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**Constrained Three Dimensional Job Assignment Model**

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**ABSTRACT**

We study the problem called “Constrained Three Dimensional Job Assignment Model (CTDJAM)”. For this we took  $M = (M_1 \cup M_2 \cup M_3 \cup \dots \cup M_p)$  sets of machines,  $W = W_1 \cup W_2 \cup W_3 \cup \dots \cup W_q$  sets of workers and  $J = J_1 \cup J_2 \cup J_3 \cup \dots \cup J_r$  sets of jobs. Let the number of elements (machines) in each set of machines be  $m_1, m_2, \dots, m_p$ , the number of elements in each set of workers be  $w_1, w_2, w_3, \dots, w_q$  and the number of elements in each set of jobs are  $n_1, n_2, n_3, \dots, n_r$ . The total number of machines, workers, jobs are in all sets are taken as  $m, w, n$  respectively. Out of  $n$  jobs the number of jobs to be assigned is  $n_0$  ( $n_0 < n$ ). Here third dimension is job. In the feasible assignment all the subsets of machines, workers and jobs should be represented. Our objective is to assign  $n_0$  jobs with minimum total cost/time with the above restrictions.

**Keywords:** Assignment, Dimension, Constrained.

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**1. INTROUDCTION**

Assignment problem is a special type of linear programming. It is concerned of assigning task to facility on a one to one basis in some optimal way. For e.g., A manager has four persons available for four separate jobs. His job is to assign each person to one and only job in such a way that the time and total cost is minimized.

So many researchers have developed the Assignment Problem.

Purusotham [1] had studied the problem called “Pattern Recognition based Lexi-Search Approach to the Variant Multi-Dimensional Assignment Problem”. The Multi Dimensional Assignment Problem is a combinatorial optimization problem that is known to be NP –Hard. On

that paper he discusses a problem with four dimensions.  $N$  jobs can be executed on  $N$  machines, at  $k$  facilities, using  $l$  concessions. Every job is to be scheduled on some machine at one of the facilities, using some concession. No two jobs can run on the same machine, at the same facility using the same concession. Furthermore, there is a specified maximum number of jobs that can be run at a given facility, and there is a maximum number of jobs that can avail of a given concession.  $C(i, j, k, l)$  be the cost of allocating job ' $i$ ' on machine ' $j$ ' at facility ' $k$ ' using the concession ' $l$ '. This is provided as a 4 dimensional array. The objective is to schedule the jobs in such a way that the constraints are met and the cost is minimized. Sobhan Babu [2] had also studied on the assignment is called "A new Approach for Variant Multi Assignment Problem".

In Hungarian method and Khun et al [4]-1955, Labeling process and the Line covering method are widely used to solve the Assignment Problem. Barr et al [5]-1977 have proposed alternating basis algorithm, while Hung et al [6]-1980 proposed a row algorithm based on Relaxation Method. Bertsekas[7]-1981 produced an algorithm for solving the Classical AP resembling the Hungarian Method in some ways but differs substantially in other aspects. A more general version of cost minimization Assignment Problem is considered by Geeta et al [8]-1993, where in addition to the (hiring) cost of workers performing the jobs, a supervisory cost is also considered.

The time minimization Assignment Problem (TMAP) is another important class of assignment problem. TMAP has been considered by many researchers like Garfinkel[9] (1971), Ravindran et al [10] (1977), Bhatia [11] (1977), Shalini Arora [12] (1997) and Balakrishna [13] (2009) under the usual assumption that work on all the  $n$  jobs starts simultaneously.

Vidhyullatha [3] had studied the problem called "Three Dimensional Group Assignment problem". In this problem there is set of machines and set of jobs are equal in number. Number of persons is the third dimension. Each person is assigned a fixed number of jobs to be performed by him. Each job has to be done on each machine by only one person. The time taken to complete all the jobs is the maximum time taken by a person to complete the jobs assigned to him. In her problem symbolically considered as  $N = \{1, 2, \dots, n\}$  set of  $n$  jobs,  $M = \{\text{set of } n \text{ machines}\}$  and  $P = \{1, 2, \dots, p\}$  set of  $p$  persons ( $p < n$ ).  $t(i, j, k)$ , the time taken by the  $k^{\text{th}}$  person to

complete  $i^{\text{th}}$  job on  $j^{\text{th}}$  machine is given. Out of  $n$  jobs each person has to complete the assigned number of jobs which totals to  $n$ . A person goes to next job only on the completion of the earlier one. Each job has to be completed individually on separate machine. All the persons start working on the assigned jobs simultaneously. The completion time of all the jobs is the maximum among  $p$  persons completion time. The objective is to assign  $n$  jobs to  $p$  persons with minimum total completion time with the above restrictions and it is a mini max problem.

## 2. PROBLEM DESCRIPTION

In this paper we study the problem called “Constrained Three Dimensional Job Assignment Model (CTDJAM)”. For this we took  $M$  machines,  $W$  workers and  $J$  jobs. Again,  $M$  machines are classified into  $p$  sets of machines. ,  $W$  Workers are classified into  $q$  sets of workers and  $J$  jobs are classified into  $r$  sets of jobs. i.e The number of sets in machines are  $p$ , The number of sets in workers are  $q$ , The number of sets in jobs are  $r$ . Let the number of elements in each set of machine are  $M_1, M_2, M_3, \dots, M_p$ . The number of elements in each set of workers are  $W_1, W_2, W_3, \dots, W_q$  and the number of elements in each set of jobs are  $N_1, N_2, N_3, \dots, N_r$ . Therefore the total number of machines in all sets taken as  $m$ , the total number of workers are taken as  $w$  and the total number of jobs are taken as  $n$ . Symbolically we can write as the following.

In this CTDJAM, we consider  $M=\{1,2,3\dots p\}$  sets of machines,  $W=\{1,2,3\dots q\}$  sets of workers and  $J=\{1,2,3\dots r\}$  set of jobs. The sets of machines  $M_1, M_2, M_3\dots, M_p$  are considered such that  $M=M_1 \cup M_2 \cup M_3 \cup \dots \cup M_p$  and  $1M_i=1m_i, 1M=1m$ . Here  $m_1+m_2+m_3+\dots+m_p=m$ . Similarly the set of workers  $W=\{1,2,3\dots q\}$  set of workers  $W_1, W_2, W_3, \dots, W_q$  are considered such that  $W=W_1 \cup W_2 \cup W_3 \cup \dots \cup W_q$  and  $1W_i=1w_i, 1W=1w$ . Here  $w_1+w_2+w_3+\dots+w_p=w$  and  $J=\{1,2,3\dots r\}$  set of jobs  $J_1, J_2, J_3, \dots, J_r$  are considered such that  $J=J_1 \cup J_2 \cup J_3 \cup \dots \cup J_p$  and  $1J_i=1n_i, 1J=1n$ . Here  $n_1+n_2+n_3+\dots+n_p=n$ .

Out of  $n$  jobs the number of jobs to be completed is  $n_0$  ( $n_0 < n$ ) i.e the total assigned number of jobs should be  $n_0 < n$  is truncation.  $t(i,j,k)$ , the cost taken by the  $i^{\text{th}}$  machine in  $M$  is used by  $j^{\text{th}}$  worker in  $W$  for doing the  $k^{\text{th}}$  job in  $J$  is given. Here third dimension is job. The total assigned number of machines in a particular case of machine ( $M_s$ ) is less than or equal to that number  $m_s$ . Similarly the total assigned number of workers in a particular case of worker ( $W_s$ ) is

less than or equal to that number  $w_s$  and the total assigned number of jobs in a particular case of job ( $J_s$ ) is less than or equal to that number  $n_s$ .

In the feasible assignment schedule all the subset of machines should be represented. i.e. in the solution there is at least one machine from each of the  $p$  sets of machines. Similarly in the feasible assignment schedule all the subset of workers should be represented. i.e. in the solution there is at least one worker from each of the  $q$  sets of workers and in the feasible assignment schedule all the subset of jobs is should be represented. i.e. in the solution there is at least one job from each of the  $r$  sets of jobs. Our objective is to assign  $n_0$  jobs with minimum total cost with the above restrictions.

### 3. MATHEMATICAL FORMULATION

$C(i,j,k)$  means that cost of the  $i^{\text{th}}$  machine in  $M$  is used by  $j^{\text{th}}$  worker in  $W$  for doing the  $k^{\text{th}}$  job in  $J$ . From  $n$  jobs we want to assign  $n_0$  jobs. Here  $n_0$  less than  $n$ .

$$\text{Minimize } Z(X) = \sum_i \sum_j \sum_k C(i, j, k) \cdot x(i, j, k) \quad \text{For } i \in M, j \in W, k \in J \dots (1)$$

**Subjected to constraints**

$$\sum_i \sum_j \sum_k x(i, j, k) = n_0 \quad \text{For } i \in M, j \in W, k \in J \dots (2)$$

Here  $n_0 < n$

$$\sum_i \sum_j \sum_k x(i, j, k) \leq m_s, i \in M_s \dots (3)$$

$$\sum_i \sum_j \sum_k x(i, j, k) \leq w_s, j \in W_s \dots (4)$$

$$\sum_i \sum_j \sum_k x(i, j, k) \leq n_s, k \in J_s \dots (5)$$

If  $x(i,j,k) = 1, i \in M_s$  then  $MI(s) = 1$  and

$$\sum_{i=1}^p MI(i) = p \quad \dots(6)$$

If  $x(i, j, k) = 1, j \in W_s$  then  $WI(s) = 1$  and

$$\sum_{i=1}^q WI(i) = q \quad \dots (7)$$

If  $x(i, j, k) = 1, i \in J_s$  then  $JI(s) = 1$  and

$$\sum_{i=1}^r JI(i) = r \quad \dots(8)$$

$$x(i, j, k) = 0 \text{ or } 1 \quad i \in M, j \in W, k \in J \quad \dots (9)$$

The constraint (1) is the objective function which measures the minimum time of completion of all the  $n_0$  jobs under the given restrictions.

The constraint (2) describes the restriction that the total number of assigned jobs ( $n_0$ ) less than or equal to the  $n$ .

The constraint (3) describes in the job assignment schedule the number of machines assigned from the set  $M_s$  should be less than its number  $m_s$ .

The constraint (4) describes in the job assignment schedule the number of workers assigned from the set  $W_s$  should be less than its number  $w_s$ .

The constraint (5) describes in the job assignment schedule the number of jobs assigned from the set  $J_s$  should be less than its number  $n_s$ .

The constraint (6) describes in the assignment schedule all the subset of machines are involved. i.e. in the solution there is at least one machine from each of the  $p$  sets of machines.

The constraint (7) describes in the assignment schedule all the subset of workers are involved. i.e. in the solution there is at least one worker from each of the  $q$  sets of workers.

The constraint (8) describes in the assignment schedule all the subset of jobs are involved. i.e. in the solution there is at least one worker from each of the r sets of jobs.

The constraint (9) describes the restriction that the  $i^{th}$  machine in M is used by  $j^{th}$  worker in W for doing the  $k^{th}$  job in J then  $x(i, j, k) = 1$  otherwise 0.

**4.NUMERICAL ILLUSTRATION**

The concept and algorithm developed by illustrated by a numerical example for which taken as the number of machine sets  $p=5$  and in each set having the machines are  $M_1=3, M_2=2, M_3=3, M_4=2, M_5=3$ . Therefore the total number of machines are  $m_1 + m_2 + m_3 + m_4 + m_5 = 3 + 2 + 3 + 2 + 3 = 13 = m$ . i.e. the total number of machines = 13. Similarly the number of workers sets  $q=4$  and in each set having the workers are  $W_1=3, W_2=4, W_3=3, W_4=3$ . Therefore the total number of workers are  $w_1 + w_2 + w_3 + w_4 = 3 + 4 + 3 + 2 = 12 = w$ . i.e. the total number of workers = 12 and the number of jobs sets  $r=5$  and in each set having the jobs are  $J_1=3, J_2=4, J_3=2, J_4=4, J_5=3$ . Therefore the total number of jobs are  $n_1 + n_2 + n_3 + n_4 + n_5 = 3 + 4 + 2 + 4 + 3 = 17 = n$ . i.e. the total number of jobs  $n=17$ .

In the following numerical example,  $C(i, j, k)$ 's taken as non- negative integers it can be easily seen that this is not a necessary condition.  $C(i, j, k)$  means the cost/time that  $i^{th}$  machine working on  $j^{th}$  worker for  $k^{th}$  job. The following table represents the requirement of the cost for do the job with respect to corresponding machine and worker. Then the cost array  $C(i, j, k)$  is given below.

TABLE-1

$D(i,j,1)=$	3	1	2	1	$D(i,j,2)=$	-	6	-	1
	2	3	3	-		4	13	1	5
	4	-	5	2		2	-	8	2
	8	7	-	6		1	3	12	-
	-	9	10	-		13	-	4	12

D(i,j,3)=	1	18	13	-
	14	7	14	19
	-	2	15	3
	6	20	5	-
	-	4	5	-

D(i,j,4)=	3	-	4	-
	11	-	20	8
	7	12	5	-
	3	-	21	16
	13	14	4	6

D(i,j,5)=	-	1	-	6
	4	5	11	3
	17	18	15	-
	5	17	4	8
	7	-	-	2

In **table-1**,  $D(4, 3, 2) = 12$  means that the cost of the 4<sup>th</sup> machine set in M is used by 3<sup>rd</sup> worker set in W for doing the 2<sup>nd</sup> job set in J is 12. ‘-’ indicates the there is no assignment of the corresponding machine, worker and job.

## 5. CONCEPTS AND DEFINITION

### 5.1 Definition of a pattern

An indicator three-dimensional array which is associated with an assignment is called a ‘pattern’. A Pattern is said to be feasible if X is a solution. The pattern represented in the table-2 is a feasible pattern. Now  $T(X)$  the value of the pattern X is defined as

$$V(x) = \sum \sum D(i, j, k) X(i, j, k)$$

The value  $V(X)$  gives the total time of the assignment for the solution represented by X. Thus the value of the feasible pattern gives the total time represented by it. In the algorithm, which is developed in the sequel, a search is made for a feasible pattern with the least value. Each pattern of the solution X is represented by the set of ordered triples  $[(i,j,k)]$  for which  $X(i,j,k)=1$ , with understanding that the other  $X(i,j,k)$ ’s are zeros.

### 5.2 Feasible solution

Consider an ordered pair set{(1,2,1),(3,4,1),(4,3,3),(5,3,2),(4,1,4),(2,2,5),(3,1,2),(2,3,1),(4,3,4),(5,3,4)} represents the pattern given in the **table-2**, which is a feasible solution for the above numerical example.

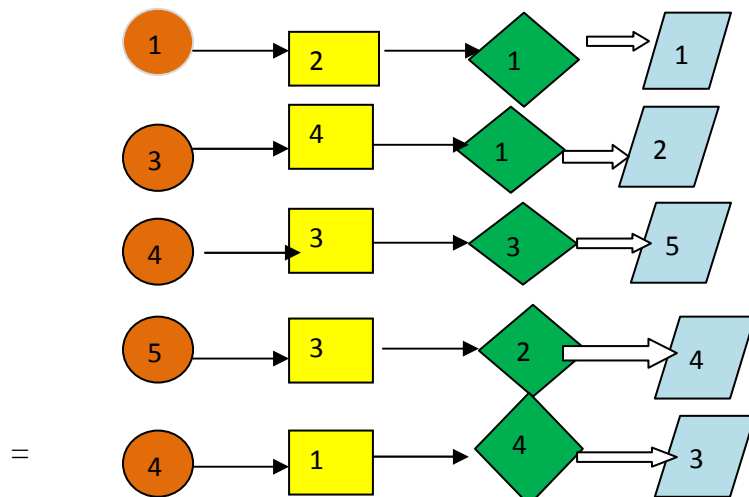
**Table-2**

$$\mathbf{X(i,j,1)} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad
 \mathbf{X(i,j,2)} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \quad
 \mathbf{X(i,j,3)} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

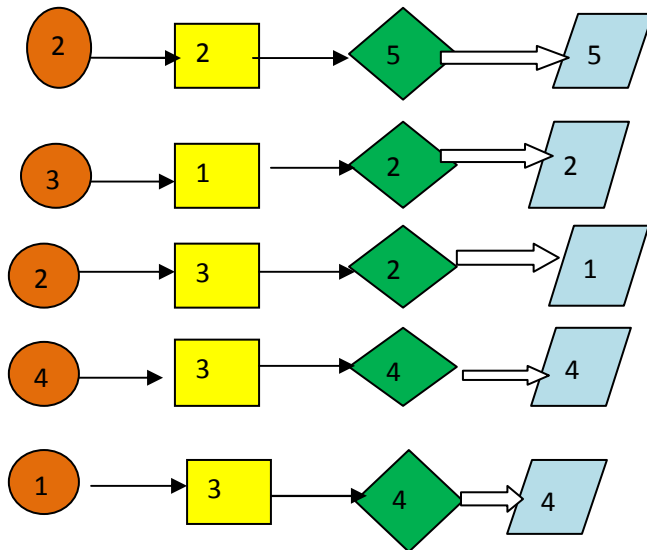
$$\mathbf{X(i,j,4)} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad
 \mathbf{X(i,j,5)} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

The following **figure-1** represents a feasible solution. The circle shapes represent machines, rectangle shapes represent workers, diamond shapes represent jobs and parallelogram shape represents the corresponding cost of machine, worker and job. The values in circles indicate name of the machine, values in rectangles indicates name of the worker and values in diamond shapes indicate name of the job.

**Figure-1**







According to the pattern represented in **figure-1** is satisfies all the constraints the section

3. The ordered tripled set represents the cost  $(1,2,1)=1,(3,4,1)=2,(4,3,3)=5,(5,3,2)=4,(4,1,4)=3,(2,2,5)=5,(3,1,2)=2,(2,3,2)=1,(4,3,4)=4,(5,3,4)=4$  . The total cost= $1+2+5+4+3+5+2+1+4+4=31$ .

### 5.3 Alphabet Table

There are  $M \times W \times J$  ordered triples in the three-dimensional array X. For convenience these are arranged in ascending order of their corresponding costs and are indexed from 1 to  $M \times W \times J$  (Sundara Murthy-1979). Let  $SN = [1, 2, 3, \dots, M \times W \times J]$  be the set of  $M \times W \times J$  indices. Let C be the corresponding array of costs. If  $a, b \in SN$  and  $a < b$  then  $C(a) \leq C(b)$ . Also let the arrays M, W, J be the array of indices of the ordered triples represented by SN and CC be the array of cumulative sum of the elements of C. For convenience same notation M, W, J are used for the corresponding array. The arrays SN, C, CC, M, W, and J for the numerical example are given in the table-3. If  $p \in SN$  then  $(M(p), W(p), J(p))$  is the ordered triple and  $C(a) = C(M(a), W(a), J(a))$  is the value of the ordered triple and  $CC(a) = \sum_{i=1}^a C(i)$ .

**TABLE-3**

SN	C	CC	M	W	J
1	1	1	1	2	1
2	1	2	1	4	1

3	1	3	2	3	2
4	1	4	1	1	3
5	1	5	1	2	5
6	2	7	1	3	1
7	2	9	2	1	1
8	2	11	3	4	1
9	2	13	3	1	2
10	2	15	3	4	2
11	2	17	3	2	3
12	2	19	5	4	5
13	3	22	1	1	1
14	3	25	2	2	1
15	3	28	2	3	4
16	3	31	4	2	2
17	3	34	3	4	3
18	3	37	1	1	4
19	3	40	4	1	4
20	3	43	2	4	5
21	4	47	3	1	1
22	4	51	2	1	2
23	4	55	5	3	2
24	4	59	5	2	3
25	4	63	1	3	4
26	4	67	5	3	4
27	4	71	2	1	5
28	4	75	4	3	5
29	5	80	3	3	1
30	5	85	2	4	2
31	5	90	4	3	3
32	5	95	5	3	3

33	5	100	3	3	4
34	5	105	2	2	5
35	5	110	4	1	5
36	6	116	4	4	1
37	6	122	1	2	2
38	6	128	4	1	3
39	6	134	5	4	4
40	6	140	1	4	5
41	7	147	4	2	1
42	7	154	1	4	2
43	7	161	2	2	3
44	7	168	3	1	4
45	7	175	5	1	5
46	8	183	4	1	1
47	8	191	3	3	2
48	8	199	2	4	4
49	8	207	4	4	5
50	9	216	5	2	1
51	10	226	5	3	1
52	11	237	4	1	2
53	11	248	2	1	4
54	11	259	2	3	5
55	12	271	4	3	2
56	12	283	5	4	2
57	12	295	3	2	4
58	13	308	2	2	2
59	13	321	5	1	2
60	13	334	1	3	3
61	13	347	5	1	4
62	14	361	2	1	3

63	14	375	2	3	3
64	14	389	5	2	4
65	15	404	3	3	3
66	15	419	3	3	5
67	16	435	4	4	4
68	17	452	3	1	5
69	17	469	4	2	5
70	18	487	1	2	3
71	18	505	3	2	5
72	19	524	2	4	3
73	20	544	4	2	3
74	20	564	2	3	1
75	21	585	4	3	4

Let us consider  $5 \in SN$ . It represents the ordered triple  $(M(5), W(5), J(5)) = (1, 2, 5)$ . Then  $C(5) = 1$  and  $CC(5) = 5$ .

**5.4. Definition of an Alphabet - Table and a word**

Let  $SN = (1, 2, \dots)$  be the set of indices,  $C$  be an array of corresponding costs of the ordered triples and  $CC$  be the array of cumulative sums of elements in  $C$ . Let arrays  $M$ ,  $W$  and  $J$  be respectively, the row, column and facility indices of the ordered triples. Let  $L_k = \{a_1, a_2, \dots, a_k\}$ ,  $a_i \in SN$  be an ordered sequence of  $k$  indices from  $SN$ . The pattern represented by the ordered triples whose indices are given by  $L_k$  is independent of the order of  $a_i$  in the sequence. Hence for uniqueness the indices are arranged in the increasing order such that  $a_i \leq a_{i+1}$ ,  $i = 1, 2, \dots, k-1$ . The set  $SN$  is defined as the "Alphabet-Table" with alphabetic order as  $(1, 2, \dots, M \times W \times J)$  and the ordered sequence  $L_k$  is defined as a "word" of length  $k$ . A word  $L_k$  is called a "sensible word". If  $a_i < a_{i+1}$ , for  $i = 1, 2, \dots, k-1$  and if this condition is called "insensible word". A word  $L_k$  is said to be feasible if the corresponding pattern  $X$  is feasible and same is with the case of infeasible and partial feasible pattern. A Partial word  $L_k$  is said to be feasible if the block of words represented by  $L_k$  has at least one feasible word or, equivalently the partial pattern represented by  $L_k$  should not have any inconsistency.

Any of the letters in SN can occupy the first place in the partial word  $L_k$ . Our interest is only in set of words of length at most  $n-1$ , since the words of length greater than  $n-1$  are necessarily infeasible, as any feasible pattern can have only  $n-1$  unit entries in it. If  $k < n$ ,  $L_k$  is called a partial word and if  $k = n$ , it is a full length word or simply a word. A partial word  $L_k$  represents, a block of words with  $L_k$  as a leader i.e. as its first  $k$  letters. A leader is said to be feasible, if the block of word, defined by it has at least one feasible word.

### 5.5. Value of the word

The value of the (partial) word  $L_k$ ,  $V(L_k)$  is defined recursively as  $V(L_k) = V(L_{k-1}) + TD(a_k)$  with  $V(L_0) = 0$  where  $TD(a_k)$  is the cost array arranged such that  $TD(a_k) < TD(a_{k+1})$ .  $V(L_k)$  and  $V(x)$  the values of the pattern  $X$  will be the same. Since  $X$  is the (partial) pattern represented  $L_k$  by Sundara Murthy – 1979.

### 5.6. Lower Bound of A partial word LB ( $L_k$ )

A lower bound  $LB(L_k)$  for the values of the block of words represented by  $L_k = (a_1, a_2, \dots, a_k)$  can be defined as follows.

$$LB(L_k) = V(L_k) + C(a_{k+n-k}) = V(L_k) + CC(a_{k+n-k}) - CC(a_k)$$

Consider the partial word  $L_4 = (3, 6, 9, 17) = V(L_4) = 1+2+2+3 = 08$

$$LB(L_4) = V(L_4) + DC(a_{4+n-k}) - DC(a_4)$$

$$= 08 + DC(17+8-4) - DC(17) = 08 + CC(21) - CC(17) = 08 + 47 - 34 = 21$$

Where  $CC(a_k) = \sum_{i=1}^k C(a_i)$ . It can be seen that  $LB(L_k)$  is the value of the complete word, which is obtained by concatenating the first  $(n-k)$  letters of SN ( $a_k$ ) to the partial word  $L_k$

### 5.7 Feasibility criterion of partial word

A recursive algorithm is developed for checking the feasibility of a partial word.  $L_{k+1} = (\alpha_1, \alpha_2, \dots, \alpha_k, \alpha_{k+1})$  given that  $L_k$  is a feasible partial word. We will introduce some more notations which will be useful in the sequel.

- IR be an array where  $IR(i) = 1, i \in M$  represents that  $i^{\text{th}}$  machine in  $M$  is assigned to  $j^{\text{th}}$  worker in  $W$  for doing the  $k^{\text{th}}$  job in  $J$  is 1, otherwise 0.
- IC be an array where  $IC(i) = 1, i \in W$  represents that  $i^{\text{th}}$  worker in  $W$  is assigned to  $j^{\text{th}}$  machine in  $M$  for doing the  $k^{\text{th}}$  job in  $J$  is 1, otherwise 0.
- IK be an array where  $IK(i) = 1, i \in J$  represents that  $i^{\text{th}}$  job in  $J$  is assigned to  $j^{\text{th}}$  machine in  $M$  for doing the  $k^{\text{th}}$  worker in  $W$  is 1, otherwise 0.
- L be an array where  $L(i)$  is the letter in  $i^{\text{th}}$  position of a partial word
- Im be an array where  $Im(i) = \alpha, i \in RA$  indicates that the machine is used  $\alpha$  times up to the  $i^{\text{th}}$  position of a partial word and  $\alpha \leq m_{RA}$ .
- Iw be an array where  $Iw(i) = \beta, i \in CA$  indicates that the worker is used  $\beta$  times up to the  $i^{\text{th}}$  position of a partial word and  $\beta \leq m_{CA}$ .
- Ij be an array where  $Ij(i) = \gamma, i \in KA$  indicates that the job is used  $\gamma$  times up to the  $i^{\text{th}}$  position of a partial word and  $\gamma \leq n_{CA}$ .
- IMX be an array where  $IMX(i) = 1, i \in M_s$  indicates that the  $M_s$  Machine is assigned.
- IWX be an array where  $IWX(i) = 1, i \in W_s$  indicates that the  $W_s$  Worker is assigned.
- IJX be an array where  $IJX(i) = 1, i \in J_s$  indicates that the  $J_s$  Worker is assigned.

Then for a given partial word  $L_k = (\alpha_1, \alpha_2, \dots, \alpha_k)$ , the values of the arrays IR, IC, IK, L, Im, Iw, Ij, IMX, IWX and IJX are as follows.

- $IR(R(\alpha_i)) = 1, i=1, 2, \dots, k$ , otherwise  $IR(j)=0$
- $IC(C(\alpha_i)) = 1, i=1, 2, \dots, k$ , otherwise  $IC(j)=0$
- $L(i) = \alpha_i, i=1, 2, \dots, k$ , otherwise  $L(j) = 0$ .
- $Im(i) = \alpha_i, i=1, 2, \dots, k$ , otherwise  $Im(j) = 0$ .
- $Iw(i) = \alpha_i, i=1, 2, \dots, k$ , otherwise  $Iw(j) = 0$ .
- $Ij(i) = \alpha_i, i=1, 2, \dots, k$ , otherwise  $Ij(j) = 0$ .
- $IMX(i) = 1, i=1, 2, \dots, k$ , otherwise  $IMX(j)=0$
- $IWX(i) = 1, i=1, 2, \dots, k$ , otherwise  $IWX(j)=0$
- $IJX(i) = 1, i=1, 2, \dots, k$ , otherwise  $IJX(j)=0$

The recursive algorithm for checking the feasibility of a partial word  $L_k$  is given as follows: In the algorithm first we equate  $IX=0$ . At the end If  $IX=1$  then the partial word is feasible, otherwise it is infeasible. For this algorithm we have  $RA=R(\alpha_k)$ ,  $CA=C(\alpha_k)$  and  $KA=K(\alpha_k)$ .

## 5.8: ALGORITHMS

### *Algorithm 1: (Checking the Feasibility)*

<i>Step 0: <math>IX=0</math></i>	<i>GOTO 2</i>
<i>Step 2: <math>MN=WN=JN = 0</math></i>	<i>GOTO 4</i>
<i>Step 4: <math>IS (IR (RA))+1 \leq m_{RA}</math></i>	<i>IF YES GOTO 6</i> <i>IF NO GOTO 20</i>
<i>Step 6: <math>IS (IC (CA))+1 \leq w_{CA}</math></i>	<i>IF YES GOTO 6</i> <i>IF NO GOTO 20</i>
<i>Step 8: <math>IS (IK (KA))+1 \leq n_{KA}</math></i>	<i>IF YES GOTO 10</i> <i>IF NO GOTO 20</i>
<i>Step 10: <math>IS (MX (RA)) = 0</math></i>	<i>IF YES, <math>MNA=MN+1</math> GOTO 12</i> <i>IF NO GOTO 12</i>
<i>Step 12: <math>IS (WX (CA)) = 0</math></i>	<i>IF YES, <math>WNB=WN+1</math> GOTO 14</i> <i>IF NO GOTO 14</i>
<i>Step 14: <math>IS (JX (KA)) = 0</math></i>	<i>IF YES, <math>JNC=JN+1</math> GOTO 16</i> <i>IF NO GOTO 16</i>
<i>Step 16: <math>Is N_0 - I \geq (p+q+r) - (MNA+ WNB+JNC)</math></i>	

*IF YES GOTO 18**IF NO GOTO 20**Step 18: IX=1**Step 20: STOP*

This recursive algorithm is used in Lexi search algorithm to check the feasibility of a partial word. We start the algorithm with a large value say ' $\infty$ ' as a trial value VT. If the value of a feasible word is known, we can as well start with that value as VT. During the search the value of VT is improved. At the end of the search the current value of VT gives the optimal feasible word. We start the partial word  $L_1 = (a_1) = (1)$ . A partial word  $L_k$  is constructed as  $L_k = L_{k-1} * (a_k)$  where \* indicates concatenation i.e. chain formation. We will calculate the values of  $V(L_k)$  and  $LB(L_k)$  simultaneously. Then two cases arise one for branching and the other for continuing the search.

1.  **$LB(L_k) < VT$ .** Then we check whether  $L_k$  is feasible or not. If it is feasible we proceed to consider a partial word of order  $(k+1)$ , which represents a sub block of the block of words represented by  $L_k$ . If  $L_k$  is not feasible then consider the next partial word of order by taking another letter which succeeds  $a_k$  in the  $k^{\text{th}}$  position. If all the words of order 'k' are exhausted then we consider the next partial word of order  $(k-1)$ .
2.  **$LB(L_k) \geq VT$ .** In this case we reject the partial word  $L_k$ . We reject the block of word with  $L_k$  as leader as not having optimum feasible solution and also reject all partial words of order 'k' that succeeds  $L_k$ .

Now we are in a position to develop a Lexi-Search algorithm to find an optimal feasible word.

### ***Algorithm -2 (Lexi - Search Algorithm (LSA))***

***Step 1: (Initialization)***



The arrays *SN, C, CC, M, W, J, m, w, p, q* and *r* are made available. *IR, IC, IK, IMX, IWX, IJX, L, V, N<sub>0</sub>* and *LB* are initialized to zero. The values  $I=1, J=0, VT=\infty, MAX$

Step 2:  $J=J+1$   
**IS ( $J>MAX$ )** **IF YES GOTO 11**  
**IF NO GOTO 3**

Step 3:  $L(I) = J$   
**IS ( $I=1$ )** **IF YES V(I) = C(J) GOTO 3B**  
**IF NO GOTO 3A**

Step 3A:  $V(I) = V(I-1) + C(J)$  **GOTO 3B**

Step 3B:  $LB(I) = V(I) + CC(J+N_0-I) - CC(J)$  **GOTO 4**

Step 4: **IS ( $LB(I) \geq VT$ )** **IF YES GOTO 11**  
**IF NO GOTO 5**

Step 5:  $RA=R(J)$   
 $CA=C(J)$   
 $KA=K(J)$  **GOTO 6**

Step 6: **Check the feasibility of *L* (using algorithm 1)**

**IS ( $IX=0$ )** **IF YES GOTO 2**  
**IF NO GOTO 7**

Step 7: **IS ( $IX=1$ )** **IF YES GOTO 8**  
**IF NO GOTO 2**

Step 8: **IS ( $I=N_0$ )** **IF YES GOTO 9**  
**IF NO GOTO 10**

Step 9:  $L(I) = J$

*L (I) is full length word and is feasible*

*VT=V (I), record L (I), VT. GOTO 13*

*Step 10: IR (RA)=1*

*IC (CA)=1*

*IK (KA)=1*

*MN= MNA*

*WN=WNB*

*JN=JNC*

*I=I+1*

*GOTO2*

*Step 11: IS (I=1)*

*IF YES GOTO 14*

*IF NO GOTO 12*

*Step 12: I=I -1*

*GOTO 13*

*Step 13: J=L(I)*

*RA=R(J)*

*CA=C(J)*

*IK (KA)=1*

*IR(RA)=0*

*IC(CA)=0*

*IR(RA)=IR(RA) -1*

*IS IR(RA)=0*

*IF NO GOTO 2*

*IF YES MN=MN-1 GOTO2*

*IC(CA)=IC(CA) -1*

*IF NO GOTO 2*

*IF YES WN=WN-1 GOTO2*

*IK(KA)=IK(KA) -1*

*IF NO GOTO 2*

*IF YES JN=JN-1 GOTO2*

**Step 14: STOP & END**

The current value of VT at the end of the search is the value of the optimal word. At the end if  $VT = \infty$ , it indicates that there is no feasible assignment.

**5.9 SEARCH TABLE**

The working details of getting an optimal word using the above algorithm for the illustrative numerical example is given in the **Table-4** The columns named (1), (2), (3),..., gives the letters in the first, second, third and so on places respectively. The columns R, C give the row, column indices of the letter. The last column gives the remarks regarding the acceptability of the partial words. In the following table A indicates ACCEPT and R indicates REJECT.

**Table-4 (Search Table)**

SN	1	2	3	4	5	6	7	8	V	LB	R	C	K	REMARK
1	1								1	11	1	2	1	A
2		2							2	11	1	4	1	A
3			3						3	11	2	3	2	A
4				4					4	11	1	1	3	A
5					5				5	11	1*	2	5	R
6					6				6	12	1*	3	1	R
7					7				6	12	2	1	1	A
8						8			8	12	3	4	1	A
9							9		10	12	3	1	2*	R
10							10		10	12	3	4	2*	R
11							11		10	12	3	2	3*	R
12							12		10	13	5	4	5	A
13								13	13	13	1	1	1*	R
14								14	13	13	2	2	1*	R
15								15	13	13	2*	3	4	R

16								16	13	13	4	2	2*	R		
17								17	13	13	3	4	3	R		
18								18	13	13	1*	1	4	R		
19								19	13	13	4*	1	4	R		
20								20	13	13	2	4	5*	R		
21								21	14	14	3	1	1*	R		
22								22	14	14	2	1	2*	R		
23								23	14	14	5	3	2*	R		
24								24	14	14	5	2	3*	R		
25								25	14	14	1*	3	4	R		
26								26	14	14	4	3	4	A, VT=14		
27							13		11	14*	1	1	1	R,=VT		
28						9			8	12	3	1	2	A		
29							10		10	12	3	4	2*	R		
30								11	10	12	3	2	3*	R		
31								12	10	13	5	4	5	A		
32									13	13	13	1	1	1	R	
33									14	13	13	2	2	1	R	
34									15	13	13	2*	3	4	R	
35									16	13	13	4	2	2*	R	
36									17	13	13	3	4	3*	R	
37									18	13	13	1*	1	4	R	
38									19	13	13	4	1	4	A,VT=13	
39								13		11	14*	1	1	1	R	
40							10			8	12	3	4	2	A	
41								11		10	12	3	2	3*	R	
42								12		10	13*	5	4	5	R,=VT	
43								11		10	13*	3	2	3	R,=VT	
44							8			6	12	3	4	1	A	
45								9			8	12	3	1	2	A

46						10		10	12	3	4	2*	R
47						11		10	12	3	2	3*	R
48						12		10	13*	5	4	5	R,=VT
49					10			8	12	3	4	2	A
50						11		10	12	3	2	3*	R
51						12		10	13*	5	4	5	R,=VT
52					11			8	13*	3	2	3	R,=VT
53				9				6	12	3	1	2	A
54					10			8	12	3	4	2	A
55						11		10	12	3	2	3*	R
56						12		10	13*	5	4	5	R,=VT
57					11			8	13*	3	2	3	R,=VT
58				10				6	13*	3	4	2	R,=VT
59			5					4	12	1	2	5	A
60				6				6	12	1*	3	1	A
61				7				6	12	2	1	1	A
62					8			8	12	3	4	1*	R
63					9			10	14*	3	1	2	R,>VT
64				8				6	12	3	4	1	A
65					9			8	12	3	1	2	A
66						10		10	12	3	4	2*	R
67						11		10	12	3	2	3*	R
68						12		10	13*	5	4	5	R,=VT
69					10			8	12	3	4	2	A
70						11		10	12	3	2	3	R
71						12		10	13*	5	4	5	R,=VT
72					11			8	13*	3	2	3	R,=VT
73				9				6	12	3	1	2	A
74					10			8	12	3	4	2*	R
75					11			8	13*	3	2	3	R,=VT

76				10				6	13*	3	4	2	R,=VT
77			6					5	13*	1	3	1	R,=VT
78		4						3	12	1	1	3	A
79			5					4	12	1*	2	5	R
80			6					5	13*	1	3	1	R,=VT
81		5						3	13*	1	2	5	R,=VT
82	3							2	12	2	3	2	A
83		4						3	12	1	1	3	A
84			5					4	12	1	2	5	A
85				6				6	12	1*	3	1	R
86				7				6	12	2	1	1	A
87					8			8	12	3	4	1	A
88						9		10	12	3	1	2	A
89							10	12	12	3	4	2*	R
90							11	12	12	3	2	3*	R
91							12	12	12	5	4	5*	R
92							13	13	13*	1	1	1	R
93						10		10	12	3	4	2	A
94							11	12	12	3	2	3*	R
95							12	12	12	5	4	3*	R
96							13	13	13*	1	1	1	R,=VT
97						11		10	12	3	2	3	A
98							12	12	12	5	4	5*	R
99							13	13	13*	1	1	1	R,=VT
100						12		10	13*	5	4	5	R,=VT
101					9			8	12	3	1	2	A
102						10		10	12	3	4	2	A
103							11	12	12	3	2	3*	R
104							12	12	12	5	4	5*	R
105							13	13	13*	1	1	1	R

106						11		10	12	3	2	3	A
107							12	10	12	5	4	5*	R
108							13	13	13*	1	1	1	R,=VT
109						12		10	13*	5	4	5	R,=VT
110					10			8	12	3	4	2	A
111						11		10	12	3*	2	3	R
112						12		10	13*	5	4	5	R,=VT
113					11			8	13*	3	2	3	R,=VT
114				8				6	12	3	4	1	A
115					9			8	12	3	1	2	A
116						10		10	12	3*	4	2	R
117						11		10	12	3*	2	3	R
118						12		10	13*	5	4	5	R,=VT
119					10			8	12	3	4	2	A
120						11		10	12	3*	2	3	R
121						12		10	13*	5	4	5	R,=VT
122					11			8	13*	3	2	3	R,=VT
123				9				6	12	3	1	2	A
124					10			8	12	3	4	2	A
125						11		10	12	3	2	3*	R
126						12		10	13*	5	4	5	R,=VT
127					11			8	13*	3	2	3	R,=VT
128				10				6	13*	3	4	2	R,=VT
129			6					5	13*	1	3	1	R,=VT
130		5						3	13*	1	2	5	R,=VT
131		4						2	13*	1	1	3	R,=VT
132	2							1	12	1	4	1	A
133		3						2	12	2	3	2	A
134			4					3	12	1	1	3	A

135				5				4	12	1	2	5	A
136					6			6	12	1*	3	1	R
137					7			6	12	2	1	1	A
138						8		8	12	3	4	1	A
139							9	10	12	3*	1	2	R
140							10	10	12	3*	4	2	R
141							11	10	12	3*	2	3	R
142							12	10	13*	5	4	5	R,=VT
143						9		8	12	3	1	2	A
144							10	10	12	3*	4	2	R
145							11	10	12	3*	2	3	R
146							12	10	13*	5	4	5	R,=VT
147						10		8	12	3	4	2	A
148							11	10	12	3*	2	3	R
149							12	10	13*	5	4	5	R,=VT
150						11		8	13*	3	2	3	R,=VT
151					8			6	12	3	4	1	A
152						9		8	12	3	1	2	A
153							10	10	12	3*	4	2	R
154							11	10	12	3*	2	3	R
155							12	10	13*	5	4	5	R,=VT
156						10		8	12	3	4*	2	R
157						11		8	13*	3	2	3	R,=VT
158					9			6	12	3	1	2	A
159						10		8	12	3	4	2	A
160							11	10	12	3*	2	3	R
161							12	10	13*	5	4	5	R,=VT
162						11		8	13*	3	2	3	R,=VT



163				10				6	13*	3	4	2	R,=VT
164			6					5	13*	1	3	1	R,=VT
165			5					3	13*	1	2	5	R,=VT
166		4						2	13*	1	1	3	R,=VT
167	3							1	13*	2	3	2	R,=VT

At the end of the search the current value of VT is  $(01+01+01+01+02+02+02+03) = 13$  and it is the value of the feasible word  $L_8 = (1, 2, 3, 4, 7, 9, 12, 19)$  it is given in 38<sup>th</sup> row of the search table – 4 and the corresponding order triples are  $(1, 2, 1), (1, 4, 1), (2, 3, 2), (1, 1, 3), (2, 1, 1), (3, 1, 2), (5, 4, 5), (4, 1, 4)$  For this optimal feasible word the arrays IR, IC, IK, Im, Iw, Ij, L,

MX, WX and JX are given in table – 5.

**Table – 5**

	1	2	3	4	5	6	7	8
<b>IR</b>	111	11	1	1	1	-	-	-
<b>IC</b>	1111	1	1	11	-	-	-	-
<b>IK</b>	111	11	1	1	1	-	-	--
<b>L</b>	1	2	3	4	7	9	12	19
<b>Im</b>	3	2	1	1	1	-	-	-
<b>Iw</b>	4	1	1	2	-	-	-	-
<b>Ij</b>	3	2	1	1	1	-	-	-
<b>MX</b>	1	1	1	1	1	-	-	-
<b>WX</b>	1	1	1	1	-	-	-	-
<b>JX</b>	1	1	1	1	1	-	-	-

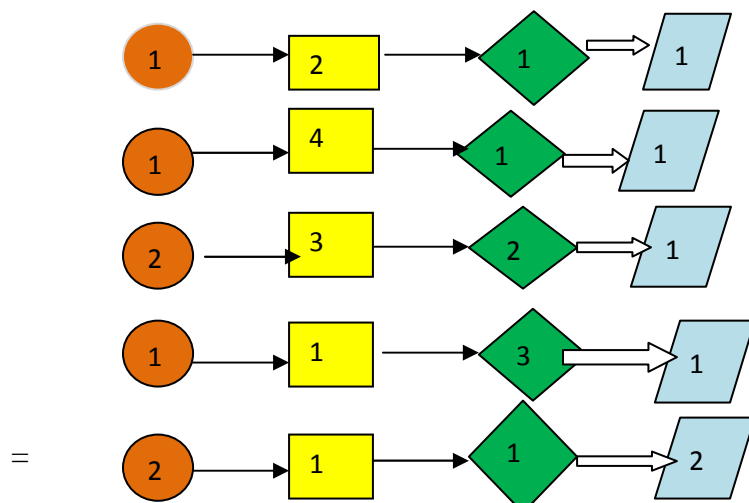
For an ordered triple set  $\{(1,2,1),(1,4,1),(2,3,2),(1,1,3),(2,1,1)(3,1,2),(5,4,5),(4,1,4)\}$  represents the pattern given in the **table-6**, which is a optimal solution for the above numerical example.

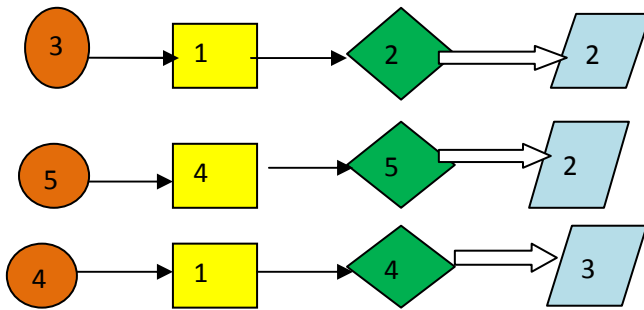
**Table-6**

$$\begin{aligned}
 \mathbf{X(i,j,1)} &= \begin{bmatrix} 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} & \mathbf{X(i,j,2)} &= \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} & \mathbf{X(i,j,3)} &= \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \\
 \mathbf{X(i,j,4)} &= \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} & \mathbf{X(i,j,5)} &= \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}
 \end{aligned}$$

The following **figure-2** represents a optimal solution. The circle shapes represent machines, rectangle shapes represent workers, diamond shapes represent jobs and parallelogram shape represents the corresponding cost of machine, worker and job. The values in circles indicate name of the machine, values in rectangles indicates name of the worker and values in diamond shapes indicate name of the job.

**Figure-2**





According to the pattern represented in **figure-2** is satisfies all the constraints the section 3. The ordered tripled set represents the cost  $(1,2,1)=1, (1,4,1)=1, (2,3,2)=1, (1,1,3)=1, (2,1,1)=2, (3,1,2)=2, (5,4,5)=2, (4,1,4)=3$ . The total cost= $1+1+1+1+2+2+2+3=13$ .

## 6. CONCLUSION

In this paper, we have studied a model namely “Constrained Three Dimensional Job Assignment Problem “. We have developed a new algorithm which is efficient, accurate and easy to understand. First the model is formulated in to a zero one programming problem. The problem is discussed in detail with help of numerical illustration.

## REFERENCES

- [1] Purusotham, S et.al. International Journal of Engineering Science and Technology (IJES Vol. 3 No. 8, 2011
- [2] Sobhan Babu. K et.al. International Journal on Computer Science and Engineering, (IJCSE) ,Vol. 02, No. 05, 2010, 1633-1640
- [3] Vidhyullatha ,A Thesis( Pattern Recognition Lexi-Search Exact Algorithms For Variant ASP and Bulk TP models), Department of mathematics, Sri Venkateswara University, Tirupati, 2011.
- [4] Kuhn, H.W, the Hungarian Method for the Assignment Problem, Navl Rec. Log.Qtly.2, 83-97, 1955.

- [5] Barr, R.S., Glover, F. & Klingman, D, the alternating basis algorithm for assignment Problem, Math. Prog., 13. 1-13, 1977.
- [6] Hung, M.S.& Rom,W.O. , Solving the Assignment problem by relaxation. Ops.Res.,18, 969-982, 1980.
- [7] Bertsekas, D.P., A New Algorithm for the Assignment problem, Math. Prog.21. 152-171, 1981.
- [8] Geetha, S. & Nair, K.P. ,A variation of the Assignment problem, Euro.J. Of Or, 68, 422-426, 1993.
- [9] Garfinkel,R.S., an improved algorithm for bottleneck assignment problem, operations research,18, 1717-1751, 1971.
- [10] Ravindran, A, Ramaswamy, V, on the bottleneck assignment problem, Journal of optimization theory and application 21, 451-458, 1988.
- [11] Bhatia, H.L, Time minimization assignment problem, Systems and cybernetics in management 6, 75-83, 1977.
- [12] Shalini Arora, Puri, M.C. A lexi search algorithm for a time minimization assignment problem 35(3), 193-213.,1997.
- [13] Balakrishna, U, Truncated time - dependent TSP M. Phil, Dissertation , S.V University, Tirupati, India.