

# Section 2.1 Matrix operations

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# Notations in a matrix

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

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- The diagonal entries in an  $m \times n$  matrix  $A = (a_{ij})$  are  $a_{11}, a_{22}, \dots$ , and they are form the main diagonal of  $A$ .

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- The two matrices are **equal** if they have the same size (i.e., the same number of rows and the same number of columns) and if their corresponding entries are equal.



# Sum, Scalar multiples, and Transpose

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- So if  $A = (a_{ij})$  and  $B = (b_{ij})$ , and  $r \in \mathbf{R}$ , then

$$A + B = (a_{ij} + b_{ij}), rA = (ra_{ij}), A^T = (a_{ji})$$

# Example

Ex1: Let  $A = \begin{bmatrix} 4 & 0 & 5 \\ -1 & 3 & 2 \end{bmatrix}$ ,  $B = \begin{bmatrix} 1 & 1 & 1 \\ 3 & 5 & 7 \end{bmatrix}$ . Find  $A + B$ ,  $3A$ ,  $B^T$ .

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- Let  $A$  and  $B$  be matrices with the same size. Then

- 1  $(A^T)^T = A$
- 2  $(A + B)^T = A^T + B^T$
- 3 For any scalar  $r$ ,  $(rA)^T = rA^T$

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- So when multiply the vector  $Bx$  by  $A$ , we have

$$A(Bx) = A(x_1 b_1) + A(x_2 b_2) + \dots + A(x_p b_p) = x_1 A b_1 + x_2 A b_2 + \dots + x_p A b_p$$

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# Examples

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$$\begin{aligned} AB &= \begin{bmatrix} 2 \cdot 4 + 3 \cdot 1 & 2 \cdot 3 + 3 \cdot (-2) & 2 \cdot 9 + 3 \cdot 3 \\ 1 \cdot 4 + (-5) \cdot 1 & 1 \cdot 3 + (-5) \cdot (-2) & 1 \cdot 9 + (-5) \cdot 3 \end{bmatrix} \\ &= \begin{bmatrix} 11 & 0 & 21 \\ -1 & 13 & -9 \end{bmatrix} \end{aligned}$$

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- In terms of transpose, we have  $(AB)^T = B^T A^T$ .

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$$A = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}, \text{ and } B = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}. \text{ Then } AB = 0$$



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Let  $A = \begin{bmatrix} 5 & 1 \\ 3 & -2 \end{bmatrix}$ , and  $B = \begin{bmatrix} 2 & 0 \\ 4 & 3 \end{bmatrix}$ . Find  $AB$  and  $BA$ .

- Even if  $AB = 0$ , we may have  $A \neq 0$  and  $B \neq 0$ :

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- If  $A$  is an  $n \times n$  matrix, and  $k$  is a positive integer, then  $A^k$  denote the product of  $k$  copies of  $A$ :  $A^k = AA \dots A$ . Moreover, we denote  $A^0$  to be the identity matrix.