## **Unit one: Matrices**

#### **1.1 Basic Definitions:**

A **matrix** is a rectangular array of numbers (or functions) enclosed in brackets. These numbers (or functions) are called the **entries** (or sometimes the *elements*) of the matrix. For example,

$$\begin{bmatrix} 0.3 & 1 & -5 \\ 0 & -0.2 & 16 \end{bmatrix}, \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}, \begin{bmatrix} e^{-x} & 2x^2 \\ e^{6x} & 4x \end{bmatrix}, [a_1 & a_2 & a_3], \begin{bmatrix} 4 \\ \frac{1}{2} \end{bmatrix}$$
(1)

The first matrix has two **rows** (horizontal lines of entries) and three **columns** (vertical lines). The second and third matrices are square matrices. Matrices having just a single row or column are called vectors. Thus the fourth and last matrices are called **row** and **column** vectors, respectively.

We shall denote matrices by capital letters A, B, C, ..., or  $A = [a_{ij}]$ 

 $m \times n$  is called the size of the matrix.

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

If m = n, we call **A** an  $n \times n$  square matrix. A matrix that is not square is called a rectangular matrix.

#### **1.2 Equality of Matrices:**

Two matrices  $\mathbf{A} = [a_{jk}]$  and  $\mathbf{B} = [b_{jk}]$  are equal (written  $\mathbf{A} = \mathbf{B}$ ) if and only if they have the same size, and the corresponding elements are equal, that is,  $a_{11} = b_{11}$ ,  $a_{12} = b_{12}$ , and so on. Matrices that are not equal are called **different**.

#### Example: Equality of Matrices

Let

Then

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \quad \text{and} \quad \mathbf{B} = \begin{bmatrix} 4 & 0 \\ 3 & -1 \end{bmatrix}.$$
$$\mathbf{A} = \mathbf{B} \quad \text{if and only if} \quad \begin{aligned} a_{11} = 4, \quad a_{12} = 0, \\ a_{21} = 3, \quad a_{22} = -1. \end{aligned}$$

#### **1.3 Addition of Matrices**

The sum of two matrices  $\mathbf{A} = [a_{ij}]$  and  $\mathbf{B} = [b_{ij}]$  of the same size is written  $\mathbf{A} + \mathbf{B}$  and has the entries  $a_{ij}+b_{ij}$  obtained by adding the corresponding entries of  $\mathbf{A}$  and  $\mathbf{B}$ . Matrices of different sizes cannot be added.

#### **Example:**

If  $\mathbf{A} = \begin{bmatrix} -4 & 6 & 3 \\ 0 & 1 & 2 \end{bmatrix}$  and  $\mathbf{B} = \begin{bmatrix} 5 & -1 & 0 \\ 3 & 1 & 0 \end{bmatrix}$ , then  $\mathbf{A} + \mathbf{B} = \begin{bmatrix} 1 & 5 & 3 \\ 3 & 2 & 2 \end{bmatrix}$ .

#### **1.4 Scalar Multiplication (Multiplication by a Number)**

The product of any  $m \times n$  matrix  $\mathbf{A} = [a_{ij}]$  and any scalar c (number c) is written  $c\mathbf{A}$  and is the  $m \times n$  matrix  $c\mathbf{A} = [ca_{jk}]$  obtained by multiplying each entry of  $\mathbf{A}$  by  $\mathbf{c}$ .

Example:

If 
$$\mathbf{A} = \begin{bmatrix} 2.7 & -1.8 \\ 0 & 0.9 \\ 9.0 & -4.5 \end{bmatrix}$$
, then  $-\mathbf{A} = \begin{bmatrix} -2.7 & 1.8 \\ 0 & -0.9 \\ -9.0 & 4.5 \end{bmatrix}$ ,  $\frac{10}{9}\mathbf{A} = \begin{bmatrix} 3 & -2 \\ 0 & 1 \\ 10 & -5 \end{bmatrix}$ ,  $0\mathbf{A} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$ .

### **Rules for Matrix Addition and Scalar Multiplication:**

(a) A + B = B + A(b) (A + B) + C = A + (B + C)(c) A + 0 = A(d) A + (-A) = 0(a) c(A + B) = cA + cB(b) (c + k)A = cA + kA(c) c(kA) = (ck)A(d) IA = A

### **1.4 Matrix multiplication**

The product  $\mathbf{C} = \mathbf{A}\mathbf{B}$  (in this order) of an  $m \times n$  matrix  $\mathbf{A} = [a_{ij}]$  times an  $r \times p$  matrix  $\mathbf{B} = [b_{ij}]$  is defined if and only if r = n and is then the  $m \times p$  matrix  $\mathbf{C} = [c_{ij}]$  with entries

$$c_{ij} = \sum_{l=1}^{n} a_{il} b_{lj} = a_{i1} b_{1j} + a_{i2} b_{2j} + \dots + a_{in} b_{nj} \qquad i = 1, \dots, m$$
$$j = 1, \dots, p$$

 $\mathbf{A} \qquad \mathbf{B} = \mathbf{C}$  $[m \times n] \qquad [n \times p] \qquad [m \times p]$ 

Example:

$$\mathbf{AB} = \begin{bmatrix} 3 & 5 & -1 \\ 4 & 0 & 2 \\ -6 & -3 & 2 \end{bmatrix} \begin{bmatrix} 2 & -2 & 3 & 1 \\ 5 & 0 & 7 & 8 \\ 9 & -4 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 22 & -2 & 43 & 42 \\ 26 & -16 & 14 & 6 \\ -9 & 4 & -37 & -28 \end{bmatrix}$$

Here  $c_{11} = 3 \cdot 2 + 5 \cdot 5 + (-1) \cdot 9 = 22$ , and so on. The entry in the box is  $c_{23} = 4 \cdot 3 + 0 \cdot 7 + 2 \cdot 1 = 14$ . The product **BA** is not defined.

*Note: Matrix Multiplication is Not Commutative,*  $AB \neq BA$  *in general.* 

#### **Rules of Matrix Multiplication:**

- (a)  $(k\mathbf{A})\mathbf{B} = k(\mathbf{A}\mathbf{B}) = \mathbf{A}(k\mathbf{B})$
- (b)  $\mathbf{A}(\mathbf{B}\mathbf{C}) = (\mathbf{A}\mathbf{B})\mathbf{C}$
- (c)  $(\mathbf{A} + \mathbf{B})\mathbf{C} = \mathbf{A}\mathbf{C} + \mathbf{B}\mathbf{C}$
- (d)  $\mathbf{C}(\mathbf{A} + \mathbf{B}) = \mathbf{C}\mathbf{A} + \mathbf{C}\mathbf{B}$

### **1.5 Transpose of a Matrix**

Given any  $m \times n$  matrix  $\mathbf{A} = [a_{ij}]$ , we define the transpose of  $\mathbf{A}$ , denoted as  $\mathbf{A}^{\mathrm{T}}$  and read as "**A**-transpose" as

$$\mathbf{A}^{\mathrm{T}} = [a_{ij}]^{\mathrm{T}} = [a_{ji}] = \begin{bmatrix} a_{11} & a_{21} & \cdots & \cdots & a_{m1} \\ a_{12} & a_{22} & \cdots & \cdots & a_{m2} \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ a_{1n} & a_{2n} & \cdots & \cdots & \vdots & a_{mn} \end{bmatrix}$$
  
Example: 
$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix}^{\mathrm{T}} = \begin{bmatrix} 1 & 3 & 5 \\ 2 & 4 & 6 \end{bmatrix}$$

# **1.6 Types of Matrices:**

| Type of Matrix             | Details  | Example  |  |  |  |  |
|----------------------------|--|--|--|--|--|--|
| Row Matrix                 | $\mathbf{A} = [a_{ij}]_{1 \times n}$   | [0 1 1 -2]   |  |  |  |  |
| Column Matrix              | $\mathbf{A} = [a_{ij}]_{m \times 1}$   | $\begin{bmatrix} -1\\ 2\\ -4\\ 5\end{bmatrix}$   |  |  |  |  |
| Zero or Null Matrix        | $\mathbf{A} = [a_{ij}]_{m \times n}$ where, $a_{ij} = 0$   | $\begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$  |  |  |  |  |
| Singleton Matrix           | $\mathbf{A} = [a_{ij}]_{m \times n}$ where, $m = n = 1$  | [2]  |  |  |  |  |
| Horizontal Matrix          | $[a_{ij}]_{m \times n}$ where $n > m$  | $\begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 5 & 1 & 1 \end{bmatrix}$   |  |  |  |  |
| Vertical Matrix            | $[a_{ij}]_{m \times n}$ where, $m > n$   | $\begin{bmatrix} 2 & 5 \\ 1 & 1 \\ 3 & 6 \\ 2 & 4 \end{bmatrix}$   |  |  |  |  |
| Square Matrix              | $[a_{ij}]_{m \times n}$ where, $m = n$   | $\begin{bmatrix} 4 & 7 \\ 9 & 13 \end{bmatrix}$  |  |  |  |  |
| Diagonal Matrix            | $\mathbf{A} = [a_{ij}]$ , $a_{ij} = 0$ , when $i \neq j$   | $\begin{bmatrix} 2 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 4 \end{bmatrix}$  |  |  |  |  |
| Identity (Unit) Matrix (I) | $\mathbf{A} = [a_{ij}]_{m 	imes n} 	ext{ where,} \ a_{ij} = egin{cases} 1, & i=j \ 0, & i  eq j \end{cases}$                             | $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$  |  |  |  |  |
| Triangular Matrix          | Can be either upper triangular<br>$(a_{ij} = 0, \text{ when } i > j) \text{ or lower}$<br>triangular $(a_{ij} = 0 \text{ when } i < j)$  | $\begin{bmatrix} 3 & 1 & 2 \\ 0 & 4 & 3 \\ 0 & 0 & 6 \end{bmatrix} \qquad \begin{bmatrix} 1 & 0 & 0 \\ 2 & 3 & 0 \\ 4 & 5 & 2 \end{bmatrix}$ Upper Matrix Lower Matrix |  |  |  |  |
| Singular Matrix            | $ \mathbf{A}  = 0$   |  |  |  |  |  |
| Non-Singular Matrix        | $ \mathbf{A}  \neq 0$  |  |  |  |  |  |
| Symmetric Matrix           | $\mathbf{A} = [a_{ij}] \text{ where, } a_{ij} = a_{ji}$<br>or $\mathbf{A} = \mathbf{A}^{\mathrm{T}}$<br>(A is a square matrix)           | $\begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 5 \\ 3 & 5 & 2 \end{pmatrix}$  |  |  |  |  |
| Skew-Symmetric Matrix      | $\mathbf{A} = [a_{ij}] \text{ where, } a_{ij} = -a_{ji}$<br>or $\mathbf{A} = -\mathbf{A}^{\mathrm{T}}$<br>( <b>A</b> is a square matrix) | $\begin{bmatrix} 0 & 2 & 1 \\ -2 & 0 & -3 \\ -1 & 3 & 0 \end{bmatrix}$   |  |  |  |  |

### **1.7 Determinants**

The scalar quantity associated with a square matrix, is called "determinant". We denote the determinant of an  $n \times n$  matrix  $\mathbf{A} = [a_{ij}]$  as

$$\det \mathbf{A} = \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} = \sum_{j=1}^{n} a_{ij} C_{ij}$$

$$|\mathbf{A}| = a_{i1} A_{i1} + a_{i2} A_{i2} + \dots + a_{in} A_{in} = \sum_{j=1}^{n} a_{ij} C_{ij} \quad \text{(row definition)}$$

$$|\mathbf{A}| = a_{1j} A_{1j} + a_{2j} A_{2j} + \dots + a_{nj} A_{nj} = \sum_{i=1}^{n} a_{ij} C_{ij} \quad \text{(column definition)}$$

where  $C_{ij}$  is called the cofactor of the  $a_{ij}$  element and is defined as

$$C_{ij} \equiv (-1)^{i+j} \operatorname{M}_{ij}$$

where  $M_{ij}$  is called the minor of  $a_{ij}$ , namely the determinant of  $(n-1) \times (n-1)$  matrix that survives when the row and column containing  $a_{ij}$  are struck out.

### **Special Cases in Determinants:**

1- Determinant of order 1: The determinant of a matrix of the first order (*n*=1) is the element itself.

Example: If A = [-1], then detA = -1.

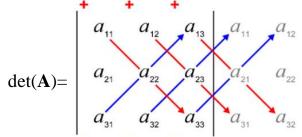
2- Determinant of order 2:

$$det \begin{bmatrix} a & c \\ b & d \end{bmatrix} = (a)(d) - (b)(c)$$
$$det \begin{bmatrix} 3 & 5 \\ 4 & 6 \end{bmatrix} = (3)(6) - (4)(5) = 18 - 20 = -2$$

3- Determinant of order 3:(a) The Leibniz formula:

$$egin{array}{ccc} a & b & c \ d & e & f \ g & h & i \end{array} = a(ei-fh) - b(di-fg) + c(dh-eg) \ = aei + bfg + cdh - ceg - bdi - afh.$$

(b) The rule of Sarrus:



 $\det(\mathbf{A}) = a_{11} a_{22} a_{33} + a_{12}a_{23}a_{31} + a_{13}a_{21}a_{32} - a_{31}a_{22}a_{13} - a_{32}a_{23}a_{11} - a_{33}a_{21}a_{12}$ 

Notes:

- i. The scheme of Sarrus for calculating the determinant of a  $3 \times 3$  matrix does not carry over into higher dimensions.
- ii. The Leibniz formula expresses the determinant of an  $n \times n$ -matrix in a manner which is consistent across higher dimensions.
  - 4- If the matrix is upper or lower triangular matrix, the determinant of the matrix is the product of the elements on the main diagonal.

## Some Properties of determinants:

1- If any row or column of a determinant det**A** only contains zero elements, then det $\mathbf{A} = 0$ .

2- If **A** is a square matrix with the transpose  $\mathbf{A}^{\mathrm{T}}$ , then det  $\mathbf{A} = \det \mathbf{A}^{\mathrm{T}}$ .

3- If each element of a row or column of a square matrix A is multiplied by a

constant k, then the value of the determinant is kdet**A**.

4- If two rows (or columns) of a square matrix are interchanged, the sign of the determinant is changed.

5- If any two rows or columns of a square matrix  $\mathbf{A}$  are proportional, then  $\det \mathbf{A} = 0$ .

**Example 1:** Evaluate the determinant of the given matrix by cofactor expansion.

 $\begin{pmatrix} 2 & 1 & -2 & 1 \\ 0 & 5 & 0 & 4 \\ 1 & 6 & 1 & 0 \\ 5 & -1 & 1 & 1 \end{pmatrix}$ Solution:  $|\mathbf{A}| = \begin{vmatrix} 2 & 1 & -2 & 1 \\ 0 & 5 & 0 & 4 \\ 1 & 6 & 1 & 0 \\ 5 & -1 & 1 & 1 \end{vmatrix} = 5 \begin{vmatrix} 2 & -2 & 1 \\ 1 & 1 & 0 \\ 5 & 1 & 1 \end{vmatrix} + 4 \begin{vmatrix} 2 & 1 & -2 \\ 1 & 6 & 1 \\ 5 & -1 & 1 \end{vmatrix} = 5(0) + 4(80) = 320$ 

**Example 2:** Explain the following results:

| 1 | <b>2</b> | 3        | 4   |   |   | 6 |       |       |   | 1        |   |       |
|---|----------|----------|---|---|---|---|-------|-------|---|----------|---|-------|
| 2 | <b>5</b> | 7        | $\begin{vmatrix} 3 \\ 6 \end{vmatrix} = 0,$ | 4 | 7 | 3 | 2     | 0 and | 6 | 7        | 6 | 9     |
| 4 | 10       | 14       | 6 = 0,                                      | 0 | 0 | 0 | 0 = 0 |       | 0 | 6        | 0 | 0 = 0 |
| 3 | 4        | <b>2</b> | 7   |   |   |   | 9     |       |   | <b>2</b> |   |       |

#### **1.8 Matrix Partitioning:**

The idea of matrix partitioning is that any matrix **A** (which is larger than  $1 \times 1$ ) may be portioned into a number of smaller matrices called *blocks*.

Example: The matrix

$$A = \begin{bmatrix} 3 & 0 & -1 & 5 & 9 & -2 \\ -5 & 2 & 4 & 0 & -3 & 1 \\ -8 & -6 & 3 & 1 & 7 & -4 \end{bmatrix}$$

Can also be written as the 2 × 3 partitioned (or block) matrix

$$A = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \end{bmatrix}$$

• Whose entries are the *blocks* (or *submatrices*)

$$A_{11} = \begin{bmatrix} 3 & 0 & -1 \\ -5 & 2 & 4 \end{bmatrix}, \quad A_{12} = \begin{bmatrix} 5 & 9 \\ 0 & -3 \end{bmatrix}, \quad A_{13} = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$$
$$A_{21} = \begin{bmatrix} -8 & -6 & 3 \end{bmatrix}, \quad A_{22} = \begin{bmatrix} 1 & 7 \end{bmatrix}, \quad A_{23} = \begin{bmatrix} -4 \end{bmatrix}$$

Example : Find AB, where

$$A = \begin{bmatrix} 2 & -3 & 1 & 0 & -4 \\ 1 & 5 & -2 & 3 & -1 \\ 0 & -4 & -2 & 7 & -1 \end{bmatrix} , B = \begin{bmatrix} 6 & 4 \\ -2 & 1 \\ -3 & 7 \\ -1 & 3 \\ 5 & 2 \end{bmatrix}$$

Solution:

$$A = \begin{bmatrix} 2 & -3 & 1 & | & 0 & -4 \\ 1 & 5 & -2 & 3 & -1 \\ \hline 0 & -4 & -2 & | & 7 & -1 \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}, B = \begin{bmatrix} 0 & 4 \\ -2 & 1 \\ \hline -3 & 7 \\ \hline -1 & 3 \\ 5 & 2 \end{bmatrix} = \begin{bmatrix} B_1 \\ B_2 \end{bmatrix}$$
$$AB = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} = \begin{bmatrix} A_{11}B_1 + A_{12}B_2 \\ A_{21}B_1 + A_{22}B_2 \end{bmatrix}$$
$$= \begin{bmatrix} -5 & 4 \\ -6 & 2 \\ \hline 2 & 1 \end{bmatrix}$$

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### **1.9 Inverse of a Matrix**

If **A** is a non-singular square matrix, there is an existence of  $n \times n$  matrix **A**<sup>-1</sup>, which is called the inverse matrix of **A**, such that it satisfies the property:

$$\mathbf{A}^{-1} = \frac{1}{A}$$
 and  $\mathbf{A}\mathbf{A}^{-1} = \mathbf{A}^{-1}\mathbf{A} = \mathbf{I}$ 

where **I** is the Identity matrix. The inverse of a nonsingular  $n \times n$  matrix  $\mathbf{A} = [a_{ij}]$  is:

$$\mathbf{A}^{-1} = \frac{1}{|A|} \begin{bmatrix} C_{ij} \end{bmatrix}^T = \frac{1}{|A|} \begin{bmatrix} c_{11} & c_{21} & \dots & c_{n1} \\ c_{12} & c_{22} & \dots & c_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ c_{1n} & c_{2n} & \dots & c_{nn} \end{bmatrix} = \frac{\operatorname{adj}(A)}{|A|}$$

where  $[C_{ij}]^T$  is the transpose of cofactors matrix of **A**,  $adj(\mathbf{A})$  is the adjoint matrix of **A**.

### **Properties of the Inverse:**

Let A and B be nonsingular matrices. Then

- (i)  $(\mathbf{A}^{-1})^{-1} = \mathbf{A}$
- (ii)  $(AB)^{-1} = B^{-1}A^{-1}$
- (iii)  $(\mathbf{A}^T)^{-1} = (\mathbf{A}^{-1})^T$

Example: Find the inverse of  $\mathbf{A} = \begin{pmatrix} 2 & 2 & 0 \\ -2 & 1 & 1 \\ 3 & 0 & 1 \end{pmatrix}$ .

**SOLUTION** Since det A = 12, we can find  $A^{-1}$  from (5). The cofactors corresponding to the entries in A are

$$C_{11} = \begin{vmatrix} 1 & 1 \\ 0 & 1 \end{vmatrix} = 1 \qquad C_{12} = -\begin{vmatrix} -2 & 1 \\ 3 & 1 \end{vmatrix} = 5 \qquad C_{13} = \begin{vmatrix} -2 & 1 \\ 3 & 0 \end{vmatrix} = -3$$
$$C_{21} = -\begin{vmatrix} 2 & 0 \\ 0 & 1 \end{vmatrix} = -2 \qquad C_{22} = \begin{vmatrix} 2 & 0 \\ 3 & 1 \end{vmatrix} = 2 \qquad C_{23} = -\begin{vmatrix} 2 & 2 \\ 3 & 0 \end{vmatrix} = 6$$
$$C_{31} = \begin{vmatrix} 2 & 0 \\ 1 & 1 \end{vmatrix} = 2 \qquad C_{32} = -\begin{vmatrix} 2 & 0 \\ -2 & 1 \end{vmatrix} = -2 \qquad C_{33} = \begin{vmatrix} 2 & 2 \\ -2 & 1 \end{vmatrix} = 6.$$

From (5) we then obtain

$$\mathbf{A}^{-1} = \frac{1}{12} \begin{pmatrix} 1 & -2 & 2\\ 5 & 2 & -2\\ -3 & 6 & 6 \end{pmatrix} = \begin{pmatrix} \frac{1}{12} & -\frac{1}{6} & \frac{1}{6}\\ \frac{5}{12} & \frac{1}{6} & -\frac{1}{6}\\ -\frac{1}{4} & \frac{1}{2} & \frac{1}{2} \end{pmatrix}.$$

#### **1.10 Solution of linear simultaneous equations:**

A system of *m* linear equations in *n* variables  $x_1, x_2, ..., x_n$ ,

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

$$\vdots$$
(1)

 $a_{m1}x_1 + a_{m2}x_2 + \ldots + a_{mn}x_n = b_m$ 

can be written completely as a matrix **AX=B**, where

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}, \quad \mathbf{X} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ \vdots \\ x_n \end{bmatrix}, \qquad \mathbf{B} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ \vdots \\ b_m \end{bmatrix}$$

Let us suppose that m = n in the equation above, so that the coefficient matrix **A** is  $n \times n$ . In particular, if **A** is nonsingular, then the system  $\mathbf{AX} = \mathbf{B}$  can be solved by:

$$\mathbf{X} = \mathbf{A}^{-1} \mathbf{B}$$

Where **X** is variable matrix, **A** coefficients matrix and **B** constant matrix. (Note: if det **A**=0, the system of linear equations is called **inconsistent**, i.e. there is no unique solution).

**Example:** Use the inverse of the coefficient matrix to solve the system

$$2x_1 + x_3 = 2$$
  
-2x<sub>1</sub> + 3x<sub>2</sub> + 4x<sub>3</sub> = 4  
-5x<sub>1</sub> + 5x<sub>2</sub> + 6x<sub>3</sub> = -1

**Solution**:

$$\mathbf{A}^{-1} = \begin{pmatrix} -2 & 5 & -3 \\ -8 & 17 & -10 \\ 5 & -10 & 6 \end{pmatrix} \qquad (Home \ work \ !)$$

Thus

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 2 & 0 & 1 \\ -2 & 3 & 4 \\ -5 & 5 & 6 \end{pmatrix}^{-1} \begin{pmatrix} 2 \\ 4 \\ -1 \end{pmatrix} = \begin{pmatrix} -2 & 5 & -3 \\ -8 & 17 & -10 \\ 5 & -10 & 6 \end{pmatrix} \begin{pmatrix} 2 \\ 4 \\ -1 \end{pmatrix} = \begin{pmatrix} 19 \\ 62 \\ -36 \end{pmatrix}$$

Consequently,  $x_1 = 19$ ,  $x_2 = 62$ , and  $x_3 = -36$ .

In addition to the *inverse matrix* method shown above, there are two another widely used important methods:

## **1-Cramer's Rule**

Let **A** be the coefficient matrix of system (1). If det  $\mathbf{A} \neq 0$ , then the solution of (1) is given by

$$x_1 = \frac{\det A_1}{\det A}$$
,  $x_2 = \frac{\det A_2}{\det A}$ , ...,  $x_n = \frac{\det A_n}{\det A}$ 

where  $A_k$ , k = 1, 2, ..., n, is defined as:

$$\mathbf{A}_{k} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1 \ k-1} & b_{1} & a_{1 \ k+1} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2 \ k-1} & b_{2} & a_{2 \ k+1} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{n \ k-1} & b_{n} & a_{n \ k+1} & \dots & a_{nn} \end{bmatrix}$$

Example: Use Cramer's rule to solve the system

$$3x_1 + 2x_2 + x_3 = 7$$
  

$$x_1 - x_2 + 3x_3 = 3$$
  

$$5x_1 + 4x_2 - 2x_3 = 1.$$

**SOLUTION** The solution requires the evaluation of four determinants:

$$\det \mathbf{A} = \begin{vmatrix} 3 & 2 & 1 \\ 1 & -1 & 3 \\ 5 & 4 & -2 \end{vmatrix} = 13, \qquad \det \mathbf{A}_1 = \begin{vmatrix} 7 & 2 & 1 \\ 3 & -1 & 3 \\ 1 & 4 & -2 \end{vmatrix} = -39,$$
$$\det \mathbf{A}_2 = \begin{vmatrix} 3 & 7 & 1 \\ 1 & 3 & 3 \\ 5 & 1 & -2 \end{vmatrix} = 78, \qquad \det \mathbf{A}_3 = \begin{vmatrix} 3 & 2 & 7 \\ 1 & -1 & 3 \\ 5 & 4 & 1 \end{vmatrix} = 52.$$

Thus, (6) gives

$$x_1 = \frac{\det \mathbf{A}_1}{\det \mathbf{A}} = -3, \quad x_2 = \frac{\det \mathbf{A}_2}{\det \mathbf{A}} = 6, \quad x_3 = \frac{\det \mathbf{A}_3}{\det \mathbf{A}} = 4.$$

## **2- Gaussian Elimination Method:**

There are three types of elementary row operations which may be performed on the rows of a matrix:

- 1- Swap the positions of two rows.
- 2- Multiply a row by a non-zero scalar.
- 3- Add to one row a scalar multiple of another.

The solution is then found by the process called **back-substitution**.

**Example**: Solve the following linear system using Gaussian elimination:

$$2x_1 + 6x_2 + x_3 = 7$$
  

$$x_1 + 2x_2 - x_3 = -1$$
  

$$5x_1 + 7x_2 - 4x_3 = 9.$$

Solution: Using row operations on the augmented matrix of the system, we obtain:

$$\begin{pmatrix} 2 & 6 & 1 & | & 7 \\ 1 & 2 & -1 & | & -1 \\ 5 & 7 & -4 & | & 9 \end{pmatrix} \xrightarrow{R_1 \leftrightarrow R_2} \begin{pmatrix} 1 & 2 & -1 & | & -1 \\ 2 & 6 & 1 & | & 7 \\ 5 & 7 & -4 & | & 9 \end{pmatrix}$$

$$\xrightarrow{-2R_1 + R_2} = \begin{pmatrix} 1 & 2 & -1 & | & -1 \\ 0 & 2 & 3 & | & 9 \\ 0 & -3 & 1 & | & 14 \end{pmatrix} \xrightarrow{\frac{1}{2}R_2} = \begin{pmatrix} 1 & 2 & -1 & | & -1 \\ 0 & 1 & \frac{3}{2} & | & \frac{9}{2} \\ 0 & -3 & 1 & | & 14 \end{pmatrix} \xrightarrow{\frac{1}{2}R_2} = \begin{pmatrix} 1 & 2 & -1 & | & -1 \\ 0 & 1 & \frac{3}{2} & | & \frac{9}{2} \\ 0 & -3 & 1 & | & 14 \end{pmatrix} \xrightarrow{\frac{3}{2}R_2 + R_3} = \begin{pmatrix} 1 & 2 & -1 & | & -1 \\ 0 & 1 & \frac{3}{2} & | & \frac{9}{2} \\ 0 & 0 & \frac{11}{2} & | & \frac{9}{2} \\ \frac{55}{2} & \xrightarrow{\frac{21}{11}R_3} & \begin{pmatrix} 1 & 2 & -1 & | & -1 \\ 0 & 1 & \frac{3}{2} & | & \frac{9}{2} \\ 0 & 0 & 1 & | & 5 \end{pmatrix}.$$

Substituting  $x_3 = 5$  into the second equation gives  $x_2 = -3$ . Substituting both these values back into the first equation finally yields  $x_1 = 10$ .

#### **Home Works:**

Exercises 8.2: Problems 1-20 (Only use Gaussian elimination).

Exercises 8.4: All Problems.

Exercises 8.5: 1-16.

Exercises 8.7: Problems 1-11.