

## Matrix Operations

Just special functions  
from  $\mathbb{R}^n$  to  $\mathbb{R}^m$

$$\{ \text{m} \times n \text{ matrices} \} = \{ \text{Linear transformations from } \mathbb{R}^n \text{ to } \mathbb{R}^m \}$$

Key observation : We can add, scale and compose functions so we can translate this to matrices.

<u>Matrices</u>	$\xrightarrow{\text{m} \times n \text{ matrices}}$
$A = (a_1 \dots a_n)$	

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$$B = (b_1 \dots b_n)$$

$$A + B := (\underline{a}_1 + \underline{b}_1, \underline{a}_2 + \underline{b}_2, \dots, \underline{a}_n + \underline{b}_n)$$

$\xrightarrow{\text{Term by term sum}}$

## Linear Transformations

$$T_A : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$T_B : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$T_A + T_B = T_{A+B}$$

$\xrightarrow{\text{Sum of Functions}}$

$$\text{ie } (T_A + T_B)(\underline{x}) := T_A(\underline{x}) + T_B(\underline{x})$$

$$\lambda A := (\lambda a_1 \lambda a_2 \dots \lambda a_n)$$

$\xrightarrow{\text{each entry multiplied by } \lambda}$

$$\lambda T_A = T_{\lambda A}$$

$\xrightarrow{\text{T}_A \text{ multiplied by constant } \lambda}$

$$\text{ie } (\lambda T_A)(\underline{x}) := \lambda T_A(\underline{x})$$

## Example

$$\begin{pmatrix} 1 & 3 & 5 \\ 2 & 4 & 6 \end{pmatrix} + \begin{pmatrix} 0 & 2 & 4 \\ 3 & -1 & 6 \end{pmatrix} = \begin{pmatrix} 1 & 5 & 9 \\ 5 & 3 & 12 \end{pmatrix}$$

$$2 \begin{pmatrix} 2 & 3 \\ 4 & 7 \end{pmatrix} = \begin{pmatrix} 4 & 6 \\ 8 & 14 \end{pmatrix}$$

Remark To sum matrices they must have same size.

Composition is more complicated:

$$A = (\underline{a}_1 \dots \underline{a}_n) - m \times n \text{ matrix}$$

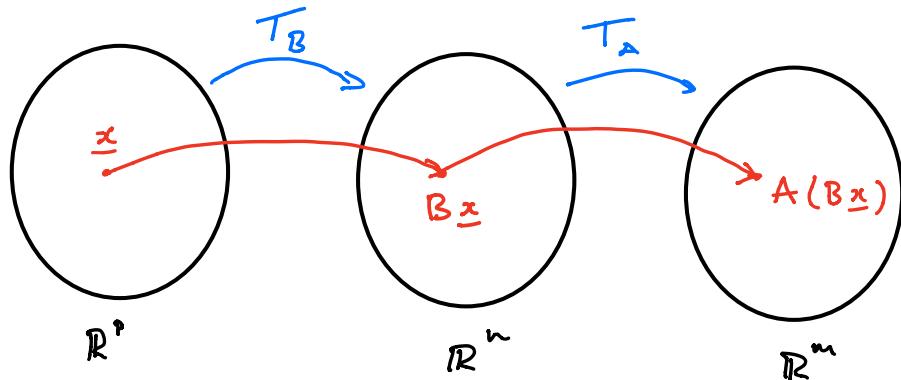
$$B = (\underline{b}_1 \dots \underline{b}_p) - n \times p \text{ matrix}$$

$$\Rightarrow T_A : \mathbb{R}^n \longrightarrow \mathbb{R}^m$$

$$\underline{x} \longmapsto A \underline{x}$$

$$T_B : \mathbb{R}^p \longrightarrow \mathbb{R}^n$$

$$\underline{x} \longmapsto B \underline{x}$$



Fact:  $T_A \circ T_B : \mathbb{R}^p \rightarrow \mathbb{R}^m$  linear.

$$T_A \circ T_B (\underline{e}_1) = A(B\underline{e}_1) = A \underline{b}_1 \quad \underline{e}_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$\vdots \qquad \vdots \qquad \vdots$$

$$T_A \circ T_B (\underline{e}_p) = A(B\underline{e}_p) = A \underline{b}_p \quad \underline{e}_p = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$\Rightarrow \text{Standard Matrix } \underset{\text{of } T_A \circ T_B}{=} (A \underline{b}_1 \ A \underline{b}_2 \ \dots \ A \underline{b}_p)$$

Definition

$$A \underset{m \times n}{\underbrace{B}} := (A \underline{b}_1, A \underline{b}_2, \dots, A \underline{b}_p) \underset{n \times p}{\underbrace{)} \quad m \times p$$

(  $A \underline{x}$  can now be interpreted as matrix multiplication. )  
 $\uparrow \quad \nwarrow$   
 $m \times n \quad n \times 1$  (vector in  $\mathbb{R}^n$ )

Key Fact : Matrix Multiplication = Composition of linear transformations

i.e.  $T_{AB} = T_A \circ T_B$

Row - Column Rule of Computing  $AB$

$$(AB)_{ij} = a_{i1} b_{1j} + a_{i2} b_{2j} + \dots + a_{in} b_{nj}$$

*j<sup>th</sup> column*

A

B

Example

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

$\uparrow \quad \uparrow$   
 $2 \times 3 \quad 3 \times 2$

$$\begin{pmatrix} 4 & 6 \\ 10 & 12 \end{pmatrix} \quad 2 \times 2$$

## Properties of Matrix Operations

1/  $A(BC) = (AB)C$

2/  $A(B+C) = AB + AC$

3/  $(A+B)C = AC + BC$

4/  $\lambda(AB) = (\lambda A)B = A(\lambda B)$

5/  $I_m A = A = A I_n$

Warning In general  $AB \neq BA$ . Also

$AB = 0$   $\cancel{\Rightarrow}^{\text{zero matrix}}$   $A = 0$  or  $B = 0$ . This means

$AC = AB \cancel{\Rightarrow} C = B$  in general. E.g.  $\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$

Definition  $A = m \times n$  matrix

$A^T = A$  "transpose"

$$(A^T)_{ij} := A_{ji}$$

Example

$$A = \begin{pmatrix} 1 & 3 & 5 \\ 2 & 4 & 6 \end{pmatrix}, A^T = \begin{pmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{pmatrix}$$

Facts

1/  $(A^T)^T = A$

2/  $(A+B)^T = A^T + B^T$

3/  $(rA)^T = r A^T$

4/  $(AB)^T = B^T A^T$