# Linear Methods (Math 211) Lecture 7 - §2.2 & §2.3

(with slides adapted from K. Seyffarth)

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# Recall

- Associated Homogeneous Systems
- Matrix Transformations

# Today

More Matrix Transformations

Matrix Multiplication

Commutativity of Matrix Multiplication

## Zero and One

• If A is the  $m \times n$  matrix of all zeros, then the transformation induced by A, namely

$$T(\mathbf{x}) = \mathbf{0}$$
 for all  $\mathbf{x} \in \mathbb{R}^n$ ,

is called the zero transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$ , and is written T=0.

② If A is the  $n \times n$  identity matrix, then the transformation induced by A is called the identity transformation on  $\mathbb{R}^n$ , and is written  $1_{\mathbb{R}^n}$ . We have

$$1_{\mathbb{R}^n}(\mathbf{x}) = \mathbf{x} \text{ for all } \mathbf{x} \in \mathbb{R}^n.$$

## Polar Coordinates

Recall that we can specify a vector  $\begin{bmatrix} a \\ b \end{bmatrix} \in \mathbb{R}^2$  either in rectangular coordinates or in polar coordinates. In polar coordinates, we specify the distance r from the origin and the angle  $\theta$  from the positive x-axis. The two systems of coordinates are related by

$$a = r\cos(\theta)$$
$$b = r\sin(\theta)$$

$$r = \sqrt{a^2 + b^2}$$

$$\theta = \begin{cases} \tan^{-1}(\frac{b}{a}) & \text{if } a > 0 \\ \pi + \tan^{-1}(\frac{b}{a}) & \text{if } a < 0 \end{cases}$$

### Rotations

Suppose we want to rotate counterclockwise by an angle  $\alpha$ . Then a point with polar coordinates  $(r, \theta)$  will map to one with polar coordinates  $(r, \theta + \alpha)$ . In rectangular coordinates, this means

so rotation by  $\boldsymbol{\alpha}$  is a matrix transformation, induced by the matrix

$$\begin{bmatrix} \cos(\alpha) & -\sin(\alpha) \\ \sin(\alpha) & \cos(\alpha) \end{bmatrix}.$$

## **Translations**

A matrix transformation must map the zero vector  $\mathbf{0}$  to itself.

### Example

Let  $T: \mathbb{R}^2 \to \mathbb{R}^2$  be defined by

$$\mathcal{T}(\mathbf{x}) = \mathbf{x} + egin{bmatrix} 1 \ -1 \end{bmatrix} ext{ for all } \mathbf{x} \in \mathbb{R}^2.$$

This translation is not a matrix transformation.

# Matrix Multiplication

Let A be an  $m \times n$  matrix and  $B = \begin{bmatrix} \mathbf{b_1} & \mathbf{b_2} & \cdots & \mathbf{b_k} \end{bmatrix}$  an  $n \times k$  matrix, whose columns are  $\mathbf{b_1}, \mathbf{b_2}, \dots, \mathbf{b_k}$ . The product of A and B is the matrix

$$AB = A \begin{bmatrix} \mathbf{b}_1 & \mathbf{b}_2 & \cdots & \mathbf{b}_k \end{bmatrix} = \begin{bmatrix} A\mathbf{b}_1 & A\mathbf{b}_2 & \cdots & A\mathbf{b}_k \end{bmatrix},$$

i.e., the first column of AB is  $A\mathbf{b}_1$ , the second column of AB is  $A\mathbf{b}_2$ , etc.

Note that the number of **columns** of the first matrix must match the number of **rows** of the second.

Let A and B be matrices.

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

Then AB has columns

$$A\mathbf{b}_1 = \begin{bmatrix} -1 & 0 & 3 \\ 2 & -1 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 \\ -1 \end{bmatrix},$$

$$A\mathbf{b}_2 = \begin{bmatrix} -1 & 0 & 3 \\ 2 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ -2 \\ 0 \end{bmatrix} = \begin{bmatrix} -1 \\ 4 \end{bmatrix},$$

$$A\mathbf{b}_3 = \begin{bmatrix} -1 & 0 & 3 \\ 2 & -1 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 4 \\ 0 \end{bmatrix} = \begin{bmatrix} -2 \\ 0 \end{bmatrix}$$

## Example (continued)

Putting the columns together, we get

$$\begin{bmatrix} -1 & 0 & 3 \\ 2 & -1 & 1 \end{bmatrix} \cdot \begin{bmatrix} -1 & 1 & 2 \\ 0 & -2 & 4 \\ 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 4 & -1 & -2 \\ -1 & 4 & 0 \end{bmatrix}.$$

### Theorem ( $\S 2.3$ Theorem 1)

Let A be an  $m \times n$  matrix, and B an  $n \times k$  matrix. Then

$$A(B\mathbf{x}) = (AB)\mathbf{x}$$
 for all k-vectors  $\mathbf{x} \in \mathbb{R}^k$ .

#### Proof.

We can write  $B\mathbf{x}$  as  $x_1\mathbf{b}_1 + \cdots + x_k\mathbf{b}_k$ , so

$$A(B\mathbf{x}) = x_1 A \mathbf{b}_1 + \cdots + x_k A \mathbf{b}_k.$$

On the other hand, the columns of AB are  $A\mathbf{b}_1, \dots, A\mathbf{b}_k$ , so we have

$$(AB)\mathbf{x} = x_1 A \mathbf{b}_1 + \cdots + x_k A \mathbf{b}_k.$$

# Transformations and Matrix Multiplication

Suppose that  $\mathbb{R}^k \xrightarrow{T_B} \mathbb{R}^m \xrightarrow{T_A} \mathbb{R}^n$  are matrix transformations, induced by an  $n \times m$  matrix A and an  $m \times k$  matrix B. The composite of  $T_A$  and  $T_B$ , written  $T_A \circ T_B$ , is defined by  $(T_A \circ T_B)(\mathbf{x}) = T_A(T_B(\mathbf{x}))$  for  $\mathbf{x} \in \mathbb{R}^k$ .

The theorem implies that the composite of two matrix transformations is a matrix transformation, induced by AB.

Any *invertible* matrix transformation of  $\mathbb{R}^2$  can be written as a composite of shears, reflections and x or y-expansions.

# Dot Products and Matrix Multiplication

## Theorem ( $\S 2.3$ Theorem 2)

Let A be an  $m \times n$  matrix and B and  $n \times k$  matrix. Then the (i,j)-entry of AB is the dot product of row i of A with column j of B.

# Dot Products and Matrix Multiplication

### Theorem (§2.3 Theorem 2)

Let A be an  $m \times n$  matrix and B and  $n \times k$  matrix. Then the (i,j)-entry of AB is the dot product of row i of A with column j of B.

#### Example

Use the above theorem to compute

$$\begin{bmatrix} -1 & 0 & 3 \\ 2 & -1 & 1 \end{bmatrix} \cdot \begin{bmatrix} -1 & 1 & 2 \\ 0 & -2 & 4 \\ 1 & 0 & 0 \end{bmatrix}$$

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# Compatible Sizes

- Let A and B be matrices. In order for the product AB to exist, the number of rows in B must be equal to the number of columns in A.
- Assuming that A is an m × n matrix, the product AB is defined if and only if B is an n × k matrix for some k. If the product is defined, then A and B are said to be compatible for (matrix) multiplication.
- Given that A is  $m \times n$  and B is  $n \times k$ , the product AB is an  $m \times k$  matrix.

### Example (revisited)

As we saw earlier,

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

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As we saw earlier,

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

Note that the product

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

does not exist.

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

- Does AB exist? If so, compute it.
- Does BA exist? If so, compute it.

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

- Does AB exist? If so, compute it.
- Does BA exist? If so, compute it.

$$AB = \begin{bmatrix} 7 & -5 & 4 & -6 \\ -3 & 3 & -6 & 0 \\ -11 & 7 & -2 & 12 \end{bmatrix}$$

Let

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

- Does AB exist? If so, compute it.
- Does BA exist? If so, compute it.

$$AB = \begin{bmatrix} 7 & -5 & 4 & -6 \\ -3 & 3 & -6 & 0 \\ -11 & 7 & -2 & 12 \end{bmatrix}$$

BA does not exist

$$G = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
 and  $H = \begin{bmatrix} 1 & 0 \end{bmatrix}$ 

- Does GH exist? If so, compute it.
- Does HG exist? If so, compute it.

$$G = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
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- Does GH exist? If so, compute it.
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- Does GH exist? If so, compute it.
- Does HG exist? If so, compute it.

$$GH = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$$

$$HG = [1]$$

Let

$$G = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
 and  $H = \begin{bmatrix} 1 & 0 \end{bmatrix}$ 

- Does GH exist? If so, compute it.
- Does HG exist? If so, compute it.

$$GH = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$$

$$\mathit{HG} = \begin{bmatrix} 1 \end{bmatrix}$$

In this example, GH and HG both exist, but they are not equal. They aren't even the same size!

$$P = \begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix}$$
 and  $Q = \begin{bmatrix} -1 & 1 \\ 0 & 3 \end{bmatrix}$ 

- Does PQ exist? If so, compute it.
- Does QP exist? If so, compute it.

$$P = \begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix}$$
 and  $Q = \begin{bmatrix} -1 & 1 \\ 0 & 3 \end{bmatrix}$ 

- Does PQ exist? If so, compute it.
- Does QP exist? If so, compute it.

$$PQ = \begin{bmatrix} -1 & 1 \\ -2 & -1 \end{bmatrix}$$

$$P = \begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix}$$
 and  $Q = \begin{bmatrix} -1 & 1 \\ 0 & 3 \end{bmatrix}$ 

- Does PQ exist? If so, compute it.
- Does QP exist? If so, compute it.

$$PQ = \begin{bmatrix} -1 & 1 \\ -2 & -1 \end{bmatrix}$$

$$QP = \begin{bmatrix} 1 & -1 \\ 6 & -3 \end{bmatrix}$$

Let

$$P = \begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix}$$
 and  $Q = \begin{bmatrix} -1 & 1 \\ 0 & 3 \end{bmatrix}$ 

- Does PQ exist? If so, compute it.
- Does QP exist? If so, compute it.

$$PQ = \begin{bmatrix} -1 & 1 \\ -2 & -1 \end{bmatrix}$$
$$QP = \begin{bmatrix} 1 & -1 \\ 6 & -3 \end{bmatrix}$$

In this example, PQ and QP both exist and are the same size, but  $PQ \neq QP$ .

#### **Fact**

The four previous examples illustrate an important property of matrix multiplication.

In general, matrix multiplication is not commutative, i.e., the order of the matrices in the product is important.

$$U = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$
 and  $V = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ 

- Does UV exist? If so, compute it.
- Does *VU* exist? If so, compute it.

# **Commuting Matrices**

#### Example

$$U = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$
 and  $V = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ 

- Does UV exist? If so, compute it.
- Does VU exist? If so, compute it.

$$UV = \begin{bmatrix} 2 & 4 \\ 6 & 8 \end{bmatrix}$$

# **Commuting Matrices**

### Example

$$U = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$
 and  $V = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ 

- Does UV exist? If so, compute it.
- Does VU exist? If so, compute it.

$$UV = \begin{bmatrix} 2 & 4 \\ 6 & 8 \end{bmatrix}$$

$$VU = \begin{bmatrix} 2 & 4 \\ 6 & 8 \end{bmatrix}$$

# **Commuting Matrices**

#### Example

Let

$$U = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$
 and  $V = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ 

- Does UV exist? If so, compute it.
- Does VU exist? If so, compute it.

$$UV = \begin{bmatrix} 2 & 4 \\ 6 & 8 \end{bmatrix}$$

$$VU = \begin{bmatrix} 2 & 4 \\ 6 & 8 \end{bmatrix}$$

In this particular example, the matrices commute, i.e., UV = VU.

# Summary

More Matrix Transformations

Matrix Multiplication

Commutativity of Matrix Multiplication