#### **Advanced Engineering Mathematics**

## **Optimization Techniques III Sem CS ( 3CS2-01)**

#### Unit V

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#### <u>UNIT-II</u>

<u>LINEAR PROGRAMMING PROBLEM</u>

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## **INTRODUCTION AND DEFINATION**

- Mathematical programming is used to find the best or optimal solution to a problem that requires a decision or set of decisions about how best to use a set of limited resources to achieve a state goal of objectives.
- **Steps involved in mathematical programming**

- Conversion of stated problem into a mathematical model that abstracts all the essential elements of the problem.
- Exploration of different solutions of the problem.
- Finding out the most suitable or optimum solution.



**Linear programming** requires that all the mathematical functions in the model be linear functions.

• <u>The Linear Programming Model (1)</u>

**<u>Requirement:</u>** Maximization of the linear function Z.

 $Z = c_1 X_1 + c_2 X_2 + c_3 X_3 + \dots + c_n X_n \quad \dots Eq (1)$ subject to the following <u>constraints</u>:  $a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \le b_1$  $a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n \le b_2$  $\vdots$  $a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n \le b_n$ all  $x_1 \ge 0$  $\dots Eq (2)$ 

Where  $a_{ij}$ ,  $b_j$  and  $c_j$  are given constants.

#### The Linear Programming Model (2)

• The linear programming model can be written in more efficient notation as:



The decision variables  $X_1, X_2, \dots, X_n$  represent levels of n competing activities.

#### Applications of linear programming

- 1. Scheduling school buses to minimize total distance traveled
- 2. Allocating police patrol units to high crime areas in order to minimize response time to 911 calls
- 3. Scheduling tellers at banks so that needs are met during each hour of the day while minimizing the total cost of labor
- 4. Selecting the product mix in a factory to make best use of machine- and labor-hours available while maximizing the firm's profit
- 5. Picking blends of raw materials in feed mills to produce finished feed combinations at minimum costs
- 6. Determining the distribution system that will minimize total shipping cost

#### <u>Requirements of an LP problem</u>

- 1. LP problems seek to maximize or minimize some quantity (usually profit or cost) expressed as an objective function
- 2. The presence of restrictions, or constraints, limits the degree to which we can pursue our objective
- 3. There must be alternative courses of action to choose from
- 4. The objective and constraints in linear programming problems must be expressed in terms of linear equations or inequalities



## Formulating LP Problems

- The product-mix problem at Shader Electronics
  - ☑ Two products
    - 1. Shader X-pod, a portable music player
    - 2. Shader BlueBerry, an internet-connected color telephone
  - Determine the mix of products that will produce the maximum profit

Formulating L	P Problems	5	
	<u>Hours R</u>	lequired	
	<u>to Produ</u>	<u>ce 1 Unit</u>	
X-pods	BlueBerrys		Available Hours
Department	(X <sub>1</sub> )	(X <sub>2</sub> )	This Week
Electronic	4	3	240
Assembly	2	1	100
Profit per unit	\$7	\$5	

**Decision Variables:** 

- $X_1$  = number of X-pods to be produced
- $X_2$  = number of Blue Berrys to be produced

## Formulating LP Problems

• Objective Function:

Maximize Profit =  $7X_1 + 5X_2$ 

There are three types of constraints

- ☑ Upper limits where the amount used is ≤ the amount of a resource
- ☑ Lower limits where the amount used is ≥ the amount of the resource
- $\square$  Equalities where the amount used is = the amount of the resource

## Formulating LP Problems

• First Constraint:

Electronic time used is ≤ Electronic time available

 $4X_1 + 3X_2 \le 240$  (hours of electronic time)

#### **Second Constraint:**

Assembly is ≤ Assembly time used is ≤ Assembly

 $2X_1 + 1X_2 \le 100$  (hours of assembly time)

## Graphical method

Can be used when there are two decision variables

- 1. Plot the constraint equations at their limits by converting each equation to an equality
- 2. Identify the feasible solution space
- 3. Create an iso-profit line based on the objective function
- 4. Move this line outwards until the optimal point is identified



#### Craphical Calution

#### **Iso-Profit Line Solution Method**

Choose a possible value for the objective function

20

0

 $210 = 7X_1 + 5X_2$ 

Solve for the axis intercepts of the function and plot the line

40

$$X_2 = 42$$
  $X_1 = 30$ 

Number of X-pods

60

80

100







Corner point method



#### Corner-Point Method

- The optimal value will always be at a corner point
- Find the objective function value at each corner point and choose the one with the highest profit

Point (1) $(X_1 = 0, X_2 = 0)$ Point (2) $(X_1 = 0, X_2 = 80)$ Point (4) $(X_1 = 50, X_2 = 0)$ Point (3) $(X_1 = 30, X_2 = 40)$ 

Profit (0) + (0) = 0Profit (0) + (0) = 400Profit (0) + (0) = 400Profit (0) + (0) = 350Profit (0) + (0) = 410

## The Simplex Method

• The geometric method of solving linear programming problems presented before. The graphical method is useful only for problems involving two decision variables and relatively few problem constraints.

• We use an algebraic method called the simplex method, which was developed by George B. DANTZIG (1914-2005) in 1947 while on assignment with the U.S. Department of the air force.

#### Standard Maximization Problems in Standard Form

A linear programming problem is said to be a standard maximization problem in standard form if its mathematical model is of the following form:

Maximize the objective function

 $Z_{\max} = P = c_1 x_1 + c_2 x_2 + \dots + c_n x_n$ 

Subject to problem constraints of the form  $a_1x_1 + a_2x_2 + ... + a_nx_n \le b$ ,  $b \ge 0$ With non-negative constraints

 $x_1, x_2, \dots, x_n \ge 0$ 

# Slack Variables

- If a constraint has a sign  $\leq$ , then in order to make it an equality we have to add some positive quantity to the left hand side .
- The positive variables which are added to left hand side of the constraints to convert into equations are called the slack variables.
- **Ex**.  $x_1 + x_2 + x_3 \le 5$   $x_1, x_2, x_3 \ge 0$

Can be converted to

**Ex**.  $x_1 + x_2 + x_3 + x_4 = 5$   $x_1, x_2, x_3, x_4 \ge 0$ 

Then  $x_4$  is called the slack variable.

# Surplus variables

- If constraint has a sign ≥ then in order to make it an equality we have to subtract some positive quantity from left hand side.
- The positive variables which are subtracted from the left hand side of the constraints to convert them into equalities are called the surplus variables.
- **Ex.**  $2x_1 + 5x_2 \ge 200$
- Can be converted to

 $2x_1 + 5x_2 - x_3 = 200$ 

• then  $x_3$  is the surplus variable.

### **Example 1:** A Simple Maximization Problem



Working steps of simplex method

we follow the following steps:

- <u>Step-1</u> if the problem is not of maximization convert it to maximization problem.
- As Min.Z = Max.Z assume z = -z. Work on the problem with objective function as Max. Z
- <u>Step-2</u> if any entry of vector  $b = (b_1, b_2, \dots, b_n)^T$  is negative, multiply the corresponding constraint by (-1) to make it positive. (all entries of b should be non-negative.)
- <u>Step-3</u> add (subtract ) slack (surplus) variables, where ever needed to convert inequalities to equalities.
- <u>Step-4</u> if needed add artificial variables to get initial basic matrix  $B = I_m$

<u>Step-5</u> in objective function, slack and surplus variables are added with coefficients zero. The coefficient of artificial variables are decided by the method.

<u>Step-6</u> calculate  $z_j - c_j$  for every (column ) vector  $a_j$ . If the most negative value of  $z_j - c_j$  occur for j equal to k, then,  $k^{th}$  vector will be the entering vector in the new basis .

- $k^{th}$  vector will be the entering vector in the new basis . <u>Step-7</u> calculate mini. Ratio  $\begin{bmatrix} (X_B)_i \\ Y_{ik} \end{bmatrix}$ . If this value of i is r then  $\beta_r$  is the departing vector .
  - <u>Step-8</u> from the new simplex table. Repeat steps 6,7 and 8 until  $z_j - c_j \ge 0$  for all j is optimal.

### Example of simplex method

• Solve the following L.P.P by simplex method

Maximize  $Z = 5x_1 + 2x_2 + x_3$ 

subject to the constraints

 $x_1 + 3x_2 - x_3 \le 6$ ,  $x_2 + x_3 \le 4$ ,  $3x_1 + x_2 \le 7$ ,  $x_1, x_2, x_3 \ge 0$ . • <u>Step-1</u> covert into standard form

$$a_{11} x_{1} + a_{12} x_{2} + \dots + a_{1n} x_{n} \le b_{1}, \qquad a_{11} x_{1} + a_{12} x_{2} + \dots + a_{1n} x_{n} + x_{n+1} = b_{1}, \\ a_{21} x_{1} + a_{22} x_{2} + \dots + a_{2n} x_{n} \ge b_{2}, \qquad a_{21} x_{1} + a_{22} x_{2} + \dots + a_{2n} x_{n} - x_{n+2} = b_{2}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} \le b_{m}, \qquad a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{n+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots + a_{mn} x_{n} + x_{m+k} = b_{m}, \\ a_{m1} x_{1} + a_{m2} x_{2} + \dots +$$

Here  $x_4, x_5, x_6$  are slack variables

Simplex: Step 2 Step 2. Start with an initial basic feasible solution (b.f.s.) and set up the initial table. In our example

Maximize **Z** =  $5x_1 + 2x_2 + x_3$ 

$$x_1 + 3x_2 - x_3 + x_4 = 6,$$

 $x_2 + x_3 + x_5 = 4,$ 

 $3x_1 + x_2 + x_6 = 7$ ,

 $x_1, x_2, x_3, x_4, x_5, x_6 \ge 0.$ 

c <sub>B</sub>	Basis			(	cj			Constants
		5	2	1	0	0	0	
		<b>X</b> <sub>1</sub>	<b>X</b> <sub>2</sub>	X3	<b>X</b> 4	X5	X <sub>6</sub>	
0	X4	1	3	-1	1	0	0	6
0	X5	0	1	1	0	1	0	4
0	X6	3	1	0	0	0	1	7
$\bar{c}$ r	'OW	5	2	1	0	0	0	Z=0

#### Step 2: Explanation

#### Adjacent Basic Feasible Solution

If we bring a non basic variable  $x_s$  into the basis, our system changes from the basis,  $x_b$ , to the following (same notation as the book):

The new value of the objective function becomes:

$$Z = \sum_{i=1}^{m} c_i (\overline{b}_i - \overline{a}_{is}) + c_s$$

Thus the change in the value of Z per unit increase in  $x_s$  is

 $\overline{c}_s =$  new value of Z - old value of Z  $c_s - \sum_{i=1}^{m} c_i \overline{a}_{is}$  This is the Inner Product rule

## Simplex: Step 3

#### Use the inner product rule to find the relative profit coefficients



Simplex: Step 4

Apply the minimum ratio rule to determine the basic variable to leave the basis.

The new values of the basis variables:

 $x_{i} = b_{i} - \overline{a}_{is} x_{s} \qquad \text{for } i = 1, ..., m$  $\max x_{s} = \min_{\overline{a}_{is} > 0} \left[ \frac{\overline{b}_{i}}{\overline{a}_{is}} \right]$ 

#### In our example:

c <sub>B</sub>	Basis			(	Cj			Constants
		5	2	1	0	0	0	
		<b>X</b> 1	<b>X</b> <sub>2</sub>	<b>X</b> 3	<b>X</b> 4	<b>X</b> 5	<b>X</b> 6	
0	X4	1	3	-1	1	0	0	6
0	<b>X</b> 5	$\underline{0}$	1	1	0	1	0	4
0	X6	(3)	1	0	0	0	1	7
<i>c</i> r	OW	5	2	1	0	0	0	Z=0

Row	Basic Variable	Ratio
1	X <sub>4</sub>	6
2	X <sub>5</sub>	-
3	X <sub>6</sub>	7/3

# Simplex: Step 5

Perform the pivot operation to get the new tableau and the b.f.s.

$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	
$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	
$0   x_5   Q   1   1   0   1   0   4$	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	
$\bar{c} \text{ row}$ 5 2 1 0 0 0 Z=0	
New iteration:	
find entering	
variable: c <sub>B</sub> Basis c <sub>j</sub> Constan	nts
$\overline{\mathbf{c}}_{\mathbf{i}} = \mathbf{c}_{\mathbf{i}} - \mathbf{c}_{\mathbf{B}} \overline{\mathbf{P}}_{\mathbf{i}} \qquad \qquad \mathbf{x}_{1}  \mathbf{x}_{2}  \mathbf{x}_{3}  \mathbf{x}_{4}  \mathbf{x}_{5}  \mathbf{x}_{6}$	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	
$C_{-} = (0 \ 0 \ 5)$ $0 \ x_{5} \ 0 \ 1 \ 1 \ 0 \ 4$	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	
$c_2 = 2 - (0) 8/3 - (0) 1 - (5) 1/3 = 1/3$ $\overline{c} row$ $0 1/3 1 0 0 -5/3$ Z=35/3	3
$c_2 = 1 - (0) (-1) - (0) 1 - (5) 0 = 1$	
$c_6 = 0 - (0) 0 - (0) 0 - (5) 1/3 = -5/3$	

## Final Table



Basis			(	2 <sub>j</sub>			Constants
	5	2	1	0	0	0	
	<b>X</b> <sub>1</sub>	<b>X</b> <sub>2</sub>	X3	<b>X</b> 4	<b>X</b> 5	X6	
<b>X</b> 4	0	8/3	-1	1	0	0	11/3
<b>X</b> 5	0	1		0	1	0	4
<b>X</b> <sub>1</sub>	1	1/3	0	0	0	1/3	7/3
ow	0	1/3	1	0	0	-5/3	Z=35/3
	Basis X4 X5 X1 DW	Basis         5 $x_1$ $x_1$ $x_4$ $0$ $x_5$ $0$ $x_1$ $1$ $\infty$ $0$	Basis         5         2 $x_1$ $x_2$ $x_4$ 0 $8/3$ $x_5$ 0         1 $x_1$ 1 $1/3$ $xw$ 0 $1/3$	Basis $5$ $2$ $1$ $x_1$ $x_2$ $x_3$ $x_4$ $0$ $8/3$ $-1$ $x_5$ $0$ $1$ $1$ $x_1$ $1/3$ $0$ $\infty$ $0$ $1/3$ $1$	Basis $c_j$ 5         2         1         0 $x_1$ $x_2$ $x_3$ $x_4$ $x_4$ 0 $8/3$ -1         1 $x_5$ 0         1         1         0 $x_1$ 1 $1/3$ 0         0 $x_0$ 0 $1/3$ 1         0	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$

#### x<sub>3</sub> enters basis, x<sub>5</sub> leaves basis

	mmmmm	annan an a			~~~~~~		~~~~~~	
c <sub>B</sub>	Basis			Constants				
		5	2	1	0	0	0	
		<b>X</b> <sub>1</sub>	<b>X</b> <sub>2</sub>	X3	<b>X</b> 4	<b>X</b> 5	X6	
0	<b>X</b> 4	0	11/3	0	1	1	0	23/3
1	X3	0	1	1	0	1	0	4
5	<b>X</b> <sub>1</sub>	1	1/3	0	0	0	1/3	7/3
$\overline{c}$ r	ow	0	-2/3	0	0	-1	-5/3	Z=47/3

The optimal solution will be







Artificial Variable Technique (The Big-M Method and Two phase Method)

- <u>Artificial variables</u>:- the simplex algorithm requires a starting basic feasible solution (BFS). We find a starting BFS by using the slack variables as our basic variables. If an LPP has ≥ or equality constraints, a starting BFS may not be readily apparent. To start, we need a canonical from:
- (1) if we have a ≤ constraints with a non-negative right hand side, it will contain an obvious basic variable after introducing a slack variable.
- (2)if we have equality constraints, it contains no obvious variable.
- (3) if we have a ≥ constraints with a non-negative right hand side, it contains no obvious basic variable even after introducing a surplus variable.
• For example :

#### $2x + 3y \le 5 \Leftrightarrow 2x + 3y + s = 5, (s, basic)$

- If 2x + 3y = 5, this is infeasible if x=y=0. And  $2x + 3y \ge 5 \iff 2x + 3y - s = 5$ ,  $(s \ge 0)$  is infeasible if x=y=0.
- An artificial variable is a non-negative quantity added to the left side of exactly one constraint of an LPP in standard form. Enough artificial variables are added to the constraints so that the resulting coefficient matrix contains an *m×m* identity sub matrix. although they are initially basic variables when we apply the simplex method, artificial variables must be driven out of the basic so that their values are zero in the final solution.if this is not possible, then the problem is infeasible (i.e. its feasible region has no extreme points)

- For instance, if we have  $4x_1 + 3x_2 \ge 12$  then introducing surplus variable we have  $4x_1 + 3x_2 - s_1 = 12$  this equation will not contribute to BFS ,therefore, we need to introduce an artificial variable  $A_1$  as  $4x_1 + 3x_2 - s_1 + A_1 = 12$ .
- There are two standard methods for handling artificial variables within the simplex method:
- The Big-M Method
- The Two-Phase Method
- These methods complete the study of the simplex method and allow us to solve any L.P.P to find its optimal solution or discover whether it is unbounded or infeasible.

### Big –M Method

- Big –M Method is also called penalty method as for introducing the artificial variables , we have to pay penalties . Following are the steps for solving a problem having artificial variables using Big-M Method :
- (1) write the given LPP in standard form .
- (2) add an artificial variable  $A_i$  to the constraints identified as  $\geq$
- or = constraints. Also add the sign restriction  $A_i > 0$ .
- (3) if the LPP is a max problem , add (for each artificial variable) MA<sub>i</sub> to the objective function where M denotes a very large positive number.

- (4) solve the transformed problem by the simplex . Since each artificial variable will be in the starting basis, all artificial variables must be eliminated from Z- row . In choosing the entering variable, remember that M is a very large positive number.
- For example, 4M-2 is more positive than 3M+900 and -6M-5 is more negative than -5M-40 .
- (5) in the final simplex table observe that :
- (a) if all artificial variables are equal to zero in the optimal solution , we have found the optimal solution to the original problem.
- (b) if any artificial variables are positive in the optimal solution , the original problem is infeasible .

Example: Minimize Z= 600X1+500X2

subject to constraints,

2X<sub>1</sub>+ X<sub>2</sub> >or= 80

X<sub>1</sub>+2X<sub>2</sub> >or= 60 and X<sub>1</sub>,X<sub>2</sub> >or= 0

# **Step1: Convert the LP problem into a system of linear equations.**

We do this by rewriting the constraint inequalities as equations by subtracting new "surplus & artificial variables" and assigning them **zero & +M** coefficients respectively in the objective function as shown below.

So the Objective Function would be:

Z=600X1+500X2+0.S1+0.S2+MA1+MA2

subject to constraints,

 $2X_1 + X_2 - S_1 + A_1 = 80$ 

 $X_1+2X_2-S_2+A_2 = 60$ 

X1,X2,S1,S2,A1,A2 >or= 0

Step 2: Obtain a Basic Solution to the problem. We do this by putting the decision variables X1=X2=S1=S2=0, so that A1= 80 and A2=60. These are the initial values of *artificial variables*.

Step 3: Form the Initial Tableau as shown.

		Cj	600	500	0	0	М	М	
Св	Basic Variab Ie (B)	Basic Soln(Хв)	X1	X2	S1	S2	Aı	A2	Min.Ratio (Хв/Pivotal Col.)
Μ	A1	80	2	1	-1	0	1	0	80
М	A2	60	1	2☆	0	-1	0	1	60
		Zj	3M	3M	—M	— M	М	М	
		Cj - Zj	600-3M	500-3M	М	М	0	0	

It is clear from the tableau that X2 will enter and A2 will leave the basis. Hence 2 is the key element in pivotal column. Now , the new row operations are as follows:

R2(New) = R2(Old)/2R1(New) = R1(Old) - 1\*R2(New)

		Cj	600	500	0	0	М	
Св	Basic Variab Ie (B)	Basic Soln(Хв)	X1	X2	S1	S2	Aı	Min.Ratio (Хв/Pivota I Col.)
Μ	A1	50	3/2☆	0	-1	1/2	1	100/3 →
500	X2	30	1/2	1	0	- 1/2	0	60
Zj		3M/2+250	500	— M	M/2-250	М		
		Cj - Zj	350-3M/2↑	0	М	250-M/2	0	

It is clear from the tableau that X<sub>1</sub> will enter and A<sub>1</sub> will leave the basis. Hence 2 is the key element in pivotal column. Now,the new row operations are as follows:

 $R_1(New) = R_1(Old)^2/3$ 

 $R_2(New) = R_2(Old) - (1/2)*R_1(New)$ 

		Cj	600	500	0	0	Min
Св	Basic Varia ble (B)	Basic Soln(Хв)	X1	X2	S1	S2	Ratio (XB/P ivotal Col.)
600	X1	100/3	1	0	-2/3	1/3	
500	Х2	40/3	0	1	1/3	- 2/3	
		Zj	600	500	-700/3	- 400/ 3	
		Cj - Zj	0	0	700/3	400/3	

Since all the values of  $(C_j-Z_j)$  are either zero or positive and also both the artificial variables have been removed, an optimum solution has been arrived at with  $X_{1=100/3}$   $X_{2=40/3}$ 



### Two-Phase Method

As the name suggests it contains two phases:

<u>Phases-I</u> In this phase , we change the objective function . All the cost values , except for the artificial variables, are zero . Artificial variables are given the cost values of (-1) each and the objective function is maximized . Constraints are taken as they are given in the equation. The problem is solved by simplex method . Once we get rid of any artificial variable from the basis , we do not even write it in the simplex table. When all the artificial variables are removed from the basis , and  $z_j - c_j \ge 0 \forall j$  then the value of the new objective function will be zero. We end the phase-I.

<u>Phase –II</u> Starting table for this phase is the same as the last table of phase-I (i.e. we start with the feasible solution obtained in phase one). But the objective function is taken as the original given objective function in maximization form and we proceed to optimize it.

Example solve the following LPP by Two –Phase Method.

$$Min.z = x_1 + x_2$$
  
s.to  
$$2x_1 + x_2 \ge 4$$
  
$$x_1 + 7x_2 \ge 7$$
  
$$x_1, x_2 \ge 0$$

Solution: we first convert the minimization problem to maximization  $Max.z^* = -Min.z$ *s.to* 

$$2x_1 + x_2 - x_3 + x_5 = 4$$
$$x_1 + 7x_2 - x_4 + x_6 = 7$$
$$x_i \ge 0, i = 1to6$$

<u>Phase –I</u> The problem of phase –I is

$$Max.z^* = 0.x_1 + 0.x_2 + 0.x_3 + 0.x_4 - x_5 - x_6$$

Here  $x_5, x_6$  are artificial variables.

### Simplex table(1)

$\boldsymbol{C}_{j}$			0 0 0 0 -1 -1	
$C_{B}$	B.V	$X_{B}$	Y1 Y2 Y3 Y4 Y5 Y6	Mini ratio
-1	X5	4	21 -1 0 1 0	4/1
-1	X6	7	1 7) -1 O 1	7/7
	$Z_j - C_J$		-3 -8 1 1 0 0	

### Simplex table(2)



13/7 is the key element

### Simplex table(3)

	$C_{j}$		0	0	0	0	-1	-1
0	<b>X1</b>	21/13	10	-7	/13 1/	/13 7/:	13 -1/	13
0	X2	10/13	0	1	1/13	-14/9	1 -1/13	<b>3</b> 14/91
	$Z_j - C_J$		0	0	0	0	1	1

All  $Z_j - C_j \ge 0$  j =1 to 6. Max  $Z_1$  =0 and no artificial variable in the basis.

### Phase –II table(1)

	$C_{j}$		-1	-1	Ο	Ο
$C_{B}$	B.V	$X_{B}$	Yı	y2	У3	У4
-1	Xı	21/13	1	0	-7/13	1/13
-1	X2	10/13	0	1	1/13	-14/13
	$Z_j - C_J$		0	0	6/13	1/13

#### $Z_j - C_J \ge 0 \forall j = 1, 2, 3, 4$ . Thus the optimal solution is given by





Q.1solve the following L.P.P by simplex method.

 $Max.Z = 4x_1 + 5x_2$ 

s.to

 $x_1 + x_2 \le 3, 3x_1 + 4x_2 \le 10, all, x_1, x_2 \ge 0$ Q.2 solve the following L.P.P by big M- method.

$$Max.Z = x_1 + 2x_2 + 3x_3 - x_4$$
  
s.to  
$$x_1 + 2x_2 + 3x_3 = 15,$$
  
$$2x_1 + x_2 + 5x_3 = 20$$
  
$$x_1 + 2x_2 + x_3 + x_4 = 10$$
  
, all,  $x_1, x_2, x_3, x_4 \ge 0$ 



Q.3 use two phase method to solve the following L.P.P.  $Mini.Z = 10x_1 + 3x_2$ s.to  $x_1 + 2x_2 \ge 3,$   $x_1 + 4x_2 \ge 4,$   $all, x_1, x_2 \ge 0$ 

# Duality in linear programming

- 1. In physics and chemistry ,wave particle duality is a conceptualization that all objects in our universe exhibits properties of both waves and particles.
- 2. According to dictionary definition, dual means double. As applied to linear programming , duality implies that a double meaning can be attached to every linear programming problem. It can be thought of as a primal problem or as the dual problem. Unless we are told , it is not possible to determine which interpretation someone is placing on a given problem. Thus each LPP can be analyzed in two different ways (primal and its dual),but having equivalent solutions.

Each LPP (both maximization and minimization) stated in its original form has association with another LPP called **dual LPP** or in short **dual**, which is unique and is based on the same data. It is immaterial which one of the two is called as the **primal** and which one as **dual** as dual of the dual is primal: if the general L.P problem is taken as

 $Max.Z_{p} = cX$ s.to  $AX \le b$   $X \ge 0$  Eq.(I)

where  $c = (c_1, c_2, ..., c_n)_{1 \times n}$ 

$$X = \begin{bmatrix} x_{1} \\ x_{2} \\ \vdots \\ x_{n} \end{bmatrix}, A = \begin{bmatrix} a_{11}, a_{12} - \cdots - a_{1n} \\ a_{21}, a_{22} - \cdots - a_{2n} \\ \cdots & \cdots & \cdots \\ a_{m1}, a_{m2} - \cdots - a_{mn} \end{bmatrix} and$$
$$b = \begin{bmatrix} b_{1} \\ b_{2} \\ \vdots \\ \vdots \\ b_{m} \end{bmatrix}_{m \times 1}$$

Then attached to it, there exist another problem,

 $Min.Z_{D} = b^{T}W$ s.to  $A^{T}W \le c^{T}$  Eq.(II)  $W \ge 0$ 

Where T, stands for the transpose of the matrix/vector and

$$W = \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ \vdots \\ w_m \end{bmatrix}_{m \times 1}$$

if (I) represent the primal problem, then (II) is the dual . In fact it is immaterial, which problem is designated as the primal, since the dual of a dual is primal. If the optimal solution of one is known, the optimal solution to the other can be obtained from it (without solving). These two problem possess very closely related properties.

**Note:** if one of the constraints , in the primal , appear with an equality sign.(say  $i^{th}$ ), then the corresponding variable  $w_i$  will be unrestricted in sign .

i.e. if  $a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n = b_i$  then  $w_i$  is unrestricted in sign. We can give the following relationship between the primal and its dual.

### Relation between the primal and dual

Primal problem	Dual problem
1. Objective function maximize $Z_p = cX$	Minimize $Z_D = b^T W$
2. constraints: $AX \leq b$	$A^T W \ge c^T$
3. <i>i<sup>th</sup></i> Constraint is an inequality <i>i<sup>th</sup></i> constraint is an equation	$\frac{W_i \ge 0}{W_i}$ is unrestricted in sign
4. $X_i \ge 0$ $X_i$ is unrestricted in sign	<i>j<sup>th</sup></i> constraint is an inequality <i>j<sup>th</sup></i> constraint is an equation
5. Right hand side vector: <b></b> <i>b</i>	Right hand side vector : $C^{T}$
6. Cost coefficients	Cost coefficients $b^T$

**Note:** for every primal constraint there is a dual variable and for every dual constraint is a primal variable.

#### About the optimal solution.

1.Max  $Z_p = \min Z_D$ 

2.(i) if the primal problem has an unbounded solution , then the dual has no feasible solution .

(ii) if the primal solution has no solution then dual has either no solution or an unbounded solution .

Example: if primal has no solution then dual also has no solution Consider the primal problem as

> $Max.Z = 3x_1 + 4x_2$ s.to  $x_1 - x_2 \le -1$  $-x_1 + x_2 \le 0$  $x_1, x_2 \ge 0$





It is clear from the above example that <sup>|</sup> Primal and dual both can have no feasible solution Example write the dual of the following L.P.P

$$Max.Z_{P} = 2x_{1} + 4x_{2}$$
  
s.to  
$$2x_{1} + 3x_{2} \le 48$$
  
$$x_{1} + 3x_{2} \le 42$$
  
$$x_{1} + x_{2} \le 21$$
  
$$x_{1}, x_{2} \ge 0$$

Solution: it is a maximization problem with all constraints having  $\leq$  Sign .

$$Max.Z_{P} = 2x_{1} + 4x_{2}$$
  
s.to  
$$2x_{1} + 3x_{2} \le 48$$
  
$$x_{1} + 3x_{2} \le 42$$
  
$$x_{1} + x_{2} \le 21$$
  
$$x_{1}, x_{2} \ge 0$$

The dual is

 $\begin{aligned} &Min.Z_{D} = 48w_{1} + 42w_{2} + 21w_{3} \\ &s.to \\ &2w_{1} + w_{2} + w_{3} \geq 2 \\ &3w_{1} + w_{2} + w_{3} \geq 4 \\ &w_{1}, w_{2}, w_{3} \geq 0 \end{aligned}$ 



https://www.youtube.com/watch?v=zDch07vVBxl https://www.youtube.com/watch?v=8IRrgDoV8Eo https://www.youtube.com/watch?v=aw7NBSje2iM https://www.youtube.com/watch?v=MZ843Vvia0A https://www.youtube.com/watch?v=oZqVDGFckZE https://www.youtube.com/watch?v=UsqtzA9XmQE https://www.youtube.com/watch?v=gmDwUCvOJQ8

### Conclusion

Conclusion of this unit is that the term linear programming define A particular class of optimization problems in which the constraints of the system can be expressed as linear equations or inequalities and the objective function is a linear function of the design variables.

Linear programming (LP) techniques are widely used to solve a number of military, economic, industrial and societal problems. The primary reasons for its wide use are the availability of commercial software to solve very large problems and the easy with which data variation (sensitivity analysis) can be handled through LP models.



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## OPTIMIZATION TECHNIQUES III Sem CS (3CS2-01) UNIT-V

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### UNIT-III TRANSPROTATION AND ASSIGNMENT PROBLEM

**CONTENTS OF UNIT-III:-**

**INTRODUCTION** 

**SOLUTION OF TRANSPORTATION PROBLEM** 

**OPTIMALITY TEST** 

**ASSIGNMENT PROBLEM** 



### What is a Transportation Problem?

- The transportation problem is a special type of LPP where the objective is to minimize the cost of distributing a product from a number of sources or origins to a number of destinations.
- Because of its special structure the usual simplex method is not suitable for solving transportation problems. These problems require special method of solution.

### **Application of Transportation Problem**

- Minimize shipping costs
- Determine low cost location
- Find minimum cost production schedule
- Military distribution system
# Introduction and definition

- Distributing any commodity from any group of supply centers, called *sources*, to any group of receiving centers, called *destinations*, in such a way as to minimize the total distribution cost (shipping cost).
- Total supply must equal total demand.
- If total supply exceeds total demand, a *dummy destination*, whose demand equals the difference between the total supply and total demand is created. Similarly if total supply is less than total demand, a *dummy source* is created, whose supply
- equals the difference.
- All unit shipping costs into a dummy destination or out of a dummy source are 0.



# Example :1

		DESTINATIONS			Supply	
		D <sub>1</sub>	$D_2$	$D_3$	$D_4$	
	S <sub>1</sub>	50	75	30	45	12
Sources	$S_2$	65	80	40	60	17
	$S_3$	40	70	50	55	11
Demand		10	10	10	10	

**Transportation Table:** 



# Initial Solution of transportation problem

## • 1. North-west Corner rule

- 1. Select the *remaining* variable in the upper left (northwest) corner and note the supply remaining in the row, s, and the demand remaining in the column, d.
- 2. Allocate the minimum of s or d to this variable. If this minimum is s, eliminate all variables in its row from future consideration and reduce the demand in its column by s; if the minimum is d, eliminate all variables in the column from future consideration and reduce the supply in its row by d.
- REPEAT THESE STEPS UNTIL ALL SUPPLIES HAVE BEEN ALLOCATED.



**Total sipping cost = 2250** 

## 2. Lowest Cost entry method

- 1. For the remaining variable with the lowest unit cost, determine the remaining supply left in its row, s, and the remaining demand left in its column, d (break ties arbitrarily).
- 2. Allocate the minimum of s or d to this variable. If this minimum is s, eliminate all variables in its row from future consideration and reduce the demand in its column by s; if the minimum is d, eliminate all variables in the column from future consideration and reduce the supply in its row by d.
- REPEAT THESE STEPS UNTIL ALL SUPPLIES HAVE BEEN ALLOCATED.

Least Cost Cell	Remaining Supply in the Row (s)	Remaining Demand in the Column (d)	Allocate	Modifications
1. $X_{13}$ Cost = 30	12	10	$X_{13} = 10$	Eliminate column 3; reduce row 1 supply to 12 — 10 = 2
2. $X_{31}$ Cost = 40	11	10	$X_{31} = 10$	Eliminate column 1; reduce row 3 supply to 11 – 10 = 1
3. $X_{14}$ Cost = 45	2	10	X <sub>14</sub> = 2	Eliminate row 1; reduce column 4 demand to 10 – 2 = 8
4. $X_{34}$ Cost = 55	1	8	X <sub>34</sub> = 1	Eliminate row 3; reduce column 4 demand to 8 — 1 = 7



Total sipping cost = 2065

## • <u>3. Vogel's Approximation Method (VAM)</u>

- 1. For each remaining row and column, determine the difference between the lowest two *remaining* costs; these are called the *row* and column penalties.
- 2. Select the row or column with the largest penalty found in step 1 and note the supply remaining for its row, s, and the demand remaining in its column, d.
- 3. Allocate the minimum of s or d to the variable in the selected row or column with the lowest remaining unit cost. If this minimum is s, eliminate all variables in its row from future consideration and reduce the demand in its column by s; if the minimum is d, eliminate all variables in the column from future consideration and reduce the supply in its row by d.
- REPEAT THESE STEPS UNTIL ALL SUPPLIES HAVE BEEN ALLOCATED.



**Total sipping cost = 2030** 



**Optimality Test on IBFS Methods : 1. Stepping Stone Method** 2. MODI Method **MODI Method Of Checking Optimality Condition To Be Satisfied :** Filled Cells = m + n - 1Where m = No. of Rowsn = No. of columns

# The Stepping-Stone Method

 1. Find the current C<sub>ij</sub>-Z<sub>ij</sub> values for each nonbasic variable and select the one with the most negative C<sub>ij</sub>-Z<sub>ij</sub> value as the entering variable; if all C<sub>ij</sub>-Z<sub>ij</sub> values are nonnegative, the current solution is optimal.

• 2. Determine which basic variable reaches 0 first when the entering variable is increased.

• 3. Determine a new basic solution and repeat the steps.

# Step 1: Determine the C<sub>ij</sub>–Z<sub>ij</sub> Values for the Nonbasic Variables

If U<sub>i</sub> is the dual variable associated with the i-th supply constraint, and V<sub>j</sub> is the dual variable associated with the j-th demand constraint, then for shipments from node i to node j, one can find the corresponding Z<sub>ij</sub> value by Z<sub>ij</sub> = U<sub>i</sub> - V<sub>j</sub>. Thus the C<sub>ij</sub>-Z<sub>ij</sub> value for variable X<sub>ij</sub> is found by

• 
$$C_{ij} - Z_{ij} = C_{ij} - (U_i - V_j) = C_{ij} - U_i + V_j$$

2. Given that there is a redundant equation among the m n constraints (and any of the m n constraints can be considered the redundant one), one can show that the U<sub>i</sub> or V<sub>j</sub> associated with the redundant equation is 0. Thus one U<sub>i</sub> or V<sub>j</sub> can arbitrarily be selected and set to 0. Arbitrarily choose U<sub>1</sub> = 0.

3. Since the C<sub>ij</sub>-Z<sub>ij</sub> values for *basic* variables are 0 (i.e., C<sub>ij</sub> - U<sub>i</sub> + V<sub>j</sub> = 0 for basic variables), we can easily solve for the remaining values of the U<sub>i</sub>'s and V<sub>j</sub>'s from the m + n - 1 equations for the basic variables.

4. Once the U<sub>i</sub>'s and V<sub>j</sub>'s have been determined, the C<sub>ij</sub>-Z<sub>ij</sub> values for the nonbasic variables can be calculated by

• 
$$C_{ij} - Z_{ij} = C_{ij} - U_i + V_j$$

Basic Cell	Basic $C_{ij}$ - $Z_{ij}$ Values = 0	Substitute	Implies
X <sub>12</sub>	$C_{12} - U_1 + V_2 = 0$	$75 - 0 + V_2 = 0$	$V_2 = -75$
X <sub>14</sub>	$C_{14} - U_1 + V_4 = 0$	$45 - 0 + V_4 = 0$	$V_4 = -45$
X <sub>22</sub>	$C_{22} - U_2 + V_2 = 0$	$80 - U_2 + (-75) = 0$	$U_2 = 5$
X <sub>23</sub>	$C_{23} - U_2 + V_3 = 0$	$40 - 5 + V_3 = 0$	$V_3 = -35$
X <sub>32</sub>	$C_{32} - U_3 + V_2 = 0$	$70 - U_3 + (-75) = 0$	$U_3 = -5$
X <sub>31</sub>	$C_{31} - U_3 + V_1 = 0$	$40 - (-5) + V_1 = 0$	$V_1 = -45$

\_\_\_\_

# Non-basic cells:

Variable Cell

 $C_{ij}$ - $Z_{ij}$  Calculation

X <sub>11</sub>	$C_{11} - U_1 + V_1 = 50 - 000$	0 - 45 = 5
X <sub>13</sub>	$C_{13} - U_1 + V_3 = 30 - 100$	0 - 35 = -5
X <sub>21</sub>	$C_{21} - U_2 + V_1 = 65 - 000$	5 - 45 = 15
X <sub>24</sub>	$C_{24} - U_2 + V_4 = 60 - 000$	5 - 45 = 10
X <sub>33</sub>	$C_{33} - U_3 + V_3 = 50 - (-$	(-5) - 35 = 20
X <sub>34</sub>	$C_{34} - U_3 + V_4 = 55 - (-$	(-5) - 45 = 15

#### Note: $X_{13}$ is the entering variable.

# Step 2: Determine Which Current Basic Variable Reaches 0 First



Note: 1. Cycle property

2. X<sub>12</sub> is the leaving variable

# Step 3: Determine the Next Transportation Table



**Total shipping cost = 2020** 

{**Improvement** = 2 (-5) = 10 }

# 2-(MODI) Method

- Find an initial basic feasible solution by some starting procedure. Then,
- 1. Set U<sub>1</sub> 0. Solve for the other U<sub>i</sub>'s and V<sub>j</sub>'s by:
- $C_{ij} U_i + V_j = 0$  for basic variables.
- Then calculate the C<sub>ij</sub>–Z<sub>ij</sub> values for nonbasic variables by:
- $\mathbf{C}_{ij} \mathbf{Z}_{ij} = \mathbf{C}_{ij} \mathbf{U}_i + \mathbf{V}_j$
- Choose the nonbasic variable with the most negative C<sub>ij</sub>-Z<sub>ij</sub> value as the entering variable. If all C<sub>ij</sub>-Z<sub>ij</sub> values are nonnegative, STOP; the current solution is optimal.
- 2. Find the cycle that includes the entering variable and some of the BASIC variables. Alternating positive and negative changes on the cycle, determine the "change amount" as the smallest allocation on the cycle at which a subtraction will be made.
- 3. Modify the allocations to the variables of the cycle found in step 2 by the "change amount" and return to step 1.

- **Note:** there must be m + n 1 basic variables for the transportation simplex method to work!
- Add dummy source or dummy destination, if necessary
- (m of sources and n of destinations)

# Example:

Building Brick Company (BBC) has orders for 80 tons of bricks at three suburban locations as follows: Northwood -- 25 tons, Westwood -- 45 tons, and Eastwood -- 10 tons. BBC has two plants, each of which can produce 50 tons per week.

How should end of week shipments be made to fill the above orders given the following delivery cost per ton:

	<u>Northwood</u>	<u>Westwood</u>	<u>Eastwood</u>
Plant 1	24	30	40
Plant 2	30	40	42

#### Initial Transportation Table

Since total supply = 100 and total demand = 80, a dummy destination is created with demand of 20 and o unit costs.



# Example:

- Lowest cost entry method
  - <u>Iteration 1:</u> Tie for least cost (o), arbitrarily select  $x_{14}$ . Allocate 20. Reduce  $s_1$  by 20 to 30 and delete the Dummy column.
  - <u>Iteration 2:</u> Of the remaining cells the least cost is 24 for  $x_{11}$ . Allocate 25. Reduce  $s_1$  by 25 to 5 and eliminate the Northwood column.
  - <u>Iteration 3:</u> Of the remaining cells the least cost is 30 for  $x_{12}$ . Allocate 5. Reduce the Westwood column to 40 and eliminate the Plant 1 row.
  - <u>Iteration 4</u>: Since there is only one row with two cells left, make the final allocations of 40 and 10 to  $x_{22}$  and  $x_{23}$ , respectively.



# Initial table



**Total transportation cost is \$2770** 

#### • Iteration 1

#### MODI Method

1. Set 
$$u_1 = 0$$

2. Since u<sub>1</sub> + v<sub>j</sub> = c<sub>1j</sub> for occupied cells in row 1, then v<sub>1</sub> = 24, v<sub>2</sub> = 30, v<sub>4</sub> = 0.
3. Since u<sub>i</sub> + v<sub>2</sub> = c<sub>i2</sub> for occupied cells in column 2, then u<sub>2</sub> + 30 = 40, hence u<sub>2</sub> = 10.
4. Since u<sub>2</sub> + v<sub>j</sub> = c<sub>2j</sub> for occupied cells in row 2,

then

 $10 + v_3 = 42$ , hence  $v_3 = 32$ .

#### • Iteration 1

## • MODI Method (continued)

Calculate the reduced costs (circled numbers on the next slide) by  $c_{ij} - u_i + v_j$ .

Unoccupied Cell	<u>Reduced Cost</u>
(1,3)	40 - 0 - 32 = 8
(2,1)	30 - 24 -10 = -4
(2,4)	0 - 10 - 0 = -10

•Since some of the reduced cost are negative, the current solution is not optimal.

• Cell (2,4) has the most negative; therefore, it will be the basic variable that must be occupied in the next iteration.

# **Iteration 1 Table:**





# • <u>Iteration 1</u>

#### <u>Stepping Stone Method</u>

The stepping stone path for cell (2,4) is (2,4), (1,4), (1,2), (2,2). The allocations in the subtraction cells are 20 and 40, respectively. The minimum is 20, and hence reallocate 20 along this path. Thus for the next tableau:

 $x_{24} = 0 + 20 = 20$  (o is its current allocation)  $x_{14} = 20 - 20 = 0$  (blank for the next tableau)  $x_{12} = 5 + 20 = 25$   $x_{22} = 40 - 20 = 20$ The other occupied cells remain the same.



Total transportation cost is \$2570 = 2770 – 10 (20)

# • <u>Iteration 2</u>

# <u>MODI Method</u>

The reduced costs are found by calculating the  $u_i$ 's and  $v_i$ 's for this tableau.

1. Set 
$$u_1 = 0$$
.

2. Since  $u_1 + v_j = c_{ij}$  for occupied cells in row 1, then

$$v_1 = 24, v_2 = 30.$$

3. Since  $u_i + v_2 = c_{i2}$  for occupied cells in column 2, then  $u_2 + 30 = 40$ , or  $u_2 = 10$ .

4. Since  $u_2 + v_j = c_{2j}$  for occupied cells in row 2, then

$$10 + v_3 = 42 \text{ or } v_3 = 32; \text{ and, } 10 + v_4 = 0 \text{ or } v_4 = -10.$$

## • <u>Iteration 2</u>

## • MODI Method (continued)

Calculate the reduced costs (circled numbers on the next slide) by  $c_{ij} - u_i + v_j$ .

Unoccupied Cell	<u>Reduced Cost</u>
(1,3)	40 - 0 - 32 = 8
(1,4)	0 - 0 - (-10) = 10
(2,1)	30 - 10 - 24 = -4

- Since there is still negative reduced cost for cell (2,1), the solution is not optimal.
- Cell (2,1) must be occupied

# **Iteration 2 Table**

#### Northwood Westwood Eastwood Dummy Ui



## • Iteration 2

## <u>Stepping Stone Method</u>

The most negative reduced cost is = -4 determined by  $x_{21}$ . The stepping stone path for this cell is (2,1),(1,1),(1,2),(2,2). The allocations in the subtraction cells are 25 and 20 respectively. Thus the new solution is obtained by reallocating 20 on the stepping stone path. Thus for the next tableau:

 $x_{21} = 0 + 20 = 20$  (o is its current allocation)

 $X_{11} = 25 - 20 = 5$ 

 $X_{12} = 25 + 20 = 45$ 

 $x_{22} = 20 - 20 = 0$  (blank for the next tableau) The other occupied cells remain the same.

#### Northwood Westwood Eastwood Dummy Supply



Total cost is \$2490 = 2570 - 4(20)

# Iteration 3

# <u>MODI Method</u>

The reduced costs are found by calculating the  $u_i$ 's and  $v_j$ 's for this tableau.

1. Set  $u_1 = 0$ 

- 2. Since  $u_1 + v_j = c_{1j}$  for occupied cells in row 1, then  $v_1 = 24$  and  $v_2 = 30$ .
- 3. Since  $u_i + v_1 = c_{i1}$  for occupied cells in column 2, then  $u_2 + 24 = 30$  or  $u_2 = 6$ .

4. Since  $u_2 + v_j = c_{2j}$  for occupied cells in row 2, then

$$6 + v_3 = 42 \text{ or } v_3 = 36$$
, and  $6 + v_4 = 0 \text{ or } v_4 = -6$ .
#### • Iteration 3

#### • MODI Method (continued)

Calculate the reduced costs (circled numbers on the next slide) by  $c_{ij} - u_i + v_j$ .

Unoccupied Cell	<u>Reduced Cost</u>
(1,3)	40-0-36 = 4
(1,4)	0 - 0 - (-6) = 6
(2,2)	40 - 6 - 30 = 4

Since all the reduced cost are nonnegative, the current solution is optimal

#### <u>Iteration 3 Table</u>

Since all the reduced costs are non-negative, this is the optimal tableau.



# Optimal Solution

<u>From</u>	<u>To</u>	Amount	<u>Cost</u>
Plant 1	Northwood	5	120
Plant 1	Westwood	45	1,350
Plant 2	Northwood	20	600
Plant 2	Eastwood	10	<u>420</u>
	Total	Cost = \$	2,490



# **Review question**

- 1. What is unbalanced transport problem?
- 2. What is degeneracy?
- 3. Solve the following transportation problem.





#### https://www.youtube.com/watch?v=Q31jKiEXxdc







- Introduction
- <u>Hungarian Method of Assignment Problems</u>.
- Special Cases in Assignment Problems.
- <u>Exercises</u>

**The Assignment Problem (AP)** 

 Assignment problem is a special case of Transportation problem with m=n and s<sub>i</sub>=d<sub>j</sub> for all i, and j.

- The Hungarian Algorithm
- solving the assignment problem of a least
- cost assignment of m workers to m jobs

# • Assumptions:

- 1. There is a cost assignment matrix for the m "people" to be assigned to m "tasks." (If necessary dummy rows or columns consisting of all 0's are added so that the numbers of people and tasks are the same.)
- 2. All costs are nonnegative.
- 3. The problem is a minimization problem.



# The Hungarian Algorithm

# Initialization

- 1. For each row, subtract the minimum number from all numbers in that row.
- 2. In the resulting matrix, subtract the minimum number in each column from all numbers in the column.

#### Iterative Steps

- 1. Make as many 0 cost assignments as possible. If all workers are assigned, STOP; this is the minimum cost assignment. Otherwise draw the minimum number of horizontal and vertical lines necessary to cover all 0's in the matrix. (A method for making the maximum number of 0 cost assignments and drawing the minimum number of lines to cover all 0's follows.)
- 2. Find the smallest value not covered by the lines; this number is the *reduction value*.
- **3.** Subtract the reduction value from all numbers not covered by any lines. Add the reduction value to any number covered by both a horizontal and vertical line.
- GO TO STEP 1.

- For small problems, one can usually determine the maximum number of zero cost assignments by observation. For larger problems, the following procedure can be used:
- Determining the Maximum Number of Zero-Cost Assignments
- 1. For each row, if only one 0 remains in the row, make that assignment and eliminate the *row* and *column* from consideration in the steps below.
- 2. For each column, if only one 0 remains, make that assignment and eliminate that *row* and *column* from consideration.

• 3. Repeat steps 1 and 2 until no more assignments can be made. (If 0's remain, this means that there are at least two 0's in each remaining row and column. Make an arbitrary assignment to one of these 0's and repeat steps 1 and 2.)

• Again, for small problems, the minimum number of lines required to cover all the 0's can usually be determined by observation. The following procedure, based on network flow arguments, can be used for larger problems:

#### • Drawing the Minimum Number of Lines to Cover All 0's

- 1. Mark all rows with no *assignments* (with a "·").
- 2. For each row just marked, mark each column that *has a* 0 in that row (with a "·").
- 3. For each column just marked, mark each row that *has an assignment* in that column (with a "·").
- 4. Repeat steps 2 and 3 until no more marks can be made.
- 5. Draw lines through *unmarked rows* and *marked columns*.



# ABCDE 1 10 4 6 10 12 2 11 7 7 9 14 3 13 8 12 14 15 4 14 16 13 17 17 5 17 11 17 20 19 Row Subtraction







#### Minimum uncovered number

	Δ	B	С	D	F	Optimal S	Solution
	~	4		0		1–C	6
1	3	1	0	2	2	2–D	9
2	3	3	0	0	3	3-F	15
3	1	0	1	1	0		1/
4	0	6	0	2	0	4-A	14
т Б	0		0	<u> </u>	4	2–B	11
Э	2	U	3	4			65



# • CONVERSION OF A MAXIMIZATION PROBLEM TO A MINIMIZATION PROBLEM

 The Hungarian algorithm works only if the matrix is a cost matrix. A maximization assignment problem can be converted to a minimization problem by creating a <u>lost opportunity matrix</u>. The problem then is to minimize the total lost opportunity.

#### Q solve the given assignment problem. given profit matrix Profit Matrix:

- J1 J2 J3 J4
- W1 67 58 90 55
- W2 58 88 89 56
- W3 74 99 80 22
- (D) 0 0 0 0

The <u>lost opportunity matrix</u> given below is derived by subtracting each number in the J1 column from 74, each number in the J2 column from 99, each number in the J3 column from 90, and each number in the J4 from 56.

- J1 J2 J3 J4
- W1 7 41 0 1
- W2 16 11 1 0
- W3 0 0 10 34
- (D) 74 99 90 56

• The Hungarian algorithm can now be applied to this lost opportunity matrix to determine the maximum profit set of assignments.

# **Review questions**

- Q.1 what is assignment problem ?
- Q.2 solve the following assignment problem

Ļ	Machine jobs	Α	В	С	D	Е
	1	10	20	4	24	8
	2	8	20	6	28	12
	3	12	26	8	32	12
	4	14	28	8	36	18
	5	16	28	12	36	24

# • Q.3 solve the following assignment problems (unbalanced problems)

/	Resources and workers	Jı	J2	J <sub>3</sub>	J4
	W1	7	5	8	4
	W2	5	6	7	4
	W3	8	7	9	8

# **Problems of maxmization**

- A maximization assignment problem can be converted to a minimization problem by creating a lost opportunity matrix using any of the following methods. The problem then is to minimize the total lost opportunity. The first method is by putting a negative sign before the values in the assignment matrix and then solves the sum as a minimization case using Hungarian methods.
- Second method is to locate the largest value in the given matrix and subtract each element in the matrix from this value .then one can solve this problem as a minimization case using the new modified matrix .

# Example:

• A company has a team of four salespersons and there are four districts where the company wants to start its business .after taking into account the capabilities of salesman and nature of districts, the company estimates the profit per day in rupees for each salesman for each district given in the following table:

District and salesman	Ι	Π	III	IV
А	16	10	14	11
В	14	11	15	15
С	15	15	13	12
D	13	12	14	15

# Travelling salesman problem(TSP)

- Travelling salesman problem is a routine problem. It can be considered as a typical assignment problem with certain restrictions . Consider a salesman who is assigned the job of visiting n different cities . He knows(is given) the distance (or time) between all pairs of cities . He is asked to visit each of the cities only once. The trip should be continuous and he should come back to the city from where he started using the shortest route . It does not matter , from which city he starts. These restrictions imply:
- (i) no assignment should be made along the diagonal ,going directly from city i to j is not permitted . Therefore diagonal elements are generally taken very large i.e. ∞

# Mathematical formulation

• A travelling salesman has to visit n cities and return to the starting point. He has to start from any one city and visit each city only once . The objective is to select the sequence which the cities are visited in such a way that the total travelling distance(or time) is minimized . Suppose he starts from the kth , city and the last he visited is m. let the cost of travel from ith , city to jth city be Cij .then the objective function

is

*Minimize*  $\sum_{i=1}^{m} \sum_{j=1}^{n} c_{ij} x_{ij}$ 

• Subject to the constraints

 $\sum_{i=1}^{n} x_{ij} = 1, \text{ for, } i = 1, 2, \dots, n, i \neq j, i \neq m$  $\sum_{j=1}^{n} x_{ij} = 1, \text{ for, } j = 1, 2, \dots, n, i \neq j, j \neq n$  $x_{mk} = 1$  $x_{ij} = 0 \text{ or } 1$ 

# Working procedure

- <u>Step-1</u> solve the problem as an assignment problem using the method used to solve the above examples.
- <u>Step-2</u> (a) if the solution thus found out are cyclic in nature ,then that is the final solution.
- (b) if it is not cyclic ,then go to step-3.
- <u>Step-3</u> select the lowest entry in the table ( other than zero). Delete the row and column of this lowest entry and again do the zero assignment in the remaining matrix.
- <u>Step-4</u> check whether cyclic assignment is available. If not, include the next higher entry in the table and the procedure is repeated until a cyclic assignment is obtained.

# Example

• Consider a four city TSP for which the cost between the city pairs are as shown in the figure below. Find the tour of the salesman so that the cost of travel is minimal.



# Solution

#### • Construct the cost matrix as follows:



Step-1



• The optimal assignment is  $1 \rightarrow 4, 2 \rightarrow 3, 3 \rightarrow 2, 4 \rightarrow 1$ Which is not cyclic.

<u>Step-2</u> consider the lowest entry '2' of the cell (2,1). If there is a tie in selecting the lowest entry, then break the tie arbitrarily. Delete the II row and I column. Do the zero assignment in the remaining matrix. The resulting table is



:-

• Next optimal assignment is  $1 \rightarrow 4, 2 \rightarrow 1, 3 \rightarrow 2, 4 \rightarrow 3$ Which is cyclic.



# https://www.youtube.com/watch?v=q7VMzOTZvqM https://www.youtube.com/watch?v=BUGIhEecipE

# **Conclusion**

- The transportation problem is a special class of the linear programming problem. It deals with the situation in which a commodity is transported from **sources(origins)** to **destinations.** The objective is to determine the amount of commodity to be transported from each source to each destination so that the total transportation cost is minimum.
- The Assignment problem is special from of the transportation problem with equal number of sources and destination and supply at each source and demand at each destination is limited to one unit.
- The name "assignment problem" originates from the classical problem where the objective is to assign a number of origins(job) to the equal number of destinations(persons) at minimum cost(or maximum profit) travelling salesman problem is a special type of assignment problem.

Thank you