

7-Dec-2015

HW 16 P1

V

Problem 1. 假定线性空间维数有限. 若线性映射 f 在某组基下有矩阵表示 A , 则 A 的特征值等于 f 的特征值.

证 $(u_1 | u_2 | \cdots | u_n)$ s.t. \downarrow 定义

$$(f(u_1) | f(u_2) | \cdots | f(u_n)) = (u_1 | u_2 | \cdots | u_n) \cdot A$$

下证 $\lambda \in \sigma(A) \Leftrightarrow \lambda \in \sigma(f)$

" \rightarrow " 由 $Av = \lambda v$ ($v \neq 0$)

$$\underbrace{(f(u_1) | f(u_2) | \cdots | f(u_n))}_{\parallel} v = (u_1 | u_2 | \cdots | u_n) \cdot \underbrace{Av}_{\parallel}$$

$$f((u_1 | u_2 | \cdots | u_n) v) \quad \lambda (u_1 | u_2 | \cdots | u_n) v$$

$$\Rightarrow (\lambda, \underbrace{(u_1 | u_2 | \cdots | u_n) v}_{\neq 0})$$

且 f 为线性.

" \Leftarrow " 由 $f(\lambda) = \lambda x$

猜： \exists $v \in V$ s.t.

$$(u_1 | u_2 | \cdots | u_n) \cdot v = x \quad \#$$

Problem 2. 记 f 是整系数的多项式,

$p(x) = x^n + a_1x^{n-1} + a_2x^{n-2} + \cdots + a_n$ 也是整系数的多项式. 将 $p(x)$ 在 \mathbb{C} 中分解作一次因子的乘积, 即 $(x - x_1)(x - x_2) \cdots (x - x_n)$. 试证明, $(x - f(x_1))(x - f(x_2)) \cdots (x - f(x_n))$ 仍是整系数多项式.

Pf. FACT 若 (λ, v) 是 A 的特征向量.

则 $(f(\lambda), v)$ 是 $f(A)$ 的特征向量.

$\Rightarrow \{A \text{ 的特征向量}\} \subseteq \{f(A) \text{ 的特征向量}\}$

\Rightarrow Prop A, B 可相似对角化, 且

$$\exists f(A) = B, \exists g(B) = A$$

$\Rightarrow \exists P \in GL_n(F)$ s.t.

$P^{-1}AP, P^{-1}B P$ 均是 对角矩阵

Thm D 是半正定矩阵.

(i)

\exists 正交矩阵 \tilde{D} s.t.

$$\tilde{D}^2 = D.$$

pf. \tilde{D} 的 迹 由 $Q^T D Q = \lambda_1 + \lambda_2$
 $= \alpha^T \sqrt{\lambda} \alpha$

\tilde{D} 为正交?

$$\text{没有 } \tilde{D}^2 = D = \tilde{D}^2$$

下证 $\tilde{D} = \tilde{D}$.

FACT $\tilde{D} = f(D) \quad (\exists f)$

$$\text{pf. } \exists f. \begin{pmatrix} \sqrt{d_1} & & \\ & \sqrt{d_2} & \\ & & \ddots \\ & & & \sqrt{d_n} \end{pmatrix} = f \left(\begin{pmatrix} d_1 & & \\ & \ddots & \\ & & d_n \end{pmatrix} \right)$$

Thm (Lagrange) $\{x\}$ 和 $\{y\}$ 构成

$$\{x_1, \dots, x_n\} \quad x_i \neq x_j$$

$$\{y_1, \dots, y_n\}$$

已知 f s.t. $f(x_k) = y_k$.

1, 1, 2, 3, 5, 8, 114514

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项式. 将 $p(x)$ 在 \mathbb{C} 中分解作一次因子的乘积, 即

$(x - x_1)(x - x_2) \cdots (x - x_n)$. 试证明, $= \det(xI - A)$

$(x - f(x_1))(x - f(x_2)) \cdots (f - f(x_n))$ 仍是整系数多项

式. $= \det(xI - f(A))$

FACT

λ, v 是 A 的特征值.

$\Rightarrow (f(\lambda), v)$ 是 $f(A)$ 的特征向量.

pf $\exists P$ s.t. $P^{-1}AP = \begin{pmatrix} x_1 & & * \\ & x_2 & \\ 0 & \ddots & x_n \end{pmatrix}$

$\Rightarrow f(A) \sim f\left(\begin{pmatrix} x_1 & & * \\ & x_2 & \\ 0 & \ddots & x_n \end{pmatrix}\right)$

$\sim \begin{pmatrix} f(x_1) & & * \\ & \ddots & \\ 0 & & f(x_n) \end{pmatrix}$

FACT $A \in \mathbb{R}^{n \times n}$

若 $\underset{P}{\text{.}} \quad x^T A x > 0 \quad (x \neq 0)$

则 $\lambda \in \sigma(A)$ 有已的实部 \checkmark

pf [1] 先证 $\sigma(A) \cap i\mathbb{R} = \emptyset$

① $0 \notin \sigma(A)$ ✓

② 若 $i \in \sigma(A)$, $\exists u \in \mathbb{C}^n$?

$\exists u \in \mathbb{C}^n$

$$Au = iu$$

$$A\bar{u} = -i\bar{u}$$

$$\Rightarrow (\operatorname{Re}(u))^T A (\operatorname{Re}(u)) = 0$$

$$(\operatorname{Im}(u))^T A (\operatorname{Im}(u)) = 0$$

而 然.

[2] A 满足 P . $\forall B = -B^T$, $A+B$ 满足 P .

$$\text{取 } B = t \cdot (A^T - A)$$

$$A(t) = A + t(A^T - A)$$

$A(0) = A$, $A(\frac{1}{2})$ 实对称且满足 P

$\Rightarrow A(\frac{1}{2})$ 已定.

8. 若 $B = -B^T$ 则 $I + B$ 可逆, 且 $\underbrace{(I - B)(I + B)^{-1}}$ 是正交矩阵.

$$\Leftrightarrow \left((I - B) (I + B)^{-1} \right)^T = \left((I - B) (I + B)^{-1} \right)^{-1}$$

$$\begin{array}{c} || \\ (I + B^T)^{-1} (I - B^T) \end{array} \quad \begin{array}{c} || \\ (I + B) (I - B)^{-1} \end{array}$$

~~$(I - B)^{-1} \cdot (I + B)$~~

$$A (A^T - A) = 0$$

$\Leftrightarrow A \in \mathbb{R}^{n \times n}, AA^T = AA$

即 $A = A^T$.

$$A^2 = AA^T = (A^T)^2$$

$$\text{tr} \left(\underbrace{(A - A^T)}_{B^T}^T \cdot \underbrace{(A - A^T)}_B \right) = 0 \quad \checkmark$$

\sim 即 $A = Q \begin{pmatrix} M & N \\ 0 & 0 \end{pmatrix} Q^T$

$\boxed{AA^T = A^TA}$

$\begin{pmatrix} MM^T + NN^T & 0 \\ 0 & 0 \end{pmatrix}$
 $\begin{pmatrix} M^2 & MN \\ 0 & 0 \end{pmatrix}$

$$\text{e.g. } A : \mathbb{R}^n \longrightarrow \mathbb{R}^n \quad \det(\mathbf{A}^T) = \underbrace{\det(\mathbf{A}\mathbf{A}^T + \mathbf{V}\mathbf{V}^T)}_{=0}$$

$$v \longmapsto Av$$

$C(A)$ の決定式を求める.

左辺 $C(A)$. $C(A)^T$ の高さを改変.

$$v_1, \dots, v_k \quad v_{k+1}, \dots, v_n$$

$$\Rightarrow (Av_1 | Av_2 | \cdots | Av_n) = (v_1 | v_2 | \cdots | v_n) \left(\frac{*}{\overline{0}} \right)$$

$\overbrace{\qquad\qquad\qquad}^{Q^T}$

$$\exists A \in \mathbb{R}^{m \times n} \Rightarrow A^T A = Q \Lambda Q^T$$

$$\Leftrightarrow (AQ)^T (AQ) = \Lambda$$

$\Leftrightarrow AQ$ の各列が独立である.

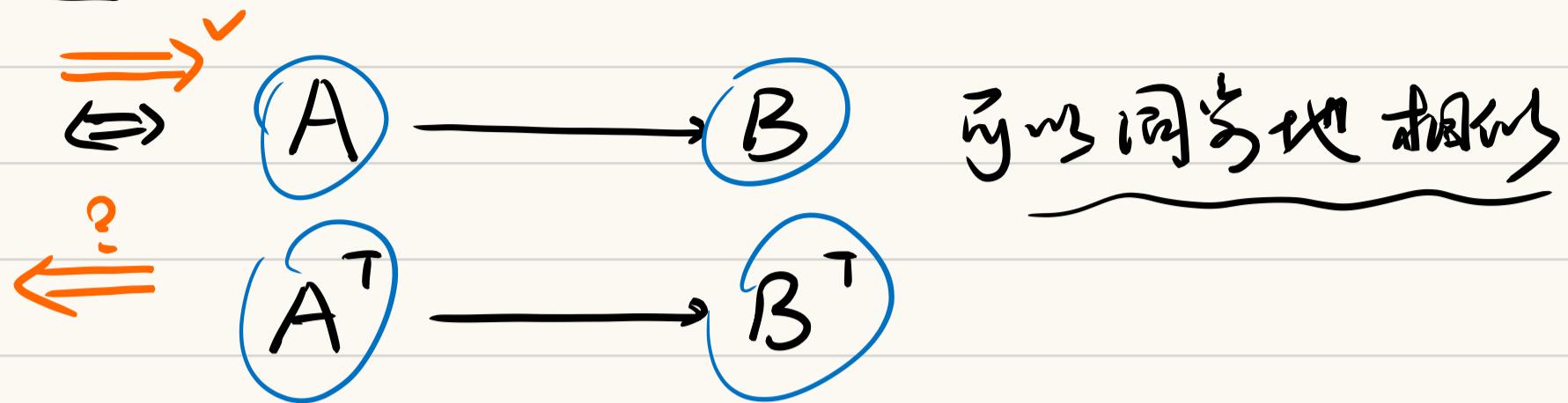
$$\text{既約 } W \cdot \begin{pmatrix} \mu_1 & & \\ & \ddots & \\ & & \mu_k \end{pmatrix}$$

$$P^{-1}AP = B \quad M^TAM = B$$

$\Leftrightarrow A, B$ の相似, 且合同

且ある正交矩陣 Q s.t. $Q^T A Q = B$.

Tlm $\exists Q \in \mathbb{R}^{n \times n}, QAQ^T = B$



$$\exists P \left\{ \begin{array}{l} PAP^{-1} = B \\ PA^TP^{-1} = B^T \end{array} \right.,$$

改：
↓ ↓
 $\Leftrightarrow \text{今 } P = Q.S$ (相乘)

$$\Rightarrow \left\{ \begin{array}{l} QSA S^{-1} Q^T = B \\ Q.SA^T S^{-1} Q^T = B^T \end{array} \right. \quad QAQ^T = B \quad \checkmark$$

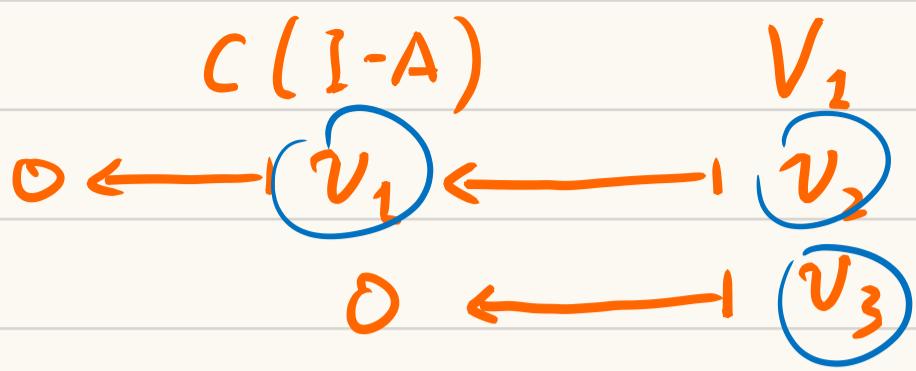
$$\Rightarrow Q(S^{-1}) A^T S Q^T = Q S A^T S^{-1} Q^T$$

$$\Rightarrow S^{-1} A^T S = S A^T S^{-1}$$

$$\Rightarrow A S^2 = S^2 A \quad \Rightarrow \quad AS = SA$$

$S = f(S^2)$

$\dim V_1 = 3, \ker N(I-A) = 2$



$$\dim [C(I-A) \cap V_1] = 1$$

Ex $A = \begin{pmatrix} 3 & 4 & 1 \\ 1 & -4 & 3 \end{pmatrix}$, 找 $U\Sigma V^T = A$

pf. (Step 1) 找 $\tilde{A}^T A$ 或 AA^T 的已知相乘向量

$$AA^T = \begin{pmatrix} 26 & -10 \\ -10 & 26 \end{pmatrix}$$

$$(26-10)^2 = 100 \\ \lambda = 16, 36.$$

$$= \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 16 & 0 \\ 0 & 36 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}^{-1}$$

$$= \underbrace{\begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \end{pmatrix}}_U \begin{pmatrix} 16 & 0 \\ 0 & 36 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \end{pmatrix}^{-1}$$

$$\Sigma = \begin{pmatrix} 4 & 0 & 0 \\ 0 & 6 & 0 \end{pmatrix}$$

(Step2) 找 V

回顧： UA 是列已正規，即

$$\text{正}\left\{\begin{array}{l} \text{的} \\ \text{向量} \end{array}\right. \quad \Sigma \cdot V^T$$

$$UA = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 3 & 4 & 1 \\ 1 & -4 & 3 \end{pmatrix}$$

$$= \frac{1}{\sqrt{2}} \begin{pmatrix} 4 & 0 & 4 \\ 2 & 8 & -2 \end{pmatrix}$$

$$= \frac{1}{\sqrt{2}} \underbrace{\begin{pmatrix} 4\sqrt{2} & 0 & 0 \\ 0 & 6\sqrt{2} & 0 \end{pmatrix}}_{\Sigma} \underbrace{\begin{pmatrix} \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \\ \frac{1}{3\sqrt{2}} & \frac{4}{3\sqrt{2}} & \frac{1}{3\sqrt{2}} \\ \frac{2}{3} & -\frac{1}{3} & -\frac{2}{3} \end{pmatrix}}_{V^T}$$

$\Leftrightarrow A \in \mathbb{C}^{n \times n}$ Jordan - Chevalley

\exists 單一形 N 實質。

D 為對角形

$$\text{s.t. } \begin{cases} A = N + D \\ ND = DN \end{cases},$$

$$\Leftrightarrow AX - XA = X, \forall X \in \mathbb{C}^n$$

