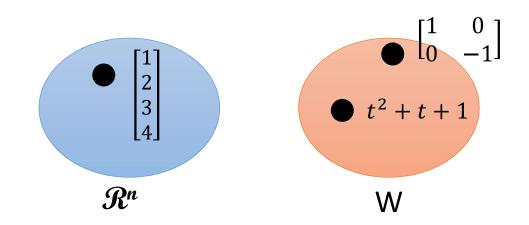
### Chapter 7 Vector Space

(除了標註※之簡報外,其餘採用李宏毅教授之投影片教材)

### Vector Spaces and Their Subspaces (Chap. 7.1)

#### Introduction

- Many things can be considered as "vectors".
  - E.g. a function can be regarded as a vector
- We can apply the concept we learned on those "vectors".
  - Linear combination
  - Span
  - Basis
  - Orthogonal .....



### (Abstract) Vector Space

For any vectors  $\mathbf{u}$ ,  $\mathbf{v}$  and  $\mathbf{w}$  in  $\mathcal{V}$ , and scalars a and b in  $\mathcal{R}$ ,  $\mathbf{u} + \mathbf{v}$  and a $\mathbf{u}$  are in  $\mathcal{V}$ , and the following axioms hold

- **1.** u + v = v + u
- 2. (u + v) + w = u + (v + w)
- 3. There is an element **0** in  $\mathcal{V}$  such that **0** + **u** = **u**
- 4. There is an element  $-\mathbf{u}$  in  $\mathcal{V}$  such that  $-\mathbf{u} + \mathbf{u} = \mathbf{0}$
- 5. 1**u** = **u**
- 6. (ab)u = a(bu)  $0 = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$  zero vector
- 7. a(**u**+**v**) = a**u** + a**v**
- 8. (a+b)**u** = a**u** + b**u**

#### Why 0u = 0 and (-1)u = -u?

- Can you prove that  $\mathbf{0}\mathbf{u} = \mathbf{0}$  (i.e., zero vector)? • **Ou** 
  - (from (3))• = 0 + 0u
  - = (-0u + 0u) + 0u (from (4))
  - = -0u + (0u + 0u) (from (2))
  - = -0u + ((0+0)u) (from (8))
  - = -0u + 0u
  - = 0

- $(0+0=0 \text{ as } 0 \text{ is in } \mathcal{R})$ (from (4))
- Can you prove that (-1)u = -u (i.e., inverse of u)?

Are they vectors?

# Are they vectors?

• A matrix

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \implies \begin{bmatrix} 2 \\ 3 \\ 4 \end{bmatrix}$$

- A linear transform
- A polynomial

$$p(x) = a_0 + a_1 x + \dots + a_n x^n$$

$$\begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{bmatrix}$$

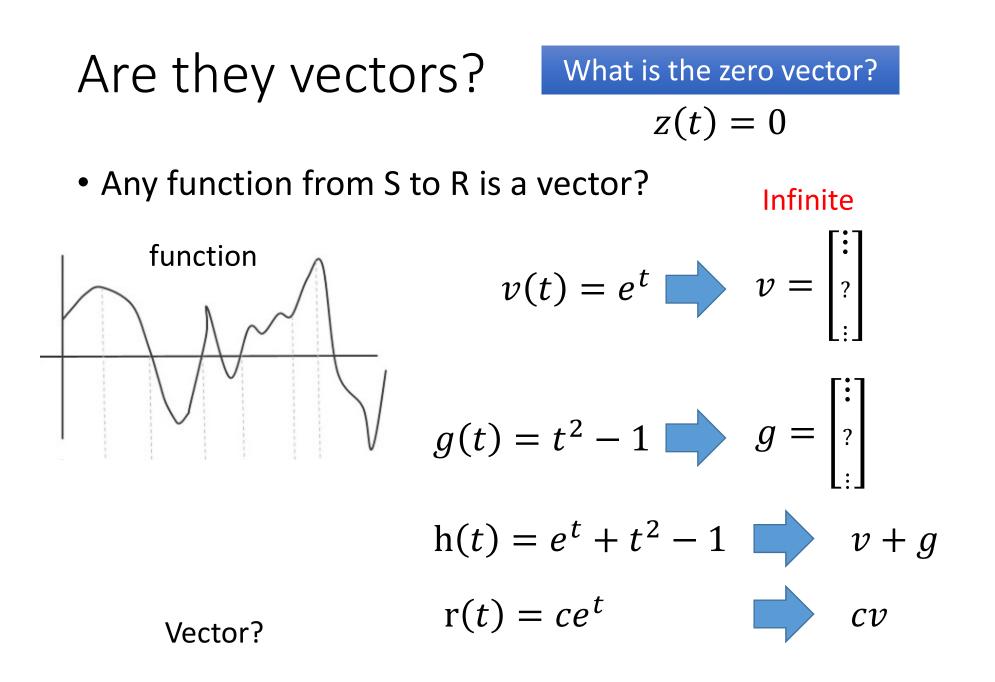
[1

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

Choose a

 $1, x, \cdots, x^n$ 

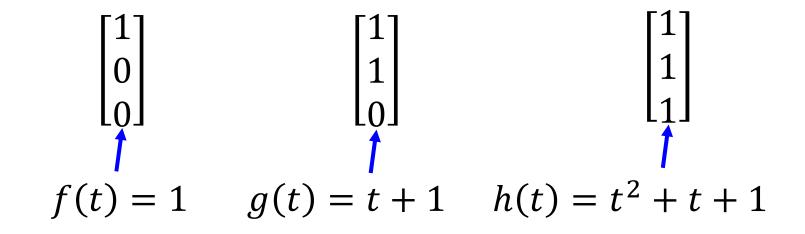
basis



#### **Objects in Different Vector Spaces**

All the polynomials with degree less than or equal to 2 form a vector space (often denoted as  $P_2$ )

w.r.t. Basis  $\{1, t, t^2\}$ 



### Subspaces

#### Review: Subspace

- A vector set V is called a subspace if it has the following three properties:
- 1. The zero vector **0** belongs to V
- 2. If **u** and **w** belong to V, then **u+w** belongs to V

Closed under (vector) addition

 3. If u belongs to V, and c is a scalar, then cu belongs to V
 Closed under scalar multiplication

#### Are they subspaces?

• All the functions pass 0 at t<sub>0</sub>

- $trace\left(\begin{bmatrix}a & b\\c & d\end{bmatrix}\right) = a + d$
- All the matrices whose trace equal to zero
- All the matrices of the form

$$\begin{bmatrix} a & a+b \\ b & 0 \end{bmatrix}$$

- All the continuous functions
- All the polynomials with degree n  $t^n, -t^n$
- All the polynomials with degree less than or equal to n

P: all polynomials,  $P_n$ : all polynomials with degree less than or equal to n

# Linear Combination and Span

#### Linear Combination and Span

• Matrices

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

Linear combination with coefficient a, b, c

$$a \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} + b \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + c \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} a & b \\ c & -a \end{bmatrix}$$

What is Span S?

All 2x2 matrices whose trace equal to zero

#### Linear Combination and Span

• Polynomials

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

Is  $f(x) = 2 + 3x - x^2$  linear combination of the "vectors" in S?

$$f(x) = 2 \cdot 1 + 3 \cdot x + (-1) \cdot x^{2}$$
  
Span{1, x, x<sup>2</sup>, x<sup>3</sup>} = P<sub>3</sub>

 $Span\{1, x, \cdots, x^n, \cdots\} = P$ 

## Linear Transformations (Chap. 7.2)

#### Linear transformation

- A mapping (function) T is called linear if for all "vectors" u, v and scalars c:
- Preserving vector addition:

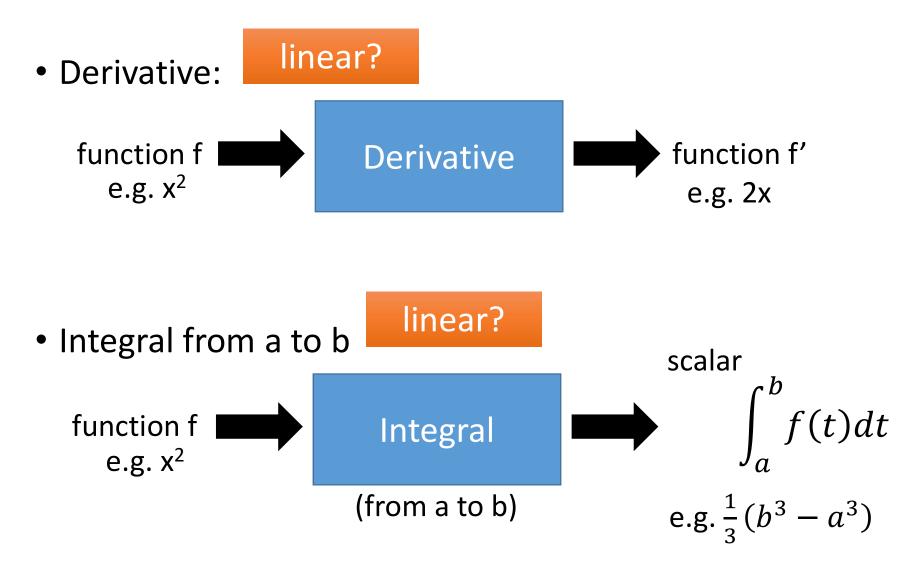
$$T(u+v) = T(u) + T(v)$$

• Preserving vector multiplication: T(cu) = cT(u)

#### Is matrix transpose linear?

Input: m x n matrices, output: n x m matrices





### Null Space and Range

- Null Space
  - The null space of T: V → W is the set of all "vectors" in V such that T(v)=0, where 0 is the zero vector in W.
  - What is the null space of matrix transpose ?

$$\mathsf{T}: \mathcal{M}_{m \times n} \to \mathcal{M}_{n \times m}$$

- Range (or Image)
  - The range of T is the set of all images of T.
  - That is, the set of all "vectors" T(v) for all v in the domain
  - What is the range of matrix transpose?



#### One-to-one and Onto

- $D: \mathscr{F}_3 \to \mathscr{F}_3$  defined by D(f) = f'
  - Is D one-to-one? no
  - Is D onto? no

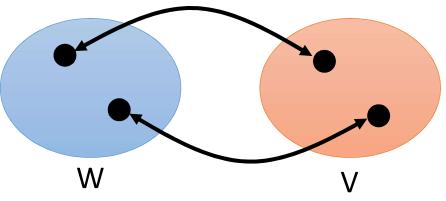
$$x^{3} + 2x + 3$$

$$3x^{2} + 2x$$

$$x^{3} + 2x$$



• Let V and W be vector space.



- A linear transformation T: V→W is called an isomorphism if it is one-to-one and onto
  - Invertible linear transform
  - W and V are isomorphic.

Example 1: U:  $\mathcal{M}_{m \times n} \to \mathcal{M}_{n \times m}$  defined by  $U(A) = A^T$ .

Example 2:  $T: \mathscr{P}_2 \rightarrow \mathscr{R}^3$ 

$$T\left(a+bx+\frac{c}{2}x^{2}\right) = \begin{bmatrix}a\\b\\c\end{bmatrix}$$

## Basis and Dimension (Chap. 7.3)

#### Independent

A basis for subspace V is a linearly independent generation set of V.

• Example

$$S = \{x^2 - 3x + 2, 3x^2 - 5x, 2x - 3\}$$
 is a subset of  $\mathscr{P}_{2}$ .

Is it linearly independent?

$$3(x^2 - 3x + 2) + (-1)(3x^2 - 5x) + 2(2x - 3) = 0$$
 No

• Example

 $S = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \right\}$ is a subset of 2x2 matrices.

Is it linearly independent?

$$a \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} + b \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + c \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$$

implies that a = b = c = 0



Independent

If  $\{v_1, v_2, ..., v_k\}$  are independent, and T is an isomorphism,  $\{T(v_1), T(v_2), ..., T(v_k)\}$  are independent

• Example

The infinite vector set  $\{1, x, x^2, \dots, x^n, \dots\}$ 

Is it linearly independent?

 $\Sigma_i c_i x^i = 0$  implies  $c_i = 0$  for all *i*. Yes

• Example

 $S = \{e^{t}, e^{2t}, e^{3t}\}$  Is it linearly independent? Yes  $ae^{t} + be^{2t} + ce^{3t} = 0 \qquad a + b + c = 0$   $ae^{t} + 2be^{2t} + 3ce^{3t} = 0 \qquad a + 2b + 3c = 0$  $ae^{t} + 4be^{2t} + 9ce^{3t} = 0 \qquad a + 4b + 9c = 0$ 

#### Independent

Theorem: If  $\{v_1, v_2, ..., v_k\}$  are independent, and T is an isomorphism,  $\{T(v_1), T(v_2), ..., T(v_k)\}$  are independent

(Proof)  
Suppose 
$$a_1 T(v_1) + a_2 T(v_2) + \dots + a_k T(v_k) = 0$$
  
 $\Rightarrow T(a_1 v_1) + T(a_2 v_2) + \dots + T(a_k v_k) = 0$   
 $\Rightarrow T(a_1 v_1 + a_2 v_2 + \dots + a_k v_k) = 0$   
 $\Rightarrow a_1 v_1 + a_2 v_2 + \dots + a_k v_k = 0$  ----- (one-to-one)  
 $\Rightarrow a_1 = a_2 = \dots = a_k = 0$ 



#### Basis

• Example

For the subspace of all 2 x 2 matrices, The basis is

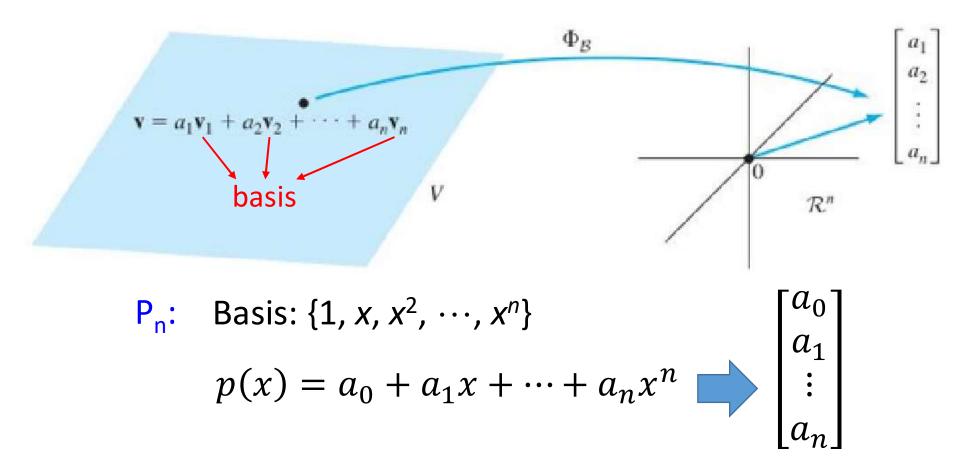
$$S = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\} \quad \mathsf{Dim} = \mathsf{4}$$

• Example

$$S = \{1, x, x^2, \dots, x^n, \dots\}$$
 is a basis of  $\mathscr{P}$ . Dim = inf

#### Vector Representation of Object

#### Coordinate Transformation



#### Coordinate Vectors

#### Definition

Let  $\mathscr{B}$  be a basis for V,  $\Phi_{\mathscr{B}} : V \to \mathscr{R}^n$  (coordinate transformation) is an isomorphism. Any vector v in V,  $\Phi_{\mathscr{B}}$  (v) is called the coordinate vector of v relative to  $\mathscr{B}$ , written as  $[v]_{\mathscr{B}}$ 

• Example

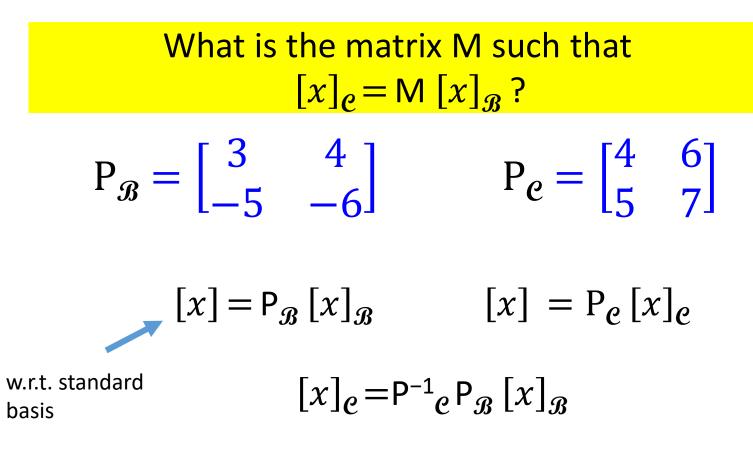
$$\mathcal{B} = \{1, x, x^2, \dots, x^n\} \text{ is a basis for } \mathcal{P}_n.$$

$$p(x) = a_0 + a_1 x + \dots + a_n x^n \qquad [p(x)]_{\mathcal{B}} = \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{bmatrix}$$



#### **Coordinate Vectors**

Let 
$$\mathcal{B} = \{ \begin{bmatrix} 3 \\ -5 \end{bmatrix}, \begin{bmatrix} 4 \\ -6 \end{bmatrix} \}$$
 and  $\mathcal{C} = \{ \begin{bmatrix} 4 \\ 5 \end{bmatrix}, \begin{bmatrix} 6 \\ 7 \end{bmatrix} \}$   
be two bases of  $\mathcal{R}^2$ 



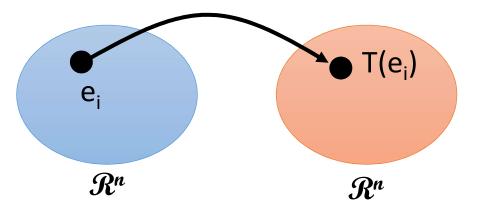


# Matrix Representations of Linear Operators (Chap. 7.4)

### Matrix Representation of Linear Operator $e_i = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix}^{i-th \ coordinate}$

Let T be a linear operator from  $\mathcal{R}^n$  to  $\mathcal{R}^n$  with standard basis  $\mathcal{E} = \{e_1, e_2, ..., e_n\}$ . Then the matrix representation of T w.r.t.  $\mathcal{E}$  is

 $[\mathsf{T}]_{\mathcal{E}} = [ [\mathsf{T}(\mathsf{e}_1)_{\mathcal{E}} \ \mathsf{T}(\mathsf{e}_2)_{\mathcal{E}} \ \dots \ \mathsf{T}(\mathsf{e}_n)_{\mathcal{E}} ]$ 





#### Linear Transformation and Matrix

$$T(\mathbf{x}) = T\left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} 3x_1 + x_3 \\ x_1 + x_2 \\ -x_1 - x_2 + 3x_3 \end{bmatrix}$$

$$Ax = \begin{bmatrix} 3 & 0 & 1 \\ 1 & 1 & 0 \\ -1 & -1 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$



# Matrix Representation of Linear Operator

Let T be a linear operator on vector space V with basis  $\mathcal{B} = \{v_1, v_2, ..., v_n\}$ . Then the matrix representation of T w.r.t.  $\mathcal{B}$  is

$$[\mathsf{T}]_{\mathscr{B}} = [ [\mathsf{T}(\mathsf{v}_1)_{\mathscr{B}} \ \mathsf{T}(\mathsf{v}_2)_{\mathscr{B}} \ \dots \ \mathsf{T}(\mathsf{v}_n)_{\mathscr{B}} ]$$

Example:  $\mathcal{B} = \{1, x\}.$ 

 $\mathbf{X}$ 

$$T(1) = 1 + 2x \quad T(x) = 3x \qquad [T]_{\mathscr{B}} = \begin{bmatrix} 1 & 0 \\ 2 & 3 \end{bmatrix}$$

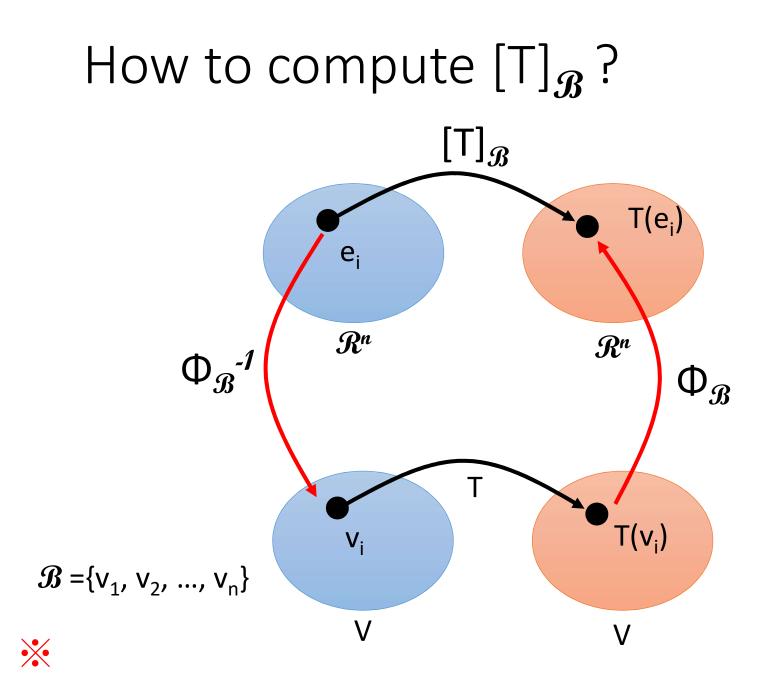
 $T(2-x) = 2(1+2x) - (3x) = 2+x \quad [T(2-x)]_{\mathcal{B}} = \begin{bmatrix} 1 & 0 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 2 \\ -1 \end{bmatrix}$ 

#### Linear Operator T: $V \rightarrow V$ $[\mathsf{T}]_{\mathscr{B}}$ $[T(x)]_{\mathcal{B}}$ $[x]_{\mathcal{B}}$ $\mathcal{R}^n$ Rn $\Phi_{\mathscr{B}}$ $\Phi_{\mathscr{B}}$ T(x) Χ V V $\mathcal{B} = \{v_1, v_2, ..., v_n\}$

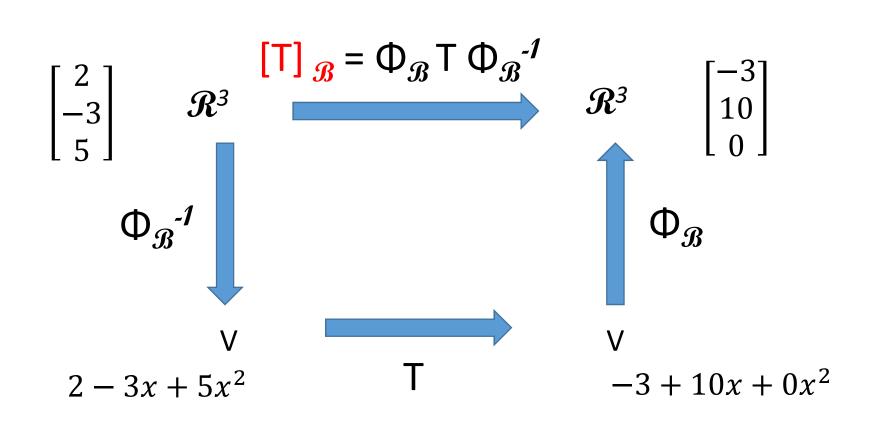
What is the matrix for T?

 $[\mathsf{T}]_{\mathscr{B}} = [ [\mathsf{T}(\mathsf{v}_1)_{\mathscr{B}} \ \mathsf{T}(\mathsf{v}_2)_{\mathscr{B}} \ \dots \ \mathsf{T}(\mathsf{v}_n)_{\mathscr{B}} ]$ 

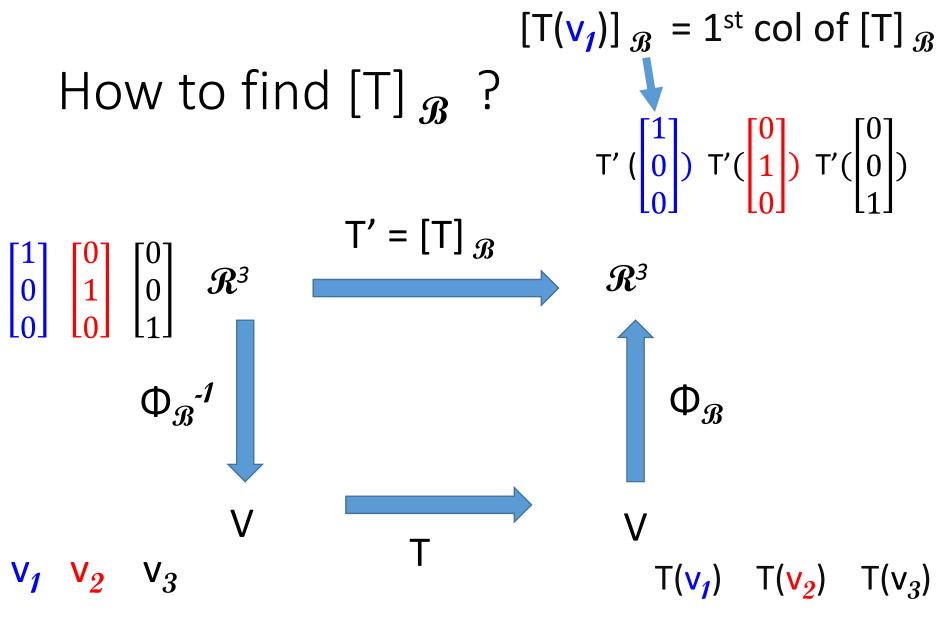


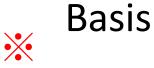


Matrix Representation<br/>of Linear OperatorRepresent it as a matrix $\mathscr{B} = \{1, x, x^2\}$  $\Phi_{\mathscr{B}}(2 - 3x + 5x^2) = (2, -3, 5)^T$ 







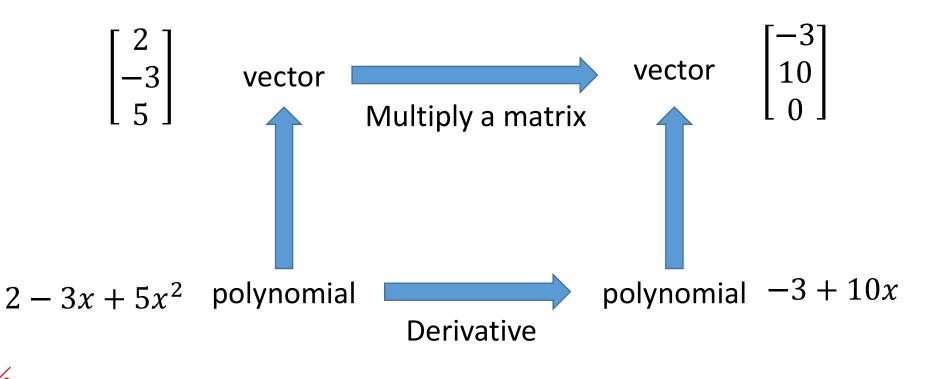


Example  
Let 
$$T: \mathscr{P}_2 \Rightarrow \mathscr{P}_2$$
 defined as  
 $T(p(x)) = p(0) + 3p(1) x + p(2)x^2$   
 $\mathscr{B} = \{1, x, x^2\}$  is a basis for  $\mathscr{P}_2$ .  
 $T(1) = 1 + 3x + x^2 \Rightarrow [T(1)]_{\mathscr{B}} = (1, 3, 1)^T$   
 $T(x) = 0 + 3x + 2x^2 \Rightarrow [T(x)]_{\mathscr{B}} = (0, 3, 2)^T$   
 $T(x^2) = 0 + 3x + 4x^2 \Rightarrow [T(x)]_{\mathscr{B}} = (0, 3, 4)^T$   
 $[T]_{\mathscr{B}} = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 3 & 3 \\ 1 & 2 & 4 \end{bmatrix}$ 

\*

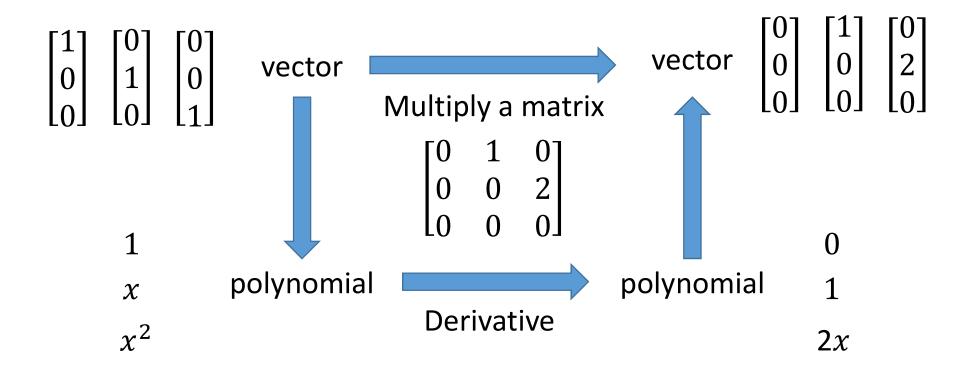
- Example:
  - D (derivative):  $P_2 \rightarrow P_2$

Represent it as a matrix



- Example:
  - D (derivative):  $P_2 \rightarrow P_2$

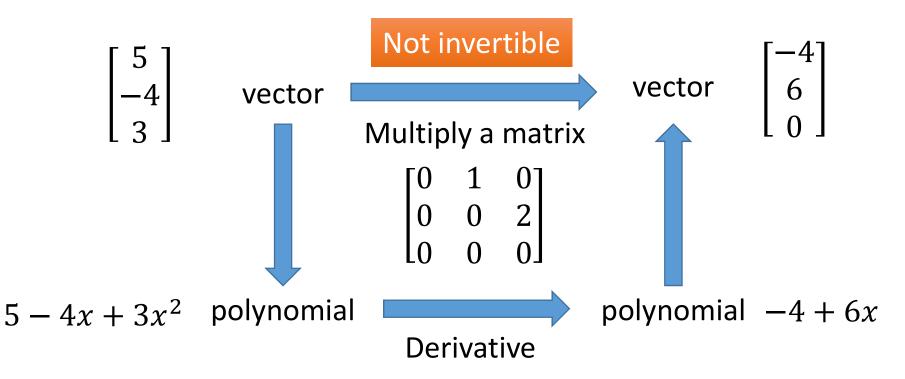
Represent it as a matrix



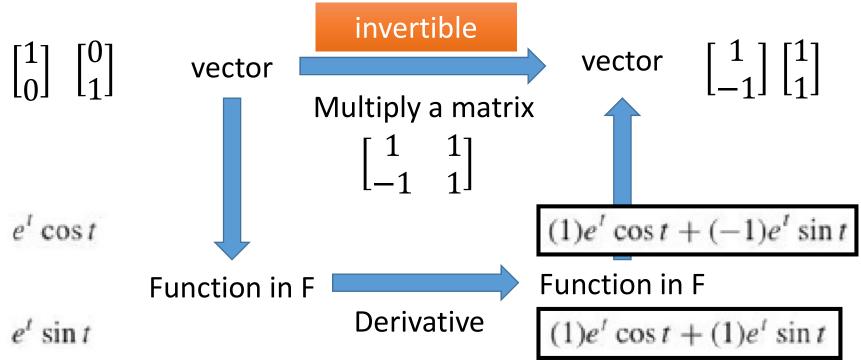
٥٦	1	[0	[5]
0	0	2	-4
LO	0	0]	[ 3 ]

- Example:
  - D (derivative):  $P_2 \rightarrow P_2$

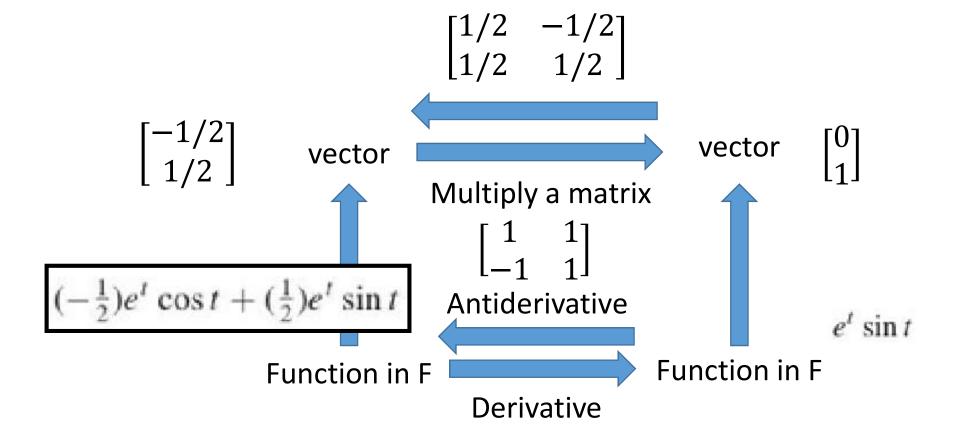




- Example:
  - D (derivative): Function set F  $\rightarrow$  Function set F
  - Basis of F is  $\{e^t \cos t, e^t \sin t\}$



## Matrix RepresentationBasis of F isof Linear Operator $\{e^t \cos t, e^t \sin t\}$

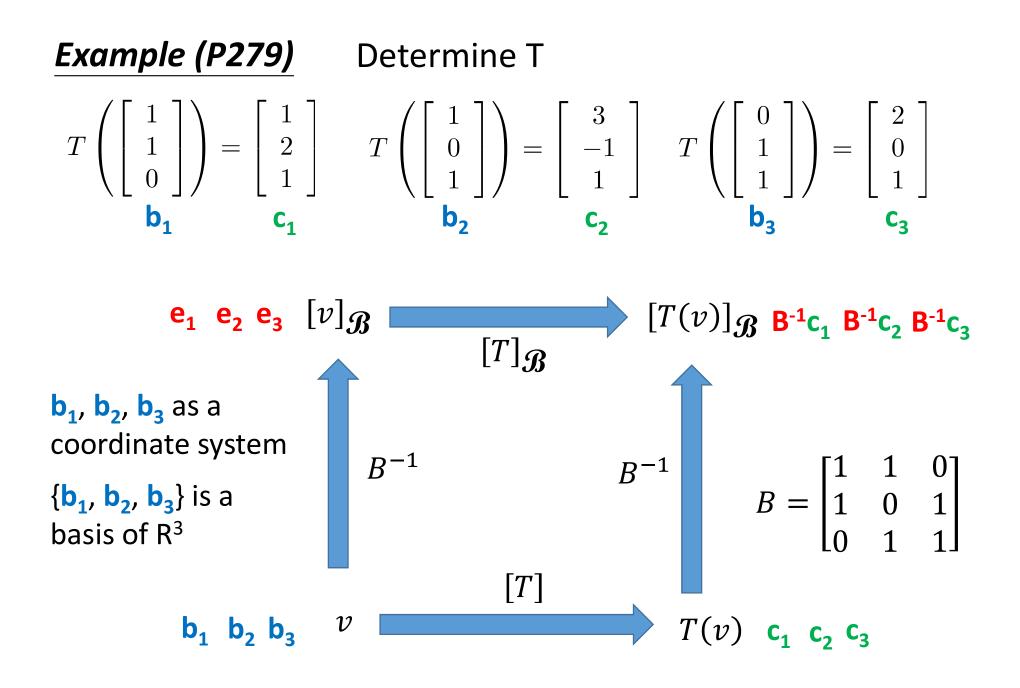


## Linear operator between two bases

Let  $T: \mathcal{V} \to \mathcal{W}$  be a linear operator between  $\mathcal{V}$  and  $\mathcal{W}$ .  $\mathcal{B} = \{b_1, b_2, \dots, b_n\}$  and  $\mathcal{B}' = \{b'_1, b'_2, \dots, b'_n\}$  bases of  $\mathcal{V}$  and  $\mathcal{W}$ .

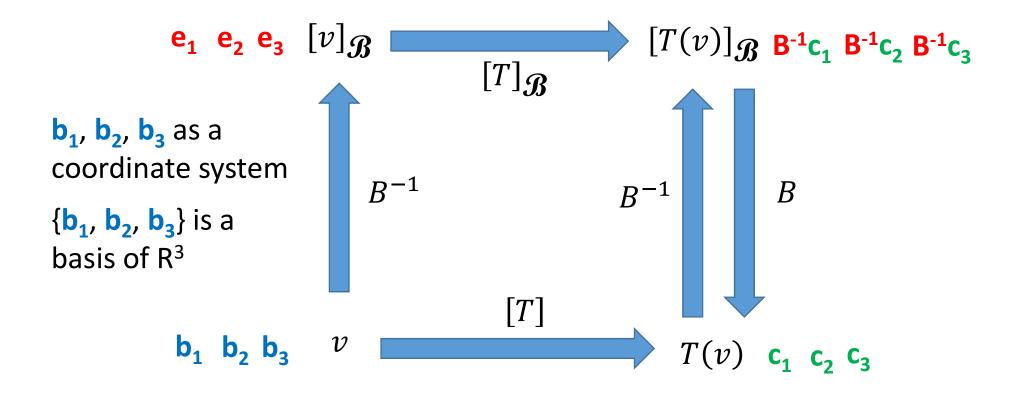
What is matrix representation  $\mathcal{M}$  of T w.r.t.  $\mathcal{B}$  and  $\mathcal{B}'$  i.e.,  $[T(x)]_{\mathcal{B}'} = \mathcal{M}[x]_{\mathcal{B}}$ ?

### $\mathcal{M} = [\mathsf{T}[\mathsf{b}_1]_{\mathcal{B}'}, \mathsf{T}[\mathsf{b}_2]_{\mathcal{B}'}, \dots, \mathsf{T}[\mathsf{b}_n]_{\mathcal{B}'}]$



#### *Example (P279)* Determine T

$$[T]_{\mathcal{B}} = \begin{bmatrix} B^{-1}c_1 & B^{-1}c_2 & B^{-1}c_3 \end{bmatrix} = B^{-1}C$$
$$[T] = B[T]_{\mathcal{B}}B^{-1} = BB^{-1}CB^{-1} = CB^{-1}$$



## Eigenvalue and Eigenvector

#### $T(\boldsymbol{v}) = \lambda \boldsymbol{v}, \, \boldsymbol{v} \neq \boldsymbol{0}, \, \boldsymbol{v}$ is eigenvector, $\lambda$ is eigenvalue

Chap. 7.4

### Eigenvalue and Eigenvector

Consider derivative (linear transformation, input & output are functions)
 Is e<sup>at</sup> an "eigenvector"? ae<sup>at</sup> What is the "eigenvalue"? a

Every scalar is an eigenvalue of derivative.

- Consider Transpose (also linear transformation, input & output are matrices)
- Is  $\lambda = 1$  an eigenvalue?

Symmetric matrices form the eigenspace

Is  $\lambda = -1$  an eigenvalue?

Skew-symmetric matrices form the eigenspace.

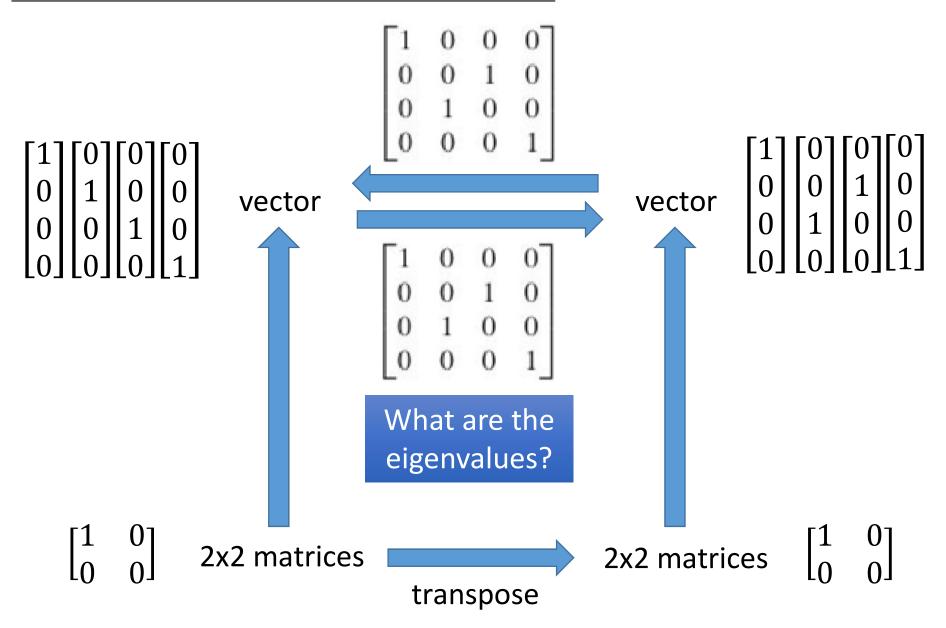
Symmetric:

 $A^T = A$ 

Skew-symmetric:

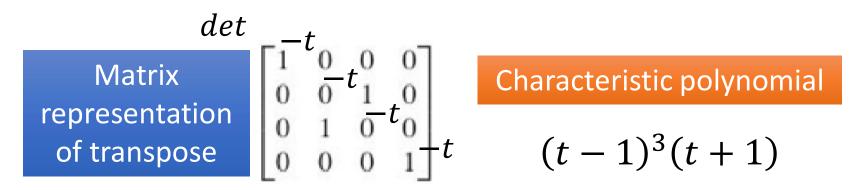
 $A^T = -A$ 

#### Consider Transpose of 2x2 matrices



### Eigenvalue and Eigenvector

#### Consider Transpose of 2x2 matrices



 $\lambda = 1$ 

 $\lambda = -1$ 

Symmetric matrices

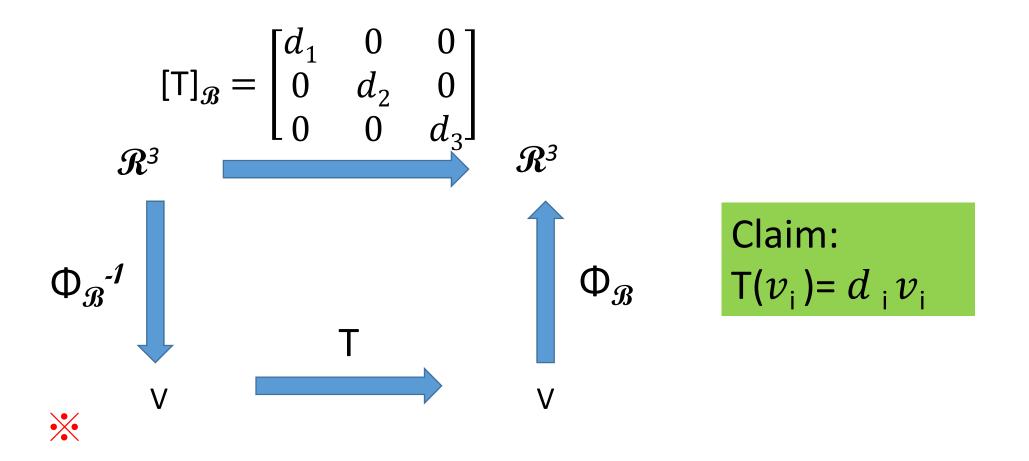
$$\begin{bmatrix} a & b \\ b & c \end{bmatrix} \quad \text{Dim=3}$$

Skew-symmetric matrices

$$\begin{bmatrix} 0 & a \\ -a & 0 \end{bmatrix}$$



Diagonalizable Linear Operator T is diagoalizable if there is a basis  $\mathcal{B} = \{v_1, v_2, \dots, v_n\}$  s.t.  $[T]_{\mathcal{B}}$  is diagonal



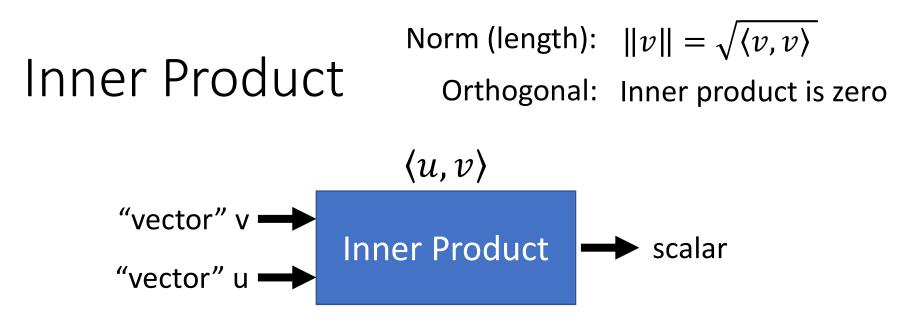
#### Diagonalizable Linear Operator

The following statements are equivalent

- T is diagoalizable
- There is a basis  $\mathcal{B} = \{v_1, v_2, \dots, v_n\}$  s.t.  $[T]_{\mathcal{B}}$  is diagonal
- T has a basis consisting of eigenvectors
- For every basis  $\mathcal{B}$ ,  $[T]_{\mathcal{B}}$  is diagoalizable



### Inner Product Spaces (Chap. 7.5)

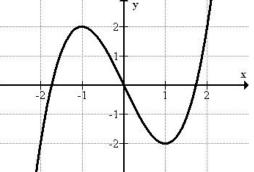


For any vectors u, v and w, and any scalar a, the following axioms hold:

1.  $\langle u, u \rangle > 0$  if  $u \neq 0$ 3.  $\langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$ 2.  $\langle u, v \rangle = \langle v, u \rangle$ 4.  $\langle au, v \rangle = a \langle u, v \rangle$ Dot product is a special case of inner product $c(u \cdot v)$ c > 0Can you define other inner product for normal vectors?

- 1.  $\langle u, u \rangle > 0$  if  $u \neq 0$ 2.  $\langle u, v \rangle = \langle v, u \rangle$ 3.  $\langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$ 4.  $\langle au, v \rangle = a \langle u, v \rangle$
- Continuous functions in C[-1, 1]

$$\langle g,h\rangle = \int_{-1}^{1} g(x)h(x)dx$$



Axiom (1):  

$$f \neq 0 \Rightarrow f^2(c) > 0$$
, for some c in  $[-1, 1]$ .  
 $\Rightarrow f^2(t) > d > 0$ , t in  $[c - \varepsilon, c + \varepsilon]$ .  
 $\Rightarrow \langle f, f \rangle = \int_{-1}^{1} f^2(t) dt = f^2(t) > 2\varepsilon d > 0$ 

Axioms (2 - 4): Easy to check

$$1. \langle u, u \rangle > 0 \text{ if } u \neq 0$$

$$2. \langle u, v \rangle = \langle v, u \rangle$$

$$3. \langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$$

$$4. \langle au, v \rangle = a \langle u, v \rangle$$

• Inner product for any function with input [-1, 1]

$$\langle g,h \rangle = \int_{-1}^{1} g(x)h(x)dx = \int_{-1}^{1} xdx = 0$$
   
 
$$\begin{aligned} |s g(x) = 1 \text{ and} \\ h(x) = x \text{ orthogonal}? \end{aligned}$$
 yes

$$\langle g,h \rangle = \sum_{i=-10}^{10} g\left(\frac{i}{10}\right) h\left(\frac{i}{10}\right)$$
 Can it be inner product for general functions?

$$u\left(\frac{i}{10}\right) = 0$$
, otherwise  $\neq 0$   $\langle u, u \rangle = 0$ , but  $u \neq 0$ 

#### Inner Product

• Inner Product of Matrix

Frobenius inner product

$$\langle A, B \rangle = trace(AB^T)$$
  
=  $trace(BA^T)$ 

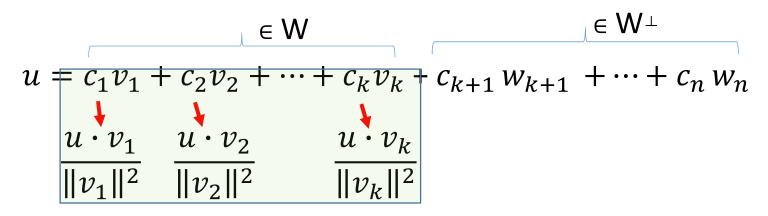
Check Axioms (1-4)

$$\begin{pmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix} \end{pmatrix} = 1 \cdot 5 + 2 \cdot 6 + 3 \cdot 7 + 4 \cdot 8 = 70$$
  
Element-wise multiplication

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad ||A|| = \sqrt{1^2 + 2^2 + 3^2 + 4^2}$$

### Orthogonal Projection

Consider a subspace W with orthogonal basis S =
 {v<sub>1</sub>, v<sub>2</sub>, …, v<sub>k</sub>} of an *n*-dim vector space V. Let u be any
 vector in V, which can be written as



orthogonal projection of u on W

• If S an orthonormal basis of W.

$$w = c_1 v_1 + c_2 v_2 + \dots + c_k v_k$$
$$u \cdot v_1 \quad u \cdot v_2 \qquad u \cdot v_k$$

 $\mathbf{X}$ 

### Orthogonal Projection

Consider a subspace W with orthonormal basis S = {v<sub>1</sub>, v<sub>2</sub>, …, v<sub>k</sub>} of an *n*-dim vector space V. The orthogonal projection of u on W can be written as

$$(u \cdot v_1)v_1 + (u \cdot v_2)v_2 + \dots + (u \cdot k)v_k$$

$$= v_1(v_1 \cdot u) + v_2(v_2 \cdot u) + \dots + v_k(v_k \cdot u)$$

$$= v_1(v_1^T u) + v_2(v_2^T u) + \dots + v_k(v_k^T u)$$

$$= (v_1v_1^T + v_2v_2^T + \dots + v_kv_k^T)u$$

$$= (v_1v_1^T + v_2v_2^T + \dots + v_kv_k^T)u$$
Recall that is *S* is an arbitrary basis, the projection matrix is  $C(C^TC)^{-1}C^T$ 

#### **Orthogonal Basis**

Let  $\{u_1, u_2, \dots, u_k\}$  be a basis of a subspace V. How to transform  $\{u_1, u_2, \dots, u_k\}$  into an orthogonal basis  $\{v_1, v_2, \dots, v_k\}$ ?



Then  $\{v_1, v_2, \dots, v_k\}$  is an orthogonal basis for W

After normalization, you can get orthonormal basis.

#### Orthogonal/Orthonormal Basis

- Find orthogonal/orthonormal basis for P<sub>2</sub>
  - Define an inner product of P<sub>2</sub> by

$$\langle f(x), g(x) \rangle = \int_{-1}^{1} f(t)g(t) dt$$
  
Find a basis {1, x, x<sup>2</sup>}  $\longrightarrow v_1, v_2, v_3$   
 $\mathbf{v}_1 = \mathbf{u}_1$   
 $\mathbf{v}_2 = \mathbf{u}_2 - \frac{\langle \mathbf{u}_2, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2} \mathbf{v}$   
 $\mathbf{v}_3 = \mathbf{u}_3 - \frac{\langle \mathbf{u}_3, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 - \frac{\langle \mathbf{u}_3, \mathbf{v}_2 \rangle}{\|\mathbf{v}_2\|^2} \mathbf{v}_2 = x^2 - \frac{1}{3}$ 

- 1

#### Orthogonal/Orthonormal Basis

- Find orthogonal/orthonormal basis for P<sub>2</sub>
  - Define an inner product of P<sub>2</sub> by

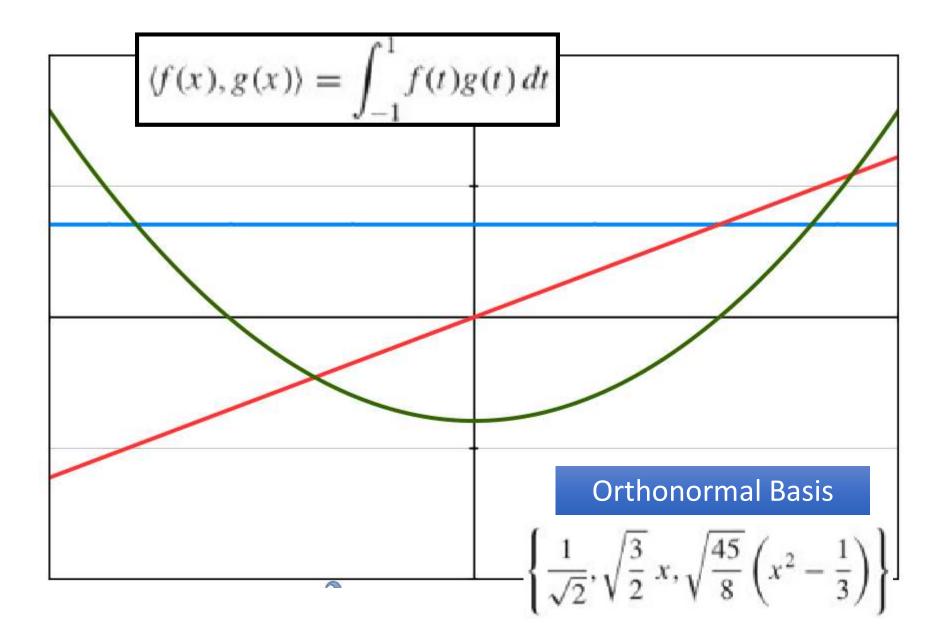
$$\langle f(x), g(x) \rangle = \int_{-1}^{1} f(t)g(t) dt$$

• Get an orthogonal basis  $\{1, x, x^2-1/3\}$ 

$$\|\mathbf{v}_1\| = \sqrt{\int_{-1}^{1} 1^2 \, dx} = \sqrt{2} \qquad \|\mathbf{v}_2\| = \sqrt{\int_{-1}^{1} x^2 \, dx} = \sqrt{\frac{2}{3}}$$

Orthonormal Basis

$$\|\mathbf{v}_3\| = \sqrt{\int_{-1}^1 \left(x^2 - \frac{1}{3}\right)^2} \, dx = \sqrt{\frac{8}{45}} \qquad \left\{\frac{1}{\sqrt{2}}, \sqrt{\frac{3}{2}} \, x, \sqrt{\frac{45}{8}} \left(x^2 - \frac{1}{3}\right)\right\}$$



Least square approx. of  $x^3$ ?  $\left\{\frac{1}{\sqrt{2}}, \sqrt{\frac{3}{2}} x, \sqrt{\frac{45}{8}} \left(x^2 - \frac{1}{3}\right)\right\}$  Orthonormal Basis Compute orthogonal projection of  $f = x^3$  on P<sub>2</sub> Orthogonal projection is  $\langle f, v_1 \rangle v_1 + \langle f, v_2 \rangle v_2 + \langle f, v_3 \rangle v_3$ Hence, the l.s.a. of  $x^3$  is 2/5 x  $\langle f, v_1 \rangle = \int_{-1}^{1} 1/\sqrt{2} t^3 dt = 0$  $\langle f, v_2 \rangle = \int_{-1}^1 \sqrt{3/2} t^4 dt = (2/5)\sqrt{3/2}$  $\langle f, v_3 \rangle = \int_{-1}^{1} \sqrt{45/8} (t^2 - 1/3) t^3 dt = 0$