

6. 证: 记 A 为秩等于 r 的对称矩阵. 则 A 合同相似于对角矩阵, 即 $A = C^T \text{diag}\{a_1, \dots, a_r, 0, \dots, 0\} C$.
 取 r 个秩为 1 的矩阵为 $C^T \text{diag}\{a_1, 0, \dots, 0\} C$, $C^T \text{diag}\{0, a_2, 0, \dots, 0\} C, \dots, C^T \text{diag}\{0, \dots, 0, a_r, 0, \dots, 0\} C$.
 则 $A = C^T a_1 E_{11} C + C^T a_2 E_{22} C + \dots + C^T a_r E_{rr} C$. \square

7. 记 $\alpha_i = (a_{i1}, a_{i2}, \dots, a_{in})^T$ ~~$X = (x_1, x_2, \dots, x_n)^T$~~ $X = (x_1, x_2, \dots, x_n)^T$ $i = 1, 2, \dots, s$.
 则 $f = (X^T \alpha_1, X^T \alpha_2, \dots, X^T \alpha_s) \cdot (X^T \alpha_1, X^T \alpha_2, \dots, X^T \alpha_s)^T$
 $= X^T (\alpha_1, \alpha_2, \dots, \alpha_s) \cdot (\alpha_1^T, \alpha_2^T, \dots, \alpha_s^T)^T X$
 $= X^T A^T A X = \cancel{(AX)^T AX}$

则 f 的秩 = $r(A^T A)$

由于 $Ax = 0 \Rightarrow A^T Ax = 0$, $\neg A^T Ax = 0 \Rightarrow X^T A^T Ax = 0 \Rightarrow (AX)^T Ax = 0 \Rightarrow Ax = 0$

故 $r(A^T A) = r(A)$ \square

(\Rightarrow) 若 $f = X^T A X$ 可以分解成 $f = gh$, 其中 $g = \alpha^T X$, $h = \beta^T X$. α, β, X 都是列向量.

则 $f = gh = \alpha^T X \cdot \beta^T X = X^T \alpha \beta^T X$. 由于 $\alpha \beta^T$ 与 $\beta \alpha^T$ 有相同秩.

故 $r(A) = r(\alpha \beta^T) = 1$.

(\Leftarrow) 若 $r(A) = 1$ 则 A 可以写成 2 个列向量 α, β 的形式 $A = \alpha \beta^T$ 故 $g = \alpha^T X$, $h = \beta^T X$, $f = gh$

若 $r(A) = 2$ 且符号差为 0, 故 $A = C^T \text{diag}\{1, -1, 0, \dots, 0\} C$, 令 $Cx = y$

$f = X^T C^T \text{diag}\{1, -1, 0, \dots, 0\} C X$. 记 $y = Cx$. 则 $f = y^T \text{diag}\{1, -1, 0, \dots, 0\} y = y_1^2 - y_2^2$

$= (y_1 - y_2)(y_1 + y_2)$ 是 2 个一次多项式之积. \square

4. 设 $F(k) = (kX_1 + (1-k)X_2)' A (kX_1 + (1-k)X_2)$, 这是一个关于 k 的多项式, $k \in [0, 1]$

$$F(0) = X_2' A X_2 < 0, F(1) = X_1' A X_1 > 0 \Rightarrow \exists k_0 \in (0, 1), \text{s.t. } F(k_0) = 0$$

取 $X_0 = k_0 X_1 + (1-k_0)X_2$ 即可 \square

5. $L_i = \alpha_i^T X$. 其中 $\alpha_i = (\alpha_{i1}, \alpha_{i2}, \dots, \alpha_{in})^T$, $X = (x_1, x_2, \dots, x_n)^T$

$$\begin{aligned} \text{则 } f &= \underbrace{(l_1^2 + \dots + l_p^2 - l_{p+1}^2 - \dots - l_{p+q}^2)}_{\substack{\text{正} \\ \text{负}}} = (l_1, l_2, \dots, l_{p+q})^T \text{diag}\{1, \dots, 1, -1, \dots, -1\} \underbrace{(l_1, \dots, l_{p+q})}_{\substack{\text{正} \\ \text{负}}} \\ &= (\alpha_1^T X, \alpha_2^T X, \dots, \alpha_{p+q}^T X)^T \text{diag}\{1, \dots, 1, -1, \dots, -1\} (\alpha_1^T X, \alpha_2^T X, \dots, \alpha_{p+q}^T X) \\ &= X^T (\alpha_1, \alpha_2, \dots, \alpha_{p+q}) \text{diag}\{1, \dots, 1, -1, \dots, -1\} (\alpha_1^T, \alpha_2^T, \dots, \alpha_{p+q}^T)^T X \end{aligned}$$

6. 由规范型结论: f 正惯性指数 $\leq p$, 负惯性指数 $\leq q$ \square

6. 不妨设 $f = X^T A X$ 为规范型, 否则进行基的子变换即可, 记 $f = X^T A X$ 的正惯性指数为 p , 则 $\dim M = "A$ 对角线上为 0 的个数" $= n - p$. \square

1. (1). $f = 99x_1^2 - 12x_1x_2 + 48x_1x_3 + 130x_2^2 - 60x_2x_3 + 71x_3^2$

$$= \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}^T \begin{pmatrix} 99 & -6 & 24 \\ -6 & 130 & -30 \\ 24 & -30 & 71 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

由于 $|99| > 0$, $\begin{vmatrix} 99 & -6 \\ -6 & 130 \end{vmatrix} > 0$, $\begin{vmatrix} 99 & -6 & 24 \\ -6 & 130 & -30 \\ 24 & -30 & 71 \end{vmatrix} = 755874 > 0$

故 f 正定 (顺序主子式都为正) \square

1. (3). $f = \sum_{i=1}^n x_i^2 + \sum_{1 \leq i < j \leq n} x_i x_j = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}^T \begin{pmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$

考虑 $D_n = \begin{vmatrix} 1 & \frac{1}{2} & \dots & \frac{1}{2} \\ \frac{1}{2} & 1 & & \\ \vdots & & \ddots & \\ \frac{1}{2} & & & 1 \end{vmatrix} = \begin{vmatrix} 1 & \frac{1}{2} & \dots & \frac{1}{2} \\ \frac{1}{2} & 1 & & \\ \vdots & & \ddots & \\ \frac{1}{2} & & & 1 \end{vmatrix} = \begin{vmatrix} \frac{1}{2} & 0 & \dots & 0 \\ \frac{1}{2} & \frac{1}{2} & & \\ \vdots & & \ddots & \\ \frac{1}{2} & & & \frac{1}{2} \end{vmatrix} = \frac{1}{2^n} > 0$

故 f 所有顺序主子式 > 0 . $\Rightarrow f$ 正定 \square

2. (1) $f = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}^T \begin{pmatrix} 1 & t \\ t & 5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \Rightarrow |1| = 1 > 0, \begin{vmatrix} 1 & t \\ t & 5 \end{vmatrix} = 5 - t^2$

正定 $\Leftrightarrow 1 - t^2 > 0 \wedge 5 - t^2 > 0 \Leftrightarrow -1 < t < 1 \wedge 0 < t < \frac{\sqrt{5}}{2} \Leftrightarrow 0 < t < \frac{\sqrt{5}}{2}$ \square

4. 设 A 的所有特征值为 $\lambda_1, \dots, \lambda_n$. 取 $t = \max\{\lambda_1, \dots, \lambda_n\}$. $t + \min\{\lambda_1, \dots, \lambda_n\} > 0$ 即有 $tE + A$ 的所有特征值 > 0 . $\Rightarrow A$ 正定 \square

7. 不妨设 A 为反证. 若 $\forall n$ 维向量 X , 都有 $X^T A X > 0$. 又 A 对称. 则 A 半正定. 则 A 的所有特征值 ≥ 0 , 则 $|A| \geq 0$ 这与 $|A| < 0$ 矛盾! \square

8. A, B 正定, 不妨设 $A = I_n$. 因为条件和结论在变换 $A \rightarrow C^T A C = I_n, B \rightarrow C^T B C$ 下保持不变. 因为 B 正定, B 正定 $\Leftrightarrow B$ 的特征值全 $> 0 \Rightarrow B + I_n$ 特征值全 $> 0 \Leftrightarrow A + B$ 正定 \square

