



GLOBAL JOURNAL OF SCIENCE FRONTIER RESEARCH: F
MATHEMATICS AND DECISION SCIENCES
Volume 22 Issue 4 Version 1.0 Year 2022
Type: Double Blind Peer Reviewed International Research Journal
Publisher: Global Journals
Online ISSN: 2249-4626 & Print ISSN: 0975-5896

An Improved Hungarian Algorithm for a Special Case of Unbalanced Assignment Problems

By Mohammad Shyfur Rahman Chowdhury

International Islamic University Chittagong

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GJSFR-F Classification: DDC Code: 005.1 LCC Code: QA76.6

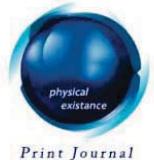


AN IMPROVED HUNGARIAN ALGORITHM FOR A SPECIAL CASE OF UNBALANCED ASSIGNMENT PROBLEMS

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An Improved Hungarian Algorithm for a Special Case of Unbalanced Assignment Problems

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Abstract- The current Hungarian approach to solving unbalanced assignment issues is based on the notion that some tasks should be delegated to fictitious or covert components, and those studies should be left unperformed. In real-world scenarios, it may be desirable to carry out all of the tasks on fundamental details. To do this, multiple tasks may be distributed to a single machine. The current research's enhanced Hungarian method for addressing unbalanced assignment challenges results in the ideal work assignment policy. An example using numbers shows how well the suggested strategy works and how effective it is. The acquired result is then likened to other current approaches to demonstrate our algorithm's superiority.

I. INTRODUCTION

One of the most significant applications of optimization theory is the Assignment Problem, where various tasks need to be distributed across components to be completed, such as spreading personnel to offices and drivers to buses, among other things. There have been numerous ways of exploring the best policy for assigning jobs to components. The Hungarian approach is the most frequently used method for determining the best assignment policy. The Hungarian Method was named by Kuhn (Kuhn, 1995) and was based in significant part on earlier work by two Hungarian mathematicians, Egervary and D. Konig. When the Hungarian approach is used to address an uneven assignment problem, the technique assigns the jobs to fake components that do not perform them (if the number of jobs is greater than the number of components). It seems impossible to leave jobs unfinished in real-world situations. As a result, rather than assigning extra work to dummy components, it is recommended to do each and every job that may be done by assigning multiple jobs to a single component.

To tackle assignment problems, the Hungarian algorithm (Chen., 2011) (W. B. Lee, 1997) and specific heuristic algorithms (e.g., simulated annealing method (B. Li, 2002) ant colony algorithm (Y. Liang, 2005) (R. K. Yin, 2008) particle swarm algorithm (W. F. Tan, 2007) and genetic algorithm (S. Q. Tao, 2004)). Heuristic approaches are frequently employed to solve problems with assignments of high complexity. However, because the search is conducted at random, it cannot guarantee that the optimal result will be obtained. The Hungarian algorithm is an algorithm with a mathematical foundation. The Hungarian algorithm is commonly used to tackle assignment problems because of its simplicity and ability to find the best solution without requiring validation

Author: Assistant Professor, Department of Business Administration, International Islamic University Chittagong. e-mail: src.dba@iiuc.ac.bd



(X. Q. Hu, 2006) (M. J. Liu, 2013) (H. Z. Zhang, 2009) (L. W. Huang, 2007) (Y. Wang, 2005).

In (J. L. Du, 2010), an enhanced Hungarian algorithm, the "add zero row approach," was presented to handle the incomplete assignment problem based on a study of the standard Hungarian algorithm. The Hungarian method was first introduced by (T. M. Chang, 2004) and, it was used to solve a common assignment problem, such as a marriage assignment. (Ma., 2014) suggested a new method, the "difference method," for solving the non-standard assignment problem: "tasks more than the number of people." This method is more straightforward than the standard algorithm because it does not require using a new matrix to replace the original coefficient matrix at the beginning and instead solves the problems directly on the old coefficient matrix. (Qiu., 2013) suggested an enhanced Hungarian algorithm for studying multiple maintenance scheduling problems in hostile environments. A quick order reduction optimization approach based on the classical Hungarian algorithm was proposed (J. X. Ren, 2014) to increase the efficiency of the distribution of cloud computing activities. The order of the matrix is quickly lowered and, the computing efficiency is increased by deleting the matrix elements that are determined. Reference (R, 2014) used the Hungarian technique to investigate the dynamic power allocation of weapon-targets by changing it into an assignment issue. Furthermore, the traditional Hungarian algorithm has been used to tackle business and technical challenges in a variety of disciplines (P. Hahn, 1998) (Kuhn., 2012) (E. M. Loiola, 2007) (T. Tassa, 2008) (S. Promparmote, 2006) (M. H. Paul, 2013).

According to many authors, the unbalanced assignment problem has many solutions, all of which assume that all jobs are finished. Kumar (Kumar, 2006) came up with a fresh approach to address the problem of uneven assignments. The decision-maker can allocate several tasks to a single component using his methodology. The Lexi Search Approach, developed by Haragopal and Yadaiah (V. Yadaiah, 2016), is a more effective technique for dealing with imbalanced assignment problems that yield the same outcomes as Kumar (Kumar, 2006). In the same year, Kumar's (Kumar, 2006), Haragopal's and Yadaiah's (V. Yadaiah, 2016) methods were surpassed by an approach provided by Betts and Vasko (Vasko, 2016).

II. MATHEMATICAL FORMULATION

Consider the processing cost of the j th job on the i th component be C_{ij} , where $i = 1, 2, \dots, m$; $j = 1, 2, \dots, n$. The challenge is to create an ideal work assignment method that ensures that every task is finished while keeping the overall cost of doing so as low as possible.

Mathematical model of an unbalanced assignment problem can be expressed as,

$$\text{Minimize: } Z = \sum_{i=1}^m \sum_{j=1}^n C_{ij} X_{ij}$$

Subject to constraints

$$\sum_{j=1}^n X_{ij} \geq 1, i = 1, 2, \dots, m \quad (1)$$

$$\sum_{i=1}^m X_{ij} \geq 1, j = 1, 2, \dots, n \quad (2)$$

$$X_{ij} = 0 \text{ or } 1$$

III. PROPOSED ALGORITHM

Think about the issue of distributing a group of "n" jobs. $J = J_1, J_2, \dots, J_n$ to an execution set with "m" components. $C = (C_1, C_2, \dots, C_m)$. X_{ij} , $i = 1, 2, \dots, m$; $j = 1, 2, \dots, n$ ($n \geq m$), indicating that there are more tasks than components. In this case, n stands for columns and m for rows.

Step-01: Enter the following values: m, n

Step-02: Each column's lowest cost should be subtracted from the column it belongs. This process results in each column reducing the cost column by having at least a single zero.

Step-03: Determine each row's lowest cost and then deduct it from the associated row.

Step-04: Analyze the feasibility of doing an ideal task so that the smallest number of lines needed to cover each zero is calculated. If the number of lines and rows equals one, proceed to Step 7; if not, proceed to Step 5.

Choose the minimum uncovered cost if the number of lines exceeds the number of rows.

1. Take the least exposed cost in the table and subtract it from each exposed cost.
2. The cost at each intersection point is added by those minimum cost.

Step 06: If the case (the total number of lines and rows is equal) fails, then continue steps 04 and 05.

To assign the work, look for a row with only one zero. Choose that zero and block the other zeros in the relevant columns (the same component can be performed on more than one job, but the same job cannot be assigned more than one component).

Step-08: Assign the value with the lowest cost in the initial problem if there is a tie, that is, if any rows have two or more zeros.

Step-09: Repeat steps 7 and 8 until all positions have been filled, that is, all jobs have been assigned to one or more processing components.

Step-10: End

IV. PARALLELISM BETWEEN PROPOSED ALGORITHM AND HUNGARIAN ALGORITHM

Table 1

Proposed	Hungarian
<ul style="list-style-type: none"> • This tactic is employed to address issues with unbalanced assignments. • If the number of jobs exceeds the number of processing components, all jobs must be completed using the available components.“ • There are no unfinished projects. • Related to at least one job that can be performed by a single component. • A single component can only be assigned to a single job. 	<ul style="list-style-type: none"> • This tactic is employed to address issues with unbalanced assignments. • If the number of jobs exceeds the number of components, the remaining jobs are performed by dummy components. • Some jobs aren't being completed. • A single component can only do one thing. • Only one component can be allocated to a single job.



V. MATHEMATICAL ANALYSIS

Let us take a problem of 8 jobs and 5 processing components with associated execution costs as given in Table 2.

Table 2

	J₁	J₂	J₃	J₄	J₅	J₆	J₇	J₈
C₁	270	260	220	190	300	320	180	250
C₂	210	190	300	200	290	180	190	310
C₃	190	260	230	220	280	190	300	290
C₄	250	210	180	190	290	240	190	300
C₅	160	180	160	140	210	170	180	200

Table 3: Follows steps 1, 2 and, 3

	J₁	J₂	J₃	J₄	J₅	J₆	J₇	J₈
C₁	50	80	60	90	40	150	0	50
C₂	50	0	130	70	40	0	0	100
C₃	60	60	50	50	10	0	100	70
C₄	40	20	10	70	80	60	0	90
C₅	0	0	0	0	0	0	0	0

Table 4: Follows step 4

	J₁	J₂	J₃	J₄	J₅	J₆	J₇	J₈
C₁	40	80	60	50	90	150	0	50
C₂	40	0	130	50	70	0	0	100
C₃	10	60	50	60	50	0	100	70
C₄	80	20	10	40	70	60	0	90
C₅	0	0	0	0	0	0	0	0

By step 5; from the uncovered costs, choose the lowest cost (i.e. 10)

- Deduct 10 from each exposed cost in the matrix above.
- To get Table 5, sum up 10 at each of the intersection points.

Table 5

	J₁	J₂	J₃	J₄	J₅	J₆	J₇	J₈
C₁	30	80	50	40	80	150	0	40
C₂	30	0	120	40	60	0	0	90
C₃	0	60	40	50	40	0	100	60
C₄	70	0	0	30	60	60	0	80
C₅	0	10	0	0	0	10	0	0

Following steps 6 and 7, we allocate job J₃ to component M₁ and cross off the remaining zeros in the row corresponding to J₃; as a result, row four has just one zero. As stated in Table 6, allocate work J₇ to component M₄.

Table 6

	J₁	J₂	J₃	J₄	J₅	J₆	J₇	J₈
C₁	30	80	50	40	80	150	0	40
C₂	30	0	120	40	60	0	0	90
C₃	0	60	40	50	40	0	100	60
C₄	70	0	0	30	60	60	0	80
C₅	0	10	0	0	0	10	0	0

Notes

According to step 8, there is a tie in the 2nd and 3rd rows (containing two zeros). We allocate J_4 to component M_2 (in Table-7) because the cost associated with this position is the lowest in the original cost matrix.

Table 7

	J_1	J_2	J_3	J_4	J_5	J_6	J_7	J_8
C_1	30	80	50	40	80	150	0	40
C_2	30	0	120	40	60	0	$0 \times$	90
C_3	0	60	40	50	40	0	100	60
C_4	70	20	0	30	60	60	$0 \times$	80
C_5	0	10	$0 \times$	0	0	10	10	0

Then following step 9, we get the final table-8.

Table 8

	J_1	J_2	J_3	J_4	J_5	J_6	J_7	J_8
C_1	30	80	50	40	80	150	0	40
C_2	30	0	120	40	60	0	$0 \times$	90
C_3	0	60	40	50	40	$0 \times$	100	60
C_4	70	20	0	30	60	60	0	80
C_5	$0 \times$	10	$0 \times$	0	0	10	10	0

VI. RESULTS AND DISCUSSION

Table 9 shows the work assignment policy that reduces the overall cost.

Component	Job	Cost
C_1	J_7	180
C_2	J_2, J_6	190, 180
C_3	J_1	180
C_4	J_3	190
C_5	J_4, J_5, J_8	140, 210, 200
Total Cost		1470

We find the total minimum cost in comparison to the other modified Hungarian methods like Kumar [26], Haragopal and Yadaiah [27], and Betts and Vasko [28].

VII. CONCLUSION

The study presents an improved Hungarian algorithm to solve a particular case (when the number of jobs is greater than the number of processing components) of an unbalanced assignment problem. Generally, most of the issues regarding assignments occur in the abovementioned case. The primary premise of our strategy is to allocate all tasks to be completed. If the created assignment plan is invalid, the virtual jobs will be changed, and the procedure will be continued until an actual optimal assignment strategy is discovered. We show that the assignment strategy found by the suggested approach is the best using mathematical analysis. It demonstrates that the revised algorithm is capable of determining the best assignment strategy.

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