Operations Research Lab Session

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Self-introduction

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(1) Use it to find a solution of an equation Example: $x_1 + 2x_2 = 4$ $2x_1 + x_2 = 5$

1 2

What is the solution of the equation?

$$\begin{bmatrix} 1 & 2 & | 4 \\ 2 & 1 & | 5 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & | 4 \\ 0 & -3 & | -3 \end{bmatrix}$$
$$\rightarrow \begin{bmatrix} 1 & 2 & | 4 \\ 0 & 1 & | 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & | 2 \\ 0 & 1 & | 1 \end{bmatrix}$$
$$\rightarrow (x_1, x_2) = (2, 1)$$

Practice:

 $x_{1} + 2x_{2} - 3x_{3} = 2$ $x_{1} - x_{3} = 0$ $x_{1} - x_{2} + 2x_{3} = 3$

What is the solution of the equation?

 $\begin{vmatrix} 1 & 2 & -3 & | 2 \\ 1 & 0 & -1 & | 0 \\ 1 & -1 & 2 & | 3 \end{vmatrix} \rightarrow \begin{vmatrix} 1 & 2 & -3 & | 2 \\ 0 & -2 & 2 & | -2 \\ 0 & -3 & 5 & | 1 \end{vmatrix}$ $\rightarrow \begin{vmatrix} 1 & 2 & -3 & | 2 \\ 0 & 1 & -1 & | 1 \\ 0 & -3 & 5 & | 1 \end{vmatrix} \rightarrow \begin{vmatrix} 1 & 0 & -1 & | 0 \\ 0 & 1 & -1 & | 1 \\ 0 & 0 & 2 & | 4 \end{vmatrix}$ $\rightarrow \begin{vmatrix} 1 & 0 & -1 & | 0 \\ 0 & 1 & -1 & | 1 \\ 0 & 0 & 1 & | 2 \end{vmatrix} \rightarrow \begin{vmatrix} 1 & 0 & 0 & | 2 \\ 0 & 1 & 0 & | 3 \\ 0 & 0 & 1 & | 2 \end{vmatrix}$

 $\rightarrow (x_1, x_2, x_3) = (2, 3, 2)$

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(2) Use it to find the inverse of a matrix

Example:

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

What is the inverse of matrix *A*?

$$\begin{bmatrix} 1 & 5 & | 1 & 0 \\ 2 & 3 & | 0 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 5 & | 1 & 0 \\ 0 & -7 & | -2 & 1 \end{bmatrix}$$
$$\rightarrow \begin{bmatrix} 1 & 5 & | 1 & 0 \\ -2 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 5 & | 1 & 0 \\ -2 & 1 \end{bmatrix}$$

 $\rightarrow A^{-1} = \frac{1}{7} \begin{bmatrix} -3 & 5\\ 2 & -1 \end{bmatrix}$

Practice:

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

What is the inverse of matrix *A*?

$$\begin{bmatrix} 1 & 0 & 1 & | & 1 & 0 & 0 \\ 4 & 1 & -2 & | & 0 & 1 & 0 \\ 3 & 1 & -1 & | & 0 & 0 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 1 & | & 1 & 0 & 0 \\ 0 & 1 & -6 & | & -4 & 1 & 0 \\ 0 & 1 & -4 & | & -3 & 0 & 1 \end{bmatrix}$$
$$\rightarrow \begin{bmatrix} 1 & 0 & 1 & | & 1 & 0 & 0 \\ -4 & 1 & 0 & | & -4 & 1 & 0 \\ 0 & 1 & -6 & | & -4 & 1 & 0 \\ 1 & -1 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 1 & | & 1 & 0 & 0 \\ -4 & 1 & 0 & | & 1 & -1 & 1 \\ 0 & 1 & -6 & | & 1 & -1 & 1 \\ 1 & 2 & 2 & 2 & 1 \end{bmatrix}$$
$$\rightarrow \begin{bmatrix} 1 & 0 & 1 & | & 1 & 0 & 0 \\ -4 & 1 & 0 & | & 1 & -1 & 1 \\ 1 & 2 & 2 & 2 & 1 \end{bmatrix} \rightarrow A^{-1} = \frac{1}{2} \begin{bmatrix} 1 & 1 & -1 \\ -2 & -4 & 6 \\ 1 & -1 & 1 \end{bmatrix}$$

Linearly independent / dependent

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$$\square \quad A = \begin{bmatrix} 1 & 2 & 5 \\ 0 & 1 & 2 \end{bmatrix}$$

- $\Box \quad \text{Column vectors of } A = \{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 5 \\ 2 \end{bmatrix} \}$
- □ Row vectors of $A = \{ \begin{bmatrix} 1 & 2 & 5 \end{bmatrix}, \begin{bmatrix} 0 & 1 & 2 \end{bmatrix} \}$
- Definition: A collection of vectors $a^1, ..., a^n$ is linearly independent if $\sum_{j=1}^n c_j a^j = 0$ imply that $c_1 = c_2 = ... = c_n = 0$

$$1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + 2 \begin{bmatrix} 2 \\ 1 \end{bmatrix} - 1 \begin{bmatrix} 5 \\ 2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

=> Column vectors of *A* are linearly dependent

 $c_1[1 \quad 2 \quad 5] + c_2[0 \quad 1 \quad 2] = 0 \text{ only when } c_1 = c_2 = 0$ => Row vectors of *A* are linearly independent

Linearly independent / dependent

- □ When row vectors of a matrix are linearly dependent
- After Gauss-Jordan elimination

there must exist at least a row that all element are 0.

Example:

$$B = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} \Rightarrow \text{No inverse!!!}$$

Rank of a matrix

- **Column rank** of matrix A is the maximal number of linearly independent column of A.
- **Row rank** of matrix A is the maximal number of linearly independent row of A.
- Since the column rank and row rank are always equal, they are simply called the rank of A.

Example:

$$A = \begin{bmatrix} 1 & 2 \\ 0 & 3 \\ 2 & -1 \end{bmatrix} \rightarrow \operatorname{Rank} = 2, \quad B = \begin{bmatrix} 1 & 2 \\ 0 & 0 \\ 2 & 4 \end{bmatrix} \rightarrow \operatorname{Rank} = 1$$

Rank of a matrix

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 5 & 4 \\ 1 & 1 & 5 \end{bmatrix}, \operatorname{rank}(A) = ? A^{-1}?$$
$$\begin{bmatrix} 1 & 0 & 7 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \end{bmatrix}$$

 \square Rank(A) = 2, no inverse!

- Step 1: Define the decision variables (and the notations you use for parameters).
- Step 2: Formulate the problem as a mathematical model by writing down the objective function and constraints.
- Step 3: Solve the model by finding the values for all decision variables in an optimal solution.
- Step 4: Interpret the optimal solution by indicating "what to do".

Scenario:

- □ You are in a market, and you have 5 dollars.
- □ There are several drinks (Coke, Pepsi, Orange juice).
- □ Each drink gives you different happiness level.
- □ Which drinks should you buy?

Step 1: Define the problem and collect relevant data

Goal: To maximize your happiness level.

Data:

Name	Price	Happiness level
Coke	2 dollars	3
Pepsi	3 dollars	6
Orange juice	2.5 dollars	4

Step 2: Formulating the problem

Parameters: 5 dollars, 3 drinks....

Decision variables: For each drink, we decide whether to buy. Let x_i be the amount of the drink we buy, i = 1,2,3.

Objected function: $3x_1 + 6x_2 + 4x_3$

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Constraint: 2x_1 + 3x_2 + 2.5x_3 \le 5
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Our model:

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$$\max 3x_1 + 6x_2 + 4x_3$$

s.t. $2x_1 + 3x_2 + 2.5x_3 \le 5$
 $x_1, x_2, x_3 \in \{0, 1\}$

Step 3: Solving the model

(1, 1, 0) will be the solution.Objected value is 9

Step 4: Interpret

To get maximum happiness, we should buy a Coke and a Pepsi. Finally we get 9 happiness level. 20

Thank you ③