

$$\begin{bmatrix} 0.707 & -0.707 & 0 \\ 1.250 & 1.250 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Mathematics - II

What You'll See in This Course

This course has two modules.

- Module 1: Linear Algebra
- Module 2: Differential Equation

LINEAR ALGEBRA

- Matrices: Elementary operations, Row reduced Echelon form, Rank of matrix, Special matrices, Matrix Inversion, Determinant, and properties. System of linear equations and equivalent systems.
- Vector spaces, sub-spaces, Linear Dependence and Independence; linear span, Basis; Dimension; Co-ordinates with respect to a basis.
- Inner Products; Norm of a vector, Cauchy-Schwarz Inequality; Orthonormal basis, Gram-Schmidt process.
- Eigen Values/Eigen Vectors, Characteristic Polynomial, Diagonalisable matrices , Similarity of matrices.

Differential Equations

- Introduction to Differential Equations., First Order ODE $y'=f(x,y)$, geometrical interpretation of solutions, Separable forms, Exact Equations, integrating factor, Linear Equations, Orthogonal Trajectories.
- Picard's Theorem, Qualitative properties and Theoretical aspects, Euler's Method, Elementary classifications of equations $F(x,y,y')=0$.
- Second Order Linear differential equations: fundamental system and general solution of homogeneous equation, reduction of order.
- Existence and uniqueness of solution for second order IVP, Wronskian and general solution of non-homogeneous equations .

Differential Equations

- Euler-Cauchy Equations, extensions of the results to higher order linear equations, Higher order Differential Equations.
- Power series method.
- Legendre Polynomials, Frobenius Method.
- Bessel equation , Properties of Bessel functions.
- Sturm Liouville BVP, Orthogonal functions, Qualitative behaviour of solutions of second order ODE, Sturm comparison Theorem.
- Laplace transform, Fourier Series and Integrals.

Text and Reference books

Text Book:

- ✓ **David C. Lay, Linear Algebra and its Applications, Pearson Education 3rd Ed, 2003.**
- ✓ Erwin Kreyszig, Advanced Engineering Mathematics, 8th edition, Wiley publishers.

Reference books:

- ✓ George F. Simmons, Steven G. Krantz, Differential Equations: Theory, Technique And Practice, Tata McGraw-Hill Education.
- ✓ Coddington, An Introduction to Ordinary Differential Equations.
- ✓ G. Strang, Linear Algebra and Its Applications, Thomson Brooks/Cole, 2007.
- S. Kumaresan, Linear Algebra, A Geometric Approach, Prentice Hall India, 2008.
- Kenneth Hoffman & R. Kunze, Linear Algebra, Prentice Hall 2nd Ed, 1971.
- Additional Resource: NPTEL, MIT Video Lectures

Evaluation Methods:

Methods	Weightage
Quizzes	20%
Midterm Exam	30%
Final Examination	50%



Definitions - Matrix

- ◆ A matrix is an rectangular array of numbers enclosed in brackets

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

- ◆ Denoted with a **bold Capital letter**

Example:

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

2 rows

3 columns

What is the order?

All matrices have an order (or **dimension**): the number of rows \times the number of columns.

$$\begin{bmatrix} 8 & -1 & 3 \\ 0 & 0 & 2 \\ 10 & 4 & -3 \end{bmatrix}$$

3 x 3

(square matrix)

$$\begin{bmatrix} 9 & -5 & 7 & 0 \end{bmatrix}$$

1 x 4

(Also called a row matrix)

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

3 x 5

$$\begin{bmatrix} -9 \\ 7 \\ 0 \\ 6 \end{bmatrix}$$

4 x 1

(Also called a column matrix)

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

2 x 2

(square matrix)

Definitions: **square matrix**

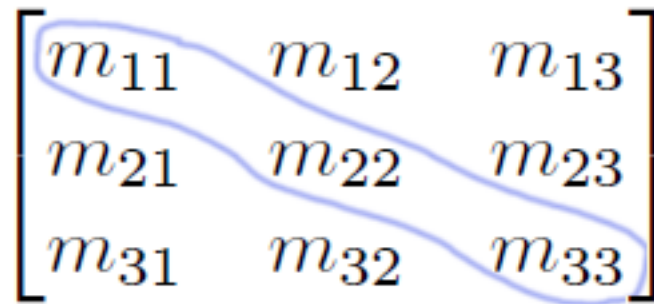
A **square matrix** is a matrix that has the same number of rows and columns ($n \times n$)

4 × 4 Square Matix

$$\begin{bmatrix} -2 & 4 & 7 & 31 \\ 6 & 9 & 12 & 6 \\ 12 & 11 & 0 & 1 \\ 9 & 10 & 2 & 3 \end{bmatrix}$$

Square Matrices: *diagonal* entries

- In a square matrix, entries m_{ii} are called the *diagonal entries*. The others are called *non-diagonal* entries

$$\begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix}$$


Diagonal Matrices

A **diagonal matrix** is a square matrix whose **non-diagonal** elements are zero.

$$A = \text{diag}(d_1, d_2, \dots, d_n) = \begin{bmatrix} d_1 & 0 & \dots & 0 \\ 0 & d_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & d_n \end{bmatrix} \in M_{n \times n}$$

$$A = \text{diag}(3, 1, -5, 2) = \begin{bmatrix} 3 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -5 & 0 \\ 0 & 0 & 0 & 2 \end{bmatrix}$$

- Diagonal matrix:

$$A = \text{diag}(d_1, d_2, \dots, d_n) = \begin{bmatrix} d_1 & 0 & \dots & 0 \\ 0 & d_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & d_n \end{bmatrix} \in M_{n \times n}$$

- Trace: If $A = [a_{ij}]_{n \times n}$

$$\text{Then } \text{Tr}(A) = a_{11} + a_{22} + \dots + a_{nn}$$

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

$$\text{Tr}(A) = 3 + 1 + (-5) + 2 = 1$$

Vectors as Matrices

- A row vector is a $1 \times n$ matrix.

- Example: 1×5 $[1 \ 2 \ 3 \ 4 \ 5]$

- A column vector is an $n \times 1$ matrix. Eg. 5×1

$$\begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{bmatrix}$$

$$[1 \ 2 \ 3]$$

$$\begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}$$

TYPES OF MATRICES

NAME	DESCRIPTION	EXAMPLE
Row matrix	A matrix with only 1 row	$[3 \quad 2 \quad 1 - 4]$
Column matrix	A matrix with only 1 column	$\begin{bmatrix} 2 \\ 3 \end{bmatrix}$
Square matrix	A matrix with same number of rows and columns	$\begin{bmatrix} 2 & 4 \\ -1 & 7 \end{bmatrix}$
Zero matrix	A matrix with all zero entries	$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$

Matrix Equality

- Two matrices are equal if and only if
 - ✓ they both have the same number of rows and the same number of columns
 - ✓ their corresponding elements are equal

$$\begin{matrix} \text{\textcolor{blue}{\#1}} \\ \begin{bmatrix} 5 & 3 & 2 \\ 33 & -5 & 11 \\ -2 & 13 & 6 \end{bmatrix} \end{matrix}$$

$$\begin{matrix} \text{\textcolor{blue}{\#2}} \\ \begin{bmatrix} 5 & 3 & 2 \\ 33 & \text{\textcolor{red}{5}} & 11 \\ -2 & 13 & 6 \end{bmatrix} \end{matrix}$$

$$\begin{matrix} \text{\textcolor{blue}{\#3}} \\ \begin{bmatrix} 5 & 3 & 2 \\ 33 & -5 & 11 \\ 2 & 13 & 6 \end{bmatrix} \end{matrix}$$

Upper triangular matrix

- **Upper triangular matrix:** A square matrix in which all the elements below the diagonal are zero i.e. a matrix of type:

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ 0 & a_{22} & a_{23} & \cdots & a_{2n} \\ 0 & 0 & a_{33} & \cdots & a_{3n} \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & 0 & 0 & a_{nn} \end{bmatrix}$$

Lower triangular matrix

- **Lower triangular matrix:** A square matrix in which all the elements above the diagonal are zero i.e. a matrix of type

$$\begin{bmatrix} a_{11} & 0 & 0 & \cdots & 0 \\ a_{21} & a_{22} & 0 & \cdots & 0 \\ a_{31} & a_{32} & a_{33} & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ a_{n1} & a_{n2} & a_{n3} & \cdots & a_{nn} \end{bmatrix}$$

Scalar matrix

Scalar matrix: A diagonal matrix in which all of the diagonal elements are equal to some constant “k”, i.e., a matrix of type

$$\begin{bmatrix} k & 0 & 0 & \dots & 0 \\ 0 & k & 0 & \dots & 0 \\ 0 & 0 & k & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & k \end{bmatrix}$$

$$\begin{pmatrix} 6 & 0 & 0 \\ 0 & 6 & 0 \\ 0 & 0 & 6 \end{pmatrix}$$

Identity matrix

Identity matrix: A diagonal matrix in which all of the diagonal elements are equal to “1” i.e. a matrix of type

$$I_n = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix}$$

An identity matrix of order $n \times n$ is denoted by I_n .

Transpose of a Matrix

- The transpose of an $r \times c$ matrix \mathbf{M} is a $c \times r$ matrix called \mathbf{M}^T .
- Take every row and rewrite it as a column.
- Equivalently, flip about the diagonal

$$\begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}^T = \begin{bmatrix} a & d & g \\ b & e & h \\ c & f & i \end{bmatrix}$$

Transpose of a Matrix: Example

$$\mathbf{b} = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}$$

$$\mathbf{b}^T = [1 \quad 1 \quad 2]$$

$$\mathbf{d} = [3 \quad 4 \quad 9]$$

$$\mathbf{d}^T = \begin{bmatrix} 3 \\ 4 \\ 9 \end{bmatrix}$$

column



row

row



column

$$\mathbf{A} = \begin{bmatrix} 1 & 2 & 3 \\ 5 & 4 & 1 \\ 6 & 7 & 4 \end{bmatrix}$$

$$\mathbf{A}^T = \begin{bmatrix} 1 & 5 & 6 \\ 2 & 4 & 7 \\ 3 & 1 & 4 \end{bmatrix}$$

Facts About Transpose

- Transpose is its own inverse: $(\mathbf{M}^T)^T = \mathbf{M}$ for all matrices \mathbf{M} .
- $\mathbf{D}^T = \mathbf{D}$ for all diagonal matrices \mathbf{D} (including the identity matrix \mathbf{I}).

Symmetric matrix

- **Symmetric matrix:** A square matrix in which corresponding elements with respect to the diagonal are equal; a matrix in which $a_{ij} = a_{ji}$ where a_{ij} is the element in the i -th row and j -th column; a matrix which is equal to its transpose; a square matrix in which a flip about the diagonal leaves it unchanged. Example:

$$\begin{bmatrix} 9 & 13 & 5 & 2 \\ 1 & 11 & 7 & 6 \\ 3 & 7 & 4 & 1 \\ 6 & 0 & 7 & 10 \end{bmatrix}$$

Not Symmetrical

$$\begin{bmatrix} 9 & 13 & 3 & 6 \\ 13 & 11 & 7 & 6 \\ 3 & 7 & 4 & 7 \\ 6 & 6 & 7 & 10 \end{bmatrix}$$

Symmetrical

Skew-symmetric matrix

- **Skew-symmetric matrix:** A square matrix in which corresponding elements with respect to the diagonal are negatives of each other; a matrix in which $a_{ij} = -a_{ji}$ where a_{ij} is the element in the i -th row and j -th column; a matrix which is equal to the negative of its transpose. The diagonal elements are always zeros. Example:

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

$$\text{If } A = \begin{bmatrix} 0 & 3 & 4 \\ -3 & 0 & 7 \\ -4 & -7 & 0 \end{bmatrix}$$

$$A^T = \begin{bmatrix} 0 & -3 & -4 \\ 3 & 0 & -7 \\ 4 & 7 & 0 \end{bmatrix} \text{ and } -A = \begin{bmatrix} 0 & -3 & -4 \\ 3 & 0 & -7 \\ 4 & 7 & 0 \end{bmatrix}$$

Orthogonal Matrix

A $n \times n$ matrix \mathbf{A} is an orthogonal matrix if $\mathbf{A}\mathbf{A}^T = \mathbf{I}$, where \mathbf{A}^T is the transpose of \mathbf{A} and \mathbf{I} is the identity matrix.

In particular, an orthogonal matrix is always invertible, and $\mathbf{A}^{-1} = \mathbf{A}^T$.

Example: $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ and $\begin{bmatrix} \cos x & -\sin x \\ \sin x & \cos x \end{bmatrix}$ are orthogonal.

Periodic matrix

Periodic matrix: A matrix A for which $A^{k+1} = A$, where k is a positive integer. If k is the least positive integer for which $A^{k+1} = A$, then A is said to be of **period k** . If $k = 1$, so that $A^2 = A$, then A is called **idempotent**.

Show that $A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$ is periodic with period 4.

Solution

$$A^2 = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix};$$

$$A^4 = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} = I;$$

$$A^5 = A^4 \cdot A = IA = A$$

Hence A is periodic and $\mathcal{P}(A) = 4$.

An idempotent matrix is a matrix $n \times n$ (square matrix A) which: $A^2 = A$

Example:

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

$$\begin{aligned} A^2 &= \begin{bmatrix} 2 & -1 \\ 2 & -1 \end{bmatrix} \cdot \begin{bmatrix} 2 & -1 \\ 2 & -1 \end{bmatrix} = \begin{bmatrix} 2 \cdot 2 - 1 \cdot 2 & 2 \cdot (-1) - 1 \cdot (-1) \\ 2 \cdot 2 - 1 \cdot 2 & 2 \cdot (-1) - 1 \cdot (-1) \end{bmatrix} \\ &= \begin{bmatrix} 4 - 2 & -2 + 1 \\ 4 - 2 & -2 + 1 \end{bmatrix} = \begin{bmatrix} 2 & -1 \\ 2 & -1 \end{bmatrix} = A \end{aligned}$$

Nilpotent matrix and Unipotent Matrix

Nilpotent matrix: A matrix A for which $A^p = 0$, where p is some positive integer. If p is the least positive integer for which $A^p = 0$, then A is said to be **nilpotent of index p** . A is said to be **unipotent** if $A - I$, where I is an identity matrix, is a nilpotent matrix

$$\text{Let } A = \begin{bmatrix} 1 & 1 & 3 \\ 5 & 2 & 6 \\ -2 & -1 & -3 \end{bmatrix} \Rightarrow A^2 = \begin{bmatrix} 0 & 0 & 0 \\ 3 & 3 & 9 \\ -1 & -1 & -3 \end{bmatrix}$$

$$\Rightarrow A^3 = A^2 \cdot A = \begin{bmatrix} 0 & 0 & 0 \\ 3 & 3 & 9 \\ -1 & -1 & -3 \end{bmatrix} \times \begin{bmatrix} 1 & 1 & 3 \\ 5 & 2 & 6 \\ -2 & -1 & -3 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}.$$

Matrix Operations

$$\begin{bmatrix} 1 & -3 \\ 7 & 5 \end{bmatrix} \times \begin{bmatrix} 10 & -8 \\ 12 & -2 \end{bmatrix} = \begin{bmatrix} (2 \cdot 10 + -3 \cdot 12) & (-16 + 6) \\ (70 + 60) & (-56 - 10) \end{bmatrix}$$

$$\begin{bmatrix} -16 & 10 \\ 130 & -66 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \times \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix} = \begin{bmatrix} 1 \cdot 5 + 2 \cdot 7 & 1 \cdot 6 + 2 \cdot 8 \\ 3 \cdot 5 + 4 \cdot 7 & 3 \cdot 6 + 4 \cdot 8 \end{bmatrix}$$

Adding Two Matrices

To add two matrices, they must have the same order. To add, you simply add corresponding entries.

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

The diagram illustrates the process of adding two matrices. A yellow curved arrow points from the top-left element of the first matrix (3) to the top-left element of the result matrix (7). Above this arrow, the calculation $3+4=7$ is shown, indicating that the top-left element of the second matrix (4) is added to the top-left element of the first matrix (3) to produce the top-left element of the result matrix (7).

Adding Two Matrices

To add two matrices, they must have the same order. To add, you simply add corresponding entries.

If $A = [a_{ij}]_{m \times n}$, $B = [b_{ij}]_{m \times n}$ Then $A + B = [a_{ij}]_{m \times n} + [b_{ij}]_{m \times n} = [a_{ij} + b_{ij}]_{m \times n}$

$$\begin{bmatrix} 5 & -3 \\ -3 & 4 \\ 0 & 7 \end{bmatrix} + \begin{bmatrix} -2 & 1 \\ 3 & 0 \\ 4 & -3 \end{bmatrix} = \begin{bmatrix} 5 + (-2) & -3 + 1 \\ -3 + 3 & 4 + 0 \\ 0 + 4 & 7 + (-3) \end{bmatrix}$$
$$= \begin{bmatrix} 3 & -2 \\ 0 & 4 \\ 4 & 4 \end{bmatrix}$$

Adding Two Matrices

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

$$= \begin{bmatrix} 8+(-1) & 0+7 & -1+5 & 3+2 \\ -5+5 & 4+3 & 2+3 & 9+(-2) \end{bmatrix}$$

$$= \begin{bmatrix} 7 & 7 & 4 & 5 \\ 0 & 7 & 5 & 7 \end{bmatrix}$$

Subtracting Two Matrices

To subtract two matrices, they must have the same order. You simply subtract corresponding entries.

$$\begin{bmatrix} 9 & -2 & 4 \\ 5 & 0 & 6 \\ 1 & 3 & 8 \end{bmatrix} - \begin{bmatrix} 4 & 0 & 7 \\ 1 & 5 & -4 \\ -2 & 3 & 2 \end{bmatrix} = \begin{bmatrix} 9-4 & -2-0 & 4-7 \\ 5-1 & 0-5 & 6-(-4) \\ 1-(-2) & 3-3 & 8-2 \end{bmatrix}$$
$$= \begin{bmatrix} 5 & -2 & -3 \\ 4 & -5 & 10 \\ 3 & 0 & 6 \end{bmatrix}$$

Subtracting Two Matrices

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

$$= \begin{bmatrix} 2-0 & -4-1 & 3-8 \\ 8-3 & 0-(-1) & -7-1 \\ 1-(-4) & 5-2 & 0-7 \end{bmatrix} = \begin{bmatrix} 2 & -5 & -5 \\ 5 & 1 & -8 \\ 5 & 3 & -7 \end{bmatrix}$$

Multiplying By a Scalar

- Can multiply a matrix by a scalar.
- Result is a matrix of the same dimension.
- To multiply a matrix by a scalar, multiply each component by the scalar.

$$k\mathbf{M} = k \begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \\ m_{41} & m_{42} & m_{43} \end{bmatrix} = \begin{bmatrix} km_{11} & km_{12} & km_{13} \\ km_{21} & km_{22} & km_{23} \\ km_{31} & km_{32} & km_{33} \\ km_{41} & km_{42} & km_{43} \end{bmatrix}$$

Multiplying a Matrix by a Scalar

- In matrix algebra, a real number is often called a **SCALAR**. To multiply a matrix by a scalar, you multiply each entry in the matrix by that scalar.

$$4 \begin{bmatrix} -2 & 0 \\ 4 & -1 \end{bmatrix} = \begin{bmatrix} 4(-2) & 4(0) \\ 4(4) & 4(-1) \end{bmatrix} \\ = \begin{bmatrix} -8 & 0 \\ 16 & -4 \end{bmatrix}$$

Matrix Multiplication

Multiplying an $r \times n$ matrix **A** by an $n \times c$ matrix **B** gives an $r \times c$ result **AB**.

$$\begin{array}{ccc} \mathbf{A} & \mathbf{B} & \mathbf{AB} \\ \begin{bmatrix} ? & ? \\ ? & ? \\ ? & ? \\ ? & ? \end{bmatrix} & \begin{bmatrix} ? & ? & ? & ? & ? \\ ? & ? & ? & ? & ? \end{bmatrix} & = \begin{bmatrix} ? & ? & ? & ? & ? \\ ? & ? & ? & ? & ? \\ ? & ? & ? & ? & ? \\ ? & ? & ? & ? & ? \end{bmatrix} \\ \begin{array}{c} r \times n \\ 4 \times 2 \end{array} & \begin{array}{c} n \times c \\ 2 \times 5 \end{array} & \begin{array}{c} r \times c \\ 4 \times 5 \end{array} \end{array}$$

Multiplication: Result

- Multiply an $r \times n$ matrix **A** by an $n \times c$ matrix **B** to give an $r \times c$ result **C** = **AB**.
- Then **C** = $[c_{ij}]$, where c_{ij} is the dot product of the i th row of **A** with the j th column of **B**.
- That is:

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

Example

$$\begin{bmatrix} c_{11} & c_{12} & c_{13} & c_{14} & c_{15} \\ c_{21} & c_{22} & c_{23} & c_{24} & c_{25} \\ c_{31} & c_{32} & c_{33} & c_{34} & c_{35} \\ c_{41} & c_{42} & c_{43} & c_{44} & c_{45} \end{bmatrix} =$$

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \\ a_{41} & a_{42} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & b_{13} & b_{14} & b_{15} \\ b_{21} & b_{22} & b_{23} & b_{24} & b_{25} \end{bmatrix}$$

$$c_{24} = a_{21}b_{14} + a_{22}b_{24}$$

Another Way of Looking at It

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \\ a_{41} & a_{42} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & b_{13} & b_{14} & b_{15} \\ b_{21} & b_{22} & b_{23} & b_{24} & b_{25} \\ c_{11} & c_{12} & c_{13} & c_{14} & c_{15} \\ c_{21} & c_{22} & c_{23} & c_{24} & c_{25} \\ c_{31} & c_{32} & c_{33} & c_{34} & c_{35} \\ c_{41} & c_{42} & c_{43} & c_{44} & c_{45} \end{bmatrix}$$

$$c_{43} = a_{41}b_{13} + a_{42}b_{23}$$

2 x 2 Case

$$\begin{aligned}\mathbf{AB} &= \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} \\ &= \begin{bmatrix} a_{11}b_{11} + a_{12}b_{21} & a_{11}b_{12} + a_{12}b_{22} \\ a_{21}b_{11} + a_{22}b_{21} & a_{21}b_{12} + a_{22}b_{22} \end{bmatrix}\end{aligned}$$

2 x 2 Example

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

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

$$= \begin{bmatrix} (-3)(-7) + (0)(4) & (-3)(2) + (0)(6) \\ (5)(-7) + (1/2)(4) & (5)(2) + (1/2)(6) \end{bmatrix}$$

$$= \begin{bmatrix} 21 & -6 \\ -33 & 13 \end{bmatrix}$$

3 x 3 Case

$$\begin{aligned}\mathbf{AB} &= \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \end{bmatrix} \\ &= \begin{bmatrix} a_{11}b_{11} + a_{12}b_{21} + a_{13}b_{31} & a_{11}b_{12} + a_{12}b_{22} + a_{13}b_{32} & a_{11}b_{13} + a_{12}b_{23} + a_{13}b_{33} \\ a_{21}b_{11} + a_{22}b_{21} + a_{23}b_{31} & a_{21}b_{12} + a_{22}b_{22} + a_{23}b_{32} & a_{21}b_{13} + a_{22}b_{23} + a_{23}b_{33} \\ a_{31}b_{11} + a_{32}b_{21} + a_{33}b_{31} & a_{31}b_{12} + a_{32}b_{22} + a_{33}b_{32} & a_{31}b_{13} + a_{32}b_{23} + a_{33}b_{33} \end{bmatrix}\end{aligned}$$

3 x 3 Example

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

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

$$= \begin{bmatrix} 1 \cdot (-8) + (-5) \cdot 7 + 3 \cdot 2 & 1 \cdot 6 + (-5) \cdot 0 + 3 \cdot 4 & 1 \cdot 1 + (-5) \cdot (-3) + 3 \cdot 5 \\ 0 \cdot (-8) + (-2) \cdot 7 + 6 \cdot 2 & 0 \cdot 6 + (-2) \cdot 0 + 6 \cdot 4 & 0 \cdot 1 + (-2) \cdot (-3) + 6 \cdot 5 \\ 7 \cdot (-8) + 2 \cdot 7 + (-4) \cdot 2 & 7 \cdot 6 + 2 \cdot 0 + (-4) \cdot 4 & 7 \cdot 1 + 2 \cdot (-3) + (-4) \cdot 5 \end{bmatrix}$$

$$= \begin{bmatrix} -37 & 18 & 31 \\ -2 & 24 & 36 \\ -50 & 26 & -19 \end{bmatrix}$$

Common Mistakes

- Does $AB = BA$ (In general)?
- Whether Matrix multiplication is commutative?
- Statement is not true in general, see example:
- Example: $A = \begin{bmatrix} 9 & 3 \\ -2 & 0 \end{bmatrix}, B = \begin{bmatrix} 1 & -4 \\ 2 & 5 \end{bmatrix}$

Caution!

- If $AB = 0$ does that mean $A = 0$, $B = 0$ or $AB = 0$?
- Statement is not true in general, see example:
- Example:
- $A = \begin{bmatrix} 1 & 1 \\ 3 & 3 \end{bmatrix}, B = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$

Caution!

- If $AC = AD$ does that mean $C = D$ (when A is non zero matrix) ?
- Statement is not true in general, see example:

Example:

$$A = \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}, \quad B = \begin{bmatrix} 2 & 1 \\ 2 & 2 \end{bmatrix}, \quad C = \begin{bmatrix} 3 & 0 \\ 1 & 3 \end{bmatrix}$$

Matrix Multiplication Facts

- Not commutative: in general $\mathbf{AB} \neq \mathbf{BA}$.

- Associative:

$$(\mathbf{AB})\mathbf{C} = \mathbf{A}(\mathbf{BC})$$

- Associates with scalar multiplication:

$$k(\mathbf{AB}) = (k\mathbf{A})\mathbf{B} = \mathbf{A}(k\mathbf{B})$$

- $(\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T$

- $(\mathbf{M}_1 \mathbf{M}_2 \mathbf{M}_3 \dots \mathbf{M}_n)^T = \mathbf{M}_n^T \dots \mathbf{M}_3^T \mathbf{M}_2^T \mathbf{M}_1^T$

- Does $\mathbf{AB} = \mathbf{BA}$?

- If $\mathbf{AB} = \mathbf{0}$, then $\mathbf{A} = ?$ Or $\mathbf{B} = ?$

- If $\mathbf{AB} = \mathbf{AC}$, Then $\mathbf{C} = ? \mathbf{B}$

Matrix

Matrix Operations

Properties of Matrix Multiplication:

- In general, $\mathbf{AB} \neq \mathbf{BA}$ if both exist, but there are special cases that this property is not true.
- If \mathbf{I} is an identity matrix $\mathbf{IB} = \mathbf{BI} = \mathbf{B}$.
- $\mathbf{A(B + C)} = \mathbf{AB + AC}$ and $\mathbf{(B + C)A} = \mathbf{BA + CA}$
- $\mathbf{A(BC)} = \mathbf{(AB)C}$
- If \mathbf{AB} exist then $\mathbf{(AB)' = B'A'}$ (this can be extended to more than 2 matrices, i.e.: $\mathbf{(ABC)' = C'B'A'}$)
- From $\mathbf{AB = \underline{0}}$ we cannot conclude necessarily that $\mathbf{A = \underline{0}}$ or $\mathbf{B = \underline{0}}$.*
- From $\mathbf{AB = AC}$ we cannot conclude necessarily that $\mathbf{B = C}$.**

Row Vector Times Matrix Multiplication

Can multiply a row vector times a matrix

$$\begin{bmatrix} x & y & z \end{bmatrix} \begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix} = \begin{bmatrix} xm_{11} + ym_{21} + zm_{31} & xm_{12} + ym_{22} + zm_{32} & xm_{13} + ym_{23} + zm_{33} \end{bmatrix}$$

Matrix Times Column Vector Multiplication

Can multiply a matrix times a column vector.

$$\begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} xm_{11} + ym_{12} + zm_{13} \\ xm_{21} + ym_{22} + zm_{23} \\ xm_{31} + ym_{32} + zm_{33} \end{bmatrix}$$

Common Mistake

$\mathbf{M}\mathbf{v}^\top \neq (\mathbf{v}\mathbf{M})^\top$, but $\mathbf{M}\mathbf{v}^\top = (\mathbf{v}\mathbf{M}^\top)^\top$ – compare the following two results:

$$\begin{aligned} \begin{bmatrix} x & y & z \end{bmatrix} \begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix} \\ = \begin{bmatrix} xm_{11} + \underbrace{ym_{21}}_{\text{red}} + \underbrace{zm_{31}}_{\text{blue}} & xm_{12} + ym_{22} + zm_{32} & xm_{13} + ym_{23} + zm_{33} \end{bmatrix} \\ \begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} xm_{11} + \underbrace{ym_{12}}_{\text{red}} + \underbrace{zm_{13}}_{\text{blue}} \\ xm_{21} + ym_{22} + zm_{23} \\ xm_{31} + ym_{32} + zm_{33} \end{bmatrix} \end{aligned}$$

Vector-Matrix Multiplication Facts 1

Associates with vector multiplication.

- Let \mathbf{v} be a row vector:

$$\mathbf{v}(\mathbf{AB}) = (\mathbf{vA})\mathbf{B}$$

- Let \mathbf{v} be a column vector:

$$(\mathbf{AB})\mathbf{v} = \mathbf{A}(\mathbf{Bv})$$

Vector-Matrix Multiplication Facts 2

- Vector-matrix multiplication distributes over vector addition:

$$(\mathbf{v} + \mathbf{w})\mathbf{M} = \mathbf{v}\mathbf{M} + \mathbf{w}\mathbf{M}$$

- That was for row vectors \mathbf{v} , \mathbf{w} . Similarly for column vectors.