

# Section 5.3 : Diagonalization

Chapter 5 : Eigenvalues and Eigenvectors

Math 1554 Linear Algebra

**Motivation:** it can be useful to take large powers of matrices, for example  $A^k$ , for large  $k$ .

**But:** multiplying two  $n \times n$  matrices requires roughly  $n^3$  computations. Is there a more efficient way to compute  $A^k$ ?

# Topics and Objectives

## Topics

1. Diagonal, similar, and diagonalizable matrices
2. Diagonalizing matrices

## Learning Objectives

For the topics covered in this section, students are expected to be able to do the following.

1. Determine whether a matrix can be diagonalized, and if possible diagonalize a square matrix.
2. Apply diagonalization to compute matrix powers.

# Diagonal Matrices

A matrix is **diagonal** if the only non-zero elements, if any, are on the main diagonal.

The following are all diagonal matrices.

$$\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}, \quad [2], \quad I_n, \quad \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

We'll only be working with diagonal square matrices in this course.

## Powers of Diagonal Matrices

If  $A$  is diagonal, then  $A^k$  is easy to compute. For example,

$$A = \begin{pmatrix} 3 & 0 \\ 0 & 0.5 \end{pmatrix}$$

$$A^2 = \begin{bmatrix} 3 & 0 \\ 0 & 0.5 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 0.5 \end{bmatrix} = \begin{bmatrix} 3^2 & 0 \\ 0 & (0.5)^2 \end{bmatrix}$$

$$A^k = \begin{bmatrix} 3^k & 0 \\ 0 & (0.5)^k \end{bmatrix}$$

But what if  $A$  is not diagonal?

Goal

$$A = P \cdot D \cdot P^{-1}$$
$$= P \cdot \begin{bmatrix} d_1 & & 0 \\ & d_2 & \\ 0 & & \ddots \\ & & & d_n \end{bmatrix} P^{-1}$$

$$A^2 = P \cdot D \cdot \underbrace{[P^{-1} \cdot P]}_{I} \cdot D \cdot P^{-1} = P \cdot D^2 \cdot P^{-1}$$

$$A^k = P \cdot D^k \cdot P^{-1}$$

# Diagonalization

Suppose  $A \in \mathbb{R}^{n \times n}$ . We say that  $A$  is **diagonalizable** if it is similar to a diagonal matrix,  $D$ . That is, we can write

$$A = P D P^{-1}$$

$$A P = P D.$$

$$A \begin{bmatrix} v_1 & v_2 & \dots & v_n \end{bmatrix} = \begin{bmatrix} v_1 & \dots & v_n \end{bmatrix} \begin{bmatrix} d_1 & & 0 \\ & d_2 & \\ 0 & & \ddots \\ & & & d_n \end{bmatrix}$$

$$= \begin{bmatrix} A v_1 & A v_2 & \dots & A v_n \end{bmatrix} = \begin{bmatrix} d_1 v_1 & d_2 v_2 & \dots & d_n v_n \end{bmatrix}$$

$$\left\{ \begin{array}{l} A v_1 = d_1 v_1 \\ A v_2 = d_2 v_2 \\ \vdots \\ A v_n = d_n v_n \end{array} \right.$$

Section 5.3 Slide 5

Always

$$A P = P D$$

$$P = \begin{bmatrix} v_1 & \dots & v_n \end{bmatrix}$$

$$D = \begin{bmatrix} d_1 & & 0 \\ & d_2 & \\ 0 & & \ddots \\ & & & d_n \end{bmatrix}$$

# Diagonalization

## Theorem

If  $A$  is diagonalizable  $\Leftrightarrow A$  has  $n$  linearly independent eigenvectors.

Note: the symbol  $\Leftrightarrow$  means “ if and only if ”.

Also note that  $A = PDP^{-1}$  if and only if

$$A = [\vec{v}_1 \ \vec{v}_2 \ \cdots \ \vec{v}_n] \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix} [\vec{v}_1 \ \vec{v}_2 \ \cdots \ \vec{v}_n]^{-1}$$

where  $\vec{v}_1, \dots, \vec{v}_n$  are linearly independent eigenvectors, and  $\lambda_1, \dots, \lambda_n$  are the corresponding eigenvalues (**in order**).

## Example 1

Diagonalize if possible.

$$A = \begin{pmatrix} 2 & 6 \\ 0 & -1 \end{pmatrix} = P \cdot D \cdot P^{-1}$$

$$\lambda_1 = 2$$

$$\lambda_2 = -1$$

$$\rightarrow D = \begin{bmatrix} 2 & 0 \\ 0 & -1 \end{bmatrix}$$

$$\underline{\underline{v_1}} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$v_2 = \begin{bmatrix} 2 \\ -1 \end{bmatrix}$$

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

$$\text{Null}(A - 2I) = \text{Null} \begin{bmatrix} 0 & 6 \\ 0 & -3 \end{bmatrix} = \text{Null} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$\text{Null}(A + I) = \text{Null} \begin{pmatrix} 3 & 6 \\ 0 & 0 \end{pmatrix} = \text{Null} \begin{pmatrix} 1 & 2 \\ 0 & 0 \end{pmatrix}$$

## Example 2

Diagonalize if possible.

$$\begin{pmatrix} 3 & 1 \\ 0 & 3 \end{pmatrix}$$



## Distinct Eigenvalues

### Theorem

If  $A$  is  $n \times n$  and has  $n$  distinct eigenvalues, then  $A$  is diagonalizable.

Why does this theorem hold?

Is it necessary for an  $n \times n$  matrix to have  $n$  distinct eigenvalues for it to be diagonalizable?

## Non-Distinct Eigenvalues

Theorem. Suppose

- $A$  is  $n \times n$
- $A$  has distinct eigenvalues  $\lambda_1, \dots, \lambda_k$ ,  $k \leq n$
- $a_i =$  algebraic multiplicity of  $\lambda_i$
- $d_i =$  dimension of  $\lambda_i$  eigenspace (“geometric multiplicity”)

Then

1.  $d_i \leq a_i$  for all  $i$
2.  $A$  is diagonalizable  $\Leftrightarrow \sum d_i = n \Leftrightarrow d_i = a_i$  for all  $i$
3.  $A$  is diagonalizable  $\Leftrightarrow$  the eigenvectors, for all eigenvalues, together form a basis for  $\mathbb{R}^n$ .

## Example 3

The eigenvalues of  $A$  are  $\lambda = 3, 1$ . If possible, construct  $P$  and  $D$  such that  $AP = PD$ .

$$A = \begin{pmatrix} 7 & 4 & 16 \\ 2 & 5 & 8 \\ -2 & -2 & -5 \end{pmatrix}$$

## Additional Example (if time permits)

Note that

$$\vec{x}_k = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \vec{x}_{k-1}, \quad \vec{x}_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad k = 1, 2, 3, \dots$$

generates a well-known sequence of numbers.

Use a diagonalization to find a matrix equation that gives the  $n^{\text{th}}$  number in this sequence.

## Basis of Eigenvectors

Express the vector  $\vec{x}_0 = \begin{bmatrix} 4 \\ 5 \end{bmatrix}$  as a linear combination of the vectors  $\vec{v}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  and  $\vec{v}_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$  and find the coordinates of  $\vec{x}_0$  in the basis  $\mathcal{B} = \{\vec{v}_1, \vec{v}_2\}$ .

$$[\vec{x}_0]_{\mathcal{B}} =$$

Let  $P = [\vec{v}_1 \ \vec{v}_2]$  and  $D = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ , and find  $[A^k \vec{x}_0]_{\mathcal{B}}$  where  $A = PDP^{-1}$ , for  $k = 1, 2, \dots$

$$[A^k \vec{x}_0]_{\mathcal{B}} =$$

## Basis of Eigenvectors - part 2

Let  $\vec{x}_0 = \begin{bmatrix} 4 \\ 5 \end{bmatrix}$ ,  $\vec{v}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  and  $\vec{v}_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$  as before.

Again define  $P = [\vec{v}_1 \ \vec{v}_2]$  but this time let  $D = \begin{bmatrix} 1 & 0 \\ 0 & -1/2 \end{bmatrix}$ , and now find  $[A^k \vec{x}_0]_{\mathcal{B}}$  where  $A = PDP^{-1}$ , for  $k = 1, 2, \dots$

$$[A^k \vec{x}_0]_{\mathcal{B}} =$$

## Basis of Eigenvectors - part 3

Let  $\vec{x}_0 = \begin{bmatrix} 4 \\ 5 \end{bmatrix}$ ,  $\vec{v}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  and  $\vec{v}_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$  as before.

Again define  $P = [\vec{v}_1 \ \vec{v}_2]$  but this time let  $D = \begin{bmatrix} 2 & 0 \\ 0 & 3/2 \end{bmatrix}$ , and now find  $[A^k \vec{x}_0]_{\mathcal{B}}$  where  $A = PDP^{-1}$ , for  $k = 1, 2, \dots$

$$[A^k \vec{x}_0]_{\mathcal{B}} =$$

## Chapter 5 : Eigenvalues and Eigenvectors

### 5.5 : Complex Eigenvalues



# Topics and Objectives

## Topics

1. Complex numbers: addition, multiplication, complex conjugate
2. Complex eigenvalues and eigenvectors.
3. Eigenvalue theorems

## Learning Objectives

1. Use eigenvalues to determine identify the rotation and dilation of a linear transform.
2. Rotation dilation matrices.
3. Find complex eigenvalues and eigenvectors of a real matrix.
4. Apply theorems to characterize matrices with complex eigenvalues.

## Motivating Question

What are the eigenvalues of a rotation matrix?

# Imaginary Numbers

**Recall:** When calculating roots of polynomials, we can encounter square roots of negative numbers. For example:

$$x^2 + 1 = 0$$

The roots of this equation are:

We usually write  $\sqrt{-1}$  as  $i$  (for “imaginary”).

## Addition and Multiplication

The imaginary (or complex) numbers are denoted by  $\mathbb{C}$ , where

$$\mathbb{C} = \{a + bi \mid a, b \text{ in } \mathbb{R}\}$$

We can identify  $\mathbb{C}$  with  $\mathbb{R}^2$ :  $a + bi \leftrightarrow (a, b)$

We can add and multiply complex numbers as follows:

$$(2 - 3i) + (-1 + i) =$$

$$(2 - 3i)(-1 + i) =$$

## Complex Conjugate, Absolute Value, Polar Form

We can **conjugate** complex numbers:  $\overline{a + bi} = \underline{\hspace{2cm}}$

The **absolute value** of a complex number:  $|a + bi| = \underline{\hspace{2cm}}$

We can write complex numbers in **polar form**:  $a + ib = r(\cos \phi + i \sin \phi)$

## Complex Conjugate Properties

If  $x$  and  $y$  are complex numbers,  $\vec{v} \in \mathbb{C}^n$ , it can be shown that:

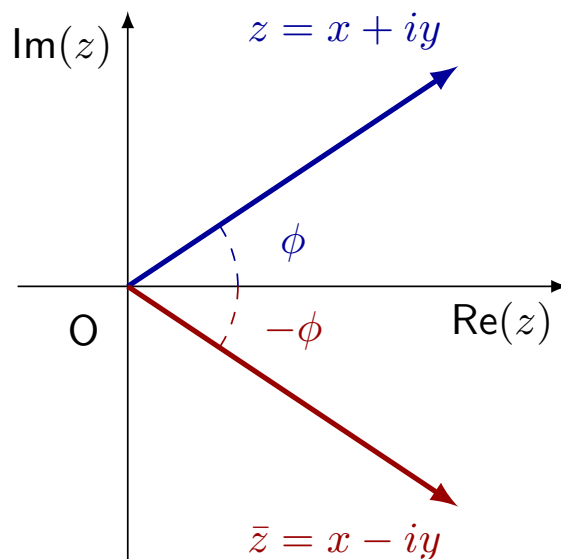
- $\overline{(x + y)} = \bar{x} + \bar{y}$
- $\overline{A\vec{v}} = A\vec{v}$
- $\text{Im}(x\bar{x}) = 0$ .

**Example** True or false: if  $x$  and  $y$  are complex numbers, then

$$\overline{(xy)} = \bar{x} \bar{y}$$

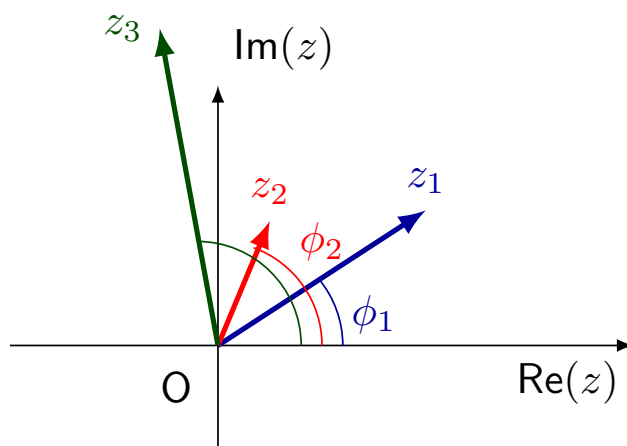
## Polar Form and the Complex Conjugate

Conjugation reflects points across the real axis.



## Euler's Formula

Suppose  $z_1$  has angle  $\phi_1$ , and  $z_2$  has angle  $\phi_2$ .



The product  $z_1 z_2$  has angle  $\phi_1 + \phi_2$  and modulus  $|z| |w|$ . Easy to remember using Euler's formula.

$$z = |z| e^{i\phi}$$

The product  $z_1 z_2$  is:

$$z_3 = z_1 z_2 = (|z_1| e^{i\phi_1})(|z_2| e^{i\phi_2}) = |z_1| |z_2| e^{i(\phi_1 + \phi_2)}$$

# Complex Numbers and Polynomials

## Theorem: Fundamental Theorem of Algebra

Every polynomial of degree  $n$  has exactly  $n$  complex roots, counting multiplicity.

## Theorem

1. If  $\lambda \in \mathbb{C}$  is a root of a real polynomial  $p(x)$ , then the conjugate  $\bar{\lambda}$  is also a root of  $p(x)$ .
2. If  $\lambda$  is an eigenvalue of real matrix  $A$  with eigenvector  $\vec{v}$ , then  $\bar{\lambda}$  is an eigenvalue of  $A$  with eigenvector  $\vec{\bar{v}}$ .



## Example

Four of the eigenvalues of a  $7 \times 7$  matrix are  $-2$ ,  $4 + i$ ,  $-4 - i$ , and  $i$ .  
What are the other eigenvalues?

## Example

The matrix that rotates vectors by  $\phi = \pi/4$  radians about the origin, and then scales (or dilates) vectors by  $r = \sqrt{2}$ , is

$$A = \begin{bmatrix} r & 0 \\ 0 & r \end{bmatrix} \begin{bmatrix} \cos \phi & -\sin \phi \\ \sin \phi & \cos \phi \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$$

What are the eigenvalues of  $A$ ? Find an eigenvector for each eigenvalue.

## Example

The matrix in the previous example is a special case of this matrix:

$$C = \begin{pmatrix} a & -b \\ b & a \end{pmatrix}$$

Calculate the eigenvalues of  $C$  and express them in polar form.

## Example

The matrix in the previous example is a special case of this matrix:

$$C = \begin{pmatrix} a & -b \\ b & a \end{pmatrix}$$

Calculate the eigenvalues of  $C$  and express them in polar form.

## Example

Find the complex eigenvalues and an associated complex eigenvector for each eigenvalue for the matrix.

$$A = \begin{pmatrix} 1 & -2 \\ 1 & 3 \end{pmatrix}$$