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耦合矩阵方程的双对称最小二乘解及其最佳逼近^①

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摘要: 由于用矩阵分解的方法求解耦合矩阵方程的双对称最小二乘解比较复杂, 所以用迭代算法来求解该方程的双对称最小二乘解并证明了算法的收敛性, 同时, 极小范数解也可通过选取特殊的初始矩阵得到. 利用此算法还可得到任意给定矩阵组的最佳逼近双对称解组.

关键词: 矩阵方程; 双对称最小二乘解组; 极小范数解组; 最佳逼近解组

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$m \times n$ 实矩阵和双对称矩阵集合分别用 $\mathbb{R}^{m \times n}$ 和 $\mathbf{BSR}^{m \times n}$ 表示. 矩阵 A 与 B 的 Kronecker 积为 $A \otimes B$; 实矩阵 $A = (a_{ij})_{n \times m}$ 与 $B = (b_{ij})_{n \times m}$ 的内积为 $\langle A, B \rangle = \text{tr}(B^T A)$, Frobenius 范数 $\|A\| = \sqrt{\langle A, A \rangle}$; $\text{vec}(A)$ 表示将矩阵 A 按行拉直构成的列向量, 单位矩阵 $I_n = (e_1 \ e_2 \ \cdots \ e_n)$, 记 $S_n = (e_n \ e_{n-1} \ \cdots \ e_1)$.

定义 1 若矩阵 $X \in \mathbb{R}^{n \times n}$ 的元素满足 $x_{ij} = x_{ji} = x_{n+1-j, n+1-i} (i, j = 1, 2, \dots, n)$, 则称 X 为双对称矩阵.

耦合矩阵方程在线性系统理论中有重要的作用. 文献[1-2]用矩阵分解的方法讨论矩阵方程的解, 但是利用矩阵分解的方法求解多变量矩阵方程的双对称解很困难. 文献[3-4]利用迭代算法讨论了多变量矩阵方程的约束解, 但在实际应用中, 方程的系数往往来源于实验, 很难保证解的存在性, 因此本文借鉴文献[5]的思想用迭代算法讨论以下问题:

问题 1 设 $A_i \in \mathbb{R}^{p \times n_i}, B_i \in \mathbb{R}^{n_i \times q} (i = 1, 2, \dots, l), C \in \mathbb{R}^{p \times q}$, 求矩阵组 $[X_1, \dots, X_l] (X_i \in \mathbf{BSR}^{n_i \times n_i}, i = 1, 2, \dots, l)$, 使得

$$\|A_1 X_1 B_1 + A_2 X_2 B_2 + \dots + A_l X_l B_l - C\| = \min$$

问题 2 给定 $[\bar{X}_1, \bar{X}_2, \dots, \bar{X}_l] (X_i \in \mathbb{R}^{n_i \times n_i}, (i = 1, 2, \dots, l))$, 设 S_E 表示问题 1 的解集合, 求

$[\hat{X}_1 \ \hat{X}_2 \ \cdots \ \hat{X}_l] \in S_E, \hat{X}_i \in \mathbf{BSR}^{n_i \times n_i}$, 使得

$$\|\hat{X}_1 - \bar{X}_1\|^2 + \|\hat{X}_2 - \bar{X}_2\|^2 + \dots + \|\hat{X}_l - \bar{X}_l\|^2 =$$

$$\min_{[X_1, X_2, \dots, X_l] \in S_E} \{ \|X_1 - \bar{X}_1\|^2 + \|X_2 - \bar{X}_2\|^2 + \dots + \|X_l - \bar{X}_l\|^2 \}$$

定理 1 矩阵方程

$$A_1 X_1 B_1 + A_2 X_2 B_2 + \dots + A_l X_l B_l = C \tag{1}$$

有双对称解的充要条件是矩阵方程组

$$\begin{cases} A_1 X_1 B_1 + A_2 X_2 B_2 + \dots + A_l X_l B_l = C \\ B_1^T X_1 A_1^T + B_2^T X_2 A_2^T + \dots + B_l^T X_l A_l^T = C^T \\ A_1 S_{n_1} X_1 S_{n_1} B_1 + A_2 S_{n_2} X_2 S_{n_2} B_2 + \dots + A_l S_{n_l} X_l S_{n_l} B_l = C \\ B_1^T S_{n_1} X_1 S_{n_1} A_1^T + B_2^T S_{n_2} X_2 S_{n_2} A_2^T + \dots + B_l^T S_{n_l} X_l S_{n_l} A_l^T = C^T \end{cases} \tag{2}$$

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相容.

证 若矩阵方程(1)有双对称解 $\mathbf{H}_i (i=1,2,\dots,l)$, 则

$$\mathbf{H}_i = \mathbf{H}_i^T = \mathbf{S}_{n_i} \mathbf{H}_i \mathbf{S}_{n_i}$$

$$\mathbf{A}_1 \mathbf{H}_1 \mathbf{B}_1 + \mathbf{A}_2 \mathbf{H}_2 \mathbf{B}_2 + \dots + \mathbf{A}_l \mathbf{H}_l \mathbf{B}_l = \mathbf{C}$$

显然 \mathbf{H}_i 就是方程组(2)的解. 反过来, 若方程组(2)相容, 即存在 $\mathbf{G}_i \in \mathbb{R}^{n_i \times n_i} (i=1,2,\dots,l)$ 满足矩阵方程组(2), 令

$$\mathbf{W}_i = \frac{\mathbf{G}_i + \mathbf{G}_i^T + \mathbf{S}_{n_i} \mathbf{G}_i \mathbf{S}_{n_i} + \mathbf{S}_{n_i} \mathbf{G}_i^T \mathbf{S}_{n_i}}{4}$$

则 $\mathbf{W}_i \in \mathbf{BSR}^{n_i \times n_i} (i=1,2,\dots,l)$, 且

$$\mathbf{A}_1 \mathbf{W}_1 \mathbf{B}_1 + \mathbf{A}_2 \mathbf{W}_2 \mathbf{B}_2 + \dots + \mathbf{A}_l \mathbf{W}_l \mathbf{B}_l = \mathbf{C}$$

因此 $\mathbf{W}_i \in \mathbf{BSR}^{n_i \times n_i} (i=1,2,\dots,l)$ 是矩阵方程(1)的解, 即矩阵方程(1)有双对称解.

1 求解问题 1 的迭代算法

算法 1

1) 给定矩阵

$$\mathbf{A}_i \in \mathbb{R}^{p \times n_i}, \mathbf{B}_i \in \mathbb{R}^{n_i \times q}, \mathbf{C} \in \mathbb{R}^{p \times q}, \mathbf{X}_i^{(0)} \in \mathbf{BSR}^{n_i \times n_i} (i=1,2,\dots,l)$$

2) 计算

$$\mathbf{R}_1^{(0)} = \mathbf{C} - \sum_{i=1}^l \mathbf{A}_i \mathbf{X}_i^{(0)} \mathbf{B}_i, \mathbf{R}_2^{(0)} = (\mathbf{R}_1^{(0)})^T, \mathbf{R}_3^{(0)} = \mathbf{C} - \sum_{i=1}^l \mathbf{A}_i \mathbf{S}_{n_i} \mathbf{X}_i^{(0)} \mathbf{S}_{n_i} \mathbf{B}_i$$

$$\mathbf{R}_4^{(0)} = (\mathbf{R}_3^{(0)})^T, \mathbf{R}^{(0)} = \text{diag}(\mathbf{R}_1^{(0)}, \mathbf{R}_2^{(0)}, \mathbf{R}_3^{(0)}, \mathbf{R}_4^{(0)}),$$

$$\mathbf{P}_i^{(0)} = \mathbf{S}_i^{(0)} = \mathbf{A}_i^T \mathbf{R}_1^{(0)} \mathbf{B}_i^T + \mathbf{B}_i \mathbf{R}_2^{(0)} \mathbf{A}_i + \mathbf{S}_{n_i} \mathbf{A}_i^T \mathbf{R}_3^{(0)} \mathbf{B}_i^T \mathbf{S}_{n_i} + \mathbf{S}_{n_i} \mathbf{B}_i \mathbf{R}_4^{(0)} \mathbf{A}_i \mathbf{S}_{n_i},$$

$$\mathbf{S}^{(0)} = \text{diag}(\mathbf{S}_1^{(0)}, \mathbf{S}_2^{(0)}, \dots, \mathbf{S}_l^{(0)}), \gamma_0 = \sum_{i=1}^l \|\mathbf{S}_i^{(0)}\|^2,$$

$$\mathbf{Q}_1^{(0)} = \sum_{r=1}^l \mathbf{A}_r \mathbf{P}_r^{(0)} \mathbf{B}_r, \mathbf{Q}_2^{(0)} = [\mathbf{Q}_1^{(0)}]^T, \mathbf{Q}_3^{(0)} = \sum_{r=1}^l \mathbf{A}_r \mathbf{S}_{n_r} \mathbf{P}_r^{(0)} \mathbf{S}_{n_r} \mathbf{B}_r,$$

$$\mathbf{Q}_4^{(0)} = [\mathbf{Q}_3^{(0)}]^T, \mathbf{Q}^{(0)} = \text{diag}(\mathbf{Q}_1^{(0)}, \mathbf{Q}_2^{(0)}, \mathbf{Q}_3^{(0)}, \mathbf{Q}_4^{(0)})$$

$$\alpha_0 = \gamma_0 / (\|\mathbf{Q}_1^{(0)}\|^2 + \|\mathbf{Q}_2^{(0)}\|^2 + \|\mathbf{Q}_3^{(0)}\|^2 + \|\mathbf{Q}_4^{(0)}\|^2) k = 0$$

3) 如果 $\gamma_k = 0$, 停止. 否则 $k = k + 1$;

4) 计算

$$\mathbf{R}_i^{(k)} = \mathbf{R}_i^{(k-1)} - \alpha_{k-1} \mathbf{Q}_i^{(k-1)} (i=1,2,3,4), \mathbf{R}^{(k)} = \text{diag}(\mathbf{R}_1^{(k)}, \mathbf{R}_2^{(k)}, \mathbf{R}_3^{(k)}, \mathbf{R}_4^{(k)}),$$

$$\mathbf{S}_i^{(k)} = \mathbf{A}_i^T \mathbf{R}_1^{(k)} \mathbf{B}_i^T + \mathbf{B}_i \mathbf{R}_2^{(k)} \mathbf{A}_i + \mathbf{S}_{n_i} \mathbf{A}_i^T \mathbf{R}_3^{(k)} \mathbf{B}_i^T \mathbf{S}_{n_i} + \mathbf{S}_{n_i} \mathbf{B}_i \mathbf{R}_4^{(k)} \mathbf{A}_i \mathbf{S}_{n_i}$$

$$\mathbf{S}^{(k)} = \text{diag}(\mathbf{S}_1^{(k)}, \mathbf{S}_2^{(k)}, \dots, \mathbf{S}_l^{(k)}), \gamma_k = \sum_{i=1}^l \|\mathbf{S}_i^{(k)}\|^2, \beta_k = \frac{\gamma_k}{\gamma_{k-1}},$$

$$\mathbf{P}_i^{(k)} = \mathbf{S}_i^{(k)} + \beta_{k-1} \mathbf{P}_i^{(k-1)}, \mathbf{X}_i^{(k)} = \mathbf{X}_i^{(k-1)} + \alpha_{k-1} \mathbf{P}_i^{(k-1)},$$

$$\mathbf{Q}_1^{(k)} = \sum_{r=1}^l \mathbf{A}_r \mathbf{P}_r^{(k)} \mathbf{B}_r, \mathbf{Q}_2^{(k)} = [\mathbf{Q}_1^{(k)}]^T, \mathbf{Q}_3^{(k)} = \sum_{r=1}^l \mathbf{A}_r \mathbf{S}_{n_r} \mathbf{P}_r^{(k)} \mathbf{S}_{n_r} \mathbf{B}_r,$$

$$\mathbf{Q}_4^{(k)} = [\mathbf{Q}_3^{(k)}]^T, \mathbf{Q}^{(k)} = \text{diag}(\mathbf{Q}_1^{(k)}, \mathbf{Q}_2^{(k)}, \mathbf{Q}_3^{(k)}, \mathbf{Q}_4^{(k)}),$$

$$\alpha_k = \frac{\gamma_k}{\|\mathbf{Q}_1^{(k)}\|^2 + \|\mathbf{Q}_2^{(k)}\|^2 + \|\mathbf{Q}_3^{(k)}\|^2 + \|\mathbf{Q}_4^{(k)}\|^2}$$

5) 返回 3).

引理 1 对算法 1 中的矩阵 $\mathbf{S}^{(i)}, \mathbf{P}_r^{(i)} (r=1,2,\dots,l)$, 若存在正数 k , 使得 $\mathbf{S}^{(i)} \neq \mathbf{O}, (i=0,1,2,\dots,k)$,

那么

$$\langle \mathbf{S}^{(i)}, \mathbf{S}^{(j)} \rangle = 0, \langle \mathbf{Q}^{(i)}, \mathbf{Q}^{(j)} \rangle = 0, \sum_{r=1}^l \langle \mathbf{P}_r^{(i)}, \mathbf{S}_r^{(j)} \rangle = 0 (i \neq j; i, j = 0, 1, 2, \dots, k)$$

证 先证明

$$\begin{aligned} \langle \mathbf{S}^{(i)}, \mathbf{S}^{(i+1)} \rangle &= 0 & \langle \mathbf{Q}^{(i)}, \mathbf{Q}^{(i+1)} \rangle &= 0 \\ \sum_{r=1}^l \langle \mathbf{P}_r^{(i)}, \mathbf{S}_r^{(i+1)} \rangle &= 0 \quad (i=0, 1, 2, \dots, k) \end{aligned}$$

用数学归纳法. 当 $i=0$ 时

$$\begin{aligned} \langle \mathbf{S}^{(0)}, \mathbf{S}^{(1)} \rangle &= \|\mathbf{S}^{(0)}\|^2 - \alpha_0 \left(\left\langle \sum_{r=1}^l \mathbf{A}_r \mathbf{P}_r^{(0)} \mathbf{B}_r, \mathbf{Q}_1^{(0)} \right\rangle + \left\langle \sum_{r=1}^l \mathbf{A}_r \mathbf{S}_{nr} \mathbf{P}_r^{(0)} \mathbf{S}_{nr} \mathbf{B}_r, \mathbf{Q}_3^{(0)} \right\rangle \right) - \\ &\quad \alpha_0 \left(\left\langle \sum_{r=1}^l \mathbf{B}_r^T \mathbf{P}_r^{(0)} \mathbf{A}_r^T, \mathbf{Q}_2^{(0)} \right\rangle + \left\langle \sum_{r=1}^l \mathbf{B}_r^T \mathbf{S}_{nr} \mathbf{P}_r^{(0)} \mathbf{S}_{nr} \mathbf{A}_r^T, \mathbf{Q}_4^{(0)} \right\rangle \right) = \\ &\quad \|\mathbf{S}^{(0)}\|^2 - \frac{\alpha_0}{\|\mathbf{Q}_1^{(0)}\|^2 + \|\mathbf{Q}_2^{(0)}\|^2 + \|\mathbf{Q}_3^{(0)}\|^2 + \|\mathbf{Q}_4^{(0)}\|^2} = 0 \\ \sum_{r=1}^l \langle \mathbf{P}_r^{(0)}, \mathbf{S}_r^{(1)} \rangle &= \|\mathbf{S}^{(0)}\|^2 - \alpha_0 \left(\left\langle \sum_{i=1}^l \mathbf{A}_i \mathbf{P}_i^{(0)} \mathbf{B}_i, \mathbf{Q}_1^{(0)} \right\rangle + \left\langle \sum_{i=1}^l \mathbf{B}_i^T \mathbf{P}_i^{(0)} \mathbf{A}_i^T, \mathbf{Q}_2^{(0)} \right\rangle \right) - \\ &\quad \alpha_0 \left(\left\langle \sum_{i=1}^l \mathbf{A}_i \mathbf{S}_{nr} \mathbf{P}_i^{(0)} \mathbf{S}_{nr} \mathbf{B}_i, \mathbf{Q}_1^{(0)} \right\rangle + \left\langle \sum_{i=1}^l \mathbf{B}_i^T \mathbf{S}_{nr} \mathbf{P}_i^{(0)} \mathbf{S}_{nr} \mathbf{A}_i^T, \mathbf{Q}_2^{(0)} \right\rangle \right) = 0 \\ \langle \mathbf{Q}^{(0)}, \mathbf{Q}^{(1)} \rangle &= \langle \mathbf{Q}_1^{(0)}, \mathbf{Q}_1^{(1)} \rangle + \langle \mathbf{Q}_2^{(0)}, \mathbf{Q}_2^{(1)} \rangle + \langle \mathbf{Q}_3^{(0)}, \mathbf{Q}_3^{(1)} \rangle + \langle \mathbf{Q}_4^{(0)}, \mathbf{Q}_4^{(1)} \rangle = \\ &\quad \sum_{r=1}^l \langle \mathbf{A}_r^T \mathbf{Q}_1^{(0)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_2^{(0)} \mathbf{A}_r + \mathbf{S}_{nr} \mathbf{A}_r^T \mathbf{Q}_3^{(0)} \mathbf{B}_r^T \mathbf{S}_{nr} + \mathbf{S}_{nr} \mathbf{B}_r \mathbf{Q}_4^{(0)} \mathbf{A}_r \mathbf{S}_{nr}, \mathbf{P}_r^{(1)} \rangle = \\ &\quad \sum_{r=1}^l \frac{1}{\alpha_0} \langle (\mathbf{S}_r^{(0)} - \mathbf{S}_r^{(1)}), (\mathbf{S}_r^{(1)} + \beta_0 \mathbf{S}_r^{(0)}) \rangle = -\frac{1}{\alpha_0} \|\mathbf{S}^{(1)}\|^2 + \frac{\beta_0}{\alpha_0} \|\mathbf{S}^{(1)}\|^2 = 0 \end{aligned}$$

假定对 $i \leq t (0 < t < k)$ 时, 结论成立, 则

$$\begin{aligned} \langle \mathbf{S}^{(t)}, \mathbf{S}^{(t+1)} \rangle &= \|\mathbf{S}^{(t)}\|^2 - \\ &\quad \alpha_t \sum_{r=1}^l \langle \mathbf{P}_r^{(t)} - \beta_{t-1} \mathbf{P}_r^{(t-1)}, \mathbf{A}_r^T \mathbf{Q}_1^{(t)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_2^{(t)} \mathbf{A}_r + \mathbf{S}_{nr} \mathbf{A}_r^T \mathbf{Q}_3^{(t)} \mathbf{B}_r^T \mathbf{S}_{nr} + \mathbf{S}_{nr} \mathbf{B}_r \mathbf{Q}_4^{(t)} \mathbf{A}_r \mathbf{S}_{nr} \rangle = \\ &\quad \|\mathbf{S}^{(t)}\|^2 - \alpha_t \left(\langle \mathbf{Q}^{(t)}, \mathbf{Q}^{(t)} \rangle - \beta_{k-1} \langle \mathbf{Q}^{(t)}, \mathbf{Q}^{(t-1)} \rangle \right) = \|\mathbf{S}^{(t)}\|^2 - \alpha_t \|\mathbf{Q}^{(t)}\|^2 = 0 \\ \sum_{r=1}^l \langle \mathbf{P}_r^{(t)}, \mathbf{S}_r^{(t+1)} \rangle &= \sum_{r=1}^l \langle \mathbf{S}_r^{(t)} + \beta_{k-1} \mathbf{P}_r^{(t)}, \mathbf{S}_r^{(t+1)} \rangle = \\ &\quad -\beta_{t-1} \alpha_t \left(\left\langle \sum_{r=1}^l \mathbf{A}_r \mathbf{P}_r^{(t-1)} \mathbf{B}_r, \mathbf{Q}_1^{(t)} \right\rangle + \left\langle \sum_{r=1}^l \mathbf{A}_r \mathbf{S}_{nr} \mathbf{P}_r^{(t-1)} \mathbf{S}_{nr} \mathbf{B}_r, \mathbf{Q}_3^{(t)} \right\rangle \right) - \\ &\quad \beta_{t-1} \alpha_t \left(\left\langle \sum_{r=1}^l \mathbf{B}_r^T \mathbf{P}_r^{(t-1)} \mathbf{A}_r^T, \mathbf{Q}_2^{(t)} \right\rangle + \left\langle \sum_{r=1}^l \mathbf{B}_r^T \mathbf{S}_{nr} \mathbf{P}_r^{(t-1)} \mathbf{S}_{nr} \mathbf{A}_r^T, \mathbf{Q}_4^{(t)} \right\rangle \right) = \\ &\quad -\beta_{t-1} \alpha_t \langle \mathbf{Q}^{(t-1)}, \mathbf{Q}^{(t)} \rangle = 0 \\ \langle \mathbf{Q}^{(t)}, \mathbf{Q}^{(t+1)} \rangle &= \langle \mathbf{Q}_1^{(t)}, \mathbf{Q}_1^{(t+1)} \rangle + \langle \mathbf{Q}_2^{(t)}, \mathbf{Q}_2^{(t+1)} \rangle + \langle \mathbf{Q}_3^{(t)}, \mathbf{Q}_3^{(t+1)} \rangle + \langle \mathbf{Q}_4^{(t)}, \mathbf{Q}_4^{(t+1)} \rangle = \\ &\quad \sum_{r=1}^l \langle \mathbf{A}_r^T \mathbf{Q}_1^{(t)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_2^{(t)} \mathbf{A}_r + \mathbf{S}_{nr} \mathbf{A}_r^T \mathbf{Q}_3^{(t)} \mathbf{B}_r^T \mathbf{S}_{nr} + \mathbf{S}_{nr} \mathbf{B}_r \mathbf{Q}_4^{(t)} \mathbf{A}_r \mathbf{S}_{nr}, \mathbf{P}_r^{(t+1)} \rangle = \\ &\quad \sum_{r=1}^l \frac{1}{\alpha_t} \langle (\mathbf{S}_r^{(t)} - \mathbf{S}_r^{(t+1)}), (\mathbf{S}_r^{(t+1)} + \beta_t \mathbf{P}_r^{(t)}) \rangle = \\ &\quad -\frac{1}{\alpha_t} \|\mathbf{S}^{(t+1)}\|^2 + \frac{\beta_t}{\alpha_t} \sum_{r=1}^l \langle \mathbf{S}_r^{(t)}, \mathbf{S}_r^{(t)} + \beta_{k-1} \mathbf{P}_r^{(t-1)} \rangle = \\ &\quad -\frac{1}{\alpha_t} \|\mathbf{S}^{(t+1)}\|^2 + \frac{\beta_t}{\alpha_t} \|\mathbf{S}^{(t)}\|^2 = 0 \end{aligned}$$

根据归纳法原理, 对 $i=0, 1, 2, \dots, k$, 都有

$$\langle \mathbf{S}^{(i)}, \mathbf{S}^{(i+1)} \rangle = 0, \quad \langle \mathbf{Q}^{(i)}, \mathbf{Q}^{(i+1)} \rangle = 0, \quad \sum_{r=1}^l \langle \mathbf{P}_r^{(i)}, \mathbf{S}_r^{(i+1)} \rangle = 0$$

第 2 步: 假定当 $0 \leq i \leq k, 1 < l < k$ 时,

$$\langle \mathbf{S}^{(i)}, \mathbf{S}^{(i+l)} \rangle = 0, \langle \mathbf{Q}^{(i)}, \mathbf{Q}^{(i+l)} \rangle = 0, \sum_{r=1}^l \langle \mathbf{P}_r^{(i)}, \mathbf{S}_r^{(i+l)} \rangle = 0 (i = 0, 1, 2, \dots, k)$$

则

$$\begin{aligned} & \langle \mathbf{S}^{(i)}, \mathbf{S}^{(i+l+1)} \rangle = \\ & -\alpha_{i+l} \sum_{r=1}^l \langle \mathbf{P}_r^{(i)} - \beta_{i-1} \mathbf{P}_r^{(i-1)}, \mathbf{A}_r^T \mathbf{Q}_2^{(i+l)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_1^{(i+l)} \mathbf{A}_r + \mathbf{S}_{nr} \mathbf{A}_r^T \mathbf{Q}_3^{(i+l)} \mathbf{B}_r^T \mathbf{S}_{nr} + \mathbf{S}_{nr} \mathbf{B}_r \mathbf{Q}_4^{(i+l)} \mathbf{A}_r \mathbf{S}_{nr} \rangle = \\ & -\alpha_{i+l} (\langle \mathbf{Q}^{(i)}, \mathbf{Q}^{(i+l)} \rangle - \beta_{i-1} \langle \mathbf{Q}^{(i-1)}, \mathbf{Q}^{(i+l)} \rangle) = 0 \\ & \sum_{r=1}^l \langle \mathbf{P}_r^{(i)}, \mathbf{S}_r^{(i+l+1)} \rangle = \\ & \sum_{r=1}^l \langle \mathbf{S}_r^{(i)} + \beta_{i-1} \mathbf{P}_r^{(i-1)}, \mathbf{S}_r^{(i+l+1)} \rangle = \\ & \sum_{r=1}^l \langle \beta_{i-1} \mathbf{P}_r^{(i-1)}, -\alpha_{i+l} (\mathbf{A}_r^T \mathbf{Q}_1^{(i+l)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_2^{(i+l)} \mathbf{A}_r + \mathbf{S}_{nr} \mathbf{A}_r^T \mathbf{Q}_1^{(i+l)} \mathbf{B}_r^T \mathbf{S}_{nr} + \mathbf{S}_{nr} \mathbf{B}_r \mathbf{Q}_2^{(i+l)} \mathbf{A}_r \mathbf{S}_{nr}) \rangle = \\ & -\alpha_{i+l} \beta_{i-1} (\langle \sum_{i=1}^l \mathbf{A}_r \mathbf{P}_r^{(i-1)} \mathbf{B}_r, \mathbf{Q}_1^{(i+l)} \rangle + \langle \sum_{i=1}^l \mathbf{B}_r^T \mathbf{P}_r^{(i-1)} \mathbf{A}_r^T, \mathbf{Q}_2^{(i+l)} \rangle) + \\ & -\alpha_{i+l} \beta_{i-1} (\langle \sum_{i=1}^l \mathbf{A}_r \mathbf{S}_{nr} \mathbf{P}_r^{(i-1)} \mathbf{S}_{nr} \mathbf{B}_r, \mathbf{Q}_3^{(i+l)} \rangle + \langle \sum_{i=1}^l \mathbf{B}_r^T \mathbf{S}_{nr} \mathbf{P}_r^{(i-1)} \mathbf{S}_{nr} \mathbf{A}_r^T, \mathbf{Q}_4^{(i+l)} \rangle) = 0 \\ & \langle \mathbf{Q}^{(i)}, \mathbf{Q}^{(i+l+1)} \rangle = \langle \mathbf{Q}_1^{(i)}, \sum_{r=1}^l \mathbf{A}_r \mathbf{P}_r^{(i+l+1)} \mathbf{B}_r \rangle + \langle \mathbf{Q}_3^{(i)}, \sum_{r=1}^l \mathbf{A}_r \mathbf{S}_{nr} \mathbf{P}_r^{(i+l+1)} \mathbf{S}_{nr} \mathbf{B}_r \rangle + \\ & \langle \mathbf{Q}_2^{(i)}, \sum_{r=1}^l \mathbf{B}_r^T \mathbf{P}_r^{(i+l+1)} \mathbf{A}_r^T \rangle + \langle \mathbf{Q}_2^{(i)}, \sum_{r=1}^l \mathbf{B}_r^T \mathbf{S}_{nr} \mathbf{P}_r^{(i+l+1)} \mathbf{S}_{nr} \mathbf{A}_r^T \rangle = \\ & \sum_{r=1}^l \langle \mathbf{A}_r^T \mathbf{Q}_1^{(i)} \mathbf{B}_r^T, \mathbf{P}_r^{(i+l+1)} \rangle + \sum_{r=1}^l \langle \mathbf{S}_{nr} \mathbf{A}_r^T \mathbf{Q}_1^{(i)} \mathbf{B}_r^T \mathbf{S}_{nr}, \mathbf{P}_r^{(i+l+1)} \rangle + \\ & \sum_{r=1}^l \langle \mathbf{S}_{nr} \mathbf{B}_r \mathbf{Q}_2^{(i)} \mathbf{A}_r \mathbf{S}_{nr}, \mathbf{P}_r^{(i+l+1)} \rangle + \sum_{r=1}^l \langle \mathbf{B}_r \mathbf{Q}_2^{(i)} \mathbf{A}_r, \mathbf{P}_r^{(i+l+1)} \rangle = \\ & \sum_{r=1}^l \frac{1}{\alpha_i} \langle (\mathbf{S}_r^{(i)} - \mathbf{S}_r^{(i+1)}), (\mathbf{S}_r^{(i+l+1)} + \beta_k \mathbf{P}_r^{(i+l)}) \rangle = 0 \end{aligned}$$

由归纳法得

$$\langle \mathbf{S}^{(i)}, \mathbf{S}^{(j)} \rangle = 0, \langle \mathbf{Q}^{(i)}, \mathbf{Q}^{(j)} \rangle = 0, \sum_{r=1}^l \langle \mathbf{P}_r^{(i)}, \mathbf{S}_r^{(j)} \rangle = 0 (i \neq j; i, j = 0, 1, 2, \dots, k)$$

引理 2 对任意初始对称矩阵组 $(\mathbf{X}_1^{(0)}, \mathbf{X}_2^{(0)}, \dots, \mathbf{X}_l^{(0)})$, 算法 1 中的矩阵 $\mathbf{S}_r^{(i)}, \mathbf{P}_r^{(i)}$ 满足

$$\sum_{r=1}^l \langle \mathbf{P}_r^{(i)}, \mathbf{S}_r^{(i)} \rangle = \|\mathbf{S}^{(i)}\|^2 \quad i = 0, 1, 2, \dots$$

证 采用数学归纳法, 当 $i = 0$ 时

$$\sum_{r=1}^l \langle \mathbf{P}_r^{(0)}, \mathbf{S}_r^{(0)} \rangle = \sum_{r=1}^l \langle \mathbf{S}_r^{(0)}, \mathbf{S}_r^{(0)} \rangle = \|\mathbf{S}^{(0)}\|^2$$

假定当 $i = k$ 时结论成立, 则当 $i = k + 1$ 时,

$$\begin{aligned} & \sum_{r=1}^l \langle \mathbf{P}_r^{(k+1)}, \mathbf{S}_r^{(k+1)} \rangle = \\ & \sum_{r=1}^l \langle \mathbf{S}_r^{(k+1)} + \beta_k \mathbf{P}_r^{(k)}, \mathbf{S}_r^{(k+1)} \rangle = \sum_{r=1}^l \langle \mathbf{S}_r^{(k+1)}, \mathbf{S}_r^{(k+1)} \rangle + \\ & \beta_k \sum_{r=1}^l \langle \mathbf{P}_r^{(k)}, \mathbf{S}_r^{(k)} - \alpha_k (\mathbf{A}_r^T \mathbf{Q}_1^{(k)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_2^{(k)} \mathbf{A}_r + \mathbf{S}_{nr} \mathbf{B}_r \mathbf{Q}_2^{(k)} \mathbf{A}_r \mathbf{S}_{nr} + \mathbf{S}_{nr} \mathbf{A}_r^T \mathbf{Q}_1^{(k)} \mathbf{B}_r^T \mathbf{S}_{nr}) \rangle = \end{aligned}$$

$$\begin{aligned} & \| \mathbf{S}^{(k+1)} \|^2 + \beta_k \| \mathbf{S}^{(k)} \|^2 - \beta_k \alpha_k \left[\sum_{r=1}^l \langle \mathbf{A}_r \mathbf{P}_r^{(k)} \mathbf{B}_r, \mathbf{Q}_1^{(k)} \rangle + \sum_{r=1}^l \langle \mathbf{A}_r \mathbf{S}_{nr} \mathbf{P}_r^{(k)} \mathbf{S}_{nr} \mathbf{B}_r, \mathbf{Q}_3^{(k)} \rangle \right] - \\ & \beta_k \alpha_k \left[\sum_{r=1}^l \langle \mathbf{B}_r^T \mathbf{P}_r^{(k)} \mathbf{A}_r^T, \mathbf{Q}_2^{(k)} \rangle + \sum_{r=1}^l \langle \mathbf{B}_r^T \mathbf{S}_{nr} \mathbf{P}_r^{(k)} \mathbf{S}_{nr} \mathbf{A}_r^T, \mathbf{Q}_4^{(k)} \rangle \right] = \\ & \| \mathbf{S}^{(k+1)} \|^2 + \beta_k \| \mathbf{S}^{(k)} \|^2 - \beta_k \gamma_k = \| \mathbf{S}^{(k+1)} \|^2 \end{aligned}$$

由归纳法原理知

$$\sum_{r=1}^l \langle \mathbf{P}_r^{(i)}, \mathbf{S}_r^{(i)} \rangle = \| \mathbf{S}^{(i)} \|^2 \quad i = 0, 1, 2, \dots$$

引理 3 若 $(\mathbf{X}_1^*, \mathbf{X}_2^*, \dots, \mathbf{X}_l^*)$ 是耦合矩阵方程的双对称最小二乘解, 则对初始矩阵组 $(\mathbf{X}_1^0, \mathbf{X}_2^0, \dots, \mathbf{X}_l^0)$ ($\mathbf{X}_i \in \mathbf{BSR}^{n \times n}$), 算法中的矩阵 $\mathbf{S}_r^{(i)}, \mathbf{P}_r^{(i)}, \mathbf{X}_r^{(i)}$ 满足

$$\sum_{r=1}^l \langle \mathbf{X}_r^* - \mathbf{X}_r^{(i)}, \mathbf{A}_r^T \mathbf{Q}_1^{(i)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_2^{(i)} \mathbf{A}_r + \mathbf{S}_{nr} \mathbf{A}_r^T \mathbf{Q}_3^{(i)} \mathbf{B}_r^T \mathbf{S}_{nr} + \mathbf{S}_{nr} \mathbf{B}_r \mathbf{Q}_4^{(i)} \mathbf{A}_r \mathbf{S}_{nr} \rangle = \| \mathbf{S}^{(i)} \|^2$$

证

$$\begin{aligned} & \sum_{r=1}^l \langle \mathbf{X}_r^* - \mathbf{X}_r^{(i)}, \mathbf{A}_r^T \mathbf{Q}_1^{(i)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_2^{(i)} \mathbf{A}_r + \mathbf{S}_{nr} (\mathbf{A}_r^T \mathbf{Q}_3^{(i)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_4^{(i)} \mathbf{A}_r) \mathbf{S}_{nr} \rangle = \\ & \langle \mathbf{R}_1^{(i)}, \sum_{r=1}^l \mathbf{A}_r \mathbf{P}_r^{(k)} \mathbf{B}_r \rangle + \langle \mathbf{R}_2^{(i)}, \sum_{r=1}^l \mathbf{B}_r^T \mathbf{P}_r^{(k)} \mathbf{A}_r^T \rangle + \\ & \langle \mathbf{R}_3^{(i)}, \sum_{r=1}^l \mathbf{A}_r \mathbf{S}_{nr} \mathbf{P}_r^{(k)} \mathbf{S}_{nr} \mathbf{B}_r \rangle + \langle \mathbf{R}_4^{(i)}, \sum_{r=1}^l \mathbf{B}_r^T \mathbf{S}_{nr} \mathbf{P}_r^{(k)} \mathbf{S}_{nr} \mathbf{A}_r^T \rangle = \\ & \sum_{r=1}^l \langle \mathbf{S}_r^{(i)}, \mathbf{P}_r^{(i)} \rangle \end{aligned}$$

由引理 2 可得

$$\sum_{r=1}^l \langle \mathbf{X}_r^* - \mathbf{X}_r^{(i)}, \mathbf{A}_r^T \mathbf{Q}_1^{(i)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_2^{(i)} \mathbf{A}_r + \mathbf{S}_{nr} \mathbf{A}_r^T \mathbf{Q}_3^{(i)} \mathbf{B}_r^T \mathbf{S}_{nr} + \mathbf{S}_{nr} \mathbf{B}_r \mathbf{Q}_4^{(i)} \mathbf{A}_r \mathbf{S}_{nr} \rangle = \| \mathbf{S}^{(i)} \|^2$$

注 1 由引理 3 若 $\mathbf{S}^{(i)} \neq \mathbf{O}$,

$$\sum_{r=1}^l \| \mathbf{A}_r^T \mathbf{Q}_1^{(i)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_2^{(i)} \mathbf{A}_r + \mathbf{S}_{nr} (\mathbf{A}_r^T \mathbf{Q}_3^{(i)} \mathbf{B}_r^T + \mathbf{B}_r \mathbf{Q}_4^{(i)} \mathbf{A}_r) \mathbf{S}_{nr} \|^2 \neq 0$$

从而

$$\mathbf{Q}^{(i)} \neq \mathbf{O}$$

即

$$\sum_{r=1}^l \| \mathbf{P}_r^{(i)} \|^2 \neq 0$$

定理 2 对任意初始矩阵组 $(\mathbf{X}_1^{(0)}, \mathbf{X}_2^{(0)}, \dots, \mathbf{X}_l^{(0)})$ ($\mathbf{X}_i \in \mathbf{BSR}^{n_i \times n_i}$, $i = 1, 2, \dots, l$), 通过算法 1 可在有限步迭代后得到问题 1 的解.

证 若

$$\mathbf{S}^{(i)} \neq \mathbf{O} \quad (i = 0, 1, \dots, t; t = \sum_{k=1}^l n_k^2)$$

由注 1 知

$$\sum_{r=1}^l \| \mathbf{P}_r^{(i)} \|^2 \neq 0$$

从而可得 $(\mathbf{X}_1^{(t)}, \mathbf{X}_2^{(t)}, \dots, \mathbf{X}_l^{(t)})$ 和 $\mathbf{S}^{(t)}$. 由引理 1

$$\langle \mathbf{S}^{(i)}, \mathbf{S}^{(t)} \rangle = 0 \quad i = 1, 2, \dots, t-1$$

而

$$\langle \mathbf{S}^{(i)}, \mathbf{S}^{(j)} \rangle = 0 \quad i, j = 0, 1, \dots, t-1, i \neq j$$

故 $\mathbf{S}^{(0)}, \mathbf{S}^{(1)}, \dots, \mathbf{S}^{(t-1)}$ 是空间 $L = \{K \mid K = \text{diag}(K_1, K_2, \dots, K_l)\}$ 的一组正交基, 其中 $K_i \in \mathbb{R}^{n_i \times n_i}$, 从而 $\mathbf{S}^{(t)} = O$, 即 $(\mathbf{X}_1^t, \mathbf{X}_2^t, \dots, \mathbf{X}_l^t)$ 为矩阵方程(1)的双对称最小二乘解组. 因此问题 1 的解组最多在 t 步计算后得到.

2 极小范数解

引理 4^[5] 设线性方程 $\mathbf{Ax} = \mathbf{b}$ 相容, 且有一个解 $\mathbf{x}^* \in \mathbb{R}(\mathbf{A}^T)$, 则该方程惟一的极小范数解为 \mathbf{x}^* .

定理 3 若初始矩阵组为

$$\mathbf{X}_i^{(0)} = \mathbf{A}_i^T \mathbf{W}_i \mathbf{B}_i^T + \mathbf{B}_i \mathbf{W}_i^T \mathbf{A}_i + \mathbf{S}_{n_i} (\mathbf{A}_i^T \mathbf{W}_i \mathbf{B}_i^T + \mathbf{B}_i \mathbf{W}_i^T \mathbf{A}_i) \mathbf{S}_{n_i} \quad (i=1, 2, \dots, l) \quad (\mathbf{W}_i \in \mathbb{R}^{n_i \times n_i}) \quad (3)$$

则问题 1 的极小范数解由算法 2 在有限步迭代后得到

证 选取初始矩阵组为(3)式, 则由算法 1 得问题 1 的解组 \mathbf{X}_i^* ($i=1, 2, \dots, l$), 且 \mathbf{X}_i^* 可表示为:

$$\mathbf{X}_i^* = \mathbf{A}_i^T \mathbf{Z}_i \mathbf{B}_i^T + \mathbf{B}_i \mathbf{Z}_i^T \mathbf{A}_i + \mathbf{S}_{n_i} (\mathbf{A}_i^T \mathbf{Z}_i \mathbf{B}_i^T + \mathbf{B}_i \mathbf{Z}_i^T \mathbf{A}_i) \mathbf{S}_{n_i} \quad (i=1, 2, \dots, l) \quad (\mathbf{Z}_i \in \mathbb{R}^{n_i \times n_i})$$

因为矩阵方程组(2)等价于线性方程组

$$\begin{pmatrix} \mathbf{B}_1^T \otimes \mathbf{A}_1 & \mathbf{B}_2^T \otimes \mathbf{A}_2 & \cdots & \mathbf{B}_l^T \otimes \mathbf{A}_l \\ \mathbf{A}_1 \otimes \mathbf{B}_1^T & \mathbf{A}_2 \otimes \mathbf{B}_2^T & \cdots & \mathbf{A}_l \otimes \mathbf{B}_l^T \\ \mathbf{B}_1^T \mathbf{S}_{n_1} \otimes \mathbf{A}_1 \mathbf{S}_{n_1} & \mathbf{B}_2^T \mathbf{S}_{n_2} \otimes \mathbf{A}_2 \mathbf{S}_{n_2} & \cdots & \mathbf{B}_l^T \mathbf{S}_{n_l} \otimes \mathbf{A}_l \mathbf{S}_{n_l} \\ \mathbf{A}_1 \mathbf{S}_{n_1} \otimes \mathbf{B}_1^T \mathbf{S}_{n_1} & \mathbf{A}_2 \mathbf{S}_{n_2} \otimes \mathbf{B}_2^T \mathbf{S}_{n_2} & \cdots & \mathbf{A}_l \mathbf{S}_{n_l} \otimes \mathbf{B}_l^T \mathbf{S}_{n_l} \end{pmatrix} \begin{pmatrix} \text{vec}(\mathbf{X}_1) \\ \text{vec}(\mathbf{X}_2) \\ \vdots \\ \text{vec}(\mathbf{X}_l) \end{pmatrix} = \begin{pmatrix} \text{vec}(\mathbf{C}) \\ \text{vec}(\mathbf{C}^T) \\ \text{vec}(\mathbf{C}) \\ \text{vec}(\mathbf{C}^T) \end{pmatrix} \quad (4)$$

又因为

$$\begin{pmatrix} \text{vec}(\mathbf{X}_1^*) \\ \text{vec}(\mathbf{X}_2^*) \\ \vdots \\ \text{vec}(\mathbf{X}_l^*) \end{pmatrix} = \mathbf{M}^T \begin{pmatrix} \text{vec}(\mathbf{Z}) \\ \text{vec}(\mathbf{Z}^T) \\ \text{vec}(\mathbf{Z}) \\ \text{vec}(\mathbf{Z}^T) \end{pmatrix} \in \mathbb{R}(\mathbf{M}^T)$$

其中

$$\mathbf{M} = \begin{pmatrix} \mathbf{B}_1^T \otimes \mathbf{A}_1 & \mathbf{B}_2^T \otimes \mathbf{A}_2 & \cdots & \mathbf{B}_l^T \otimes \mathbf{A}_l \\ \mathbf{A}_1 \otimes \mathbf{B}_1^T & \mathbf{A}_2 \otimes \mathbf{B}_2^T & \cdots & \mathbf{A}_l \otimes \mathbf{B}_l^T \\ \mathbf{B}_1^T \mathbf{S}_{n_1} \otimes \mathbf{A}_1 \mathbf{S}_{n_1} & \mathbf{B}_2^T \mathbf{S}_{n_2} \otimes \mathbf{A}_2 \mathbf{S}_{n_2} & \cdots & \mathbf{B}_l^T \mathbf{S}_{n_l} \otimes \mathbf{A}_l \mathbf{S}_{n_l} \\ \mathbf{A}_1 \mathbf{S}_{n_1} \otimes \mathbf{B}_1^T \mathbf{S}_{n_1} & \mathbf{A}_2 \mathbf{S}_{n_2} \otimes \mathbf{B}_2^T \mathbf{S}_{n_2} & \cdots & \mathbf{A}_l \mathbf{S}_{n_l} \otimes \mathbf{B}_l^T \mathbf{S}_{n_l} \end{pmatrix}$$

由引理 4 知 $(\text{vec}(\mathbf{X}_1^*), \text{vec}(\mathbf{X}_2^*), \dots, \text{vec}(\mathbf{X}_l^*))$ 是矩阵方程(4)的极小范数解, 因此, 可得到问题 1 的极

小范数解 $[\mathbf{X}_1^*, \mathbf{X}_2^*, \dots, \mathbf{X}_l^*]$.

3 问题 2 的解

设 $[\overline{\mathbf{X}}_1, \overline{\mathbf{X}}_2, \dots, \overline{\mathbf{X}}_l]$ ($\overline{\mathbf{X}}_i \in \mathbb{R}^{n_i \times n_i}$ ($i=1, 2, \dots, l$)) 为任意给定矩阵组, \mathbf{S}_E 非空, 若 $\mathbf{X}_i \in \mathbf{S}_E$ ($i=1, 2, \dots, l$), 则问题 2 等价于

$$\|\mathbf{X}_1 - \overline{\mathbf{X}}_1\|^2 + \|\mathbf{X}_2 - \overline{\mathbf{X}}_2\|^2 + \cdots + \|\mathbf{X}_l - \overline{\mathbf{X}}_l\|^2 = \min$$

其中

$$\overline{\mathbf{X}}_i = \frac{\overline{\mathbf{X}}_i + \overline{\mathbf{X}}_i^T + \mathbf{S}_{n_i} (\overline{\mathbf{X}}_i + \overline{\mathbf{X}}_i^T) \mathbf{S}_{n_i}}{4} \quad i=1, 2, \dots, l$$

改写矩阵方程组(1)为

$$\begin{cases} \mathbf{A}_1 \tilde{\mathbf{X}}_1 \mathbf{B}_1 + \mathbf{A}_2 \tilde{\mathbf{X}}_2 \mathbf{B}_2 + \cdots + \mathbf{A}_l \tilde{\mathbf{X}}_l \mathbf{B}_l = \tilde{\mathbf{C}} \\ \mathbf{B}_1^T \tilde{\mathbf{X}}_1 \mathbf{A}_1^T + \mathbf{B}_2^T \tilde{\mathbf{X}}_2 \mathbf{A}_2^T + \cdots + \mathbf{B}_l^T \tilde{\mathbf{X}}_l \mathbf{A}_l^T = \tilde{\mathbf{C}}^T \\ \mathbf{A}_1 \mathbf{S}_{n_1} \tilde{\mathbf{X}}_1 \mathbf{S}_{n_1} \mathbf{B}_1 + \mathbf{A}_2 \mathbf{S}_{n_2} \tilde{\mathbf{X}}_2 \mathbf{S}_{n_2} \mathbf{B}_2 + \cdots + \mathbf{A}_l \mathbf{S}_{n_l} \tilde{\mathbf{X}}_l \mathbf{S}_{n_l} \mathbf{B}_l = \tilde{\mathbf{C}} \\ \mathbf{B}_1^T \mathbf{S}_{n_1} \tilde{\mathbf{X}}_1 \mathbf{S}_{n_1} \mathbf{A}_1^T + \mathbf{B}_2^T \mathbf{S}_{n_2} \tilde{\mathbf{X}}_2 \mathbf{S}_{n_2} \mathbf{A}_2^T + \cdots + \mathbf{B}_l^T \mathbf{S}_{n_l} \tilde{\mathbf{X}}_l \mathbf{S}_{n_l} \mathbf{A}_l^T = \tilde{\mathbf{C}}^T \end{cases} \quad (5)$$

令

$$\tilde{\mathbf{X}}_i = \mathbf{X}_i - \overline{\overline{\mathbf{X}}_i}, \quad \tilde{\mathbf{C}} = \mathbf{C} - \sum_{i=1}^l \mathbf{A}_i \overline{\overline{\mathbf{X}}_i} \mathbf{B}_i \quad i = 1, 2, \dots, l$$

选初始矩阵(3)可求得矩阵方程组(5)的极小范数解 $[\mathbf{X}_1^*, \mathbf{X}_2^*, \dots, \mathbf{X}_l^*]$, 问题 2 的解

$$[\hat{\mathbf{X}}_1, \hat{\mathbf{X}}_2 \cdots \hat{\mathbf{X}}_l] = [\tilde{\mathbf{X}}_1^* + \overline{\overline{\mathbf{X}}_1}, \tilde{\mathbf{X}}_2^* + \overline{\overline{\mathbf{X}}_2} \cdots \tilde{\mathbf{X}}_l^* + \overline{\overline{\mathbf{X}}_l}]$$

4 数值实验

设 $n = 2$, 给定 $\mathbf{A}_1, \mathbf{B}_1, \mathbf{A}_2, \mathbf{B}_2, \mathbf{C}$ 如下.

$$\mathbf{A}_1 = \begin{bmatrix} 3 & -5 & 3 & -5 & -2 & 3 \\ 4 & 3 & -1 & 3 & -1 & 4 \\ 3 & -5 & 3 & -5 & -2 & 3 \\ 4 & 3 & -1 & 3 & -1 & 4 \\ 3 & -5 & 3 & -5 & 2 & 3 \\ 4 & 3 & -1 & 3 & -1 & 4 \end{bmatrix}$$

$$\mathbf{B}_1 = \begin{bmatrix} 3 & 9 & 3 & 9 & 3 & 9 \\ 7 & 4 & 7 & 4 & 7 & 4 \\ 4 & 3 & 4 & 3 & 4 & 3 \\ -5 & 4 & -5 & 4 & -5 & 4 \\ 3 & 4 & 3 & 4 & 3 & 4 \\ 4 & -5 & 4 & -5 & 4 & -5 \end{bmatrix}$$

$$\mathbf{A}_2 = \begin{bmatrix} 4 & 3 & -1 & 3 & -1 & 4 \\ 3 & -5 & 3 & -5 & 2 & 5 \\ 4 & 3 & -1 & 3 & -1 & 4 \\ 3 & -5 & 3 & -5 & 2 & 5 \\ 4 & 3 & -1 & 3 & -1 & 4 \\ 3 & -5 & 3 & -5 & 2 & 5 \end{bmatrix}$$

$$\mathbf{B}_2 = \begin{bmatrix} 3 & -4 & 3 & -4 & 3 & -4 \\ 5 & -3 & 5 & -3 & 5 & -3 \\ 1 & 4 & 1 & 4 & 1 & 4 \\ -5 & 2 & -5 & 2 & -5 & 2 \\ -3 & 4 & -3 & 4 & -3 & 4 \\ 4 & -3 & 4 & -3 & 4 & -3 \end{bmatrix}$$

$$\mathbf{C} = \begin{bmatrix} 68 & 91 & 54 & 77 & 51 & 83 \\ 81 & 72 & 81 & 72 & 81 & 72 \\ 68 & 91 & 54 & 77 & 51 & 83 \\ 81 & 72 & 81 & 72 & 81 & 72 \\ 68 & 91 & 54 & 77 & 51 & 83 \\ 81 & 72 & 81 & 72 & 81 & 72 \end{bmatrix}$$

$$\bar{\mathbf{X}}_1 = \begin{bmatrix} -1 & 2 & 1 & -2 & 1 & 5 \\ 2 & 0 & 1 & 0 & -1 & 1 \\ 1 & 4 & 0 & -1 & 5 & 2 \\ 2 & 1 & -3 & 1 & 0 & 1 \\ 3 & -1 & 0 & -1 & 0 & 2 \\ 2 & -1 & -2 & 1 & 2 & -1 \end{bmatrix}$$

$$\bar{\mathbf{X}}_2 = \begin{bmatrix} -2 & 0 & 3 & -2 & 1 & 1 \\ 0 & -1 & 0 & -2 & 3 & -1 \\ 3 & 2 & -1 & 2 & 0 & 1 \\ 3 & 1 & 2 & -2 & 0 & 1 \\ 2 & -1 & 1 & 0 & 5 & 3 \\ 1 & 4 & -2 & 1 & 5 & -2 \end{bmatrix}$$

1) 问题 1 的极小范数解:

$$\mathbf{X}_1^{(40)} = \begin{bmatrix} 0.7319 & 0.5058 & 0.3761 & 0.0664 & 0.4632 & 0.7319 \\ 0.5058 & -0.3734 & -0.0234 & -0.1392 & -0.43498 & 0.4632 \\ 0.3761 & -0.0234 & 0.0823 & -0.0980 & -0.1392 & 0.0664 \\ 0.0664 & -0.1392 & -0.0980 & 0.0823 & -0.0234 & 0.3761 \\ 0.4632 & -0.4349 & -0.1392 & -0.0234 & -0.3734 & 0.5058 \\ 0.7319 & 0.4632 & 0.0664 & 0.3761 & 0.5058 & 0.7319 \end{bmatrix}$$

$$\bar{\mathbf{X}}_2^{(40)} = \begin{bmatrix} -0.0476 & 0.1135 & 0.0313 & 0.1517 & 0.0930 & -0.0869 \\ 0.1135 & -0.0751 & -0.0369 & 0.0371 & 0.0399 & 0.0930 \\ 0.0313 & -0.0369 & 0.1889 & -0.1639 & 0.0371 & 0.1517 \\ 0.1517 & 0.0371 & -0.1639 & 0.1889 & -0.0751 & 0.1135 \\ 0.0930 & 0.0399 & 0.0371 & -0.0369 & -0.0751 & 0.1135 \\ -0.0869 & 0.0930 & 0.1517 & 0.0313 & 0.1135 & -0.0476 \end{bmatrix}$$

$$\|\mathbf{R}^{(40)}\| = 5.539 \times 10^{-11}, \quad \|\mathbf{A}_1 \mathbf{X}_1^{(40)} \mathbf{B}_1 + \mathbf{A}_2 \mathbf{X}_2^{(40)} \mathbf{B}_2 - \mathbf{C}\|^2 = 28.1069$$

2) 应用算法 1, 得到问题 2 的解:

$$\bar{\mathbf{X}}_1 = \begin{bmatrix} -1.6974 & 1.3851 & 0.8018 & -0.4814 & 0.9160 & 2.8026 \\ 1.3851 & 1.0136 & 1.0895 & 1.5305 & -0.3492 & 0.9160 \\ 0.8018 & 1.0895 & -0.5111 & -1.2573 & 1.5305 & -0.4814 \\ -0.4814 & 1.5305 & -1.2573 & -0.5111 & 1.0895 & 0.8018 \\ 0.9160 & -0.3492 & 1.5305 & 1.0895 & 1.0136 & 1.3851 \\ 2.8026 & 0.9160 & -0.4814 & 0.8018 & 1.3851 & -1.6974 \end{bmatrix}$$

$$\bar{\mathbf{X}}_2 = \begin{bmatrix} -1.4699 & 1.5668 & 1.5695 & -0.4404 & 1.2922 & 1.5626 \\ 1.5668 & 0.9683 & 1.0444 & -0.2963 & 1.9599 & 1.2922 \\ 1.5695 & 1.0444 & -1.0309 & 1.9353 & -0.2963 & -0.4404 \\ -0.4404 & -0.2963 & 1.9353 & -1.0309 & 1.0444 & 1.5695 \\ 1.2922 & 1.9599 & -0.2963 & 1.0444 & 0.9683 & 1.5668 \\ 1.5626 & 1.2922 & -0.4404 & 1.5695 & 1.5668 & -1.4699 \end{bmatrix}$$

$$\min_{(\mathbf{X}, \mathbf{Y}) \in SE} \{ \|\mathbf{X}_1 - \bar{\mathbf{X}}_1\|^2 + \|\mathbf{X}_2 - \bar{\mathbf{X}}_2\|^2 \} = 18.4280$$

参考文献:

- [1] XU G P, WEI M S, ZHENG D S. On Solutions of Matrix Equation $\mathbf{AXB} + \mathbf{CYD} = \mathbf{F}$ [J]. Linear Alg Appl, 1998, 279: 93–109.
- [2] 袁仕芳, 廖安平, 雷 渊. 矩阵方程 $\mathbf{AXB} + \mathbf{CYD} = \mathbf{F}$ 的对称极小范数最小二乘解 [J]. 计算数学, 2007, 29(2): 203–216.
- [3] 彭卓华, 胡锡炎, 张 磊. 矩阵方程 $\mathbf{A}_1\mathbf{X}_1\mathbf{B}_1 + \mathbf{A}_2\mathbf{X}_2\mathbf{B}_2 + \cdots + \mathbf{A}_l\mathbf{X}_l\mathbf{B}_l = \mathbf{C}$ 的中心对称解及其最佳逼近 [J]. 数学物理学报, 2009, 29A(1): 193–207.
- [4] 郑风芹, 张凯院. 求多变量线性矩阵方程组自反解的迭代算法 [J]. 数值计算与计算机应用, 2010, 31(1): 39–42.
- [5] 魏木生. 广义最小二乘问题的理论和计算 [M]. 北京: 科学出版社, 2006.

The Least Squares Bisymmetric Solution Group of the Coupled Matrix Equation and Its Optimal Approximation

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Abstract: Owing to the difficulty of obtaining the bisymmetric solution of a coupled matrix equation by matrix decomposition, an iterative method is proposed in this paper to solve the least squares bisymmetric solution group of the coupled matrix equation, and its convergence is proved. Its least-norm bisymmetric solution group can also be obtained by choosing the special kind of initial iterative matrices. In addition, the unique optimal approximation solution group to the given matrix group in Frobenius norm can be obtained.

Key words: matrix equation; least squares bisymmetric solution group; least norm solution group; optimal approximation solution group

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