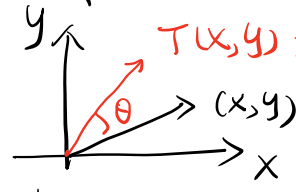


Linear Transformation  $T: V \rightarrow W$   
 $\vec{v} \mapsto T(\vec{v})$

$T$  preserves two operations  $\left\{ \begin{array}{l} \textcircled{1} T(\vec{u} + \vec{v}) = T(\vec{u}) + T(\vec{v}), \quad \forall \vec{u}, \vec{v} \in V \\ \textcircled{2} T(a\vec{v}) = aT(\vec{v}), \quad \forall a \in \mathbb{R} \Rightarrow T(\vec{0}) = \vec{0} \end{array} \right.$

Examples of Linear Transformation from  $\mathbb{R}^2 \rightarrow \mathbb{R}^2$

① Rotation   $T(x, y) = (\cos\theta x - \sin\theta y, \sin\theta x + \cos\theta y)$   
 $= \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$

② Reflection   $T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$

③ Projection   $T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$

Ex 1:  $V = C(\mathbb{R}) = \{\text{all continuous functions}\}$

$T: V \rightarrow \mathbb{R}$   
 $f(x) \mapsto \int_a^b f(t) dt$

Ex 2:  $V = \{\text{all differentiable functions}\}$

$W = \{\text{all functions}\}$

$T: V \rightarrow W$   
 $f(x) \mapsto f'(x)$

Def ordered basis: a basis with an order.

Ex:  $\mathbb{R}^3$ ,  $e_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ ,  $e_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ ,  $e_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$

Ex:  $P_n(\mathbb{R})$ ,  $\{1, x, x^2, \dots, x^n\}$

Def: Given an ordered basis  $\beta = \{\vec{u}_1, \dots, \vec{u}_n\}$  of  $V$   
 $\forall \vec{v} \in V, \vec{v} = a_1 \vec{u}_1 + \dots + a_n \vec{u}_n$   
 $[\vec{v}]_\beta = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}$  is the coordinate of  $\vec{v}$  under basis  $\beta$ .

Ex:  $P_2(\mathbb{R}), \beta = \{1, x, x^2\}$

$$f(x) = 4 + 6x - 7x^2$$

$$[f]_\beta = \begin{bmatrix} 4 \\ 6 \\ -7 \end{bmatrix}$$

Def  $T: V \rightarrow W$

$\beta = \{\vec{v}_1, \dots, \vec{v}_n\}$  is an ordered basis of  $V$

$\gamma = \{\vec{w}_1, \dots, \vec{w}_m\}$  is an ordered basis of  $W$

Then  $T(\vec{v}_j) = \sum_{i=1}^m a_{ij} \vec{w}_i, \forall j=1, \dots, n$ .

$A = [a_{ij}]_{m \times n}$  is the matrix representation

of  $T$  under bases  $\beta$  and  $\gamma$ .

We write  $A = [T]_\beta^\gamma$

$$A = \begin{bmatrix} a_{1j} \\ \vdots \\ a_{mj} \end{bmatrix}_{m \times n}$$

$$\Downarrow$$

$$[T(\vec{v}_j)]_\gamma$$

$T(\vec{v}_j) \in W$ .

Ex:  $T: P_3(\mathbb{R}) \rightarrow P_2(\mathbb{R})$

$$f(x) \mapsto f'(x)$$

$\beta = \{1, x, x^2, x^3\}$

$\gamma = \{1, x, x^2\}$

$$T(1) = 0 = 0 \cdot 1 + 0 \cdot X + 0 \cdot X^2$$

$$T(X) = 1 = 1 \cdot 1 + 0 \cdot X + 0 \cdot X^2$$

$$T(X^2) = 2X = 0 \cdot 1 + 2 \cdot X + 0 \cdot X^2$$

$$T(X^3) = 3X^2 = 0 \cdot 1 + 0 \cdot X + 3 \cdot X^2$$

$$[T]_{\beta}^{\beta} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \end{bmatrix}_{3 \times 4}$$

Ex:  $I_V: V \rightarrow V$   
 $\vec{v} \mapsto \vec{v} \quad \beta = \{\vec{v}_1, \dots, \vec{v}_n\}$

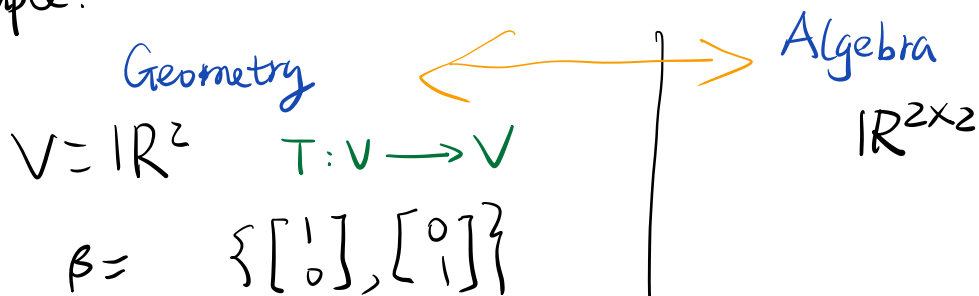
$$[I_V]_{\beta}^{\beta} = \begin{pmatrix} 1 & 0 \\ 0 & \ddots \\ 0 & & 1 \end{pmatrix}$$

Def  $T: V \rightarrow V \quad \beta = \{v_1, \dots, v_n\}$

$$[T]_{\beta}^{\beta} \in \mathbb{R}^{n \times n}$$

Remark: Given an  $n$ -dimensional  $V$  with an ordered basis  $\beta$ , any linear transformation  $T: V \rightarrow V$  corresponds to a square matrix  $[T]_{\beta}^{\beta}$

Example:



① Rotation by  $30^\circ$  counterclockwise

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{pmatrix} \cos\frac{\pi}{6} & -\sin\frac{\pi}{6} \\ \sin\frac{\pi}{6} & \cos\frac{\pi}{6} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

② Reflection w.r.t. x-axis

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

③ Projection to x-axis

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

④ Identity

$$[T]_{\beta}^{\beta} = \begin{pmatrix} \cos\frac{\pi}{6} & -\sin\frac{\pi}{6} \\ \sin\frac{\pi}{6} & \cos\frac{\pi}{6} \end{pmatrix}$$

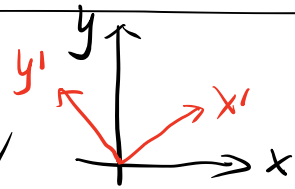
$$[T]_{\beta}^{\beta} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

$$[T]_{\beta}^{\beta} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

$$[I]_{\beta}^{\beta} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

### Change of Coordinates

Given  $\beta = \{\vec{v}_1, \dots, \vec{v}_n\}$  } bases of  $V$   
 $\gamma = \{\vec{w}_1, \dots, \vec{w}_m\}$



Q: Given  $[\vec{v}]_{\beta}$ , what is  $[\vec{v}]_{\gamma}$ ?

$$I : V \rightarrow V$$

$$\vec{v} \mapsto \vec{v}$$

$$[\vec{v}]_{\gamma} = [I(\vec{v})]_{\gamma} = [I]_{\beta}^{\gamma} [\vec{v}]_{\beta}$$

$$T : \begin{matrix} n \\ V \\ \beta \end{matrix} \rightarrow \begin{matrix} m \\ W \\ \gamma \end{matrix}$$

$$T(\vec{v}_j) = \sum_{i=1}^m a_{ij} \vec{w}_i$$

$$\vec{v} \mapsto T(\vec{v})$$

$$[T]_{\beta}^{\gamma} = (a_{ij})_{m \times n}$$

$$[I]_{\beta}^{\gamma} = \left( \begin{array}{c} \vdots \\ \text{j-th col} \\ \vdots \end{array} \right)_{n \times n} = \left( \begin{array}{ccc|c} 1 & & & 1 \\ & 1 & & 2 \\ & & 1 & 1 \\ & & & 0 \end{array} \right)$$

Ex:  $P_2(\mathbb{R})$      $\beta = \{1, x, x^2\}$   
 $\gamma = \{1, (x-1), (x-1)^2\}$

$$I(1) = 1 = 1 \cdot 1 + 0 \cdot (x-1) + 0 \cdot (x-1)^2$$

$$I(x) = x = 1 \cdot 1 + 1 \cdot (x-1) + 0 \cdot (x-1)^2$$

$$I(x^2) = x^2 = a \cdot 1 + b(x-1) + c(x-1)^2$$

$$\text{Solve for } a, b, c \Rightarrow \begin{cases} a=1 \\ b=2 \\ c=1 \end{cases}$$

$$= 1 \cdot 1 + 2 \cdot (x-1) + 1 \cdot (x-1)^2$$

Given  $f(x) = ax^2 + bx + c$

want  $f(x) = A(x-1)^2 + B(x-1) + C$

$$[f]_{\beta} = \begin{pmatrix} c \\ b \\ a \end{pmatrix}, \quad [f]_{\gamma} = \begin{bmatrix} C \\ B \\ A \end{bmatrix}$$

$$[f]_{\gamma} = [I]_{\beta}^{\gamma} [f]_{\beta}$$

$$\begin{bmatrix} C \\ B \\ A \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} c \\ b \\ a \end{bmatrix} = \begin{pmatrix} a+b+c \\ b+2a \\ a \end{pmatrix}$$

Now just focus on  $L_A: \mathbb{R}^n \rightarrow \mathbb{R}^n$   
 $\vec{v} \mapsto A\vec{v}$

for a square matrix  $A \in \mathbb{R}^n$

Theorem

Let  $\beta = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$  be an ordered basis of  $\mathbb{R}^n$

Let  $V = [\vec{v}_1 \ \vec{v}_2 \ \dots \ \vec{v}_n] \in \mathbb{R}^{n \times n}$

For  $L_A: \mathbb{R}^n \rightarrow \mathbb{R}^n$ , its matrix  
representation under basis  $\beta$

$$\text{is } [L_A]_{\beta}^{\beta} = V^{-1}AV$$

Remark: If  $V^{-1}AV = D$  is diagonal ( $\Leftrightarrow A = VDV^{-1}$ )  
then it means the matrix representation of

the linear transformation  $L_A$  is diagonal under  
basis  $\beta = \{\vec{v}_1, \dots, \vec{v}_n\}$   $L_A(\vec{v}_i) = d_i \vec{v}_i$

Question: how to find  $\beta$  if this is possible?

Answer: find  $n$  independent eigenvectors of  $A$

## Theorem

Let  $\beta = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$  be an ordered basis of  $\mathbb{R}^n$   
 $\gamma = \{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_n\}$  be an ordered basis of  $\mathbb{R}^n$

Let  $V = [\vec{v}_1 \ \vec{v}_2 \ \dots \ \vec{v}_n] \in \mathbb{R}^{n \times n}$   
 $U = [\vec{u}_1 \ \vec{u}_2 \ \dots \ \vec{u}_n] \in \mathbb{R}^{n \times n}$

For  $L_A: \mathbb{R}^n \rightarrow \mathbb{R}^n$ , its matrix  
representation under bases  $\beta$  and  $\gamma$

$$\text{is } [L_A]_{\beta}^{\gamma} = U^{-1}AV$$

## Corollary:

If  $\beta = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$  are left singular vectors  
 $\gamma = \{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_n\}$  are right singular vectors,

then  $\underline{A = U\Sigma V^T} \Rightarrow [L_A]_{\beta}^{\gamma} = U^{-1}AV = \Sigma$

Matrix representation of  $L_A$  is diagonal!  
 $L_A(\vec{v}_i) = \sigma_i \vec{u}_i$

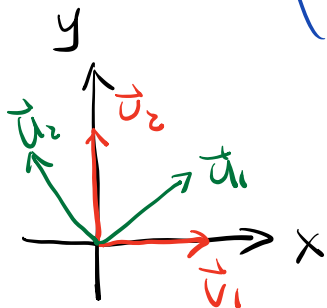
Remark: ① Diagonalization  $A = VDV^{-1}$  is not always possible

② SVD  $A = U\Sigma V^T$  is always true!

Example: Rotation by  $\theta = 30^\circ$

$$A = \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix} = U\Sigma V^T = \begin{pmatrix} \vec{u}_1 & \vec{u}_2 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} \vec{v}_1^T \\ \vec{v}_2^T \end{pmatrix}$$

$LA: \mathbb{R}^2 \rightarrow \mathbb{R}^2$   
 $\begin{pmatrix} x \\ y \end{pmatrix} \mapsto A \begin{pmatrix} x \\ y \end{pmatrix} \begin{cases} LA(\vec{v}_1) = \vec{u}_1 \\ LA(\vec{v}_2) = \vec{u}_2 \end{cases}$



$$AA^T = \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix} \begin{pmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = A^T A$$

For this case, we can pick any orthonormal basis  $\beta = \{\vec{v}_1, \vec{v}_2\}$

and define

$$\begin{cases} \vec{u}_1 = A\vec{v}_1 \\ \vec{u}_2 = A\vec{v}_2 \\ \gamma = \{\vec{u}_1, \vec{u}_2\} \end{cases}$$

s.t.  $[LA]_{\beta}^{\gamma} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$