

## 2.1 Matrix Operations

### Matrix Notation:

Two ways to denote  $m \times n$  matrix  $A$ :

In terms of the *columns* of  $A$ :

$$A = \left[ \mathbf{a}_1 \ \mathbf{a}_2 \ \cdots \ \mathbf{a}_n \right]$$

In terms of the *entries* of  $A$ :

$$A = \begin{bmatrix} a_{11} & \cdots & a_{1j} & \cdots & a_{1n} \\ \vdots & & & & \vdots \\ a_{i1} & \cdots & a_{ij} & \cdots & a_{in} \\ \vdots & & \vdots & & \vdots \\ a_{m1} & \cdots & a_{mj} & \cdots & a_{mn} \end{bmatrix}$$

**Main diagonal entries:** \_\_\_\_\_

**Zero matrix:**

$$0 = \begin{bmatrix} 0 & \cdots & 0 & \cdots & 0 \\ \vdots & & & & \vdots \\ 0 & \cdots & 0 & \cdots & 0 \\ \vdots & & \vdots & & \vdots \\ 0 & \cdots & 0 & \cdots & 0 \end{bmatrix}$$

### THEOREM 1

Let  $A$ ,  $B$ , and  $C$  be matrices of the same size, and let  $r$  and  $s$  be scalars. Then

- |                                |                         |
|--------------------------------|-------------------------|
| a. $A + B = B + A$             | d. $r(A + B) = rA + rB$ |
| b. $(A + B) + C = A + (B + C)$ | e. $(r + s)A = rA + sA$ |
| c. $A + 0 = A$                 | f. $r(sA) = (rs)A$      |

## Matrix Multiplication

Multiplying  $B$  and  $\mathbf{x}$  transforms  $\mathbf{x}$  into the vector  $B\mathbf{x}$ . In turn, if we multiply  $A$  and  $B\mathbf{x}$ , we transform  $B\mathbf{x}$  into  $A(B\mathbf{x})$ . So  $A(B\mathbf{x})$  is the composition of two mappings.

Define the product  $AB$  so that  $A(B\mathbf{x}) = (AB)\mathbf{x}$ .

Suppose  $A$  is  $m \times n$  and  $B$  is  $n \times p$  where

$$B = [\mathbf{b}_1 \ \mathbf{b}_2 \ \cdots \ \mathbf{b}_p] \text{ and } \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_p \end{bmatrix}.$$

Then

$$B\mathbf{x} = x_1\mathbf{b}_1 + x_2\mathbf{b}_2 + \cdots + x_p\mathbf{b}_p$$

and

$$A(B\mathbf{x}) = A(x_1\mathbf{b}_1 + x_2\mathbf{b}_2 + \cdots + x_p\mathbf{b}_p)$$

$$= A(x_1\mathbf{b}_1) + A(x_2\mathbf{b}_2) + \cdots + A(x_p\mathbf{b}_p)$$

$$= x_1A\mathbf{b}_1 + x_2A\mathbf{b}_2 + \cdots + x_pA\mathbf{b}_p = [A\mathbf{b}_1 \ A\mathbf{b}_2 \ \cdots \ A\mathbf{b}_p] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_p \end{bmatrix}.$$

Therefore,

$$A(B\mathbf{x}) = [A\mathbf{b}_1 \ A\mathbf{b}_2 \ \cdots \ A\mathbf{b}_p]\mathbf{x}.$$

and by defining

$$AB = [A\mathbf{b}_1 \ A\mathbf{b}_2 \ \cdots \ A\mathbf{b}_p]$$

we have  $A(B\mathbf{x}) = (AB)\mathbf{x}$ .

**EXAMPLE:** Compute  $AB$  where  $A = \begin{bmatrix} 4 & -2 \\ 3 & -5 \\ 0 & 1 \end{bmatrix}$  and  $B = \begin{bmatrix} 2 & -3 \\ 6 & -7 \end{bmatrix}$ .

*Solution:*

$$\begin{aligned} A\mathbf{b}_1 &= \begin{bmatrix} 4 & -2 \\ 3 & -5 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 6 \end{bmatrix}, & A\mathbf{b}_2 &= \begin{bmatrix} 4 & -2 \\ 3 & -5 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} -3 \\ -7 \end{bmatrix} \\ &= \begin{bmatrix} -4 \\ -24 \\ 6 \end{bmatrix} & &= \begin{bmatrix} 2 \\ 26 \\ -7 \end{bmatrix} \\ & & \Rightarrow AB &= \begin{bmatrix} -4 & 2 \\ -24 & 26 \\ 6 & -7 \end{bmatrix} \end{aligned}$$

Note that  $A\mathbf{b}_1$  is a linear combination of the columns of  $A$  and  $A\mathbf{b}_2$  is a linear combination of the columns of  $A$ .

Each column of  $AB$  is a linear combination of the columns of  $A$  using weights from the corresponding columns of  $B$ .

**EXAMPLE:** If  $A$  is  $4 \times 3$  and  $B$  is  $3 \times 2$ , then what are the sizes of  $AB$  and  $BA$ ?

*Solution:*

$$AB = \begin{bmatrix} * & * & * \\ * & * & * \\ * & * & * \\ * & * & * \end{bmatrix} \begin{bmatrix} * & * \\ * & * \\ * & * \end{bmatrix} = \begin{bmatrix} & \\ & \\ & \\ & \end{bmatrix}$$

$$BA \text{ would be } \begin{bmatrix} * & * \\ * & * \\ * & * \end{bmatrix} \begin{bmatrix} * & * & * \\ * & * & * \\ * & * & * \\ * & * & * \end{bmatrix} \text{ which is } \underline{\hspace{2cm}}.$$

If  $A$  is  $m \times n$  and  $B$  is  $n \times p$ , then  $AB$  is  $m \times p$ .

## Row-Column Rule for Computing AB (alternate method)

The definition

$$AB = [Ab_1 \quad Ab_2 \quad \cdots \quad Ab_p]$$

is good for theoretical work.

When  $A$  and  $B$  have small sizes, the following method is more efficient when working by hand.

If  $AB$  is defined, let  $(AB)_{ij}$  denote the entry in the  $i$ th row and  $j$ th column of  $AB$ . Then

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

$$\begin{bmatrix} a_{i1} & a_{i2} & \cdots & a_{in} \end{bmatrix} \begin{bmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{nj} \end{bmatrix} = \begin{bmatrix} (AB)_{ij} \end{bmatrix}$$

**EXAMPLE**  $A = \begin{bmatrix} 2 & 3 & 6 \\ -1 & 0 & 1 \end{bmatrix}$ ,  $B = \begin{bmatrix} 2 & -3 \\ 0 & 1 \\ 4 & -7 \end{bmatrix}$ . Compute  $AB$ , if it is defined.

*Solution:* Since  $A$  is  $2 \times 3$  and  $B$  is  $3 \times 2$ , then  $AB$  is defined and  $AB$  is  $\underline{\quad} \times \underline{\quad}$ .

$$AB = \begin{bmatrix} 2 & 3 & 6 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & -3 \\ 0 & 1 \\ 4 & -7 \end{bmatrix} = \begin{bmatrix} 28 & \blacksquare \\ \blacksquare & \blacksquare \end{bmatrix}, \quad \begin{bmatrix} 2 & 3 & 6 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & -3 \\ 0 & 1 \\ 4 & -7 \end{bmatrix} = \begin{bmatrix} 28 & -45 \\ \blacksquare & \blacksquare \end{bmatrix}$$

$$\begin{bmatrix} 2 & 3 & 6 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & -3 \\ 0 & 1 \\ 4 & -7 \end{bmatrix} = \begin{bmatrix} 28 & -45 \\ 2 & \blacksquare \end{bmatrix}, \quad \begin{bmatrix} 2 & 3 & 6 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & -3 \\ 0 & 1 \\ 4 & -7 \end{bmatrix} = \begin{bmatrix} 28 & -45 \\ 2 & -4 \end{bmatrix}$$

So  $AB = \begin{bmatrix} 28 & -45 \\ 2 & -4 \end{bmatrix}$ .

## THEOREM 2

Let  $A$  be  $m \times n$  and let  $B$  and  $C$  have sizes for which the indicated sums and products are defined.

- a.  $A(BC) = (AB)C$  (associative law of multiplication)
- b.  $A(B + C) = AB + AC$  (left - distributive law)
- c.  $(B + C)A = BA + CA$  (right-distributive law)
- d.  $r(AB) = (rA)B = A(rB)$   
for any scalar  $r$
- e.  $I_m A = A = A I_n$  (identity for matrix multiplication)

## WARNINGS

Properties above are analogous to properties of real numbers. But **NOT ALL** real number properties correspond to matrix properties.

1. It is not the case that  $AB$  always equal  $BA$ . (see Example 7, page 114)
2. Even if  $AB = AC$ , then  $B$  may not equal  $C$ . (see Exercise 10, page 116)
3. It is possible for  $AB = 0$  even if  $A \neq 0$  and  $B \neq 0$ . (see Exercise 12, page 116)

## Powers of $A$

$$A^k = \underbrace{A \cdots A}_k$$

### EXAMPLE:

$$\begin{bmatrix} 1 & 0 \\ 3 & 2 \end{bmatrix}^3 = \begin{bmatrix} 1 & 0 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 3 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 3 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 21 & 8 \end{bmatrix}$$

If  $A$  is  $m \times n$ , the **transpose** of  $A$  is the  $n \times m$  matrix, denoted by  $A^T$ , whose columns are formed from the corresponding rows of  $A$ .

### EXAMPLE:

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

**EXAMPLE:** Let  $A = \begin{bmatrix} 1 & 2 & 0 \\ 3 & 0 & 1 \end{bmatrix}$ ,  $B = \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ -2 & 4 \end{bmatrix}$ . Compute  $AB$ ,  $(AB)^T$ ,  $A^T B^T$  and  $B^T A^T$ .

*Solution:*

$$AB = \begin{bmatrix} 1 & 2 & 0 \\ 3 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ -2 & 4 \end{bmatrix} = \begin{bmatrix} \quad & \quad \\ \quad & \quad \end{bmatrix}$$

$$(AB)^T = \begin{bmatrix} \quad & \quad \\ \quad & \quad \end{bmatrix}$$

$$A^T B^T = \begin{bmatrix} 1 & 3 \\ 2 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & -2 \\ 2 & 1 & 4 \end{bmatrix} = \begin{bmatrix} 7 & 3 & 10 \\ 2 & 0 & -4 \\ 2 & 1 & 4 \end{bmatrix}$$

$$B^T A^T = \begin{bmatrix} 1 & 0 & -2 \\ 2 & 1 & 4 \end{bmatrix} \begin{bmatrix} 1 & 3 \\ 2 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} \quad & \quad \\ \quad & \quad \end{bmatrix}$$

### THEOREM 3

Let  $A$  and  $B$  denote matrices whose sizes are appropriate for the following sums and products.

- $(A^T)^T = A$  (i.e., the transpose of  $A^T$  is  $A$ )
- $(A + B)^T = A^T + B^T$
- For any scalar  $r$ ,  $(rA)^T = rA^T$
- $(AB)^T = B^T A^T$  (i.e. the transpose of a product of matrices equals the product of their transposes in reverse order.)

**EXAMPLE:** Prove that  $(ABC)^T = \underline{\hspace{2cm}}$ .

*Solution:* By Theorem 3d,

$$(ABC)^T = ((AB)C)^T = C^T( \quad )^T = C^T( \quad ) = \underline{\hspace{2cm}}.$$