

An Analysis of Improved Version of Vogel's Approach: An Approach to Discover Basic Explanation for Transportation Problem

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Abstract—The transport problem is the unique class in the field of functional Mathematics & Operating Work in linear programming. Existing transport problem algorithms like North West corner rule (NWC), the Last Cost Method (LCM), & Vogel Method for Approximation (VAM). It is available to find the simple feasible solution utilizing a Vogel Method for Approximation (VAM). This paper discusses Vogel's Approximation Method (VAM) limitations & has deployed a superior approach after overcoming this transport problem restriction. The Vogel Approximation (VAM) approach is the most effective transportation algorithm but it has a certain limit if two or more rows or columns end up with the greatest penalty costs. VAM has no better solution in this situation. In this paper, we have introduced an improved approach to this issue and an algorithm called the improved Vogel approach method (IV-VAM) solution, where the viable solution of this method is far closer to an optimum solution than VAM.

Index Terms—IV-VAM, LPP, VAM, Penalty, Transportation Cost, Transportation Problem (TP),.

I. INTRODUCTION

The problem of transportation (TP) is an unusual kind of problem with linear programming. This issue includes the transportation of the commodity from m to n terms, with a $1.a_2 \dots a_m$ and $b_1 \dots b_n$ capacities, respectively. There is also a penalty CIJ for the transport of the unit from Origin I to designations j . This penalty can be either cost, delivery time, delivery protection, etc. The unknown volume of x_{ij} from Origin to Target is the vector x_i . The problem of transportation (TP) is a special kind of problem with linear programming. This issue includes the transportation of the commodity from m to n terms, with a $1.a_2 \dots a_m$ and $b_1 \dots b_n$ capacities, respectively. There is also a penalty CIJ for the transport of the unit from Origin to designations. The unknown volume of x_{ij} from Origin I to Target j is the vector x_i .

Theory suggests a mathematical methodology to address imprecise concepts and problems there are several potential solutions. Zadeh developed the theory of fuzzy sets in 1965. Tanaka et al first proposed the idea of mathematical

fluid programming at a general level in the process work of Bellman and Zadeh's fluidized decision [1]. Chanas, Kolodziejczk, and Machaj [3] have given the concept of the fuzzy method to the transportation problem. Transportation issues in the fuzzy environment given some ideas by Saad and Abbas [2]. & Abdul Razak & Nagoorgani [4] found a fuzzy explanation for a 2 Stage with trapezoidal. Dinakar & Keerthivasan [3] obtained a fuzzy solution with the Best Candidate Method using periodesteemed triangular fuzzy no.

In this work, the fuzzy transportation problems using interval-valued triangular fuzzy numbers have been discussed. The initial basic feasible solution of the same transportation problem is obtained by different methods such as North West Corner rule, North East Corner rule, Least cost approach and Best Candidate Method [5].

Transportation problems are well-known in Operations study for its extensive applicability in genuine life. This is a particular form of n/w optimization problem, where goods are transported from several Origin to several Targets subject to Origin and Target demand & supply respectively, to minimize the overall transport costs. In 1941 Hitchcock introduced the fundamental problem of transport [6]. Effective methods have been developed to find a solution, mainly in 1951 by Dantzig [7] and then in 1953 by Charnes, Cooper, and Henderson [8]. The transportation question-solving protocol (TP) consists of three steps:

Step (i): Mathematical Formulation of the TP.

Step (ii): To find a feasible initial solution.

Step (iii): Optimize the initial solution that is possible in step (ii).

The Transportation issue is to define transportation expense, to establish the number of trips of the raw material that hit a certain Origin point for a particular time while meeting the supply & demand limits from Origin to Target in the factory. The transportation issue is It is also

function. We provide easy solutions to complex performance and convergence problems with an association of energy Origin and constant power loads using the port-controlled Hamiltonian method. An FC boost converter (2,5kW 2-phase interleaved converter) to support the proposed control law[12].

R. Liu et al.[2019]proposed The distributional algorithm estimation method (EDA) for the solution of Route Planning problems for AUVs in a complex setting is known as the fixed-height histogram (LFHH). To order to improve its accuracy and convergence, To speed up the discovery of alternative routes, a smooth approach is used. To control complex variables, a design window is often used. LFHH is tested with dynamic factor variations in complex 2-D and 3-D settings, and experimental findings confirm the efficacy of LFHH[13].

P. He, G. Jiang, S. Lam &D. Tang [2019]investigate the Travel-time forecast,which takes into account the time of the commuter on many bus journeys, as well as their waiting time at travel points. A new system is introduced, in which the time and waiting time of a given journey are individually calculated from various data sets (i.e. lines, busses & road networks), and the results are compiled into a final travel time forecast. We assess the impact factors of bus driving times empirically& establish a long-term memory model that can precisely forecast the driving time for each section of the bus routes.We also show that waiting times at points of transition have a major effect on the overall time and that calculating the period of waiting is not negligible since a fixed time of waiting for the distribution can not be taken into account. We implement a novel historical average interval approach to reliably predict waiting time, which can effectively fix correlation and sensitivity problems with waiting time forecasts. Real-world data studies have shown that the proposed method substantially exceeds six baseline approaches for all considered scenarios[14].

M. Sam'anetal. [2018] intended to allocate fuzzy transportation expenses that havethe same ranking value on Fuzzy Transportation Algorithm is randomly chosen to deal with the problem of full-scale transport. This way, however, affects the base cell of which the full unit of estimated fuzzy quantities must be calculated. Then, by adding weight with the SAW technique, the Modified Fuzzy Transports Algorithm showed no such foggy transport costs. A case study is resolved & the results are compared to the solutionsof the current algorithm to explain the algorithm suggested modifications. Since the proposed algorithm is an immediate extension of the classical method, it is simple to understand and practical to the plannedimprovement[15].

A. Vinyl and D. F. Silva[2018] discussed the Conduct of a Monte Carlo simulation test to spread the possibility of

route lengths b/w small no. of random location for Traveling-Salesman-problems (TSP). We regard as a convex field where a fixed no. of random locations are generated and the respective euclidean TSP route is located from a known probability distribution. This approach is thoroughly simulated and the resulting experimental distribution for the TSP Tours was analyzed both quantitatively and qualitatively. We show that the duration of the TSP tour is well some assumptions of the geographic shape and probability distribution of locations[16].

Ivaschenko, I. Syusin, and P. Sitnikov.[2017] proposed a new concept for TISP which develops intelligent software solutions for transport logistics. The TISP is the intermediary service platform for transport. The key emphasis is on the need to improve the quality of transport services offered. The Smart Transportation Platform is a software system that provides virtual decision points for smart transport companies to compete with and cooperate in an interconnected environment. For the distribution of products and services between fixed numbers of consumers and pickup and delivery service to unplanned customers, the examples of effective TISP implementation in practice are provided[17].

M. A. Manzoor and Y. Morgan[2017] proposed a method based on Linear Support for this problem by a vector machine. In this job, the algorithm Scale-Invariant Transform Function (SIFT) will be used The word bag model is used to represent local features as a permanent length vector representing an image. to extract and reflect local interest. This approach is evaluated on the available vehicle production and model data and is achieved with promising results [18].

IV. PROBLEM DOMAIN

The transportation issue is one of the subclasses of Linear Programmable Problem(LPP) to carry different quantities of a single homogeneous product originally stored in different locations in a manner that reduces overall transport costs. To achieve this aim we need to know how big and where supplies are available and how much is needed.Moreover, the cost of moving a single commodity unit from different Origin to different Targets must be understood.The transportation problem is an important class of linear programming problem aimed at transporting various amounts of a single homogenous product which are transported to different Targets at different Origins in such a way as to reduce transport costs.Transportation problem arises in situations involving physical movements of goods e.g. milk and milk products from plants to cold storages, cold storages to wholesalers, wholesalers to retailers and retailers to customers. The solution of a TP is to determine the quantity to be shifted from each plant to each cold storage to maintain the supply and demand

requirements at the lowest transportation cost.

V. PROPOSED METHODOLOGY

We stated at the outset that certain methods exist to overcome transport problems such as North West Corner Law (NWC), Least Cost System (LCM), and VAM, etc. We address Vogel's approach method (VAM) in this section & our proposed approach improved Vogel's Approximation approach.

A. For Vogel's Approximation Approach Existing Algorithm

Vogel Approach (VA) is a recursive method to compute an appropriate feasible alternative of a transportation hitch. This approach is better than the other 2 approaches i.e. North West Corner Rule (NWC) & Least cost approach (LCA), Because obtained fundamental feasible answer. The optimal explanation is closer to this method. The current Vogel Approximation Method (VAM) algorithm follows:

Step-1: Identify the boxes with high and high transport costs write in each line the penalty against the corresponding row on the table side.

Step-2: Identify the minimum and next to minimize transport costs in each column and enter the penalty on the Opposite column side of the table. When the minimum cost is displayed in a row or column two or more times, pick the same costs as the minimum and next to minimize costs and penalty will be 0.

Step-3: a. List the column and row with the highest penalty and arbitrarily sever ties. Assign the variable to the minimum cost in the selected row or column as much as possible. Change supply & demand & delete the column or lines. If a row and a column are met, only one is excluded and a zero supply or demand is applied to the remaining row or column. b. When two or more charges are of the same size, select one (or select the top or far-left row).

Step-4: If the supply or demand of exactly 1 row or 1 column remains uncrossed, Stop. where one row or column with +ve demand or supply is left out, basic variables shall be calculated by the lowest cost approach in the row or column. c. If (rest) 0 supply or demand occurs for all uncrossed rows or columns, the zero basic variables are calculated by the least cost rule. Stop. d. or else, back to Step-1.

B. Finding Limitations Of Vogel's Approximation Method (Vam)

In the VAM method, the penalties are based on the difference between each row and each column of two minimum costs. One of the 2 min. costs are lowest & the other is too elevated. The maximum penalty implies that. Choose this row or column in the VAM algorithm, which includes the highest penalty, to ensure that the current

Recurrence is less expensive [19]. The VAM algorithm selects either one (or selects the most top row or extreme connection column) [(2.) Step-3(b)] if the maximum cost of penalties appears in a row or column. Nevertheless, the biggest drawback is not necessarily guaranteed the lowest price because the difference between the two pairs can be equal if the one pair is smaller than the other. The difference between 10 and 5 and the difference between 7 and 2 are the same, but the second pair has the smallest number. Of this reason, it can be assumed that the lowest costs in the current VAM algorithm will not be chosen so that total transport costs will not be reduced in the topmost or far-left position. VAM can not be issued in this case.

C. Proposed Algorithm For Improved Vogel's Approximation Method (IV-Vam)

In the debate above, we have resolved this issue and put forward an enhanced algorithm called a Vogel's logical approximate creation (IV-VAM) as the basis for Vogel's approach process. We have also suggested an improved algorithm. We have solved this VAM problem by choosing the row or column the contains the least possible sum and the maximum allocation if two or four or more columns or rows are maximum penalties. In this algorithm, we have resolved this problem. The following is the algorithm of IV-VAM: Set o_i to be supply amount of the i^{th} Origin & t_j be the sum of demand of j^{th} Target and c_{ij} be unit transportation.

Step-1: Varyify: if $o_i < 0$ & $t_j < 0$ then end

Step-2: If $\sum_i o_i > \sum_j t_j$ or if $\sum_i o_i < \sum_j t_j$ Then balance the issue of transport by adding demand or supply of dummies.

Step-3: Recognize the smallest & lowest expense per column and row and measure the difference that is the penalty. P_i is the penalty for rows and P_j is the penalty for columns.

$$P_i = |C_{ih} - C_{ik}| \text{ and } P_j = |C_{hj} - C_{kj}|$$

Step-4: Select $\max(P_i, P_j)$ Choose the lowest row or column cost that has a high penalty & max. possible amount x_{ij} i.e. $\min(o_i, t_j)$. If in two or more cells in the same column, the lowest cost appears and then the far left or lowest cost cell is selected.

Step-5: If a tie occurs in any rows or columns in the biggest penalties, pick the line or column at a lower cost.

Step-6: Change the supply & demand & eliminate the column or lines. If row & column are filled at the same time, then one of them is crossed out & the remaining column or row is supplied or demand-free.

Step-7: If exactly one row or one column remains unregulated with zero supply or demand, stop.

a. If one row or column has +ve volume or request, the

basic variables shall be calculated by the lower-cost approach in the row or column.

- b. If (rest) zero supply or demand in all uncrossed line(s) or columns, the simple zero variables are calculated using the lowest cost process. Halt.
- c. Otherwise, go to Step-3.

D. NUMERICAL ILLUSTRATION

Consider several special kinds of transport issues where the greatest penalty occurs in two or more rows or columns, solving these by using the Vogel Approximation Method (VAM) and the modified version of the Vogel Approximation Method (IV-VAM) suggested method.

EX.-1:

Consider a transport problem mathematical model in the following

Table-1.1:

Origin	Target					Supply
	T1	T2	T3	T4	T5	
O1	10	8	9	5	13	100
O2	7	9	8	10	4	80
O3	9	3	7	10	6	70
O4	11	4	8	3	9	90
Demand	60	40	100	50	90	

RESULT OF EX.-1 USING Improved Version OF VOGEL APPROXIMATION METHOD :

Expenses are shown in the allocations & right corner, in the bottom left corner are shown

Recurrence-1:

Origin	Target					Supply	Penalty of Row
	T1	T2	T3	T4	T5		
O1	10	8	9	5	13	100	3
O2	7	9	8	10	4	80	3
O3	9	3 40	7	10	6	30	3
O4	11	4	8	3	9	90	1
Demand	60		100	50	90		
Penalty of Column	2	1	1	2	2		

The biggest fines in Recurrence-1 can be found in O1, O2, O3 rows three times, but the lowest costs are shown in cells (O3, T2).

Recurrence-2:

Origin	Target					Supply	Penalty of Row
	T1	T2	T3	T4	T5		
O1	10	8	9	5	13	100	4

O2	7	9	8	10	4	80	3
O3	9	3 40	7	10	6	30	1
O4	11	4	8	3 50	9	40	5
Demand	60		100		90		
Penalty of Column	2		1	2	2		

Recurrence-3:

Origin	Target					Supply	Penalty of Row
	T1	T2	T3	T4	T5		
O1	10	8	9	5	13	100	1
O2	7	9	8	10	4 80		3
O3	9	3 40	7	10	6	30	1
O4	11	4	8	3 50	9	40	1
Demand	60		100		10		
Column Penalty	2		1		2		

Recurrence-4:

Origin	Target					Supply	Penalty of Row
	T1	T2	T3	T4	T5		
O1	10	8	9	5	13	100	1
O2	7	9	8	10	4 80		
O3	9	3 40	7	10	6 10	20	1
O4	11	4	8	3 50	9	40	1
Demand	60		100				
Column Penalty	1		1		3		

Recurrence-5:

Origin	Target					Supply	Row Penalty
	T1	T2	T3	T4	T5		
O1	10	8	9	5	13	100	1
O2	7	9	8	10	4 80		
O3	9	3 40	7	10	6 10	20	2
O4	11	4	8 40	3 50	9		3
Demand	60		60				
Penalty of Column	1		1				

Recurrence-6:

Origin	Target					Supply	Penalty of Row
	T1	T2	T3	T4	T5		
O1	10	8	9	5	13	100	1
O2	7	9	8	10	4	80	
O3	9	3	7	10	6	40	2
O4	11	4	8	3	9	40	
Demand	60		40				
Penalty of column	1		2				

The cost is shown in the right and the lower-left corner, the allocations are shown.

Total Transportation Cost (By VAM):

$$(5 \times 50) + (10 \times 50) + (9 \times 10) + (7 \times 50) + (6 \times 10) + (4 \times 40) + (8 \times 50) + (4 \times 80) = 2130$$

Observation: We considered the feasible solution given by VAM for Ex. 1 to 2130 & IV-VAM to 2070 to be less than VAM. It was considered that

Ex.-2:

Assume a Mathematical Model for a Transportation Problem in below:

Recurrence-7:

Origin	Target					Supply	Penalty of Row
	T1	T2	T3	T4	T5		
O1	10	8	9	5	13	60	
O2	7	9	8	10	4	80	
O3	9	3	7	10	6	40	
O4	11	4	8	3	9	40	
Demand	0						
Penalty of Column							

Table-2.1:

Origin	Target				Supply
	T1	T2	T3	T4	
O1	7	5	9	11	30
O2	4	3	8	6	25
O3	3	8	10	5	20
O4	2	6	7	3	15
Demand	30	30	20	10	

In Recurrence-6, only one row has remains with +vedemand & supply Then the sum is distributed by Least Cost Method by the IV-VAM algorithm. The following is the final workable solution table:

RESULT OF EXAMPLE-2 USING Improved Version Of VOGEL'S APPROXIMATION METHOD (IV-VAM):

Costs are representing in the right-top corner & allocations are representing in the bottom-left corner.

Table-2.1:

Origin	Target					Supply
	T1	T2	T3	T4	T5	
O1	10	8	9	5	13	100
O2	7	9	8	10	4	80
O3	9	3	7	10	6	70
O4	11	4	8	3	9	90
Demand	60	40	100	50	90	

Origin	Target				Supply
	T1	T2	T3	T4	
O1	7	5	9	11	30
O2	4	3	8	6	25
O3	3	8	10	5	20
O4	2	6	7	3	15
Demand	30	30	20	10	

Sum Transportation Cost (By Improved Version-VAM):

$$(9 \times 40) + (10 \times 60) + (4 \times 80) + (3 \times 40) + (7 \times 20) + (8 \times 40) + (3 \times 50) + (6 \times 10) = 2070$$

Sum Transportation Cost (By IV-VAM):

$$(5 \times 5) + (7 \times 5) + (9 \times 20) + (3 \times 20) + (2 \times 5) + (3 \times 25) + (3 \times 10) = 415$$

RESULT OF EXAMPLE-1 by VOGEL'S APPROXIMATION METHOD(VAM):

Result OF EX.-2 by VOGEL'S APPROXIMATION METHOD(VAM):

Source	Target					Supply	Penalty of Row					
	T1	T2	T3	T4	T5							
O1	10	8	9	5	1	100	3	1	1	1	1	1
O2	7		9	8	4	80	3	3				
O3	9	3	7	1	6	70	3	3	1	1	2	2
O4	11	4	8	3	9	90	1	4	1	1	3	
Demand	60	40	10	5	9							
Penalty Column	2	1	1	2	2							
	2	1	1		2							
	2		1		2							
	1		1		3							
	1		1									
	1		2									

Expenses are displayed in the left and allocations in the bottom-left corner.

Table-2.2:

Source	Target				Supply
	T1	T2	T3	T4	
O1	7	5	9	11	30
O2	4	3	8	6	25
O3	3	8	10	5	20
O4	2	6	7	3	15
Demand	30	30	20	10	

Observation: By analyzing that VAM gives feasible result for Ex.-2 is 470 & IV-VAM gives 415 which is lesser than VAM

EX.-3:

Assume a Mathematical form of a Transportation issue in below:

Table-3.1:

Origin	Targets							Supply
	T1	T2	T3	T4	T5	T6	T7	
O1	12	7	3	8	10	6	6	60
O2	6	9	7	12	8	12	4	80
O3	10	12	8	4	9	9	3	70
O4	8	5	11	6	7	9	3	100
O5	7	6	8	11	9	5	6	90
Demand	20	30	40	70	60	80	100	

Results Of EX.-4 by Improved Version OF VOGEL'S APPROXIMATION METHOD (IV- VAM):

The expenses are shown in the right corner & in the lower-left corner allocations are shown

Table-3.2:

Origin	Targets							Supply
	T1	T2	T3	T4	T5	T6	T7	
O1	12	7	3	8	10	6	6	60
O2	6	9	7	12	8	12	4	80
O3	10	12	8	4	9	9	3	70
O4	8	5	11	6	7	9	3	100
O5	7	6	8	11	9	5	6	90
Demand	20	30	40	70	60	80	100	

Total transportation Cost:

700 + 200 + 300 + 400 + 400 + 800 + 800 + 600 + 800 + 600 + 600 + 200 + 900 = 7000

VI. EXPERIMENTAL ANALYSIS

For the above, we have found that feasible solutions by the enhanced Vogel Approximation Method version (IV-VAM), some of which are the same solution as the Vogel Approximation Method (VAM), are lower, and others very close to the optimal result. The table below shows the analysis of these solutions:

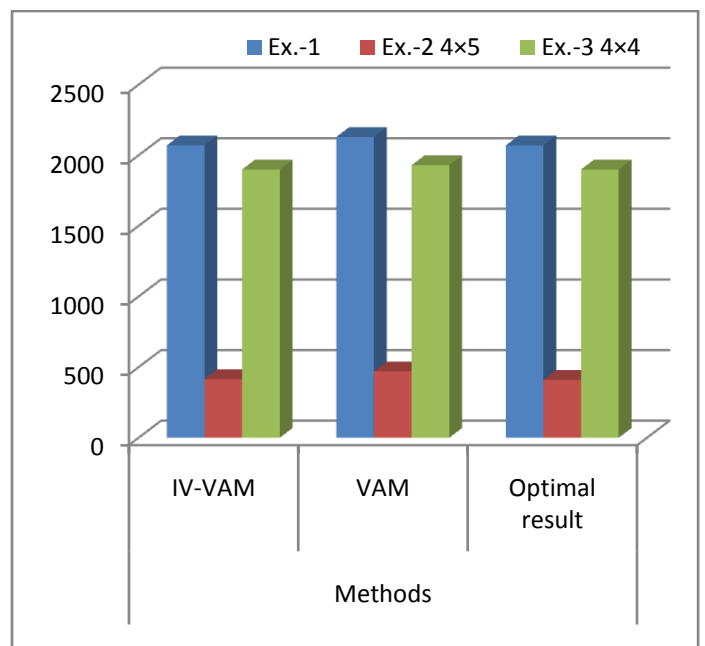


Figure 1.1

Table-6:

Transportation Problem	Problem range	Methods		
		IV-VAM	VAM	Optimal result
Ex.-1	4×5	2070	2130	2070
Ex.-2	4×4	415	470	410
Ex.-3	5×7	1900	1930	1900

VI. CONCLUSION

Vogel method is preferred over the NWCM and VAM because the starting fundamental feasible solution obtained by this approach is either an optimal solution or extremely nearer to the optimal explanation. This method helps to reduce transport costs by interpreting the transport costs from one place to another in a mathematical table. The column reflects the centers of demand, the row the points of supply. In this paper, we come across a restriction of Vogel’s Approximation Approach & deployed a superior algorithm by settling this restriction approach —Improved version of Vogel’s Approximation Method (IV-VAM) using fixed point for defining Transportation Problem. From the given ex. & other transportation problems, IV-VAM is the least feasible answer than VAM, is extremely similar to an optimal result & at times the same as an optimal solution.

VII. FUTURE SCOPE

A new alternative approach has been developed to solve transport problems that offer either a near-optimal solution or an optimal solution. The new alternate methods are used only for Vogel’s Approximation Method. We could not, however, seek to address other transport- and transshipment-related optimization issues. Thus, the unbalanced problem can be overcome with new alternative methods built in this study.

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