# CHAPTER 6

Matrices and systems of Linear Equations

#### 6-1 Matrixes

#### 1. Matrix

A matrix is any rectangular array of objects arranged in rows and in columns.

 $A=[a_{ij}] = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1m} \\ a_{21} & a_{22} & \cdots & a_{2m} \\ \vdots & & & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nm} \end{bmatrix}$ 

Here aij is the entry or element of matrix A in row i and column j.

n x m	Matrix	example
n≠m	Rectangular	$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$
n = m	Square	$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$
n =1	Row	$\begin{bmatrix} 1 & 2 & 3 & 4 \end{bmatrix}_{1\times 4}$
m =1	Column	$\begin{bmatrix} 1\\2\\3\\4 \end{bmatrix}_{4\times 1}$

2. Equality of matrix

$$A = \left[a_{ij}\right]_{m \times n} = B = \left[b_{ij}\right]_{p \times q}$$

iff m = p, n = q and  $a_{ij} = b_{ij}$ 

- 3. Matrix algebra
  - (1) matrix addition

$$A = \begin{bmatrix} a_{ij} \end{bmatrix}_{m \times n}, B = \begin{bmatrix} b_{ij} \end{bmatrix}_{n \times m}$$
$$A + B = \begin{bmatrix} a_{ij} + b_{ij} \end{bmatrix}_{m \times n}$$

(2) product of matrix and a scalar  $\alpha$ 

$$\alpha A = \left[\alpha a_{ij}\right]_{n \times m}$$

(3) multiplication of matrices

$$A_{n \times r} B_{r \times m} = \left[ \sum_{s=1}^{r} a_{ij} b_{ij} \right]$$
  
=  $C_{n \times m}$   
=  $\left[ C_{ij} \right]_{n \times m}$   
$$C_{ij} = a_{i1} b_{1j} + a_{i2} b_{2j} + \dots + a_{ir} b_{rj}$$
  
=  $(row \ i \ of \ A) \bullet (column \ j \ of \ B)$   
$$\begin{cases} AA = A^{2} \\ AAA = A^{3} \qquad A^{100} = ? \quad if \qquad A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$
  
 $AA \dots = A^{n}$ 

(4) Theorems

- i. Commutative law of addition A+B=B+A
- ii. Associative law of addition A+(B+C) = (A+B)+C
- iii. Distributive law  $A \cdot (B+C) = AB + AC$
- iv. Distributive law  $(A+B) \cdot C = AC + BC$
- v. Associative law of multiplication (AB)C=A (BC)

(5) Other properties

i. 
$$AB \neq BA$$
  
ii.  $AB = AC$  and  $B \neq C$   
Ex.  

$$\begin{bmatrix} 1 & 1 \\ 3 & 3 \end{bmatrix} \begin{bmatrix} 4 & 12 \\ 3 & 6 \end{bmatrix} = \begin{bmatrix} 7 & 18 \\ 21 & 54 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 3 & 3 \end{bmatrix} \begin{bmatrix} 2 & 7 \\ 5 & 11 \end{bmatrix}$$
but  $\begin{bmatrix} 4 & 12 \\ 3 & 6 \end{bmatrix} \neq \begin{bmatrix} 2 & 7 \\ 5 & 11 \end{bmatrix}$   
iii.  $AB = 0$  and  $A \neq 0, B \neq 0$ 

(6) Special Matrices

(i) zero matrix

$$0 = \left[a_{ij}\right]_{n \times m} \text{ where } a_{ij} = 0$$

(a) A + 0 = 0 + A = 0(b) A + (-A) = 0(ii) Identity matrix,  $I_n$ 

$$I_{n} = \left[ I_{ij} \right]_{n \times n} \quad \text{Square matrix} \iff I_{ij} = \begin{cases} 0 & i \neq j \\ 1 & i = j \end{cases}$$
$$\boxed{I_{n}A_{n \times m} = A_{n \times m}I_{m} = A_{n \times m}}$$

(iii) Transpose(轉置) of matrix

$$A = \begin{bmatrix} a_{ij} \end{bmatrix}_{n \times m} \qquad A^{t} = \begin{bmatrix} a_{ij} \end{bmatrix}_{m \times n}$$
  
ij element of  $A^{t} = ji$  element of  $A$   
(a)  $I_{n}^{t} = I_{n}$   
(b)  $(A^{t})^{t} = A$   
(c)  $(AB)^{t} = B^{t}A^{t}$ 

### 4. Matrices and systems of linear equations

(1) System of linear algebra equation

$$\begin{cases} 2x_1 + x_2 = 1\\ 3x_1 + 2x_2 = 2 \end{cases}$$

In matrix form

$$\begin{bmatrix} 2 & 1 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \quad Ax = B$$

where A=matrix of coefficients of system

X=matrix of unknown

B=matrix of constants

(2) system of linear differential eqn.

$$\begin{cases} x_1' + tx_2' - x_3' = f(t) \\ t^2 x_1' - \cos(t)x_2' - x_3' = g(t) \end{cases}$$

In matrix form

$$\begin{bmatrix} 1 & t-1 \\ t^2 & -\cos(t)-1 \end{bmatrix} \begin{bmatrix} x_1' \\ x_2' \\ x_3' \end{bmatrix} = \begin{bmatrix} f(t) \\ g(t) \end{bmatrix}$$
  

$$AX' = F$$
  
Ex. System AX=0 
$$\begin{bmatrix} -1 & 0 & 1 & 1 & 2 \\ 0 & 1 & 3 & 0 & 4 \\ 1 & 2 & 1 & 1 & 1 \\ -3 & 1 & 0 & 0 & 4 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ X_5 \end{bmatrix} = 0, \text{ the solution space}=?$$
  
Sol. A<sub>R</sub>=
$$\begin{bmatrix} 1 & 0 & 0 & 0 & -\frac{9}{8} \\ 0 & 1 & 0 & 0 & \frac{5}{8} \\ 0 & 0 & 1 & 0 & \frac{9}{8} \\ 0 & 0 & 0 & 1 & \frac{-1}{4} \end{bmatrix}$$

 $\therefore m = 5 \operatorname{rank}(A) = \operatorname{rank}(A_R) = 4$ 

: Dimension of solution space of AX=0

m-rank(A)=5-4=1

General solution is

$$\begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ X_5 \end{bmatrix} = \begin{bmatrix} \frac{-9}{8} \\ \frac{5}{8} \\ \frac{9}{8} \\ \frac{-1}{4} \\ 1 \end{bmatrix}$$
 One arbitrary constant.

3. Trivial solution

The solution  $X_1 = X_2 = \dots = X_n = 0$  is the trivial solution of a homogeneous system AX=0.

4. Existence of nontrivial solution

(1)Theorem

Let A be  $n \times m$  matrix. The system AX=0 has a nontrivial solution iff

m>rank(A).

: Dimension of solution space of AX=0=m-rank(A)>0

Then system AX=0 has a nonzero solution!

(2)Corollary

A. AX=0 always has a nontrivial solution if the number of unknowns exceed the number of equations.

A.  $A_{n \times m} X_{m \times 1} = 0$ , n is the number of equations; m is the number of unknowns.

 $\therefore$  n<m and rank(A)  $\le$  n. $\therefore$  m-rank(A)  $\ge$  m-n>0.

B. A is an n×m matrix of real number. Then the system AX=0 has only the trivial solution iff  $A_R=I_n$ 

 $:: A_R = I_n$ 

 $\therefore$  rank(A)=rank(A<sub>R</sub>)=n, and m=n.

 $\therefore$  m-rank(A)n-n=0

 $\Rightarrow$  System AX=0 has only zero solution.

# 6-7 Nonhomogeneous Systems of Linear Equations

1. Nonhomogeneous linear system

```
\begin{array}{l}
(a_{11}x_{1} + a_{12}x_{2} + \dots + a_{1m}x_{m} = b_{1} \\
a_{21}x_{1} + a_{22}x_{2} + \dots + a_{2m}x_{m} = b_{2} \\
\dots \\
a_{n1}x_{1} + a_{n2}x_{2} + \dots + a_{nm}x_{m} = b_{n}
\end{array}
```

In matrix from AX=B

When  $A = [a_{ij}] = n \times m$  coefficient matrix

 $X=[X_j]=m \times 1$  unknowns

 $B = [b_i] = n \times 1$  matrix (at least one of which is nonzero).

Augmented matrix of system AX=B

2.Structure of solution of AX=B

(1)Theorem

Let Up be a solution of AX=B, then every solution of AX=Bi the form of  $U_p$ +H. in which H is a solution of AX=0.

[Proof]Assume AW=B  $\therefore$   $AU_p=B$ 

 $\therefore AW = AU_p = A(W - U_P) = B - B = 0$ 

- : W-U<sub>P</sub>=H is the solution of AX=0
- $\therefore W{=}U_P{+}H$
- (2) Remark on  $U_P$  and H
  - A. U<sub>P</sub>=particular solution of AX=0
  - B. H=general solution of AX=0
  - C. U<sub>P</sub>+H=general solution of AX=0
- 3. Existence and Uniquences of solutions
  - (1) Consistent system of equation

Definition:

AX=B is said to be consistent if there exists a solution, otherwise the system is inconsistent.

- (2) Reduced form of Augmented matrix  $[A:B]_{12} = [AR:C]$
- (3) Existence of a solution

Theorem:

System AX=B has a solution iff A and [A:B] have the same rank.

Rank(A)=rank([A:B])

(4)Uniquence of a solution

Let A be n×n. Then system AX=B have a unique solution iff

 $A_R = In \text{ or } rank(A) = n$ 

$$A_n \times_n X = B \Longrightarrow A_R X = C$$

4. Procedure for solving AX=B

Step 1. Find reduced matrix of  $[A:B] \rightarrow [A_R:_C]$ 

Step 2. Check if rank(A)=rank([A:B]) for existence of solution.

Step3. Identify the dependent unknowns(作法同 system AX=0)

Step4. Find general solution by assigning independent unknowns any constant.

Ex. 
$$\begin{bmatrix} -1 & 1 & 3 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix} = \begin{bmatrix} -2 \\ 4 \end{bmatrix}$$
  
Sol.  $[A \vdots B]_R = \begin{bmatrix} -1 & 1 & 3 & -2 \\ 0 & 1 & 2 & 4 \end{bmatrix}$ , reduced form  $[A \vdots B]$   
 $[A \vdots B]_R = \begin{bmatrix} 1 & 0 & -1 & 6 \\ 0 & 1 & 2 & 4 \end{bmatrix} = [A_R \vdots C]$   
rank $(A) = \operatorname{rank}(A_R) = 2$   
rank $([A \vdots B]) = \operatorname{rank}([A_R \vdots C]) = 2$ 

 $\therefore$  rank(A)=rank([A:B])

 $\Rightarrow$  consistent system  $\Rightarrow$  a solution exists.

From A<sub>R</sub>, we have

 $X_1$  and  $X_2$ = dependent unknowns

 $X_3$  = independent unknowns.

Ex. 
$$\begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix} = \begin{bmatrix} 6 \\ 4 \end{bmatrix}$$

Sol. A<sub>R</sub>X=C

$$\therefore X_{1} = X_{3} + 6$$

$$X_{2} = -2X_{3} + 4$$

$$\Rightarrow X = \begin{bmatrix} X_{1} \\ X_{2} \\ X_{3} \end{bmatrix} = \begin{bmatrix} X_{3} + 6 \\ -2X_{3} + 4 \\ X_{3} \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} X_{3} + \begin{bmatrix} 6 \\ 4 \\ 0 \end{bmatrix}$$

Let  $X_3 = \alpha$ , general solution is

$$X = \alpha \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} + \begin{bmatrix} 6 \\ 4 \\ 0 \end{bmatrix}$$
  
= H + U<sub>P</sub>  
H: general solution of AX=0  
U<sub>P</sub>: particular solution of AX=B

# 6-9 Matrix Inverses

Inverse of a square matrix. If AB=BA, then B is an inverse matrix of A.

(1)Definition of nonsingular matrix

A matrix that has an inverse is called nonsingular; otherwise called singular.

(2)Theorem of uniquence of an inverse

A nonsingular matrix has exactly one inverse.

$$\begin{bmatrix} Proof \end{bmatrix} \begin{cases} BA = I \\ CA = I \end{cases} \Rightarrow B=C \\ B=BI=BAC=IC=C \end{cases}$$

(3) Theorem of existence of an inverse matrix  $An \times m$  is nonsingular iff  $A_R=In$ 2. Properties

Let A and B be nonsingular matrices

- (1)  $(AB)^{-1} = (BA)^{-1}$
- (2)  $(A^{-1})^{-1} = A$
- (3)  $(A^t)^{-1} = (A^{-1})^t$
- (4)  $(In)^{-1} = In$

(5) AB and BA are singular if A and B are  $n \times n$  matrix and either is singular.

3.Method for Finding Inverse

(1) Row operation method 
$$[In : A] \to [\Omega : A_R] = [A^{-1} : In]$$
  
 $\Omega A = A_R, \text{ If } A_R = In \therefore \Omega = A^{-1} \quad (\Omega A = In)$   
Ex.  $A = \begin{bmatrix} 5 & -1 \\ 6 & 8 \end{bmatrix} , A^{-1} = ?$   
Sol.  $[I_2 : A] = \begin{bmatrix} 1 & 0 & | & 5 & -1 \\ 0 & 1 & | & 6 & 8 \end{bmatrix} \xrightarrow{\text{row} \text{operation}} \begin{bmatrix} \frac{8}{46} & \frac{1}{46} & | & 1 & 0 \\ \frac{-6}{46} & \frac{5}{46} & | & 0 & 1 \end{bmatrix} = [\Omega : A_R] = [A^{-1} : I_2]$   
 $\therefore A^{-1} = \Omega = \begin{bmatrix} \frac{8}{46} & \frac{1}{46} \\ \frac{-6}{46} & \frac{5}{46} \end{bmatrix}$ 

(2) Determinants method (refer to section 7.7)

4. Use the Inverse to solve linear system

Let A be an  $n \times n$  matrix. Then AX=B has a solution iff A is nonsingular.

 $\therefore$  A is nonsingular  $\therefore$  A has inverse A<sup>-1</sup>

 $\Rightarrow$  A<sup>-1</sup>AX=A<sup>-1</sup>B X=A<sup>-1</sup>B

#### 6.2 Elementary Row operations and Elementary Matrices AX=0

#### 1. Elementary row operations

- (1) type I : Interchange any two rows
- (2) type II : multiply a row by a nonzero scalar
- (3) type III : Add a scalar multiple of one row to another row

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} = \begin{bmatrix} 5 \\ 6 \end{bmatrix}$$

type I operation (row 1  $\iff$  row 2)

$$\begin{bmatrix} 3 & 4 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} = \begin{bmatrix} 6 \\ 5 \end{bmatrix}$$

type II operation (- $1/2 \times row2$ )

$$\begin{bmatrix} 1 & 2 \\ -\frac{3}{2} & -2 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} = \begin{bmatrix} 5 \\ -3 \end{bmatrix}$$

type III operation  $(-1/2 \times row2 + row1)$ 

$$\begin{bmatrix} -\frac{1}{2} & 0\\ 3 & -2 \end{bmatrix} \begin{bmatrix} X_1\\ X_2 \end{bmatrix} = \begin{bmatrix} 2\\ 6 \end{bmatrix}$$

## 2. Elementary matrix

A matrix formed by performing an elementary row operation an identity matrix  $I_n$ .

Ex. 
$$\mathbf{I}_{3}$$
=  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$   
Type 1 row operation  $\begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$  elementary matrix  
Type 2 row operation  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -4 & 0 & 1 \end{bmatrix}$  elementary matrix

(1) Theorem1

B=EA

Where **B**=matrix formed from A by elementary row operation

E=elementary matrix formed from I by the same row operation on A.

Ex.

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \text{ type3 operation } -\frac{1}{2} \times \text{ row2} \rightarrow \text{ row1}$$
  
Sol. 
$$\mathbf{B} = \begin{bmatrix} -\frac{1}{2} & 0 \\ 3 & 4 \end{bmatrix}$$
$$\mathbf{I}_{2} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \rightarrow \mathbf{E} = \begin{bmatrix} 1 & -\frac{1}{2} \\ 0 & 1 \end{bmatrix}$$
Sol. 
$$\mathbf{E} \mathbf{A} = \begin{bmatrix} 1 & -\frac{1}{2} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} & 0 \\ 3 & 4 \end{bmatrix} = \mathbf{B}$$

(2) Theorem 2.

**B=Ω**A.

Where

 $\mathbf{B}$  = matrix produced from  $\mathbf{A}$  by any finite sequence of elementary row operations.

 $\boldsymbol{\Omega}=\text{matrix}$  of a product of elementary matrices formed form I by the same

sequence of row operations on A.

Ex. 
$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$
  $\mathbf{O}_1 = (\frac{1}{2} \times \operatorname{row2} \to \operatorname{row1})$  on  $\mathbf{A}$  to form  $\mathbf{A}_1$ .  
 $\mathbf{O}_2 = (6 \times \operatorname{row1})$  on  $\mathbf{A}_1$  to form  $\mathbf{A}_2$ .  
 $\mathbf{O}_3 = (\operatorname{row1} \to \operatorname{row2})$  on  $\mathbf{A}_2$  to form  $\mathbf{B}$ .  
If  $\mathbf{B} = \mathbf{\Omega} \mathbf{A}$  then  $\mathbf{B} = ?$   $\mathbf{\Omega} = ?$ 

Sol.

$$\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \xrightarrow{\mathbf{O}_1} \mathbf{A}_1 = \begin{bmatrix} -1/2 & 0 \\ 3 & 4 \end{bmatrix} \xrightarrow{\mathbf{O}_2} \mathbf{A}_2 = \begin{bmatrix} -3 & 0 \\ 3 & 4 \end{bmatrix} \xrightarrow{\mathbf{O}_3} \mathbf{B} = \begin{bmatrix} -3 & 0 \\ 0 & 4 \end{bmatrix}$$
$$\mathbf{I}_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \xrightarrow{\mathbf{O}_1} \begin{bmatrix} 1 & -\frac{1}{2} \\ 0 & 1 \end{bmatrix} \xrightarrow{\mathbf{O}_2} \begin{bmatrix} 6 & -3 \\ 0 & 1 \end{bmatrix} \xrightarrow{\mathbf{O}_3} \begin{bmatrix} 6 & -3 \\ 6 & -2 \end{bmatrix} = \mathbf{\Omega}$$

CHECK

$$\Omega \mathbf{A} = \begin{bmatrix} 6 & -3 \\ 6 & -2 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} -3 & 0 \\ 0 & 4 \end{bmatrix} = \mathbf{B}$$
$$\Omega = \mathbf{E}_1 \mathbf{E}_2 \mathbf{E}_3$$
$$\begin{cases} \mathbf{I} \stackrel{\mathbf{O}_1}{\longrightarrow} \mathbf{E}_1 \\ \mathbf{I} \stackrel{\mathbf{O}_2}{\longrightarrow} \mathbf{E}_2 \\ \mathbf{I} \stackrel{\mathbf{O}_3}{\longrightarrow} \mathbf{E}_3 \end{cases}$$

#### 3. Row equivalence of matrices

Matrix A is row equivalent to matrix B if B can be obtained from A by a sequence of elementary row operations.

Theorems.

(1)Reflexive property	A $\xrightarrow{\text{row equivalent}}$ A
(2)Symmetric property	A $\leftarrow$ row equivalent B
(3)Transitive property	$A \xrightarrow{r.e} B \xrightarrow{r.e} C \implies A \xrightarrow{r.e} C$

## 6.3 The reduced Form of a Matrix

### 1. Zero row

All of its elements (entries) of a row of a matrix are Zero

#### 2. Nonzero row

At least one element of a row is not Zero

## 3. Leading entry

The first nonzero element of a nonzero row (left  $\rightarrow$  right)

	row	zero row	leading entry
$\begin{bmatrix} 0 & 2 & 7 \end{bmatrix}$			
$Ex:\begin{bmatrix} 0 & 2 & 7 \\ 0 & -2 & 0 \\ 0 & 0 & 0 \end{bmatrix}$	1	Х	2 (a12)
$\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$	2	Х	-2 (a22)
0 0 9	3	О	-
	4	Х	9 (a <sub>43</sub> )

#### 4. Reduced Matrix

A matrix satisfies the following conditions

- 1. leading entry of nonzero row is 1
- 2. leading entry in column j , all other elements of column j are zero
- 3. if  $k\ a\ zero\ row$  , i a nonzero row , then  $i{<}k$
- 4. if leading entry of row  $r_1$  is in column  $c_1$ , leading entry of row  $r_2$  is in column  $c_2$ , and  $r_1 < r_2$ , then  $c_1 < c_2$

Ex $A = \begin{pmatrix} 1 & -4 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$		row	zero row	Leading entry	elements of column in	
	$\rightarrow$ A is redu	ced matrix				leading entry
		1	Х	$\frac{1}{1} (a_{11}) \\ \frac{1}{1} (a_{24})$	0	
			2	Х	1 (a <sub>24</sub> )	0
Ex i	. –	2 0 0 0 1 0 0 0 1 0 0 0 Leading entry	Other	entry of col	lumn of leading e	entry
1	Х	1 (a <sub>12</sub> )			0	
2	Х	$ \begin{array}{cccc} 1 & (a_{24}) \\ 1 & (a_{35}) \\ & - \\ (1) & (4) \end{array} $			0	
3	Х	1 (a <sub>35</sub> )	0			
4	0	-	-			
	(3)	(1) (4)	(2)			
→reduced matrix						

Ex. C= $\begin{bmatrix} 0 & 1 & 5 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$	0 0 0 1		
row	Zero row	Leading entry	Other entry
1	Х	1 $(a_{12})$	0
2	Х	1 (a <sub>23</sub> )	5
3	Х	1 $(a_{34})$	0
4	Х	1 $(a_{45})$	0

Condition (2) not satisfied, NG.  $\rightarrow$  not reduced matrix

$$\mathbf{C} = \begin{bmatrix} 0 & 1 & 5 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \rightarrow (-5 \times \text{row}2 + \text{row}1) \rightarrow \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} = \mathbf{B}$$

#### 5.Theorems

(1)Equivalence to a Reduced Matrix

Every matrix **A** is row equivalent to a reduced matrix

(2)Reduced matrix A<sub>R</sub>

A is a matrix, then there is exactly one reduced matrix  $A_R$  that is row equivalent to A. (Reduced matrix of A)

Ex. 
$$\mathbf{A} = \begin{bmatrix} -2 & 1 & 3 \\ 0 & 1 & 1 \\ 2 & 0 & 1 \end{bmatrix}$$
 A is not a reduced matrix

 $A_R$ =reduced form of A=?

Sol.

$$\mathbf{A} = \begin{bmatrix} -2 & 1 & 3 \\ 0 & 1 & 1 \\ 2 & 0 & 1 \end{bmatrix} \rightarrow \frac{1}{2} (\text{row 1}) \rightarrow \begin{bmatrix} 1 & -\frac{1}{2} & -\frac{3}{2} \\ 0 & 1 & 1 \\ 2 & 0 & 1 \end{bmatrix}$$

$$\rightarrow -2(\text{row 1}) + (\text{ row 3}) \rightarrow \begin{bmatrix} 1 & -\frac{1}{2} & -\frac{3}{2} \\ 0 & 1 & 1 \\ 0 & 1 & 4 \end{bmatrix}$$

$$\rightarrow (\text{row 2}) + (\text{row 1}) \cdot -1(\text{row 2}) + (\text{row 3}) \rightarrow \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 3 \end{bmatrix}$$

$$\rightarrow \frac{1}{3}(\text{row 3}) \rightarrow \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\rightarrow -1(\text{row 3}) + (\text{row 2}) \cdot (\text{row 3}) + (\text{row 1}) \rightarrow \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \mathbf{I}_{3} = \mathbf{A}_{R}$$

(3)  $\Omega \mathbf{A} = \mathbf{A}_{\mathbf{R}}$ 

Let A be an  $n \times m$  matrix. Then there is an  $n \times m$  matrix  $\Omega$  such that  $\Omega A = A_{R}$ . How to find  $\Omega$ ?

Method

Augmented matrix 
$$n \times (n + m) [\mathbf{I}_n : \mathbf{A}_{n \times m}]$$
  
 $\downarrow$   
A sequence of row operations

$$\begin{array}{c} \text{Matrix } [\Omega \vdots A_{R}] \\ \downarrow \\ \hline \Omega A = A_{R} \end{array}$$

Ex. 6.20

Given A 
$$\begin{pmatrix} -3 & 1 & 0 \\ 4 & -2 & 1 \end{pmatrix}$$
  
Find (1) A<sub>R</sub>  
(2)  $\Omega$  such that  $\Omega$  A= A<sub>R</sub>

Sol.

$$\begin{bmatrix} \mathbf{I}_2 & \vdots & \mathbf{A}_{2\times 4} \end{bmatrix} = \left\langle \begin{array}{cc|c} 1 & 0 & -3 & 1 & 0 \\ 0 & 1 & 4 & 2 & 1 \end{array} \right\rangle$$
$$\rightarrow -\frac{1}{3} (\text{row } 1) \rightarrow \left\langle \begin{array}{cc|c} -1/3 & 0 & 1 & -1/3 & 0 \\ 0 & 1 & 4 & -2 & 1 \end{array} \right\rangle$$

$$\rightarrow -4(\text{row 1}) + (\text{row 2}) \rightarrow \begin{pmatrix} -1/3 & 0 & | & 1 & -1/3 & 0 \\ 4/3 & 1 & | & 0 & -2/3 & 1 \end{pmatrix} \rightarrow -\frac{2}{3}(\text{row 2}) \rightarrow \begin{pmatrix} -1/3 & 0 & | & 1 & -1/3 & 0 \\ -2 & -3/2 & | & 0 & 1 & -3/2 \end{pmatrix} \rightarrow \frac{1}{3}(\text{row 2}) + (\text{row 1}) \rightarrow \begin{pmatrix} -1 & -1/2 & | & 1 & 0 & -1/2 \\ -2 & -3/2 & | & 0 & 1 & -3/2 \end{pmatrix} = [\Omega \vdots \mathbf{A}_{\mathbf{R}}] \Rightarrow \Omega = \begin{bmatrix} -1 & 1/2 \\ -2 & -3/2 \end{bmatrix} \text{ and } \mathbf{A}_{\mathbf{R}} = \begin{bmatrix} 1 & 0 & -1/2 \\ 0 & 1 & -3/2 \end{bmatrix}$$

check if  $\Omega \mathbf{A} = \mathbf{A}_{\mathbf{R}}$ 

### 6.4 Row and Column Spaces of a Matrix and Rank of a Matrix

1. Row space of a matrix

If A is any  $n \times m$  matrix, the rows of A can be thought as vector in  $\mathbb{R}^m$ . The set of all linear combinations of these row vectors forms the row space of A, and is a subspace of  $\mathbb{R}^m$ .

EX:

Matrix A = 
$$\begin{pmatrix} 1 - 1 & 4 & 2 \\ 0 & 1 & 3 & 2 \\ 3 - 2158 \end{pmatrix}$$

sol: row space of A

 $\alpha$  (1,-1,4,2)+ $\beta$  (0,1,3,2)+ $\gamma$  (3,-2,15,8)

Dimension of row space of A?

: row vectors of A are linearly dependent.

(3,-2,15,8)=3(1,-1,4,2)+(0,1,3,2)

- $\therefore$  row space of A is of the following form.
  - $\alpha$  (1,-1,4,2)+ $\beta$  (0,1,3,2)
- $\therefore$  vectors (1,-1,4,2) & (0,1,3,2) are linearly independent.
  - => (1,-1,4,2) & (0,1,3,2) form the basis of row space of A.
  - => Dimension=2.
- 2. column space of a matrix

If A is any  $n \times m$  matrix, the column of A can be thought as vector in  $\mathbb{R}^n$ . The set of all linear combinations of these column vectors forms the column space of A, and is a subspace of  $\mathbb{R}^n$ .

$$\mathbf{A} = \begin{pmatrix} a_{11} \cdots a_{1m} \\ \vdots \\ a_{n1} & a_{nm} \end{pmatrix}_{n \times m}$$

column vectors of A are

$$\begin{pmatrix} a_{11} \\ \vdots \\ a_{n1} \end{pmatrix} \begin{pmatrix} a_{12} \\ \vdots \\ a_{n2} \end{pmatrix} \dots \begin{pmatrix} a_{1m} \\ \vdots \\ a_{nm} \end{pmatrix} \text{ in } \mathbb{R}^{n}.$$

column space of A

$$C_{1}\begin{pmatrix}a_{11}\\\vdots\\a_{n1}\end{pmatrix}+C_{2}\begin{pmatrix}a_{12}\\\vdots\\a_{n2}\end{pmatrix}+\cdots+C_{m}\begin{pmatrix}a_{1m}\\\vdots\\a_{nm}\end{pmatrix}$$

EX:

$$A = \begin{pmatrix} -1 & 4 & 0 & 1 & 6 \\ -2 & 8 & 0 & 2 & 12 \end{pmatrix}$$

Column space of a

$$C_{1}\begin{pmatrix} -1\\ 2 \end{pmatrix} + C_{2}\begin{pmatrix} 4\\ 8 \end{pmatrix} + C_{3}\begin{pmatrix} 0\\ 0 \end{pmatrix} + C_{4}\begin{pmatrix} 1\\ 2 \end{pmatrix} + C_{5}\begin{pmatrix} 6\\ 12 \end{pmatrix}$$
  

$$\therefore \text{ each column vector} = \alpha \begin{pmatrix} -1\\ -2 \end{pmatrix}$$
  

$$\therefore \text{ column space of A} - C_{1} + 4C_{2} + C_{4} + 6C_{5} = 0$$
  

$$\alpha \begin{pmatrix} -1\\ -2 \end{pmatrix} \text{ in } \mathbb{R}^{2} - 2C_{1} + 8C_{2} + 2C_{4} + 12C_{5} = 0$$
  

$$\Rightarrow \begin{pmatrix} -1\\ -2 \end{pmatrix} \text{ form the basis} \Rightarrow \text{ Dimension} = 1$$

3. Theorems

- (1) If A is a matrix , then the row and column spaces of A has the same dimension.
- (2) If matrix B is formed from matrix A by an elementary row operation, then matrix A and B have the same row space.

4. Rank, rank(A) 秩

The rank of a matrix is the number of nonzero rows of the reduced from of a matrix

Rank(A) = rank(A<sub>R</sub>) = number of nonzero rows of A<sub>R</sub>

EX:

$$A = \begin{pmatrix} 1 & -1 & 4 & 2 \\ 0 & 1 & 3 & 2 \\ 3 & -2 & 15 & 8 \end{pmatrix} \quad rank(A) = ?$$

sol: Find reduced from of A

$$A_R = \begin{pmatrix} 1 & 0 & 7 & 4 \\ 0 & 1 & 3 & 2 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

 $\therefore$  rank(A) = rank( $A_R$ )=2

(1) Theorem

Let A be an nxn matrix. Then rank(A) = n iff  $A_R = I_n$ 

(2) Lemma

If A is a reduced matrix, then rank of A equals the dimension of the row space of A.

 $\operatorname{rank}(A) = \operatorname{rank}(A_R)$ 

= Dimension of row space of A

= Dimension of column space of A

## 6.5 solutions of homogeneous systems of linear equations

1. system of linear equations

Homogeneous linear system

 $\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1m}x_m = 0 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2m}x_m = 0 \\ \vdots \\ a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nm}x_m = 0 \end{cases}$  n equations in m unknowns

In matrix from

AX= O

where

$$A = \begin{bmatrix} a_{ij} \end{bmatrix} = \begin{bmatrix} a_{11} \cdots a_{1m} \\ a_{21} & \vdots \\ \vdots & \vdots \\ a_{n1} & a_{nm} \end{bmatrix} = \text{matrix off coefficients}$$

$$\mathbf{X} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix} = \text{matrix of unknowns } \circ$$

 $O = n \times 1$  zero matrix

The reduced system of AX=0 is

$$A_R X = 0$$

which has the same solution as AX=0

1. Theorem

Let A be <code>nxm</code> matrix  $\,\circ\,$  Then the liner homogeneous system AX= 0 has the same solution as the reduced system A\_RX= 0  $\,\circ\,$ 

- 2. Gauss Jordan Reduction method for AX = 0
  - (1) Find  $A_R$
  - (2) Determine the dependent and independent unknowns

If column j of  $A_R$  contains the leading entry of some row , then  $x_j$  is

independent unknowns •

- (3) Solve for each dependent unknowns as a sum of constants times independent unknowns in each nonzero row of A<sub>R</sub>.
- (4) Assign the independent unknowns any values and yield the general solution.

EX:

Given : linear system

$$\begin{cases} x_1 - 3x_2 + x_3 - 7x_4 + 4x_5 = 0\\ x_1 + 2x_2 - 3x_3 = 0\\ x_2 - 4x_3 + x_5 = 0 \end{cases}$$

Find : general solution

sol :

In matrix from

$$\begin{bmatrix} 1 & -3 & 1 & -7 \\ 1 & 2 & -3 & 0 \\ 0 & 1 & -4 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = 0$$

$$\therefore \mathbf{A} = \begin{bmatrix} 1 & -3 & 1 & -7 & 4 \\ 1 & 2 & -3 & 0 & 0 \\ 0 & 1 & -4 & 0 & 1 \end{bmatrix}$$

Step (1) Find AR

$$\begin{pmatrix} 1-3 \ 1 \ -74 \\ 1 \ 2 \ -3 \ 0 \ 0 \\ 0 \ 1 \ 4 \ 0 \ 1 \end{pmatrix} \longrightarrow A_{R} = \begin{pmatrix} -\frac{35}{16} \ \frac{13}{16} \\ 0 \ 10 \ \frac{28}{16} \ -\frac{20}{16} \\ 0 \ 1 \ \frac{28}{16} \ -\frac{20}{16} \\ 0 \ 1 \ \frac{7}{16} \ -\frac{9}{16} \end{pmatrix}$$

(2) Determine dependent and independent unknowns columns 1,2,3,of AR contain the leading entries

Result to dependent unknowns : X1,X2,X3

Independent unknowns: X4,X5

(3)Solve reduced system ArX=0

$$\begin{pmatrix} -\frac{35}{16} & \frac{13}{16} \\ 100 & \frac{28}{16} & -\frac{20}{16} \\ 010 & \frac{28}{16} & -\frac{20}{16} \\ 001 & \frac{7}{16} & -\frac{9}{16} \\ \end{pmatrix} \begin{pmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ X_5 \\ \end{pmatrix} = 0$$
$$X_1 = 35/16 \times 4 - 13/16 \times 5$$
$$X_2 = -28/16 \times 4 + 20/16 \times 5$$

$$X_2 = -7/16x4 + 9/16x5$$

(4) Assign any value to independent unknowns:

 $X_4 = 16 \alpha$ ,  $X_5 = 16 \beta$ ,  $\alpha$ ,  $\beta$ : arbitrary constants General solution

$$\begin{pmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ X_5 \end{pmatrix} = \alpha \begin{pmatrix} 35 \\ -28 \\ -7 \\ 16 \\ 0 \end{pmatrix} \beta \begin{pmatrix} -13 \\ 20 \\ 9 \\ 0 \\ 16 \end{pmatrix}$$

Rank(A) = rank(RA)

=3

=number of dependent unknowns

Back to column space

$$A = \begin{pmatrix} 1 - 1 & 4 & 2 \\ 0 & 1 & 3 & 2 \\ 3 - 2158 \end{pmatrix}$$

Column space of A

$$C_{1}\begin{pmatrix}1\\0\\3\end{pmatrix}+C_{2}\begin{pmatrix}-1\\1\\-2\end{pmatrix}+C_{3}\begin{pmatrix}4\\3\\15\end{pmatrix}+C_{4}\begin{pmatrix}2\\2\\8\end{pmatrix}$$

Let

$$C_{1}\begin{pmatrix}1\\0\\3\end{pmatrix}+C_{2}\begin{pmatrix}-1\\1\\-2\end{pmatrix}+C_{3}\begin{pmatrix}4\\3\\15\end{pmatrix}+C_{4}\begin{pmatrix}2\\2\\8\end{pmatrix}=0$$

$$\begin{pmatrix}1-1 & 4 & 2\\0 & 1 & 3 & 2\\3-215 & 8\end{pmatrix}\begin{pmatrix}C_{1}\\C_{2}\\C_{3}\\C_{4}\end{pmatrix}=0$$

Homogeneous linear equation Reduced system is

$$\begin{pmatrix} 1074 \\ 0132 \\ 0000 \end{pmatrix} \begin{pmatrix} C_1 \\ C_2 \\ C_3 \\ C_4 \end{pmatrix} = 0$$

Result to 
$$C_1 = -7C_3 - 4C_4$$
  
 $C_2 = -3C_3 - 2C_4$   
 $C_1\vec{F}_1 + C_2\vec{F}_2 = (-7C_3 - 4C_4)\vec{F}_1 + (-3C_3 - 2C_4)\vec{F}_2$   
 $= -C_3(7\vec{F}_1 + 3\vec{F}_2) - C_4(4\vec{F}_1 + 2\vec{F}_2) = -C_3\vec{F}_3 - C_4\vec{F}_4$ 

Result to  $\vec{F}_1, \vec{F}_2, \vec{F}_3, \vec{F}_4$  linearly dependent

$$\vec{F}_3 = 7\vec{F}_1 + 3\vec{F}_2$$
  
 $\vec{F}_4 = 4\vec{F}_1 + 2\vec{F}_2$ 

Column space of A is

$$\alpha \begin{pmatrix} 1 \\ 0 \\ 3 \end{pmatrix} + \beta \begin{pmatrix} -1 \\ 1 \\ -2 \end{pmatrix}$$

$$\begin{pmatrix} 1\\0\\3 \end{pmatrix}, \begin{pmatrix} -1\\1\\-2 \end{pmatrix}$$

form the basis of column space of, Dimension of column

space of A=2

# 6.6 The solution space of AX=0

1. solution space

A is a nxm matrix and AX=0

Any solution (mx1 column matrix) of AX=0can be thought of as a vector in RM. Then solutions of AX=0 are a subspace of RM and called the solution space of system AX=0

pf:

The set of solution of AX=0 AX<sub>1</sub>=0 AX<sub>2</sub>=0 A(X<sub>1</sub>+X<sub>2</sub>)=AX<sub>1</sub>+AX<sub>2</sub>=0 A sum of solutions is a solution A( $\alpha$  X<sub>1</sub>)=  $\alpha$  (AX<sub>2</sub>)=0

Scalar multiplication of a solution is a solution.

#### 2.Dimension of solution space

Let A be an nxm matrix of real number.

Then the solution space of AX=0 has dimension m=rank(A)

pf:

general solution of AX=0=linear combination of linearly independent vectors..

No. of linearly independent vectors

= No. of arbitrary constants

= m-rank(A)

No. of linearly independent vectors

= a basis of solution space

= dimension of solution space

Ex. System AX=0 
$$\begin{bmatrix} -1 & 0 & 1 & 1 & 2 \\ 0 & 1 & 3 & 0 & 4 \\ 1 & 2 & 1 & 1 & 1 \\ -3 & 1 & 0 & 0 & 4 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ X_5 \end{bmatrix} = 0, \text{ the solution space}=?$$
  
Sol. A<sub>R</sub>=
$$\begin{bmatrix} 1 & 0 & 0 & 0 & -\frac{9}{8} \\ 0 & 1 & 0 & 0 & \frac{5}{8} \\ 0 & 0 & 1 & 0 & \frac{9}{8} \\ 0 & 0 & 0 & 1 & \frac{-1}{4} \end{bmatrix}$$
  
 $\therefore m = 5 \quad \text{rank}(A) = \text{rank}(A_R) = 4$ 

 $\therefore$  Dimension of solution space of AX=0

m-rank(A)=5-4=1

General solution is

$$\begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ X_5 \end{bmatrix} = \begin{bmatrix} \frac{-9}{8} \\ \frac{5}{8} \\ \frac{9}{8} \\ \frac{-1}{4} \\ 1 \end{bmatrix}$$
 One arbitrary constant.

3. Trivial solution

The solution  $X_1 = X_2 = \dots = X_n = 0$  is the trivial solution of a homogeneous system AX=0.

4. Existence of nontrivial solution

(1)Theorem

Let A be  $n \times m$  matrix. The system AX=0 has a nontrivial solution iff

m>rank(A).

:: Dimension of solution space of AX=0=m-rank(A)>0

Then system AX=0 has a nonzero solution!

(2)Corollary

A. AX=0 always has a nontrivial solution if the number of unknowns exceed the number of equations.

A.  $A_{n \times m} X_{m \times 1} = 0$ , n is the number of equations; m is the number of unknowns.  $\therefore$  n<m and rank(A)  $\le$  n. $\therefore$  m-rank(A)  $\ge$  m-n>0. B. A is an n×m matrix of real number. Then the system AX=0 has only the trivial solution iff  $A_R=I_n$ 

$$:: A_R = I_n$$

 $\therefore$  rank(A)=rank(A<sub>R</sub>)=n, and m=n.

 $\therefore$  m-rank(A)n-n=0

 $\Rightarrow$  System AX=0 has only zero solution.

# 6-7 Nonhomogeneous Systems of Linear Equations

1. Nonhomogeneous linear system

 $\begin{cases}
a_{11}x_1 + a_{12}x_2 + \dots + a_{1m}x_m = b_1 \\
a_{21}x_1 + a_{22}x_2 + \dots + a_{2m}x_m = b_2 \\
\dots \\
a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nm}x_m = b_n
\end{cases}$ 

In matrix from AX=B

When  $A = [a_{ij}] = n \times m$  coefficient matrix

 $X=[X_i]=m \times 1$  unknowns

 $B = [b_i] = n \times 1$  matrix (at least one of which is nonzero).

2.Structure of solution of AX=B

(1)Theorem

Let Up be a solution of AX=B, then every solution of AX=Bi the form of  $U_p$ +H. in which H is a solution of AX=0.

 $\begin{array}{l} [Proof] \\ Assume \boxed{AW=B} & \because \boxed{AU_p=B} \\ & \therefore AW=AU_p=A(W-U_P)=B-B=0 \\ & \therefore W-U_P=H \text{ is the solution of } AX=0 \\ & \therefore W=U_P+H \end{array}$ 

(2) Remark on  $U_P$  and H

A.  $U_P$ =particular solution of AX=0

- B. H=general solution of AX=0
- C.  $U_P$ +H=general solution of AX=0
- 3. Existence and Uniquences of solutions
  - (1) Consistent system of equation
    - Definition:

AX=B is said to be consistent if there exists a solution, otherwise the system is inconsistent.

- (2) Reduced form of Augmented matrix  $[A:B]_{12} = [AR:C]$
- (3) Existence of a solution
  - Theorem:

System AX=B has a solution iff A and [A:B] have the same rank.

Rank(A)=rank([A:B])

(4)Uniquence of a solution

Let A be  $n \times n$ . Then system AX=B have a unique solution iff

$$A_{R}=In \text{ or } rank(A)=n$$

$$A_{n \times n}X=B \Rightarrow A_{R}X=C$$

$$InX=C$$

4. Procedure for solving AX=B

Step 1. Find reduced matrix of  $[A:B] \rightarrow [A_R:C]$ 

Step 2. Check if rank(A)=rank([A:B]) for existence of solution.

Step3. Identify the dependent unknowns(作法同 system AX=0)

Step4. Find general solution by assigning independent unknowns any constant.

Ex. 
$$\begin{bmatrix} -1 & 1 & 3 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix} = \begin{bmatrix} -2 \\ 4 \end{bmatrix}$$
  
Sol. 
$$[A:B]_R = \begin{bmatrix} -1 & 1 & 3 & -2 \\ 0 & 1 & 2 & 4 \end{bmatrix}$$
, reduced form  $[A:B]$ 
$$[A:B]_R = \begin{bmatrix} 1 & 0 & -1 & 6 \\ 0 & 1 & 2 & 4 \end{bmatrix} = [A_R:C]$$
rank(A)=rank(A\_R)=2  
rank([A:B])=rank([A\_R:C])=2  
 $\therefore$  rank(A)=rank([A:B])  
 $\Rightarrow$  consistent system  $\Rightarrow$  a solution exists.  
From A<sub>R</sub>, we have  
X<sub>1</sub> and X<sub>2</sub>= dependent unknowns  
X<sub>3</sub> = independent unknowns.

Ex. 
$$\begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix} = \begin{bmatrix} 6 \\ 4 \end{bmatrix}$$

Sol.  $A_R X = C$   $\therefore X_1 = X_3 + 6$   $X_2 = -2X_3 + 4$  $\Rightarrow X = \begin{bmatrix} X_1 \\ X_2 \\ X_3 \end{bmatrix} = \begin{bmatrix} X_3 + 6 \\ -2X_3 + 4 \\ X_3 \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} X_3 + \begin{bmatrix} 6 \\ 4 \\ 0 \end{bmatrix}$ 

Let  $X_3 = \alpha$ , general solution is

$$X = \alpha \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} + \begin{bmatrix} 6 \\ 4 \\ 0 \end{bmatrix}$$
$$= H + U_P$$
H: general solution of AX=0
$$U_P$$
: particular solution of AX=B

# 6-9 Matrix Inverses

Inverse of a square matrix. If AB=BA, then B is an inverse matrix of A.

(1)Definition of nonsingular matrix

A matrix that has an inverse is called nonsingular; otherwise called singular.

(2)Theorem of uniquence of an inverse

A nonsingular matrix has exactly one inverse.

$$[Proof] \begin{cases} BA = I \\ CA = I \end{cases} \Rightarrow B=C \\ B=BI=BAC=IC=C \end{cases}$$

(3) Theorem of existence of an inverse matrix An  $\times$  m is nonsingular iff  $A_R$ =In

#### 2. Properties

Let A and B be nonsingular matrices

(1) 
$$(AB)^{-1} = (BA)^{-1}$$
  
(2)  $(A^{-1})^{-1} = A$   
(3)  $(A^{t})^{-1} = (A^{-1})^{t}$ 

(4)  $(In)^{-1} = In$ 

(5) AB and BA are singular if A and B are  $n \times n$  matrix and either is singular.

## 3.Method for Finding Inverse

(2) Row operation method [In: A] 
$$\rightarrow$$
 [ $\Omega$ :  $A_R$ ]=[A<sup>-1</sup>: In]  
 $\Omega$  A=A<sub>R</sub>, If A<sub>R</sub>=In  $\therefore \Omega$ =A<sup>-1</sup> ( $\Omega$  A=In)  
Ex. A= $\begin{bmatrix} 5 & -1 \\ 6 & 8 \end{bmatrix}$ , A<sup>-1</sup>=?  
Sol. [I<sub>2</sub>: A]= $\begin{bmatrix} 1 & 0 & | & 5 & -1 \\ 0 & 1 & | & 6 & 8 \end{bmatrix}$   $\xrightarrow{row\_operation}$   $\begin{bmatrix} \frac{8}{46} & \frac{1}{46} & | & 1 & 0 \\ \frac{-6}{46} & \frac{5}{46} & | & 0 & 1 \end{bmatrix}$ =[ $\Omega$ : A<sub>R</sub>]=[A<sup>-1</sup>: I<sub>2</sub>]  
 $\therefore$  A<sup>-1</sup>= $\Omega$ = $\begin{bmatrix} \frac{8}{46} & \frac{1}{46} \\ \frac{-6}{46} & \frac{5}{46} \end{bmatrix}$ 

(2) Determinants method (refer to section 7.7)

4. Use the Inverse to solve linear system
Let A be an n×n matrix. Then AX=B has a solution iff A is nonsingular.
∴ A is nonsingular ∴ A has inverse A<sup>-1</sup>
⇒ A<sup>-1</sup>AX=A<sup>-1</sup>B X=A<sup>-1</sup>B