

MATRIX ALGEBRA

A **matrix** is a rectangular array of numbers. The numbers in the array are called **entries** in the matrix. The **order** or **dimension** or **size** of a matrix is described by specifying the **number of rows** and the **number of columns**.

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 0 \\ -1 & 4 \end{bmatrix} \quad \text{Here } A \text{ is a matrix of order } 3 \times 2, \text{ ie, 3 rows and 2 columns.}$$

$[2 \ 1 \ 0 \ -3]$ is a (1×4) – matrix.

$$\text{Let } A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}$$

Here A is a matrix of **order** $m \times n$. If $m = n$, then we say that A is a **square** matrix. Thus a square matrix is a matrix with **equal numbers of columns and rows**.

Notation: We write $A = [a_{ij}]_{m \times n}$ or simply $[a_{ij}]$.

a_{ij} – is the **entry** (or element) in row i and column j .

Row vector: a matrix of order $1 \times n$

Column vector: a matrix of order $m \times 1$.

$$\text{Eg: } \begin{matrix} [5 & 0 & 2 & 5 & -1] ; \\ 1 \times 5 \end{matrix} \quad \begin{matrix} \begin{bmatrix} 5 \\ 2 \\ 3 \\ -4 \end{bmatrix} \\ 4 \times 1 \end{matrix}$$

- A **zero** matrix is a matrix whose entries are all zero.

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \quad \text{Zero matrix of order } 3 \times 2$$

- Two matrices $A = [a_{ij}]$ and $B = [b_{ij}]$ are **equal** if they have the **same** order and the **corresponding entries are equal**. **Notation:** $A = B$.

- Consider $A = \begin{bmatrix} 2 & 1 \\ 3 & x \end{bmatrix}$, $B = \begin{bmatrix} 2 & 1 \\ 3 & 5 \end{bmatrix}$ and $C = \begin{bmatrix} 2 & 1 & 0 \\ 3 & 4 & 0 \end{bmatrix}$. If $x = 5$ then $A = B$. For all other values of x , $A \neq B$. A and C have **different** sizes. Therefore they are not equal matrices.

MATRIX OPERATIONS (§ 8.5)

THE SUM OF MATRICES

Suppose A and B are matrices of order $m \times n$. Let $A = [a_{ij}]$ and $B = [b_{ij}]$.

The **sum** $A + B$ is a matrix obtained by **adding the corresponding** entries.

Therefore, $A + B = C = [c_{ij}]$ where $c_{ij} = a_{ij} + b_{ij}$

- The order of $A + B$ is also $m \times n$.
- Two matrices **cannot be added** if their **sizes are not the same**.

Exercise: Read the examples from the text .

MULTIPLICATION BY A SCALAR

Suppose $A = [a_{ij}]$ is an $m \times n$ matrix. Let k be a real number (**scalar**) then **kA is a matrix** whose $(i, j)^{\text{th}}$ element is ka_{ij}

Example:

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

$$3A = \begin{bmatrix} 6 & 9 & 12 \\ 3 & -9 & 3 \end{bmatrix}$$

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

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

NOTATION: $-A$ means $(-1)A$
 $A - B$ means $A + (-1)B$

MATRIX MULTIPLICATION

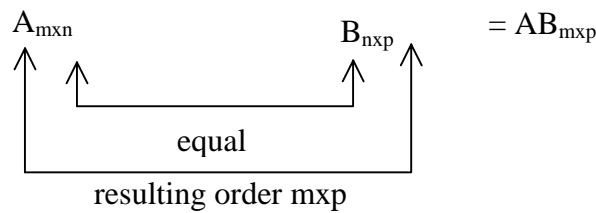
Suppose $A = [a_{ij}]$ is an $m \times n$ matrix and $B = [b_{ij}]$ is an $n \times p$ matrix.

The **product** AB is defined as a matrix $C = [c_{ij}]$ of order $m \times p$ where $c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}$

- The $(i,j)^{\text{th}}$ - entry of the matrix $C (= AB)$ is computed using the i^{th} row

$$[a_{i1} \ a_{i2} \ \dots \ a_{in}] \text{ of } A \text{ and the } j^{\text{th}} \text{ column } \begin{bmatrix} b_{1j} \\ b_{2j} \\ \dots \\ b_{nj} \end{bmatrix} \text{ of } B.$$

- The product AB of two matrices A and B is **only defined if the number of columns in A = the number of rows in B .**



- The order of AB is $m \times p$.

Example

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

A is a 2×3 matrix; B is a 3×4 matrix. Therefore AB is a 2×4 matrix.

$$AB = \begin{bmatrix} 1 & 2 & 4 \\ 2 & 6 & 0 \end{bmatrix} \begin{bmatrix} 4 & 1 & 4 & 3 \\ 0 & -1 & 3 & 1 \\ 2 & 7 & 5 & 2 \end{bmatrix} = \begin{bmatrix} \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & 26 & \cdot \end{bmatrix}$$

The (2,3)-entry $= (2 \times 4) + (6 \times 3) + (0 \times 5) = 26$.

The entry in row 1 and column 4 of AB is given by

$$(1 \times 3) + (2 \times 1) + (4 \times 2) = 13 \text{ etc.}$$

Thus $AB = \begin{bmatrix} 12 & 27 & 30 & 13 \\ 8 & -4 & 26 & 12 \end{bmatrix}$ (Verify). Here BA is **not** defined

Remark: Even when AB and BA are both defined for matrices A and B , **they need not be equal.**

For example, $A = \begin{bmatrix} 1 & 4 \\ 2 & 3 \end{bmatrix}$ $B = \begin{bmatrix} 2 & -1 \\ 3 & 0 \end{bmatrix}$ $AB = \begin{bmatrix} 14 & -1 \\ 13 & -2 \end{bmatrix}$ and $BA = \begin{bmatrix} 0 & 5 \\ 3 & 12 \end{bmatrix}$

PROPERTIES OF MATRIX OPERATIONS

Assume that the orders of the matrices are such that the following operations are defined.

1. $A + B = B + A$ (addition is commutative)
2. $A + (B + C) = (A + B) + C$ (addition is associative)
3. $A(BC) = (AB)C$ (multiplication is associative)
4. $(A + B)C = AC + BC$ (distributive)
5. $k(A + B) = kA + kB$

etc.

Read the text; pp 661-666

TRANSPOSE OF A MATRIX

Suppose $A = [a_{ij}]$ is an $m \times n$ matrix. Then the **transpose** A^T of A is the $n \times m$ matrix defined as

$$A^T = [a'_{ij}] \quad \text{where}$$

$$a'_{ij} = a_{ji} \quad (\text{the } (i,j)\text{-entry of } A^T \text{ is the } (j,i)\text{-entry of } A)$$

i.e. the rows of A^T are the columns of A .

e.g. Let $A = \begin{bmatrix} 2 & 3 & 1 \\ 4 & 2 & 3 \end{bmatrix}$ be a (2×3) -matrix. Then

$$A^T = \begin{bmatrix} 2 & 4 \\ 3 & 2 \\ 1 & 3 \end{bmatrix}. \quad \text{Its dimension is } 3 \times 2, \text{ i.e., it has 3 rows and 2 columns.}$$

Def: A square matrix A is said to be a **symmetric matrix** if $A^T = A$, and a **skew-symmetric matrix** if $A^T = -A$.

PROPERTIES OF THE TRANSPOSE

- (a) $(A^T)^T = A$
- (b) $(A + B)^T = A^T + B^T$
- (c) $(kA)^T = kA^T$, for any scalar k
- (d) $(AB)^T = B^T A^T$

Proof of (d): Suppose $A = [a_{ij}]_{m \times n}$, $B = [b_{ij}]_{n \times p}$ and $C = [c_{ij}]_{m \times p}$ such that $C = AB$. Then

$$c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}.$$

$$C^T = [c'_{ij}] \text{ where } c'_{ij} = c_{ji}. \text{ The } (i,j) \text{ entry in } C^T = c'_{ij} = c_{ji} = \sum_{k=1}^n a_{jk} b_{ki}.$$

$$\begin{aligned} \text{Now the } (i,j) \text{ entry in } B^T A^T &= \sum_{k=1}^n b'_{ik} a'_{kj} \\ &= \sum_{k=1}^n b_{ki} a_{jk} \\ &= (i,j) \text{ entry in } C^T. \end{aligned}$$

$$\therefore C^T = B^T A^T$$

$$\text{i.e. } (AB)^T = B^T A^T.$$

Definition: Suppose A is a square matrix. (i.e. number of rows of A = number of columns of A = n , say).

$$A = [a_{ij}]_{n \times n}$$

The entries $a_{11}, a_{22}, \dots, a_{nn}$ are said to be on the **main diagonal of A** .

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \quad \text{main diagonal}$$

$$\begin{aligned} \text{Notation: } A^2 &= A \cdot A \\ A^m &= \underbrace{A \cdot A \cdots A}_{m \text{ factors}} \end{aligned}$$

Definition: An **identity matrix** is a square matrix with 1's on the main diagonal and 0's elsewhere.

- I_n : identity matrix of order $n \times n$.

$$\bullet \quad I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \text{ and } I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Theorem: For any $m \times n$ matrix A , we have $AI_n = I_m A = A$.

Definition: Suppose A and B are square matrices.

If $AB = BA = I$, the identity matrix, then A is said to be **invertible** and B is called an **inverse** of A .

Note: If B is an inverse of A , then B is invertible and A is an inverse of B .

Example:

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

$$AB = \begin{bmatrix} 2 & -5 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 3 & 5 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 6-5 & 10-10 \\ -3+3 & -5+6 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

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

Here A, B are invertible.

Note that not all matrices are invertible.

Suppose $A = \begin{bmatrix} 1 & 4 & 0 \\ 2 & 5 & 0 \\ 3 & 6 & 0 \end{bmatrix}$.

We will show that A is not invertible.

Suppose $B = \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \end{bmatrix}$ is a 3×3 matrix. Consider BA . It is easy to check that

the 3rd column of BA is $\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$. $\therefore BA \neq I$.

Thus there is no matrix B such that $BA = I$. Hence the matrix A is not invertible.

THEOREM 1: If a matrix is invertible then its inverse is **unique**.

Proof: Suppose A is invertible. Let B and C be inverses of A .

$$\therefore \quad \quad \quad AB = BA = I \quad (1)$$

$$\text{and} \quad \quad \quad AC = CA = I \quad (2)$$

We will show that $B = C$.

Since $AB = I$, $C(AB) = CI = C$ (3)

But $C(AB) = (CA)B = IB = B$ (4)

From (3) and (4) we have $B = C$.

∴ The inverse of A is **unique**.

NOTATION: The inverse of A is denoted as A^{-1} .

Therefore $AA^{-1} = A^{-1}A = I$.

Note that the inverse of A^{-1} is A , that is, $(A^{-1})^{-1} = A$.

Result: If A and B are invertible matrices of the same order then the inverse of AB is $B^{-1}A^{-1}$.

Proof: $(AB)(B^{-1}A^{-1}) = A(BB^{-1})A^{-1}$
 $= (AI)A^{-1}$
 $= AA^{-1} = I$

Similarly we can show that $(B^{-1}A^{-1})(AB) = I$.

Therefore, AB is invertible and $(AB)^{-1} = B^{-1}A^{-1}$.

SYSTEMS OF LINEAR EQUATIONS (§ 8.4)

The equation

$$\alpha_1x_1 + \alpha_2x_2 + \dots + \alpha_nx_n = \beta \quad \text{-----} \quad (*)$$

is a **linear equation** in n variables x_1, x_2, \dots, x_n .

Here α_i 's and β are real numbers.

A collection of linear equations in x_1, \dots, x_n is referred to as a **system of linear equations**.

For example,

$$\begin{aligned} 2x_1 + 3x_2 + 5x_3 &= 7 \\ 4x_1 + x_2 - 3x_3 &= 6 \end{aligned}$$

is a system of **two** equations in **three** variables (unknowns) x_1, x_2 and x_3 .

A sequence of numbers s_1, s_2, \dots, s_n is a **solution to the system** (*) if

$$\alpha_1s_1 + \alpha_2s_2 + \dots + \alpha_ns_n = \beta.$$

Example: Consider the system of equations:

$$\begin{aligned} x + y + z &= 3 \\ 2x + y + 3z &= 1 \end{aligned}$$

$$\text{Clearly } \left. \begin{array}{l} x = -2 \\ y = 5 \\ z = 0 \end{array} \right\} \quad \text{and} \quad \left. \begin{array}{l} x = 0 \\ y = 4 \\ z = -1 \end{array} \right\} \text{ are solutions to the above system.}$$

Consider the system

$$\begin{array}{rclcl} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n & = & b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n & = & b_2 \\ \vdots & & \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n & = & b_m \end{array} \quad \text{-----(1)}$$

This system is constructed using the rectangular array

$$\left[\begin{array}{cccc|c} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & b_2 \\ \vdots & \vdots & & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} & b_n \end{array} \right]$$

This $m \times (n + 1)$ -matrix is called the **augmented matrix** for the system (1). To solve the system (1) we **eliminate** the variables.

Example: Solve the system given below:

$$\begin{array}{rcl} 2x + y - 2z & = & 10 \quad \text{-----(1)} \\ 3x + 2y + 2z & = & 1 \quad \text{-----(2)} \\ 5x + 4y + 3z & = & 4 \quad \text{-----(3)} \end{array}$$

ELEMENTARY ROW OPERATIONS (e.r.o.) (§ 8.4)

An **elementary row operation (e.r.o)** is an operation performed on the rows of a matrix and is one of **3 types**.

- 1. Multiply a row by a non-zero constant;**
- 2. Interchange any two rows;**
- 3. Add a multiple of one row to another row.**

By these operations we can transform the given matrix into the so called **row echelon form**. To solve a system of equations we transform the augmented matrix into **row echelon form**. This form facilitates the solving for the unknown variables.

DEFINITION:

- The **leading entry** of a row in a matrix is the **first non-zero** entry in the row.

- A matrix is in **row-echelon form** if the **leading** entry in **each row** (**except the first**) is to the **right** of the leading entry of the **preceding row**. The leading entry in each row is 1. All rows consisting of zeros are at the bottom of the matrix.
- A matrix is in **reduced row echelon form** if, in addition, every number above and below each leading entry is a 0.

Examples: Consider the matrices given below: The first one is in row echelon form; the second matrix is in reduced row echelon form; the third matrix is not in row echelon form.

$$\begin{bmatrix} \textcircled{1} & 2 & 0 & 5 & -6 \\ 0 & \textcircled{1} & 1 & 2 & 0 \\ 0 & 0 & 0 & \textcircled{1} & 2 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

The circled entries are the leading entries.

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

$$\begin{bmatrix} 0 & 1 & -0.5 & 0 \\ 1 & 0 & 3 & 4 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{bmatrix}$$

DEFINITION: Two matrices A and B are **row-equivalent** when one can be obtained from the other using elementary row operations.

Notation: $A \sim B$

THE RANK OF A MATRIX

DEFINITION: A set of row vectors $\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_n$ is a **linearly dependent** set if there are scalars, c_1, c_2, \dots, c_n , **not all zero**, such that:

$$c_1 \cdot \mathbf{r}_1 + c_2 \cdot \mathbf{r}_2 + \dots + c_n \cdot \mathbf{r}_n = 0. \quad \text{-----} (*)$$

The set of vectors $\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_n$ is **linearly independent** if the equation (*) implies that:

$$c_1 = c_2 = \dots = c_n = 0.$$

DEFINITION: The **rank** of a matrix A is the **maximum number of linearly independent rows of A**.

Notation: $r(A)$.

Theorem: The rank of a matrix A, **in row echelon form**, is the number of non-zero rows of A.

Theorem: If A and B are row equivalent, then $r(A) = r(B)$.

TYPES OF SOLUTIONS OF A SYSTEM OF LINEAR EQUATIONS (§ 8.4)

Consider the system:

$$\begin{matrix} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_n \end{matrix} \quad \text{----- (1)}$$

Let $A = [a_{ij}]_{m \times n}$, $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$ and $\mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$

The system (1) can be rewritten as $A\mathbf{x} = \mathbf{b}$ ----- (2)

The augmented matrix for the given system is:

$$[A | \mathbf{b}] = \left[\begin{array}{cccc|c} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & b_2 \\ \dots & & & & \\ a_{m1} & a_{m2} & \dots & a_{mn} & b_n \end{array} \right]$$

- (i) If $r(A) \neq r(A | \mathbf{b})$ then the equations in system (1) are **inconsistent and therefore there are no solutions.**
- (ii) If $r(A) = r(A | \mathbf{b}) = n$, the system is **consistent and has a unique solution.**
- (iii) If $r(A) = r(A | \mathbf{b}) = r < n$, then the system is **consistent. It has more than one solution.** In this case there are **n-r arbitrary variables.**

NOTE: When the augmented matrix of a system of linear equations has been reduced to the row-echelon form, the variables corresponding to columns containing a leading entry are called **leading variables**. The non-leading variables are taken as the **arbitrary variables**. These provide the **parameters** in the final solution.

HOMOGENOUS SYSTEM OF LINEAR EQUATIONS

A system of equations of the form

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = 0 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = 0 \\ \dots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = 0 \end{cases} \text{-----} (*)$$

is said to be a **homogenous** system of equations.

The system (*) can be rewritten as $A \mathbf{x} = \mathbf{0}$

NOTES:

Clearly $x_1 = x_2 = \dots = x_n = 0$ is **always** a solution of the system (*). It is referred to as the **trivial solution**.

If $m < n$ then the system (*) has infinitely many non-trivial solutions.

THEOREM:

Let A be an $n \times n$ matrix. The following are equivalent:

- (a) A is invertible.
- (b) $A \mathbf{x} = \mathbf{0}$ has only the trivial solution.
- (c) A is **row equivalent** to I.

Definition: A **non-invertible** matrix is said to be **singular**. If a matrix is invertible then it is said to be **non-singular**.