

证明 重开

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1. 比较两种变换

① 子空间 V 中的两组基.

$(\alpha_1, \dots, \alpha_n), (\beta_1, \dots, \beta_n)$ 可相互表示.

设有 $\beta_i = k_{i1}\alpha_1 + \dots + k_{in}\alpha_n$.

$$\text{则有 } \begin{cases} \beta_1 = k_{11}\alpha_1 + \dots + k_{1n}\alpha_n \\ \vdots \\ \beta_n = k_{n1}\alpha_1 + \dots + k_{nn}\alpha_n \end{cases}$$

$$\text{写为 } \begin{pmatrix} \beta_1 \\ \vdots \\ \beta_n \end{pmatrix} = \begin{pmatrix} k_{11} & \dots & k_{1n} \\ \vdots & \ddots & \vdots \\ k_{n1} & \dots & k_{nn} \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix}$$

$$\text{也即 } (\beta_1, \dots, \beta_n) = (\alpha_1, \dots, \alpha_n) \begin{pmatrix} k_{11} & \dots & k_{1n} \\ \vdots & \ddots & \vdots \\ k_{n1} & \dots & k_{nn} \end{pmatrix} = (\alpha_1, \dots, \alpha_n) T.$$

对原向量 η ,

$$\text{记为 } \eta = (\alpha_1, \dots, \alpha_n) \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = (\beta_1, \dots, \beta_n) \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

$$= (\alpha_1, \dots, \alpha_n) X = (\beta_1, \dots, \beta_n) Y,$$

X, Y 为坐标, 则有 $X = TY$.

② 线性变换的两种坐标.

$A(\alpha_1, \dots, \alpha_n)$ 是对 $(\alpha_1, \dots, \alpha_n)$ 的一种线性变换.

$$\text{其中 } A(\alpha_i) = k_{i1}\alpha_1 + \dots + k_{in}\alpha_n = (\alpha_1, \dots, \alpha_n) \begin{pmatrix} k_{i1} \\ \vdots \\ k_{in} \end{pmatrix}.$$

$$\text{则 } A(\alpha_1, \dots, \alpha_n) = (\alpha_1, \dots, \alpha_n) \begin{pmatrix} k_{11} & \dots & k_{1n} \\ \vdots & \ddots & \vdots \\ k_{n1} & \dots & k_{nn} \end{pmatrix} = (\alpha_1, \dots, \alpha_n) T.$$

$$\text{设原向量 } \eta = (\alpha_1, \dots, \alpha_n) \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix},$$

$$\text{变换后 } A(\eta) = (A\alpha_1, \dots, A\alpha_n) \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}.$$

$$\text{则对 } A(\eta) = (\alpha_1, \dots, \alpha_n) T \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = (\alpha_1, \dots, \alpha_n) \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

即变换前后坐标 $TX = Y$.

2. 可对角化

3. 实对称矩阵

① 定义.

$$\text{基 } \alpha_1, \dots, \alpha_n \text{ 下, } \alpha = (\alpha_1, \dots, \alpha_n) X = \sum_{i=1}^n \alpha_i x_i$$

$$\beta = (\alpha_1, \dots, \alpha_n) Y = \sum_{i=1}^n \alpha_i y_i$$

$$(\alpha, \beta) = \left(\sum_{i=1}^n \alpha_i x_i, \sum_{j=1}^n \alpha_j y_j \right)$$

$$= \sum_{i=1}^n x_i \sum_{j=1}^n y_j (\alpha_i, \alpha_j)$$

$$= X^T G Y, \text{ 取 } G = ((\alpha_i, \alpha_j))_{n \times n}.$$

② 坐标变换.

$$\text{基 } (\beta_1, \dots, \beta_n) \text{ 下, } \alpha = (\beta_1, \dots, \beta_n) \tilde{X} = \sum_{i=1}^n \beta_i \tilde{x}_i$$

$$\beta = (\beta_1, \dots, \beta_n) \tilde{Y} = \sum_{i=1}^n \beta_i \tilde{y}_i.$$

$$(\alpha, \beta) = \tilde{X}^T \tilde{G} \tilde{Y}, \tilde{G} = ((\beta_i, \beta_j))_{n \times n}.$$

变换关系 $(\beta_1, \dots, \beta_n) = (\alpha_1, \dots, \alpha_n) T$.

则对坐标, 如 $\alpha = (\alpha_1, \dots, \alpha_n) X = (\beta_1, \dots, \beta_n) \tilde{X}$,

$$\alpha = (\alpha_1, \dots, \alpha_n) T \tilde{X} = (\alpha_1, \dots, \alpha_n) X, X = T \tilde{X}, X^T = \tilde{X}^T T^T$$

$$\therefore (\alpha, \beta) = \tilde{X}^T \tilde{G} \tilde{Y} = X^T T^T G T Y = (\tilde{X}^T T^T) G (T Y)$$

$$= \tilde{X}^T (T^T G T) \tilde{Y}, \therefore \tilde{G} = T^T G T.$$

这又 $\tilde{G} = T^T G T$ 总是相容.

③ 实对称矩阵可逆.

证明: 可逆 $\Leftrightarrow \text{rank } G = n \Leftrightarrow$ 行秩, 列秩为 n

\Leftrightarrow 各行 / 列 线性无关.

$$G = \begin{pmatrix} (\alpha_1, \alpha_1) & (\alpha_1, \alpha_2) & \dots & (\alpha_1, \alpha_n) \\ (\alpha_2, \alpha_1) & (\alpha_2, \alpha_2) & \dots & (\alpha_2, \alpha_n) \\ \vdots & \vdots & \ddots & \vdots \\ (\alpha_n, \alpha_1) & (\alpha_n, \alpha_2) & \dots & (\alpha_n, \alpha_n) \end{pmatrix}$$

验证各行向量

$$k_1[(\alpha_1, \alpha_1) \ (\alpha_1, \alpha_2) \ \dots \ (\alpha_1, \alpha_n)] + k_2[(\alpha_2, \alpha_1) \ \dots \ (\alpha_2, \alpha_n)]$$

$$+ \dots + k_n[(\alpha_n, \alpha_1) \ (\alpha_n, \alpha_2) \ \dots \ (\alpha_n, \alpha_n)] = 0 \text{ 的 } k_i \text{ 的 } \forall \text{ 解.}$$

行线性性

$$\Leftrightarrow (k_1(\alpha_1, \alpha_1) \ k_1(\alpha_1, \alpha_2) \ \dots \ k_1(\alpha_1, \alpha_n)) + \dots$$

$$+ (k_n(\alpha_n, \alpha_1) \ k_n(\alpha_n, \alpha_2) \ \dots \ k_n(\alpha_n, \alpha_n)) = 0.$$

由线性性

$$\Leftrightarrow (k_1(\alpha_1, \alpha_1) \ k_1(\alpha_1, \alpha_2) \ \dots \ k_1(\alpha_1, \alpha_n)) + \dots$$

$$+ (k_n(\alpha_n, \alpha_1) \ k_n(\alpha_n, \alpha_2) \ \dots \ k_n(\alpha_n, \alpha_n)) = 0.$$

行线性性

$$\Leftrightarrow (k_1(\alpha_1, \alpha_1) \ k_1(\alpha_1, \alpha_2) \ \dots \ k_1(\alpha_1, \alpha_n)) = 0.$$

$$\text{记 } \alpha = \sum_{i=1}^n k_i \alpha_i, (\alpha, \alpha_1) \ (\alpha, \alpha_2) \ \dots \ (\alpha, \alpha_n) = 0.$$

$$\text{则 } \sum_{i=1}^n k_i (\alpha, \alpha_i) = 0, \quad k_i (\alpha, \alpha_i) = 0.$$

$$\therefore \sum_{i=1}^n k_i (\alpha, \alpha_i) = 0, \quad (\alpha, \sum_{i=1}^n k_i \alpha_i) = 0, \quad (\alpha, \alpha) = 0.$$

$$\text{由积正定性} \Rightarrow \alpha = 0.$$

$$\text{即 } \sum_{i=1}^n k_i \alpha_i = 0, \quad \alpha_i \neq 0, \quad \therefore k_i = 0.$$

$$\therefore G \text{ 各行向量线性无关, 可逆.}$$

4. 标准正交基

① 定义.

正交向量组: 两两正交的那组向量.

正交基: 正交向量组成为一组基.

标准正交基: 单位向量构成的正交基.

② 性质: $\alpha_1, \dots, \alpha_n$ 两两正交, 则构成的正交向量组线性无关 (可构成基).

证明: 对方程 $\lambda_1 \alpha_1 + \lambda_2 \alpha_2 + \dots + \lambda_n \alpha_n = 0$.

两边同时取与 α_i 的内积,

$$\text{得到 } \lambda_i (\alpha_i, \alpha_i) + \lambda_1 (\alpha_1, \alpha_i) + \dots = 0.$$

两两正交, 故只有 $\lambda_i (\alpha_i, \alpha_i) = 0$,

其中 $\alpha_i \neq 0, (\alpha_i, \alpha_i) \neq 0$, 故 $\lambda_i = 0$,

对 $\forall i \in \{1, \dots, n\}$ 成立.

$$\text{即 } \lambda_1 = \lambda_2 = \dots = \lambda_n = 0.$$

③ 存在性: 取 n -组基, 都可构造.

求标准正交基的方法: Schmidt 正交化.

① 单位化: $\beta_i = \frac{\alpha_i}{|\alpha_i|}, \alpha_i = |\alpha_i| \beta_i$.

正交化: 作垂直形成 $\tilde{\beta}_i$,

$$|\tilde{\beta}_i| = |\beta_i| \cos \langle \tilde{\beta}_i, \beta_i \rangle$$

$$= |\beta_i| \frac{(\tilde{\beta}_i, \beta_i)}{|\tilde{\beta}_i| |\beta_i|}$$

$$\tilde{\beta}_i = |\tilde{\beta}_i| \frac{\tilde{\beta}_i}{|\tilde{\beta}_i|} = \frac{(\tilde{\beta}_i, \beta_i)}{(\tilde{\beta}_i, \tilde{\beta}_i)} \tilde{\beta}_i$$

$$\beta_i - \tilde{\beta}_i = \beta_i - \frac{(\tilde{\beta}_i, \beta_i)}{(\tilde{\beta}_i, \tilde{\beta}_i)} \tilde{\beta}_i.$$

$$\tilde{\beta}_i = \beta_i - \frac{(\tilde{\beta}_i, \beta_i)}{(\tilde{\beta}_i, \tilde{\beta}_i)} \tilde{\beta}_i.$$

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5. 正交变换

① 正交变换性质.

② 正交变换性质.

6. 对称变换.

① 定义.

② 对称矩阵.

③ 可对角化.