

Higher Dimensional P - Path Route Minimum Distance Supply To The Destinations From The Central Station

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ABSTRACT

We take into consideration one of the non-linear polynomial combinatorial programming issues known as P-path route minimum distance supply to the destinations from the central station. The distance matrix $D(i, j, k)$ is given from the i th destination to the j th destination utilizing the k th facility, and there are n destinations. A facility, denoted by the letter K , can be an individual component that affects travel distances or costs. Depending on the needs of the many destinations, the central station has the potential to supply the load. The challenge is to determine the lowest distance, while taking the aforementioned factors into account, to connect all of the destinations to the central station (let's say one). We created a Pattern Recognition Technique based Lexi Search Method for this issue, and we built the suggested algorithm in C. We compared it to the models already in use and came to the conclusion that it was recommended for handling higher dimensional issues.

Keywords: Lexi search method, Pattern recognition technique, Facility

1. INTRODUCTION

Recently, there has been a lot of focus on the relevance of network advancements in the domain of computing and telecommunications. The design method aims, in part, to achieve maximum connectivity with minimal effort and expense. Some routes contain all of the connections (say P). Planning a road network or an integrated circuit has similar challenges. By imposing limits on the possible degrees a node might have, we can simulate the technological limitation that the number of connections at a node is limited. As Garey [2] shown, the resulting degree-constrained minimum is the lowest possible value. Here, we look into a subclass of Minimum spanning networks. We give computing results for a Lexi-algorithm we created using the "Pattern Recognition Technique" to handle the problem's elementary combinatorial structure.

Minimum Spanning Tree (MST) variants were the focus of the research of a few of the team members. In addition to Pop, et al, [6] and Karger [3], who discovered a randomized linear-time algorithm based on a hybrid of Boruvka's and the reverse-delete method, there are a few others worth mentioning. Chazelle [1] provides a deterministic linear solution to the problem. When the function grows extremely slowly, its duration is $o(m, (m, n))$. So, Chazelle's algorithm is nearly linear in execution time. Seth Pette [4,5] has discovered a minimum spanning tree methodology that relies on deterministic comparisons and is thus likely optimal. Different spanning models were investigated by Sobhan Babu [8] using a pattern recognition approach [9]. Let there be N nodes that need to reach out to the hub in destination 1. A facility K represents an individual factor that affects the distances and costs. If two destinations, j_1 and j_2 , are linked to i_1 , then k_1 and k_2 must be identical. Taking the opposite approach of [11], Suresh Babu [10] investigated a different kind of spanning model. The goal is to find the optimal solution for the minimum distance/cost using k facilities between all destinations connected to headquarters 1.

Problem Description

In this manuscript, we examine the “higher dimensional p – path route minimum distance supply to the destinations from the central station” problem. From the central station “1,” the P-trucks (say $p = 3$) deliver supplies to the other destinations, and the goal is to do so at the lowest possible distance. Each path cannot need more than 100 units of requirement, and there must be at least P destinations connected to the central station via P paths. Central station can accommodate a total of 300 units capacity. To illustrate, assume that $N=[1, 2, 3, 4, \dots, n]$ destinations have been given distances in the form $[NXN]$. Using a pattern recognition based lexical search approach, we solve the minimum spanning network connectivity problem in this manuscript. One of the requirements is that all N1 destinations use the same facility. For this manuscript, a precise algorithm is proposed. When applied to the problem of minimizing total travel distance and expenses, the algorithm provides a solution.

Let N denote the set of n stations with the parameters $N=1,2,3,4, \dots, n$ and K denote the set of k facilities with the parameters $1,2, \dots, q$. If i, j Type equation here. N and k K, then we may write the cost to go from destination i to destination j utilizing facility k as $D(i, j, k)$. Let $M=[1, 2, 3, \dots, m]$ be the cluster, and let M be a subset of N if and only if M contains at least m destinations. To illustrate, consider destination 1 to be the central station. The goal is to use P-paths to link the central station to the other (n-1) destinations. Every destination had direct or indirect ties to the central station 1. The P-trucks depart from P-paths in central station 1. The number of available P-trucks exceeds or meets the minimum demand of the destinations. The goal of this challenge is to determine the shortest route between all n destinations and the central station, 1. A appropriate numerical example for three different routes demonstrates our Lexi-Search algorithm, which is based on the pattern recognition technique.

Mathematical Formulation

$$\text{Minimize } Z(X) = \sum_{i \in N} \sum_{j \in N} \sum_{k \in K} D(i, j, k) X(i, j, k) \quad \dots (1)$$

Subject to the constraints

$$\sum_{k \in K} \sum_{j=2}^n X(1, j, k) = p (= 3) \quad \dots (2) \quad \sum_{i \in N} \sum_{j \in N} \sum_{k \in K} X(i, j, k) = n - 1$$

I(3)

Let $\alpha_{r1}, \alpha_{r2}, \alpha_{r3}, \dots, \alpha_{rn}$ are n_r destinations in r^{th} path ($r=1,2, \dots, P$) then

$$\sum_{s=1}^{nr-1} x(\alpha_{rs}, \alpha_{r,s+1}) = n_r - 1 \quad \dots (4)$$

Let $\beta_1, \beta_2 \in M$

If $X(\alpha_1, \beta_1, \gamma_1) = X(\alpha_2, \beta_2, \gamma_2) = 1$, Then

$$\gamma_1 = \gamma_2 \quad \dots (5) \quad \sum_{i=1}^p n_i = n - 1$$

$$1 \quad \dots (6),$$

$$\sum_{s=1}^{nr} Q(\alpha_{rs}) \leq V \quad (r=1, 2, \dots, P) \quad \dots (7)$$

$$\sum_{i \in N} Q(i) \leq PV \quad \dots (8)$$

Where $Q(1) = 0$

$$X(i, j, k) = 0 \text{ or } 1 \quad \dots (9)$$

Numerical Illustration

To give a concrete illustration of the theory and approach established, let's assume that out of a total of $N=1,2,3,4,5,6,7,8,9$ destinations, we'll designate destination 1 as the central station and group destinations 2,4,9 into a single cluster ($M=2,4,9$). It is easy to demonstrate that this is not a required condition when we take the following example in which $D(i, j, k)$ are all non-negative integers. Distance to link Destination 4 with Destination 3 via Facility 1 is shown to be 13 in Table 1. If two destinations are separated by a hyphen ('-'), it means they are not linked in any way. The following table illustrates the destinations' needs. Direct or indirect P-paths link all of the other destinations to the central station. The associated distance matrix is listed below.

Table-1

-	1	∞	2	7	10	∞	15	18
∞	-	21	∞	25	∞	28	∞	27
∞	23	-	13	28	∞	24	35	40
∞	4	30	-	∞	20	∞	31	2
∞	∞	∞	7	-	22	26	50	29
∞	20	7	5	25	-	10	16	∞
∞	∞	10	∞	20	16	-	12	17
∞	∞	15	6	28	∞	23	-	24
∞	28	12	∞	20	14	17	22	-

$D(i,j,1)=$

Table-2

-	14	3	22	∞	19	∞	∞	∞
∞	-	20	12	3	12	26	4	32
∞	11	-	18	15	6	4	7	19
∞	∞	29	-	∞	28	∞	18	9
∞	30	∞	∞	-	22	3	24	∞
∞	18	16	16	26	-	∞	24	22
∞	16	15	∞	20	16	-	14	∞
∞	13	∞	30	13	6	18	-	19
∞	24	8	16	∞	26	∞	4	-

$D(i,j,2) =$

We simplify matters by imagining that all destination data is stored in a B-tree of only two dimensions. Then, consider the array B that we provide below:

Table-3

1	2	3	4	5	6	7	8	9
0	1	0	1	0	0	0	0	1

Table 2's numerical example shows that destination I is part of cluster M if and only if $B(I_{ss}) = 1$. However, if $B(i) = 0$, then the condition is false. Assume that destination 2 is part of M-based cluster if and only if $B(2)=1$. Take into account the following matrix Q as table-3, which contains the needs of the destinations.

Table-4

1	2	3	4	5	6	7	8	9
-	30	30	20	30	60	40	30	40

From the above table-4, $QR(j) = \alpha$ means that the requirement of destination j is α . Suppose $Q(6) = 60$ means that the requirement of destination 6 is 60. Here $QR(1) = '-'$ means that the destination 1 act as central station so no need to requirement at 1 but it has the availability 300.

Concepts and Definitions

Definition of a pattern

In this context, the term “pattern” refers to a three-dimensional array of indicators that is linked to a certain task. If Pattern X has a solution (X is a solution), then Pattern is feasible. Specifically: $V(x) = D I j, k$. An expression of the form $X(i, j, k)$, where $i > N$, $j > N$, and $k > K$. The layout shown in table 4 is a workable one. The cost of connecting the destinations in X is given by the function $V(X)$. Because of this, the entire cost represented by the viable pattern is equal to its value. The method is built in the next chapter, and it looks for the cheapest possible pattern. Sets of ordered triples $I(j), (k)]$ for which $X(i), (j)$, and $(k)=1$ are representative of patterns in the solution X , with the other $X(i), (j)$, and (k) values being zeros.

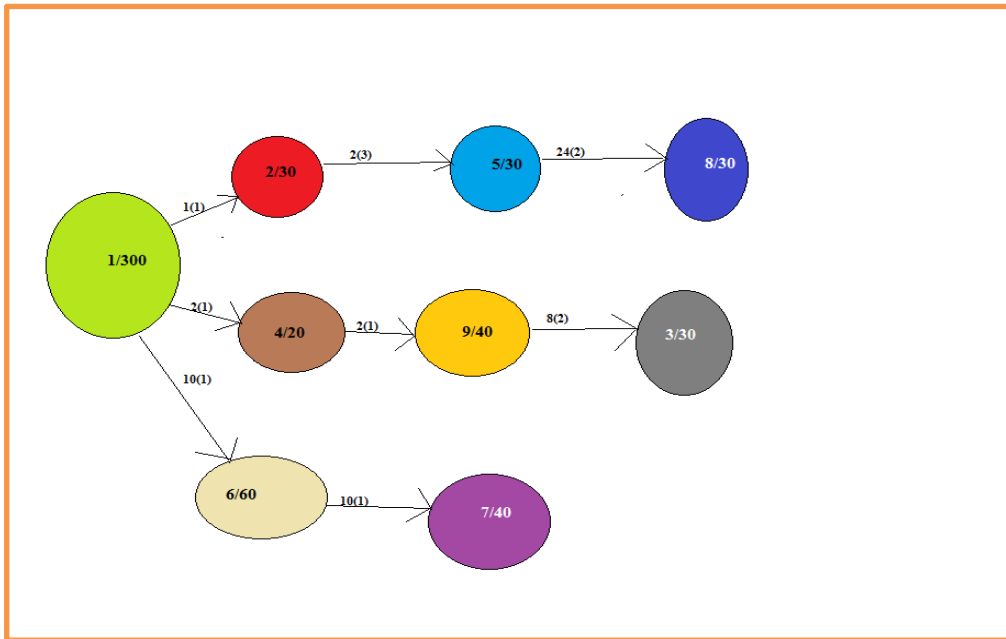
Feasible Solution

Consider an ordered triple set $\{(1,2,2),(2,5,1),(5,8,1),(1,4,2),(4,9,2),(9,3,1),(1,6,1),(6,7,1)\}$ Represents the pattern given in the table-5, which is a feasible solution.

Table- 5

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

In this context, the term “pattern” refers to a three-dimensional array of indicators that is linked to a certain task. If Pattern X has a solution (X is a solution), then Pattern is feasible. Specifically: $V(x) = D (i, j, k)$. An expression of the form $X(i, j, k)$, where $i > N$, $j > N$, and $k > K$. The layout shown in table 4 is a workable one. The cost of connecting the destinations in X is given by the function $V(X)$. Because of this, the entire cost represented by the viable pattern is equal to its value. Sets of ordered triples $I(j), (k)]$ for which $X(i), (j)$, and $(k)=1$ are representative of patterns in the solution X , with the other $X(i), (j)$, and (k) values being zeros.



From the above figure, three trucks are started from central station destination 1 with 100 units of capacity.. In first path the destination 1 is connected to destination 2 at facility 1 with 30 units, destination 2 is connected to destination 5 at facility 1, with 30 units and destination 5 is connected to destination 8 at facility 2, with 30 units. In second path destination 1 is connected to destination 4 at facility 1 with 20 units, destination 4 is connected to destination 9 with facility1 with 40 units; destination 9 is connected to destination 3 at facility 2 with 30 units. In third path destination 1 is connected to destination 6 at facility 1 with 60 units, destination 6 is connected to destination 7 at facility1 with 40 units. So that all the destinations in cluster M (2, 4,9) using the same facility 1 to connect from other destinations. Hence the solution is

$$Z = D(1,2,1) + D(2,5,2) + D(5,8,2) + D(1,4,1) + D(4,9,1) + D(9,3,2) + D(1,6,1) + D(6,7,1) = 1 + 2 + 24 + 2 + 2 + 8 + 10 + 10 = 59.$$

Alphabet Table:-

Hihger-dimensional array X contains minimum spanning ordered triples. For our use, they have been indexed from 1 to Z in ascending order of their corresponding distance. Let Z be the number of indices, and let S.No.=[1, 2, 3... Z]. Let D represent the associated cost matrix. To put it another way, if a and b are both S.No. and an is smaller than b, then D (a) is smaller than D (b). Let CD be an array containing the cumulative sum of the components of D, R, C, and F stand in for the row, column, and facility indices of the ordered triples represented by S.No., respectively. Table S-6 lists the numerical example's arrays SN, D, CD, R, C, and F If pSN, then the ordered triple is (R(p),C(p),F(p)) and the value of the ordered triple is D(a)=D(R(a),C(a),F(a)).

Table- 6

S.No.	D	CD	R	C	K
1	1	1	1	2	1
2	2	3	1	4	1
3	2	5	4	9	1
4	3	8	1	3	2
5	3	11	2	5	2
6	3	14	5	7	2

7	4	18	4	2	1
8	4	22	2	8	2
9	4	26	3	6	2
10	4	30	9	8	2
11	5	35	6	4	1
12	6	41	8	4	1
13	6	47	3	6	2
14	6	53	8	6	2
15	7	60	1	5	1
16	7	67	5	4	1
17	7	74	6	3	1
18	7	81	3	8	2
19	8	89	9	3	2
20	9	98	4	9	2
21	10	108	1	6	1
22	10	118	6	7	1
23	10	128	7	3	1
24	11	139	3	2	2
25	12	151	7	8	1
26	12	163	9	3	1
27	12	175	2	4	2
28	12	187	2	6	2
29	13	200	3	4	1
30	13	213	8	2	2
31	13	226	8	5	2
32	14	240	9	6	1
33	14	254	1	2	2
34	14	268	7	8	2
35	15	283	1	8	1
36	15	298	8	3	1
37	15	313	3	5	2
38	15	328	7	3	2
39	16	344	6	8	1
40	16	360	7	6	1
41	16	376	6	3	2

42	16	392	7	2	2
43	16	408	7	6	2
44	16	424	9	4	2
45	17	441	7	9	1
46	17	458	9	7	1
47	18	476	1	9	1
48	18	494	3	4	2
49	18	512	4	8	2
50	18	530	6	2	2
51	18	548	8	7	2
52	19	567	1	6	2
53	19	586	3	9	2
54	19	605	6	5	2
55	19	624	8	9	2
56	20	644	4	6	1
57	20	664	6	2	1
58	20	684	7	5	1
59	20	704	9	6	1
60	20	724	2	3	2
61	20	744	7	5	2
62	21	765	2	3	1
63	22	787	5	6	1
64	22	809	7	9	1
65	22	831	1	4	2
66	22	853	5	6	2
67	23	876	3	2	1
68	23	899	8	7	1
69	24	923	3	7	1
70	24	947	8	9	1
71	24	971	5	8	2
72	24	995	6	8	2
73	24	1019	9	2	2
74	25	1044	2	5	1
75	25	1069	6	5	1
76	26	1095	5	7	1

77	26	1121	2	7	2
78	26	1147	6	4	2
79	26	1173	9	6	2
80	27	1200	2	9	1
81	28	1228	2	7	1
82	28	1256	3	5	1
83	28	1284	8	5	1
84	28	1312	9	2	1
85	28	1340	4	6	2
86	29	1369	5	9	1
87	29	1398	4	3	2
88	30	1428	4	3	1
89	30	1458	5	2	2
90	30	1488	8	4	2
91	31	1519	4	8	1
92	32	1551	2	9	2
93	35	1586	3	8	1
94	40	1626	3	9	2

Let us consider 5th S.No. It represents the ordered triple (R(5),C(5),(5))=(2,5,2).Then D(5) = 3 and CD(5) = 11.

4.4 Definition of an Alphabet – Table and a word

D is an array of the corresponding costs of the ordered triples, and DC is an array of the cumulative sums of components in D, where SN = (1,2,...) is the set of indices. Let R, C, and k represent the row, column, and facility indices of the arrays containing the sorted triples, respectively. Let the series of k indices from SN be denoted by $L_k = a_1, a_2, \dots, a_k$, where a_i is an index from SN. The sequence order a_i has no bearing on the pattern depicted by the ordered triples whose indices are given by L_k . Therefore, the indices are organized in a way that guarantees uniqueness: $a_i \neq a_{i+1}$, where $i = 1, 2, \dots, k-1$. In this context, the ordered sequence L_k is a “word” of length k, and the set SN is the “Alphabet-Table,” where the indices range from 1 to n. A “sensible word” L_k is one that makes sense. An insensible word exists if and only if (for all i from 1 to k-1) a_i is less than a_{i+1} . If the related pattern X is feasible, then the word L_k is said to be feasible. The same holds true for infeasible and partially feasible patterns. If the block of words that L_k represents contains at least one feasible word, or if the partial pattern that L_k represents does not contain any inconsistencies, then we say that L_k is a viable partial word.

The initial position of the letters SN in the unfinished word L_k can be any of the letters in SN. Any conceivable pattern can only have n-1 unit entries, hence we’re only interested in the set of words up to and including n-1 characters in length. Whenever k is less than n, L_k is referred to as a short word, and whenever k equals n, it is referred to as a long word. An L_k -representing word fragment is a set of words that follows L_k as its leader letters (the first k). If the set of words that the leader specifies contains at least one possible word, we say that the leader is feasible.

4.5 Algorithm- : (Lexi – Search algorithm)

STEP 0: (Initialization)

The arrays SN, D, DC, R, C, T and LN the values of N, M are made available IR, IC, SW, SWI, F, IK, L, V, LB are initialized to zero. The values $I=1, J=0, LN(TR)=0, TV=0, MAX=(n*n)/2$

STEP I:	J=J+1 MAX= (n*n)/2LDP=0 IS (J>MAX)	IF NO GOTO II IF YES GOTO XXX
STEP II:	L (I) =J	
	TR= (J)TC= C (J)TK= T (J)	GOTO III
STEP III:	V (I) =V (I-1) +D (J) LB (I) =V (I) +CD (J+N-1-I)-CD (J))	GOTO IV
STEP IV:	IS (LB (I)>=VT)	IF NO GOTO V IF YES GOTO XXXIV
STEP V:	IS (TR= =HC)	IF YES GOTO VI
STEP VI:	IS (IR [TR] <P)	IF NO GOTO VII IF YES GOTO VIII
STEP VII:	IS (IR [TR] = =1)	IF NO GOTO II IF YES GOTO II
STEP VIII:	IS(IC [TC] = =1)	IF NO GOTO VIII IF YES GO TO II
STEP IX:	W=TC	IF NO GOTO IX GOTO X

STEP X:	LDP=LDP + q[w] IS (LD<LDP)	IF NO GOTO XII IF YES GOTO II
STEP XII:	IS (SW[w] = =0)	IF NO GOTO XIII IF YES GOTO XV
STEP XIII:	IS (W= =TR))	IF NO GOTO XIV IF YES GOTO II
STEP XIV:	W=SW [W]	GOTO X
STEP XV:	W=TR	GOTO XVI
STEP XIV: STEP XVII:	LDP=LDP + q[w]IS (LD<LDP) IS (SWI[w] = =0)	IF NO GOTO XVII IF YES GOTO II IF NO GOTO XVIII IF YES GOTO XX
STEP XVIII:	IS (W= =TC))	IF NO GOTO XIX IF YES GOTO II

STEPXXIX:	W=SWI [W]	GOTOXVI
STEPXX:	IS (b [TC] = =1)	IF NO GOTO XXV IF YES GOTOXXI
STEPXXI:	IS (NB= =0)	IF NO GOTO XXIII IF YES GOTOXXII
STEPXXII:	B=b [TC]	
	F [B] = TK	GOTOXXIV
STEPXXIII:	IS (F [B] = = TK)	IF NO GOTO II IF YES GOTOXXIV
STEPXXIV:	NB= NB+1	GOTO XXV
STEPXXV:	IS (i= =n-1)	IF NO GOTO XXVI IF YES GOTOXXVIII
STEPXXVI:	L [I] =J	
	IC [TC] =1 SW [TR] =TC SWI [TC] =TRIS (TR= =HC)	{IR [TR] =IR [TR] +1 GOTO XXVII}
		IR [TR] =1 GOTO XXVII}
STEPXXVII:	I=I+1	
	Z2=P-IR [HC] IS (Z2<=n-I)	IF YES GOTOII
		IF NO GOTO XXIX
STEPXXVIII:	T=V [I]	
	L [I] =J	GOTO XXIX
STEPXXIX:	I=I-1	GOTO XXX
STEPXXX:	J=L [I] TR=R [J]TC=C [J]	

	IR [TR] =0 IC [TC] =0 SW [TR] =0 SWI [TC] =0	
	L [I+1] =0 IS (TR= =HC)	IR [HC] = IR [HC]-1 GOTO XXXI}
STEP XXXI:	IS (b [TC] = =1)	IR [TR] =0 GOTO XXXI} IF YES GOTOXXXII
STEP XXXII:	NB=NB-1	IF NO GOTO 2 GOTO XXXIII

STEPXXXIII:	IS (I= =1)	IF YES GOTO XXXIV IF NO GOTO XXIX
STEP XXXIV:	STOP	

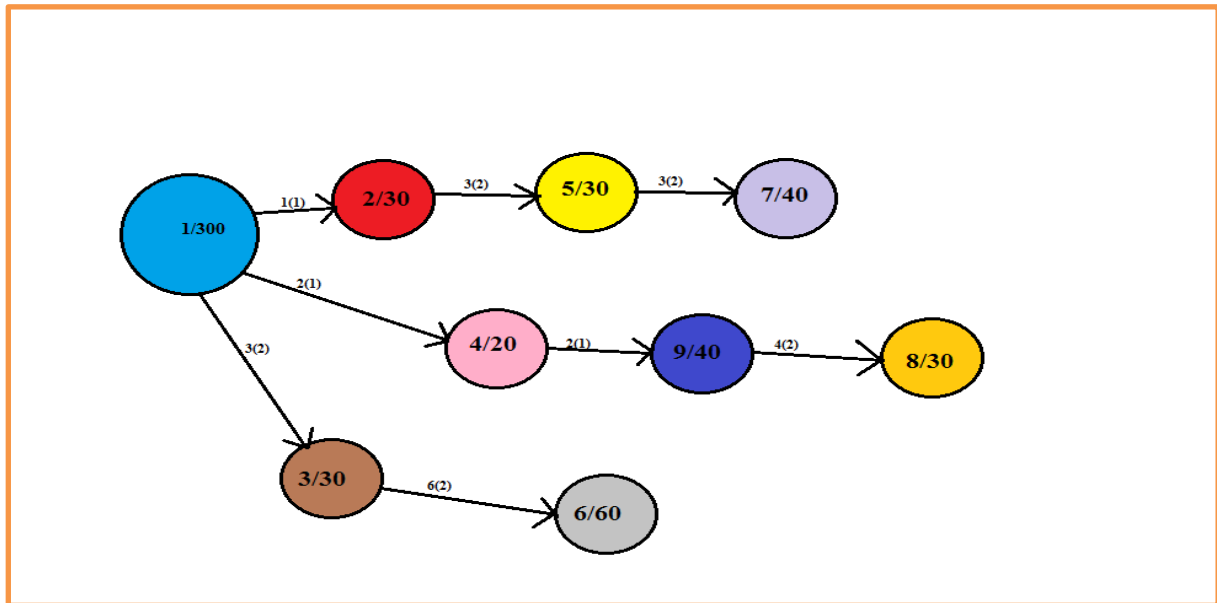
4.6 Search-Table:

The working details of getting an optimal word using the above algorithm for the illustrative numerical example is given in the Table-7. The columns named (1), (2), (3),..., gives the letters in the first, second, third and so on places respectively. The columns R, C and K give the row, column and facility indices of the letter. The last column gives the remarks regarding the acceptability of the partial words. In the following table -7, Y indicates ACCEPT, N indicates REJECT and TV is the feasible solution or optimal solution..

TABLE-7

SN	1	2	3	4	5	6	7	8	V	LB	R	C	K	REMARK
1	1								1	22	1	2	1	Y,(F=1)
2		2							3	22	1	4	1	Y (F=1)
3			3						5	22	4	9	1	Y,(F=1)
4				4					8	22	1	3	2	Y
5					5				11	22	2	5	2	Y
6						6			14	22	5	7	2	Y
7							7		18	22	4 [#]	2	1	N
8							8		18	22	2 [#]	8	2	N
9							9		18	22	3	6	2	Y
10								10	24	24	9	8	2	Y,TV=24
11							10		18	24	9	8	2	N, _≥ TV
12						7			15	24	4 [#]	2	1	N, _≥ TV
13					6				11	24	5	7	2	N, _≥ TV
14				5					8	24	2	5	2	N, _≥ TV
15			4						6	25	1	3	2	N, _≥ TV
16		3							3	26	4	9	1	N, _≥ TV
17	2								2	24	1	4	1	N

At the end of the search the current value of TV is 23 and it is the value of the optimal feasible word $L_8=(1, 2, 3, 4, 5, 6, 7, 9, 10)$. It is given in the 10th row of the search table. The optimal feasible word $L_8 = (1, 2, 3, 4, 5, 6, 7, 9, 10)$ and the respective ordered triple set is $\{(1,2,1),(1,4,1),(4,9,1),(1,3,2),(2,5,2),(5,7,2),(3,6,2),(9,8,2)\}$.The paths represented by the above pattern is $\{(1,2,1),(1,4,1),(4,9,1),(1,3,2),(2,5,2),(5,7,2),(3,6,2),(9,8,2)\}$ The destinations $\{2, 4, 9\}$ using the same facility 1. The diagrammatic representation of this solution can also see in the following figure-1.



The above figure-1 represents a optimal solution. In first path the central station 1 is connected to the destination 2 at facility 1 with 30units, destination 2 is connected to destination 5 at facility 2, with 30 units , destination 5 is connected to destination 7 at facility 2, with 40 units and the sum of the total supply to the destinations in first path is 100 units, which is less than or equal to truck capacity100. In second path, the central station 1 is connected to destination 4 at facility 1 with 20 units, destination 4 is connected to destination 9 with facility 1 with 40 units, and destination 9 is connected to destination 8 at facility 2 with 30 units and the sum of the total supply to the destinations in second path is 90 units(≤ 100), which is less than or equal to the truck capacity100. In third path central station is connected to destination 2 at facility 1 with 30 units, destination 2 is connected to destination 5 at facility2 with 30 units; destination 5 is connected to destination 7 at facility 2 with 40 units and the sum of the total supply to the destinations in first path is 80 units, which is equal to the truck capacity100. So, in each path sum of the supply to the destinations less than or equal to the truck capacity100 and all the destinations in cluster M (2, 4, 9) using the same facility 1 to connect from other destinations Hence the optimal solution is $Z = D(1,2,1) + D(2,5,2) + D(5,7,2) + D(1,4,1) + D(4,9,1) + D(9,8,2) + D(1,3,2) + D(3,6,2) = 1+3+3+2+2+4+3+6 = 24$

2. 5 EXPERIMENTAL RESULTS

Results obtained for the suggested algorithm, a pattern recognition tool built on the Lexi-search algorithm, are shown in the table below. We implement this technique as a C programme and test it on a COMPAQ dx2280 MT computer. Through testing with a range of input sizes, we validate this algorithm's effectiveness. We populate the distance matrix with arbitrary numbers. $D(i,j,k)$ in the distance matrix is assigned uniformly random numbers between 0 and 1000. Changing the parameters N, K, and CL allowed us to test a variety of scenarios. A tabular presentation of the findings (Table -10) is provided below. Seven to nine data sets are examined in each case. It can be observed that the time needed to find the best option is relatively short.

Table-10

SN	Problem dimension			TPN	CPU runtime in seconds	
	N	K	LC		Avg TA	Avg ST
1	4	2	1	7	0	0
2	4	2	2	7	0	0
3	5	2	1	7	0	0
4	5	3	1	7	0.054945	0
5	10	2	1	7	0.109890	0
6	10	3	1	7	0.21978	0

7	10	3	2	7	0.219780	0
8	15	2	1	9	0.384615	
9	15	3	1	9	0.659341	0.054945
10	15	3	2	9	0.329670	0.054945

In the above table the notation N represent number of destinations, K represent number of facilities, LC represent number of clusters to be taken, TPN represents number of problems tried. In the next columns Avg. TA represent average central processing unit run time to form an alphabet table. Avg. TS represents average central processing unit run time to form search table for obtaining the optimal solution. We observe that the time requirement of search table very less comparatively of time duration of alphabet table.

3. CONCLUSION

In this manuscript, we looked at the model " higher dimensional p - path route minimum distance supply to the destinations from the central station" and created a new algorithm based on a pattern recognition technique to find the optimal solution. Then, the model is called as a zero-one programming issue. Each and every step of the new algorithm and related concepts are discussed, along with a relevant numerical example. To implement the our proposed algorithm, we used the C programming language.. Since higher dimensional problems require less CPU time to run, the best possible answer can be found using this method. In addition to that, Lexi search techniques have been shown to be more effective in a wide variety of combinatorial issues. In numerous studies, researchers have demonstrated the effectiveness and speed of their own Lexi - search algorithms by employing various alphabet tables. From our observations, we conclude that this approach is quite efficient and can handle problems of much greater sizes.

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