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TEACHING MATERIAL ON



**MATHEMATICS
SCHOOL OF SCIENCE**

Dr. Dhrub Kumar Singh (Department Of Mathematics) ,School of Science YBN University , Ranchi

Intro.

Quadratic Programming:

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We studied LPP and simplex method (efficient method) 75

Unlike the LPP case, no such general algorithm exists for solving ~~all~~ all NLPP. However for problems with certain suitable structures, efficient algorithms have been developed. Also, it is often possible to convert the given non-linear programming problem into one in which these structures becomes visible.

The general mathematical programming Problem (GMPP) can be defined as the problem of finding $\bar{x} \in \mathbb{R}^n$ so that to

Optimize the Obj. $f(\bar{x})$

$$(a) \bar{x} = f(\bar{x})$$

s.t. constraints (b) $g_i(\bar{x}) (\leq, = \text{ or } \geq) b_i, i=1, \dots, m$

$$\text{and (c)} \bar{x} \geq 0,$$

Where $f(\bar{x})$ & $g_i(\bar{x})$ are the real valued functions of \bar{x} for $i=1, 2, \dots, m$ & b_i 's are constants.

It may be observed that the above GMPP reduces to general NLPP (GNLPP) if either $f(\bar{x})$ or $g_i(\bar{x})$ for some or all $i=1, \dots, m$ or $f(\bar{x})$ only

or $g_i(\bar{x})$ only for some or all $i=1, 2, \dots, m$ are non-linear in \bar{x} . Further, there

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funtions are known to be continuously differentiable.

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Optimal sols to a NLPP can be found anywhere on the boundary of the feasible region unlike LPP, but there is no such, as yet, any simplex method developed so far to find the optimal sols of a NLPP.

A well known quadratic programming model, dealing with the problems of selecting an investment portfolio that will yield a given expected total return with a minimum variance was developed by 'Markowitz'. The problem referred to as the portfolio selection model assumes that the investor wishes to maximize his anticipated return while he considers variance of return as undesirable.

Wolf's Method (Simple Method) 153 14

Let the quadratic programming problem be

$$\text{Max } z = f(\bar{x}) = \sum_{j=1}^n g_j x_j + \frac{1}{2} \sum_{j=1}^n \sum_{k=1}^n g_{kj} x_j x_k.$$

s.t. $\sum_{j=1}^n a_{ij} x_j \leq b_i, x_j \geq 0 \quad (i=1, \dots, m; j=1, \dots, n)$

where $g_{kj} = g_{kj} + g_{jk}$, $b_i > 0 \quad k=1, \dots, m$

Also we know that the quadratic form

$$\sum_{j=1}^n \sum_{k=1}^n g_{kj} x_j x_k$$
 be negative

semi-definite.

Then, the Wolf's iterative procedure may be obtained in the following ways.

Step 1: First, convert inequality constraints into equations, by introducing slack variables q_i^1 in the i^{th} constraint ($i=1, \dots, m$) and the slack variables r_j^2 in the j^{th} non-negativity constraint ($j=1, \dots, n$)

Step 2: Then construct the Lagrangian function

$$L(\bar{x}, \bar{q}, \bar{r}, \lambda, \mu) = f(\bar{x}) - \sum_{i=1}^m \lambda_i \left[\sum_{j=1}^n a_{ij} x_j - b_i + q_i^1 \right] - \sum_{j=1}^n \mu_j \left[-x_j + r_j^2 \right]$$

The only time necessary will be optimum solution for app. //

$$\text{When } \tau = (x_1, x_2, \dots, x_m), q = (q^1, q^2, \dots, q^m) \\ r = (r^1, r^2, \dots, r^m) \text{ & } z = (z_1, z_2, \dots, z_m) \\ M = (M_1, M_2, \dots, M_m).$$

Differentiating the above for 'L' partially w.r.t the components $x_1, x_2, x_3, \dots, x_m$, equating the first order partial derivative to zero, we derive Kuhn-Tucker conditions from the resulting equations.

- B. Wolfe (1959) suggested to introduce the non-negative artificial variables v_j , $j=1, 2, \dots, n$ in the Kuhn-Tucker conditions

$$g + \sum_{k=1}^n c_k x_k - \sum_{j=1}^m \lambda_j q_{kj} + \mu_j = 0$$

for $j=1, 2, \dots, n$ and constant.

and objective fn.

$$Z_v = v_1 + v_2 + \dots + v_n$$

Step-4, In this step, obtain the initial basic feasible solution to the following linear programming problem

$$\text{Min } Z_v = v_1 + v_2 + \dots + v_n.$$

Sit. 

$$\sum_{k=1}^m g_k x_k - \sum_{i=1}^m \lambda_i q_i + \mu_j + v_j = g_j \quad (j=1, \dots, n)$$

$$\sum_{j=1}^n q_j x_j + q_i^2 = b_i \quad (i=1, \dots, m)$$

$$x_j, \lambda_j, \mu_j, v_j \geq 0 \quad (j=1, \dots, n)$$

and satisfying the complementary slackness conditions.

$$\sum_{j=1}^n \mu_j x_j + \sum_{i=1}^m \lambda_i s_i = 0$$

(where $s_i = q_i^2$)

$$A_i s_i = 0 \quad \mu_j x_j = 0 \quad (\text{for } i=1, \dots, m \text{ and } j=1, \dots, n)$$

Step 5: Now, apply simplex method to find an optimal soln to the LPP constructed in step 4. The soln must satisfy the above complementary slackness condition.

Step 6: The optimum soln thus obtained in step 5 gives the optimum soln of given QPP.

and the following are connected with the 2010. In (ii) are connected

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Wolfe's modified simplex method:-

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Let the quadratic programming problem be

$$\text{Maximize } Z = f(\bar{x}) = \sum_{j=1}^m a_j x_j + \frac{1}{2} \sum_{j=1}^m \sum_{k=1}^n c_{jk} x_j x_k$$

Subject to the constraints $\sum_{j=1}^m a_{ij} x_j \leq b_i$, $x_j \geq 0$, ($i = 1, \dots, m$, $j = 1, \dots, n$)

where $c_{jk} = c_{x_j x_k}$ & a_{ij} and $b_i, a_{ij} \geq 0$ for all $i = 1, 2, \dots, m$.

We also assume that the quadratic form $\sum_{j=1}^m \sum_{k=1}^n c_{jk} x_j x_k$ be ~~non-negative~~ semi-definite.

Step-1 first convert the inequality constraints into equations by introducing slack-variables q_i^2 in the i th constraint ($i = 1, \dots, m$) and the slack variable r_j^2 in the j th non-negativity constraint ($j = 1, \dots, n$).

Step-2 Then we construct the Lagrangian function

$$L(\bar{x}, \bar{q}, \bar{r}, \bar{\lambda}, \bar{\mu}) = f(\bar{x}) - \sum_{i=1}^m \lambda_i \left[\sum_{j=1}^n a_{ij} x_j - b_i + q_i^2 \right] - \sum_{j=1}^n \mu_j \left[-x_j + r_j^2 \right]$$

When $\bar{x} = (x_1, \dots, x_n)$, $\bar{q} = (q_1, \dots, q_m)$, $\bar{r} = (r_1, \dots, r_n)$, $\bar{\lambda} = (\lambda_1, \dots, \lambda_m)$, $\bar{\mu} = (\mu_1, \dots, \mu_n)$

Differentiating the above function "L" partially w.r.t $\bar{x}, \bar{q}, \bar{r}, \bar{\lambda}, \bar{\mu}$, and equating the 1st order partial derivative to zero, we derive Kuhn-Tucker conditions from the resulting conditions.

Step-3. Wolfe (1959) suggested to introduce the non-negative artificial variable v_j , $j = 1, \dots, n$ in the Kuhn-Tucker condition:

$$a_j + \sum_{k=1}^n c_{jk} x_k - \sum_{i=1}^m \lambda_i a_{ij} + \mu_j = 0 \quad (j = 1, \dots, n)$$

and to construct an objective fn $Z_v = v_1 + v_2 + \dots + v_n$.

Step-4 In this step we obtain the initial basic solution to the following linear programming problem.

Max $Z_v = v_1 + v_2 + \dots + v_n$, subject to the constraints

$$\sum_{k=1}^n c_{jk} x_k - \sum_{i=1}^m \lambda_i a_{ij} + \mu_j + v_j = -a_j \quad (j = 1, \dots, n)$$

$$\sum_{j=1}^m a_{ij} x_j + q_i^2 = b_i \quad (i = 1, \dots, m)$$

$$v_j, \lambda_j, \mu_j, x_j \geq 0 \quad (j = 1, \dots, n), \quad (i = 1, \dots, m)$$

and satisfying the complementary slackness conditions

$$\sum_{j=1}^m \mu_j x_j + \sum_{i=1}^m \lambda_i q_i^2 = 0 \quad (\text{where } s_i = q_i^2), \quad \lambda_i q_i = 0 \quad (i = 1, \dots, m)$$

Step-5 - Now apply two phase simplex method to find optimum soln to this LPP. The soln may not satisfy the above Complementary Slackness Condition, which will be optimum soln for QPP. //

Example-1 Apply Wolfe method for solving the quadratic programming problem: — 158 (17)

Max $Z_x = 4x_1 + 6x_2 - 2x_1^2 - 2x_1x_2 - 2x_2^2$, subject to
 $x_1 + 2x_2 \leq 2$, and $x_1, x_2 \geq 0$.

Soln:-

Step 1 First we write all the constraint inequalities with \leq sign as follows: — $x_1 + 2x_2 \leq 2$, $-x_1 \leq 0$, $-x_2 \leq 0$.

Step 2. Now, introducing the slack variables $q_1^2, q_2^2, r_1^2, r_2^2$, our problem becomes of the form: —

$$\text{Max } Z_x = 4x_1 + 6x_2 - 2x_1^2 - 2x_1x_2 - 2x_2^2$$

$$\text{Subject to } x_1 + 2x_2 + q_1^2 = 2$$

$$-x_1 + r_1^2 = 0$$

$$-x_2 + r_2^2 = 0$$

Step 3. Here to obtain the Kuhn-Tucker conditions, we construct the Lagrange function

$$L(x_1, x_2, q_1^2, r_1^2, r_2^2, \lambda_1, \mu_1, \mu_2)$$

$$= (4x_1 + 6x_2 - 2x_1^2 - 2x_1x_2 - 2x_2^2) - \lambda_1(x_1 + 2x_2 + q_1^2 - 2) - \mu_1(-x_1 + r_1^2) - \mu_2(-x_2 + r_2^2)$$

The necessary and sufficient conditions are: —

$$\frac{\partial L}{\partial x_1} = 4 - 4x_1 - 2x_2 - \lambda_1 + \mu_1 = 0 \quad \frac{\partial L}{\partial x_2} = 6 - 2x_1 - 4x_2 - 2\lambda_1 + \mu_2 = 0$$

Defining $s_1 = q_1^2$, we have $\lambda_1 s_1 = 0$, $\mu_1 x_1 = 0$, $\mu_2 x_2 = 0$.

Also, $x_1 + 2x_2 + s_1 = 2$ and finally $x_1, x_2, s_1, \lambda_1, \mu_1, \mu_2 \geq 0$.

Ques. To construct the modified linear programming problem we will introduce the artificial variables v_1 and v_2 . The modified linear programming problem becomes: —

$$\text{Max } Z_v = -v_1 - v_2 \quad \text{subject to}$$

$$4x_1 + 2x_2 + \lambda_1 - \mu_1 + v_1 - v_2 = 4$$

$$2x_1 + 4x_2 + 2\lambda_1 - \mu_2 + v_2 - v_1 = 6$$

$$x_1 + 2x_2 + s_1 = 2$$

where all variables are non-negative and $\lambda_1 x_1 = 0$, $\mu_2 x_2 = 0$, $\lambda_1 s_1 = 0$. Now all these constraints equations can be written in matrix form as follows: —

Assignment Problem

Introduction:- It is a special type of linear programming problem in which the objective is to find the optimum allocation of a number of tasks (jobs) to an equal number of facilities (persons).

Here we have a general assumption that each person can perform each job but with varying degree of efficiency. For example a departmental head may have four persons available for assignment and four jobs to fill. Then this will be his interest to find out the best assignment in the interest of his department.

Matrix form (standard form) of Assignment problem:-

The assignment problem can be stated in the form of $n \times n$ matrix $[c_{ij}]$ called the cost or effective matrix and means that it is the cost of assigning i -th facility (person) to the j -th job, also called as effectiveness matrix.

: Effectiveness Matrix:

	Jobs			\dots					
	1	2	3	\dots	\dots	j	\dots	n	
Persons (Facilities)	1	c_{11}	c_{12}	c_{13}	\dots	\dots	c_{1j}	\dots	c_{1n}
	2	c_{21}	c_{22}	c_{23}	\dots	\dots	c_{2j}	\dots	c_{2n}
	3	\dots	\dots	\dots	\dots	\dots	\dots	\dots	\dots
	i	c_{i1}	c_{i2}	c_{i3}	\dots	\dots	c_{ij}	\dots	c_{in}
	n	c_{n1}	c_{n2}	c_{n3}	\dots	\dots	c_{nj}	\dots	c_{nn}

(13) (2)

Mathematical formulation of an Assignment Problem :-

Defn Minimize the total cost $Z = \sum_{i=1}^n \sum_{j=1}^m c_{ij} x_{ij}$ (60)

where, $x_{ij} = \begin{cases} 1 & \text{if } i\text{-th person is assigned to} \\ & \text{the } j\text{-th job.} \\ 0 & \text{if } i\text{-th person is not assigned to} \\ & \text{the } j\text{-th job.} \end{cases}$

Subject to the conditions that

(i) $\sum_{j=1}^m x_{ij} = 1, j=1, 2, \dots, n$, means only job is done by the i -th person, $i=1, 2, \dots, n$

(ii) $\sum_{i=1}^n x_{ij} = 1, i=1, 2, \dots, n$; means that only one person should be assigned to the j -th job, $j=1, 2, \dots, m$.

Important Theorems:-

Theorem 1 If in an assignment problem, a constant is added or subtracted to every element of a row (or column) of the cost matrix $[c_{ij}]$, then an assignment which minimizes the total cost for one matrix also minimizes the total cost for the other matrix.

Mathematically,

If $x'_{ij} = x_{ij}$, minimizes $Z = \sum_{i=1}^n \sum_{j=1}^m c_{ij} x_{ij}$ over all x_{ij} such that $\sum_{i=1}^n x_{ij} = 1 = \sum_{j=1}^m x_{ij}$ and $x_{ij} \geq 0$

then $x'_{ij} = x_{ij}$ also minimizes $Z' = \sum_{i=1}^n \sum_{j=1}^m c'_{ij} x_{ij}$ where $c'_{ij} = c_{ij} + a_i + b_j$, a_i, b_j are constants, $i=1, 2, \dots, n$; $j=1, 2, \dots, m$.

Theorem 2 : If all $c_{ij} \geq 0$ and if a solution $x'_{ij} = x_{ij}$ s.t $\sum_{i=1}^n \sum_{j=1}^m c_{ij} x_{ij} = 0$ then this solution is an optimal solution (i.e., this solution minimizes Z).

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Assignment Problem

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Q2

Assignment Problem is a particular or special type of Transportation problem in which a number of operations are to be assigned to an equal number of operators.

Hungarian Method (Reduced Matrix Method) :-

Step-1 Subtract the minimum elt of each row in the cost matrix $[C_{ij}]$ from every element of the corresponding row. (called row - operations)

Step-2 Do column operations as said in row operation in step-1. Here at this stage we will get row & column reduced matrix.

- Step-3
- (a) Now we examine rows successively until a row having exactly one zero is found. Mark (\square) at this zero, as an assignment made there. Mark (\times) at all other zeros in the column (in which we have marked \square) to indicate that they can't be used to make other assignments. We proceed in this way until the last row is examined.
 - (b) we adopt the same rule for examining the columns also as done above in case of rows.
 - (c) We continue these operations (a) and (b) successively until we reach to any of the two situations:
 - (i) all the rows are marked \square or \times or
 - (ii) the remaining unmarked zeros lies at least two in each row and column.

For case (i) we have a maximal assignment and still to improve.

(ii) still we have some zeros ..

Now there are two possibilities :-

- (i) If we have got an assignment in every row and every column i.e. total No. of $\square = n$ (the order of matrix)

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If such condition is reached we say that we have got the complete optimal assignment. (4)

(ii) If above (i) condition [no. of $\square \neq n$] then we will have to modify the cost (effectiveness/matrix) by adding or subtracting to create some more zeros in it. For this we proceed as in step 4 below.

Step-4: We draw minimum number of horizontal and vertical lines necessary to cover all zeros at least once. For this the following is adopted:—

- (i) Mark (\checkmark) all rows for which assignment have not been made.
- (ii) Mark (\checkmark) column which have zero in marked rows.
- (iii) Mark (\checkmark) rows (not already marked) which have assignment in marked columns.

(iv) Repeat Step (ii) and (iii) until the chain of marking ends.

(v) Draw minimum no. of lines through these marked rows and columns to cover all the zeros.

This procedure will yield the minimum number of lines (equal to the no. of assignments) in the maximal assignment obtained in Step 3 that will pass through all zeros.

Step-5 Now we select the smallest of the elements that do not have a line through them, subtract it from all the elements that do not have a line passing through them and add it to every element that lies at the \cap (intersection) of two lines and leave the remaining elements of the matrix unchanged.

Step-6: At the end of step-5, no. of zeros are increased (never decreased) in the matrix than that in Step-3 so we repeat Step 3 to the modified matrix obtain in step-5 to get the optimal assignment as above ways.

Example 1: Solve the following minimal assignment problem:- A department head has four subordinates and four tasks to be performed. The subordinates differ in efficiency and the tasks differ in their intrinsic difficulty. His estimate of the times each man would take to perform each task is given in the effectiveness matrix below. How should the task be allocated, one to a man, so as to minimize the total man hours?

Subordinates

	I	II	III	IV
A	8	26	17	11
B	13	28	4	26
C	38	19	18	15
D	19	26	24	10

Solution:- Step-1 We first find that whether it is balanced or unbalanced assignment problem. Obviously no. of rows = no. of columns, Hence it is a balanced Assignment problem. So we choose smallest element in each row and subtract them from every ~~other~~ element of the corresponding row. So we get the following matrix:-

	I	II	III	IV
A	0	18	9	3
B	9	24	0	22
C	23	4	3	0
D	9	16	14	0

Step 2 We subtract the smallest element in each column of the row-reduced matrix obtained in step-1, from every element of the corresponding column, we get the following column reduced matrix.

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Step 3. Now we test whether it is possible to make an assignment using the zeros by the method described in step-3 (Algorithm)

	I	II	III	IV
A	0	14	9	3
B	9	20	0	22
C	23	0	3	0
D	9	12	14	0

Starting with row I, we mark \square (ie, make an assignment) in the row containing only one zero and cross (\times) the zeros in the corresponding column in which \square lies. Thus, we have the following matrix as under:

Again starting with column I, we mark \square (ie, make assignment) in the column containing only one unmarked or uncrossed zero in the above table and cross out the zeros in the corresponding rows in which this assignment \square is marked. Thus we have the following table:

	I	II	III	IV
A	\square	14	9	3
B	9	20	\square	22
C	23	0	3	\times
D	9	12	14	\square

Let us now see that every row and every column have one assignment or the no. of $\square = \text{no. of order of the matrix}$. Thus we have the complete optimal zeros assignment.

	I	II	III	IV
A	\square	14	9	3
B	9	20	\square	22
C	23	\square	3	\times
D	9	12	14	\square

So we have the complete optimal zeros assignment.

\therefore Job I II III IV
Man 1 3 2 4

which is the optimal assignment in

Assignment Problem (Maximization Problem)

Hungarian Method :- Convert it into minimization problem by subtracting all the elements from the largest element. (Same Algo will be applied for Minimization problem).

The above process : → Phase-1 :- Reduction by Row and Column.

Phase-2 :- Optimizations.

Balanced Assignment Problem.

Q1: Solve the following assignment problems to maximize sales.

Assignment problem

Each source ~~can~~ should have the capacity to fulfill the demand of any of the.

Territories.

	I	II	III	IV
A	45	38	30	22
B	35	29	20	14
C	35	29	20	14
D	27	20	15	10

Q2. Solve the following assignment : How the task should be allocated to each person so as to minimize the total man hours?

	I	II	III	IV
1	8	26	17	11
2	13	28	4	26
3	38	19	18	15
4	19	26	24	10

Q3: There are two types of Assignment problems (1) Balanced and (2) Unbalanced. If there is equal no. of rows and columns then it is called balanced otherwise unbalanced Assignment problem. For Unbalanced Assignment problem we will have to add either

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'a dummy row ~~or~~ (65) dummy columns fitting each row ~~by~~ or column by zero element.'

S0/11 Hungarian Method:-

Step-1 Subtracting the smallest element from each row.

row reduce	I	II	III	IV
A	0	18	9	3
B	9	24	0	22
C	23	4	3	0
D	9	16	14	0

Since 0 is the minimum
so subtract from
each element of the
first row and zero
Similarly in each
row we get matrix again

Column reduction: Similarly we will do Column reduction

	I	II	III	IV
0	14	9	13	
9	29	0	22	
23	0	3	0	
9	12	14	0	

Reduced Matrix,

Here we have to see
which Subordinates has to
execute what tasks.
Next then let us Assign now wise
then column wise.

If in a row or column if we have single zero
& then we will have to make square (I)
ie, we will have to make square &

Cross if there is any row ~~having~~ single
zero. and in that column if there
~~is any all zeros~~ then we make a cross

cross (to all zeros) square means
Assignment has been made or completed

In first row we have single zero
so we may assign (I) and ~~in~~ in correspond-

~~if in their 1st column there is~~

will be any or all zeros then we

had to make cross to all but

there is no zero in ~~first~~ column

so we will not make zero cross anywhere

Assignment: ~~A-I~~, B-II, C-II, D-IV Total
 Man hours
 Implementation 8 167 4 19 10 = 41 hrs.

Note:- If ~~the~~ no of Assignment \neq the order of matrix then we will have to improve the matrix.

Ex 2 For that let us consider the following problem:-

	I	II	III	IV
A	18	26	17	11
B	13	28	14	26
C	38	19	18	15
D	19	26	24	10

Row reduction

7	15	6	0
0	15	1	13
23	4	3	0
9	16	14	0

Column Reduction

7	15	5	0
0	11	13	
23	0	2	
9	12	13	

Reduced Matrix

Row taken and wise assignment :-

Here we have only three Assignment
 w/ its order
 a) in each row/column we have not
 get at least one assignmt.
 So optimality is not reached
 & since 4th row and 3rd column
 do not have any assignment
 so it is not an optimal assignment.
 Therefore we improve the above
 matrix.

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In second row there is a single zero. (9)
 So we will make an assignment
 by putting square (1) here. Again
 there is no any zero in the remaining
 elment of the ^(completely) 2nd row. So we skip.

In third row we see that - there
 are more than a single zero so
 we skip.

In the next row we will see that - It a
 single zero. So we assign here and
 in this particular column there is
 one zero so we make a cross.

So we have completed row wise
 assignment.

Similarly we repeat column wise
 assignment.

In the 1st column & any single zero.
 In 2nd column, we assign a single
 zero in this column. We have to
 check there is any zero in this particular
 row & zeros. There was 2 zeros already
~~cross~~ crossed. Now we find the
 we had all zero which has
 been assigned already.

Condition of Optimality: — If ~~No. of non-zero~~
~~No. of Assignment (3)~~

= No. of order of
~~the matrix (4)~~

Hence Optimal Assignment has
 been made.
Or In each rows / columns there must be
 at least one assignment. Then it
 will be an optimal assignment.

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let us make the table again & 80

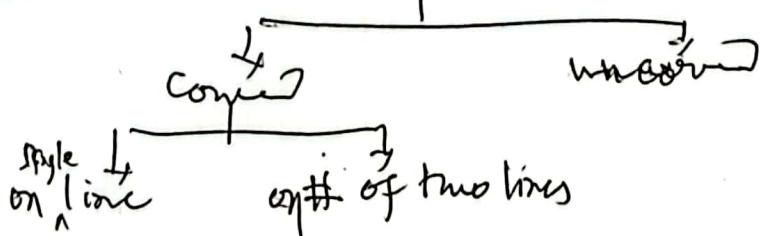
7	11	5	10
10	11	X	13
23	10	2	X
9	12	13	X

① we have to draw minimum no. of lines on Assign zero to that row.

let us mark ① that row in which no assignment(x) has been made (4th row). To find row in the 4th column which row has zero assignment earlier - yes if there is one zero so mark 2nd by the

Assigned zero (10). Then we have to see unselected row as that column in which we have ticked. we have to draw a line in which row there has not any tick so 2nd & 3rd row will have a line. Now we see that column in which we have ticked so the fourth column so that all zeros has been covered. Now after drawing the minimum no. of lines we found for the matrix those elements which has been covered or uncovered.

Covered & Elements.



Now we find the smallest element among uncovered elements which is 5.

No elements on line will remain same (as it is) as all uncovered elements must be subtracted by 5 and add on elements on it.

21	A	2	6	10	7	14
22	B	6	11	X	18	
23	C	10	2	5		
24	D	4	7	8	10	

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(170) Do Assignment again
as done initially

is the optimal Assignment

Opt Assignment $A = \text{III}$, $B = \text{I}$, $C = \text{II}$, $D = \text{IV}$ Total by

Man hours
check in
original
method.

$$30 + 35 + 28 + 10 = \underline{\underline{104}}$$

Assignment Problem (17)
Solv. problem of Transformation Problem.

85 (1)

- (1) Balanced (2) Unbalanced
 $\text{No. of rows} = \text{No. of columns}$ Convert by adding rows/columns

Hungarian Method - Algorithm

Phase-1 Rows & Column Reduction

- (1) ~~Solve using the Hungarian~~.

Step-1, Subtract the minimum value of each row from the entries of that row.

Step-II Subtract the minimum value of each column from the entries of that column.

Phase-2 Optimization of the Problem - Algo.

Step-1 - draw max no. of lines to cover all the zeros of the matrix.

- Q1. Solve the following assignment problem using Hungarian method. The matrix entries represent the processing times in hours. There are five operators & five jobs.

Balance Method.

	1	2	3	4	5	Row minimum.
1	9	11	14	11	7	7
2	6	15	13	13	10	6
3	12	13	6	8	8	6
4	11	9	10	12	9	9
5	7	12	14	10	-	7

If unbalanced then add dummy rows/column having 0 entries.

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

row minimum reduce matrix after step-I using 2nd step.

Showing Matrix.

n n n 2 0

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

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New matrix after column minimization.

and Please Optimization of the Problem.

Step-1. Draw minimum no. of lines to cover all the zeros of the matrix.

There are two procedures (a) row scanning.
(b) column scanning.

(a) row scanning:-

- (i) Solving from the first row, ask the following question. Is there exactly one zero in that row? If yes mark a square around that zero in that row and draw a vertical line passing through that zero; otherwise skip that row.
- (ii) After scanning the last row, check whether all the zeros are covered with lines. If yes, go to step-2; otherwise, do column scanning.

(b) Column scanning:-

- (i) Start from the first column, ask the following question: Is there exactly one zero in that column? If yes mark a square around that zero entry and draw a horizontal line passing through that zero; otherwise skip that column.
- (ii) After scanning the last column check whether all the zeros are covered with lines.

After rows & column reduction we get new matrix, then.

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

After row camp
go for column stamp
from 3rd column).

Step 10. (row...)

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Step-2 \rightarrow Check whether the no. of square marked is equal to the no. of rows of the matrin \rightarrow if yes, go to Step-5, else go to Step-3.

Step-3. Identify the minimum value of the undeleted cell values.

- Add the minimum undeleted cell value at the intersection points of the present matrin.
- Subtract the minimum undeleted cell value from all the undeleted cell values.
- All other entries remain same.

Step-4. Go to Step-1

Step-5. Treat the solution as marked by the square as the optimal solution.

Since no. of squares is $4 \neq 5$ rows. Optimally
 \rightarrow not reached
 \rightarrow so go to 3rd step.
 6, 7, 8 in 3rd row.

See Step-3.

- (1) \rightarrow add the minimum undeleted cell.
- (2) \rightarrow subtract it
- (3) \checkmark

	1	2	3	4	5
1	2	4	6	1	9
2	8	9	6	4	7
3	7	8	0	0	5
4	2	10	0	0	0
5	0	5	6	0	7

New matrin after step-3.
Go for Step-4.

Step-4. Go to Step-1

All the zeros are deleted.

Check no. of squares = no. of rows.
 $S = 5$ Hence the

sols is optimal

Job	1	2	3	4	5
1					
2					
3					
4					
5					

Job	operator	Time	operator	Time
1	5	7	6	8
2	1	5	3	6
3	3	6	4	9
4	2	9	7	11
5	4	7		not

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Transportation Problem:- It is a special kind of LPP in which the goods are transported from a set of sources to a set of destinations subject to the supply and demand of the sources and destinations respectively. Such that the total cost of transportation is minimised.

Type-I Balanced TP $S = D$

Type-II Unbalanced T.P. $S \neq D$

METHODS :- (I) Finding the initial basic feasible solution.
 (II) Finding optimisation.

		A	B	C	Supply
Source	I	2	7	5	200
	II	3	4	2	300
	III	5	4	7	500
Demn	200	400	400	1000	Balance $D = S$

I Phase Transportation problem.

- (1) Northwest corner cell method
- (2) Least cost cell method
- (3) Vary

(1) N.C.C. Method different source & diff destination.

		A	B	C	D	Supply	(2)
N.W corner	1	250	50	7	4	300 - 250 = 50	
	2	2	300	6	5	400 - 300 = 100	
3	8	3	3	3	2	500 - 300 = 200	
	250	350	400	200	1200	$S = D$ Balance.	
	= 300	= 300	-100	0	0		

Multiply as follow

$$(250 \times 3) + (50 \times 1) + (300 \times 6) + (100 \times 5) + (300 \times 3) + (200 \times 2) \\ = \text{Rs } 4400$$

II Least Cost Cell Method (176)

Least Cost among all various cell.

		Destination				Supply
		A	B	C	D	
Source	1	300	1	7	4	300.0
	2	250	50	100	5	400.150.80.0
	3	2	5	3	2	100.50.300.0
Demand	300	350	400	200	0	
	0	50	100	0		

$$(300 \times 1) + (250 \times 2) + (50 \times 6) + (100 \times 5) + (300 \times 3) + (200 \times 2) = 2900 \text{ Rs}$$

III Vogel's Approximation method. ① We need to calculate Row Diff. for each column.

		Destination				Supply	Row Diff. for each column			
		A	B	C	D		1	2	3	4
Source	I	3	1	7	4	300.0	2	3	-	-
	II	250	5	150	5	400 (150)	3	1	1	1
	III	2	5	3	2	100.50.300.200	1	1	1	0
Demand	300	350	400	200	0					
Column Diff.	1	2	2	2						
	2	2	2	2						
	3	2	1	1						
	(3)	2								

Total Pr. Cost with the allocated cell = $(200 \times 1) + (250 \times 2) + (150 \times 5) + (50 \times 3)$
 $+ (250 \times 3) + (200 \times 2) = 2850 \text{ Rs}$