

# Chapter 5

## Eigenvalues, Eigenvectors, and Diagonozation

(除了標註✖之簡報外，其餘採用李宏毅教授之投影片教材)

# Eigenvalues and eigenvectors

(Chapter 5.1)

# Eigenvalues and Eigenvectors

- If  $Av = \lambda v$  ( $v$  is a vector,  $\lambda$  is a scalar)  $A\mathbf{0} = \lambda\mathbf{0}$ 
  - $v$  is an **eigenvector** of  $A$  **excluding zero vector**
  - $\lambda$  is an **eigenvalue** of  $A$  that corresponds to  $v$

A must be square

$$\begin{bmatrix} 5 & 2 & 1 \\ -2 & 1 & -1 \\ 2 & 2 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 \\ -4 \\ 4 \end{bmatrix} = 4 \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

Eigen value

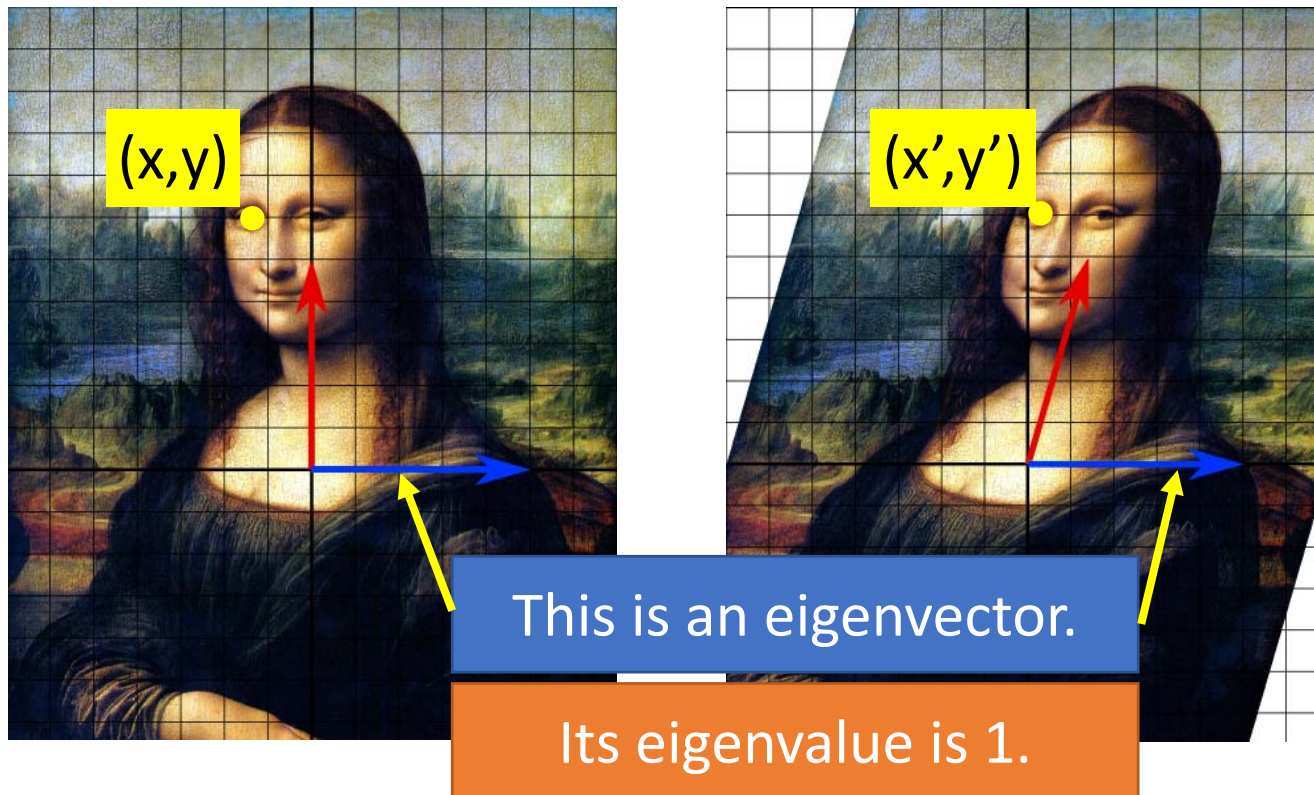
Eigen vector

# Eigenvalues and Eigenvectors

- If  $Av = \lambda v$  ( $v$  is a vector,  $\lambda$  is a scalar)
  - $v$  is an eigenvector of  $A$  **excluding zero vector**
  - $\lambda$  is an eigenvalue of  $A$  that corresponds to  $v$
- $T$  is a **linear operator**. If  $T(v) = \lambda v$  ( $v$  is a vector,  $\lambda$  is a scalar)
  - $v$  is an eigenvector of  $T$  **excluding zero vector**
  - $\lambda$  is an eigenvalue of  $T$  that corresponds to  $v$

# Eigenvalues and Eigenvectors

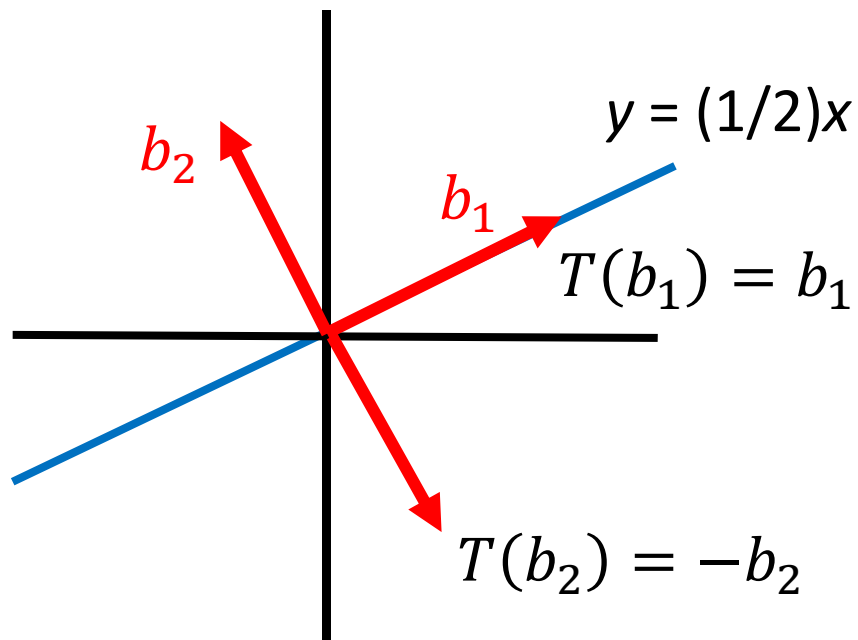
- Example: Shear Transform  $\begin{bmatrix} x' \\ y' \end{bmatrix} = T \left( \begin{bmatrix} x \\ y \end{bmatrix} \right)$



# Eigenvalues and Eigenvectors

- Example: Reflection

reflection operator  $T$  about the line  $y = (1/2)x$



$\mathbf{b}_1$  is an eigenvector of  $T$

Its eigenvalue is 1.

$\mathbf{b}_2$  is an eigenvector of  $T$

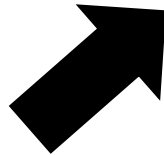
Its eigenvalue is -1.

# Eigenvalues and Eigenvectors

- Example:

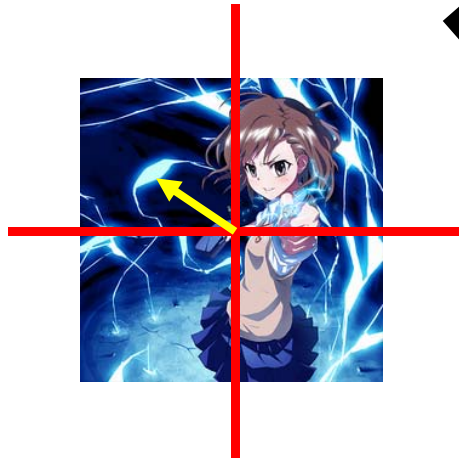
*Expansion and  
Compression*

$$\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

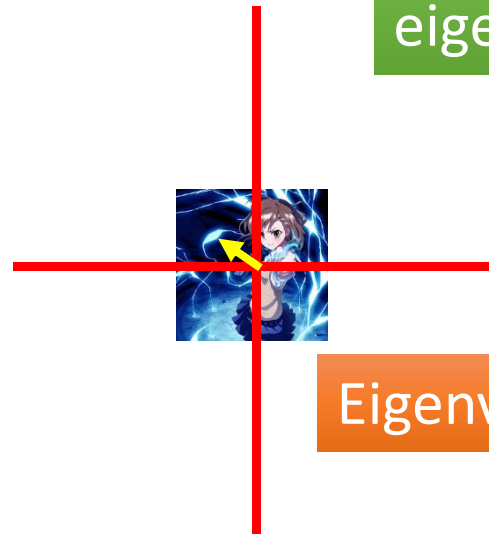
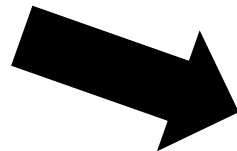


Eigenvalue is 2

All vectors are  
eigenvectors.



$$\begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix}$$



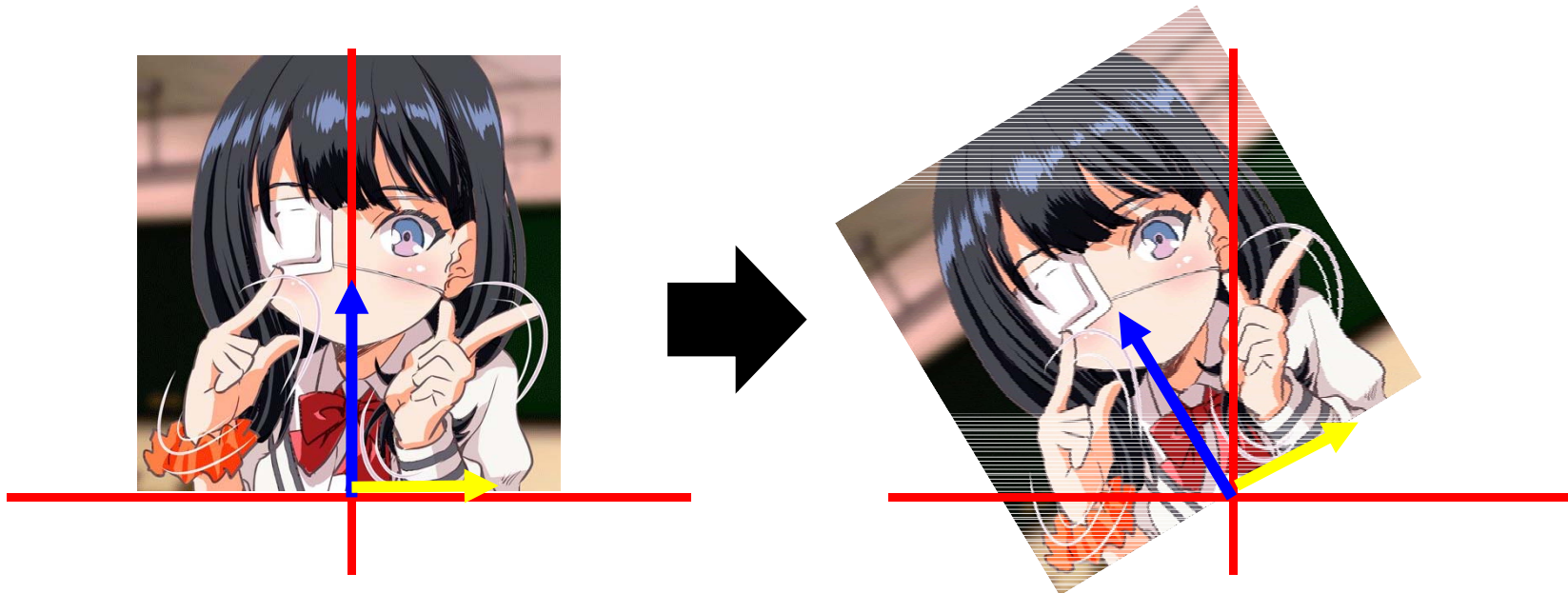
Eigenvalue is 0.5

# Eigenvalues and Eigenvectors

- Example: Rotation

Source of image:

<https://twitter.com/circleponiponi/status/1056026158083403776>



Do any  $n \times n$  matrix or linear operator have eigenvalues?



# How to find eigenvectors (given eigenvalues)

(Chapter 5.1)

# Eigenvalues and Eigenvectors

- An eigenvector of  $A$  corresponds to a unique eigenvalue.
- An eigenvalue of  $A$  has infinitely many eigenvectors.

Example:

$$A = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 1 \end{bmatrix} \quad v = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \quad u = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$$

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

Eigenvalue= -1

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

Eigenvalue= -1

Do the eigenvectors corresponding to the same eigenvalue form a subspace?



$$Av = \lambda v$$

$$Au = \lambda u$$

$$A(cv) = \lambda(cv)$$

$$A(u + v) = \lambda(u + v)$$

# Eigenspace

- Assume we know  $\lambda$  is the eigenvalue of matrix  $A$
- Eigenvectors corresponding to  $\lambda$

$$A\mathbf{v} = \lambda\mathbf{v}$$

$$A\mathbf{v} - \lambda\mathbf{v} = \mathbf{0}$$

$$A\mathbf{v} - \lambda I_n \mathbf{v} = \mathbf{0}$$

$$\underline{(A - \lambda I_n)}\mathbf{v} = \mathbf{0}$$

matrix

Eigenvectors corresponding to  $\lambda$   
are **nonzero** solution of

$$(A - \lambda I_n)\mathbf{v} = \mathbf{0}$$

Eigenvectors corresponding to  $\lambda$

$$= \underline{\text{Null}(A - \lambda I_n)} - \{\mathbf{0}\}$$

eigenspace

Eigenspace of  $\lambda$ :

Eigenvectors corresponding to  $\lambda + \{\mathbf{0}\}$

Check whether a scalar is  
an eigenvalue  
(Chapter 5.1)

# Check Eigenvalues

$Null(A - \lambda I_n)$ :  
eigenspace of  $\lambda$

- How to know whether a scalar  $\lambda$  is the eigenvalue of A?

Check the dimension of eigenspace of  $\lambda$

If the dimension is 0

➡ Eigenspace only contains  $\{0\}$

➡ No eigenvector

➡  $\lambda$  is not eigenvalue

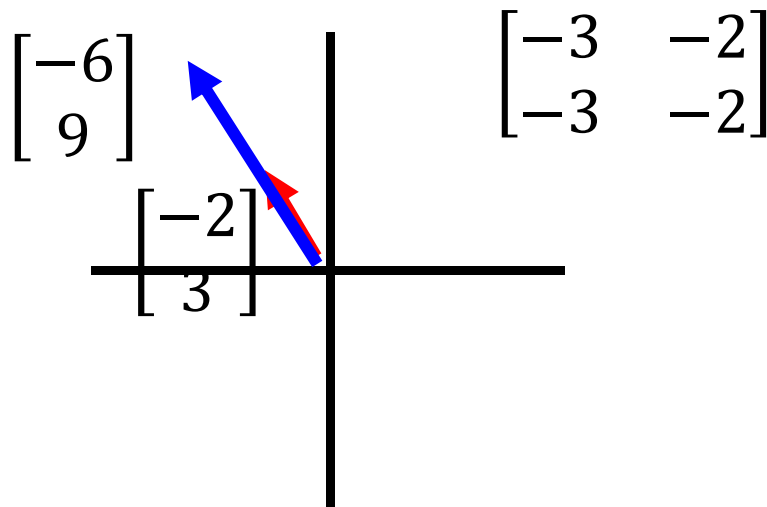
# Check Eigenvalues

$Null(A - \lambda I_n)$ :  
eigenspace of  $\lambda$

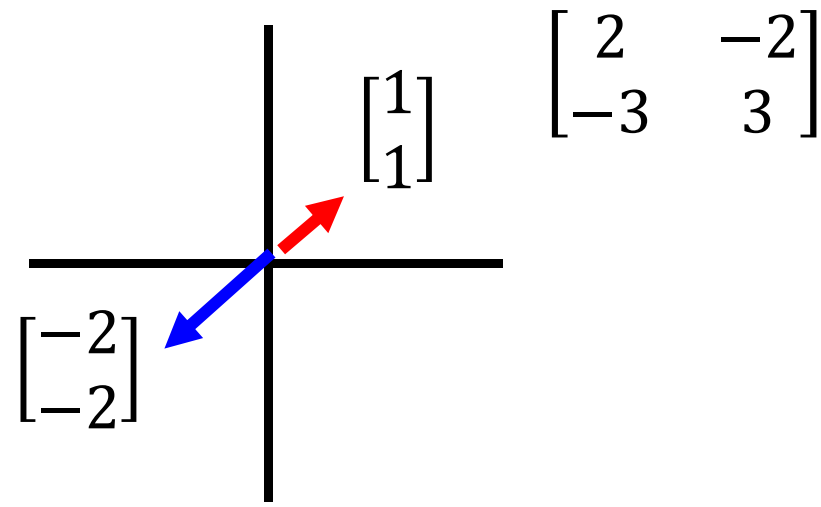
- Example: to check 3 and  $-2$  are eigenvalues of the linear operator  $T$

$$T \left( \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \right) = \begin{bmatrix} -2x_2 \\ -3x_1 + x_2 \end{bmatrix} \quad A = \begin{bmatrix} 0 & -2 \\ -3 & 1 \end{bmatrix}$$

$$Null(A - 3I_n) = ?$$



$$Null(A + 2I_n) = ?$$



# Check Eigenvalues

$Null(A - \lambda I_n)$ :  
eigenspace of  $\lambda$

- Example: check that 3 is an eigenvalue of  $B$  and find a basis for the corresponding eigenspace

$$B = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 1 \end{bmatrix} \quad \text{find the solution set of } (B - 3I_3)\mathbf{x} = \mathbf{0}$$

find the RREF of  
 $B - 3I_3$

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

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_3 \\ x_3 \end{bmatrix}$$

$$= x_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$$



# Looking for eigenvalues

(Chapter 5.2)

# Looking for Eigenvalues

A scalar  $t$  is an eigenvalue of  $A$



Existing  $v \neq 0$  such that  $Av = tv$



Existing  $v \neq 0$  such that  $Av - tv = 0$



Existing  $v \neq 0$  such that  $(A - tI_n)v = 0$



$(A - tI_n)v = 0$  has multiple solution



The columns of  $(A - tI_n)$  are **Dependent**



$(A - tI_n)$  is not invertible



$$\det(A - tI_n) = 0$$

# Looking for Eigenvalues

- Example 1: Find the eigenvalues of  $A = \begin{bmatrix} -4 & -3 \\ 3 & 6 \end{bmatrix}$

A scalar  $t$  is an eigenvalue of  $A$   $\iff$   $\det(A - tI_n) = 0$

$$A - tI_2 = \begin{bmatrix} -4 - t & -3 \\ 3 & 6 - t \end{bmatrix}$$

$$\det(A - tI_2) = 0$$

$$\implies t = -3 \text{ or } 5$$

The eigenvalues of  $A$  are -3 or 5.

# Looking for Eigenvalues

- Example 1: Find the eigenvalues of  $A = \begin{bmatrix} -4 & -3 \\ 3 & 6 \end{bmatrix}$

The eigenvalues of  $A$  are -3 or 5.

## **Eigenspace of -3**

$$Ax = -3x \quad \longrightarrow \quad (A + 3I)x = 0$$

find the solution

## **Eigenspace of 5**

$$Ax = 5x \quad \longrightarrow \quad (A - 5I)x = 0$$

find the solution

# Looking for Eigenvalues

- Example 2: find the eigenvalues of linear operator

$$T \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} -x_1 \\ 2x_1 - x_2 - x_3 \\ -x_3 \end{bmatrix} \xrightarrow{\text{standard matrix}} A = \begin{bmatrix} -1 & 0 & 0 \\ 2 & -1 & -1 \\ 0 & 0 & -1 \end{bmatrix}$$

A scalar  $t$  is an eigenvalue of  $A$   $\iff \det(A - tI_n) = 0$

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

$$\implies \det(A - tI_n) = (-1 - t)^3$$

# Looking for Eigenvalues

- Example 3: linear operator on  $\mathcal{R}^2$  that rotates a vector by  $90^\circ$

A scalar  $t$  is an eigenvalue of  $A$   $\iff det(A - tI_n) = 0$

standard matrix of the  $90^\circ$ -rotation:  $\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$

$$\det \left( \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} - tI_2 \right)$$

No eigenvalues, no eigenvectors

# Characteristic Polynomial

A scalar  $t$  is an eigenvalue of  $A$   $\iff det(A - tI_n) = 0$


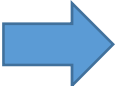
$A$  is the standard matrix of linear operator  $T$

$det(A - tI_n)$ : Characteristic polynomial of  $A$   
linear operator  $T$

$det(A - tI_n) = 0$ : Characteristic equation of  $A$   
linear operator  $T$

Eigenvalues are the roots of characteristic polynomial or solutions of characteristic equation.

# Characteristic Polynomial

- In general, a matrix  $A$  and RREF of  $A$  have different characteristic polynomials.  Different Eigenvalues
- Similar matrices have the same characteristic polynomials  The same Eigenvalues

$$\begin{aligned} \det(B - tI) &= \det(P^{-1}AP - P^{-1}(tI)P) && B = P^{-1}AP \\ &= \det(P^{-1}(A - tI)P) \\ &= \det(P^{-1})\det(A - tI)\det(P) \\ &= \left(\frac{1}{\det(P)}\right)\det(A - tI)\det(P) = \det(A - tI) \end{aligned}$$



# Characteristic Polynomial

- Question: What is the order of the characteristic polynomial of an  $n \times n$  matrix  $A$ ?
  - The characteristic polynomial of an  $n \times n$  matrix is indeed a polynomial with degree  $n$
  - Consider  $\det(A - tI_n)$

$$\det \begin{bmatrix} ? - t & \cdots & ? \\ \vdots & \ddots & \vdots \\ ? & \cdots & ? - t \end{bmatrix}$$

- Question: What is the number of eigenvalues of an  $n \times n$  matrix  $A$ ?
  - Fact: An  $n \times n$  matrix  $A$  have less than or equal to  $n$  eigenvalues
  - Consider complex roots and multiple roots

# Characteristic Polynomial vs. Eigenspace

- Characteristic polynomial of A is

$$\det(A - tI_n) \xrightarrow{\text{Factorization}} \text{multiplicity}$$
$$= (t - \lambda_1)^{m_1} (t - \lambda_2)^{m_2} \dots (t - \lambda_k)^{m_k} (\dots)$$

Eigenvalue:  $\lambda_1$        $\lambda_2$        $\lambda_k$

Eigenspace:  $d_1$        $d_2$        $d_k$

(dimension)       $\leq m_1$        $\leq m_2$        $\leq m_k$

# Characteristic Polynomial

- The eigenvalues of an upper triangular matrix are its diagonal entries.

Characteristic Polynomial:

$$\begin{bmatrix} a & * & * \\ 0 & b & * \\ 0 & 0 & c \end{bmatrix} \quad \det \begin{bmatrix} a - t & * & * \\ 0 & b - t & * \\ 0 & 0 & c - t \end{bmatrix}$$
$$= (a - t)(b - t)(c - t)$$

The determinant of an upper triangular matrix is the product of its diagonal entries.

# Diagonalization

(Chapter 5.3)

# Review

- If  $Av = \lambda v$  ( $v$  is a vector,  $\lambda$  is a scalar)
  - $v$  is an eigenvector of  $A$  **excluding zero vector**
  - $\lambda$  is an eigenvalue of  $A$  that corresponds to  $v$

- Eigenvectors corresponding to  $\lambda$  are **nonzero**  
solution of  $(A - \lambda I_n)\mathbf{v} = \mathbf{0}$

Eigenvectors

corresponding to  $\lambda$

$$= \underline{\text{Null}(A - \lambda I_n)} - \{\mathbf{0}\}$$

**eigenspace**

**Eigenspace of  $\lambda$ :**

Eigenvectors

corresponding to  $\lambda + \{\mathbf{0}\}$

- A scalar  $t$  is an eigenvalue of  $A$



$$\det(A - tI_n) = 0$$

# Review

- Characteristic polynomial of A is

$$\det(A - tI_n)$$

Factorization

Algebraic multiplicity

$$= (t - \lambda_1)^{m_1} (t - \lambda_2)^{m_2} \dots (t - \lambda_k)^{m_k} (\dots)$$

Eigenvalue:

$\lambda_1$

$\lambda_2$

$\lambda_k$

Eigenspace:

$d_1$

$d_2$

$d_k$

(dimension)

$\leq m_1$

$\leq m_2$

$\leq m_k$

# Outline

- An  $n \times n$  matrix  $A$  is called **diagonalizable** if  $A = PDP^{-1}$ 
  - $D$ :  $n \times n$  diagonal matrix
  - $P$ :  $n \times n$  invertible matrix
- Is a matrix  $A$  **diagonalizable**?
  - If yes, find  $D$  and  $P$

# Diagonalizable

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

- Not all matrices are diagonalizable

$$\Rightarrow A^2 = 0 \quad (?)$$

If  $A = PDP^{-1}$  for some invertible  $P$  and diagonal  $D$

$$\Rightarrow A^2 = PD^2P^{-1} = 0 \quad \Rightarrow D^2 = 0 \quad \Rightarrow D = 0$$

$$\Rightarrow A = 0 \quad \times$$

D is diagonal

$$D = \begin{bmatrix} \overset{=0}{d_1} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \underset{=0}{d_n} \end{bmatrix}$$

$$D^2 = \begin{bmatrix} \overset{=0}{(d_1)^2} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \underset{=0}{(d_n)^2} \end{bmatrix}$$



# Diagonalizable

$$P = [p_1 \quad \cdots \quad p_n]$$

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

- If A is diagonalizable

$$A = PDP^{-1} \quad \longrightarrow \quad AP = PD$$

$$\longrightarrow AP = [\underline{Ap_1} \quad \cdots \quad \underline{Ap_n}]$$

$$\longrightarrow PD = P[d_1e_1 \quad \cdots \quad d_ne_n]$$

$$= [Pd_1e_1 \quad \cdots \quad Pd_ne_n]$$

$$= [d_1Pe_1 \quad \cdots \quad d_nPe_n]$$

$$= [\underline{d_1p_1} \quad \cdots \quad \underline{d_np_n}] \quad \longrightarrow \quad Ap_i = d_ip_i$$

$p_i$  is an eigenvector of A corresponding to eigenvalue  $d_i$

# Diagonalizable

- If A is diagonalizable

$$A = PDP^{-1}$$

||

There are n eigenvectors that form an invertible matrix

||

There are n independent eigenvectors

||

The eigenvectors of A can form a basis for  $\mathbb{R}^n$ .

$$P = [p_1 \quad \cdots \quad p_n]$$

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

$p_i$  is an eigenvector of A  
corresponding to eigenvalue  $d_i$

# Diagonalizable

- If  $A$  is diagonalizable

$$A = PDP^{-1}$$

$$P = [p_1 \quad \cdots \quad p_n]$$

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

$p_i$  is an eigenvector of  $A$   
corresponding to eigenvalue  $d_i$

## How to diagonalize a matrix $A$ ?

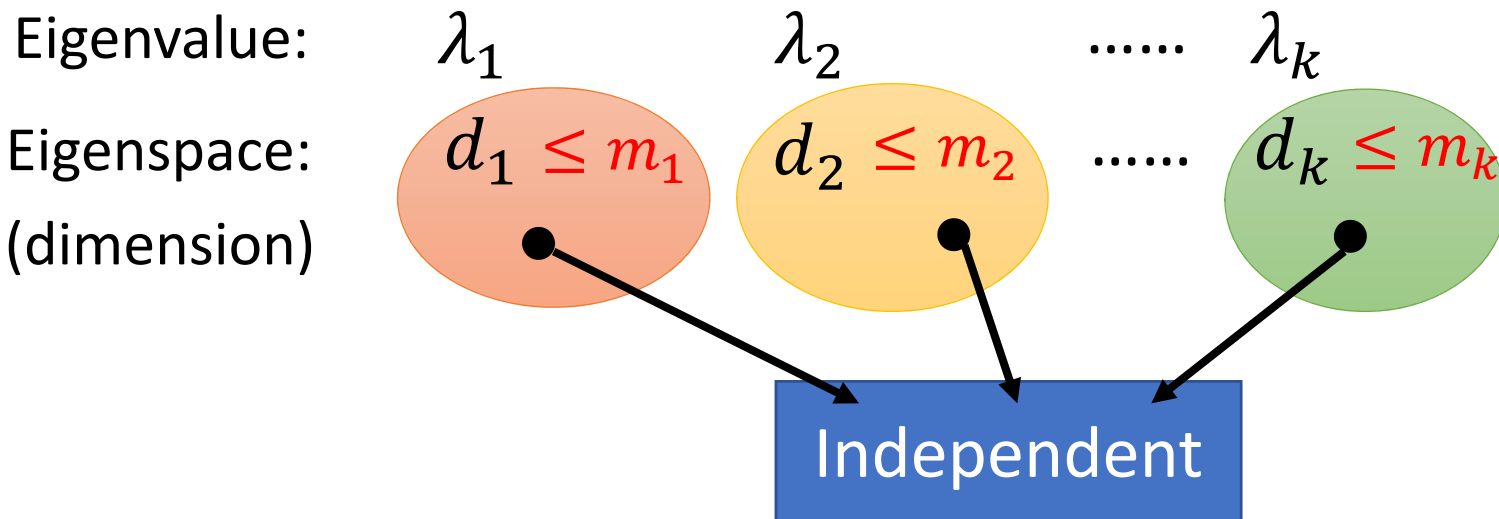
- Step 1: Find  $n$  independent eigenvectors corresponding if possible, and form an invertible  $P$
- Step 2: The eigenvalues corresponding to the eigenvectors in  $P$  form the diagonal matrix  $D$ .

# Diagonalizable

A set of eigenvectors that correspond to distinct eigenvalues is linearly independent.

$$\det(A - tI_n) \quad \text{Factorization}$$

$$= (t - \lambda_1)^{\underline{m_1}} (t - \lambda_2)^{\underline{m_2}} \dots (t - \lambda_k)^{\underline{m_k}} (\dots \dots)$$



# Diagonalizable

A set of eigenvectors that correspond to distinct eigenvalues is linearly independent.

Eigenvalue:  $\lambda_1$     $\lambda_2$    .....    $\lambda_m$

Assume dependent

Eigenvector:  $v_1$     $v_2$    .....    $v_m$

➡ a contradiction

$$\mathbf{v}_k = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_{k-1} \mathbf{v}_{k-1}$$

$$A\mathbf{v}_k = c_1 A\mathbf{v}_1 + c_2 A\mathbf{v}_2 + \dots + c_{k-1} A\mathbf{v}_{k-1}$$

$$\lambda_k \mathbf{v}_k = c_1 \lambda_1 \mathbf{v}_1 + c_2 \lambda_2 \mathbf{v}_2 + \dots + c_{k-1} \lambda_{k-1} \mathbf{v}_{k-1}$$

$$- \lambda_k \mathbf{v}_k = c_1 \lambda_k \mathbf{v}_1 + c_2 \lambda_k \mathbf{v}_2 + \dots + c_{k-1} \lambda_k \mathbf{v}_{k-1}$$

$(\lambda_k)$

---


$$\mathbf{0} = c_1 (\lambda_1 - \lambda_k) \mathbf{v}_1 + c_2 (\lambda_2 - \lambda_k) \mathbf{v}_2 + \dots + c_{k-1} (\lambda_{k-1} - \lambda_k) \mathbf{v}_{k-1}$$

Not  $c_1 = c_2 = \dots = c_{k-1} = 0$  ➡ Same eigenvalue ➡ a contradiction

# Diagonalizable

$$P = [p_1 \quad \cdots \quad p_n]$$

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

- If A is diagonalizable

$$A = PDP^{-1}$$

$p_i$  is an eigenvector of A  
corresponding to eigenvalue  $d_i$

$$\det(A - tI_n)$$

$$= (t - \lambda_1)^{m_1} (t - \lambda_2)^{m_2} \cdots (t - \lambda_k)^{m_k} (\dots)$$

Eigenvalue:

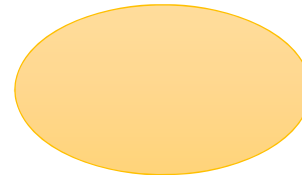
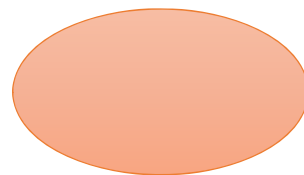
$\lambda_1$

$\lambda_2$

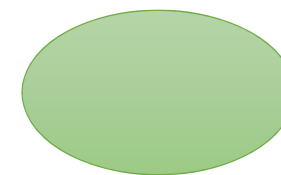
.....

$\lambda_k$

Eigenspace:



.....



Basis for  $\lambda_1$

Basis for  $\lambda_2$

Basis for  $\lambda_3$




Independent Eigenvectors

You can't find more!

# Diagonalizable - Example

- Diagonalize a given matrix  $A = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 1 \end{bmatrix}$

characteristic polynomial is  $-(t + 1)^2(t - 3)$   eigenvalues: 3, -1

eigenvalue 3

$$B_1 = \left\{ \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \right\}$$

eigenvalue -1

$$B_2 = \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} \right\}$$

$$A = PDP^{-1},$$

where

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

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

# Test for a Diagonalizable Matrix

- An  $n \times n$  matrix  $A$  is diagonalizable if and only if both the following conditions are met.

The characteristic polynomial of  $A$  factors into a product of linear factors.

$$\begin{aligned} \det(A - tI_n) & \text{ Factorization} \\ & = (t - \lambda_1)^{m_1} (t - \lambda_2)^{m_2} \dots (t - \lambda_k)^{m_k} \end{aligned}$$

For each eigenvalue  $\lambda$  of  $A$ , the multiplicity of  $\lambda$  (**algebraic multiplicity**) equals the dimension of the corresponding eigenspace (**geometric multiplicity**).



# Independent Eigenvectors

An  $n \times n$  matrix  $A$  is diagonalizable

||

The eigenvectors of  $A$  can form a basis for  $\mathbb{R}^n$ .

||

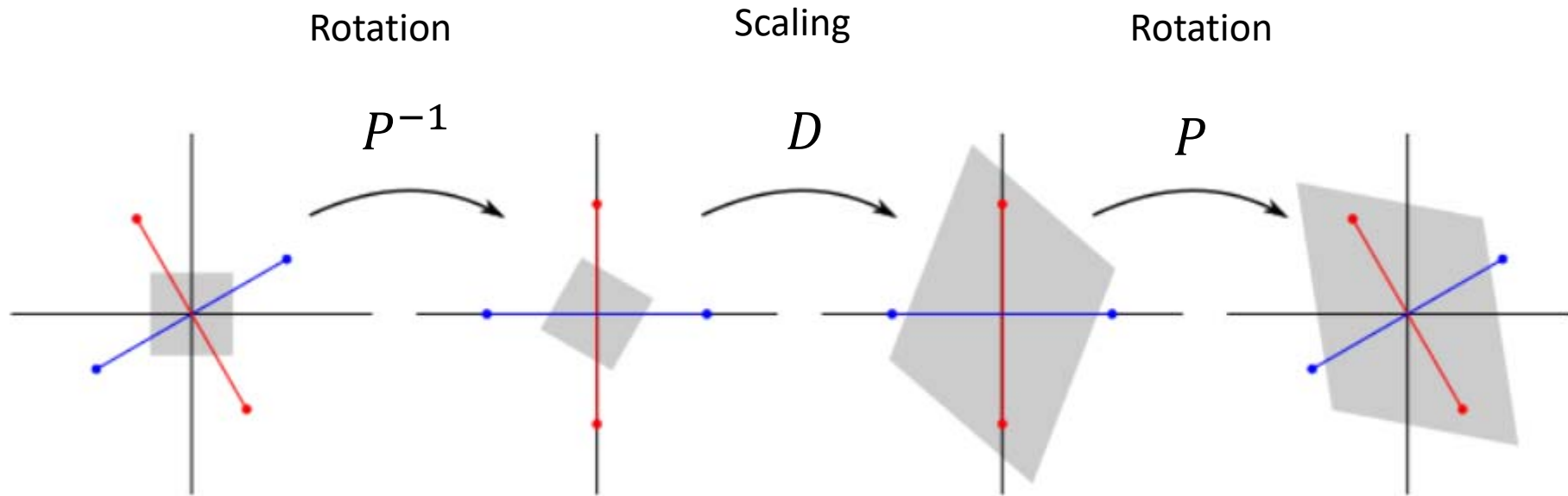
$$\det(A - tI_n)$$

$$= (t - \lambda_1)^{m_1} (t - \lambda_2)^{m_2} \dots (t - \lambda_k)^{m_k} \text{ (.....)}$$

Eigenvalue:	$\lambda_1$	$\lambda_2$	.....	$\lambda_k$
Eigenspace: (dimension)	$d_1 = m_1$	$d_2 = m_2$	.....	$d_k = m_k$

$$d_1 + d_2 + \dots + d_k = n$$

# Geometric Meaning of Diagonalization $A = PDP^{-1}$



Red and blue axes correspond to the directions of two eigenvectors



# How to Cope with Non-diagonalizable Matrices

**Question:** If  $A$  is not diagonalizable (i.e.,  $A \neq PDP^{-1}$ ), can we write  $A = UTU^{-1}$ , where  $T$  is “near diagonal”?

**Schur Decomposition:** Any square matrix  $A$  can be written as  $A = UTU^{-1}$ , where  $U$  is an “orthonormal matrix” and  $T$  is “upper triangular” with eigenvalues on the diagonal.

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



# Jordan Normal Form

**Theorem:** For every  $n \times n$  matrix  $A$ , there exists an invertible matrix  $Q$  such that  $Q^{-1}AQ = J$  where  $J$  is in **Jordan Normal Form**.

Jordan Block

$$J = \begin{bmatrix} \lambda & 1 & 0 & \cdots & 0 \\ 0 & \lambda & 1 & & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda & 1 \\ 0 & 0 & 0 & \cdots & \lambda \end{bmatrix}_{k \times k}$$



Eigenspace of 2  
has dim = 1

Eigenspace of 3  
has dim = 2

Jordan Normal Form

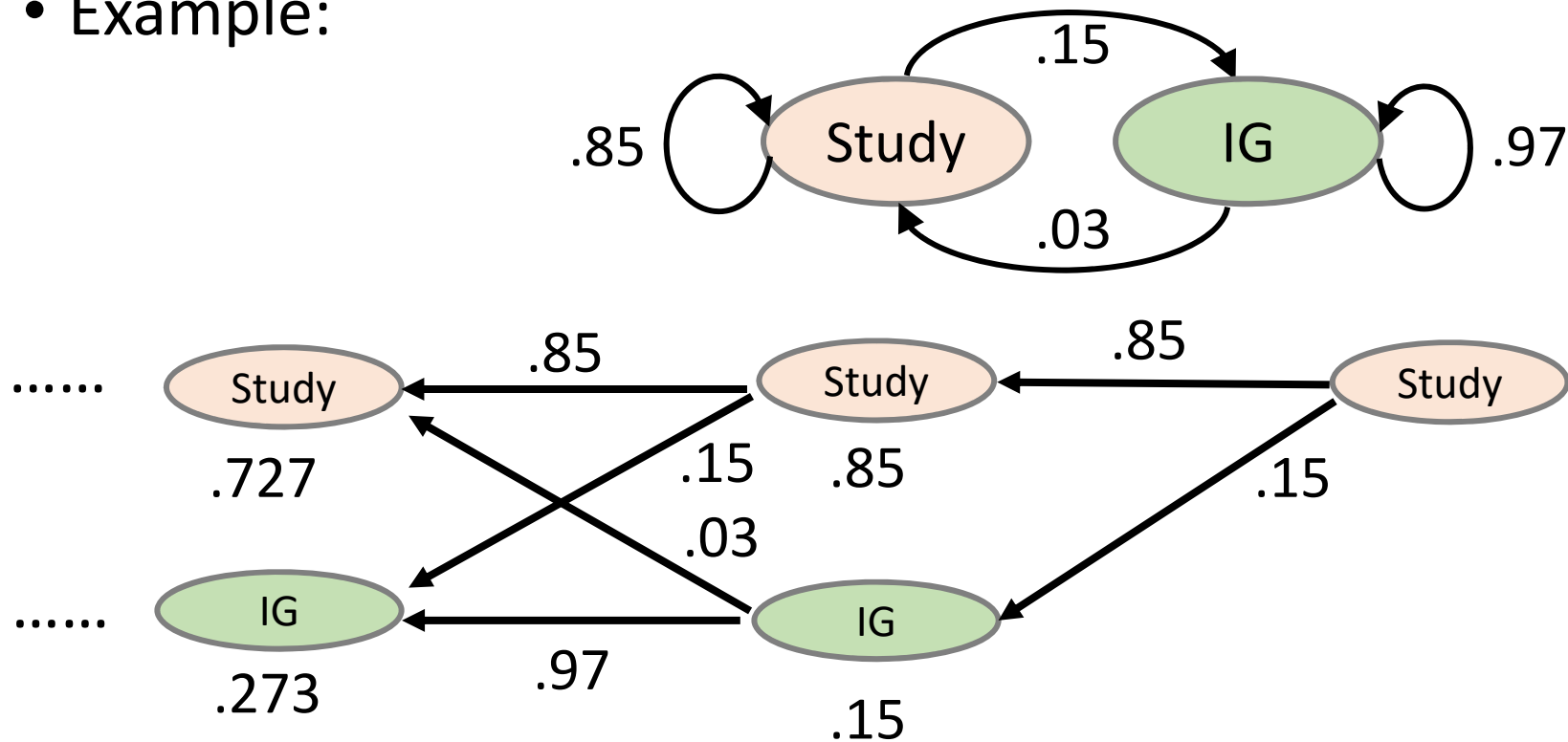
$$J = \begin{bmatrix} 2 & 1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 3 & 1 & 0 \\ 0 & 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 0 & 3 \end{bmatrix}$$

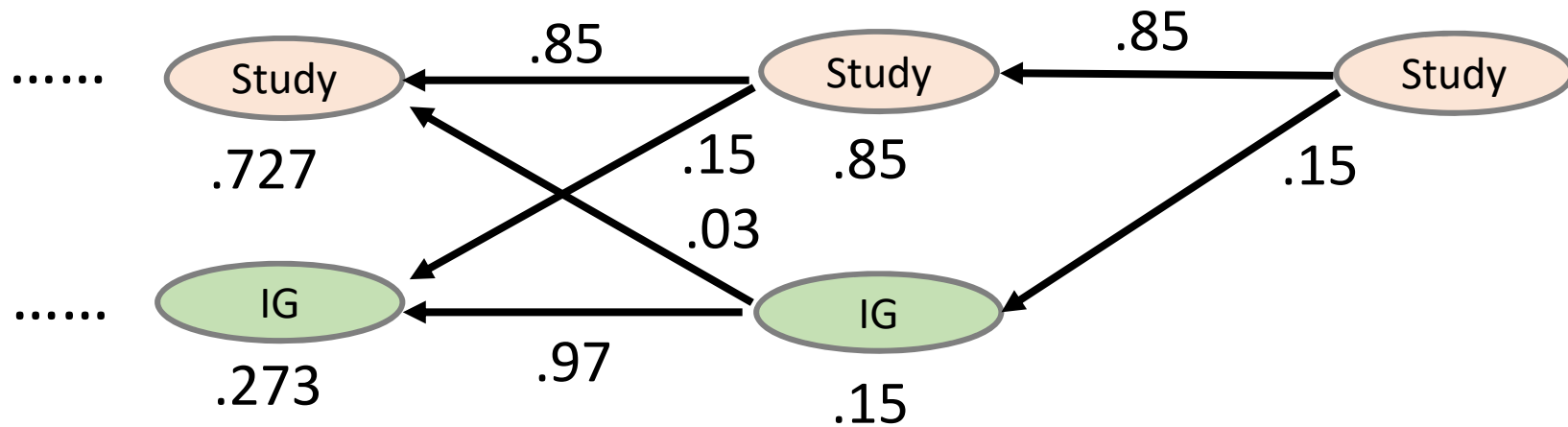
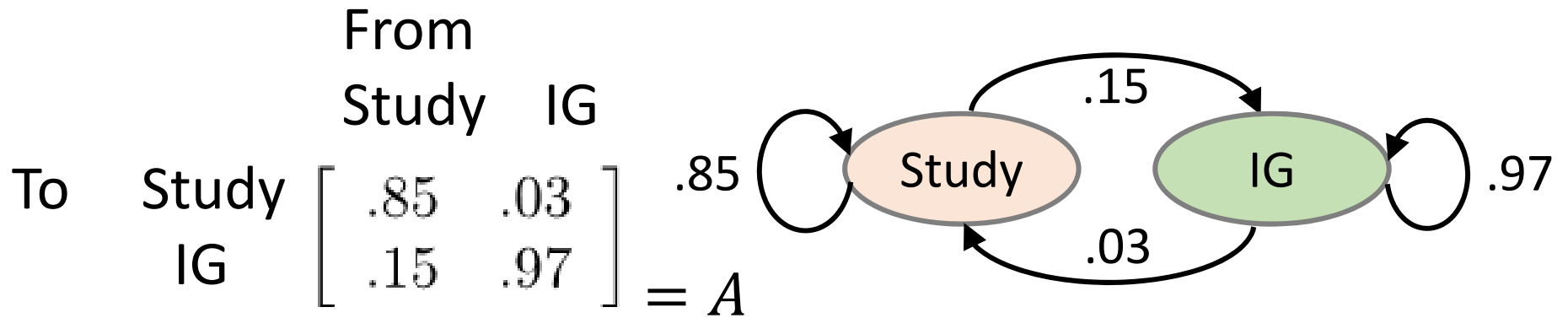
# Application of Diagonalization

- If  $A$  is diagonalizable,

$$A = PDP^{-1} \longrightarrow A^m = PD^mP^{-1}$$

- Example:





$$\begin{bmatrix} .727 \\ .273 \end{bmatrix} \begin{bmatrix} .85 & .03 \\ .15 & .97 \end{bmatrix} \begin{bmatrix} .85 \\ .15 \end{bmatrix} \begin{bmatrix} .85 & .03 \\ .15 & .97 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$A = PDP^{-1} \longrightarrow A^m = PD^mP^{-1}$$

# Diagonalizable

- Diagonalize a given matrix  $A = \begin{bmatrix} .85 & .03 \\ .15 & .97 \end{bmatrix}$

$$\det (A - tI_2)$$

$$\begin{array}{l} A - .82I_2 \xrightarrow{\text{RREF}} \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \Rightarrow \mathbf{p}_1 = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \\ A - I_2 \xrightarrow{\text{RREF}} \begin{bmatrix} 1 & -.2 \\ 0 & 0 \end{bmatrix} \Rightarrow \mathbf{p}_2 = \begin{bmatrix} 1 \\ 5 \end{bmatrix} \end{array} \Rightarrow P = \begin{bmatrix} -1 & 1 \\ 1 & 5 \end{bmatrix} \text{ (invertible)}$$
$$D = \begin{bmatrix} .82 & 0 \\ 0 & 1 \end{bmatrix}$$

$$A = PDP^{-1} \text{ where } P = \begin{bmatrix} -1 & 1 \\ 1 & 5 \end{bmatrix}, D = \begin{bmatrix} .82 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\begin{aligned} A^m &= PD^m P^{-1} \\ &= \begin{bmatrix} -1 & 1 \\ 1 & 5 \end{bmatrix} \begin{bmatrix} (.82)^m & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 1 & 5 \end{bmatrix}^{-1} \\ &= \frac{1}{6} \begin{bmatrix} 1 + 5(.82)^m & 1 - (.82)^m \\ 5 - 5(.82)^m & 5 + (.82)^m \end{bmatrix} \end{aligned}$$

When  $m \rightarrow \infty$ ,

$$A^m = \begin{bmatrix} 1/6 & 1/6 \\ 5/6 & 5/6 \end{bmatrix}$$

The beginning condition does not influence.

$$\begin{bmatrix} 1/6 & 1/6 \\ 5/6 & 5/6 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1/6 \\ 5/6 \end{bmatrix} \quad \begin{bmatrix} 1/6 & 1/6 \\ 5/6 & 5/6 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1/6 \\ 5/6 \end{bmatrix}$$



# Diagonalization of linear operators\*

(Chapter 5.4)

# Diagonalization of Linear Operator

• Example 1:  $T \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} 8x_1 + 9x_2 \\ -6x_1 - 7x_2 \\ 3x_1 + 3x_2 - x_3 \end{bmatrix}$

The standard matrix is  $A = \begin{bmatrix} 8 & 9 & 0 \\ -6 & -7 & 0 \\ 3 & 3 & -1 \end{bmatrix}$

$\Rightarrow$  the characteristic polynomial is  $-(t + 1)^2(t - 2)$

Eigenvalue -1:

$$\mathcal{B}_1 = \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$$

Eigenvalue 2:

$$\mathcal{B}_2 = \left\{ \begin{bmatrix} 3 \\ -2 \\ 1 \end{bmatrix} \right\}$$

$\Rightarrow \mathcal{B}_1 \cup \mathcal{B}_2$  is a basis of  $\mathcal{R}^3$

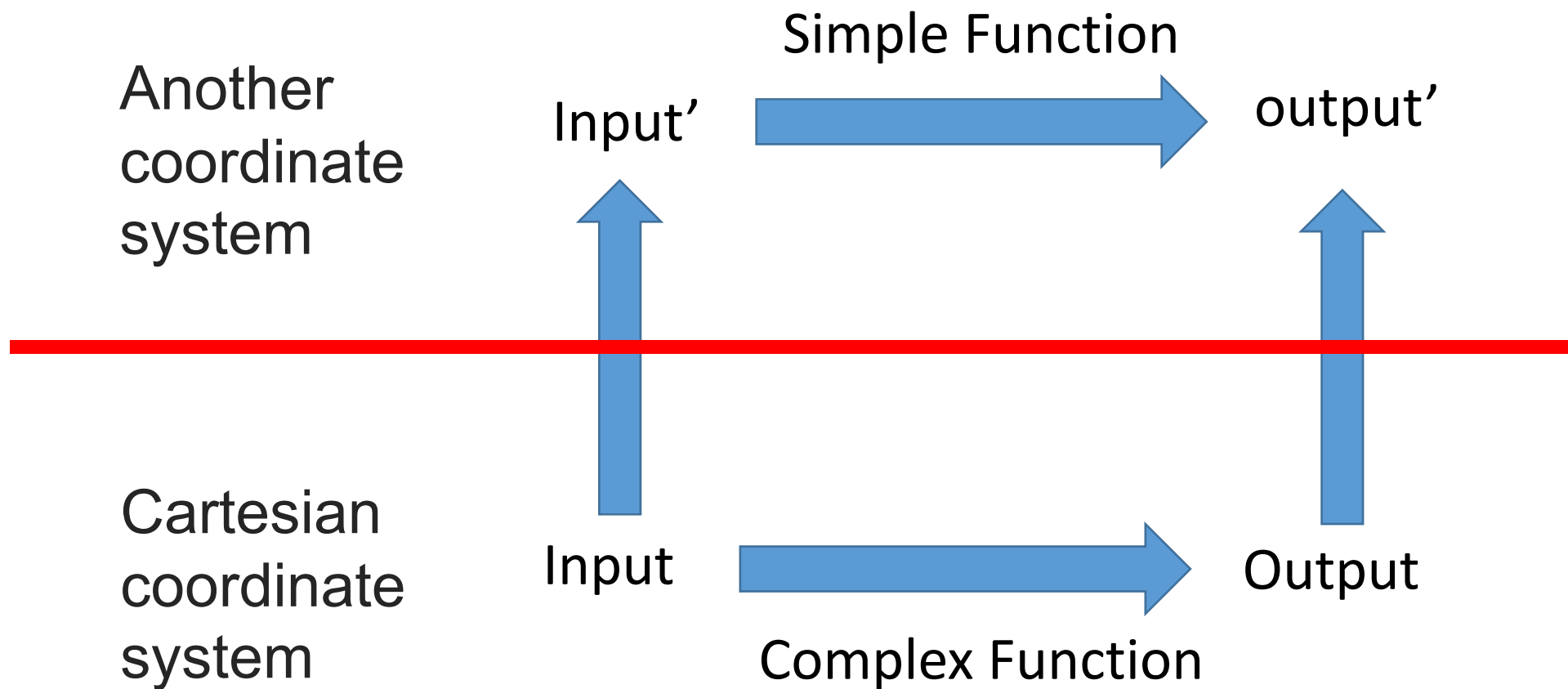
# Diagonalization of Linear Operator

• Example 2:  $T \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} -x_1 + x_2 + 2x_3 \\ x_1 - x_2 \\ 0 \end{bmatrix}$

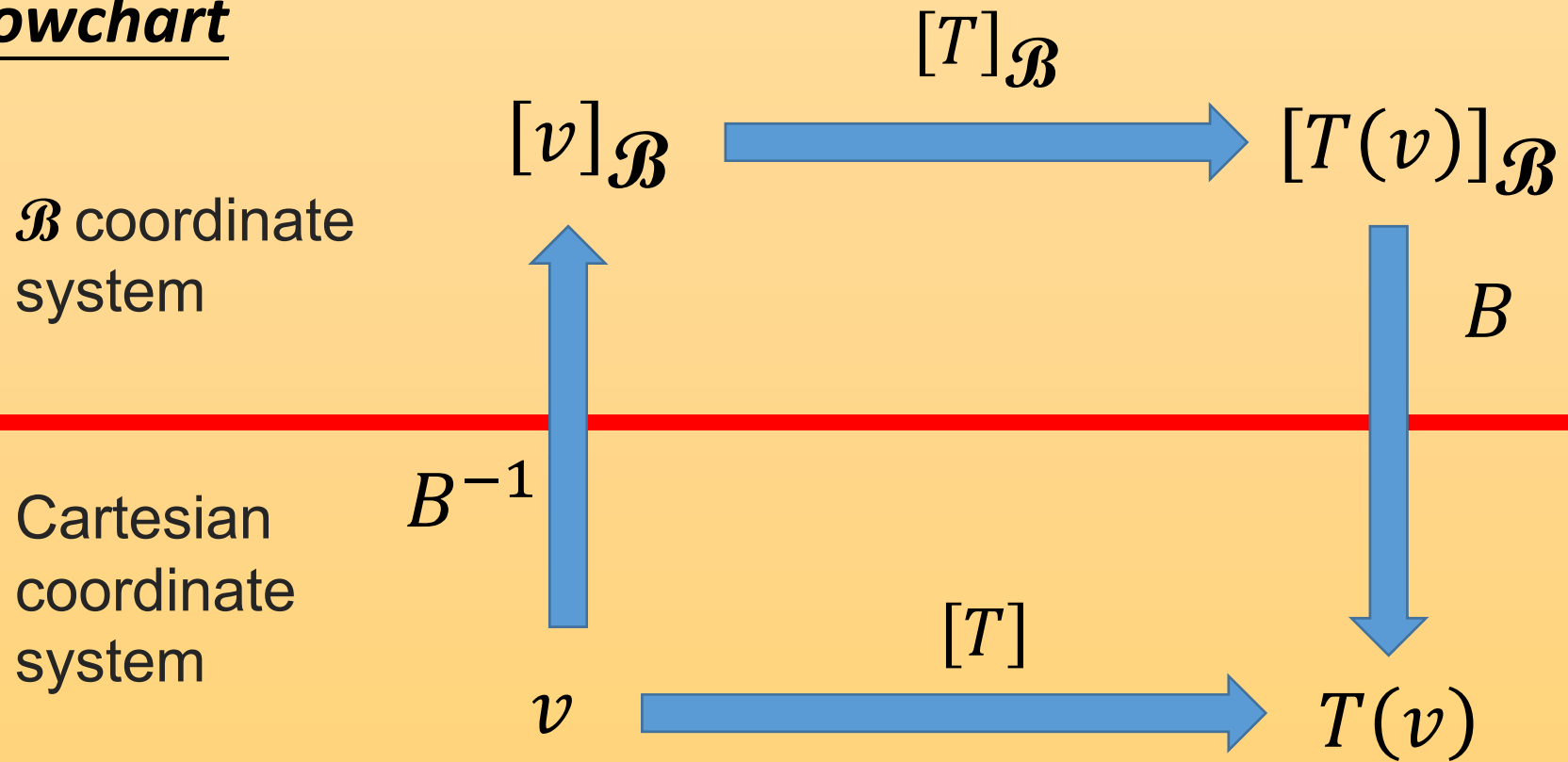
The standard matrix is  $A = \begin{bmatrix} -1 & 1 & 2 \\ 1 & -1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$

$$\begin{aligned} x_1 - x_2 &= 0 \\ x_3 &= 0 \end{aligned} \quad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad ; \quad \begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

# Review



## Flowchart



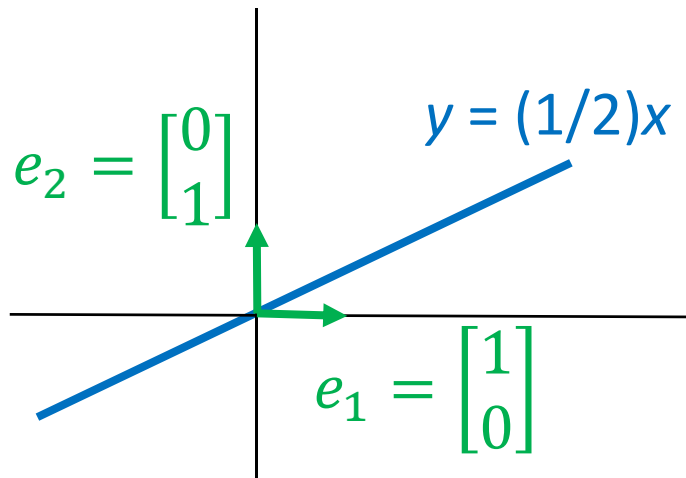
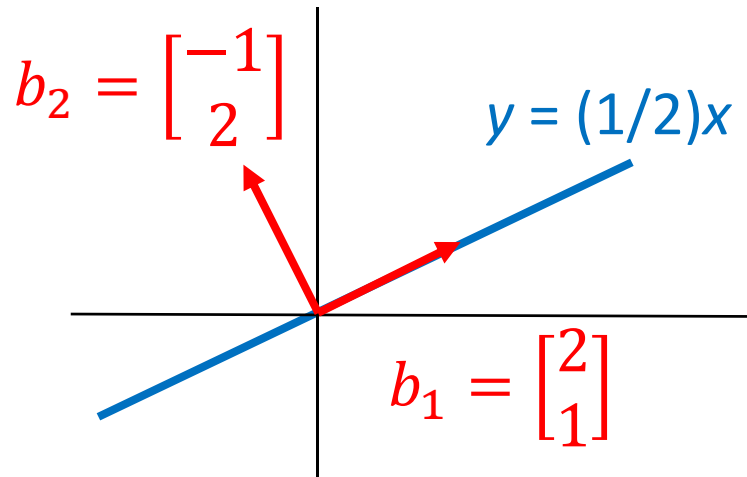
$$[T] = B [T]_{\mathcal{B}} B^{-1}$$

similar

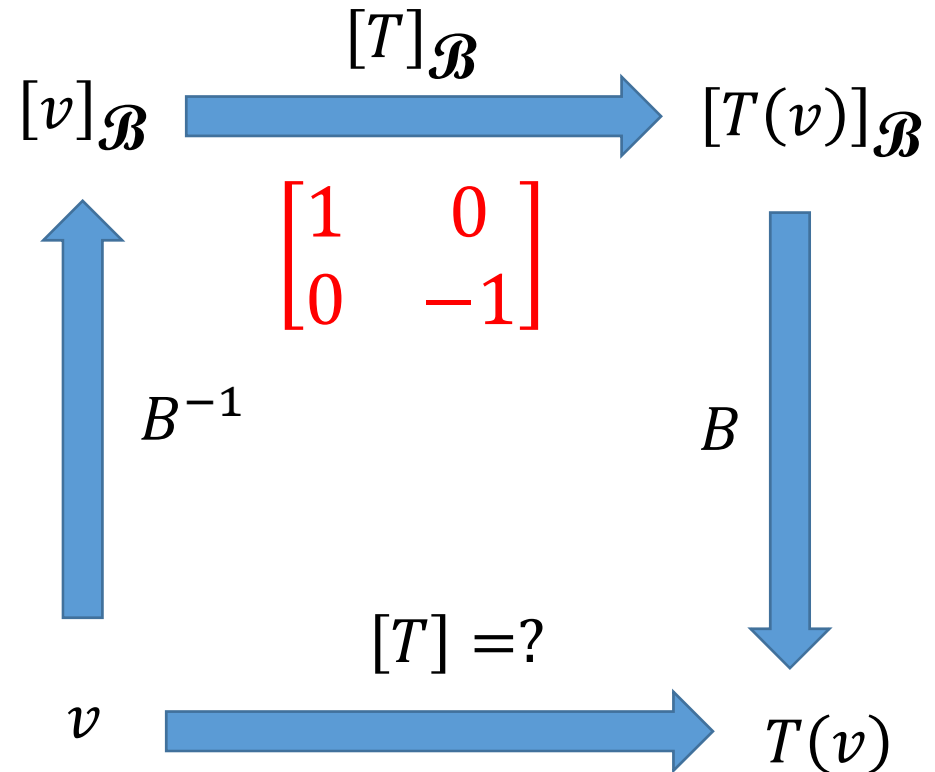
$$[T]_{\mathcal{B}} = B^{-1} [T] B$$

similar

- Example: reflection operator  $T$  about the line  $y = (1/2)x$



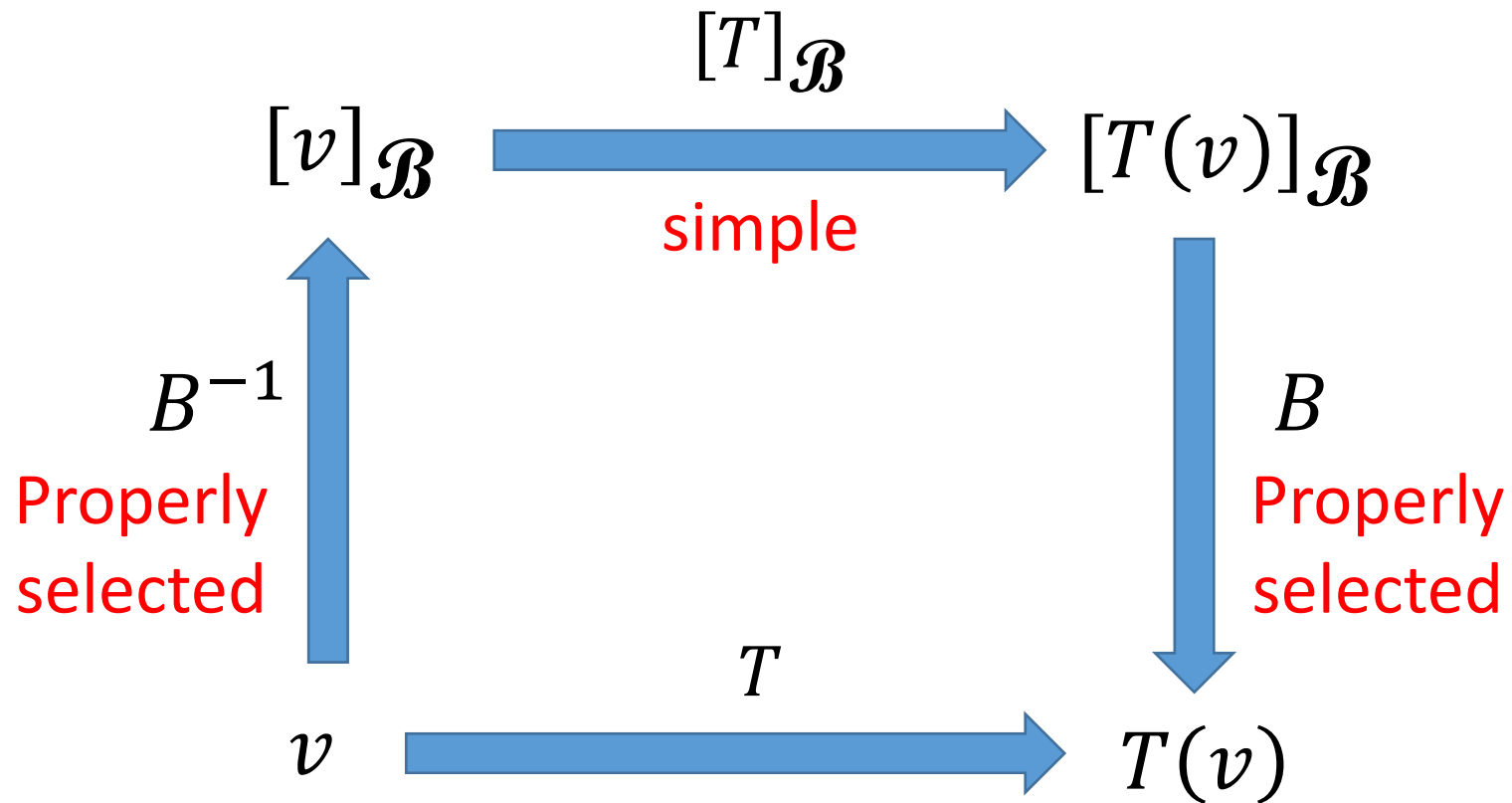
$$B^{-1} = \begin{bmatrix} 0.4 & 0.2 \\ -0.2 & 0.4 \end{bmatrix} \quad B = \begin{bmatrix} 2 & -1 \\ 1 & 2 \end{bmatrix}$$



$$[T] = B[T]_{\mathcal{B}}B^{-1}$$

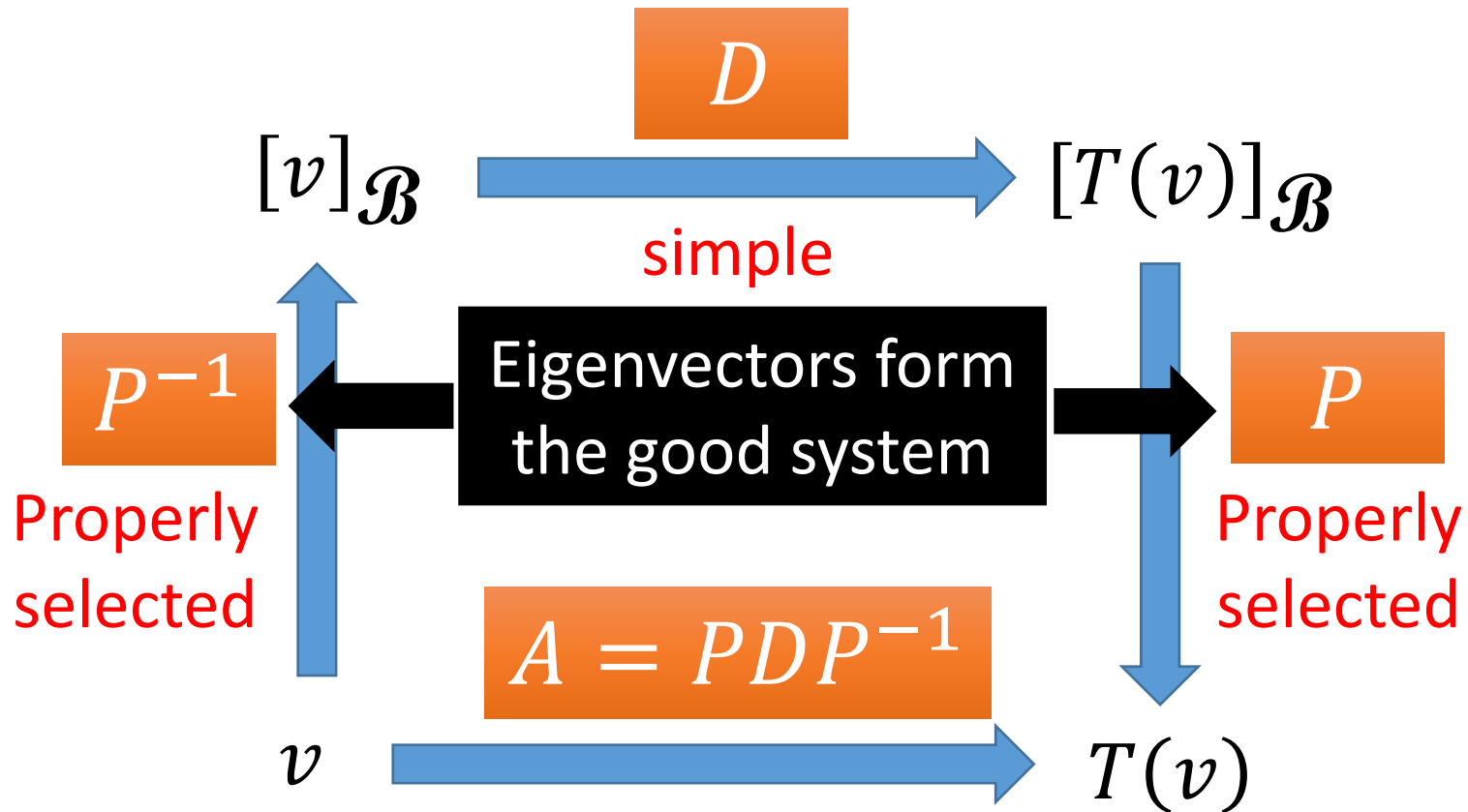
# Diagonalization of Linear Operator

- Reference: Chapter 5.4



# Diagonalization of Linear Operator

- If a linear operator  $T$  is diagonalizable





# Diagonalization of Linear Operator

$$T \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} 8x_1 + 9x_2 \\ -6x_1 - 7x_2 \\ 3x_1 + 3x_2 - x_3 \end{bmatrix} \quad \begin{matrix} -1: \\ \mathcal{B}_1 = \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\} \end{matrix} \quad \begin{matrix} 2: \\ \mathcal{B}_2 = \left\{ \begin{bmatrix} 3 \\ -2 \\ 1 \end{bmatrix} \right\} \end{matrix}$$

