September 24 Math 3260 sec. 53 Fall 2025

Chapter 3: Matrix Algebra

For $m \times n$ matrices $A = [a_{ij}]$ and $B = [b_{ij}]$ and scalar c, we defined

the row and column vectors,

$$\mathsf{Row}_i(A) = \langle a_{i1}, a_{i2}, \dots, a_{in} \rangle, \quad \mathsf{for} \quad i = 1, \dots, m$$
 and $\mathsf{Col}_i(A) = \langle a_{1j}, a_{2j}, \dots, a_{mj} \rangle, \quad \mathsf{for} \quad i = j, \dots, n$

- ▶ matrix addition: $A + B = [a_{ij} + b_{ij}]$, and
- ▶ scalar multiplication $cA = [ca_{ij}]$.

Recall that the entries can be referenced using two notations,

$$a_{ij}=A_{(i,j)}.$$



3.3 Multiplication of Two Matrices

Suppose *A* is an $m \times p$ matrix and *B* is a $p \times n$ matrix. Then the product *AB* is the $m \times n$ matrix

$$AB = [(AB)_{(i,j)}], \quad \text{where} \quad (AB)_{(i,j)} = \text{Row}_i(A) \cdot \text{Col}_j(B).$$

If the number of columns of *A* does not match the number of rows of *B*, then *AB* is not defined.

Transpose

Suppose $A = [a_{ij}]$ is an $m \times n$ matrix. The matrix A^T , called the **transpose** of A, is the $n \times m$ matrix defined by

$$\left(\mathbf{A}^{T}\right)_{(i,j)}=\mathbf{A}_{(j,i)}.$$

$$Row_i(A) = Col_i(A^T)$$
 and $Col_i(A) = Row_i(A^T)$.

Algebraic Properties

Let *A*, *B*, and *C* be matrices such that the appropriate sums and products are defined, and let *r* be a scalar. Then

(i)
$$A(B+C) = AB + AC$$
 and $(A+B)C = AC + BC$

(ii)
$$A(rB) = (rA)B = r(AB)$$

(iii)
$$(A^T)^T = A$$

(iv)
$$(A + B)^T = A^T + B^T$$

$$(v) (rA)^T = rA^T$$

(vi)
$$(AB)^T = B^T A^T$$



Example:
$$A = \begin{bmatrix} 1 & 0 \\ 2 & -1 \\ 3 & 3 \end{bmatrix}, B = \begin{bmatrix} -2 & 1 \\ 2 & -4 \end{bmatrix}$$

Evaluate

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

3.
$$BA^{T} = \begin{bmatrix} -2 & -5 & -3 \\ 2 & 8 & -6 \end{bmatrix}$$



3.5 Multiplication of a Vector by a Matrix

The Product $A\vec{x}$

Suppose *A* is an $m \times n$ matrix and let \vec{x} be a vector in R^n . Then the **matrix-vector product** $A\vec{x}$ is the vector in R^m given by

$$A\vec{x} = \langle \mathsf{Row}_1(A) \cdot \vec{x}, \mathsf{Row}_2(A) \cdot \vec{x}, \dots, \mathsf{Row}_m(A) \cdot \vec{x} \rangle$$
.

- For $A\vec{x}$ to be defined, \vec{x} must have the same number of entries as A has columns.
- ► The vector $A\vec{x}$ has the same number of entries as A has rows.

$$A_{m \times n} \overset{\overrightarrow{R}^{n}}{\underset{R^{m}}{\bigvee}}$$

Example

Find the product
$$A\vec{x}$$
 if $A = \begin{bmatrix} 3 & 0 & 1 & 3 \\ 1 & -1 & 2 & 0 \\ 0 & 2 & 0 & -1 \end{bmatrix}$ and $\vec{x} = \langle 1, -3, 0, 2 \rangle$.

$$A\vec{x} = \left\langle Pow_{1}(A) \cdot \vec{x}, Pow_{2}(A) \cdot \vec{x}, Pow_{3}(A) \cdot \vec{x} \right\rangle$$

$$Pow_{1}(A) \cdot \vec{x} : \left\langle 3, 0, 1, 3 \right\rangle \cdot \left\langle 1, 3, 0, 2 \right\rangle = 3 + 0 + 0 + 6 = 9$$

$$Pow_{3}(A) \cdot \vec{x} = \left\langle 1, -1, 2, 0 \right\rangle \cdot \left\langle 1, -3, 0, 2 \right\rangle = 1 + 3 + 0 + 0 = 9$$

$$Pow_{3}(A) \cdot \vec{x} = \left\langle 0, 2, 0, -1 \right\rangle \cdot \left\langle 1, -3, 0, 2 \right\rangle = 6 - 6 + 0 - 2 = -8$$



Compatibility w/ Matrix Multiplication

If $\vec{x} = \langle x_1, x_2, \dots, x_n \rangle$ is a vector in \mathbb{R}^n , and we define an $n \times 1$ matrix X via $Col_1(X) = \vec{x}$, i.e.,

then the matrix-vector product

$$AX = \begin{bmatrix} \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{1}(X) \\ \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{1}(X) \\ \vdots \\ \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{1}(X) \end{bmatrix} = \begin{bmatrix} \operatorname{Row}_{1}(A) \cdot \vec{x} \\ \operatorname{Row}_{2}(A) \cdot \vec{x} \\ \vdots \\ \operatorname{Row}_{m}(A) \cdot \vec{x} \end{bmatrix}$$

is the $m \times 1$ matrix such that $Col_1(AX) = A\vec{x}$.



Alternate Formulation of $A\vec{x}$

Let
$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix}$$
 and $\vec{x} = \langle x_1, x_2, x_3 \rangle$.

Evaluate the product $A\vec{x}$ and find three vectors \vec{u}_1 , \vec{u}_2 and \vec{u}_3 such that $A\vec{x}$ is a linear combination $x_1\vec{u}_1 + x_2\vec{u}_2 + x_3\vec{u}_3$.

$$\begin{array}{lll}
A\vec{\chi} & A\vec{\chi} = \left\langle Ro\omega_{1}(A) \cdot \vec{\chi} \right\rangle Ro\omega_{2}(A) \cdot \vec{\chi} \\
\downarrow^{2} & = \left\langle a_{11} \times_{1} + a_{12} \times_{2} + a_{13} \times_{3} \right\rangle a_{21} \times_{1} + a_{22} \times_{2} + a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{1} + \left\langle a_{12} \times_{2} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{1} + \left\langle a_{12} \times_{2} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{1} + \left\langle a_{12} \times_{2} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{1} + \left\langle a_{12} \times_{2} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{1} + \left\langle a_{12} \times_{2} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{1} + \left\langle a_{12} \times_{2} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{1} + \left\langle a_{12} \times_{2} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{1} + \left\langle a_{12} \times_{2} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{1} + \left\langle a_{12} \times_{2} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{2} + \left\langle a_{12} \times_{3} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{2} + \left\langle a_{12} \times_{3} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{2} + \left\langle a_{12} \times_{3} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{2} + \left\langle a_{12} \times_{3} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} \\
&= \left\langle a_{11} \times_{1} \right\rangle a_{21} \times_{2} + \left\langle a_{12} \times_{3} \right\rangle a_{22} \times_{2} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_{3} + \left\langle a_{13} \times_{3} \right\rangle a_{23} \times_$$

$$\vec{u}_{1} = (a_{11}, a_{21})$$
 $\vec{u}_{2} = (a_{12}, a_{22})$
 $\vec{u}_{3} = (a_{13}, a_{23})$

 $A = \left| \begin{array}{ccc} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{array} \right|$

Alternate Formulation of $A\vec{x}$

For $\vec{x} = \langle x_1, x_2, \dots, x_n \rangle$ in \mathbb{R}^n and A and $m \times n$ matrix,

$$A\vec{x} = x_1 \operatorname{Col}_1(A) + x_2 \operatorname{Col}_2(A) + \cdots + x_n \operatorname{Col}_n(A).$$

That is, $A\vec{x}$ is the vector in R^m that is the linear combination of the columns of A with the entries of \vec{x} as the weights.

Remark: We have two equivalent definitions for the product $A\vec{x}$. One focuses on the role of the rows of A, and the other focuses on the role of the columns of A.

Example

Find the product $A\vec{x}$ using the second formulation, where

$$A = \begin{bmatrix} 3 & 0 & 1 & 3 \\ 1 & -1 & 2 & 0 \\ 0 & 2 & 0 & -1 \end{bmatrix}$$
 and $\vec{x} = \langle 1, -3, 0, 2 \rangle$.

$$A_{X}: x_{1}Gl_{1}(A) + X_{2}Gl_{2}(A) + X_{3}Gl_{3}(A) + X_{4}Gl_{4}(A)$$

$$= 1 (3,1,0) + (-3)(0,-1,2) + 0(1,2,0) + 2(3,0,-1)$$

$$= (3,1,0) + (0,3,-6) + (0,0,0) + (6,0,-2)$$

$$= (3+0+0+6, 1+3+0+0,0-6-0-2)$$

$$= (9,4,-8)$$

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AB = \begin{bmatrix} \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{1}(B) & \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{2}(B) & \cdots & \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{n}(B) \\ \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{1}(B) & \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{2}(B) & \cdots & \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{n}(B) \\ \vdots & \vdots & \ddots & \vdots \\ \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{1}(B) & \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{2}(B) & \cdots & \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{n}(B) \end{bmatrix}
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Look at the first column:

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\begin{bmatrix} \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{1}(B) \\ \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{1}(B) \\ \vdots \\ \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{1}(B) \end{bmatrix}
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$$AB = \begin{bmatrix} \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{1}(B) & \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{2}(B) & \cdots & \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{n}(B) \\ \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{1}(B) & \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{2}(B) & \cdots & \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{n}(B) \\ \vdots & \vdots & \ddots & \vdots \\ \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{1}(B) & \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{2}(B) & \cdots & \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{n}(B) \end{bmatrix}$$

Look at the first column:

$$\left[\begin{array}{c} \operatorname{Row}_1(A) \cdot \operatorname{Col}_1(B) \\ \operatorname{Row}_2(A) \cdot \operatorname{Col}_1(B) \\ \vdots \\ \operatorname{Row}_m(A) \cdot \operatorname{Col}_1(B) \end{array} \right] \quad \text{looks like} \quad \left[\begin{array}{c} \operatorname{Row}_1(A) \cdot \operatorname{Col}_1(X) \\ \operatorname{Row}_2(A) \cdot \operatorname{Col}_1(X) \\ \vdots \\ \operatorname{Row}_m(A) \cdot \operatorname{Col}_1(X) \end{array} \right]$$

$$AB = \begin{bmatrix} \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{1}(B) & \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{2}(B) & \cdots & \operatorname{Row}_{1}(A) \cdot \operatorname{Col}_{n}(B) \\ \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{1}(B) & \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{2}(B) & \cdots & \operatorname{Row}_{2}(A) \cdot \operatorname{Col}_{n}(B) \\ \vdots & \vdots & \ddots & \vdots \\ \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{1}(B) & \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{2}(B) & \cdots & \operatorname{Row}_{m}(A) \cdot \operatorname{Col}_{n}(B) \end{bmatrix}$$

Look at the first column:

This is the matrix-vector product of A with the vector $Col_1(B)$

$$Col_1(AB) = A Col_1(B).$$



Column Vectors of the Product AB

If A in an $m \times p$ matrix and B is a $p \times n$ matrix, then the product AB is the $m \times n$ matrix whose columns are defined via the matrix-vector products

$$Col_i(AB) = A Col_i(B).$$

This says that each column vector of the product AB is a matrix-vector product that looks like " $A\vec{x}$ " where the matrix is the matrix A from the left and the vector is a column vector from the matrix on the right, B.

Example

$$A = \begin{bmatrix} 1 & -3 \\ -2 & 2 \end{bmatrix}$$
, and $B = \begin{bmatrix} 2 & 0 & 2 \\ 1 & -4 & 6 \end{bmatrix}$

Determine the product AB by computing the column vectors as the matrix-vector products $Col_i(AB) = A Col_i(B)$.

AB
$$Col_{1}(AB) = AGl_{1}(B)$$

$$= \begin{bmatrix} 1 & -3 \\ -2 & z \end{bmatrix} \langle 2,1 \rangle$$

$$= \langle 2-3, -4+2 \rangle = \langle -1, -2 \rangle$$

$$Gl_{2}(AB) = AGl_{2}(B) = \begin{bmatrix} 1 & -3 \\ -2 & z \end{bmatrix} \langle 0, -4 \rangle$$

$$= \langle 0+12, 0-8 \rangle = \langle 12, -8 \rangle$$

$$A = \begin{bmatrix} 1 & -3 \\ -2 & 2 \end{bmatrix}, \text{ and } B = \begin{bmatrix} 2 & 0 & 2 \\ 1 & -4 & 6 \end{bmatrix}$$

$$(31_3(AB) = AG1_3(B)$$

= $\begin{bmatrix} 1 & -3 \\ -2 & z \end{bmatrix}(2,6) = (2-18,-4+12) = (-16,8)$

$$AB = \begin{bmatrix} -1 & 12 & -16 \\ -2 & -8 & 8 \end{bmatrix}$$

The Product AB

If *A* is an $m \times p$ matrix and *B* is a $p \times n$ matrix, then the product *AB* is the $m \times n$ matrix whose row vectors are given by

$$Row_i(AB) = B^T Row_i(A), \quad i = 1, ..., m.$$

We know that $(XY)^T = Y^TX^T$ and $Col_i(XY) = X Col_i(Y)$.



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The Product $A^T \vec{x}$

Suppose A is an $m \times n$ matrix and \vec{x} is a vector in R^m . Then the product $A^T \vec{x}$ is the vector in R^n given by

$$A^T \vec{x} = \langle \mathsf{Col}_1(A) \cdot \vec{x}, \mathsf{Col}_2(A) \cdot \vec{x}, \dots, \mathsf{Col}_n(A) \cdot \vec{x} \rangle$$

$$= x_1 \, \mathsf{Row}_1(A) + x_2 \, \mathsf{Row}_2(A) + \dots + x_m \, \mathsf{Row}_m(A).$$

Since $Row_i(A^T) = Col_i(A)$ and $Col_i(A^T) = Row_i(A)$, these match our previous representations for A^T instead of A:

$$A^{T}\vec{x} = \left\langle \underbrace{\text{Row}_{1}(A^{T})}_{\text{Col}_{1}(A)} \cdot \vec{x}, \underbrace{\text{Row}_{2}(A^{T})}_{\text{Col}_{2}(A)} \cdot \vec{x}, \dots, \underbrace{\text{Row}_{n}(A^{T})}_{\text{Col}_{n}(A)} \cdot \vec{x} \right\rangle$$

$$A^{T}\vec{x} = x_{1} \underbrace{\text{Col}_{1}(A^{T})}_{\text{Row}_{1}(A)} + x_{2} \underbrace{\text{Col}_{1}(A^{T})}_{\text{Row}_{2}(A)} + \dots + x_{m} \underbrace{\text{Col}_{m}(A^{T})}_{\text{Row}_{m}(A)}$$

Example

Let
$$A = \begin{bmatrix} 1 & -1 \\ 0 & 2 \\ -2 & 1 \end{bmatrix}$$
, $\vec{x} = \langle 3, 1 \rangle$ and $\vec{y} = \langle -2, -1, 1 \rangle$. Evaluate

$$A\vec{x}$$
, $A^T\vec{y}$, $(A\vec{x}) \cdot \vec{y}$, and $\vec{x} \cdot (A^T\vec{y})$.

$$A_{X}^{2} = \begin{bmatrix} 1 & -1 \\ 0 & 2 \\ -2 & 1 \end{bmatrix} (3,1) = (3-1,0+2,-6+1)$$

$$= (2,2,-5)$$

$$A^{T}\vec{y} = \begin{bmatrix} 1 & 0 & -2 \\ -1 & 2 & 1 \end{bmatrix} (-2,-1,1) = (-2+0-2,2-2+1) \\
= (-4,1)$$

$$A = \begin{bmatrix} 1 & -1 \\ 0 & 2 \\ -2 & 1 \end{bmatrix}, \vec{x} = \langle 3, 1 \rangle \text{ and } \vec{y} = \langle -2, -1, 1 \rangle$$

$$A\vec{x} = \langle 2, 2, -5 \rangle \quad A\vec{y} = \langle -4, 1 \rangle$$

$$(Ax)\cdot y = (2,2,-5)\cdot (-2,-1,1)$$

= -4-2-5 = -11

We found that

$$(A\vec{x})\cdot\vec{y}=\vec{x}\cdot(A^T\vec{y})$$

Dot Products & Transposes

Suppose A is an $m \times n$ matrix, \vec{x} is a vector in \mathbb{R}^n and \vec{y} is a vector in \mathbb{R}^m . Then

$$(A\vec{x})\cdot\vec{y}=\vec{x}\cdot(A^T\vec{y}).$$

Note that the left side of the equation is a dot product of vectors in R^m , whereas the right side is a dot product in R^n . So the number of products and sums on each side is different when $m \neq n$, but the final value is the same.