

# Lecture 4: Matrix Operations

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# Recap

- General System of linear equations and their solutions

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- Defined various matrices

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- General System of linear equations and their solutions
- Defined various matrices
- Proceeding further - Look at operations on matrices

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- $A$  and  $B$  - Same size
- Example:

$$A = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 4 & 2 & -1 & 3 \end{pmatrix}, B = \begin{pmatrix} 1 & -1 & 1 & 1 \\ 2 & 3 & 4 & 5 \end{pmatrix}, C = A + B = \begin{pmatrix} 2 & 1 & 4 & 5 \\ 6 & 5 & 3 & 8 \end{pmatrix}$$

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- if  $A$  and  $B$  are of the same size, then  $A - B = A + (-1)B$ .
- Subtraction of two matrices - Effected as matrix addition with  $(-1)$  multiple of  $B$

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- Note that we have only defined the scalar multiplication and not multiplication of two matrices.

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- Symmetric matrices - Wide applications in Data Science etc

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- Example:  $A = \begin{pmatrix} 0 & 2 \\ -2 & 0 \end{pmatrix}$  is a skew-symmetric matrix.
- $A = \begin{pmatrix} 1 & 2 & 3 \\ -2 & 4 & 5 \\ -3 & -5 & 7 \end{pmatrix}$  is not a skew-symmetric matrix

# Properties of Transpose Operation

For any two matrix of same size  $A$  and  $B$

1.  $(A \pm B)^T = A^T \pm B^T$
2.  $(A^T)^T = A$
3.  $(k_1 A)^T = k_1(A^T)$
4. For a square matrix  $A$ ,  $A + A^T$  is always symmetric  
Proof:  $(A + A^T)^T = A^T + (A^T)^T = A^T + A = A + A^T$ .
5. For a square matrix  $A$ ,  $A - A^T$  is always skew - symmetric.  
Proof:  $(A - A^T)^T = A^T - (A^T)^T = -(A - A^T)$

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$$A + A^T = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} + \begin{pmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{pmatrix} = \begin{pmatrix} 2 & 6 & 10 \\ 6 & 10 & 14 \\ 10 & 14 & 18 \end{pmatrix}$$

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

# Trace of a matrix

- Given a matrix  $A$ , Trace of  $A$  as the sum of the diagonal elements of  $A$ .
- $\text{Trace}(A) = \sum_{i=1}^n a_{ii}$
- $Tr(A)$  or  $\text{Trace}(A)$ .
- For example:

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \Rightarrow Tr(A) = 1 + 5 + 9 = 15$$

## Properties of Trace of a matrix

- For any two square matrices of the same size,  $A, B$ ,  $\text{Trace}(A + B) = \text{Trace}(A) + \text{Trace}(B)$
- $\text{Trace}(A^T) = \text{Trace}(A)$
- $\text{Trace}(k_1 A) = k_1(\text{Trace}(A))$ , for some scalar  $k_1$
- Trace of a matrix = sum of the eigenvalues of  $A$

# Summary

- Matrix Operations such as addition of matrices and scalar multiplication of a matrix
- Matrix Transpose, Symmetric and skew-symmetric matrix
- Trace of matrix and properties
- Matrix Multiplication, Homogeneous system of equations - Next Lecture

**Namaste!!!**

# Lecture 5: Matrix Multiplication, Homogeneous System of Equations

M. Krishna Kumar

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- Focus of this lecture: Matrix Multiplication, Homogeneous system of equations

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- Done slightly differently

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Let  $A$  and  $B$  be two matrices

- $A$ , if it has  $n$  columns, then  $B$  must have  $n$  rows to obtain the product matrix  $C = AB$

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- For  $A^{m \times n}, B^{n \times p}$ ,  $C = AB$  is an  $m \times p$  matrix

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- $C$ , then, has  $m$  rows and  $p$  columns
- For  $A^{m \times n}, B^{n \times p}$ ,  $C = AB$  is an  $m \times p$  matrix
- The product  $AB$  is defined only if **the number of rows of  $B$  = Number of columns of  $A$**

# Matrix Multiplication - How is it done?

- Let  $A$  be a  $2 \times 3$  matrix :  $A = \begin{pmatrix} a_{1,1} & a_{1,2} & a_{1,3} \\ a_{2,1} & a_{2,2} & a_{2,3} \end{pmatrix}$

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- Let  $B$  be a  $3 \times 2$  matrix:  $B = \begin{pmatrix} b_{1,1} & b_{1,2} \\ b_{2,1} & b_{2,2} \\ b_{3,1} & b_{3,2} \end{pmatrix}$

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- Number of columns of  $A = 3$ , Number of rows of  $B = 3$

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- Let  $B$  be a  $3 \times 2$  matrix:  $B = \begin{pmatrix} b_{1,1} & b_{1,2} \\ b_{2,1} & b_{2,2} \\ b_{3,1} & b_{3,2} \end{pmatrix}$
- Number of columns of  $A = 3$ , Number of rows of  $B = 3$
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$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{i1} & a_{i2} & a_{i3} & \dots & a_{in} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{pmatrix}$$

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- Example: Both  $A$  and  $B$  are square matrices of the same size

# Properties of Multiplication of matrices

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Example:

$$A = \begin{pmatrix} 2 & 1 \\ 6 & 3 \end{pmatrix}, B = \begin{pmatrix} -1 & 0 \\ 5 & 2 \end{pmatrix}, C = \begin{pmatrix} 3 & 1 \\ -3 & 0 \end{pmatrix}$$

$$AB = \begin{pmatrix} 2 & 1 \\ 6 & 3 \end{pmatrix} \begin{pmatrix} -1 & 0 \\ 5 & 2 \end{pmatrix} = \begin{pmatrix} 3 & 2 \\ 9 & 6 \end{pmatrix}$$

$$AC = \begin{pmatrix} 2 & 1 \\ 6 & 3 \end{pmatrix} \begin{pmatrix} 3 & 1 \\ -3 & 0 \end{pmatrix} = \begin{pmatrix} 3 & 2 \\ 9 & 6 \end{pmatrix}$$

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- Observe: If all the unknowns,  $x_i$ 's are 0, then the RHS of the  $m$  linear equations are all 0.
- All  $x_i$ 's = 0 : **TRIVIAL Solution**
- Homogeneous system of equations - **Always consistent**
- Any non zero solution - **NON TRIVIAL** solution

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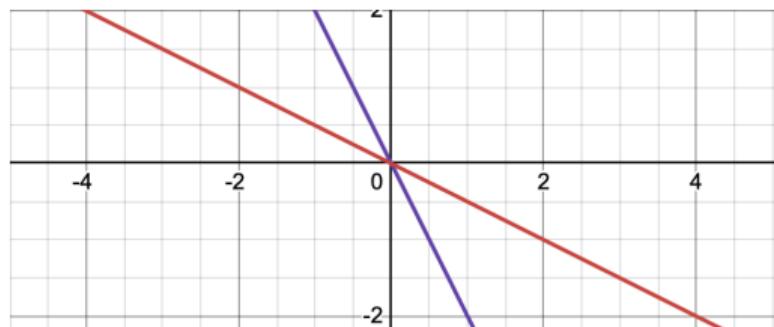


Figure: Homogeneous System with unique solution. The two lines intersect at the origin (0, 0)

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- Study of homogeneous system of linear equations and its solution - Reveals lot of information about the matrix rank, invertibility if the coefficient matrix is a square matrix etc

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**Namaste!!!**