

Eigenvalues and Eigenvectors

- Eigenvalues and Eigenvectors
- Diagonalization

Eigenvalues and Eigenvectors

- Eigenvalue problem (one of the most important problems in the linear algebra):

If A is an $n \times n$ matrix, do there exist nonzero vectors \mathbf{x} in R^n such that $A\mathbf{x}$ is a scalar multiple of \mathbf{x} ?

(The term eigenvalue is from the German word *Eigenwert*, meaning “proper value”)

- Eigenvalue and Eigenvector:

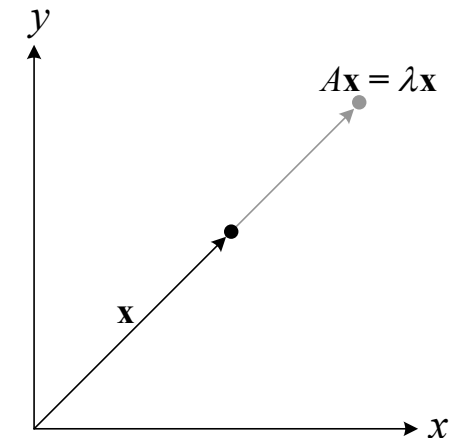
A : an $n \times n$ matrix

λ : a scalar (could be **zero**)

\mathbf{x} : a **nonzero** vector in R^n

$$\begin{array}{c} \text{Eigenvalue} \\ \downarrow \\ A\mathbf{x} = \lambda\mathbf{x} \\ \uparrow \quad \uparrow \\ \text{Eigenvector} \end{array}$$

※ Geometric Interpretation



Example: Verifying eigenvalues and eigenvectors

$$A = \begin{bmatrix} 2 & 0 \\ 0 & -1 \end{bmatrix} \quad \mathbf{x}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \mathbf{x}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$A\mathbf{x}_1 = \begin{bmatrix} 2 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 0 \end{bmatrix} = 2\mathbf{x}_1$$

Eigenvalue
↓
Eigenvalue
↑
Eigenvalue
↑
Eigenvalue

※ In fact, for each eigenvalue, it has infinitely many eigenvectors. For $\lambda = 2$, $[3 \ 0]^T$ or $[5 \ 0]^T$ are both corresponding eigenvectors. Moreover, $([3 \ 0] + [5 \ 0])^T$ is still an eigenvector.

$$A\mathbf{x}_2 = \begin{bmatrix} 2 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \end{bmatrix} = -1 \begin{bmatrix} 0 \\ 1 \end{bmatrix} = (-1)\mathbf{x}_2$$

Eigenvalue
↓
Eigenvalue
↑
Eigenvalue
↑
Eigenvalue

Theorem: The eigenspace corresponding to λ of matrix A

If A is an $n \times n$ matrix with an eigenvalue λ , then the set of all eigenvectors of λ **together with the zero vector** is a subspace of R^n . This subspace is called the eigenspace of λ .

Proof:

(1) Note 0 is in the set. Let \mathbf{x}_1 and \mathbf{x}_2 are eigenvectors corresponding to λ

(i.e., $A\mathbf{x}_1 = \lambda\mathbf{x}_1$, $A\mathbf{x}_2 = \lambda\mathbf{x}_2$)

(2) $A(\mathbf{x}_1 + \mathbf{x}_2) = A\mathbf{x}_1 + A\mathbf{x}_2 = \lambda\mathbf{x}_1 + \lambda\mathbf{x}_2 = \lambda(\mathbf{x}_1 + \mathbf{x}_2)$

(i.e., $\mathbf{x}_1 + \mathbf{x}_2$ is also an eigenvector corresponding to λ)

(3) $A(c\mathbf{x}_1) = c(A\mathbf{x}_1) = c(\lambda\mathbf{x}_1) = \lambda(c\mathbf{x}_1)$

(i.e., $c\mathbf{x}_1$ is also an eigenvector corresponding to λ)

Since this set is closed under vector addition and scalar multiplication, this set is a subspace of R^n

Example: Examples of eigenspaces on the xy -plane

For the matrix A as follows, the corresponding eigenvalues are $\lambda_1 = -1$ and $\lambda_2 = 1$:

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

Solution:

For the eigenvalue $\lambda_1 = -1$, corresponding vectors are any vectors on the x -axis

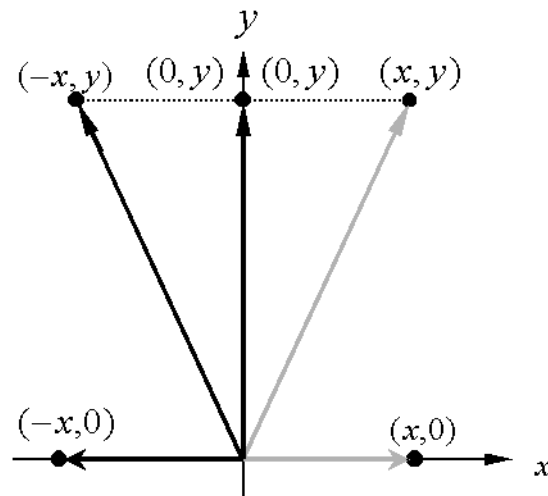
$$A \begin{bmatrix} x \\ 0 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ 0 \end{bmatrix} = \begin{bmatrix} -x \\ 0 \end{bmatrix} = \textcircled{-1} \begin{bmatrix} x \\ 0 \end{bmatrix} \quad \text{※ Thus, the eigenspace corresponding to } \lambda = -1 \text{ is the } x\text{-axis, which is a subspace of } \mathbb{R}^2$$

For the eigenvalue $\lambda_2 = 1$, corresponding vectors are any vectors on the y -axis

$$A \begin{bmatrix} 0 \\ y \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ y \end{bmatrix} = \textcircled{1} \begin{bmatrix} 0 \\ y \end{bmatrix} \quad \text{※ Thus, the eigenspace corresponding to } \lambda = 1 \text{ is the } y\text{-axis, which is a subspace of } \mathbb{R}^2$$

※ Geometrically speaking, multiplying a vector (x, y) in R^2 by the matrix A corresponds to a reflection to the y -axis, i.e., left multiplying A to \mathbf{v} can transform \mathbf{v} to another vector in the same vector space

$$\begin{aligned} A\mathbf{v} &= A\begin{bmatrix} x \\ y \end{bmatrix} = A\left(\begin{bmatrix} x \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ y \end{bmatrix}\right) = A\begin{bmatrix} x \\ 0 \end{bmatrix} + A\begin{bmatrix} 0 \\ y \end{bmatrix} \\ &= -1\begin{bmatrix} x \\ 0 \end{bmatrix} + 1\begin{bmatrix} 0 \\ y \end{bmatrix} = \begin{bmatrix} -x \\ y \end{bmatrix} \end{aligned}$$



Theorem: Finding eigenvalues and eigenvectors of a matrix $A \in M_{n \times n}$

Let A be an $n \times n$ matrix.

(1) An eigenvalue of A is a scalar λ such that $\det(\lambda I - A) = 0$

(2) The eigenvectors of A corresponding to λ are the nonzero solutions of $(\lambda I - A)\mathbf{x} = \mathbf{0}$

Note: following the definition of the eigenvalue problem

$$A\mathbf{x} = \lambda\mathbf{x} \Rightarrow A\mathbf{x} = \lambda I\mathbf{x} \Rightarrow (\lambda I - A)\mathbf{x} = \mathbf{0} \text{ (homogeneous system)}$$

$$(\lambda I - A)\mathbf{x} = \mathbf{0} \text{ has nonzero solutions for } \mathbf{x} \text{ iff } \det(\lambda I - A) = 0$$

- **Characteristic equation of A :**

$$\det(\lambda I - A) = 0$$

- **Characteristic polynomial of $A \in M_{n \times n}$:**

$$\det(\lambda I - A) = |(\lambda I - A)| = \lambda^n + c_{n-1}\lambda^{n-1} + \cdots + c_1\lambda + c_0$$

Theorem: Eigenvalues and Invertibility

A square matrix A is invertible if and only if $\lambda = 0$ is not an eigenvalue.

Theorem: Eigenvalues of Powers of a Matrix

If k is a positive integer and λ is an eigenvalue of a matrix A , and \mathbf{x} is a corresponding eigenvector, then λ^k is an eigenvalue of A and \mathbf{x} is a corresponding eigenvector.

Proof: For $k = 2$,

suppose that λ is an eigenvalue of A and \mathbf{x} is a corresponding eigenvector.

$$A^2\mathbf{x} = A(A\mathbf{x}) = A(\lambda\mathbf{x}) = \lambda(A\mathbf{x}) = \lambda(\lambda\mathbf{x}) = \lambda^2\mathbf{x}$$

Example: Finding eigenvalues and eigenvectors

$$A = \begin{bmatrix} 2 & -12 \\ 1 & -5 \end{bmatrix}$$

Solution: Characteristic equation:

$$\begin{aligned} \det(\lambda I - A) &= \begin{vmatrix} \lambda - 2 & 12 \\ -1 & \lambda + 5 \end{vmatrix} \\ &= \lambda^2 + 3\lambda + 2 = (\lambda + 1)(\lambda + 2) = 0 \end{aligned}$$

$$\Rightarrow \lambda = -1, -2$$

Eigenvalue: $\lambda_1 = -1, \lambda_2 = -2$

$$(1) \lambda_1 = -1 \Rightarrow (\lambda_1 I - A)\mathbf{x} = \begin{bmatrix} -3 & 12 \\ -1 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} -3 & 12 \\ -1 & 4 \end{bmatrix} \xrightarrow{\text{G.-J. E.}} \begin{bmatrix} 1 & -4 \\ 0 & 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 4t \\ t \end{bmatrix} = t \begin{bmatrix} 4 \\ 1 \end{bmatrix}, \quad t \neq 0$$

$$(2) \lambda_2 = -2 \Rightarrow (\lambda_2 I - A)\mathbf{x} = \begin{bmatrix} -4 & 12 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} -4 & 12 \\ -1 & 3 \end{bmatrix} \xrightarrow{\text{G.-J. E.}} \begin{bmatrix} 1 & -3 \\ 0 & 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3s \\ s \end{bmatrix} = s \begin{bmatrix} 3 \\ 1 \end{bmatrix}, \quad s \neq 0$$

Example: Finding eigenvalues and eigenvectors

Find the eigenvalues and corresponding eigenvectors for the matrix A . What is the dimension of the eigenspace of each eigenvalue?

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

Solution: Characteristic equation:

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

Eigenvalue: $\lambda = 2$

The eigenspace of $\lambda = 2$:

$$(\lambda I - A)\mathbf{x} = \begin{bmatrix} 0 & -1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} s \\ 0 \\ t \end{bmatrix} = s \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \quad s, t \neq 0$$

$$\left\{ s \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \middle| s, t \in R \right\} : \text{the eigenspace of } A \text{ corresponding to } \lambda = 2$$

Thus, the dimension of its eigenspace is 2

Notes:

- (1) If an eigenvalue λ_1 occurs as a multiple root (k times) for the characteristic polynomial, then λ_1 has multiplicity k
- (2) The multiplicity of an eigenvalue is greater than or equal to the dimension of its eigenspace. (In Ex, k is 3 and the dimension of its eigenspace is 2)

Example: Finding Eigenvalues and Eigenvectors

Find the eigenvalues of the matrix A and find a basis for each of the corresponding eigenspaces

$$A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 5 & -10 \\ 1 & 0 & 2 & 0 \\ 1 & 0 & 0 & 3 \end{bmatrix}$$

Solution: Characteristic equation:

$$\begin{aligned} |\lambda I - A| &= \begin{vmatrix} \lambda - 1 & 0 & 0 & 0 \\ 0 & \lambda - 1 & -5 & 10 \\ -1 & 0 & \lambda - 2 & 0 \\ -1 & 0 & 0 & \lambda - 3 \end{vmatrix} \\ &= (\lambda - 1)^2 (\lambda - 2) (\lambda - 3) = 0 \end{aligned}$$

Eigenvalues: $\lambda_1 = 1, \lambda_2 = 2, \lambda_3 = 3$

※ According to the previous slide, the dimension of the eigenspace of $\lambda_1 = 1$ is at most to be 2

※ For $\lambda_2 = 2$ and $\lambda_3 = 3$, the dimensions of their eigenspaces are at most to be 1

$$(1) \lambda_1 = 1 \Rightarrow (\lambda_1 I - A)\mathbf{x} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & -5 & 10 \\ -1 & 0 & -1 & 0 \\ -1 & 0 & 0 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\stackrel{\text{G.-J.E.}}{\Rightarrow} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -2t \\ s \\ 2t \\ t \end{bmatrix} = s \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} -2 \\ 0 \\ 2 \\ 1 \end{bmatrix}, \quad s, t \neq 0$$

$$\Rightarrow \left\{ \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -2 \\ 0 \\ 2 \\ 1 \end{bmatrix} \right\} \text{ is a basis for the eigenspace corresponding to } \lambda_1 = 1$$

※ The dimension of the eigenspace of $\lambda_1 = 1$ is 2

$$(2) \lambda_2 = 2 \Rightarrow (\lambda_2 I - A)\mathbf{x} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & -5 & 10 \\ -1 & 0 & 0 & 0 \\ -1 & 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{array}{l} \text{G.-J.E.} \\ \Rightarrow \end{array} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 5t \\ t \\ 0 \end{bmatrix} = t \begin{bmatrix} 0 \\ 5 \\ 1 \\ 0 \end{bmatrix}, \quad t \neq 0$$

$$\Rightarrow \left\{ \begin{bmatrix} 0 \\ 5 \\ 1 \\ 0 \end{bmatrix} \right\} \text{ is a basis for the eigenspace} \\ \text{corresponding to } \lambda_2 = 2$$

※ The dimension of the eigenspace of $\lambda_2 = 2$ is 1

$$(3) \lambda_3 = 3 \Rightarrow (\lambda_3 I - A)\mathbf{x} = \begin{bmatrix} 2 & 0 & 0 & 0 \\ 0 & 2 & -5 & 10 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\stackrel{\text{G.-J.E.}}{\Rightarrow} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ -5t \\ 0 \\ t \end{bmatrix} = t \begin{bmatrix} 0 \\ -5 \\ 0 \\ 1 \end{bmatrix}, \quad t \neq 0$$

$$\Rightarrow \left\{ \begin{bmatrix} 0 \\ -5 \\ 0 \\ 1 \end{bmatrix} \right\} \text{ is a basis for the eigenspace} \\ \text{corresponding to } \lambda_3 = 3$$

※ The dimension of the eigenspace of $\lambda_3 = 3$ is 1

Theorem: Eigenvalues for triangular matrices

If A is an $n \times n$ triangular matrix, then its eigenvalues are the entries on its main diagonal

Example: Finding eigenvalues for triangular and diagonal matrices

$$(a) A = \begin{bmatrix} 2 & 0 & 0 \\ -1 & 1 & 0 \\ 5 & 3 & -3 \end{bmatrix} \quad (b) A = \begin{bmatrix} -1 & 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & -4 & 0 \\ 0 & 0 & 0 & 0 & 3 \end{bmatrix}$$

Solution:

$$(a) |\lambda I - A| = \begin{vmatrix} \lambda - 2 & 0 & 0 \\ 1 & \lambda - 1 & 0 \\ -5 & -3 & \lambda + 3 \end{vmatrix} = (\lambda - 2)(\lambda - 1)(\lambda + 3) = 0$$

$$\Rightarrow \lambda_1 = 2, \lambda_2 = 1, \lambda_3 = -3$$

$$(b) \lambda_1 = -1, \lambda_2 = 2, \lambda_3 = 0, \lambda_4 = -4, \lambda_5 = 3$$

※ According to Thm., the determinant of a triangular matrix is the product of the entries on the main diagonal

Example: Finding eigenvalues and eigenvectors for standard matrices

Find the eigenvalues and corresponding eigenvectors for

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

Solution:

$$|\lambda I - A| = \begin{vmatrix} \lambda - 1 & -3 & 0 \\ -3 & \lambda - 1 & 0 \\ 0 & 0 & \lambda + 2 \end{vmatrix} = (\lambda + 2)^2 (\lambda - 4) = 0$$

\Rightarrow eigenvalues $\lambda_1 = 4$, $\lambda_2 = -2$

For $\lambda_1 = 4$, the corresponding eigenvector is $(1, 1, 0)$.

For $\lambda_2 = -2$, the corresponding eigenvectors are $(1, -1, 0)$
and $(0, 0, 1)$.

Diagonalization

- **Diagonalization problem:**

For a square matrix A , does there exist an invertible matrix P such that $P^{-1}AP$ is diagonal?

- **Diagonalizable matrix:**

Definition 1: A square matrix A is called **diagonalizable** if there exists an invertible matrix P such that $P^{-1}AP$ is a diagonal matrix (i.e., P diagonalizes A)

Definition 2: A square matrix A is called **diagonalizable** if A is **similar** to a diagonal matrix

※ Remember two square matrices A and B are **similar** if there exists an invertible matrix P such that $B = P^{-1}AP$.

Notes:

This section shows that the eigenvalue and eigenvector problem is closely related to the diagonalization problem

Theorem: **Similar matrices have the same eigenvalues**

If A and B are similar $n \times n$ matrices, then they have the same eigenvalues

Proof:

A and B are similar $\Rightarrow B = P^{-1}AP$

For any diagonal matrix in the form of $D = \lambda I$, $P^{-1}DP = D$

Consider the characteristic equation of B :

$$\begin{aligned} |\lambda I - B| &= |\lambda I - P^{-1}AP| \stackrel{\text{blue arrow}}{=} |P^{-1}\lambda IP - P^{-1}AP| = |P^{-1}(\lambda I - A)P| \\ &= |P^{-1}| |\lambda I - A| |P| = |P^{-1}| |P| |\lambda I - A| = |P^{-1}P| |\lambda I - A| \\ &= |\lambda I - A| \end{aligned}$$

Since A and B have the same characteristic equation, they are with the same eigenvalues

※ Note that the eigenvectors of A and B are not necessarily identical

Example: Eigenvalue problems and diagonalization programs

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

Solution: Characteristic equation:

$$|\lambda I - A| = \begin{vmatrix} \lambda - 1 & -3 & 0 \\ -3 & \lambda - 1 & 0 \\ 0 & 0 & \lambda + 2 \end{vmatrix} = (\lambda - 4)(\lambda + 2)^2 = 0$$

The eigenvalues : $\lambda_1 = 4$, $\lambda_2 = -2$, $\lambda_3 = -2$

$$(1) \lambda = 4 \Rightarrow \text{the eigenvector } \mathbf{p}_1 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$$

$$(2) \lambda = -2 \Rightarrow \text{the eigenvector } \mathbf{p}_2 = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \quad \mathbf{p}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$P = [\mathbf{p}_1 \quad \mathbf{p}_2 \quad \mathbf{p}_3] = \begin{bmatrix} 1 & 1 & 0 \\ 1 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \text{ and } P^{-1}AP = \begin{bmatrix} 4 & 0 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & -2 \end{bmatrix}$$

▪ **Note:** If $P = [\mathbf{p}_2 \quad \mathbf{p}_1 \quad \mathbf{p}_3]$

$$= \begin{bmatrix} 1 & 1 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow P^{-1}AP = \begin{bmatrix} -2 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & -2 \end{bmatrix}$$

※ The above example can verify Thm. since the eigenvalues for both A and $P^{-1}AP$ are the same to be 4, -2, and -2

※ The reason why the matrix P is constructed with the eigenvectors of A is demonstrated in next Thm. on the next slide

Theorem: **Condition for diagonalization**

An $n \times n$ matrix A is diagonalizable if and only if it has n linearly independent eigenvectors.

- ※ If there are n linearly independent eigenvectors, it does not imply that there are n distinct eigenvalues. It is possible to have only one eigenvalue with the multiplicity n , and there are n linearly independent eigenvectors
- ※ On the other hand, if there are n distinct eigenvalues, then there are n linearly independent eigenvectors, and thus A must be diagonalizable

Proof: (\Rightarrow)

Since A is diagonalizable, there exists an invertible P s.t. $D = P^{-1}AP$ is diagonal. Let $P = [\mathbf{p}_1 \ \mathbf{p}_2 \ \cdots \ \mathbf{p}_n]$ and $D = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n)$, then

$$\begin{aligned} PD &= [\mathbf{p}_1 \ \mathbf{p}_2 \ \cdots \ \mathbf{p}_n] \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix} \\ &= [\lambda_1 \mathbf{p}_1 \ \lambda_2 \mathbf{p}_2 \ \cdots \ \lambda_n \mathbf{p}_n] \end{aligned}$$

$$\because AP = PD \text{ (since } D = P^{-1}AP \text{)}$$

$$\therefore [A\mathbf{p}_1 \ A\mathbf{p}_2 \ \cdots \ A\mathbf{p}_n] = [\lambda_1\mathbf{p}_1 \ \lambda_2\mathbf{p}_2 \ \cdots \ \lambda_n\mathbf{p}_n]$$

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

(The above equations imply the column vectors \mathbf{p}_i of P are eigenvectors of A , and the diagonal entries λ_i in D are eigenvalues of A)

Because A is diagonalizable $\Rightarrow P$ is invertible

\Rightarrow Columns in P , i.e., $\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_n$, are linearly independent
(see Slide 4.101 in the lecture note)

Thus, A has n linearly independent eigenvectors

(\Leftarrow)

Since A has n linearly independent eigenvectors $\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_n$ with corresponding eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ (could be the same), then

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

$$\text{Let } P = [\mathbf{p}_1 \ \mathbf{p}_2 \ \cdots \ \mathbf{p}_n]$$

$$\begin{aligned}
 AP &= A[\mathbf{p}_1 \ \mathbf{p}_2 \ \cdots \ \mathbf{p}_n] = [A\mathbf{p}_1 \ A\mathbf{p}_2 \ \cdots \ A\mathbf{p}_n] \\
 &= [\lambda_1\mathbf{p}_1 \ \lambda_2\mathbf{p}_2 \ \cdots \ \lambda_n\mathbf{p}_n] \\
 &= [\mathbf{p}_1 \ \mathbf{p}_2 \ \cdots \ \mathbf{p}_n] \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix} = PD
 \end{aligned}$$

Since $\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_n$ are linearly independent

$\Rightarrow P$ is invertible

$$\because AP = PD \therefore P^{-1}AP = D$$

$\Rightarrow A$ is diagonalizable

(according to the definition of the diagonalizable matrix)

✱ Note that \mathbf{p}_i 's are linearly independent eigenvectors and the diagonal entries λ_i in the resulting diagonalized D are eigenvalues of A

Example: A matrix that is not diagonalizable

Show that the following matrix is not diagonalizable

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

Solution: Characteristic equation:

$$|\lambda I - A| = \begin{vmatrix} \lambda - 1 & -2 \\ 0 & \lambda - 1 \end{vmatrix} = (\lambda - 1)^2 = 0$$

The eigenvalue $\lambda_1 = 1$, and then solve $(\lambda_1 I - A)\mathbf{x} = \mathbf{0}$ for eigenvectors

$$\lambda_1 I - A = I - A = \begin{bmatrix} 0 & -2 \\ 0 & 0 \end{bmatrix} \Rightarrow \text{eigenvector } \mathbf{p}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

Since A does not have two linearly independent eigenvectors,
 A is not diagonalizable

- Steps for diagonalizing an $n \times n$ square matrix:

Step 1: Find n linearly independent eigenvectors $\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_n$
for A with corresponding eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$

Step 2: Let $P = [\mathbf{p}_1 \ \mathbf{p}_2 \ \dots \ \mathbf{p}_n]$

Step 3:

$$P^{-1}AP = D = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

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

Example: **Diagonalizing a matrix**

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

Find a matrix P such that $P^{-1}AP$ is diagonal.

Solution: Characteristic equation:

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

The eigenvalues : $\lambda_1 = 2$, $\lambda_2 = -2$, $\lambda_3 = 3$

$$\lambda_1 = 2 \Rightarrow \lambda_1 I - A = \begin{bmatrix} 1 & 1 & 1 \\ -1 & -1 & -1 \\ 3 & -1 & 3 \end{bmatrix} \xrightarrow{\text{G.-J. E.}} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -t \\ 0 \\ t \end{bmatrix} \Rightarrow \text{eigenvector } \mathbf{p}_1 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

$$\lambda_2 = -2 \Rightarrow \lambda_2 I - A = \begin{bmatrix} -3 & 1 & 1 \\ -1 & -5 & -1 \\ 3 & -1 & -1 \end{bmatrix} \xrightarrow{\text{G.-J. E.}} \begin{bmatrix} 1 & 0 & -\frac{1}{4} \\ 0 & 1 & \frac{1}{4} \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \frac{1}{4}t \\ -\frac{1}{4}t \\ t \end{bmatrix} \Rightarrow \text{eigenvector } \mathbf{p}_2 = \begin{bmatrix} 1 \\ -1 \\ 4 \end{bmatrix}$$

$$\lambda_3 = 3 \Rightarrow \lambda_3 I - A = \begin{bmatrix} 2 & 1 & 1 \\ -1 & 0 & -1 \\ 3 & -1 & 4 \end{bmatrix} \xrightarrow{\text{G.-J. E.}} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -t \\ t \\ t \end{bmatrix} \Rightarrow \text{eigenvector } \mathbf{p}_3 = \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix}$$

$$P = [\mathbf{p}_1 \quad \mathbf{p}_2 \quad \mathbf{p}_3] = \begin{bmatrix} -1 & 1 & -1 \\ 0 & -1 & 1 \\ 1 & 4 & 1 \end{bmatrix} \text{ and it follows that}$$

$$P^{-1}AP = \begin{bmatrix} 2 & 0 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$$

Note: a quick way to calculate A^k based on the diagonalization technique

$$(1) D = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix} \Rightarrow D^k = \begin{bmatrix} \lambda_1^k & 0 & \cdots & 0 \\ 0 & \lambda_2^k & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n^k \end{bmatrix}$$

$$(2) D = P^{-1}AP \Rightarrow D^k = \underbrace{P^{-1}AP}_{\text{repeat } k \text{ times}} \underbrace{P^{-1}AP} \cdots \underbrace{P^{-1}AP} = P^{-1}A^kP$$

$$A^k = PD^kP^{-1}, \text{ where } D^k = \begin{bmatrix} \lambda_1^k & 0 & \cdots & 0 \\ 0 & \lambda_2^k & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n^k \end{bmatrix}$$

Theorem: **Sufficient conditions for diagonalization**

If an $n \times n$ matrix A has n distinct eigenvalues, then the corresponding eigenvectors are linearly independent and thus A is diagonalizable according to last Thm.

Proof:

Let $\lambda_1, \lambda_2, \dots, \lambda_n$ be distinct eigenvalues and corresponding eigenvectors be $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$. In addition, consider that the first m eigenvectors are linearly independent, but the first $m+1$ eigenvectors are linearly dependent, i.e.,

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

where c_i 's are not all zero. Multiplying both sides of Eq. (1) by A yields

$$\begin{aligned} A\mathbf{x}_{m+1} &= Ac_1 \mathbf{x}_1 + Ac_2 \mathbf{x}_2 + \dots + Ac_m \mathbf{x}_m \\ \lambda_{m+1} \mathbf{x}_{m+1} &= c_1 \lambda_1 \mathbf{x}_1 + c_2 \lambda_2 \mathbf{x}_2 + \dots + c_m \lambda_m \mathbf{x}_m \end{aligned} \quad (2)$$

On the other hand, multiplying both sides of Eq. (1) by λ_{m+1} yields

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

Now, subtracting Eq. (2) from Eq. (3) produces

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

Since the first m eigenvectors are linearly independent, we can infer that all coefficients of this equation should be zero, i.e.,

$$c_1 (\lambda_{m+1} - \lambda_1) = c_2 (\lambda_{m+1} - \lambda_2) = \cdots = c_m (\lambda_{m+1} - \lambda_m) = 0$$

Because all the eigenvalues are distinct, it follows all c_i 's equal to 0, which contradicts our assumption that \mathbf{x}_{m+1} can be expressed as a linear combination of the first m eigenvectors. So, the set of n eigenvectors is linearly independent given n distinct eigenvalues, and according to previous Thm., we can conclude that A is diagonalizable

Example: **Determining whether a matrix is diagonalizable**

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

Solution: Because A is a triangular matrix, its eigenvalues are
 $\lambda_1 = 1, \lambda_2 = 0, \lambda_3 = -3$

According to Thm., because these three values are distinct, A is diagonalizable

For an eigenvalue λ_0 : let E_{λ_0} be the eigenspace corresponding to λ_0

Geometric multiplicity of $\lambda_0 := \dim(E_{\lambda_0})$

Algebraic multiplicity of $\lambda_0 :=$ number of times $\lambda - \lambda_0$ appears in the characteristic polynomial.

THEOREM Geometric and Algebraic Multiplicity

If A is a square matrix, then:

- (a) For every eigenvalue of A , the geometric multiplicity is less than or equal to the algebraic multiplicity.*
- (b) A is diagonalizable if and only if the geometric multiplicity of every eigenvalue is equal to the algebraic multiplicity.*

Example 3: Solve the eigenvalue problem $A\mathbf{x} = \lambda\mathbf{x}$ and find the eigenspace, algebraic multiplicity, and geometric multiplicity for each eigenvalue.

$$A = \begin{bmatrix} -4 & -3 & 6 \\ 0 & -1 & 0 \\ -3 & -3 & 5 \end{bmatrix}$$

Step 1: Write down the characteristic equation of A and solve for its eigenvalues.

$$p(\lambda) = |\lambda I - A| = \begin{vmatrix} \lambda + 4 & 3 & -6 \\ 0 & \lambda + 1 & 0 \\ 3 & 3 & \lambda - 5 \end{vmatrix} = (-1)^4 (\lambda + 1) \begin{vmatrix} \lambda + 4 & -6 \\ 3 & \lambda - 5 \end{vmatrix}$$

$$\begin{aligned} p(\lambda) &= (\lambda + 1)((\lambda + 4)(\lambda - 5) + 18) = 0 \\ &= (\lambda + 1)(\lambda^2 - \lambda - 2) = -(\lambda + 1)(\lambda - 2)(\lambda + 1) = 0 \\ &= (\lambda - 2)(\lambda + 1)^2 = 0. \end{aligned}$$

So the eigenvalues are $\lambda_1 = 2, \lambda_2 = -1$.

Since the factor $(\lambda - 2)$ is first power, $\lambda_1 = 2$ is not a repeated root. $\lambda_1 = 2$ has an algebraic multiplicity of 1. On the other hand, the factor $(\lambda + 1)$ is squared, $\lambda_2 = -1$ is a repeated root, and it has an algebraic multiplicity of 2.

Step 2: Use Gaussian elimination with back-substitution to solve $(\lambda I - A) \mathbf{x} = \mathbf{0}$ for λ_1 and λ_2 .

For $\lambda_1 = 2$, the augmented matrix for the system is

$$[2I - A | \vec{0}] = \begin{bmatrix} 6 & 3 & -6 & 0 \\ 0 & 3 & 0 & 0 \\ 3 & 3 & -3 & 0 \end{bmatrix} \sim \begin{matrix} \frac{1}{6}r1 \rightarrow r1 \\ \frac{1}{3}r2 \rightarrow r2 \\ r3 \end{matrix} \begin{bmatrix} 1 & 1/2 & -1 & 0 \\ 0 & 1 & 0 & 0 \\ 3 & 3 & -3 & 0 \end{bmatrix}$$

$$\sim \begin{matrix} r1 \\ r2 \\ -3r1 + r3 \rightarrow r3 \end{matrix} \begin{bmatrix} 1 & 1/2 & -1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 3/2 & 0 & 0 \end{bmatrix}$$

$$\sim \begin{matrix} r1 \\ r2 \\ -\frac{3}{2}r2 + r3 \rightarrow r3 \end{matrix} \begin{bmatrix} 1 & 1/2 & -1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$

In this case,

$x_3 = r$, $x_2 = 0$, and

$x_1 = -1/2(0) + r$

$= 0 + r = r.$

Thus, the eigenvector corresponding to $\lambda_1 = 2$ is

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} r \\ 0 \\ r \end{bmatrix} = r \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, r \neq 0. \text{ If we choose } \vec{p}_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix},$$

then $B_1 = \left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \right\}$ is a basis for the eigenspace of $\lambda_1 = 2$.

$E_{\lambda_1} = \text{span}(\{\vec{p}_1\})$ and $\dim(E_{\lambda_1}) = 1$, so the geometric multiplicity is 1.

$$A\vec{x} = 2\vec{x} \text{ or } (2I - A)\vec{x} = \vec{0}.$$

$$\begin{bmatrix} -4 & -3 & 6 \\ 0 & -1 & 0 \\ -3 & -3 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} -4 + 6 \\ 0 \\ -3 + 5 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ 2 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}.$$

For $\lambda_2 = -1$, the augmented matrix for the system is

$$\begin{aligned} [(-1)I - A | \vec{0}] &= \begin{bmatrix} 3 & 3 & -6 & 0 \\ 0 & 0 & 0 & 0 \\ 3 & 3 & -6 & 0 \end{bmatrix} \sim \begin{matrix} \frac{1}{3}r1 \rightarrow r1 \\ r2 \\ r3 \end{matrix} \begin{bmatrix} 1 & 1 & -2 & 0 \\ 0 & 0 & 0 & 0 \\ 3 & 3 & -6 & 0 \end{bmatrix} \\ &\sim \begin{matrix} r1 \\ r2 \\ -3r1 + r3 \rightarrow r3 \end{matrix} \begin{bmatrix} 1 & 1 & -2 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \end{aligned}$$

$x_3 = t$, $x_2 = s$, and $x_1 = -s + 2t$. Thus, the solution has two linearly independent eigenvectors for $\lambda_2 = -1$ with

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -s + 2t \\ s \\ t \end{bmatrix} = s \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix}, s \neq 0, t \neq 0.$$

If we choose $\vec{p}_2 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$, and $\vec{p}_3 = \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix}$, then $B_2 = \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} \right\}$

is a basis for $E_{\lambda_2} = \text{span}(\{\vec{p}_2, \vec{p}_3\})$ and $\dim(E_{\lambda_2}) = 2$,

so the geometric multiplicity is 2.

Since the geometric multiplicity is equal to the algebraic multiplicity for each distinct eigenvalue, we found three linearly independent eigenvectors. The matrix A is diagonalizable since $P = [\mathbf{p}_1 \ \mathbf{p}_2 \ \mathbf{p}_3]$ is nonsingular.