# Operations Research Lab Session 

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## Self－introduction

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## Gauss-Jordan elimination (1)

(1) Use it to find a solution of an equation

Example:

$$
\begin{aligned}
x_{1}+2 x_{2} & =4 \\
2 x_{1}+x_{2} & =5
\end{aligned}
$$

What is the solution of the equation?

## Gauss-Jordan elimination (1)

$$
\begin{aligned}
& {\left[\begin{array}{ll|l}
1 & 2 & 4 \\
2 & 1 & 5
\end{array}\right] \rightarrow\left[\begin{array}{cc|c}
1 & 2 & 4 \\
0 & -3 & -3
\end{array}\right] } \\
\rightarrow & {\left[\begin{array}{ll|l}
1 & 2 & 4 \\
0 & 1 & 1
\end{array}\right] \rightarrow\left[\begin{array}{lll|}
1 & 0 & 2 \\
0 & 1 & 1
\end{array}\right] } \\
\rightarrow & \left(x_{1}, x_{2}\right)=(2,1)
\end{aligned}
$$

## Gauss-Jordan elimination (1)

## Practice:

$$
\begin{aligned}
& x_{1}+2 x_{2}-3 x_{3}=2 \\
& x_{1}-x_{3}=0 \\
& x_{1}-x_{2}+2 x_{3}=3
\end{aligned}
$$

What is the solution of the equation?

## Gauss-Jordan elimination (1)

$$
\begin{aligned}
& {\left[\begin{array}{ccc|c}
1 & 2 & -3 & 2 \\
1 & 0 & -1 & 0 \\
1 & -1 & 2 & 3
\end{array}\right] \rightarrow\left[\begin{array}{ccc|c}
1 & 2 & -3 & 2 \\
0 & -2 & 2 & -2 \\
0 & -3 & 5 & 1
\end{array}\right]} \\
& \rightarrow\left[\begin{array}{ccc|c}
1 & 2 & -3 & 2 \\
0 & 1 & -1 & 1 \\
0 & -3 & 5 & 1
\end{array}\right] \rightarrow\left[\begin{array}{ccc|c}
1 & 0 & -1 & 0 \\
0 & 1 & -1 & 1 \\
0 & 0 & 2 & 4
\end{array}\right] \\
& \rightarrow\left[\begin{array}{ccc|c}
1 & 0 & -1 & 0 \\
0 & 1 & -1 & 1 \\
0 & 0 & 1 & 2
\end{array}\right] \rightarrow\left[\begin{array}{lll|l}
1 & 0 & 0 & 2 \\
0 & 1 & 0 & 3 \\
0 & 0 & 1 & 2
\end{array}\right] \\
& \rightarrow\left(x_{1}, x_{2}, x_{3}\right)=(2,3,2)
\end{aligned}
$$

## Gauss-Jordan elimination (2)

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

Example:

$$
A=\left[\begin{array}{ll}
1 & 5 \\
2 & 3
\end{array}\right]
$$

What is the inverse of matrix $A$ ?

## Gauss-Jordan elimination (2)

$$
\begin{aligned}
& {\left[\begin{array}{ll|ll}
1 & 5 & 1 & 0 \\
2 & 3 & 0 & 1
\end{array}\right] \rightarrow\left[\begin{array}{cc|cc}
1 & 5 & 1 & 0 \\
0 & -7 & -2 & 1
\end{array}\right] } \\
\rightarrow & {\left[\begin{array}{ll|ll}
1 & 5 & 1 & 0 \\
0 & 1 & \frac{-1}{7} & \frac{-1}{7}
\end{array}\right] \rightarrow\left[\begin{array}{ll|ll}
1 & 0 & \frac{-3}{7} & \frac{5}{7} \\
0 & 1 & \frac{2}{7} & \frac{-1}{7}
\end{array}\right] } \\
\rightarrow & A^{-1}=\frac{1}{7}\left[\begin{array}{cc}
-3 & 5 \\
2 & -1
\end{array}\right]
\end{aligned}
$$

## Gauss-Jordan elimination (2)

## Practice:

$A=\left[\begin{array}{ccc}1 & 0 & 1 \\ 4 & 1 & -2 \\ 3 & 1 & -1\end{array}\right]$

What is the inverse of matrix $A$ ?

## Gauss-Jordan elimination (2)

$$
\begin{aligned}
& {\left[\begin{array}{ccc|ccc}
1 & 0 & 1 & 1 & 0 & 0 \\
4 & 1 & -2 & 0 & 1 & 0 \\
3 & 1 & -1 & 0 & 0 & 1
\end{array}\right]}
\end{aligned} \rightarrow\left[\begin{array}{cccc|ccc}
1 & 0 & 1 & 1 & 0 & 0 \\
0 & 1 & -6 & -4 & 1 & 0 \\
0 & 1 & -4 & -3 & 0 & 1
\end{array}\right]
$$

## Linearly independent / dependent

- $A=\left[\begin{array}{lll}1 & 2 & 5 \\ 0 & 1 & 2\end{array}\right]$
- Column vectors of $A=\left\{\left[\begin{array}{l}1 \\ 0\end{array}\right],\left[\begin{array}{l}2 \\ 1\end{array}\right],\left[\begin{array}{l}5 \\ 2\end{array}\right]\right\}$
$\square$ Row vectors of $A=\left\{\left[\begin{array}{lll}1 & 2 & 5\end{array}\right],\left[\begin{array}{lll}0 & 1 & 2\end{array}\right]\right\}$
$\square$ Definition: A collection of vectors $\mathrm{a}^{1}, \ldots, \mathrm{a}^{\mathrm{n}}$ is linearly independent if $\sum_{j=1}^{n} c_{j} a^{j}=0$ imply that $c_{1}=c_{2}=\ldots=c_{n}=0$
- $1\left[\begin{array}{l}1 \\ 0\end{array}\right]+2\left[\begin{array}{l}2 \\ 1\end{array}\right]-1\left[\begin{array}{l}5 \\ 2\end{array}\right]=\left[\begin{array}{l}0 \\ 0\end{array}\right]$
=> Column vectors of $A$ are linearly dependent
$\square c_{1}\left[\begin{array}{lll}1 & 2 & 5\end{array}\right]+c_{2}\left[\begin{array}{lll}0 & 1 & 2\end{array}\right]=0$ only when $c_{1}=c_{2}=0$
=> Row vectors of $A$ are linearly independent


## Linearly independent / dependent

$\square$ When row vectors of a matrix are linearly dependent
$\square$ After Gauss-Jordan elimination there must exist at least a row that all element are 0 .

Example:
$B=\left[\begin{array}{ll}1 & 2 \\ 2 & 4\end{array}\right] \Rightarrow\left[\begin{array}{ll}1 & 2 \\ 0 & 0\end{array}\right] \Rightarrow$ No inverse!!!

## Rank of a matrix

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

Example:
$A=\left[\begin{array}{cc}1 & 2 \\ 0 & 3 \\ 2 & -1\end{array}\right] \rightarrow \operatorname{Rank}=2, \quad B=\left[\begin{array}{ll}1 & 2 \\ 0 & 0 \\ 2 & 4\end{array}\right] \rightarrow \operatorname{Rank}=1$

## Rank of a matrix

$\square A=\left[\begin{array}{lll}1 & 2 & 3 \\ 2 & 5 & 4 \\ 1 & 1 & 5\end{array}\right], \operatorname{rank}(A)=? A^{-1} ?$
$\square\left[\begin{array}{ccc}1 & 0 & 7 \\ 0 & 1 & -2 \\ 0 & 0 & 0\end{array}\right]$
$\square \operatorname{Rank}(A)=2$, no inverse!

## DFSI principle

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

## DFSI principle

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

## DFSI principle

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 |

## DFSI principle

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: $3 x_{1}+6 x_{2}+4 x_{3}$
Constraint: $2 x_{1}+3 x_{2}+2.5 x_{3} \leq 5$

Our model:

$$
\begin{aligned}
& \max 3 x_{1}+6 x_{2}+4 x_{3} \\
& \text { s.t. } 2 x_{1}+3 x_{2}+2.5 x_{3} \leq 5 \\
& x_{1}, x_{2}, x_{3} \in\{0,1\}
\end{aligned}
$$

## DFSI principle

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.

Thank you ©

