
Math 22 –
Linear Algebra and its
applications

- Lecture 9 -

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GENERAL INFORMATION

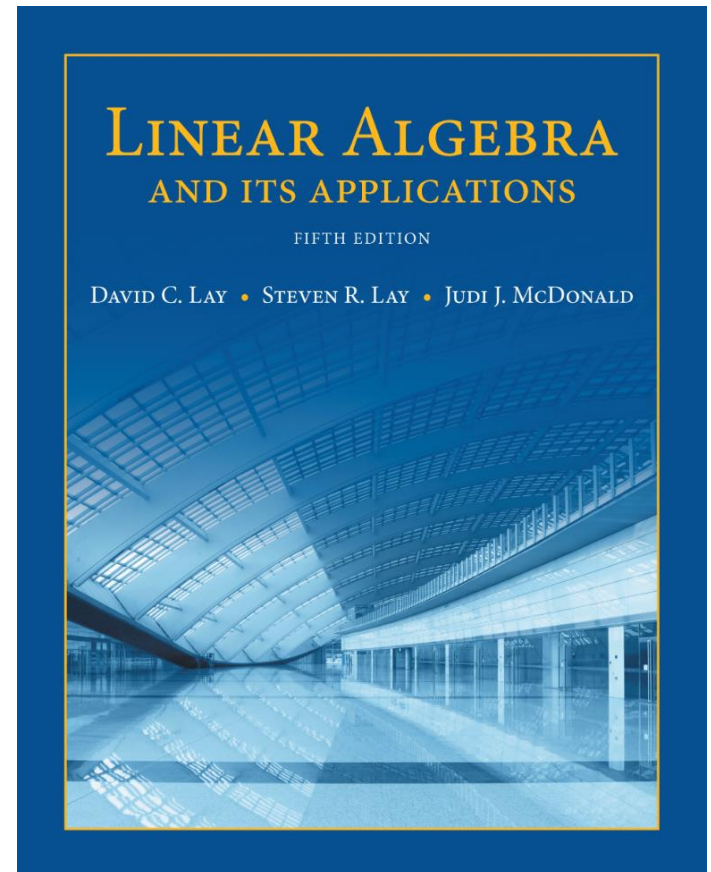
- **Office hours:** Tu 1-3 pm, Th, Sun 2-4 pm in KH 229
- **Tutorial:** Tu, Th, Sun 7-9 pm in KH 105
- **Midterm 1:** Monday Oct 7 from 4-6 pm in Carpenter 013
Topics: till this Thursday (included)
You can find the **practice exam** online

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Matrix Algebra

2.1

MATRIX OPERATIONS



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- **Summary:** The **composition of linear maps** is equal to the **multiplication of the associated standard matrices.**

MATRICES

- If A is an $m \times n$ matrix, then A has m rows and n columns.
The entry in the i -th row and j -th column of A is denoted by a_{ij} and is called the (i, j) -entry of A .
- Each column of A is a vector in \mathbb{R}^m . The j -th column vector is denoted by \mathbf{a}_j

$$\begin{array}{c} \text{Column} \\ j \end{array} \begin{array}{c} \left[\begin{array}{ccccc} a_{11} & \cdots & a_{1j} & \cdots & a_{1n} \\ \vdots & & \vdots & & \vdots \\ \text{Row } i & a_{i1} & \cdots & a_{ij} & \cdots & a_{in} \\ \vdots & & \vdots & & \vdots \\ a_{m1} & \cdots & a_{mj} & \cdots & a_{mn} \end{array} \right] \end{array} = A = [a_1 \ a_2 \ \cdots \ a_n]$$

\uparrow \uparrow \uparrow
 \mathbf{a}_1 \mathbf{a}_j \mathbf{a}_n

MATRICES

- **Note:** The **diagonal entries** of an $m \times n$ matrix $A = [a_{ij}]$ are $a_{11}, a_{22}, a_{33}, \dots$, and they form the **main diagonal** of A .
A **diagonal matrix** is a square $n \times n$ matrix whose nondiagonal entries are all zero.
- **Example:** I_n , the $n \times n$ identity matrix and $\mathbf{0}$ the $n \times n$ **zero** matrix.

$$I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \mathbf{0} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

SUM AND SCALAR MULTIPLICATION

- **Definition:** If A and B are $m \times n$ matrices, then the **sum** $A + B$ is the $m \times n$ matrix whose entries are the sums of the corresponding entries in A and B .
- **Note:** The sum $A + B$ is **defined only** when A and B are the **same size**.
- **Definition:** If r is a scalar and A is a matrix, then the **scalar multiple** rA is the matrix whose entries are r times the corresponding entries in A .

SUM AND SCALAR MULTIPLICATION

- **Theorem 1:** Let A , B , and C be matrices of the same size, and let r and s be scalars.
 - a. $A + B = B + A$
 - b. $(A + B) + C = A + (B + C)$
 - c. $A + 0 = A$
 - d. $r(A + B) = rA + rB$
 - e. $(r + s)A = rA + sA$
 - f. $r(sA) = (rs)A$

This is easily verified as all operations are performed **entrywise**.

SUM AND SCALAR MULTIPLICATION

Example: For $A = \begin{bmatrix} 4 & 0 & 5 \\ -1 & 3 & 2 \end{bmatrix}$, $B = \begin{bmatrix} 1 & 1 & 1 \\ 3 & 5 & 7 \end{bmatrix}$, $C = \begin{bmatrix} 2 & -3 \\ 0 & 1 \end{bmatrix}$

- 1.) Find $A+B$ and $A+C$.
- 2.) Find $3C$ and $A+2B$.

■ **Solution:**

COMPOSITION OF LINEAR MAPS

- **Theorem:** Let $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ and $S: \mathbb{R}^m \rightarrow \mathbb{R}^p$ be linear transformations with standard matrices B and A , respectively. Then the **composition** $S \circ T: \mathbb{R}^n \rightarrow \mathbb{R}^p$ is again a **linear transformation**.
- **Proof:**

- **What is the standard matrix C of the composite transformation $S \circ T: \mathbb{R}^n \rightarrow \mathbb{R}^p$?**

We know by **Theorem 10** that for the basis $\{e_1, e_2, \dots, e_n\}$ in \mathbb{R}^n

$$C = [S \circ T(e_1) \dots S \circ T(e_n)] .$$

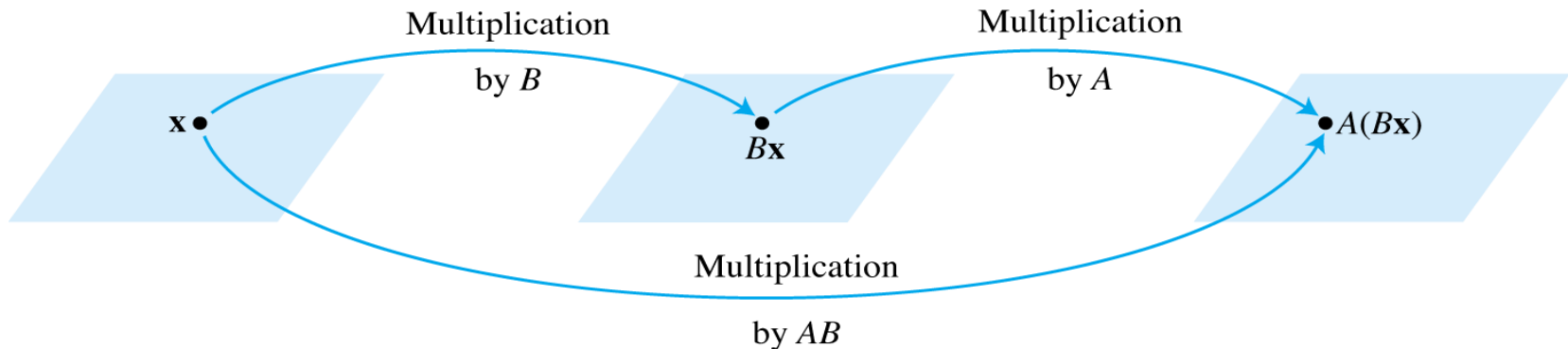
COMPOSITION OF LINEAR MAPS

- **Can we express C in terms of A and B ?**
- To this end we translate our problem into matrix notation:
- The image $T(\mathbf{x})$ of a vector \mathbf{x} is $B\mathbf{x}$.
- To get the image $S \circ T(\mathbf{x}) = S(T(\mathbf{x}))$ of \mathbf{x} , we must apply S to $B\mathbf{x}$.
- This is equal to multiplying A with the vector $B\mathbf{x}$. Hence
$$S \circ T(\mathbf{x}) = A(B\mathbf{x}).$$
- Thus $A(B\mathbf{x})$ is obtained from \mathbf{x} by the *composition of two linear mappings*.

MATRIX MULTIPLICATION

- **Goal: Define** matrix multiplication AB , such that AB is the matrix C of the composite transformation $S \circ T$.

$$A(Bx) = (AB)x = Cx$$



Multiplication by AB .

MATRIX MULTIPLICATION

- We can find C in terms of A and B by looking at the images of the basis $\{e_1, e_2, \dots, e_n\}$ in \mathbb{R}^n . We know that

$$C = [S \circ T(e_1) \dots S \circ T(e_n)] .$$

We have:

MATRIX MULTIPLICATION

Therefore we **define**:

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

$$AB = [Ab_1, Ab_2 \dots, Ab_n].$$

- **This way multiplication of matrices corresponds to composition of linear transformations.**

MATRIX MULTIPLICATION

- **Example:** Compute AB , where

$$A = \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} 4 & 3 & 1 \\ 1 & 0 & 3 \end{bmatrix}.$$

- **Solution:**

MATRIX MULTIPLICATION

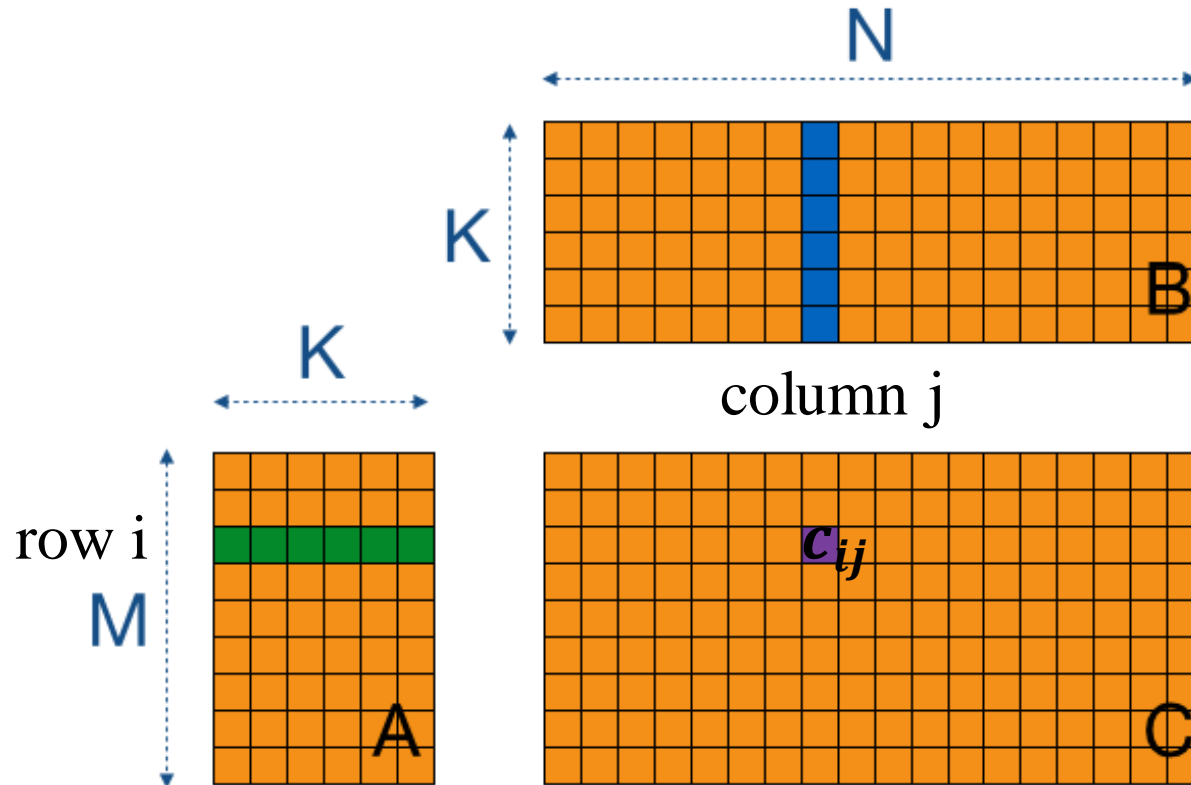
Row-column rule for computing $C=AB$

- If A is an $m \times k$ matrix, and if B is a $k \times n$ matrix then the product $C=AB$ is defined.
- In this case the
entry c_{ij} in row i and column j of $C=AB$,
is the **sum** of the **products** of corresponding
entries from **row i of A and column j of B .**

If $c_{ij}=(AB)_{ij}$, then

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

Row-column rule for computing $C=AB$



PROPERTIES OF MATRIX MULTIPLICATION

- **Theorem 2:** Let A , B and C be matrices that have sizes for which the indicated sums and products are defined. Then
 - 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)

PROPERTIES OF MATRIX MULTIPLICATION

- **Proof:** (a) follows from the fact that matrix multiplication corresponds to composition of linear mappings and as the composition of mappings is associative.
- (b)-(e) are easily verified using the definition.

PROPERTIES OF MATRIX MULTIPLICATION

- **Definition:** If $AB = BA$, we say that A and B **commute** with one another.

- **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 $B = 0$.

POWERS OF A MATRIX

- **Definition:** 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 :

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

- We set $A^0 = I_n$, the identity matrix.

THE TRANSPOSE OF A MATRIX

- **Definition:** Given an $m \times n$ matrix 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 .

Theorem 3: Let A and B denote matrices whose sizes are appropriate for the following sums and products.

a. $(A^T)^T = A$

b. $(A + B)^T = A^T + B^T$

c. For any scalar r , $(rA)^T = rA^T$

d. $(AB)^T = B^T A^T$

Example: For $A = \begin{bmatrix} 2 & 3 \\ 1 & -5 \end{bmatrix}$ and $B = \begin{bmatrix} 4 & 3 & 6 \\ 1 & -2 & 3 \end{bmatrix}$, find A^2 and B^T .

- **Solution:**

PROPERTIES OF MATRIX MULTIPLICATION

- **Exercise:** For each of the following statements find different nontrivial 2×2 matrices, that satisfy them.
 - 1.) $AB = BA$.
 - 2.) $AB \neq BA$.
 - 3.) $AB = AC$, but $B \neq C$.
 - 4.) $AB = 0$ but $A \neq 0$ and $B \neq 0$.

