

# Eigenvalues and Eigenvectors

## §5.2 Diagonalization

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# Goals

Suppose  $A$  is square matrix of order  $n$ .

- ▶ Provide necessary and sufficient condition when there is an invertible matrix  $P$  such that  $P^{-1}AP$  is a diagonal matrix.

# Definitions

- ▶ Two square matrices  $A, B$  are said to be **similar**, if there is an invertible matrix  $P$ , such that  $A = P^{-1}BP$ .
- ▶ A square matrix  $A$  said to be **diagonalizable**, if there is an invertible matrix  $P$ , such that  $P^{-1}AP$  is a diagonal matrix. That means, if  $A$  is similar to a diagonal matrix, we say that  $A$  is **diagonalizable**.

## Theorem 5.2.1

Suppose  $A, B$  are two similar matrices. Then,  $A$  and  $B$  have same eigenvalues.

**Proof.** Write  $A = P^{-1}BP$ . Then

$$\begin{aligned} |\lambda I - A| &= |\lambda I - P^{-1}BP| = |\lambda(P^{-1}P) - P^{-1}BP| = |P^{-1}(\lambda I - B)P| \\ &= |P^{-1}| |\lambda I - B| |P| = |P|^{-1} |\lambda I - B| |P| = |\lambda I - B| \end{aligned}$$

So,  $A$  and  $B$  has same characteristic polynomials. So, they have same eigenvalues. The proof is complete. ■

## Theorem 5.2.2: Diagonalizability

We ask, when a square matrix is diagonalizable?

**Theorem 5.2.2** A square matrix  $A$ , of order  $n$ , is diagonalizable if and only if  $A$  has  $n$  linearly independent eigenvectors.

**Proof.** There are two statements to prove. First, suppose  $A$  is diagonalizable.

$$\text{Then } P^{-1}AP = D, \quad \text{and hence } AP = PD$$

where  $P$  is an invertible matrix and  $D$  is a diagonal matrix.

$$\text{Write, } D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}, \quad P = ( \mathbf{p}_1 \quad \mathbf{p}_2 \quad \cdots \quad \mathbf{p}_n )$$

## Continued

- ▶ Since  $AP = PA$ , we have

$$\begin{aligned}
 & A \left( \mathbf{p}_1 \quad \mathbf{p}_2 \quad \cdots \quad \mathbf{p}_n \right) \\
 &= \left( \mathbf{p}_1 \quad \mathbf{p}_2 \quad \cdots \quad \mathbf{p}_n \right) \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}.
 \end{aligned}$$

Or

$$\left( A\mathbf{p}_1 \quad A\mathbf{p}_2 \quad \cdots \quad A\mathbf{p}_n \right) = \left( \lambda_1\mathbf{p}_1 \quad \lambda_2\mathbf{p}_2 \quad \cdots \quad \lambda_n\mathbf{p}_n \right)$$

## Continued

- ▶ So,

$$A\mathbf{p}_i = \lambda_i\mathbf{p}_i \quad \text{for } i = 1, 2, \dots, n$$

Since  $P$  is invertible,  $\mathbf{p}_i \neq \mathbf{0}$  and hence  $\mathbf{p}_i$  is an eigenvector of  $A$ , for  $\lambda$ .

- ▶ Also,  $\text{rank}(P) = n$ . So, its columns  $\{\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_n\}$  are linearly independent.
- ▶ So, it is established that if  $A$  is diagonalizable, then  $A$  has  $n$  linearly independent eigenvectors.

## Continued

- ▶ Now, we prove the converse. So, we assume  $A$  has  $n$  linearly independent eigenvectors:

$$\{\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_n\}$$

- ▶ So,

$$A\mathbf{p}_1 = \lambda_1\mathbf{p}_1, A\mathbf{p}_2 = \lambda_2\mathbf{p}_2, \dots, A\mathbf{p}_n = \lambda_n\mathbf{p}_n \quad \text{for some } \lambda_i.$$



## Continued

- ▶ Write,

$$P = ( \mathbf{p}_1 \quad \mathbf{p}_2 \quad \cdots \quad \mathbf{p}_n ) \text{ and } D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}.$$

- ▶ It follows from the equations  $A\mathbf{p}_i = \lambda_i\mathbf{p}_i$  that

$$AP = PD. \text{ So, } P^{-1}AP = D \text{ is diagonal.}$$

The proof is complete. ■

# Steps for Diagonalizing

Suppose  $A$  is a square matrix of order  $n$ .

- ▶ If  $A$  does not have  $n$  linearly independent eigenvectors, then  $A$  is not diagonalizable.
- ▶ When possible, find  $n$  linearly independent eigenvectors  $\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_n$  for  $A$  with corresponding eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$ .
- ▶ Then, write

$$P = \left( \begin{array}{cccc} \mathbf{p}_1 & \mathbf{p}_2 & \cdots & \mathbf{p}_n \end{array} \right) \text{ and } D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}$$

- ▶ We have  $D = P^{-1}AP$  is a diagonal matrix.

## Corollary 4.4.3

Suppose  $V$  is a vectors space and  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$  be vectors in  $V$ . Then,  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$  are linearly dependent if and only if there is an integer  $m \leq n$  such that (1)  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m$  are linearly dependent and (2)  $\mathbf{x}_m \in \text{span}(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{m-1})$ .

**Proof.** Suppose  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$  are linearly dependent. By Theorem 4.4.2, one of these vectors is a linear combination of the rest. By relabeling, we can assume  $\mathbf{x}_n$  is a linear combination of  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1}$ . Let

$$m = \min\{k : \mathbf{x}_k \in \text{span}(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{k-1})\}$$

## Continued

If  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{m-1}$  are linearly dependent, then we could apply Theorem 4.4.2 again, which would lead to a contradiction, that  $m$  is minimum. So,  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{m-1}$  are linearly independent. This establishes one way implication. Conversely, suppose there is an  $m \leq n$  such that (1) and (2) holds. Then,

$$\mathbf{x}_m = c_1\mathbf{x}_1 + \cdots + c_{m-1}\mathbf{x}_{m-1} \quad \text{for some } c_1, \dots, c_{m-1} \in \mathbb{R}$$

So,

$$c_1\mathbf{x}_1 + \cdots + c_{m-1}\mathbf{x}_{m-1} + (-1)\mathbf{x}_m = \mathbf{0}$$

which is a nontrivial linear combination. So,  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m, \dots, \mathbf{x}_n$  are linearly dependent.

## Theorem 5.2.3: With Distinct Eigenvalues

Let  $A$  be a square matrix  $A$ , of order  $n$ . Suppose  $A$  has  $n$  **distinct** eigenvalues. Then

- ▶ the corresponding eigenvectors are linearly independent
- ▶ and  $A$  is diagonalizable.

### Proof.

- ▶ The second statement follows from the first, by theorem 5.2.2. So, we prove the first statement only.
- ▶ Let  $\lambda_1, \lambda_2, \dots, \lambda_n$  be distinct eigenvalues of  $A$ .
- ▶ So, for  $i = 1, 2, \dots, n$  we have

$$A\mathbf{x}_i = \lambda_i\mathbf{x}_i \quad \text{where } \mathbf{x}_i \neq 0 \quad \text{are eigenvectors.}$$

## Continued

- ▶ We need to prove that  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$  are linearly independent. We prove by contra-positive argument.
  - ▶ So, assume they are linearly dependent.
  - ▶ By Corollary 4.4.3 there is an  $m < n$  such that  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m$  are mutually linearly independent and  $\mathbf{x}_{m+1}$  is in can be written as a linear combination of  $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m\}$ . So,

$$\mathbf{x}_{m+1} = c_1\mathbf{x}_1 + c_2\mathbf{x}_2 + \cdots + c_m\mathbf{x}_m \quad (1)$$

Here, at least one  $c_i \neq 0$ . Re-labeling  $\mathbf{x}_i$ , if needed, we can assume  $c_1 \neq 0$ .

## Continued

- ▶ Multiply (1) by  $A$  on the left:

$$A\mathbf{x}_{m+1} = c_1 A\mathbf{x}_1 + c_2 A\mathbf{x}_2 + \cdots + c_m A\mathbf{x}_m \quad (2)$$

Now, use  $A\mathbf{x}_i = \lambda_i \mathbf{x}_i$ :

$$\lambda_{m+1} \mathbf{x}_{m+1} = \lambda_1 c_1 \mathbf{x}_1 + \lambda_2 c_2 \mathbf{x}_2 + \cdots + \lambda_m c_m \mathbf{x}_m \quad (3)$$

- ▶ Also, multiply (1) by  $\lambda_{m+1}$ , we have

$$\lambda_{m+1} \mathbf{x}_{m+1} = \lambda_{m+1} c_1 \mathbf{x}_1 + \lambda_{m+1} c_2 \mathbf{x}_2 + \cdots + \lambda_{m+1} c_m \mathbf{x}_m \quad (4)$$

## Continued

- ▶ Subtract (3) from (4):

$$(\lambda_{m+1} - \lambda_1)c_1\mathbf{x}_1 + (\lambda_{m+1} - \lambda_2)c_2\mathbf{x}_2 + \cdots + (\lambda_{m+1} - \lambda_m)c_m\mathbf{x}_m = \mathbf{0}.$$

- ▶ Since these vectors are linearly independent, and hence

$$(\lambda_{m+1} - \lambda_i)c_i = 0 \quad \text{for } i = 1, 2, \dots, m.$$

- ▶ Since  $c_1 \neq 0$  we get  $\lambda_{m+1} - \lambda_1 = 0$  or  $\lambda_{m+1} = \lambda_1$ . This contradicts that  $\lambda_i$ s are distinct. So, we conclude that  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$  are linearly independent. The proof is complete. ■



## Example 5.2.2

$$\text{Let } A = \begin{pmatrix} 2 & 3 & 1 \\ 0 & -1 & 2 \\ 0 & 0 & 3 \end{pmatrix} \quad \text{and} \quad P = \begin{pmatrix} 1 & 1 & 5 \\ 0 & -1 & 1 \\ 0 & 0 & 2 \end{pmatrix}.$$

Verify that  $A$  is diagonalizable, by computing  $P^{-1}AP$ .

**Solution:** We do it in a two steps.

1. Use TI to compute

$$P^{-1} = \begin{pmatrix} 1 & 1 & -3 \\ 0 & -1 & .5 \\ 0 & 0 & .5 \end{pmatrix}. \text{ So, } P^{-1}AP = \begin{pmatrix} 2 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 3 \end{pmatrix}.$$

So, it is verified that  $P^{-1}AP$  is a diagonal matrix.

## Example 5.2.3

$$\text{Let } A = \begin{pmatrix} 3 & 1 \\ -9 & -3 \end{pmatrix}.$$

Show that  $A$  is not diagonalizable.

**Solution:** Use Theorem 5.2.2 and show that  $A$  **does not have 2 linearly independent eigenvectors**. To do this, we have find and count the dimensions of all the eigenspaces  $E(\lambda)$ . We do it in a few steps.

- ▶ First, find all the eigenvalues. To do this, we solve

$$\det(\lambda I - A) = \begin{vmatrix} \lambda - 3 & -1 \\ 9 & \lambda + 3 \end{vmatrix} = \lambda^2 = 0.$$

So,  $\lambda = 0$  is the only eigenvalue of  $A$ .

## Continued

- ▶ Now we compute the eigenspace  $E(0)$  of the eigenvalue  $\lambda = 0$ . We have  $E(0)$  is solution space of

$$(0I - A) \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \text{ or } \begin{pmatrix} -3 & -1 \\ 9 & 3 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Using TI (or by hand), a parametric solution of this system is given by  $x = -.5t$   $y = t$ .

$$\text{So } E(0) = \{(t, -3t) : t \in \mathbb{R}\} = \mathbb{R}(1, -3).$$

So, the (sum of) dimension(s) of the eigenspace(s)  
 $= \dim E(0) = 1 < 2$ .

Therefore  $A$  is not diagonalizable.

## Example 5.2.3

$$\text{Let } A = \begin{pmatrix} 1 & 1 & 1 \\ 0 & -3 & 1 \\ 0 & 0 & -3 \end{pmatrix}.$$

Show that  $A$  is not diagonalizable.

**Solution:** Use Theorem 5.2.2 and show that  $A$  **does not have 3 linearly independent eigenvectors**.

- ▶ To find the eigenvalues, we solve

$$\det(\lambda I - A) = \begin{vmatrix} \lambda - 1 & -1 & -1 \\ 0 & \lambda + 3 & -1 \\ 0 & 0 & \lambda + 3 \end{vmatrix} = (\lambda - 1)(\lambda + 3)^2 = 0.$$

So,  $\lambda = 1, -3$  are the only eigenvalues of  $A$ .

## Continued

- ▶ We have  $E(1)$  is solution space of

$$(I - A) \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\text{Or } \begin{pmatrix} 0 & -1 & -1 \\ 0 & 4 & -1 \\ 0 & 0 & 4 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

*(As an alternative approach, avoid solving this system.)*

The (column) rank of the coefficient matrix is 2. So,  
 $\dim(E(1)) = \text{nullity} = 3 - \text{rank} = 3 - 2 = 1.$

## Continued

- ▶ Now we compute the dimension  $\dim E(-3)$ .  $E(-3)$  is the solution space of

$$(-3I - A) \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \quad \text{or}$$

$$\begin{pmatrix} -4 & -1 & -1 \\ 0 & 0 & -1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

The rank of the coefficient matrix is 2 (use TI, if you need). So,

$$\dim(E(-3)) = \text{nullity} = 3 - \text{rank} = 3 - 2 = 1.$$

# Continued

- ▶ So, the sum of dimensions of the eigenspaces  
$$= \dim E(1) + \dim E(-3) = 2 < 3.$$

Therefore  $A$  is not diagonalizable.

## Example 5.2.4

Let  $A = \begin{pmatrix} 17 & 113 & -2 \\ 0 & \sqrt{2} & 1 \\ 0 & 0 & \pi \end{pmatrix}$  Find its eigenvalues

and determine (use Theorem 5.2.3), if  $A$  is diagonalizable. If yes, write down a an invertible matrix  $P$  so that  $P^{-1}AP$  is a diagonal matrix.

**Solution:** To find eigenvalues solve

$$\begin{aligned} \det(\lambda I - A) &= \begin{vmatrix} \lambda - 17 & -113 & 2 \\ 0 & \lambda - \sqrt{2} & -1 \\ 0 & 0 & \lambda - \pi \end{vmatrix} \\ &= (\lambda - 17)(\lambda - \sqrt{2})(\lambda - \pi) = 0. \end{aligned}$$



## Continued

So,  $A$  has three distinct eigenvalues  $\lambda = 17, \sqrt{2}, \pi$ . Since  $A$  is a  $3 \times 3$  matrix, by Theorem 5.2.3,  $A$  is diagonalizable. We will proceed to compute the matrix  $P$ , by computing bases of  $E(17)$ ,  $E(\sqrt{2})$  and  $E(\pi)$ . ■

## Continued

To compute  $E(17)$ , we solve:  $(17I_3 - A)\mathbf{x} = \mathbf{0}$ , which is

$$\begin{pmatrix} 0 & -113 & 2 \\ 0 & 17 - \sqrt{2} & -1 \\ 0 & 0 & 17 - \pi \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

So,  $z = y = 0$  and  $x = t$ , for any  $t \in \mathbb{R}$ . So,

$$E(17) = \left\{ \begin{pmatrix} t \\ 0 \\ 0 \end{pmatrix} : t \in \mathbb{R} \right\}$$

# Continued

with  $t = 1$  a basis of  $E(17)$  is  $\left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \right\}$

## Continued

To compute  $E(\sqrt{2})$ , we solve:  $(\sqrt{2}I_3 - A)\mathbf{x} = \mathbf{0}$ , which is

$$\begin{pmatrix} \sqrt{2} - 17 & -113 & 2 \\ 0 & 0 & -1 \\ 0 & 0 & \sqrt{2} - \pi \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

So,  $z = 0$  and  $x = t$  and  $y = \frac{\sqrt{2}-17}{113}t$  for any  $t \in \mathbb{R}$ . So,

$$E(\sqrt{2}) = \left\{ \begin{pmatrix} t \\ \frac{\sqrt{2}-17}{113}t \\ 0 \end{pmatrix} : t \in \mathbb{R} \right\}$$

# Continued

with  $t = 113$  a basis of  $E(\sqrt{2})$  is  $\left\{ \begin{pmatrix} 113 \\ \sqrt{2} - 17 \\ 0 \end{pmatrix} \right\}$

# Continued

To compute  $E(\pi)$ , we solve:  $(\pi I_3 - A)\mathbf{x} = \mathbf{0}$ , which is

$$\begin{pmatrix} \pi - 17 & -113 & 2 \\ 0 & \pi - \sqrt{2} & -1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\begin{cases} z = t \\ y = \frac{1}{\pi - \sqrt{2}} z = \frac{1}{\pi - \sqrt{2}} t \\ x = \frac{113}{\pi - 17} y - \frac{2}{\pi - 17} z = \frac{113 + 2\sqrt{2} - 2\pi}{(\pi - 17)(\pi - \sqrt{2})} t \end{cases}$$

$$E(\pi) = \left\{ \begin{pmatrix} \frac{113 + 2\sqrt{2} - 2\pi}{(\pi - 17)(\pi - \sqrt{2})} t \\ \frac{1}{\pi - \sqrt{2}} t \\ t \end{pmatrix} : t \in \mathbb{R} \right\}$$

# Continued

With  $t = (\pi - 17)(\pi - \sqrt{2})$  a basis of  $E(\pi)$  is

$$\left\{ \begin{pmatrix} 113 + 2\sqrt{2} - 2\pi \\ \pi - 17 \\ (\pi - 17)(\pi - \sqrt{2}) \end{pmatrix} \right\}$$

## Continued

We form the matrix of the eigenvectors.

$$P = \begin{pmatrix} 1 & 113 & 113 + 2\sqrt{2} - 2\pi \\ 0 & \sqrt{2} - 17 & \pi - 17 \\ 0 & 0 & (\pi - 17)(\pi - \sqrt{2}) \end{pmatrix}.$$

We check

$$P^{-1}AP = \begin{pmatrix} 17 & 0 & 0 \\ 0 & \sqrt{2} & 0 \\ 0 & 0 & \pi \end{pmatrix} \quad \text{Or} \quad AP = P \begin{pmatrix} 17 & 0 & 0 \\ 0 & \sqrt{2} & 0 \\ 0 & 0 & \pi \end{pmatrix}$$



## Continued

We have

$$\begin{aligned}
 AP &= \begin{pmatrix} 17 & 113 & -2 \\ 0 & \sqrt{2} & 1 \\ 0 & 0 & \pi \end{pmatrix} \begin{pmatrix} 1 & 113 & 113 + 2\sqrt{2} - 2\pi \\ 0 & \sqrt{2} - 17 & \pi - 17 \\ 0 & 0 & (\pi - 17)(\pi - \sqrt{2}) \end{pmatrix} \\
 &= \begin{pmatrix} 17 & 113\sqrt{2} & \pi(113 + 2\sqrt{2} - 2\pi) \\ 0 & \sqrt{2}(\sqrt{2} - 17) & \pi(\pi - 17) \\ 0 & 0 & \pi(\pi - 17)(\pi - \sqrt{2}) \end{pmatrix} \\
 &= P \begin{pmatrix} 17 & 0 & 0 \\ 0 & \sqrt{2} & 0 \\ 0 & 0 & \pi \end{pmatrix}
 \end{aligned}$$