

# 正交矩阵

向量空间  $(V; +, \cdot)$

内积:  $(\cdot, \cdot): V \times V \rightarrow \mathbb{R}$

长度:  $|\alpha| := \sqrt{(\alpha, \alpha)}$

夹角:  $\langle \alpha, \beta \rangle$

$$= \arccos \frac{(\alpha, \beta)}{|\alpha| |\beta|}$$

0 or  $\pi$ : 共线

$\frac{\pi}{2}$ : 垂直 (正交)

## 定义 3.7

$n \times n$  实矩阵  $A$  被称为 **正交矩阵** 如果  $\alpha, \beta$  正交  $\Leftrightarrow (\alpha, \beta) = 0$

$$A^T A = E.$$

正交矩阵  $A$  满足以下性质:

- ▶  $A^{-1} = A^T$
- ▶  $A^{-1}$  和  $A^T$  也是正交矩阵
- ▶  $|A| = \pm 1$   $|A|^2 = |A| \cdot |A| = |A^T| \cdot |A| = |A^T A| = |E| = 1$
- ▶ 正交矩阵的乘积也是正交

$$\begin{aligned} & (A \cdot B)^T (A \cdot B) \\ &= B^T A^T A B = E. \end{aligned}$$

基  $\{\alpha_1, \dots, \alpha_n\}$

$$\text{正交基 } \{\alpha_i\} \quad (\alpha_i, \alpha_j) = \begin{cases} 0 & i \neq j \\ |\alpha_i|^2 & i = j \end{cases}$$

单强化

$$\text{单强正交基 } \{\alpha_i\} \quad (\alpha_i, \alpha_j) = \delta_{ij} = \begin{cases} 0 & i \neq j \\ 1 & i = j \end{cases}$$

如何构造单强正交基

$$\begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix} \rightarrow \begin{pmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_n \end{pmatrix} \rightarrow \begin{pmatrix} \gamma_1 \\ \gamma_2 \\ \vdots \\ \gamma_n \end{pmatrix}$$

$\gamma_1 = \frac{\beta_1}{|\beta_1|}$   
 $\gamma_2 = \frac{\beta_2}{|\beta_2|}$   
 $\gamma_n = \frac{\beta_n}{|\beta_n|}$

正交矩阵  $A = (\alpha_1, \dots, \alpha_n)$

# 定理 3.3

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记  $n$  阶实矩阵  $A = (\alpha_1, \dots, \alpha_n)$ .  $A$  是正交矩阵, 当且仅当,  $\alpha_1, \dots, \alpha_n$  为  $\mathbb{R}^n$  的一组标准正交基.

$A = (\alpha_1, \dots, \alpha_n)$  是正交.

$$\Leftrightarrow A^T A = E$$

$$\begin{pmatrix} \alpha_1^T \\ \vdots \\ \alpha_n^T \end{pmatrix}_{n \times 1} (\alpha_1 \cdots \alpha_n)_{1 \times n} = \begin{pmatrix} \alpha_1^T \alpha_1 & \cdots & \alpha_1^T \alpha_n \\ \vdots & \ddots & \vdots \\ \alpha_n^T \alpha_1 & \cdots & \alpha_n^T \alpha_n \end{pmatrix} = \begin{pmatrix} 1 & & \\ & \ddots & \\ & & 1 \end{pmatrix}$$

$$\Leftrightarrow (\alpha_i, \alpha_j) = \alpha_i^T \alpha_j = \delta_{ij}$$

$$\Leftrightarrow \{\alpha_i\} \text{ 标准正交}$$

$$\text{正交 } A = (\alpha_1, \dots, \alpha_n) = \begin{pmatrix} \beta_1 \\ \vdots \\ \beta_n \end{pmatrix} \Leftrightarrow \text{正交 } A^T = (\beta_1^T, \dots, \beta_n^T)$$

$$\Leftrightarrow \text{行向量 } \{\alpha_i\} \text{ 正交}$$

$$\Downarrow \\ \{\beta_j^T\} \text{ 正交}$$

$$\Leftrightarrow \text{列向量 } \{\beta_j\} \text{ 正交}$$

# 定理 3.4

$$A \in \mathbb{R}^{n \times n} \subseteq \mathbb{C}^{n \times n}$$

Thm 3.4: 实对称 A

存在 n 个实特征值

(重根)

## 定理 3.4

$|\lambda E - A| = 0$   
有 n 个根 (可能复数重根)

实对称矩阵的特征值均为实数

## 证明

$$\because A \in \mathbb{R}^{n \times n} \subseteq \mathbb{C}^{n \times n}$$

$$\therefore |\lambda E - A| = 0 \text{ 总有解}$$

不妨设  $\lambda_1$

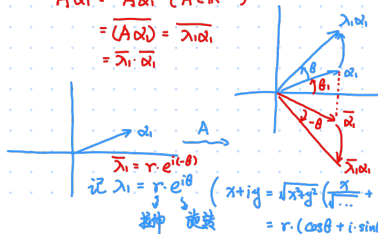
$$\text{s.t. } A\alpha_1 = \lambda_1\alpha_1$$

反证: 假设  $\lambda_1 \notin \mathbb{R}$

$$A\bar{\alpha}_1 = \bar{A}\bar{\alpha}_1 \quad (A \in \mathbb{R}^{n \times n})$$

$$= \overline{(A\alpha_1)} = \overline{\lambda_1\alpha_1}$$

$$= \bar{\lambda}_1 \cdot \bar{\alpha}_1$$



$\alpha_1$  是  $\lambda_1$  的特征向量

$$(a+ib)(a-ib) =$$

$\Leftrightarrow k\alpha_1$  也是  $\lambda_1$  的特征向量

$$\text{特别地 } k = \frac{\alpha_1^T \bar{\alpha}_1}{\alpha_1^T \alpha_1} = \frac{|\alpha_1|^2}{\alpha_1^T \alpha_1}$$

$$\Rightarrow k\alpha_1 = \frac{|\alpha_1|^2}{\alpha_1^T \alpha_1} \cdot \alpha_1$$

$$\text{记 } \alpha_1 = r \cdot e^{i\theta} \\ \Rightarrow \frac{r^2}{r^2 \cdot e^{i(2\theta)}} \cdot r \cdot e^{i\theta} \\ = r \cdot e^{-i\theta} = \bar{\alpha}_1$$

$$(\alpha_1, \bar{\alpha}_1)_A = \alpha_1^T A \bar{\alpha}_1 = \alpha_1^T \bar{\lambda}_1 \bar{\alpha}_1 = \bar{\lambda}_1 \alpha_1^T \bar{\alpha}_1$$

$\parallel A$  对称

$$= \bar{\lambda}_1 \cdot |\alpha_1|^2$$

$$(\bar{\alpha}_1, \alpha_1)_A = \bar{\alpha}_1^T A \alpha_1 = \bar{\alpha}_1^T \lambda_1 \alpha_1 = \lambda_1 |\alpha_1|^2$$

$$\Rightarrow \lambda_1 = \bar{\lambda}_1 \Rightarrow \lambda_1 \in \mathbb{R}$$

# 定理 3.5

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实对称矩阵中, 属于不同特征值的特征向量必正交

回顾:  $A$  特征值  $\lambda_i \neq \lambda_j$

$\Rightarrow$  特征向量  $\alpha_i, \alpha_j$  线性无关

Pf:  $(\alpha_i, \alpha_j)_A = \alpha_i^T A \alpha_j = \alpha_i^T \lambda_j \alpha_j = \lambda_j \alpha_i^T \alpha_j$   
           $\parallel (A \text{ 对称})$

$(\alpha_j, \alpha_i)_A = \alpha_j^T A \alpha_i = \alpha_j^T \lambda_i \alpha_i = \lambda_i \alpha_j^T \alpha_i$

$\Rightarrow \lambda_j (\alpha_i, \alpha_j) = \lambda_i (\alpha_i, \alpha_j)$

又  $\lambda_i \neq \lambda_j$

$\therefore (\alpha_i, \alpha_j) = 0 \text{ if } i \neq j.$

# 定理 3.6

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对任意  $n$  阶实对阵矩阵  $A$ , 则存在正交矩阵  $T$ ,  
s.t.  $T^{-1}AT$  为对角矩阵.

$A$  是可对角化 ( $A \sim \Lambda$ )

$\exists$  可逆  $P = \{\alpha_1, \dots, \alpha_n\}$

s.t.  $P^{-1}AP = \text{diag}(\lambda_1, \dots, \lambda_n)$

Pf: 矩阵  $A_0 = A$

是某个线性映射  $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$

在  $\{e_1, \dots, e_n\}$  下的表示

目标: 构造  $\{\alpha_1, \dots, \alpha_n\}$

s.t.  $\alpha_i$  是  $A$  的特征向量

$(\alpha_i, \alpha_j) = \delta_{ij}$

STEP 1:  $\alpha_1$  存在的

$(A \in \mathbb{R}^{n \times n} \subseteq \mathbb{C}^{n \times n})$

总有  $\lambda_1 \in \mathbb{C}$  特征值

又  $\lambda_1 \in \mathbb{R} \therefore \exists \alpha_1 \in \mathbb{R}^n$  s.t.  $A\alpha_1 = \lambda_1\alpha_1$

$\therefore \alpha_1$  与  $\{e_i\}$  的  
一个向量线性相关

$\therefore$  不妨设  $\alpha_1$  和  $\{e_2, \dots, e_n\}$   
线性无关

$\therefore \{\alpha_1, e_2, \dots, e_n\}$

线性独立  $\rightarrow \{\alpha_1, e_2^{(1)}, \dots, e_n^{(1)}\}$

$\leadsto T_1 = (\alpha_1, e_2^{(1)}, \dots, e_n^{(1)})$

$T_1^{-1}AT_1 = T_1^T AT_1$

$= \begin{pmatrix} \alpha_1^T \\ e_2^{(1)T} \\ \vdots \\ e_n^{(1)T} \end{pmatrix} A (\alpha_1, e_2^{(1)}, \dots, e_n^{(1)})$

$= \begin{pmatrix} \alpha_1^T A \alpha_1 & \alpha_1^T A e_2^{(1)} & \dots & \alpha_1^T A e_n^{(1)} \\ e_2^{(1)T} A \alpha_1 & e_2^{(1)T} A e_2^{(1)} & \dots & e_2^{(1)T} A e_n^{(1)} \\ \vdots & \vdots & \ddots & \vdots \\ e_n^{(1)T} A \alpha_1 & e_n^{(1)T} A e_2^{(1)} & \dots & e_n^{(1)T} A e_n^{(1)} \end{pmatrix}$

(续)  $= \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & & & \\ \vdots & & A_2 & \\ 0 & & & \end{pmatrix}$

Claim:  $A_2$  还是实对称

STEP 2:  $T_2 = \begin{pmatrix} 1 & & \\ & \ddots & \\ & & T_2^{-1} \end{pmatrix}$

$T_2 T_1 = (\alpha_1, \alpha_2, e_3^{(2)}, \dots, e_n^{(2)})$

STEP i:  $T_i = \begin{pmatrix} E_{i-1} & \\ & T_i^{-1} \end{pmatrix}$

STEP n:  $\begin{pmatrix} & & \\ & & \\ & & T_n^{-1} \end{pmatrix}$

$T = T_n \cdots T_2 T_1$

$\Rightarrow T^{-1}AT = \begin{pmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \ddots \\ & & & \lambda_n \end{pmatrix}$

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# 例题 3.3

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设

$$A = \begin{pmatrix} 1 & 2 & 2 \\ 2 & 1 & 2 \\ 2 & 2 & 1 \end{pmatrix}.$$

求正交矩阵  $T$ , 使得  $T^{-1}AT$  为对角矩阵.

STEP 1: 求特征值

$$|\lambda E - A| = 0$$

$$\begin{vmatrix} 1-\lambda & 2 & 2 \\ 2 & 1-\lambda & 2 \\ 2 & 2 & 1-\lambda \end{vmatrix} = 0$$

- $\lambda = -1$
- $\lambda = 5$

STEP 2: 求特征向量

$$\bullet \lambda = -1$$

$$x_1 + x_2 + x_3 = 0$$

$$\alpha_1 = \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}, \alpha_2 = \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}$$

$$\bullet \lambda = 5$$

$$\begin{pmatrix} -2 & 1 & 1 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = 0$$

$$\hookrightarrow \alpha_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

STEP 3: 对  $\alpha_1, \alpha_2, \alpha_3$  单位正交化

$$\bullet \beta_1 = \alpha_1 \xrightarrow{\text{单位化}} \gamma_1 = \frac{\beta_1}{|\beta_1|} = \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix}$$

$$\begin{aligned} \bullet \beta_2 &= \alpha_2 - (\alpha_2, \gamma_1) \gamma_1 \\ &= \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} - (-1, 0, 1) \cdot \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix} \cdot \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{pmatrix} = \dots \\ &= \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} - \begin{pmatrix} -\frac{1}{2} \\ -\frac{1}{2} \\ 0 \end{pmatrix} = \begin{pmatrix} -\frac{1}{2} \\ \frac{1}{2} \\ 1 \end{pmatrix} \end{aligned}$$

$$\beta_3 = \alpha_3 - (\alpha_3, \gamma_1) \gamma_1 - (\alpha_3, \gamma_2) \gamma_2$$

$$\gamma_3 = \frac{\beta_3}{|\beta_3|}$$

$$T = (\gamma_1, \gamma_2, \gamma_3)$$

## 例题 3.4

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求证: 如果实对称  $A \sim B$ , 那么存在正交矩阵  $T$   
s.t.

$$T^{-1}AT = B.$$