## Eigenvalues of linear transformation and representation matrix | #lemma

Let V be a finitely generated vector space over  $\mathbb{F}$ 

Let B be a basis of V

Let  $T: V \to V$  be a linear transformation

Then  $\lambda$  is an eigenvalue of  $T \iff \lambda$  is an eigenvalue of  $[T]_B^B$ 

#### **Proof:**

 $\lambda$  is an eigenvalue of  $T \iff \exists v \neq 0 \in V : T(v) = \lambda v$  $\iff \exists v \neq 0 \in V: [T]_B^B[v]_B = [\lambda v]_B = \lambda [v]_B \iff \lambda \text{ is an eigenvalue of } [T]_B^B$ 

### Cayley-Hamilton theorem #theorem

$$ext{Let } A \in \mathbb{F}^{n imes n} \ ext{Then } P_A(A) = 0$$

Explanation, not proof:

$$P_A(A) = \det(AI - A) = \det(0) = 0$$

Proof:

$$P_A(\lambda) = \lambda^n + \sum_{i=0}^{n-1} a_i \lambda^i$$

$$(\lambda I - A) \cdot \operatorname{adj}(\lambda I - A) = \det(\lambda I - A)I = P_A(\lambda)I$$

$$\operatorname{adj}(\lambda I - A) \in \mathbb{F}_{n-1}[\lambda]^{n \times n}$$

$$\implies \exists \{B_0, \dots, B_{n-1}\} \subseteq \mathbb{F}^{n \times n} : \operatorname{adj}(\lambda I - A) = \sum_{i=0}^{n-1} \lambda^i B_i$$

$$\implies (\lambda I - A) \cdot \sum_{i=0}^{n-1} \lambda^i B_i = P_A(\lambda)I$$

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$$\implies (\lambda I - A) \cdot \sum_{i=0}^{n-1} \lambda^i A_i = \sum_$$

Diagonalization and minimal polynomial #theorem

Let 
$$A \in \mathbb{F}^{n imes n}$$

 $A \sim D \iff m_A ext{ is factorizable into distinct linear factors}$ 

$$egin{aligned} &\operatorname{Proof:} \ &\Longrightarrow \operatorname{Let} A \sim D \ \ &\Longrightarrow P_A(\lambda) = \prod_{i=1}^k (\lambda - lpha_i)^{\mu_A(lpha_i)} \ \ &\Longrightarrow m_A(\lambda) = \prod_{i=1}^k (\lambda - lpha_i)^{t_i}, t_i \leq \mu_A(lpha_i) \ \ A_I = D \end{aligned}$$

 $\implies$  Largest Jordan block corresponding to any eigenvalue of A is of size 1

$$\implies orall i \in [1,k]: t_i = 1 \implies oxedom{m_A(\lambda) = \prod_{i=1}^k (\lambda - lpha_i)}$$

 $\longleftarrow$  Let  $m_A$  be factorizable into distinct linear factors

$$\implies m_A(\lambda) = \prod_{i=1}^k (\lambda - lpha_i)$$
  $\implies P_A(\lambda) = \prod_{i=1}^k (\lambda - lpha_i)^{\mu_A(lpha_i)} \implies A \sim A_J$ 

 $\implies$  Largest Jordan block corresponding to any eigenvalue of A is of size 1

$$\implies A_J = D \implies \boxed{A \sim D}$$

# Pythagorean theorem #theorem

Let B be an orthogonal basis of V

$$egin{aligned} \operatorname{Let} v \in V \ v &= \sum_{i=1}^n lpha_i v_i \ \implies &\sum_{i=1}^n lpha_i v_i \ \end{array} = \sum_{i=1}^n \|lpha_i v_i\|^2 = \sum_{i=1}^n |lpha_i|^2 \|v_i\|^2$$

Proof:

$$egin{aligned} \sum_{i=1}^n lpha_i v_i & = \left\langle \sum_{i=1}^n lpha_i v_i, \sum_{i=1}^n lpha_i v_i 
ight
angle = \sum_{i=1}^n \sum_{j=1}^n \left\langle lpha_i v_i, lpha_j v_j 
ight
angle = \ & = \sum_{i=1}^n \sum_{j=1}^n lpha_i \overline{lpha_j} \langle v_i, v_j 
angle = \sum_{i=1}^n lpha_i \overline{lpha_i} \langle v_i, v_i 
angle = \sum_{i=1}^n |lpha_i|^2 \|v_i\|^2 \end{aligned}$$

# Cauchy-Schwarz inequality #lemma

Let V be an inner product space

$$egin{aligned} \operatorname{Let} v, u \in V \ \|v\| \cdot \|u\| & \geq |\langle v, u 
angle| \ \|v\| \cdot \|u\| & = |\langle v, u 
angle| \iff u = lpha v \end{aligned}$$

$$egin{aligned} \operatorname{Case} & 1. \ u = v = 0 \ & \|v\| \cdot \|u\| \geq |\langle v, u 
angle| \end{aligned}$$
 $\operatorname{Case} & 2. \ egin{bmatrix} v 
eq 0 \ u 
eq 0 \end{bmatrix} \operatorname{Let} v 
eq 0 \end{aligned}$ 
 $\operatorname{Let} A = \left\{ rac{v}{\|v\|} 
ight\}$ 

A is an orthonormal set

By Bessel's inequality: 
$$\|u\|^2 \ge \left\langle u, \frac{v}{\|v\|} \right\rangle^2 = \frac{1}{\|v\|^2} |\langle u, v \rangle|^2 \underset{|z| = |\overline{z}|}{=} \frac{1}{\|v\|^2} |\langle v, u \rangle|^2$$

$$\implies \|u\|^2 \cdot \|v\|^2 \ge |\langle v, u \rangle|^2$$

$$\implies \overline{\|v\| \cdot \|u\| \ge |\langle v, u \rangle|}$$

$$\|v\|\cdot\|u\| = |\langle v,u
angle| \iff \underbrace{\|u\|^2 = rac{1}{\|v\|^2} |\langle v,u
angle|^2}_{ ext{Bessel's equality case}} \ \iff u \in spA \iff \overline{u = lpha v}$$

## Riesz Representation theorem #theorem

Let V be an inner product space over  $\mathbb F$ 

Let  $T:V \to \mathbb{F}$  be a linear transformation

Then  $\exists ! h \in V : \forall v \in V : T(v) = \langle v, h \rangle$ 

#### Proof:

Let B be an orthonormal basis of V

Let  $C = \{1\}$  be a standard basis of  $\mathbb{F}$ 

$$T(v) = [T(v)]_C = [T]_C^B[v]_B \in \mathbb{F} \implies T(v) = ([T]_C^B[v]_B)^T = [v]_B^T([T]_C^B)^T$$

$$\langle v, h \rangle = [v]_B^T G_B[\overline{h}]_B = [v]_B^T[\overline{h}]_B$$

$$[\cdot]_B \text{ is surjective } \implies \boxed{\exists h \in V : [h]_B = \overline{([T]_C^B)^T} = (\overline{[T]_C^B})^T}$$

$$\text{Let } \forall v \in V : \langle v, h_1 \rangle = T(v) = \langle v, h_2 \rangle$$

$$\implies [v]_B^T[\overline{h}_1]_B = [v]_B^T[\overline{h}_2]_B$$

$$\implies [h_1]_B = [h_2]_B \implies \boxed{h_1 = h_2}$$

Existence and uniqueness of adjoint linear transformation (#theorem

Let V,U be inner product spaces over  $\mathbb F$ Let T:V o U be a linear transformation

Let  $S:U\to V$  be a linear transformation

Then  $\exists ! S : S$  is a adjoint linear transformation of T

$$egin{aligned} \operatorname{Let} u &\in U \ &\operatorname{Let} K_u : V 
ightarrow \mathbb{F}, K_u(v) = \langle T(v), u 
angle \ &\Longrightarrow ext{ By Riesz theorem } \exists h_u \in V : \forall v \in V : K_u(v) = \langle v, h_u 
angle \ &\operatorname{Let} S : U 
ightarrow V, S(u) = h_u \ &\Longrightarrow ext{ } \forall v \in V, u \in U : \langle T(v), u 
angle = \langle v, h_u 
angle = \langle v, S(u) 
angle \end{aligned}$$

$$\operatorname{Let} v \in V, u_1, u_2 \in U, lpha \in \mathbb{F} \ \langle v, S(u_1 + lpha u_2) 
angle = \langle T(v), u_1 + lpha u_2 
angle = \langle T(v), u_1 
angle + \overline{lpha} \langle T(v), u_2 
angle = K_{u_1}(v) + \overline{lpha} K_{u_2}(v) = \ = \langle v, h_{u_1} 
angle + \overline{lpha} \langle v, h_{u_2} 
angle = \langle v, S(u_1) 
angle + \overline{lpha} \langle v, S(u_2) 
angle = \langle v, S(u_1) + lpha S(u_2) 
angle \ \forall v \in V : \langle v, S(u_1 + lpha u_2) 
angle = \langle v, S(u_1) + lpha S(u_2) 
angle \ \Rightarrow S(u_1 + lpha u_2) = S(u_1) + lpha S(u_2) \Rightarrow \boxed{S \text{ is a linear transformation}} \ \text{Let } \hat{S} : U \to V, \forall v \in V, u \in U : \langle T(v), u 
angle = \langle v, S(u) 
angle \ \Rightarrow \forall v \in V, u \in U : \langle v, S(u) 
angle = \langle v, \hat{S}(u) 
angle \Rightarrow \forall u \in U : S(u) = \hat{S}(u) \ \Rightarrow \boxed{S = \hat{S}}$$

# Representation matrix of adjoint linear transformation #lemma

B orthonormal basis of V

C orthonormal basis of U

 $T:V \to U$  is a linear transformation

$$[T^*]_B^C = ([T]_C^B)^*$$

Proof:

$$\dim V = n$$
  $\dim U = m$   $\mathbb{T}^{n} \in \mathbb{F}^{m imes n}$ 

$$[T]_C^B \in \mathbb{F}^{m imes n} \ [T^*]_B^C \in \mathbb{F}^{n imes m}$$

$$orall v \in V, u \in U: \langle T(v), u 
angle = \langle v, T*(u) 
angle$$

$$egin{aligned} \langle T(v),u
angle &= [T(v)]_C^TG_C\overline{[u]_C} = [T(v)]_C^T\overline{[u]_C} = [v]_B^T([T]_C^B)^T\overline{[u]_C} \ \langle v,T^*(u)
angle &= [v]_B^TG_B\overline{[T^*(u)]_B} = [v]_B^T\overline{[T^*]_B^C}[u]_C = [v]_B^T\overline{[T^*]_B^C}\overline{[u]_C} \end{aligned}$$

$$\implies ([T]_C^B)^T = \overline{[T^*]_B^C} \implies \overline{([T]_C^B)^* = [T^*]_B^C}$$

**Gram-Schmidt matrix of two bases** #lemma

Let 
$$B, \hat{B}$$
 be bases of  $V$   
Let  $C = [I]_B^{\hat{B}}$   
Then  $G_{\hat{B}} = C^T G_B \overline{C}$ 

$$\begin{split} \langle v,u\rangle &= [v]_B^T G_B \overline{[u]_B} = ([I]_B^{\hat{B}}[v]_{\hat{B}})^T G_B \overline{[I]_B^{\hat{B}}[u]_{\hat{B}}} = [v]_{\hat{B}}^T \cdot C^T G_B \overline{C} \cdot \overline{[u]_{\hat{B}}} \\ &= \langle v,u\rangle = [v]_{\hat{B}}^T G_{\hat{B}} \overline{[u]_{\hat{B}}} \\ &\Longrightarrow [v]_{\hat{B}}^T \cdot C^T G_B \overline{C} \cdot \overline{[u]_{\hat{B}}} = [v]_{\hat{B}}^T \cdot G_{\hat{B}} \cdot \overline{[u]_{\hat{B}}} \\ & \text{Let } \hat{B} = \{v_1,\dots,v_n\} \\ & \forall i,j \in [1,n] : [v_i]_{\hat{B}}^T \cdot C^T G_B \overline{C} \cdot \overline{[v_j]_{\hat{B}}} = [v_i]_{\hat{B}}^T \cdot G_{\hat{B}} \cdot \overline{[v_j]_{\hat{B}}} \\ &\Longrightarrow \forall i,j \in [1,n] : e_i^T \cdot C^T G_B \overline{C} \cdot e_j = e_i^T \cdot G_{\hat{B}} \cdot e_j \\ &\Longrightarrow \forall i,j \in [1,n] : (C^T G_B \overline{C})_{ij} = (G_{\hat{B}})_{ij} \\ &\Longrightarrow \overline{C}^T G_B \overline{C} = G_{\hat{B}} \end{split}$$

## Polar norm equations

### #lemma

$$egin{align} ext{Re}(\langle v,u
angle) &= rac{1}{2}(\|v+u\|^2 - \|v\|^2 - \|u\|^2) \ ext{Im}(\langle v,u
angle) &= irac{1}{2}(\|v+iu\|^2 - \|v\|^2 - \|u\|^2) \ \end{aligned}$$

## Normal linear operator criterion #theorem

Let V be a finitely generated inner product space over  $\mathbb F$ 

Let B be an orthonormal basis of V

Let  $T: V \to V$  be a linear operator

Then T is normal  $\iff \forall v \in V : \|T(v)\| = \|T^*(v)\|$ 

 $\implies$  Let T be normal

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

$$\text{Let } v \in V$$

$$\|T(v)\| = \sqrt{\langle Tv, Tv \rangle} = \sqrt{\langle v, T^*Tv \rangle}$$

$$\|T^*(v)\| = \sqrt{\langle T^*v, T^*v \rangle} = \sqrt{\langle v, TT^*v \rangle} = \sqrt{\langle v, T^*Tv \rangle} = \|T(v)\|$$

$$\implies \boxed{\forall v \in V : \|T^*(v)\| = \|T(v)\|}$$

$$\text{Let } v, u \in V$$

$$\langle u, TT^*v \rangle = \langle T^*u, T^*v \rangle =$$

$$= \frac{1}{2}(\|T^*u + T^*v\|^2 - \|T^*u\|^2 - \|T^*v\|^2) + i\frac{1}{2}(\|T^*u + iT^*v\|^2 - \|T^*u\|^2 - \|T^*v\|^2)$$

$$\langle u, T^*Tv \rangle = \langle Tu, Tv \rangle =$$

$$= \frac{1}{2}(\|Tu + Tv\|^2 - \|Tu\|^2 - \|Tv\|^2) + i\frac{1}{2}(\|Tu + iTv\|^2 - \|Tu\|^2 - \|Tv\|^2)$$

$$\implies \langle u, T^*Tv \rangle - \langle u, T^*v \rangle =$$

$$= \frac{1}{2}(\|T^*u + T^*v\|^2 - \|Tu + Tv\|^2) + i\frac{1}{2}(\|T^*u + iT^*v\|^2 - \|Tu + iTv\|^2) =$$

$$= \frac{1}{2}(\|T^*u + T^*v\|^2 - \|Tu + Tv\|^2) + i\frac{1}{2}(\|T^*u + iT^*v\|^2 - \|Tu + iTv\|^2) =$$

$$= \frac{1}{2}(\|T^*(u + v)\|^2 - \|T(u + v)\|^2) + i\frac{1}{2}(\|T^*(u + iv)\|^2 - \|T(u + iv)\|^2) = 0 - 0i = 0$$

$$\implies \forall v, u \in V : \langle u, TT^*v \rangle = \langle u, T^*Tv \rangle \implies \boxed{TT^* = T^*T}$$