

# Review

## Definition

An *eigenvector* of an  $n \times n$  matrix  $A$  is a nonzero vector  $\mathbf{x} \in \mathbb{R}^n$  (or in  $\mathbb{C}^n$ ) such that  $A\mathbf{x} = \lambda\mathbf{x}$  for some  $\lambda \in \mathbb{R}$  (or  $\lambda \in \mathbb{C}$ ). A number  $\lambda$  is called an *eigenvalue* of  $A$  if there is a nontrivial solution  $\mathbf{x}$  of  $A\mathbf{x} = \lambda\mathbf{x}$ ; such an  $\mathbf{x}$  is called an *eigenvector corresponding to  $\lambda$* .

## Theorem

$\lambda$  is an eigenvalue of  $A$  if and only if  $\det(A - \lambda I) = 0$ .

## Definition

$p(\lambda) = \det(A - \lambda I)$  is called the *characteristic polynomial* of  $A$ .  
The equation

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

is called the *characteristic equation* of  $A$ .

# Complex Eigenvalues

of  $2 \times 2$   
matrices.

## Definition

Let  $A$  be an  $n \times n$  matrix. A number  $\lambda \in \mathbb{C}$  is called a *complex eigenvalue* of  $A$  if there exists a nonzero vector  $\mathbf{x} \in \mathbb{C}^n$  such that  $A\mathbf{x} = \lambda\mathbf{x}$ .

## → Abuse of Notation

Every real number is a complex number. However, we often use the word “complex eigenvalue” to refer to an eigenvalue which is in  $\mathbb{C} \setminus \mathbb{R}$ .

I will however, try to explicitly mention that an eigenvalue is of the form  $a + bi$  where  $b \neq 0$ , when dealing with such eigenvalues.

$$\begin{vmatrix} \cos \theta - \lambda & -\sin \theta \\ \sin \theta & \cos \theta - \lambda \end{vmatrix} = (\cos \theta - \lambda)^2 + \sin^2 \theta$$

The eigenvalues of a (non-trivial)  $2 \times 2$  rotation matrix

$$\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

are complex.

Any  $2 \times 2$  matrix which has complex eigenvalues is similar to the product of a scaling and a rotation.

$$p(\lambda) = \lambda^2 - 2\cos \theta \lambda + 1$$

$$\text{roots: } \frac{2\cos \theta \pm \sqrt{4\cos^2 \theta - 4}}{2} = \cos \theta \pm i \sin \theta$$

$$\begin{aligned} & \cos^2 \theta + \lambda^2 \\ & - 2\cos \theta \lambda \\ & + \sin^2 \theta \\ & = \lambda^2 - 2\cos \theta \lambda + 1 \end{aligned}$$

$$e^{\pm i\theta}$$

$$re^{i\theta} = r\cos\theta + i(r\sin\theta)$$

$$re^{\pm i\theta}$$

$\mathbb{C}$

$A:$

$2 \times 2$

matrix

$$\mathbb{C} \xrightarrow{f} \mathbb{C}$$

$T$

$$x \mapsto Ax$$

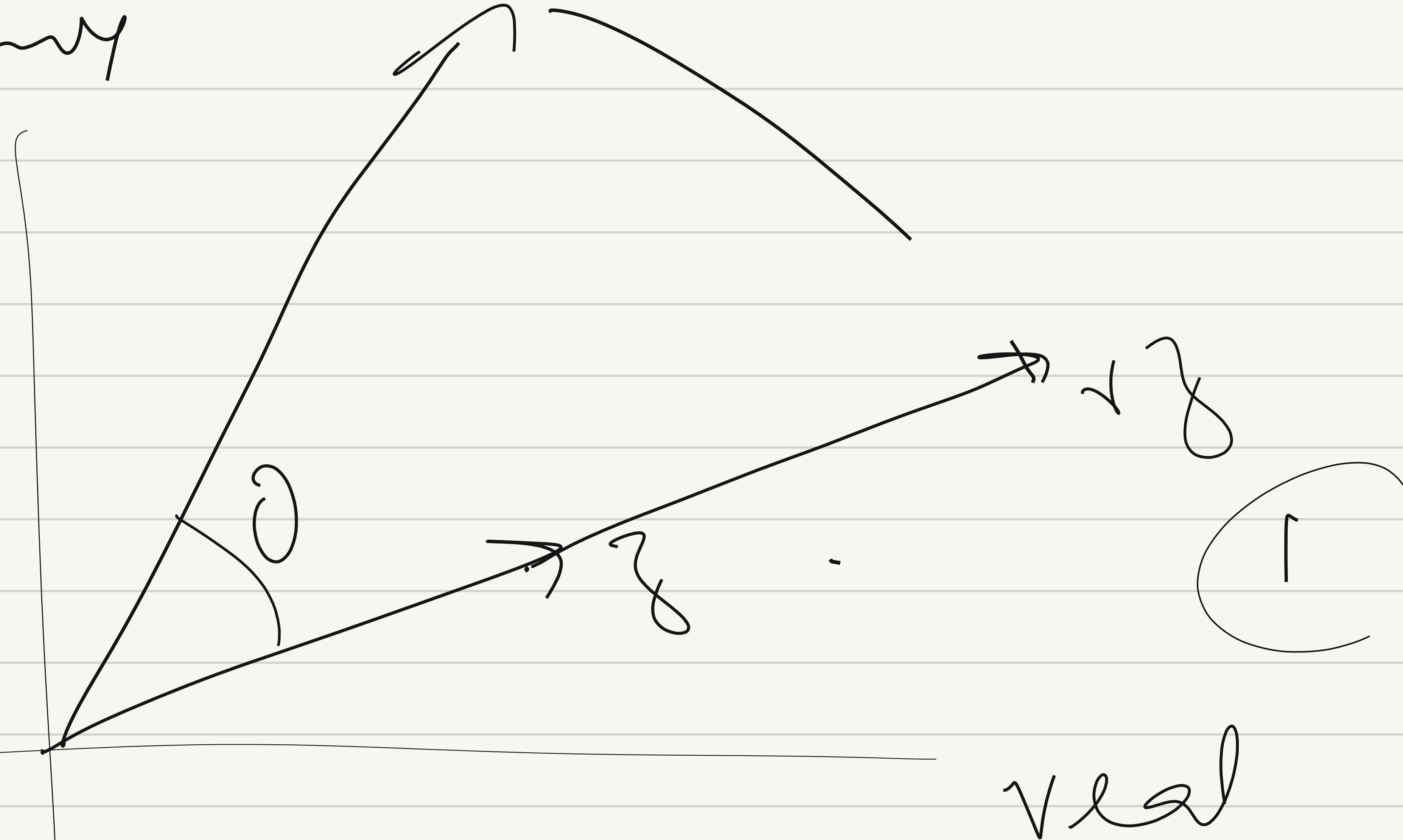
Fix

$$\lambda \in \mathbb{C}$$

$$z \mapsto \lambda z$$

$$f(z) = \lambda z$$

imaginary



For any

$$z \in \mathbb{C}$$

$$e^{i\theta} z$$

$$f(z) = \lambda z$$

$$\lambda = r e^{i\theta}$$

$$|\lambda| = r$$

$$\arg \lambda = \theta$$

$$\mathbb{R}^2 \rightarrow \mathbb{R}^2$$

$$X \mapsto AX$$

## Theorem

If  $A$  is a  $2 \times 2$  matrix and  $\lambda = re^{-i\theta}$  is a complex eigenvalue of  $A$ , and  $\mathbf{v} \in \mathbb{C}^2$  is a complex eigenvector of  $A$  corresponding to  $\lambda$ , then

$$P^{-1}AP = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} r & 0 \\ 0 & r \end{bmatrix},$$

where  $P = [\operatorname{Re}(\mathbf{v}) \quad \operatorname{Im}(\mathbf{v})]$ .

Any complex number represents a scaling and a rotation. So we extract that information from the eigenvalues. Because of some strange coincidence which I don't know how to explain, the same scaling and rotation happen if we look at the coordinate system determined by the real and imaginary parts of the complex eigenvectors.   
 but opposite (negative of  $\arg \lambda$ )

$$\begin{bmatrix} a + bi \\ c + di \end{bmatrix} = \begin{bmatrix} a \\ c \end{bmatrix} + i \begin{bmatrix} b \\ d \end{bmatrix}$$

$$= \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 1 \\ i \end{bmatrix}$$

Proof  
 Let  $v$  be eigenvector of  $A$   
 corresponding to  $\lambda = \underline{re^{-i\theta}}$ .

$$\therefore Av = \lambda v$$

$$\begin{aligned} \lambda v &= AP \begin{bmatrix} 1 \\ i \end{bmatrix} \\ &= A \begin{bmatrix} \operatorname{Re}(v) & \operatorname{Im}(v) \end{bmatrix} \begin{bmatrix} 1 \\ i \end{bmatrix} \end{aligned}$$

- (1)

$$v = \operatorname{Re}(v) + i \operatorname{Im}(v)$$

$$v = \begin{bmatrix} \operatorname{Re}(v) & \operatorname{Im}(v) \end{bmatrix} \begin{bmatrix} 1 \\ i \end{bmatrix}$$

$$v = P \begin{bmatrix} 1 \\ i \end{bmatrix}$$


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$$\lambda v = \begin{bmatrix} \operatorname{Re}(v) & \operatorname{Im}(v) \end{bmatrix} \begin{bmatrix} 1 \\ i \end{bmatrix} \lambda$$



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

$$\Rightarrow \overline{A v} = \overline{\lambda} \overline{v}$$

$$\Rightarrow A P \begin{bmatrix} 1 \\ -i \end{bmatrix} = A \begin{bmatrix} \operatorname{Re}(v) & \operatorname{Im}(v) \end{bmatrix} \begin{bmatrix} 1 \\ -i \end{bmatrix}$$

$$= \overline{\lambda} \overline{v}$$

②

$$\overline{v} = \overline{\operatorname{Re}(v)}$$

$$- i \operatorname{Im}(v)$$

$$= \begin{bmatrix} \operatorname{Re}(v) & \operatorname{Im}(v) \end{bmatrix} \begin{bmatrix} 1 \\ -i \end{bmatrix}$$

$$\overline{\lambda} \overline{v} = \begin{bmatrix} \operatorname{Re}(v) & \operatorname{Im}(v) \end{bmatrix} \begin{bmatrix} 1 \\ -i \end{bmatrix}$$

$\leftarrow$   $\text{arr}$       (1)       $\rightarrow$       (2)

$$AP \begin{bmatrix} 1 & 1 \\ i & -i \end{bmatrix} = A \begin{bmatrix} \operatorname{Re}(v) & \operatorname{Im}(v) \end{bmatrix} \begin{bmatrix} 1 & 1 \\ i & -i \end{bmatrix}$$

$$= \begin{bmatrix} \operatorname{Re}(v) & \operatorname{Im}(v) \end{bmatrix} \begin{bmatrix} \lambda & \lambda \\ \lambda i & -\lambda i \end{bmatrix}$$

$$= \begin{bmatrix} \begin{bmatrix} \lambda & \lambda \\ \lambda i & -\lambda i \end{bmatrix} \end{bmatrix}$$

$$AP = P \begin{bmatrix} \lambda & \bar{\lambda} \\ x_i & -\bar{x}_i \end{bmatrix} \begin{bmatrix} 1 & 1 \\ i & -i \end{bmatrix}$$

$$= \begin{pmatrix} 1 \\ -\frac{1}{2i} \end{pmatrix} P \begin{bmatrix} \lambda & \bar{\lambda} \\ x_i & -\bar{x}_i \end{bmatrix} \begin{bmatrix} -i & -1 \\ -i & 1 \end{bmatrix}$$

$$= \begin{pmatrix} 1 \\ -\frac{1}{2i} \end{pmatrix} P \begin{bmatrix} i(\lambda + \bar{\lambda}) & \lambda - \bar{\lambda} \\ \lambda - \bar{\lambda} & -\lambda_i - \bar{\lambda}_i \end{bmatrix}$$

$$= P \begin{bmatrix} \operatorname{Re}(\lambda) & \operatorname{Im}(\lambda) \\ -\operatorname{Im}(\lambda) & \operatorname{Re}(\lambda) \end{bmatrix}$$

In general:  
 $\lambda \rightarrow \lambda = a + bi$

$$= a + bi$$

$$- (a - bi)$$

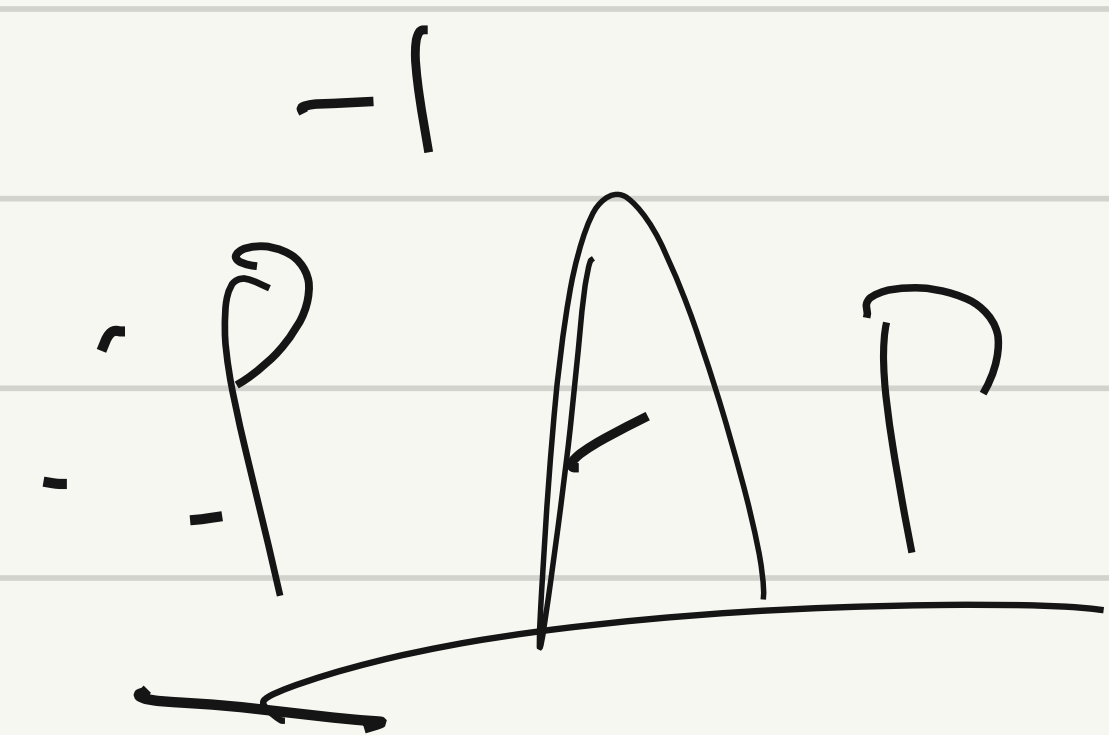
$$= 2bi$$

$$= P \begin{bmatrix} r \cos \theta & -r \sin \theta \\ r \sin \theta & r \cos \theta \end{bmatrix}$$

$$\lambda = 1e^{-i\theta}$$

$$= r \cos \theta$$

$$-i r \sin \theta$$



$$\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} r & 0 \\ 0 & r \end{bmatrix}$$

# Inner Product in $\mathbb{R}^n$

$$|\hat{u} \cdot \hat{v}| = |\cos \theta|$$

## Definition

If  $\mathbf{u} = (u_1, \dots, u_n), \mathbf{v} = (v_1, \dots, v_n) \in \mathbb{R}^n$  then the *inner product* or the *dot product* of  $\mathbf{u}$  and  $\mathbf{v}$  is the real number

$$\mathbf{u} \cdot \mathbf{v} = \sum_{i=1}^n u_i v_i$$

$$\sum_{i=1}^n u_i^2 = \mathbf{u} \cdot \mathbf{u}$$

$$\mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$$

bilinear

## Theorem

Let  $\mathbf{u}, \mathbf{v}$ , and  $\mathbf{w}$  be vectors in  $\mathbb{R}^n$ , and let  $c$  be a scalar. Then

- 1  $\mathbf{u} \cdot \mathbf{v} = \mathbf{v} \cdot \mathbf{u}$  — symmetry
- 2  $(\mathbf{u} + \mathbf{v}) \cdot \mathbf{w} = \mathbf{u} \cdot \mathbf{w} + \mathbf{v} \cdot \mathbf{w}$
- 3  $(c\mathbf{u}) \cdot \mathbf{v} = c(\mathbf{u} \cdot \mathbf{v}) = \mathbf{u} \cdot (c\mathbf{v})$  } linearity
- 4  $\mathbf{u} \cdot \mathbf{u} \geq 0$ , and  $\mathbf{u} \cdot \mathbf{u} = 0$  if and only if  $\mathbf{u} = \mathbf{0}$

verification  
is HW  
← positive definite

# Inner Product in other Vector Spaces

## Definition

Let  $V$  be a real vector space. An inner product  $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}$  is a function satisfying the following conditions:

- (1)  $\langle \mathbf{v} + \mathbf{w}, \mathbf{z} \rangle = \langle \mathbf{v}, \mathbf{z} \rangle + \langle \mathbf{w}, \mathbf{z} \rangle$ , for every  $\mathbf{v}, \mathbf{w}, \mathbf{z} \in V$  ✓
- (2)  $\langle c\mathbf{v}, \mathbf{w} \rangle = c\langle \mathbf{v}, \mathbf{w} \rangle$  for every  $c \in \mathbb{R}$  and  $\mathbf{v}, \mathbf{w} \in V$
- (3)  $\langle \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{w}, \mathbf{v} \rangle$ , for every  $\mathbf{v}, \mathbf{w} \in V$
- (4)  $\langle \mathbf{v}, \mathbf{v} \rangle \geq 0$  for every  $\mathbf{v} \in V$ , and  $\langle \mathbf{v}, \mathbf{v} \rangle = 0$  holds if and only if  $\mathbf{v} = \mathbf{0}$

### Example

Let  $V = C[a, b]$  the space of continuous real-valued functions defined on interval  $[a, b]$ . Define

$$\langle f, g \rangle := \int_a^b f(t)g(t) dt.$$

(1)  $f, g, h \in C[a, b]$

$$\langle f + g, h \rangle = \int_a^b (f(t) + g(t)) h(t) dt$$

$$= \int_a^b (f(t)h(t) + g(t)h(t)) dt$$

$$= \underbrace{\int_a^b f(t)h(t) dt}_{[a} + \underbrace{\int_a^b g(t)h(t) dt}_{a]$$

$$= \langle f, h \rangle + \langle g, h \rangle$$



$$\textcircled{2} \quad c \in \mathbb{R}, \quad f, g \in V.$$

$$\langle cf, g \rangle = \int_a^b c f(t) g(t) dt$$

$$= c \int_a^b f(t) g(t) dt$$

$$= c \langle f, g \rangle.$$

3

$f, g \in V$

$$\langle f, g \rangle = \int_a^b f(t)g(t)dt$$

$$= \int_a^b g(t)f(t)dt$$

$$= \langle g, f \rangle$$

4

If  $f = 0$

$$\text{then } \langle f, f \rangle = \int_a^b 0 \, dt = 0.$$

Conversely, if  $\langle f, f \rangle = 0$   
 $\Rightarrow \int_a^b (f(t))^2 \, dt = 0$

$$\Rightarrow f = 0$$

Fact from calculus:

If the definite integral of a non-negative

continuous function is zero

then the function must  
be zero.