

## 5.A Invariant Subspaces

### Definition: invariant subspace

Suppose  $T \in \mathcal{L}(V)$ . A subspace  $U$  of  $V$  is called **invariant** under  $T$  if  $u \in U$  implies  $Tu \in U$ .

### Definition: eigenvalue

Suppose  $T \in \mathcal{L}(V)$ . A number  $\lambda \in \mathbb{F}$  is called an **eigenvalue** of  $T$  if there exists  $v \in V$  such that  $v \neq 0$  and  $Tv = \lambda v$ .

### Equivalent conditions to be an eigenvalue

Suppose  $V$  is finite-dimensional,  $T \in \mathcal{L}(V)$ , and  $\lambda \in \mathbb{F}$ . Then TFAE:

- (a)  $\lambda$  is an eigenvalue of  $T$ .
- (b)  $T - \lambda I$  is not injective.
- (c)  $T - \lambda I$  is not surjective.
- (d)  $T - \lambda I$  is not invertible.

### Definition: eigenvector

Suppose  $T \in \mathcal{L}(V)$  and  $\lambda \in \mathbb{F}$  is an eigenvalue of  $T$ . A vector  $v \in V$  is called an **eigenvector** of  $T$  corresponding to  $\lambda$  if  $v \neq 0$  and  $Tv = \lambda v$ .

### Linearly independent eigenvectors

Let  $T \in \mathcal{L}(V)$ . Suppose  $\lambda_1, \dots, \lambda_m$  are distinct eigenvalues of  $T$  and  $v_1, \dots, v_m$  are corresponding eigenvectors. Then  $v_1, \dots, v_m$  is linearly independent.

### Number of eigenvalues

Suppose  $V$  is finite-dimensional. Then each operator on  $V$  has at most  $\dim V$  distinct eigenvalues.

### Definition: $T|_U$ and $T/U$

Suppose  $T \in \mathcal{L}(V)$  and  $U$  is a subspace of  $V$  invariant under  $T$ .

- The **restriction operation**  $T|_U \in \mathcal{L}(U)$  is defined by

$$T|_U(u) = Tu \quad \forall u \in U.$$

- The **quotient operator**  $T/U \in \mathcal{L}(V/U)$  is defined by

$$(T/U)(v + U) = Tv + U \quad \forall v \in V.$$

## 5.B Eigenvectors and Upper-Triangular Matrices

### Definition: $T^m$

Suppose  $T \in \mathcal{L}(V)$  and  $m$  is a positive integer.

- $T^m$  is defined by

$$T^m = \underbrace{T \cdots T}_{m \text{ times}}$$

- $T^0$  is defined to be the identity operator  $I$  on  $V$ .
- If  $T$  is invertible with inverse  $T^{-1}$ , then  $T^{-m}$  is defined by

$$T^{-m} = (T^{-1})^m.$$

### Definition: $p(T)$

Suppose  $T \in \mathcal{L}(V)$  and  $p \in \mathcal{P}(\mathbb{F})$  is a polynomial given by

$$p(z) = a_0 + a_1z + a_2z^2 + \cdots + a_mz^m$$

for  $z \in \mathbb{F}$ . Then  $p(T)$  is the operator defined by

$$p(T) = a_0I + a_1T + a_2T^2 + \cdots + a_mT^m.$$

### Definition: *product of polynomials*

If  $p, q \in \mathcal{P}(\mathbb{F})$ , then  $pq \in \mathcal{P}(\mathbb{F})$  is the polynomial defined by

$$(pq)(z) = p(z)q(z)$$

for  $z \in \mathbb{F}$ .

## 6.A Inner Products and Norms

### Definition: *norm*

The length of a vector  $x$  is called the **norm** of  $x$  denoted  $\|x\|$ . The norm is non-linear.

$$\|x\| = \sqrt{x_1^2 + \cdots + x_n^2}.$$

### Definition: *dot product*

For  $x, y \in \mathbb{R}^n$ , the **dot product** of  $x$  and  $y$ , denoted  $x \cdot y$  (or  $x.y$ ) is defined by

$$x \cdot y = x_1y_1 + \cdots + x_ny_n,$$

where  $x = (x_1, \dots, x_n)$  and  $y = (y_1, \dots, y_n)$ .

Note that the dot product of two vectors in  $\mathbb{R}^n$  is a scalar (number (i.e. 1, 7, 5.3,  $\pi$ )), not another vector.

$$x \cdot x = \|x\|^2 \quad \forall x \in \mathbb{R}^n$$

**dot product properties** For a dot product on  $\mathbb{R}^n$ ,

- (a)  $x \cdot x \geq 0 \quad \forall x \in \mathbb{R}^n$
- (b)  $x \cdot x = 0 \iff x = 0$
- (c) For a fixed  $y \in \mathbb{R}^n$ , the map from  $\mathbb{R}^n \rightarrow \mathbb{R}$  that sends  $x \in \mathbb{R}^n$  to  $x \cdot y$  is linear
- (d)  $x \cdot y = y \cdot x \quad \forall x, y \in \mathbb{R}^n$  (commutative)

**Definition:**  $\operatorname{Re} z$ ,  $\operatorname{Im} z$

Suppose  $z = a + bi$  where  $a, b \in \mathbb{R}$ .

- The **real part** of  $z$ , denoted  $\operatorname{Re} z$ , is defined by  $\operatorname{Re} z = a$ .
- The **imaginary part** of  $z$ , denoted  $\operatorname{Im} z$ , is defined by  $\operatorname{Im} z = b$ .

Thus for every  $z \in \mathbb{C}$ , we have

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

**Definition: complex conjugate**

Suppose  $z \in \mathbb{C}$ . The **complex conjugate** of  $z \in \mathbb{C}$ , denoted  $\bar{z}$ , is defined by

$$\bar{z} = \operatorname{Re} z - (\operatorname{Im} z)i.$$

**Definition: absolute value**

Suppose  $z \in \mathbb{C}$ . The **absolute value** of  $z \in \mathbb{C}$ , denoted  $|z|$ , is defined by

$$|z| = \sqrt{(\operatorname{Re} z)^2 + (\operatorname{Im} z)^2}.$$

**Complex properties**

- sum of  $z$  and  $\bar{z}$

$$z + \bar{z} = 2 \operatorname{Re} z$$

- difference of  $z$  and  $\bar{z}$

$$z - \bar{z} = 2(\operatorname{Im} z)i$$

- product of  $z$  and  $\bar{z}$

$$z\bar{z} = |z|^2$$

- additivity and multiplicativity of complex conjugate

$$\overline{w + z} = \bar{w} + \bar{z} \quad \text{and} \quad \overline{wz} = \bar{w}\bar{z}$$

- conjugate of a conjugate

$$\overline{\bar{z}} = z$$

- real and imaginary parts are bounded by  $|z|$

$$|\operatorname{Re} z| \leq |z| \quad \text{and} \quad |\operatorname{Im} z| \leq |z|$$

- absolute value of a complex conjugate

$$|\bar{z}| = |z|$$

- multiplicativity of absolute value

$$|wz| = |w| |z|$$

- Triangle Inequality

$$|w + z| \leq |w| + |z|$$

Recall that for  $\lambda = a + bi$  where  $a, b \in \mathbb{R}$ , then

- The absolute value of  $\lambda$ , denoted  $|\lambda|$ , is defined as

$$|\lambda| = \sqrt{a^2 + b^2}.$$

- The complex conjugate of  $\lambda$ , denoted  $\bar{\lambda}$ , is defined as

$$\bar{\lambda} = a - bi.$$

- The absolute value squared is defined as

$$|\lambda|^2 = \lambda \bar{\lambda}$$

For  $z = (z_1, \dots, z_n) \in \mathbb{C}^n$ , we define the norm of  $z$  by

$$\|z\| = \sqrt{|z_1|^2 + \dots + |z_n|^2}.$$

These absolute values are needed because we want  $\|z\|$  to be a nonnegative scalar. Note that

$$\|z\|^2 = z_1 \bar{z}_1 + \dots + z_n \bar{z}_n.$$

We want to think of  $\|z\|^2$  as an inner product of  $z$  with itself, like we did in  $\mathbb{R}^n$ . The equation above suggests that for  $w = (w_1, \dots, w_n) \in \mathbb{C}^n$ , the inner product of  $w$  with  $z$  should be

$$\langle w, z \rangle = w_1 \bar{z}_1 + \dots + w_n \bar{z}_n$$

$$\langle z, w \rangle = z_1 \overline{w_1} + \cdots + z_n \overline{w_n}$$

If  $\lambda \in \mathbb{C}$  then the notation  $\lambda \geq 0$  means that  $\lambda$  is a non-negative real number.

We use the common notation  $\langle u, v \rangle$  with angle brackets to denote an inner product. Parenthesis can also be used, but then  $(u, v)$  becomes confusing.

**Definition: inner product**

An **inner product** on  $V$  is a function that takes each ordered pair  $(u, v)$  of elements of  $V$  to a number  $\langle u, v \rangle \in \mathbb{F}$  and has the following properties:

**positivity:**

$$\langle v, v \rangle \geq 0 \quad \forall v \in V$$

**definiteness:**

$$\langle v, v \rangle = 0 \iff v = 0$$

**additivity in the first slot:**

$$\langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle \quad \forall u, v, w \in V$$

**homogeneity in the first slot:**

$$\langle \lambda u, v \rangle = \lambda \langle u, v \rangle \quad \forall \lambda \in \mathbb{F} \text{ and } u, v \in V$$

**conjugate symmetry:**

$$\langle u, v \rangle = \overline{\langle v, u \rangle} \quad \forall u, v \in V$$

Note:  $x = \bar{x} \quad \forall x \in \mathbb{R}$

**Examples:**

A **Euclidean inner product** on  $\mathbb{F}^n$  is defined by

$$\langle (w_1, \dots, w_n), (z_1, \dots, z_n) \rangle = w_1 \overline{z_1} + \cdots + w_n \overline{z_n}.$$

If  $c_1, \dots, c_n$  are all positive numbers, then an inner product can be defined on  $\mathbb{F}^n$  by

$$\langle (w_1, \dots, w_n), (z_1, \dots, z_n) \rangle = c_1 w_1 \overline{z_1} + \cdots + c_n w_n \overline{z_n}.$$

An inner product can be defined on the vector space of continuous real-valued functions on the interval  $[-1, 1]$  by

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

An inner product can be defined on  $\mathcal{P}(\mathbb{R})$  by

$$\langle p, q \rangle = \int_0^\infty p(x)q(x)e^{-x} dx$$

**Definition: inner product space**

An **inner product space** is a vector space  $V$  along with an inner product on  $V$ .

Unless otherwise stated, you can assume the inner product on  $\mathbb{F}^n$  is defined as

$$\langle w, z \rangle = w_1 \overline{z_1} + \cdots + w_n \overline{z_n}$$

**Notation:** For the rest of this chapter,  $V$  denotes an inner product space over  $\mathbb{F}$ .

### Basic properties of an inner product

- (a) For each fixed  $u \in V$ , the function that takes  $v$  to  $\langle v, u \rangle$  is a linear map from  $V$  to  $\mathbb{F}$ .
- (b)  $\langle 0, u \rangle = 0 \quad \forall u \in V$
- (c)  $\langle u, 0 \rangle = 0 \quad \forall u \in V$
- (d)  $\langle u, v + w \rangle = \langle u, v \rangle + \langle u, w \rangle \quad \forall u, v, w \in V$
- (e)  $\langle u, \lambda v \rangle = \bar{\lambda} \langle u, v \rangle \quad \forall \lambda \in \mathbb{F} \text{ and } u, v \in V$
- (f)  $\langle u, v \rangle = 0 \iff \langle v, u \rangle = 0 \quad \forall u, v \in V$

### Definition: norm

For  $v \in V$ , the **norm** of  $v$ , denoted  $\|v\|$ , is defined by

$$\|v\| = \sqrt{\langle v, v \rangle}.$$

In other words,  $\|v\|^2 = \langle v, v \rangle$ .

### Basic properties of the norm

Suppose  $v \in V$ . Then

- (a)  $\|v\| = 0 \iff v = 0$ .
- (b)  $\|\lambda v\| = |\lambda| \|v\| \quad \forall \lambda \in \mathbb{F}$ .

### Examples:

- (a) If  $(z_1, \dots, z_n) \in \mathbb{F}^n$  (with the usual Euclidean inner product), then

$$\|(z_1, \dots, z_n)\| = \sqrt{|z_1|^2 + \cdots + |z_n|^2}.$$

- (b) In the vector space of continuous real-valued functions on  $[1, -1]$  (with the inner product defined as  $\langle f, g \rangle = \int_{-1}^1 f(x)g(x) dx$ ), then

$$\|f\| = \sqrt{\int_{-1}^1 (f(x))^2 dx}.$$

**Definition: orthogonal**

Two vectors  $u, v \in V$  are called **orthogonal** if  $\langle u, v \rangle = 0$ .

Instead of saying “ $u$  and  $v$  are orthogonal”, we sometimes say “ $u$  is orthogonal to  $v$ ”.

For two vectors  $u, v \in \mathbb{R}^2$ , then

$$\langle u, v \rangle = \|u\| \|v\| \cos \theta,$$

where  $\theta$  is the angle between the two vectors. If  $\theta = 90^\circ = \frac{\pi}{2}$  radians then  $\cos \theta = 0$ . That is, if  $u$  and  $v$  are perpendicular the inner product is 0.

**Orthogonality and 0**

- (a)  $0$  is orthogonal to every vector in  $V$ .
- (b)  $0$  is the only vector in  $V$  that is orthogonal to itself.

**Pythagorean Theorem**

If  $u$  and  $v$  are orthogonal vectors in  $V$ . Then

$$\|u + v\|^2 = \|u\|^2 + \|v\|^2$$

The converse is true if  $V$  is a real inner product space. That is, if  $\|u + v\|^2 = \|u\|^2 + \|v\|^2$  then  $u$  and  $v$  are orthogonal.

**An orthogonal decomposition**

Suppose  $u, v \in V$ , with  $v \neq 0$ . Set  $c = \frac{\langle u, v \rangle}{\|v\|^2}$  and  $w = u - \frac{\langle u, v \rangle}{\|v\|^2}v$ . Then

$$\langle w, v \rangle = 0 \quad \text{and} \quad u = cv + w.$$

**Cauchy-Schwarz Inequality**

Suppose  $u, v \in V$ . Then

$$|\langle u, v \rangle| \leq \|u\| \|v\|.$$

Further, this is an *equality* if and only if  $u$  and  $v$  are scalar multiples of one another.

**Triangle Inequality**

Suppose  $u, v \in V$ . Then

$$\|u + v\| \leq \|u\| + \|v\|.$$

This is an equality if and only if  $u, v$  are a nonnegative multiples of the other.

**Parallelogram Equality**

Suppose  $u, v \in V$ . Then

$$\|u + v\|^2 + \|u - v\|^2 = 2(\|u\|^2 + \|v\|^2).$$

## 6.B Orthonormal Bases

### Definition: orthonormal

- A list of vectors is called **orthonormal** if each vector in the list has norm 1 and is orthogonal to all the other vectors in the list.
- In other words, a list  $e_1, \dots, e_m$  of vectors in  $V$  is orthonormal if

$$\langle e_j, e_k \rangle = \begin{cases} 1 & \text{if } j = k, \\ 0 & \text{if } j \neq k. \end{cases}$$

Example, the standard basis in  $\mathbb{F}^n$ .

### The norm of an orthonormal linear combination

If  $e_1, \dots, e_m$  is an orthonormal list of vectors in  $V$ , then

$$\|a_1 e_1 + \dots + a_m e_m\|^2 = |a_1|^2 + \dots + |a_m|^2$$

for all  $a_1, \dots, a_m \in \mathbb{F}$ .

### Corollary: An orthonormal list is linearly independent

Every orthonormal list of vectors is linearly independent.

### Definition: orthonormal basis

An **orthonormal basis** of  $V$  is an orthonormal list of vectors in  $V$  that is also a basis of  $V$ .

For example, the standard basis is an orthonormal basis of  $\mathbb{F}^n$ .

### An orthonormal list of the right length is an orthonormal basis

Every orthonormal list of vectors in  $V$  with length  $\dim V$  is an orthonormal basis of  $V$ .

### Writing a vector as linear combination of orthonormal basis

Suppose  $e_1, \dots, e_n$  is an orthonormal basis of  $V$  and  $v \in V$ . Then

$$v = \langle v, e_1 \rangle e_1 + \dots + \langle v, e_n \rangle e_n$$

and

$$\|v\|^2 = |\langle v, e_1 \rangle|^2 + \dots + |\langle v, e_n \rangle|^2.$$

### Gram-Schmidt Procedure

Suppose  $v_1, \dots, v_m$  is a linearly independent list of vectors in  $V$ . Let  $e_1 = \frac{v_1}{\|v_1\|}$ .

For  $j = 2, \dots, m$ , define  $e_j$  inductively by

$$e_j = \frac{v_j - \langle v_j, e_1 \rangle e_1 - \dots - \langle v_j, e_{j-1} \rangle e_{j-1}}{\|v_j - \langle v_j, e_1 \rangle e_1 - \dots - \langle v_j, e_{j-1} \rangle e_{j-1}\|}.$$



Then  $e_1, \dots, e_m$  is an orthonormal list of vectors in  $V$  such that

$$\text{span}(v_1, \dots, v_j) = \text{span}(e_1, \dots, e_j)$$

for  $j = 1, \dots, m$ .

### Existence of orthonormal basis

Every finite-dimensional inner product space has an orthonormal basis.

### Corollary: Orthonormal list extends to orthonormal basis

Suppose  $V$  is finite-dimensional. Then every orthonormal list of vectors in  $V$  can be extended to an orthonormal basis of  $V$ .

### Upper-triangular matrix with respect to orthonormal basis

Suppose  $T \in \mathcal{L}(V)$ . If  $T$  has an upper-triangular matrix with respect to some basis of  $V$ , then  $T$  has an upper-triangular matrix with respect to some orthonormal basis of  $V$ .

### Schur's Theorem

Suppose  $V$  is a finite-dimensional complex vector space and  $T \in \mathcal{L}(V)$ . Then  $T$  has an upper-triangular matrix with respect to some orthonormal basis of  $V$ .

Because linear maps into a scalar field play a special role, we give them a special name.

### Definition: linear functional

A **linear functional** on  $V$  is a linear map from  $V$  to  $\mathbb{F}$ . In other words, a linear function is an element of  $\mathcal{L}(V, \mathbb{F})$ .

### Riesz Representation Theorem

Suppose  $V$  is finite-dimensional and  $\varphi$  is a linear functional on  $V$ . Then there is a unique vector  $u \in V$  such that

$$\varphi(v) = \langle v, u \rangle \quad \forall v \in V,$$

where

$$u = \overline{\varphi(e_1)}e_1 + \dots + \overline{\varphi(e_n)}e_n.$$

## 6.C Orthogonal Complements and Minimization Problems

### Definition: orthogonal complement, $U^\perp$

If  $U$  is a subset of  $V$ , then the **orthogonal complement** of  $U$ , denoted  $U^\perp$ , is the set of all vectors in  $V$  that are orthogonal to every vector in  $U$ :

$$U^\perp = \{v \in V : \langle v, u \rangle = 0 \text{ for every } u \in U\}.$$

For example, if  $U$  is a line in  $\mathbb{R}^3$  then  $U^\perp$  is the plane containing 0 that is perpendicular to  $U$ . If  $U$  was a plane in  $\mathbb{R}^3$  then  $U^\perp$  is a line passing through 0 that is perpendicular to  $U$ .

### Basic properties of orthogonal complement

- (a) If  $U$  is a subset of  $V$ , then  $U^\perp$  is a subspace of  $V$ .
- (b)  $\{0\}^\perp = V$ .
- (c)  $V^\perp = \{0\}$ .
- (d) If  $U$  is a subset of  $V$ , then  $U \cap U^\perp \subseteq \{0\}$ .
- (e) If  $U$  and  $W$  are subsets of  $V$  and  $U \subseteq W$ , then  $W^\perp \subseteq U^\perp$ .

### Direct sum of a subspace and its orthogonal complement

Suppose  $U$  is a finite-dimensional subspace of  $V$ . Then

$$V = U \oplus U^\perp$$

### Dimension of the orthogonal complement

Suppose  $V$  is finite-dimensional and  $U$  is a subspace of  $V$ . Then

$$\dim U^\perp = \dim V - \dim U.$$

### The orthogonal complement of the orthogonal complement

Suppose  $U$  is a finite-dimensional subspace of  $V$ . Then

$$U = (U^\perp)^\perp.$$

### Definition: orthogonal projection, $P_U$

Suppose  $U$  is a finite-dimensional subspace of  $V$ . The **orthogonal projection** of  $V$  onto  $U$  is the operator  $P_U \in \mathcal{L}(V)$  defined as follows:

For  $v \in V$ , write  $v = u + w$ , where  $u \in U$  and  $w \in U^\perp$ . Then  $P_U v = u$ .

### Properties of the orthogonal projection $P_U$

Suppose  $U$  is a finite-dimensional subspace of  $V$  and  $v \in V$ . Then

- (a)  $P_U \in \mathcal{L}(V)$
- (b)  $P_U u = u \quad \forall u \in U$
- (c)  $P_U w = 0 \quad \forall w \in U^\perp$
- (d)  $\text{range } P_U = U$
- (e)  $\text{null } P_U = U^\perp$
- (f)  $v - P_U v \in U^\perp$
- (g)  $P_U^2 = P_U$

(h)  $\|P_U v\| \leq \|v\|$

(i) for every orthonormal basis  $e_1, \dots, e_m$  of  $U$ ,

$$P_U v = \langle v, e_1 \rangle e_1 + \dots + \langle v, e_m \rangle e_m.$$

### Minimizing the distance to a subspace

Suppose  $U$  is a finite-dimensional subspace of  $V$ ,  $v \in V$ , and  $u \in U$ . Then

$$\|v - P_U v\| \leq \|v - u\|.$$

Furthermore, the inequality is an equality if and only if  $u = P_U v$ .

## 7.A Self-Adjoint and Normal Operators

**Definition:** *adjoint*,  $T^*$

Suppose  $T \in \mathcal{L}(V, W)$ . The *adjoint* of  $T$  is the function  $T^* : W \rightarrow V$  such that

$$\langle Tv, w \rangle = \langle v, T^* w \rangle$$

for every  $v \in V$  and every  $w \in W$ .

### The adjoint is a linear map

If  $T \in \mathcal{L}(V, W)$ , then  $T^* \in \mathcal{L}(W, V)$ .

### Properties of the adjoint

- (a)  $(S + T)^* = S^* + T^*$  for all  $S, T \in \mathcal{L}(V, W)$
- (b)  $(\lambda T)^* = \bar{\lambda} T^*$  for all  $\lambda \in F$  and  $T \in \mathcal{L}(V, W)$
- (c)  $(T^*)^* = T$  for all  $T \in \mathcal{L}(V, W)$
- (d)  $I^* = I$ , where  $I$  is the identity operator on  $V$
- (e)  $(ST)^* = T^* S^*$  for all  $T \in \mathcal{L}(V, W)$  and  $S \in \mathcal{L}(W, U)$  (here  $U$  is an inner product space over  $\mathbb{F}$ ).

### Null space and range of $T^*$

- (a)  $\text{null } T^* = (\text{range } T)^\perp$
- (b)  $\text{range } T^* = (\text{null } T)^\perp$
- (c)  $\text{null } T = (\text{range } T^*)^\perp$
- (d)  $\text{range } T = (\text{null } T^*)^\perp$

**Definition: conjugate transpose**

The **conjugate transpose** of an  $m$ -by- $n$  matrix is the  $n$ -by- $m$  matrix obtained by transposing the matrix and taking the complex conjugate of each entry.

**The matrix of  $T^*$** 

Let  $T \in \mathcal{L}(V, W)$ . Suppose  $e_1, \dots, e_n$  is an orthonormal basis of  $V$  and  $f_1, \dots, f_m$  is an orthonormal basis of  $W$ . Then

$$\mathcal{M}(T^*, (f_1, \dots, f_m), (e_1, \dots, e_n))$$

is the conjugate transpose of

$$\mathcal{M}(T, (e_1, \dots, e_n), (f_1, \dots, f_m))$$

**Definition: self-adjoint**

An operator  $T \in \mathcal{L}(V)$  is called **self-adjoint** if  $T = T^*$ . In other words,  $T \in \mathcal{L}(V)$  is self-adjoint if and only if

$$\langle Tv, w \rangle = \langle v, Tw \rangle$$

for all  $v, w \in V$ .

**Eigenvalues of self-adjoint operators are real**

Every eigenvalue of a self-adjoint operator is real

**Over  $\mathbb{C}$ ,  $Tv$  is orthogonal to  $v$  for all  $v$  only for the 0 operator**

Suppose  $V$  is a complex inner product space and  $T \in \mathcal{L}(V)$ . Suppose

$$\langle Tv, v \rangle = 0$$

for all  $v \in V$ . Then  $T = 0$ .

**Over  $\mathbb{C}$ ,  $\langle Tv, v \rangle$  is real for all  $v$  only for self-adjoint operators**

Suppose  $V$  is a complex inner product space and  $T \in \mathcal{L}(V)$ . Then  $T$  is self-adjoint if and only if

$$\langle Tv, v \rangle \in \mathbb{R}$$

for every  $v \in V$ .

**If  $T = T^*$  and  $\langle Tv, v \rangle = 0$  for all  $v$ , then  $T = 0$** 

Suppose  $T$  is a self-adjoint operator on  $V$  such that

$$\langle Tv, v \rangle = 0$$

for all  $v \in V$ . Then  $T = 0$ .

**Definition: normal**

An operator on an inner product space is called **normal** if it commutes with its adjoint. In other words,  $T \in \mathcal{L}(V)$  is normal if

$$TT^* = T^*T.$$

**$T$  is normal if and only if  $\|Tv\| = \|T^*v\|$  for all  $v$**

An operator  $T \in \mathcal{L}(V)$  is normal if and only if

$$\|Tv\| = \|T^*v\|$$

for all  $v \in V$ .

**For  $T$  normal,  $T$  and  $T^*$  have the same eigenvectors**

Suppose  $T \in \mathcal{L}(V)$  is normal and  $v \in V$  is an eigenvector of  $T$  with eigenvalue  $\lambda$ . Then  $v$  is also an eigenvector of  $T^*$  with eigenvalue  $\bar{\lambda}$ .

**Orthogonal eigenvectors for normal operators**

Suppose  $T \in \mathcal{L}(V)$  is normal. Then the eigenvectors of  $T$  corresponding to distinct eigenvalues are orthogonal.