

# Using LP to solve IP

Instructor: Kwei-Long Huang

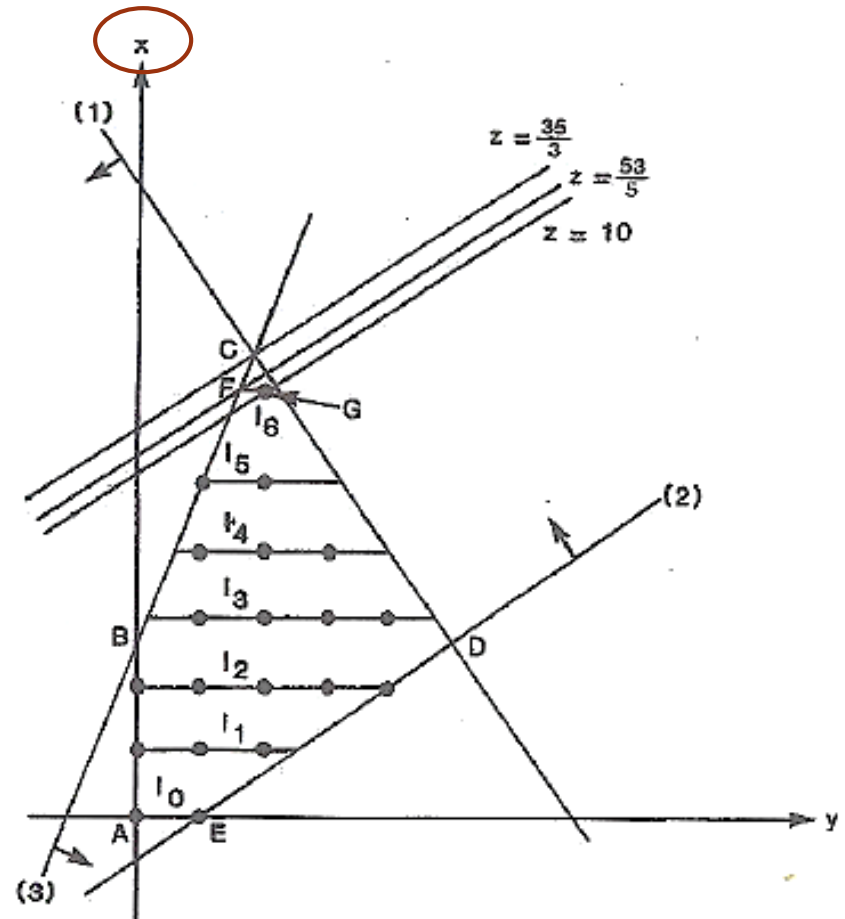
Course No: 546 U6110

# Agenda

- Graphical Solutions
- Unimodularity
- Rounding LP Solutions

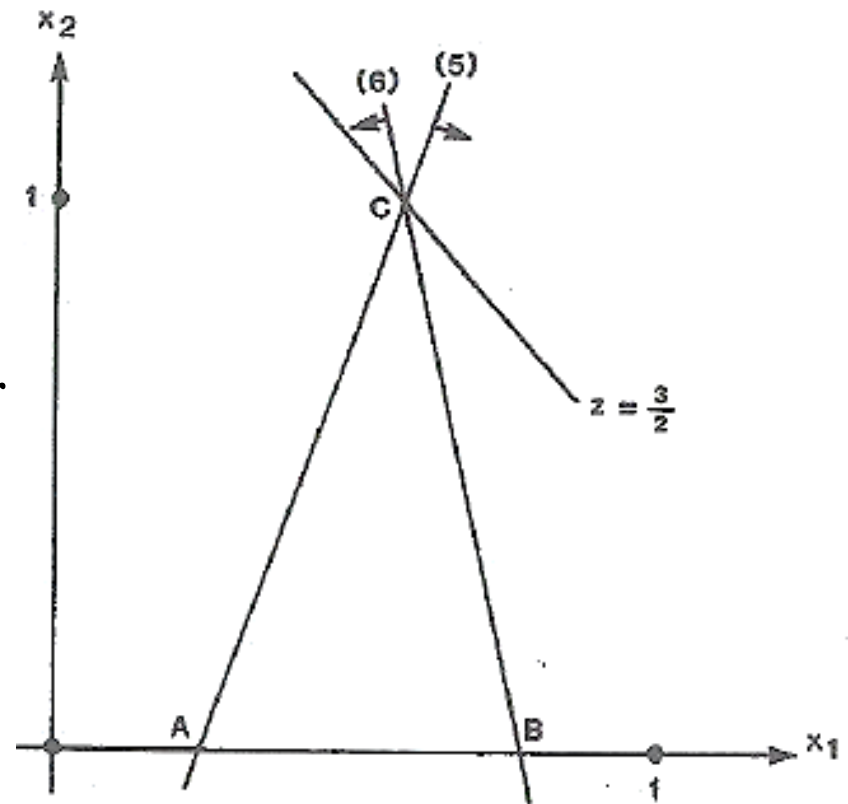
# Graphical Solutions

Maximize  $2x - y = z$   
 subject to  $5x + 7y \leq 45$ ,  
 $-2x + y \leq 1$ ,  
 $2x - 5y \leq 5$ ,  
 $x, y \geq 0$ ,  
 and  $x$  integer.



# No Solution

Maximize  $x_1 + x_2 = z$   
subject to  $-4x_1 + x_2 \leq -1$ ,  
 $4x_1 + x_2 \leq 3$ ,  
and  $x_1, x_2 \geq 0$  and integer



# Some Insights

- The maximal value of the objective function to the MIP (IP) solved as a LP is \_\_\_\_\_ on the value of any feasible solution to MIP (IP).
- If the optimal solution to the MIP solved as a LP is integer in its integer constrained variables, it solves the MIP
- If the MIP solved as a linear one is **infeasible**, so is the MIP.

# Agenda

- Graphical Solutions
- Unimodularity
- Rounding LP Solutions

# Unimodularity

- Matrix  $A$  is totally unimodular *if and only if* the determinant of every square of  $A$  is either 0, +1 or -1.

# Examples

- The following matrices are TU.

$$\begin{pmatrix} 1 & -1 & -1 & 0 \\ -1 & 0 & 0 & 1 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 0 \end{pmatrix}, \quad \begin{pmatrix} 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 0 \end{pmatrix}$$

- The following matrices are not TU

$$\begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}, \quad \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix}$$



# Simplest Determinant

For a 2\*2 matrix  $A$ , the determinant, denoted by  $|A|$  or  $\det(A)$ , is given by

$$|A| = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{12}a_{21}$$

# Example

$$A = \begin{bmatrix} -5 & 4 \\ 2 & 7 \end{bmatrix}$$

$$|A| = ?$$

$$|A| = -5*7 - 2*4 = -43$$

# Cofactor Expansion

Formal definition of determinants of  $n \times n$  matrices will **adopt** the definition of the determinant of a  $2 \times 2$  matrix.

Minor of  $a_{ij}$ :  $M_{ij}$

For a square  $n \times n$  matrix  $A$ ,  $M_{ij}$  is the  
determinants of the matrix after  
remove row  $i$  and column  $j$ .

# Example

$$A = \left[ \begin{array}{c|c|c} & & \\ \hline & & \\ \hline a_{21} & & \dots a_{2n} \\ \vdots & & \vdots \\ a_{n1} & & \dots a_{nn} \\ \hline \end{array} \right]$$

$$M_{12} = ?$$

# Example

$$A = \begin{bmatrix} 1 & 0 & \text{red} \\ \text{red} & \text{red} & \text{red} \\ 0 & -2 & \text{red} \end{bmatrix} \quad M_{23} = ?$$

$$M_{23} = \left\| \begin{bmatrix} 1 & 0 \\ 0 & -2 \end{bmatrix} \right\| = -2$$

Cofactor of  $a_{ij}$ :  $C_{ij}$

$$C_{ij} = (-1)^{i+j} M_{ij}$$

$C_{ij}$  and  $M_{ij}$  might be different only by a **sign**.

# Example

$$A = \begin{bmatrix} 1 & 0 & \text{red box} \\ \text{red box} & \text{red box} & \text{red box} \\ 0 & -2 & \text{red box} \end{bmatrix} \quad C_{23} = ?$$

$$C_{23} = (-1)^{2+3} \left| \begin{bmatrix} 1 & 0 \\ 0 & -2 \end{bmatrix} \right| = 2$$



## Cofactor Expansions on first row (column)

For an  $n \times n$  matrix,  $A$ , the determinant is the sum of the products of the elements of the **first** row (column) and their cofactors.

# Cofactor Expansions on First Row/Column

$$|A| = a_{11}C_{11} + a_{12}C_{12} + a_{13}C_{13} \\ + \dots + a_{1n}C_{1n}$$

Or

$$|A| = a_{11}C_{11} + a_{21}C_{21} + a_{31}C_{31} \\ + \dots + a_{n1}C_{n1}$$

# Example

$$A = \begin{bmatrix} 1 & 0 & 3 \\ 4 & -1 & 2 \\ 0 & -2 & 1 \end{bmatrix} \quad |A| = ?$$

$$\begin{aligned} |A| &= a_{11}C_{11} + a_{12}C_{12} + a_{13}C_{13} \\ &= 1(-1)^{1+1}(-1+4) + 0 + \\ &\quad 3(-1)^{1+3}(-8+0) = -21 \end{aligned}$$

# Cofactor Expansions on $i$ th Row or $j$ th Column

$i$ th row expansion:

$$|A| = a_{i1}C_{i1} + a_{i2}C_{i2} + \dots + a_{in}C_{in}$$

$j$ th column expansion:

$$|A| = a_{1j}C_{1j} + a_{2j}C_{2j} + \dots + a_{nj}C_{nj}$$

## Example

$$A = \begin{bmatrix} 1 & 0 & 3 \\ 4 & -1 & 2 \\ 0 & -2 & 1 \end{bmatrix} \quad |A| = ?$$

$$\begin{aligned} |A| &= a_{12}C_{12} + a_{22}C_{22} + a_{32}C_{32} \\ &= 0 + (-1)(-1)^{2+2}(1-0) + \\ &\quad (-2)(-1)^{3+2}(2-12) = -21 \end{aligned}$$

# Example

$$A = \begin{bmatrix} 2 & 1 & 0 & 4 \\ 0 & -1 & 0 & 2 \\ 7 & -2 & 3 & 5 \\ 0 & 1 & 0 & -3 \end{bmatrix} \quad |A| = ?$$

$$\begin{aligned} |A| &= 3(-1)^{3+3}[2(-1)^{1+1}(3-2)] \\ &= 6 \end{aligned}$$

# Matrix of Cofactors

For a square  $n \times n$  matrix  $A$ ,

$$\text{Matrix of Cofactor} = \begin{bmatrix} C_{11} & C_{12} & \dots & C_{1n} \\ C_{21} & C_{22} & \dots & C_{2n} \\ \vdots & \vdots & & \vdots \\ C_{n1} & C_{n2} & \dots & C_{nn} \end{bmatrix}$$

## Example

$$A = \begin{bmatrix} 1 & 0 & 3 \\ 4 & -1 & 2 \\ 0 & -2 & 1 \end{bmatrix} \text{ Matrix of Cofactor} = ?$$

$$\begin{array}{l} C_{11}=3, C_{12}=-4, C_{13}=-8, \\ C_{21}=-6, C_{22}=1, C_{23}=2, \\ C_{31}=3, C_{32}=10, C_{33}=-1 \end{array} \quad \begin{bmatrix} 3 & -4 & -8 \\ -6 & 1 & 2 \\ 3 & 10 & -1 \end{bmatrix}$$



# Adjoint of $A$

For a square  $n \times n$  matrix  $A$ ,

$$\text{Adj}(A) = (\text{Matrix of Cofactors})^t$$

$$\text{Adj}(A) = \begin{bmatrix} C_{11} & C_{21} & \cdots & C_{n1} \\ C_{12} & C_{22} & \cdots & C_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ C_{1n} & C_{2n} & \cdots & C_{nn} \end{bmatrix}$$

## Example

$$A = \begin{bmatrix} 1 & 0 & 3 \\ 4 & -1 & 2 \\ 0 & -2 & 1 \end{bmatrix} \quad Adj(A) = ?$$

$$\begin{array}{l} \text{Matrix of} \\ \text{Cofactors} \\ = \end{array} \begin{bmatrix} 3 & -4 & -8 \\ -6 & 1 & 2 \\ 3 & 10 & -1 \end{bmatrix} \Rightarrow \begin{bmatrix} 3 & -6 & 3 \\ -4 & 1 & 10 \\ -8 & 2 & -1 \end{bmatrix}$$

# The Inverse and the Determinant

Let  $A$  be a **square** matrix with  $|A| \neq 0$  ( $A$  is invertible )

$$A^{-1} = \frac{1}{|A|} \text{Adj}(A)$$

## Example 3

- Use the formula for the inverse of a matrix to compute the inverse of the matrix

$$A = \begin{bmatrix} 2 & 0 & 3 \\ -1 & 4 & -2 \\ 1 & -3 & 5 \end{bmatrix}$$

## Cofactor of A

$$C_{11} = \begin{vmatrix} 4 & -2 \\ -3 & 5 \end{vmatrix} = 14 \quad C_{12} = -\begin{vmatrix} -1 & -2 \\ 1 & 5 \end{vmatrix} = 3 \quad C_{13} = \begin{vmatrix} -1 & 4 \\ 1 & -3 \end{vmatrix} = -1$$

$$C_{21} = -\begin{vmatrix} 0 & 3 \\ -3 & 5 \end{vmatrix} = -9 \quad C_{22} = \begin{vmatrix} 2 & 3 \\ 1 & 5 \end{vmatrix} = 7 \quad C_{23} = -\begin{vmatrix} 2 & 0 \\ 1 & -3 \end{vmatrix} = 6$$

$$C_{31} = \begin{vmatrix} 0 & 3 \\ 4 & -2 \end{vmatrix} = -12 \quad C_{32} = -\begin{vmatrix} 2 & 3 \\ -1 & -2 \end{vmatrix} = -1 \quad C_{33} = \begin{vmatrix} 2 & 0 \\ -1 & 4 \end{vmatrix} = 8$$

The adjoint of  $A$  is the transpose of this matrix.

$$\text{adj}(A) = \begin{bmatrix} 14 & -9 & -12 \\ 3 & 7 & 1 \\ -1 & 6 & 8 \end{bmatrix}$$

## Solution

$|A| = 25$ , so the inverse of  $A$  exists. Thus,

$$A^{-1} = \frac{1}{|A|} \text{adj}(A) = \frac{1}{25} \begin{bmatrix} 14 & -9 & -12 \\ 3 & 7 & 1 \\ -1 & 6 & 8 \end{bmatrix} = \begin{bmatrix} \frac{14}{25} & \frac{-9}{25} & \frac{-12}{25} \\ \frac{3}{25} & \frac{7}{25} & \frac{1}{25} \\ \frac{-1}{25} & \frac{6}{25} & \frac{8}{25} \end{bmatrix}$$

# Cramer's Rule

Given  $AX=B$  and  $|A| \neq 0$ ,

$$x_1 = \frac{|A_1|}{|A|}, x_2 = \frac{|A_2|}{|A|}, \dots x_n = \frac{|A_n|}{|A|}$$

where  $A_i$  is from  $A$  by replacing column  $i$  with  $B$ .

$$|A_i|$$

$$A = \begin{bmatrix} \boxed{\begin{matrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{matrix}} & a_{12} & \cdots & \boxed{\begin{matrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{matrix}} & a_{1n} \\ & a_{22} & \cdots & & a_{2n} \\ & \vdots & & & \vdots \\ & a_{m2} & \cdots & & a_{mn} \end{bmatrix} \quad B = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

$$|A_1| = ? \quad |A_j| = ?$$



# Unimodularity Theorem

Consider

$$\begin{aligned} &\text{maximize } z = c^T x, \\ &\text{subject to } Ax = b, \\ &\quad x \geq 0, \text{ integer,} \end{aligned}$$

Where  $A$  is unimodular and  $b$  is integer

## **Unimodularity Theorem:**

Let  $A$  be an  $m \times n$  integer matrix with  $\text{rank}(A) = m$  which is smaller than  $n$ .

The following are equivalent:

- For each basis matrix  $B$  of  $A$ ,  $\det B = +1$  or  $-1$ .
- For each integer vector  $b$ , every feasible extreme point of  $X = \{x \mid Ax = b, x \geq 0\}$  is integer.
- For each basis matrix  $B$  of  $A$ ,  $B^{-1}$  is integer.

# Summary

- If an ILP has a coefficient matrix  $A$  that is totally unimodular and the right-hand-side vector  $b$  is integer, then every basic feasible solution (BFS) is integer. Thus, the integrality constraints can be relaxed and the ILP can be solved as an LP.
- A sufficient condition for TU:
  - Let  $A$  be an  $m \times n$  matrix
  - All entries in  $A$  are  $+1$ ,  $-1$  or  $0$
  - Each column contain at most 2 nonzero entries
  - The row of  $A$  can be partitioned into 2 sets. For each column, if the entries are of the same sign, there are in different sets; otherwise, there are in the same set.

# Assignment Problem

$$\text{Minimize} \quad \sum_{i=1}^3 \sum_{j=1}^3 \mathbf{c}_{ij} \mathbf{x}_{ij}$$

$$\text{subject to } \sum_{i=1}^3 x_{ij} = 1 \quad (j = 1, 2, 3)$$

$$\sum_{j=1}^3 x_{ij} = 1 \quad (i = 1, 2, 3)$$

and  $x_{ij} = 0 \text{ or } 1 \quad (\text{all } i, j).$

$$\begin{array}{cccccccccc}
 & x_{11} & x_{12} & & \dots\dots & & x_{32} & x_{33} \\
 \text{Machine} & \left[ \begin{array}{ccccccccc} 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right] \\
 \\
 \text{Job} & \left[ \begin{array}{ccccccccc} 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 \end{array} \right]
 \end{array}$$

# Minimum Cost Network Flow Problem

$$\text{maximize } z = \sum_{i=1}^m \sum_{j=1}^m c_{ij} x_{ij}$$

$$\text{subject to } \sum_{j=1}^m x_{ij} - \sum_{k=1}^m x_{ki} = b_i, \quad i = 1, \dots, m,$$

$$x_{ij} \geq 0, \text{ integer}$$

# Agenda

- Graphical Solutions
- Unimodularity
- Rounding LP Solutions

# Rounding LP Solutions

- Example: facility location (assume all produced items have to be shipped out)

		Shopping costs Customers ( $n$ )					Production Capacity $M_i$
		1	2	3	4	5	
Sources ( $m$ )	1	93	70	48	68	81	2
	2	45	89	97	85	96	3
	3	92	93	58	37	99	2
	4	55	103	55	57	38	3
	5	74	60	78	54	52	2
Demands $d_j$		1	1	1	1	1	

# LP Solution

LP optimal solution, cost = 228

$$x_1 = x_3 = x_5 = 1/2, \quad x_2 = x_4 = 1/3,$$

$$z_{13} = z_{21} = z_{34} = z_{45} = z_{52} = 1$$

Facility variables	Shipping variables	Total cost
$x_1 = x_2 = 1$	$z_{12}=z_{13}=z_{21}=z_{24}=z_{25}=1$	344
$x_1 = x_4 = 1$	$z_{12}=z_{13}=z_{41}=z_{44}=z_{45}=1$	268
$x_3 = x_2 = 1$	$z_{21}=z_{22}=z_{25}=z_{33}=z_{34}=1$	325
$x_3 = x_4 = 1$	$z_{32}=z_{34}=z_{41}=z_{43}=z_{45}=1$	278
$x_5 = x_2 = 1$	$z_{21}=z_{22}=z_{23}=z_{54}=z_{55}=1$	337
$x_5 = x_4 = 1$	$z_{41}=z_{43}=z_{45}=z_{52}=z_{54}=1$	<b>262</b>

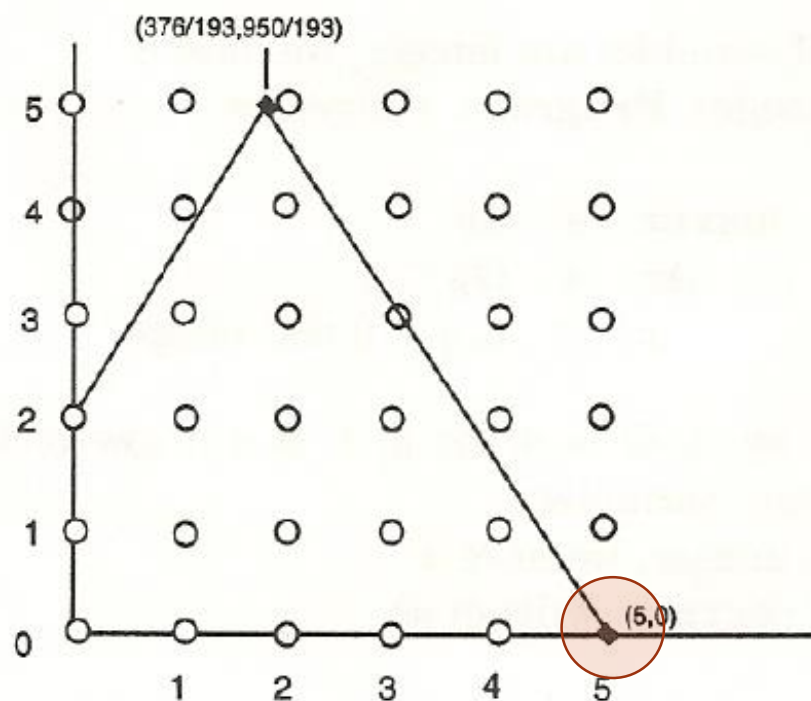
## Example 2

$$\max 1.00x_1 + 0.64x_2$$

$$50x_1 + 31x_2 \leq 250$$

$$3x_1 - 2x_2 \geq -4$$

$$x_1, x_2 \geq 0 \text{ and integer.}$$





# Questions

- Homework 2 (**due on 3/30**). Please download it from Ceiba.