as $f_{ij,1}$ the Balinski matrix is obtained by formulating a linear version of FCTP by relaxing the integer restriction on y_{ij} in (1) as follows, where $M_{ij} = Min(s_i, d_j)$:

$$y_{ij} = x_{ij} / M_{ij}$$

The linear version of FCTP will have the unit transportation cost of shipping through the route (i, j) as follows:

$$C_{ij} = f_{ij,1} / M_{ij} + c_{ij} \tag{9}$$

Since there is no algorithm for SFCTP, any heuristic method which provides a good solution is considered useful. In this direction Kowalski & Lev [14] have put in efforts to propose two heuristic algorithms. In both the algorithms, the objective was to get a "good initial solution" and using this perturbing each load using single stepping-stone moves. In the first algorithm, the integer restriction considered in (9) by Balinski [2] has been replaced by C_{ij} as represented in (10).

$$C_{ij} = (f_{ij,1} + f_{ij,2}) / M_{ij} + c_{ij} \tag{10}$$

In the second formula, the integer restriction considered in (9) by Balinski has been replaced by C_{ij} as represented in (11).

$$C_{ij} = f_{ij,2} / (M_{ij} - A_{ij}) + c_{ij} \tag{11}$$

A critical look at (11) reveals that the formulation fails to consider the cases when $A_{ij} = M_{ij}$ and $A_{ij} > M_{ij}$ as the values will be infinity when $A_{ij} = M_{ij}$ and assumes negative value in case $A_{ij} > M_{ij}$.

As illustrated in Fig. 1, [14] in the case of FCTP, for every loaded route (i, j) the cost of the fixed-charge step function formulation is greater than the corresponding cost of the relaxed integer restriction function. The situation in case of SFCTP is illustrated in Fig. 2 [14]. The total cost (TC_{ij}) for shipping M_{ij} units through route (i, j) can be calculated as represented in (12).

$$TC_{ij} = f_{ij,1} + c_{ij}A_{ij,1} + f_{ij,2} + c_{ij}(M_{ij} - A_{ij}) \tag{12}$$

Further, (12) can be represented as in (13)

$$TC_{ij} = f_{ij,1} + f_{ij,2} + c_{ij}M_{ij} \tag{13}$$

The cost per unit (C_{ij}) can be calculated by dividing (13) by M_{ij} and can be represented as in (14).

$$C_{ij} = (f_{ij,1} + f_{ij,2}) / M_{ij} + c_{ij} \tag{14}$$

There are two cases of the shipped quantities which are $A_{ij} \geq M_{ij}$ and $A_{ij} < M_{ij}$. Therefore (14) can be represented as in (15).

$$C_{ij} = \begin{cases} f_{ij,1} / M_{ij} + c_{ij} & \text{if } A_{ij} \geq M_{ij} \\ (f_{ij,1} + f_{ij,2}) / M_{ij} + c_{ij} & \text{if } A_{ij} < M_{ij} \end{cases} \quad \forall i, j \tag{15}$$

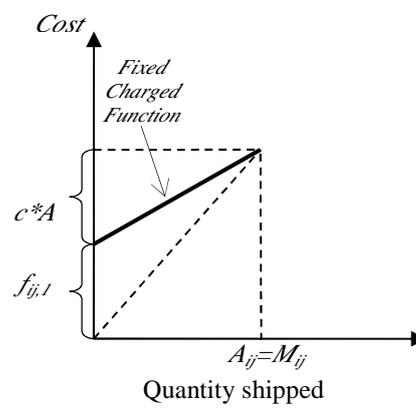


Fig. 1. Shipping costs as function of quantity shipped along route (i, j) for FCTP

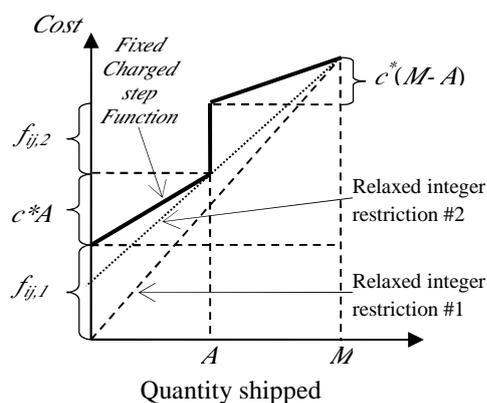


Fig. 2. Shipping costs as function of quantity shipped along route (i, j) for SFCTP

Alternately, by considering that only $(M_{ij} - A_{ij})$ units will be shipped through route (i, j) , (14) can be represented as (16).

$$C_{ij} = \begin{cases} f_{ij,1} / M_{ij} + c_{ij} & \text{if } A_{ij} \geq M_{ij} \\ f_{ij,2} / (M_{ij} - A_{ij}) + c_{ij} & \text{if } A_{ij} < M_{ij} \end{cases} \quad \forall i, j \tag{16}$$

In the situation when only $(M_{ij}-A_{ij})$ units will be shipped through route (i, j) , and another $(M_{ij}-A_{ij})$ units shipped through another route, and if we consider that the cost of shipping all units as the cost through the route (i, j) ; the unit cost of this route can be represented as (17).

$$C_{ij} = \begin{cases} f_{ij,1}/M_{ij} + c_{ij} & \text{if } A_{ij} \geq M_{ij} \\ f_{ij,2}/A_{ij} + f_{ij,1}/(M_{ij} - A_{ij}) + c_{ij} & \text{if } A_{ij} < M_{ij} \end{cases} \quad \forall i, j$$

This paper aims to propose three formulae as in (15), (16), and (17) for constructing intermediate coefficient matrix as a base for finding a local solution for SFCTP. Further a comparison of the performances and quality of these proposed formulae is undertaken with the earlier proposed formula (9) proposed by Balinski [2] and also with the two formulae (10) and (11) proposed by Kowalski & Lev [14]. However, it has been pointed out that (11) fails to consider the cases when $A_{ij} = M_{ij}$ and $A_{ij} > M_{ij}$ and hence the performance and quality comparison is restricted against (10) only.

4 Illustrative Examples

This section represents two illustrative examples used to compare the proposed three formulae (15), (16), and (17) with the previously proposed two (9) and (10).

In the first illustration, a 4x5 step transportation problem has been considered with parameters, viz., supplies s_i , demand d_j , variable costs $f_{ij,1}$, fixed costs $f_{ij,2}$, and step values A_{ij} as in Table 1.

The coefficient matrix generated using the formula (9) is shown in Table 2. The corresponding solution using QM for Windows Version 2.1 is presented in Table 3. The total fixed costs $f_{ij,1}$ and $f_{ij,2}$ are 150 and 200 respectively, the total variable cost $\sum_{i=1}^m \sum_{j=1}^n c_{ij}x_{ij}$ is 580, and the total cost is 930.

Table 1: The parameters and variables of example 4x5

	1	2	3	4	5	1	2	3	4	5
d_j	40	20	70	10	60					
s_i	Variable cost c_{ij}					Fixed cost $f_{ij,1}$				
10	5	3	2	4	6	40	20	30	20	10
100	3	5	3	4	3	10	20	20	30	20
20	3	4	6	5	2	40	30	10	20	30
70	2	5	4	3	4	10	40	40	10	10
	Fixed cost $f_{ij,2}$					Step value A_{ij}				
	50	70	80	70	80	20	20	20	20	20
	60	70	60	80	60	20	20	20	20	20
	60	80	80	70	70	20	20	20	20	20
	80	40	50	50	50	20	20	20	20	20

Table 2: The coefficient matrix using formula (9)

	d_1	d_2	d_3	d_4	d_5
s_1	9.0	5.0	5.0	6.0	7.0
s_2	3.3	6.0	3.3	7.0	3.3
s_3	5.0	5.5	6.5	7.0	3.5
s_4	2.3	7.0	4.6	4.0	4.2

Table 3: Optimal distribution for formula (9)

	d_1	d_2	d_3	d_4	d_5
s_1		10			
s_2			70		30
s_3		10			10
s_4	40			10	20

Similarly, the coefficient matrix generated using the formula (10) is shown in Table 4. The corresponding solution is presented in Table 5. The total fixed costs $f_{ij,1}$ and $f_{ij,2}$ are 110 and 250 respectively, the total variable cost is 620, and the total cost is 980.

Table 4: The coefficient matrix using formula (9)

	d_1	d_2	d_3	d_4	d_5
s_1	14.0	12.0	13.0	13.0	15.0
s_2	4.8	9.5	4.1	15.0	4.3
s_3	8.0	9.5	10.5	14.0	7.0
s_4	4.3	9.0	5.3	9.0	5.0

The coefficient matrix generated using the formulae (15) and (16) are shown in Tables 6 and 7 respectively. It is observed that using (15) and (16), we obtain the same local optimal solution, as presented in Table 8. The total fixed costs $f_{ij,1}$ and $f_{ij,2}$ are 140 and 140 respectively, the total variable cost is 580, and the total cost is 860.

Table 5: Optimal distribution for formula (10)

	d_1	d_2	d_3	d_4	d_5
s_1		0		10	
s_2			70		30
s_3		20			0
s_4	40				30

Table 6: The coefficient matrix using formula (15)

	d_1	d_2	d_3	d_4	d_5
s_1	9.0	5.0	5.0	6.0	7.0
s_2	4.8	6.0	4.1	7.0	4.3
s_3	5.0	5.5	6.5	7.0	3.5
s_4	4.3	7.0	5.3	4.0	5.0

Table 7: The coefficient matrix using formula (16)

	d_1	d_2	d_3	d_4	d_5
s_1	9.0	5.0	5.0	6.0	7.0
s_2	6.0	6.0	4.2	7.0	4.5
s_3	5.0	5.5	6.5	7.0	3.5
s_4	6.0	7.0	5.0	4.0	5.3

Table 8: Optimal distribution for formula (15) and (16)

	d_1	d_2	d_3	d_4	d_5
s_1		10			
s_2		10	70		20
s_3					20
s_4	40			10	20

The coefficient matrix generated using the formula (17) is shown in Table 9. The corresponding solution is presented in Table 10. The total fixed costs $f_{ij,1}$ and $f_{ij,2}$ are 150 and 140 respectively, the total variable cost is 590, and the total cost is 880.

Table 9: The coefficient matrix using formula (17)

	d_1	d_2	d_3	d_4	d_5
s_1	9.0	5.0	5.0	6.0	7.0
s_2	6.5	6.0	6.4	7.0	6.5
s_3	5.0	5.5	6.5	7.0	3.5
s_4	6.5	7.0	7.3	4.0	6.8

Table 10: Optimal distribution for formula (17)

	d_1	d_2	d_3	d_4	d_5
s_1			10		
s_2		20	60		20
s_3					20
s_4	40			10	20

The comparative statement of the total costs for the illustration using the different formulae is summarized in Table 11.

Table 11: Summary of total costs using different formulae.

Formula	$f_{ij,1}$	$f_{ij,2}$	$\sum_{i=1}^m \sum_{j=1}^n c_{ij}x_{ij}$	Total Cost
(10)	150	200	580	930
(11)	110	250	620	980
(14)	140	140	580	860
(15)	140	140	580	860
(16)	150	140	590	880

As summarized in Table 11, the results using the proposed three formulae 15, 16 and 17 are superior to the ones proposed earlier. Further, the quality of the results using the formulae 14 and 15 are superior to the results obtained using 16.

In the next illustration, a problem with a higher dimension, viz., 5x10 has been considered. Table 12 gives the parameters and variables used for the example of size 5x10.

Table 12: The parameters and variables of example 5x10

	1	2	3	4	5	6	7	8	9	10
d_j	40	20	50	10	10	20	30	30	50	40
s_i	Variable cost c_{ij}									
20	4	5	5	2	2	4	4	2	8	4
40	4	4	7	5	6	5	7	6	7	5
90	4	6	3	8	4	3	3	3	5	7
60	5	6	3	6	6	4	6	8	2	2
90	3	5	5	8	3	8	5	7	4	6
	Fixed cost $f_{ij,1}$									
	100	170	190	100	170	150	190	170	150	200
	110	170	170	200	180	160	180	180	170	140
	120	120	170	100	120	170	130	160	110	190
	130	120	130	180	160	140	170	180	190	110
	110	180	160	170	130	120	110	160	160	120
	A _{ij}									
	40	30	40	50	40	30	20	40	50	40
	10	50	30	40	30	50	20	30	20	10
	50	40	40	10	50	20	30	10	30	20
	40	10	30	20	20	40	50	20	20	30
	20	30	20	20	10	30	50	20	40	50
	Fixed cost $f_{ij,2}$									
	210	400	280	370	320	210	300	220	230	210
	290	340	340	280	360	330	200	390	310	400
	360	300	330	290	290	400	310	210	350	390
	390	220	220	250	330	290	370	310	350	280
	340	320	270	270	270	320	360	220	370	280

The coefficient matrix and the corresponding solution generated using the formula (9) is shown in Tables 13 and 14. The total fixed costs $f_{ij,1}$ and $f_{ij,2}$ are 1790 and 1200 respectively, the total variable cost is 960, and the total cost is 3950.

Table 13: The coefficient matrix using formula (9)

	d_1	d_2	d_3	d_4	d_5	d_6	d_7	d_8	d_9	d_{10}
s_1	9.0	13.5	14.5	12.0	19.0	11.5	13.5	10.5	15.5	14.0
s_2	6.8	12.5	11.3	25.0	24.0	13.0	13.0	12.0	11.3	8.5
s_3	7.0	12.0	6.4	18.0	16.0	11.5	7.3	8.3	7.2	11.8
s_4	8.3	12.0	5.6	24.0	22.0	11.0	11.7	14.0	5.8	4.8
s_5	5.8	14.0	8.2	25.0	16.0	14.0	8.7	12.3	7.2	9.0

Table 14: Optimal distribution for formula (9)

	d_1	d_2	d_3	d_4	d_5	d_6	d_7	d_8	d_9	d_{10}
s_1				10		10				
s_2	10	20				10				
s_3			30				30	30		
s_4			20						0	40
s_5	30				10					50

Tables 15 and 16 give the coefficient matrix and the corresponding solution generated using the formula (10). The total fixed costs $f_{ij,1}$ and $f_{ij,2}$ are 1500 and 1640 respectively, the total variable cost is 1140, and the total

cost is 4280.

Table 15: The coefficient matrix using formula (10)

	d_1	d_2	d_3	d_4	d_5	d_6	d_7	d_8	d_9	d_{10}
s_1	19.5	33.5	28.5	49.0	51.0	22.0	28.5	21.5	27.0	24.5
s_2	14.0	29.5	19.8	53.0	60.0	29.5	19.7	25.0	19.0	18.5
s_3	16.0	27.0	13.0	47.0	45.0	31.5	17.7	15.3	14.2	21.5
s_4	18.0	23.0	10.0	49.0	55.0	25.5	24.0	24.3	12.8	11.8
s_5	14.3	30.0	13.6	52.0	43.0	30.0	20.7	19.7	14.6	16.0

Tables 17 and 18 give the coefficient matrix and the corresponding solution generated using the formula (15). The total fixed costs $f_{ij,1}$ and $f_{ij,2}$ are 1770 and 220 respectively, the total variable cost is 1150, and the total cost is 3140.

Table 16: Optimal distribution for formula (10)

	d_1	d_2	d_3	d_4	d_5	d_6	d_7	d_8	d_9	d_{10}
s_1				0		20				
s_2	40						0			
s_3			20	10			30	30		
s_4		20	0							40
s_5			30		10				50	

Table 17: The coefficient matrix using formula (15)

	d_1	d_2	d_3	d_4	d_5	d_6	d_7	d_8	d_9	d_{10}
s_1	9.0	13.5	14.5	12.0	19.0	11.5	13.5	10.5	15.5	14.0
s_2	14.0	12.5	19.8	25.0	24.0	13.0	19.7	12.0	19.0	18.5
s_3	7.0	12.0	13.0	18.0	16.0	11.5	7.3	15.3	14.2	21.5
s_4	8.3	23.0	10.0	24.0	22.0	11.0	11.7	24.3	12.8	11.8
s_5	14.3	14.0	13.6	25.0	16.0	14.0	8.7	19.7	14.6	9.0

Table 18: Optimal distribution for formula (15)

	d_1	d_2	d_3	d_4	d_5	d_6	d_7	d_8	d_9	d_{10}
s_1				10		0		10		
s_2		20						20		
s_3	40					20	30			
s_4			50			0			10	
s_5				10					40	40

Corresponding Tables using formula (16) are Tables 19 and 20, resulting in the total fixed costs $f_{ij,1}$ and $f_{ij,2}$ as 1770 and 620 respectively, the total variable cost $\sum_{i=1}^m \sum_{j=1}^n c_{ij}x_{ij}$ is 1190, and the total cost is 3580.

Table 19: The coefficient matrix using formula (16)

	d_1	d_2	d_3	d_4	d_5	d_6	d_7	d_8	d_9	d_{10}
s_1	9.0	13.5	14.5	12.0	19.0	11.5	13.5	10.5	15.5	14.0
s_2	13.7	12.5	41.0	25.0	24.0	13.0	27.0	12.0	22.5	18.3
s_3	7.0	12.0	36.0	18.0	16.0	11.5	7.3	13.5	22.5	26.5
s_4	8.3	28.0	14.0	24.0	22.0	11.0	11.7	39.0	13.7	30.0
s_5	20.0	14.0	14.0	25.0	16.0	14.0	8.7	29.0	41.0	9.0

Table 20: Optimal distribution for formula (16)

	d_1	d_2	d_3	d_4	d_5	d_6	d_7	d_8	d_9	d_{10}
s_1				10				10		
s_2		20						20		
s_3	40	0			10	10	30			
s_4			0			10			50	
s_5			50							40

Corresponding Tables using formula (17) are Tables 21 and 22, resulting in the total fixed costs $f_{ij,1}$ and $f_{ij,2}$ as 1810 and 220 respectively, the total variable cost $\sum_{i=1}^m \sum_{j=1}^n c_{ij}x_{ij}$ is 1460, and the total cost is 3490.

Table 21: The coefficient matrix using formula (17)

	d_1	d_2	d_3	d_4	d_5	d_6	d_7	d_8	d_9	d_{10}
s_1	9.0	13.5	14.5	12.0	19.0	11.5	13.5	10.5	15.5	14.0
s_2	36.7	12.5	35.3	25.0	24.0	13.0	35.0	12.0	31.0	49.7
s_3	7.0	12.0	28.3	18.0	16.0	11.5	7.3	32.0	22.2	36.0
s_4	8.3	40.0	16.8	24.0	22.0	11.0	11.7	41.5	25.8	22.3
s_5	25.5	14.0	23.8	25.0	16.0	14.0	8.7	34.0	29.3	9.0

Table 22: Optimal distribution for formula (17)

	d_1	d_2	d_3	d_4	d_5	d_6	d_7	d_8	d_9	d_{10}
s_1									20	
s_2		10						30		
s_3	40	0		10		10			30	
s_4			50			10				
s_5		10			10		30			40

The comparative statement of the total costs for the illustration using the different formulae is summarized in Table 23.

Table 23: Summary of total costs using different formulae.

Formula	$f_{ij,1}$	$f_{ij,2}$	$\sum_{i=1}^m \sum_{j=1}^n c_{ij}x_{ij}$	Total Cost
(10)	1790	1200	960	3950
(11)	1500	1640	1140	4280
(14)	1770	220	1150	3140
(15)	1770	620	1190	3580
(16)	1810	220	1460	3490

As summarized in Table 23, the results using the proposed three formulae 15, 16 and 17 are superior to the ones proposed earlier. Further, the quality of the results using the formula 15 is superior to the results obtained using rest of the formulae.

From the above two illustrations it can be observed that there exist formulations of the intermediate coefficient matrix C_{ij} which yield superior coefficient matrix as a base for finding a local solution for SFCTPs as compared to the earlier proposed formulations. In order to further explore the effectiveness of the proposed formulae, the results based on different problems with

eight dimensions ranging from 3x3 to 20x20 and with different A_{ij} were analyzed. The values of A_{ij} were considered as both fixed as well as variable values for different problems. The details of analysis and the results are presented in the next section.

5 Parametric analysis

In this section different illustrative examples are considered to discover the best formulation of the function for determining the intermediate coefficient matrix, C_{ij} from among the earlier proposed two and the newly proposed three formulations. Random numbers were generated using Excel for determining the problem parameters and for generating the coefficient matrix corresponding to each formulation. The problem is solved using the Transportation Module using QM for Windows, Version 2.1. These solutions are taken to Excel sheet to find the corresponding fixed costs, variable costs and total costs for each problem. Because the scale of the functions in each problem will be different, they cannot be compared directly. Therefore, the Relative Percentage Deviation (RPD) is used for each combination in [15]. RPD is calculated by using (18).

$$RPD = \frac{Alg_{sol} - Min_{sol}}{Min_{sol}} \times 100 \quad (18)$$

where Alg_{sol} and Min_{sol} are the obtained TC_{ij} values for each replication of trial for a given dimension and the obtained best solution, respectively. After converting the objective values to $RPDs$, the mean RPD is calculated for each dimension. Problems with eight dimensions ranging from 3x3 to 15x20 were considered for illustrations. For each dimension, five problems with different values for each characteristic (both fixed and variable values) have been generated and used to calculate the average costs and RPD values for each dimension. Thus, a sample of 40 problems have been generated and solved. The characteristics of the test problems considered are presented in Table 24.

Table 24: Characteristics of SFCT test problems

Pro. size	Range of (d_j) , (s_i)	Rang of (c_{ij})		Rang of $(f_{ij,1})$		Rang of $(f_{ij,2})$		
		LL	UL	LL	UL	LL	UL	
3x3	50	100	1	3	10	20	20	50
4x5	150	250	1	9	10	40	30	70
5x10	200	500	1	9	10	50	30	90
10x10	300	500	1	9	100	200	200	400
10x15	500	1000	1	9	100	500	200	600
15x15	500	2000	1	9	100	500	200	600
15x20	1000	3000	1	9	100	500	200	700
20x20	1000	3000	1	9	100	500	200	700

All the 40 problems considered were solved to find the total cost of the associated SFCTP and subsequently the corresponding $RPDs$ for each of the earlier proposed two and the newly proposed three formulae. The values of average $RPDs$, based on five illustrative examples for each of the eight dimensions considered using the five formulae and the overall mean RPD for each of the formulae are presented in Table 25.

Based on the results presented in Table 25, the overall mean RPD of formula (15) is providing the least value as compared to the other formulae. This is followed by the other two proposed formulae, viz., (17) and (16) respectively. Hence, it can be concluded that the newly proposed three formulae are superior and can be used as a better alternative for constructing coefficient matrix as a base for finding a local solution for SFCTPs as compared to the earlier used formulae (9) and (10).

Table 25: The comparative results of the average RPD for the proposed formulae

Formulas	Average RPD of the test problems								Mean RPD
	3x3	4x5	5x10	10x10	10x15	15x15	15x20	20x20	
(10)	2.1	4.0	10.5	10.2	9.1	2.3	2.4	14.5	6.89
(11)	4.6	3.4	12.4	8.6	8.7	6.6	1.3	16.2	7.73
(15)	0.0	2.9	0.0	0.0	0.0	0.0	0.0	0.0	0.36
(16)	0.5	3.4	7.3	11.0	8.1	0.5	2.9	7.7	5.18
(17)	3.6	0.0	9.5	3.8	8.4	1.7	9.4	3.8	5.03

In addition to the above, in order to statistically test the significance of effectiveness of the results using different formulae, paired sample t-tests were used to determine the significant differences in the RPD values obtained using the five formulations, for each of the pairs. For the purpose of comparisons the RPD values obtained using all the 40 problems were used. The results of the tests are summarized in Table 26.

Table 26: The p-values of paired sample t-tests

Formulae	p -value(2-Tailed)			
	(17)	(16)	(15)	(11)
(10)	0.293	0.188	0.000	0.530
(11)	0.155	0.107	0.000	
(15)	0.000	0.000		
(16)	0.993			

As illustrated in Table 26, it can be concluded at 0.01 level of significance the quality of the results using the proposed formula (15) provides the coefficient matrix which yields the total cost which is significantly lower than those provided by the rest of the formulae. Hence, the proposed formula (15) can be considered as the best alternative as compared to the formulae provided by

Balinski [2] and Kowalski & Lev [14] for solving SFCTPs.

6 Conclusion

Three formulae have been proposed in this paper for constructing intermediate coefficient matrix as a basis for finding a local solution for SFCTPs. In addition, a comparison of the performances and quality of these proposed formulae is undertaken with the earlier proposed formulae proposed by Balinski [2] and also with the two formulae proposed by Kowalski & Lev [14]. It is proved that one of the formulae (11) proposed by Kowalski & Lev [14] fails to consider the cases when $A_{ij} = M_{ij}$ and $A_{ij} > M_{ij}$ as the values will be infinity when $A_{ij} = M_{ij}$ and assumes negative value in case $A_{ij} > M_{ij}$. In order to compare the formulae for their effectiveness, the results based on different problems with differing dimensions and with different A_{ij} were analyzed. Tests of hypotheses were performed and proved that one of the proposed formulae (15) provides the intermediate coefficient matrix C_{ij} which yields significantly lower total costs as compared to the remaining formulae.

Further work includes more experiments with the parameters of SFCTP and testing the proposed SFCTP on other real life problems. In addition, investigating the usage of metaheuristic techniques such as artificial immune systems, tabu search, particle swarm, simulated annealing and genetic algorithms for solving SFCTP will be explored.

7 Acknowledgement

This paper is supported by the Research Center at the College of Business Administration and the Deanship of Scientific Research at King Saud University, Riyadh.

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