

3.58: 证明 Schur 引理 $\square \Leftrightarrow \forall z \in \mathbb{R}^{n \times k}, z^T (C - B^T A^{-1} B) z \leq 0$

$\Rightarrow X_1 = \begin{bmatrix} A_1 & B_1 \\ B_1^T & C_1 \end{bmatrix} \quad X_2 = \begin{bmatrix} A_2 & B_2 \\ B_2^T & C_2 \end{bmatrix}$

$\Rightarrow X \in S_{++}^n \Rightarrow \begin{bmatrix} A & B \\ B^T & C \end{bmatrix} \in S_{++}^{n+k}$

证明 $\forall \lambda < 1, X_1, X_2 \in S_{++}^n, \forall z \in \mathbb{R}^n$

$$z^T (C_1 + (1-\lambda)C_2 - (\lambda B_1 + (1-\lambda)B_2)^T (\lambda A_1 + (1-\lambda)A_2)^{-1} (\lambda B_1 + (1-\lambda)B_2)) z$$

$$= \lambda z^T C_1 + (1-\lambda) z^T C_2 - (\lambda B_1 + (1-\lambda)B_2)^T (\lambda A_1 + (1-\lambda)A_2)^{-1} (\lambda B_1 + (1-\lambda)B_2) z$$

证明 $\forall z, z^T B^T A^{-1} B z \leq 0$

$A = D^T D, A^{-1} = D^{-T} D^{-1}$

$B^T A^{-1} B = B^T D^{-T} D^{-1} B$

$z^T B^T A^{-1} B z = (B D^{-1} z)^T (B D^{-1} z) \geq 0$

$\therefore B^T A^{-1} B \in S_{++}^k$

$\therefore A \in S_{++}^n, B \in \mathbb{R}^{n \times k}$

$\therefore A^{-1} \in S_{++}^n, D \in \mathbb{R}^{n \times n}$

$\therefore B^T A^{-1} B \in S_{++}^k$

$\therefore B^T A^{-1} B \in S_{++}^k$

$z^T B^T A^{-1} B z \leq 0$

$\Rightarrow B^T A^{-1} B \in S_{++}^k$

$\Rightarrow A \in S_{++}^n, B \in \mathbb{R}^{n \times k}$

①: $\forall z \neq 0, z^T B^T A^{-1} B z > 0$

$= (Bz)^T A^{-1} (Bz)$

$\in S_{++}^k$

$\therefore A^{-1} \in S_{++}^n$

$z^T B^T A^{-1} B z > 0$

$B^T A^{-1} B \in S_{++}^k$

$z^T B^T A^{-1} B z \leq 0$

$\Rightarrow B^T A^{-1} B \in S_{++}^k$

$\Rightarrow A \in S_{++}^n, B \in \mathbb{R}^{n \times k}$

②: $f(z, X) = z^T B^T A^{-1} B z$ 在 $\mathbb{R}^{n \times k} \times S_{++}^n$ 上凸

必有 $\forall z, f(z, X)$ 在 S_{++}^n 上凸

epi $f = \{(z, X, t) \mid X \succ 0, z^T B^T A^{-1} B z \leq t\}$

$= \{(z, X, t) \mid \begin{bmatrix} B^T A^{-1} B & z \\ z^T & t \end{bmatrix} \preceq 0, X \succ 0\}$

\therefore epi f 为凸集

$\therefore f(z, X)$ 为凸函数

$\forall z, z^T B^T A^{-1} B z$ 为凸函数, Schur 引理 \square 得证, \square



3-56
5.2.6



可解: $\{(1, 0)\}$
即 $\partial f = \{1\}$

$x^* = (1, 0)$
 $p = 1$

2) $\nabla(x_1^2 + x_2^2) = (2x_1, 2x_2)$

$\nabla((x_1-1)^2 + (x_2-1)^2 - 1) = (2x_1-2, 2x_2-2)$

$\nabla((x_1+1)^2 + (x_2+1)^2 - 1) = (2x_1+2, 2x_2+2)$

构造 Lagrangian:

$(2x_1^*, 2x_2^*) + \lambda_1 (2x_1^* - 2, 2x_2^* - 2) + \lambda_2 (2x_1^* + 2, 2x_2^* + 2) = 0$

即:
$$\begin{cases} (\lambda_1^* + \lambda_2^* + 1)x_1 = 0 & \text{--- (1)} \\ (\lambda_1^* + \lambda_2^* + 1)x_2 = 0 & \text{--- (2)} \\ (x_1^* - 1)^2 + (x_2^* - 1)^2 = 1 & \text{--- (3)} \\ (x_1^* + 1)^2 + (x_2^* + 1)^2 = 1 & \text{--- (4)} \\ \lambda_1^* \geq 0 & \text{--- (5)} \\ \lambda_2^* \geq 0 & \text{--- (6)} \\ \lambda_1^* ((x_1 - 1)^2 + (x_2 - 1)^2 - 1) = 0 & \text{--- (7)} \\ \lambda_2^* ((x_1 + 1)^2 + (x_2 + 1)^2 - 1) = 0 & \text{--- (8)} \end{cases}$$

$\lambda_1 + \lambda_2 = \lambda_1 + \lambda_2 + 1$

$\begin{cases} x_1 = 1, x_2 = 0 \\ \lambda_1 = 0, \lambda_2 = 0 \end{cases}$
 $\lambda_1 = 0$
 $\therefore x_2 = 0, \text{由 (2) } \lambda_1^* = \lambda_2^*$

由 (1), (4), (7), (8) 可得
由 (1), (5), (6) 可得
 $2\lambda_1 + 1 = 2\lambda_1 = 0$, 不存在
故 1

(不存在)





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5.27: $\|Ax-b\|_2^2 = (Ax-b)^T(Ax-b)$

$$\frac{\partial \|Ax-b\|_2^2}{\partial x} = x^T A^T A x - \cancel{x^T A^T b} - b^T A x + b^T b$$

$$= 2A^T A x - A^T b$$

$$\frac{\partial (Gx-h)}{\partial x} = G^T$$

① KKT条件:

$$\begin{cases} Gx^* - h = 0 & \text{--- ①} \\ u^* \geq 0 & \text{--- ②} \quad (u \in \mathbb{R}^p) \\ 2A^T A x^* - 2A^T b + G^T u^* = 0 & \text{--- ③} \\ \cancel{2(Ax^* - b)^T A u^* + G^T u^* = 0} \\ \cancel{2(A^T A x^* - A^T b)^T u^*} \\ \cancel{2(Ax^* - b)^T A u^* = 0} \\ \cancel{(2A^T A x^* - 2A^T b - G^T u^*)^T u^* = 0} \end{cases}$$

②: $2A^T A x^* = 2A^T b - G^T u^*$

$r(A) = n, n \leq m$
 $A \in \mathbb{R}^{m \times n}$

$$x^* = \frac{1}{2} (A^T A)^{-1} (2A^T b - G^T u^*)$$

代入①有

$$\frac{G}{2} (A^T A)^{-1} (2A^T b - G^T u^*) - h = 0$$

$$\frac{G}{2} (A^T A)^{-1} G^T u^* = h - G(A^T A)^{-1} A^T b$$

$r(G) = p, p \leq n,$

$$\therefore G(A^T A)^{-1} G^T \in \mathbb{R}^{p \times p}$$

$$u^* = 2(G(A^T A)^{-1} G^T)^{-1} (h - G(A^T A)^{-1} A^T b)$$

$$x^* = (A^T A)^{-1} (2A^T b - G^T (G(A^T A)^{-1} G^T)^{-1} (h - G(A^T A)^{-1} A^T b))$$



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5.28: KKT 条件:

$$A^* \lambda \leq 0$$

$$\begin{bmatrix} 47 \\ 93 \\ 17 \\ 93 \end{bmatrix} + \begin{bmatrix} 1 & -1 & 0 & -6 & 1 \\ -6 & 2 & 3 & -1 & 6 \\ 1 & 7 & -6 & 2 & -1 \\ 3 & 1 & -1 & 12 & -3 \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \\ \lambda_3 \\ \lambda_4 \\ \lambda_5 \end{bmatrix} = 0 \quad \text{--- (1)}$$

$$\begin{bmatrix} \lambda_1^* \\ \lambda_2^* \\ \lambda_3^* \\ \lambda_4^* \\ \lambda_5^* \end{bmatrix} \leq 0 \quad \text{--- (2)}$$

$$\begin{bmatrix} -1 & -6 & 1 & 3 \\ -1 & -2 & 7 & 1 \\ 0 & 3 & -6 & -1 \\ -6 & -11 & -2 & 12 \\ 1 & 6 & -1 & -3 \end{bmatrix} \begin{bmatrix} x_1^* \\ x_2^* \\ x_3^* \\ x_4^* \end{bmatrix} - \begin{bmatrix} -3 \\ 5 \\ -6 \\ -7 \end{bmatrix} \leq 0 \quad \text{--- (3)}$$

由 (1), 有

$$\begin{bmatrix} 1 & 1 & 0 & 6 & -1 \\ 0 & 1 & \frac{1}{2} & 3 & 0 \\ 0 & 0 & 1 & 2 & 0 \\ 0 & 0 & 0 & 11 & 0 \end{bmatrix} \begin{bmatrix} \lambda_1^* \\ \lambda_2^* \\ \lambda_3^* \\ \lambda_4^* \\ \lambda_5^* \end{bmatrix} = \begin{bmatrix} 47 \\ 24 \\ 16 \\ 77 \end{bmatrix}$$

$$\begin{bmatrix} 1 & -6 & 1 & 3 \\ -1 & -2 & 7 & 1 \\ 0 & 3 & -6 & -1 \\ -6 & -11 & -2 & 12 \\ 1 & 6 & -1 & -3 \end{bmatrix} \begin{bmatrix} x_1^* \\ x_2^* \\ x_3^* \\ x_4^* \end{bmatrix} - \begin{bmatrix} -3 \\ 5 \\ -6 \\ -7 \end{bmatrix} = 0$$

$$\begin{bmatrix} \lambda_1^* \\ \lambda_2^* \\ \lambda_3^* \\ \lambda_4^* \\ \lambda_5^* \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \\ 7 \\ x_5^* \end{bmatrix}$$

$\lambda_1^* \geq 0$
 \therefore 必有 $\lambda_1^*, \lambda_2^*, \lambda_3^*, \lambda_4^* > 0$

有

$$\begin{bmatrix} -1 & -6 & 1 & 3 \\ -1 & -2 & 7 & 1 \\ 0 & 3 & -6 & -1 \\ -6 & -11 & -2 & 12 \end{bmatrix} \begin{bmatrix} x_1^* \\ x_2^* \\ x_3^* \\ x_4^* \end{bmatrix} = \begin{bmatrix} -3 \\ 5 \\ -6 \\ -7 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 6 & -1 & -3 \\ 0 & 1 & \frac{1}{2} & -\frac{1}{2} \\ 0 & 0 & 1 & -\frac{1}{24} \\ 0 & 0 & 0 & \frac{143}{24} \end{bmatrix} \begin{bmatrix} x_1^* \\ x_2^* \\ x_3^* \\ x_4^* \end{bmatrix} = \begin{bmatrix} -3 \\ 2 \\ \frac{2}{24} \\ \frac{143}{24} \end{bmatrix}$$

得到最优解

$$\begin{bmatrix} x_1^* \\ x_2^* \\ x_3^* \\ x_4^* \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}$$



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