Section 2.1 Matrix Operations

O₁ <u>Definitions</u>:

- 1. The diagonal entries in an $m \times n$ matrix $A = [a_{ij}]$ are a_{11}, a_{22}, \cdots , and they form the main diagonal of A.
- 2. A diagonal matrix is a square $n \times n$ matrix whose nondiagonal entris are zero.
- 3. An $m \times n$ matrix whose entries are all zero is a zero matrix and is written as 0.
- 4. Two matrices are equal if they have the same size and if their corresponding entries are equal.

Sums and Scalar Multiples:

- 1. If A and B are $m \times n$ matrices, then the sum A + B is the $m \times n$ matrix whose columns are the sums of the corresponding columns in A and B.
- 2. If r is a scalar and A is a matrix, then the scalar multiple rA is the matrix whose columns are r times the corresponding columns in A.

Theorem: Let *A*, *B* and *C* be matrices of the same size, and let *r* and *s* be scalars.

$$A + B = B + A$$
 $r(A + B) = rA + rB$ $(A + B) + C = A + (B + C)$ $(r + s)A = rA + sA$ $A + 0 = A$ $r(sA) = rsA$

Example 1: Let $A = \begin{bmatrix} 2 & 0 \\ 4 & -5 \end{bmatrix}$ and $B = \begin{bmatrix} 7 & -5 \\ 1 & -4 \end{bmatrix}$, compute -3A and A - 2B. $-3A = -3 \begin{bmatrix} 2 & 0 \\ 4 & -5 \end{bmatrix} = \begin{pmatrix} -3 & 2 & 3 & 0 \\ -3 & 4 & 3 & -5 \end{bmatrix} = \begin{pmatrix} -4 & 0 \\ -12 & 15 \end{bmatrix}$ $A - 2B = \begin{pmatrix} 2 & 0 \\ 4 & -5 \end{pmatrix} - 2 \begin{pmatrix} 7 & -5 \\ -3 & 4 & 3 & -5 \end{bmatrix} = \begin{pmatrix} 2 & 0 \\ 4 & -3 \end{pmatrix} - \begin{pmatrix} 14 & -10 \\ 2 & -8 \end{pmatrix}$ Matrix Multiplication:

Recall: When a matrix B multiplies a vector \mathbf{x} , it transforms \mathbf{x} into the vector $B\mathbf{x}$.

If this vector is then multiplied in turn by a matrix A, the resulting vector is $A(B\mathbf{x})$. Thus $A(B\mathbf{x})$ is produced from \mathbf{x} by a composition of linear transformations.

In the dat product of
$$\vec{v}_1 = (a_1 \ v_2 \ \dots \ a_p) \ dp^T$$

 $S \ \vec{v}_2 = (b_1, \ b_2, \dots \ b_p) \ dp^P$
 $\vec{v}_1 \ \vec{v}_2 = (a_1 \ b_1 + c_2 \ b_2 + \dots \ a_p \ b_p.$
I. $A \ t \ p^{(n,n)}, \ B \ c \ p^{(n,n)}, \ B \ c \ p^{(n,n)}, \ b \ c \ dp^{(n,n)}, \ dp^{(n,n)}, \ B \ c \ p^{(n,n)}, \ b \ c \ dp^{(n,n)}, \ dp^{(n,n)}, \ dp^{(n,n)}, \ B \ c \ p^{(n,n)}, \ dp^{(n,n)}, \ d$

pe Vew

I. Dot product between $\vec{a} = (a_1 \ a_2 \ \dots \ a_n)$ 8 $\vec{b} = (b_1 \ b_2 \ \dots \ b_n)$; $a_1 \ b_1 + a_2 \ b_2 + \dots + a_n \ b_n$

I. A EIR", B E R" &
A B is unfined only if
$$n = p$$

(# of cols of the first matrix A
= # of rows of the second matrix B)

$$\mathbf{H} \cdot (\mathbf{A} \cdot \mathbf{B}) \quad \mathbf{13} \quad \mathbf{s} + \mathbf{10} \quad \mathbf{a} \quad \mathbf{matpix}, \quad \mathbf{fine} \quad \mathbf{encion} \quad \mathbf{is} : \quad \mathbf{m} \cdot \mathbf{g}$$

$$\mathbf{H} \cdot \mathbf{f} \quad \mathbf{rous} \quad \mathbf{cg} \quad \mathbf{A} \cdot \mathbf{g} = \mathbf{H} \quad \mathbf{of} \quad \mathbf{rous} \quad \mathbf{of} \quad \mathbf{A} = \mathbf{m}$$

$$\mathbf{H} \cdot \mathbf{f} \quad \mathbf{cols} \quad \mathbf{of} \quad \mathbf{A} \cdot \mathbf{g} = \mathbf{H} \quad \mathbf{of} \quad \mathbf{cols} \quad \mathbf{of} \quad \mathbf{B} = \mathbf{g}.$$

We would like to represent this composite mapping as multiplication by a single matrix, denoted by AB, so that

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

Definition: If *A* is an $m \times n$ matrix, and if *B* is an $n \times p$ matrix with columns $\mathbf{b}_1, \dots, \mathbf{b}_p$, then the product *AB* is an $m \times p$ matrix whose columns are $A\mathbf{b}_1, \dots, A\mathbf{b}_p$, i.e.

Remark:

- Each column of *AB* is a linear combination of the columns of *A* using weights from the corresponding column of *B*.
- *AB* has the same number of rows as *A* and the same number of columns as *B*.

Example 2: If a matrix A is 4×5 and the product AB is 4×6 , what is the size of B?

Row-column rule for computing *AB***:**

If the product AB is defined, then the entry in row *i* and column *j* of AB is the sum of the products of corresponding entries from row *i* of *A* and column *j* of *B*.

If $(AB)_{ij}$ denotes the (i, j)-entry in AB, and if A is an $m \times n$ matrix, then

Example 3: Let $A = \begin{bmatrix} 2 & 0 \\ 4 & -5 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 2 \\ -2 & 1 \end{bmatrix}$, compute the product AB by the row-column rule for computing AB. A $\cdot \mathbf{b}_{2}, \mathbf{2}$ $= \begin{bmatrix} \mathbf{a} \cdot \mathbf{c} \cdot \mathbf{$

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

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 3 & 4 & 16 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 2 \\ 4 & 5 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 4 & 16 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 4 & 6 \end{bmatrix} = \begin{bmatrix}$$

Theorem: Properties of Matrix Multiplication:

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

(left distributive law)

(right distributive law

(identity for matrix multiplication)

 $\mathbf{I}_{\mathbf{J}} = \begin{bmatrix} \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{1} \end{bmatrix}$

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

WARNINGS:

- **1.** In general, $AB \neq BA$.
- 2. The cancellation laws do *not* hold for matrix multiplication. That is, if AB = AC, then it is *not* true in general that B = C.
- 3. If a product *AB* is the zero matrix, you *cannot* conclude in general that either A = 0 or B = 0.

Power of a Matrix:

A^k = A·A·A...A k copies of A

If A is an $n \times n$ matrix and if k is a positive integer, then A^k denotes the product of k copies of A.

Example 4: Let $A = \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix}$ and $D = \begin{bmatrix} 1 & 0 \\ 0 & 3 \end{bmatrix}$. Compute *AD* and *DA*. Find a 2 × 2 matrix $B \neq I_2$, such that AB = BA.

The transpose of a Matrix:

Given an $m \times n$ matrix \overline{A} , the transpose of A is the $n \times m$ matrix, denoted by A^T , whose columns are formed from the corresponding rows of A.

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

 A^{t}

- a. $(A^T)^T = A$
- $b. \ (A+B)^T = A^T + B^T$
- c. For any scalar r, $(rA)^T = rA^T$
- d. $(AB)^T = B^T A^T$

Remark: (d)The transpose of a product of matrices equals the product of their transposes in the reverse order.

Example 6: Let
$$\mathbf{u} = \begin{bmatrix} -3 \\ 2 \end{bmatrix}$$
 and $\mathbf{v} = \begin{bmatrix} a \\ b \end{bmatrix}$, compute $\mathbf{u}^T \mathbf{v}, \mathbf{v}^T \mathbf{u}, \mathbf{u} \mathbf{v}^T, \mathbf{v} \mathbf{u}^T$.
A = $\begin{pmatrix} \mathbf{v} \\ \mathbf{v} \\ \mathbf{v} \end{pmatrix}$