

# 一类 $3 \times 3$ 块鞍点系统解的结构向后误差分析

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## 摘要

近年来, 一些作者提出了很多求解一类特殊的  $3 \times 3$  块鞍点系统的有效迭代方法。为了便于评估这些数值算法的强稳定性, 本文对这种类型的  $3 \times 3$  块鞍点系统进行了结构向后误差分析, 并给出了结构向后误差的可计算的具体表达式。数值实验表明, 该表达式可用于检验实际算法的稳定性。

## 关键词

$3 \times 3$  块鞍点问题, 向后误差, 结构向后误差

## Structured Backward Error Analysis for a Class of Block Three-by-Three Saddle Point System

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## Abstract

In recent years, a number of authors have proposed efficient iteration methods for

solving a special class of block  $3 \times 3$  saddle point systems. In order to evaluate the strong stability of these numerical algorithms, this paper perform the structured backward error analysis for this type of block  $3 \times 3$  saddle point system and present an explicit and computable formula for the structured backward error. Numerical example show that the expressions are useful for testing the stability of practical algorithms.

## Keywords

Block  $3 \times 3$  Saddle Point Problem, Backward Error, Structured Backward Error

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## 1. 引言

本文研究了以下  $3 \times 3$  块鞍点问题的结构向后误差

$$\begin{bmatrix} A & B^T & 0 \\ B & 0 & C^T \\ 0 & C & D \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} f \\ g \\ h \end{bmatrix}, \quad (1.1)$$

其中  $A \in \mathbb{R}^{n \times n}$  是对称正定矩阵,  $B \in \mathbb{R}^{m \times n}$  是行满秩矩阵, 并且  $m \leq n$ ,  $D \in \mathbb{R}^{l \times l}$  是对称半正定矩阵,  $C \in \mathbb{R}^{l \times m}$ , 且当  $D$  不是对称正定矩阵时,  $C$  是行满秩矩阵.  $B^T$  代表  $B$  的转置,  $C^T$  代表  $C$  的转置. 对  $A, B, C, D$  的假设保证了线性系统 (1.1) 有唯一解. 向后误差分析可以回答实际解决的问题与我们想要解决的问题有多接近, 并揭示数值算法的稳定性 [1, 2]. 线性系统 (1.1) 出现于众多科学与工程领域中, 比如线性等式约束的最小二乘问题 [3, 4], 求解非连续系数时变 Maxwell 方程的全离散有限元法 [5, 6], 求解发散形式线性二阶椭圆方程的双对偶混合有限元法 [7, 8] 等.

$3 \times 3$  块鞍点系统 (1.1) 可以恰当地分为广义鞍点问题, 即:

$$\left[ \begin{array}{cc|c} A & B^T & 0 \\ B & 0 & C^T \\ \hline 0 & C & D \end{array} \right] \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} f \\ g \\ h \end{bmatrix} \quad \text{或} \quad \left[ \begin{array}{c|cc} A & B^T & 0 \\ \hline B & 0 & C^T \\ 0 & C & D \end{array} \right] \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} f \\ g \\ h \end{bmatrix}.$$

近年来, 很多学者 [9–13] 对广义鞍点系统进行了结构向后误差分析, 证明了相应的数值算法具有较强的稳定性. 尽管  $3 \times 3$  块鞍点系统 (1.1) 可以被认为的广义的 ( $2 \times 2$  块) 鞍点问题, 因为它具有特

殊的  $3 \times 3$  块结构, 上述结构向后误差分析并不能准确地表示系统 (1.1). 在本文中, 我们将对  $3 \times 3$  块鞍点系统 (1.1) 进行结构向后误差分析.

为了简便, 把 (1.1) 记作

$$\mathcal{A}t = d.$$

对任意计算解  $\tilde{t} = [\tilde{x}^T, \tilde{y}^T, \tilde{z}^T]^T$ , 定义其无结构的向后误差  $\eta(\tilde{t})$  为

$$\eta(\tilde{t}) = \min_{\Delta \mathcal{A}, \Delta d} \left\{ \left\| \left( \frac{\|\Delta \mathcal{A}\|_F}{\|\mathcal{A}\|_F}, \frac{\|\Delta d\|_2}{\|d\|_2} \right) \right\|_2 : (\mathcal{A} + \Delta \mathcal{A})\tilde{t} = d + \Delta d \right\},$$

上面的式子可以进一步表示为 [14]

$$\eta(\tilde{t}) = \frac{\|d - \mathcal{A}\tilde{t}\|_2}{\sqrt{\|\mathcal{A}\|_F^2 \|\tilde{t}\|_2^2 + \|d\|_2^2}}, \quad (1.2)$$

其中  $\|\cdot\|_F$  和  $\|\cdot\|_2$  分别表示矩阵的 Frobenius 范数和 2-范数. 若  $\eta(\tilde{t})$  是机器精度的同量级, 则计算解  $\tilde{t}$  的计算过程是向后稳定的, 即  $\tilde{t}$  是一个向后稳定解. 需要注意的是, 如果系数矩阵  $\mathcal{A}$  具有某种特殊的结构, 自然要求  $\mathcal{A} + \Delta \mathcal{A}$  也具有与  $\mathcal{A}$  相同的结构. 在这种情况下, 自然要考虑结构后向误差.

令  $\tilde{t} = (\tilde{x}^T, \tilde{y}^T, \tilde{z}^T)^T$  是系统 (1.1) 的计算解, 定义结构向后误差  $\eta_S(\tilde{x}, \tilde{y}, \tilde{z})$  为

$$\eta_S(\tilde{x}, \tilde{y}, \tilde{z}) = \min_{\left( \begin{array}{c} \Delta A, \Delta B, \Delta C, \\ \Delta D, \Delta f, \Delta g, \\ \Delta h \end{array} \right) \in \mathcal{F}} \left\| \left[ \begin{array}{ccc} \frac{\|\Delta A\|_F}{\|\mathcal{A}\|_F} & \frac{\|\Delta B\|_F}{\|\mathcal{B}\|_F} & \frac{\|\Delta C\|_F}{\|\mathcal{C}\|_F} \\ \frac{\|\Delta D\|_F}{\|\mathcal{D}\|_F} & \frac{\|\Delta f\|_2}{\|f\|_2} & \frac{\|\Delta g\|_2}{\|g\|_2} \\ \frac{\|\Delta h\|_2}{\|h\|_2} & 0 & 0 \end{array} \right] \right\|_F,$$

其中

$$\mathcal{F} = \left\{ \left( \begin{array}{c} \Delta A, \Delta B, \Delta C, \\ \Delta D, \Delta f, \Delta g, \\ \Delta h \end{array} \right) : \left[ \begin{array}{ccc} A + \Delta A & (B + \Delta B)^T & 0 \\ B + \Delta B & 0 & (C + \Delta C)^T \\ 0 & C + \Delta C & D + \Delta D \end{array} \right] \left[ \begin{array}{c} \tilde{x} \\ \tilde{y} \\ \tilde{z} \end{array} \right] = \left[ \begin{array}{c} f + \Delta f \\ g + \Delta g \\ h + \Delta h \end{array} \right], \begin{array}{l} \Delta A = \Delta A^T, \\ \Delta D = \Delta D^T \end{array} \right\}. \quad (1.3)$$

若计算解  $\tilde{t}$  的结构向后误差是机器精度的同量级, 则计算解  $\tilde{t}$  是一个结构向后稳定解, 相应的数值算法是结构向后稳定的 (或强稳定 [15]). 因此, 给出结构向后误差  $\eta_S(\tilde{x}, \tilde{y}, \tilde{z})$  的可计算的具体表达式将有助于测试实际数值算法的稳定性. 为此, 进一步定义  $\eta^{(\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2, \lambda_3)}(\tilde{x}, \tilde{y}, \tilde{z})$  为

$$\begin{aligned} & \eta^{(\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2, \lambda_3)}(\tilde{x}, \tilde{y}, \tilde{z}) \\ &= \min_{\left( \begin{array}{c} \Delta A, \Delta B, \Delta C, \\ \Delta D, \Delta f, \Delta g, \\ \Delta h \end{array} \right) \in \mathcal{F}} \left\| \left[ \begin{array}{ccc} \theta_1 \|\Delta A\|_F & \theta_2 \|\Delta B\|_F & \theta_3 \|\Delta C\|_F \\ \theta_4 \|\Delta D\|_F & \lambda_1 \|\Delta f\|_2 & \lambda_2 \|\Delta g\|_2 \\ \lambda_3 \|\Delta h\|_2 & 0 & 0 \end{array} \right] \right\|_F, \end{aligned} \quad (1.4)$$

其中  $\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2$  和  $\lambda_3$  为正参数. 若  $A \neq 0, B \neq 0, C \neq 0, D \neq 0, f \neq 0, g \neq 0$  和  $h \neq 0$ , 则取

$$\tilde{\theta}_1 = \frac{1}{\|A\|_F}, \tilde{\theta}_2 = \frac{1}{\|B\|_F}, \tilde{\theta}_3 = \frac{1}{\|C\|_F}, \tilde{\theta}_4 = \frac{1}{\|D\|_F}, \tilde{\lambda}_1 = \frac{1}{\|f\|_2}, \tilde{\lambda}_2 = \frac{1}{\|g\|_2}, \text{ and } \tilde{\lambda}_3 = \frac{1}{\|h\|_2},$$

从而有

$$\eta_S(\tilde{x}, \tilde{y}, \tilde{z}) = \eta^{(\tilde{\theta}_1, \tilde{\theta}_2, \tilde{\theta}_3, \tilde{\theta}_4, \tilde{\lambda}_1, \tilde{\lambda}_2, \tilde{\lambda}_3)}(\tilde{x}, \tilde{y}, \tilde{z}). \quad (1.5)$$

本文结构安排如下:

在第 2 节中, 主要介绍了 Kronecker 积, 和后面证明中需要用到的几个关键的引理.

在第 3 节中, 推导了  $3 \times 3$  块鞍点系统 (1.1) 结构向后误差的具体的可计算的表达式.

在第 4 节中, 通过数值实验来证明推导出的表达式对于测试求解  $3 \times 3$  块鞍点系统的实际数值算法稳定性是有用的.

在第 5 节中, 对全文工作进行了总结.

## 2. 预备知识

本文用  $\mathbb{R}^{m \times n}$  和  $\mathbb{S}\mathbb{R}^{n \times n}$  分别表示  $m \times n$  阶实矩阵的集合和  $n \times n$  阶实对称矩阵的集合, 用  $A^\dagger$  表示矩阵  $A$  的 Moore-Penrose 逆, 设  $X = (x_{ij}) \in \mathbb{R}^{m \times n}$ ,  $Z \in \mathbb{R}^{p \times q}$ .  $X$  和  $Z$  之间的 Kronecker 积定义为(见文献 [16], 第 4 章)

$$X \otimes Z = (x_{ij}Z) \in \mathbb{R}^{mp \times nq}.$$

由文献 [16]知

$$\text{vec}(XYZ) = (Z^T \otimes X) \text{vec}(Y), \quad (2.1)$$

$$(X \otimes Z)^T = X^T \otimes Z^T, \quad (2.2)$$

$$(X \otimes Z)(C \otimes G) = (XC \otimes ZG), \quad (2.3)$$

其中  $Y \in \mathbb{R}^{n \times p}$ ,  $C$  和  $G$  为合适阶数的矩阵,  $\text{vec}(Y)$  定义为将矩阵  $Y$  的所有列依次堆叠成一个向量后得到的向量.

在本节中, 我们将介绍一些引理, 这些引理将在接下来的章节中使用.

**引理 2.1** ([11, 17]) 设  $u \in \mathbb{R}^m$  和  $p \in \mathbb{R}^n$  已知. 定义

$$\mathcal{X} = \{X \in \mathbb{R}^{n \times m} : Xu = p\}.$$

则,  $\mathcal{X} \neq \emptyset$  当且仅当  $pu^\dagger u = p$ , 且在此情况下, 任意的  $X \in \mathcal{X}$  可表示为

$$X = pu^\dagger + Z(I_m - uu^\dagger), Z \in \mathbb{R}^{n \times m}.$$

引理2.2 ([11, 17]) 设  $b, c \in \mathbb{R}^n$  已知. 定义

$$\mathcal{H} = \{H \in \mathbb{S}\mathbb{R}^{n \times n} : Hb = c\}.$$

则,  $\mathcal{H} \neq \emptyset$  当且仅当  $b$  和  $c$  满足  $cb^\dagger b = c$ , 且在此条件下, 任意的  $H \in \mathcal{H}$  可表示为

$$H = cb^\dagger + (b^\dagger)^T c^T (I_n - bb^\dagger) + (I_n - bb^\dagger) T (I_n - bb^\dagger),$$

其中  $T \in \mathbb{S}\mathbb{R}^{n \times n}$ .

引理2.3 ([11, 17]) 假设  $F \in \mathbb{R}^{p \times m}$ ,  $G \in \mathbb{R}^{n \times q}$  和  $K \in \mathbb{R}^{p \times q}$  已知, 并令  $X^* = F^\dagger K G^\dagger$ . 则

$$\min_{X \in \mathbb{R}^{m \times n}} \|FXG - K\|_F = \|FX^*G - K\|_F.$$

### 3. 求解系统 (1.1) 的结构向后误差问题

为了给出  $\eta^{(\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2, \lambda_3)}(\tilde{x}, \tilde{y}, \tilde{z})$  的结构向后误差的明确表达式. 我们首先研究了部分结构向后误差  $\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z})$ , 其定义为

$$\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z}) = \min_{(\Delta A, \Delta B, \Delta C, \Delta D) \in \mathcal{F}_0} \left\| \begin{bmatrix} \theta_1 \|\Delta A\|_F & \theta_2 \|\Delta B\|_F \\ \theta_3 \|\Delta C\|_F & \theta_4 \|\Delta D\|_F \end{bmatrix} \right\|_F, \quad (3.1)$$

其中

$$\mathcal{F}_0 = \left\{ \left( \begin{array}{c} \Delta A, \Delta B, \\ \Delta C, \Delta D \end{array} \right) : \begin{bmatrix} A + \Delta A & (B + \Delta B)^T & 0 \\ B + \Delta B & 0 & (C + \Delta C)^T \\ 0 & C + \Delta C & D + \Delta D \end{bmatrix} \begin{bmatrix} \tilde{x} \\ \tilde{y} \\ \tilde{z} \end{bmatrix} = \begin{bmatrix} f \\ g \\ h \end{bmatrix}, \begin{array}{l} \Delta A = \Delta A^T, \\ \Delta D = \Delta D^T \end{array} \right\}. \quad (3.2)$$

下面给出  $\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z})$  的明确表达式.

定理 3.1 假设  $(\tilde{x}^T, \tilde{y}^T, \tilde{z}^T)^T$  满足  $\tilde{x} \neq 0$  和  $\tilde{z} \neq 0$  为系统 (1.1) 的一个计算解. 则

$$\begin{aligned} [\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z})]^2 &= \frac{2\theta_1^2\theta_2^2}{\gamma_1} \|r_f\|_2^2 + \frac{\theta_2^2\theta_3^2}{\gamma_2} \|r_g\|_2^2 + \frac{2\theta_3^2\theta_4^2}{\gamma_3} \|r_h\|_2^2 + \frac{\theta_1^2\gamma_5}{\gamma_1\gamma_4} (r_f^T \tilde{x})^2 \\ &+ \frac{\gamma_6}{\gamma_2\gamma_4} (r_g^T \tilde{y})^2 + \frac{\theta_4^2\gamma_7}{\gamma_3\gamma_4} (r_h^T \tilde{z})^2 - \frac{2\theta_1^2\gamma_8}{\gamma_4} (r_f^T \tilde{x}) (r_g^T \tilde{y}) \\ &+ \frac{2\theta_1^2\theta_4^2\|\tilde{y}\|_2^2}{\gamma_4} (r_f^T \tilde{x}) (r_h^T \tilde{z}) - \frac{2\theta_4^2\gamma_9}{\gamma_4} (r_g^T \tilde{y}) (r_h^T \tilde{z}), \end{aligned} \quad (3.3)$$

其中

$$r_f = f - A\tilde{x} - B^T\tilde{y}, \quad r_g = g - B\tilde{x} - C^T\tilde{z}, \quad r_h = h - C\tilde{y} - D\tilde{z},$$

$$\begin{aligned}
\gamma_1 &= \theta_2^2 \|\tilde{x}\|_2^2 + 2\theta_1^2 \|\tilde{y}\|_2^2, \quad \gamma_2 = \theta_3^2 \|\tilde{x}\|_2^2 + \theta_2^2 \|\tilde{z}\|_2^2, \quad \gamma_3 = \theta_3^2 \|\tilde{z}\|_2^2 + 2\theta_4^2 \|\tilde{y}\|_2^2, \\
\gamma_4 &= \theta_1^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^4 + \theta_3^2 \|\tilde{x}\|_2^4 \|\tilde{z}\|_2^2 + \theta_2^2 \|\tilde{x}\|_2^2 \|\tilde{z}\|_2^4 + \theta_4^2 \|\tilde{x}\|_2^4 \|\tilde{y}\|_2^2, \\
\gamma_5 &= 2\theta_1^2 \theta_3^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 + 2\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^4 - \theta_2^2 \theta_3^2 \|\tilde{x}\|_2^2 \|\tilde{z}\|_2^2 - \theta_2^4 \|\tilde{z}\|_2^4 - \theta_2^2 \theta_4^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2, \\
\gamma_6 &= \theta_1^2 \theta_3^4 \|\tilde{x}\|_2^2 \|\tilde{z}\|_2^2 + \theta_1^2 \theta_3^2 \theta_4^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 + \theta_1^2 \theta_2^2 \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 + \theta_2^4 \theta_4^2 \|\tilde{x}\|_2^2 \|\tilde{z}\|_2^2, \\
\gamma_7 &= 2\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^4 + 2\theta_2^2 \theta_4^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 - \theta_1^2 \theta_3^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 - \theta_3^4 \|\tilde{x}\|_2^4 - \theta_2^2 \theta_3^2 \|\tilde{x}\|_2^2 \|\tilde{z}\|_2^2, \\
\gamma_8 &= \theta_3^2 \|\tilde{z}\|_2^2 + \theta_4^2 \|\tilde{y}\|_2^2, \quad \gamma_9 = \theta_2^2 \|\tilde{x}\|_2^2 + \theta_1^2 \|\tilde{y}\|_2^2.
\end{aligned}$$

**证明** 由 (3.2) 知,  $(\Delta A, \Delta B, \Delta C, \Delta D) \in \mathcal{F}_0$  当且仅当  $\Delta A, \Delta B, \Delta C$  和  $\Delta D$  满足

$$\Delta A \tilde{x} = r_f - \Delta B^T \tilde{y}, \quad \Delta B \tilde{x} = r_g - \Delta C^T \tilde{z}, \quad \Delta D \tilde{z} = r_h - \Delta C \tilde{y}, \quad \Delta A = \Delta A^T, \quad \Delta D = \Delta D^T. \quad (3.4)$$

将引理 2.1 应用于 (3.4) 的第二个等式, 可得

$$\Delta B^T = (\tilde{x}^\dagger)^T (r_g - \Delta C^T \tilde{z})^T + (I_n - \tilde{x} \tilde{x}^\dagger) Z, \quad Z \in \mathbb{R}^{n \times m}. \quad (3.5)$$

考虑到  $\Delta D = \Delta D^T$ , 应用引理 2.2 于 (3.4) 的第三个等式, 对于任意  $\Delta C \in \mathbb{R}^{l \times m}$ , 可得

$$\Delta D = (r_h - \Delta C \tilde{y}) \tilde{z}^\dagger + (\tilde{z}^\dagger)^T (r_h - \Delta C \tilde{y})^T (I_l - \tilde{z} \tilde{z}^\dagger) + (I_l - \tilde{z} \tilde{z}^\dagger) T_2 (I_l - \tilde{z} \tilde{z}^\dagger), \quad (3.6)$$

其中  $T_2 \in \mathbb{S}\mathbb{R}^{l \times l}$ .

考虑到  $\Delta A = \Delta A^T$ , 把 (3.5) 带入到 (3.4) 的第一个等式, 再次应用引理 2.2, 对于任意  $\Delta C \in \mathbb{R}^{l \times m}$ , 可推出

$$\Delta A = r \tilde{x}^\dagger + (\tilde{x}^\dagger)^T r^T (I_n - \tilde{x} \tilde{x}^\dagger) + (I_n - \tilde{x} \tilde{x}^\dagger) T_1 (I_n - \tilde{x} \tilde{x}^\dagger), \quad T_1 \in \mathbb{S}\mathbb{R}^{n \times n}, \quad (3.7)$$

其中

$$r = r_f - (\tilde{x}^\dagger)^T (r_g^T \tilde{y} - \tilde{z}^T \Delta C \tilde{y}) - (I_n - \tilde{x} \tilde{x}^\dagger) Z \tilde{y}$$

考虑到  $\tilde{x}^\dagger (I_n - \tilde{x} \tilde{x}^\dagger) = 0$  这一事实, 对 (3.5), (3.6) 和 (3.7) 的等号两边同时取 Frobenius 范数, 可得

$$\|\Delta B\|_F^2 = \frac{\|r_g - \Delta C^T \tilde{z}\|_2^2}{\|\tilde{x}\|_2^2} + \|(I_n - \tilde{x} \tilde{x}^\dagger) Z\|_F^2, \quad (3.8)$$

$$\begin{aligned}
\|\Delta D\|_F^2 &= \frac{\|r_h - \Delta C \tilde{y}\|_2^2}{\|\tilde{z}\|_2^2} + \|(I_l - \tilde{z} \tilde{z}^\dagger) T_2 (I_l - \tilde{z} \tilde{z}^\dagger)\|_F^2 + \frac{\|(I_l - \tilde{z} \tilde{z}^\dagger) (r_h - \Delta C \tilde{y})\|_2^2}{\|\tilde{z}\|_2^2} \\
&= \frac{2 \|r_h - \Delta C \tilde{y}\|_2^2}{\|\tilde{z}\|_2^2} - \frac{(r_h^T \tilde{z} - \tilde{z}^T \Delta C \tilde{y})^2}{\|\tilde{z}\|_2^4} + \|(I_l - \tilde{z} \tilde{z}^\dagger) T_2 (I_l - \tilde{z} \tilde{z}^\dagger)\|_F^2,
\end{aligned} \quad (3.9)$$

和

$$\begin{aligned} \|\Delta A\|_F^2 &= \frac{\|r\|_2^2}{\|\tilde{x}\|_2^2} + \|(I_n - \tilde{x}\tilde{x}^\dagger) T_1 (I_n - \tilde{x}\tilde{x}^\dagger)\|_F^2 + \frac{\|(I_n - \tilde{x}\tilde{x}^\dagger) r\|_2^2}{\|\tilde{x}\|_2^2} \\ &= \frac{2\|r_f - (\tilde{x}^\dagger)^T (r_g^T \tilde{y} - \tilde{z}^T \Delta C \tilde{y}) - (I_n - \tilde{x}\tilde{x}^\dagger) Z \tilde{y}\|_2^2}{\|\tilde{x}\|_2^2} \\ &\quad - \frac{(r_f^T \tilde{x} - r_g^T \tilde{y} + \tilde{z}^T \Delta C \tilde{y})^2}{\|\tilde{x}\|_2^4} + \|(I_n - \tilde{x}\tilde{x}^\dagger) T_1 (I_n - \tilde{x}\tilde{x}^\dagger)\|_F^2, \end{aligned} \quad (3.10)$$

根据  $\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z})$  的定义 (3.1), 以及表达式 (3.8), (3.9) 和 (3.10) 可以得出

$$\begin{aligned} &[\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z})]^2 \\ &= \min_{\substack{\Delta C \in \mathbb{R}^{l \times m}, Z \in \mathbb{R}^{n \times m} \\ T_1 \in \mathcal{S} \mathbb{R}^{n \times n}, T_2 \in \mathcal{S} \mathbb{R}^{l \times l}}} \left\{ \theta_1^2 \|\Delta A\|_F^2 + \theta_2^2 \|\Delta B\|_F^2 + \theta_3^2 \|\Delta C\|_F^2 + \theta_4^2 \|\Delta D\|_F^2 \right\} \\ &= \min_{\Delta C \in \mathbb{R}^{l \times m}, Z \in \mathbb{R}^{n \times m}} p(\Delta C), \end{aligned} \quad (3.11)$$

其中

$$p(\Delta C) = p_1(Z, \Delta C) + p_2(\Delta C), \quad (3.12)$$

并且

$$p_1(Z, \Delta C) = \frac{2\theta_1^2}{\|\tilde{x}\|_2^2} \|r_f - (\tilde{x}^\dagger)^T (r_g^T \tilde{y} - \tilde{z}^T \Delta C \tilde{y}) - (I_n - \tilde{x}\tilde{x}^\dagger) Z \tilde{y}\|_2^2 + \theta_2^2 \|(I_n - \tilde{x}\tilde{x}^\dagger) Z\|_F^2,$$

和

$$\begin{aligned} p_2(\Delta C) &= \frac{2\theta_4^2}{\|\tilde{z}\|_2^2} \|r_h - \Delta C \tilde{y}\|_2^2 + \frac{\theta_2^2}{\|\tilde{x}\|_2^2} \|r_g - \Delta C^T \tilde{z}\|_2^2 - \frac{\theta_4^2}{\|\tilde{z}\|_2^4} (r_h^T \tilde{z} - \tilde{z}^T \Delta C \tilde{y})^2 \\ &\quad - \frac{\theta_1^2}{\|\tilde{x}\|_2^4} (r_f^T \tilde{x} - r_g^T \tilde{y} + \tilde{z}^T \Delta C \tilde{y})^2 + \theta_3^2 \|\Delta C\|_F^2. \end{aligned} \quad (3.13)$$

为了便于书写, 记

$$\hat{g} = r_f - (\tilde{x}^\dagger)^T (r_g^T \tilde{y} - \tilde{z}^T \Delta C \tilde{y}), \quad Y = (I_n - \tilde{x}\tilde{x}^\dagger) Z$$

则

$$\begin{aligned} p_1(Z, \Delta C) &= \frac{2\theta_1^2 \|\hat{g} - Y \tilde{y}\|_2^2}{\|\tilde{x}\|_2^2} + \theta_2^2 \|Y\|_F^2 \\ &= \frac{2\theta_1^2 \|\hat{g}\|_2^2}{\|\tilde{x}\|_2^2} + \text{tr} \left[ Y \left( \theta_2^2 I_m + \frac{2\theta_1^2 \tilde{y} \tilde{y}^T}{\|\tilde{x}\|_2^2} \right) Y^T - \frac{2\theta_1^2 Y \tilde{y} \hat{g}^T}{\|\tilde{x}\|_2^2} - \frac{2\theta_1^2 \hat{g} \tilde{y}^T Y^T}{\|\tilde{x}\|_2^2} \right]. \end{aligned}$$

另外, 记

$$M = Y \left( \theta_2^2 I_m + \frac{2\theta_1^2 \tilde{y} \tilde{y}^T}{\|\tilde{x}\|_2^2} \right)^{\frac{1}{2}}, \quad N^T = \frac{2\theta_1^2}{\|\tilde{x}\|_2^2} \left( \theta_2^2 I_m + \frac{2\theta_1^2 \tilde{y} \tilde{y}^T}{\|\tilde{x}\|_2^2} \right)^{-\frac{1}{2}} \tilde{y} \hat{g}^T.$$

利用引理 2.3, 并注意到

$$\|N\|_F^2 - \|\tilde{x}\tilde{x}^\dagger N\|_F^2 = \|(I_n - \tilde{x}\tilde{x}^\dagger) N\|_F^2,$$

则

$$\begin{aligned}
& \min_{Z \in \mathbb{R}^{n \times m}} p_1(Z, \Delta C) \\
&= \min_{Z \in \mathbb{R}^{n \times m}} \left\{ \frac{2\theta_1^2 \|\hat{g}\|_2^2}{\|\tilde{x}\|_2^2} + \|M - N\|_F^2 - \|N\|_F^2 \right\} \\
&= \frac{2\theta_1^2 \|\hat{g}\|_2^2}{\|\tilde{x}\|_2^2} - \|(I_n - \tilde{x}\tilde{x}^\dagger)N\|_F^2 \\
&= \frac{2\theta_1^2 \|\hat{g}\|_2^2}{\|\tilde{x}\|_2^2} - \frac{4\theta_1^4}{\|\tilde{x}\|_2^4} \operatorname{tr} \left[ (I_n - \tilde{x}\tilde{x}^\dagger) \hat{g}\tilde{y}^T \left( \theta_2^2 I_m + \frac{2\theta_1^2 \tilde{y}\tilde{y}^T}{\|\tilde{x}\|_2^2} \right)^{-1} \tilde{y}\hat{g}^T (I_n - \tilde{x}\tilde{x}^\dagger) \right].
\end{aligned}$$

此外, 使用 Sherman-Morrison-Woodbury 公式(见文献 [18])可以将上面的式子改写为下面的形式

$$\begin{aligned}
& \min_{Z \in \mathbb{R}^{n \times m}} p_1(Z, \Delta C) \\
&= \frac{2\theta_1^2 \|\hat{g}\|_2^2}{\|\tilde{x}\|_2^2} - \frac{4\theta_1^4}{\theta_2^2 \|\tilde{x}\|_2^4} \operatorname{tr} \left[ (I_n - \tilde{x}\tilde{x}^\dagger) \hat{g}\tilde{y}^T \left( I_m - \frac{2\theta_1^2 \tilde{y}\tilde{y}^T}{\theta_2^2 \|\tilde{x}\|_2^2 + 2\theta_1^2 \|\tilde{y}\|_2^2} \right) \tilde{y}\hat{g}^T (I_n - \tilde{x}\tilde{x}^\dagger) \right] \\
&= \frac{2\theta_1^2 \|\hat{g}\|_2^2}{\|\tilde{x}\|_2^2} - \frac{4\theta_1^4 \|\tilde{y}\|_2^2 \|(I_n - \tilde{x}\tilde{x}^\dagger)\hat{g}\|_2^2}{\|\tilde{x}\|_2^2 (\theta_2^2 \|\tilde{x}\|_2^2 + 2\theta_1^2 \|\tilde{y}\|_2^2)} \\
&= \frac{2\theta_1^2 \theta_2^2 \|r_f\|_2^2}{\theta_2^2 \|\tilde{x}\|_2^2 + 2\theta_1^2 \|\tilde{y}\|_2^2} + \frac{4\theta_1^4 \|\tilde{y}\|_2^2 (r_f^T \tilde{x})^2}{\|\tilde{x}\|_2^4 (\theta_2^2 \|\tilde{x}\|_2^2 + 2\theta_1^2 \|\tilde{y}\|_2^2)} - \frac{4\theta_1^2 (r_f^T \tilde{x}) (r_g^T \tilde{y})}{\|\tilde{x}\|_2^4} + \frac{2\theta_1^2 (r_g^T \tilde{y})^2}{\|\tilde{x}\|_2^4} \\
&\quad + \frac{4\theta_1^2 (r_f^T \tilde{x}) (\tilde{z}^T \Delta C \tilde{y})}{\|\tilde{x}\|_2^4} - \frac{4\theta_1^2 (r_g^T \tilde{y}) (\tilde{z}^T \Delta C \tilde{y})}{\|\tilde{x}\|_2^4} + \frac{2\theta_1^2 (\tilde{z}^T \Delta C \tilde{y})^2}{\|\tilde{x}\|_2^4}.
\end{aligned} \tag{3.14}$$

把 (3.13) 和 (3.14) 带入 (3.12) 中, 可以得到

$$\begin{aligned}
& p(\Delta C) \\
&= \frac{2\theta_1^2 \theta_2^2 \|r_f\|_2^2}{\theta_2^2 \|\tilde{x}\|_2^2 + 2\theta_1^2 \|\tilde{y}\|_2^2} + \frac{4\theta_1^4 \|\tilde{y}\|_2^2 (r_f^T \tilde{x})^2}{\|\tilde{x}\|_2^4 (\theta_2^2 \|\tilde{x}\|_2^2 + 2\theta_1^2 \|\tilde{y}\|_2^2)} - \frac{4\theta_1^2 (r_f^T \tilde{x}) (r_g^T \tilde{y})}{\|\tilde{x}\|_2^4} + \frac{2\theta_1^2 (r_g^T \tilde{y})^2}{\|\tilde{x}\|_2^4} \\
&\quad + \frac{4\theta_1^2 (r_f^T \tilde{x}) (\tilde{z}^T \Delta C \tilde{y})}{\|\tilde{x}\|_2^4} - \frac{4\theta_1^2 (r_g^T \tilde{y}) (\tilde{z}^T \Delta C \tilde{y})}{\|\tilde{x}\|_2^4} + \frac{2\theta_1^2 (\tilde{z}^T \Delta C \tilde{y})^2}{\|\tilde{x}\|_2^4} + \frac{2\theta_4^2 \|r_h - \Delta C \tilde{y}\|_2^2}{\|\tilde{z}\|_2^2} \\
&\quad + \frac{\theta_2^2 \|r_g - \Delta C^T \tilde{z}\|_2^2}{\|\tilde{x}\|_2^2} - \frac{\theta_4^2 (r_h^T \tilde{z} - \tilde{z}^T \Delta C \tilde{y})^2}{\|\tilde{z}\|_2^4} - \frac{\theta_1^2 (r_f^T \tilde{x} - r_g^T \tilde{y} + \tilde{z}^T \Delta C \tilde{y})^2}{\|\tilde{x}\|_2^4} + \theta_3^2 \|\Delta C\|_F^2 \\
&= \frac{2\theta_1^2 \theta_2^2 \|r_f\|_2^2}{\theta_2^2 \|\tilde{x}\|_2^2 + 2\theta_1^2 \|\tilde{y}\|_2^2} + \frac{\theta_2^2 \|r_g\|_2^2}{\|\tilde{x}\|_2^2} + \frac{2\theta_4^2 \|r_h\|_2^2}{\|\tilde{z}\|_2^2} + \frac{\theta_1^2 (2\theta_1^2 \|\tilde{y}\|_2^2 - \theta_2^2 \|\tilde{x}\|_2^2) (r_f^T \tilde{x})^2}{\|\tilde{x}\|_2^4 (\theta_2^2 \|\tilde{x}\|_2^2 + 2\theta_1^2 \|\tilde{y}\|_2^2)} \\
&\quad - \frac{2\theta_1^2 (r_f^T \tilde{x}) (r_g^T \tilde{y})}{\|\tilde{x}\|_2^4} + \frac{\theta_1^2 (r_g^T \tilde{y})^2}{\|\tilde{x}\|_2^4} - \frac{\theta_4^2 (r_h^T \tilde{z})^2}{\|\tilde{z}\|_2^4} + \frac{2\theta_1^2 (r_f^T \tilde{x}) (\tilde{z}^T \Delta C \tilde{y})}{\|\tilde{x}\|_2^4} - \frac{2\theta_1^2 (r_g^T \tilde{y}) (\tilde{z}^T \Delta C \tilde{y})}{\|\tilde{x}\|_2^4} \\
&\quad + \frac{2\theta_4^2 (r_h^T \tilde{z}) (\tilde{z}^T \Delta C \tilde{y})}{\|\tilde{z}\|_2^4} + \frac{(\theta_1^2 \|\tilde{z}\|_2^4 - \theta_4^2 \|\tilde{x}\|_2^4) (\tilde{z}^T \Delta C \tilde{y})^2}{\|\tilde{x}\|_2^4 \|\tilde{z}\|_2^4} - \frac{2\theta_2^2 (\tilde{z}^T \Delta C r_g)}{\|\tilde{x}\|_2^2} + \frac{\theta_2^2 \|\tilde{z}^T \Delta C\|_2^2}{\|\tilde{x}\|_2^2} \\
&\quad - \frac{4\theta_4^2 (r_h^T \Delta C \tilde{y})}{\|\tilde{z}\|_2^2} + \frac{2\theta_4^2 \|\Delta C \tilde{y}\|_2^2}{\|\tilde{z}\|_2^2} + \theta_3^2 \|\Delta C\|_F^2.
\end{aligned}$$



记  $s = \text{vec}(\Delta C) \in \mathbb{R}^{ml}$ , 利用 Kronecker 积的性质 (2.1) 和 (2.2), 上面的式子可以进一步化为

$$\begin{aligned}
& p(\Delta C) \\
&= \frac{2\theta_1^2\theta_2^2\|r_f\|_2^2}{\theta_2^2\|\tilde{x}\|_2^2 + 2\theta_1^2\|\tilde{y}\|_2^2} + \frac{\theta_2^2\|r_g\|_2^2}{\|\tilde{x}\|_2^2} + \frac{2\theta_4^2\|r_h\|_2^2}{\|\tilde{z}\|_2^2} + \frac{\theta_1^2(2\theta_1^2\|\tilde{y}\|_2^2 - \theta_2^2\|\tilde{x}\|_2^2)(r_f^T x)^2}{\|\tilde{x}\|_2^4(\theta_2^2\|\tilde{x}\|_2^2 + 2\theta_1^2\|\tilde{y}\|_2^2)} \\
&\quad - \frac{2\theta_1^2(r_f^T \tilde{x})(r_g^T \tilde{y})}{\|\tilde{x}\|_2^4} + \frac{\theta_1^2(r_g^T \tilde{y})^2}{\|\tilde{x}\|_2^4} - \frac{\theta_4^2(r_h^T \tilde{z})^2}{\|\tilde{z}\|_2^4} + \theta_3^2 s^T I_{ml} s + \frac{2\theta_1^2(r_f^T \tilde{x})(\tilde{y}^T \otimes \tilde{z}^T) s}{\|\tilde{x}\|_2^4} \\
&\quad - \frac{2\theta_1^2(r_g^T \tilde{y})(\tilde{y}^T \otimes \tilde{z}^T) s}{\|\tilde{x}\|_2^4} + \frac{2\theta_4^2(r_h^T \tilde{z})(\tilde{y}^T \otimes \tilde{z}^T) s}{\|\tilde{z}\|_2^4} \\
&\quad + \frac{(\theta_1^2\|\tilde{z}\|_2^4 - \theta_4^2\|\tilde{x}\|_2^4) s^T (\tilde{y} \otimes \tilde{z})(\tilde{y}^T \otimes \tilde{z}^T) s}{\|\tilde{x}\|_2^4\|\tilde{z}\|_2^4} - \frac{2\theta_2^2(r_g^T \otimes \tilde{z}^T) s}{\|\tilde{x}\|_2^2} \\
&\quad + \frac{\theta_2^2 s^T (I_m \otimes \tilde{z})(I_m \otimes \tilde{z}^T) s}{\|\tilde{x}\|_2^2} - \frac{4\theta_4^2(\tilde{y}^T \otimes r_h^T) s}{\|\tilde{z}\|_2^2} + \frac{2\theta_4^2 s^T (\tilde{y} \otimes I_l)(\tilde{y}^T \otimes I_l) s}{\|\tilde{z}\|_2^2}
\end{aligned}$$

将上面的等式带入到(3.11)中, 并考虑(2.2), 可以得到

$$\begin{aligned}
[\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{z}, \tilde{x}, \tilde{y})]^2 &= \frac{2\theta_1^2\theta_2^2\|r_f\|_2^2}{\theta_2^2\|\tilde{x}\|_2^2 + 2\theta_1^2\|\tilde{y}\|_2^2} + \frac{\theta_2^2\|r_g\|_2^2}{\|\tilde{x}\|_2^2} + \frac{2\theta_4^2\|r_h\|_2^2}{\|\tilde{z}\|_2^2} - \frac{2\theta_1^2(r_f^T \tilde{x})(r_g^T \tilde{y})}{\|\tilde{x}\|_2^4} \\
&\quad + \frac{\theta_1^2(2\theta_1^2\|\tilde{y}\|_2^2 - \theta_2^2\|\tilde{x}\|_2^2)(r_f^T x)^2}{\|\tilde{x}\|_2^4(\theta_2^2\|\tilde{x}\|_2^2 + 2\theta_1^2\|\tilde{y}\|_2^2)} + \frac{\theta_1^2(r_g^T \tilde{y})^2}{\|\tilde{x}\|_2^4} - \frac{\theta_4^2(r_h^T \tilde{z})^2}{\|\tilde{z}\|_2^4} \\
&\quad + \min_{s \in \mathbb{R}^{ml}} H(s),
\end{aligned}$$

这里  $H(s) = s^T E s - 2e^T s$ , 其中

$$\begin{aligned}
E &= \theta_3^2 I_{ml} + \frac{\theta_2^2(I_m \otimes \tilde{z})(I_m \otimes \tilde{z}^T)}{\|\tilde{x}\|_2^2} + \frac{2\theta_4^2(\tilde{y} \otimes I_l)(\tilde{y}^T \otimes I_l)}{\|\tilde{z}\|_2^2} \\
&\quad + \frac{(\theta_1^2\|\tilde{z}\|_2^4 - \theta_4^2\|\tilde{x}\|_2^4)(\tilde{y} \otimes \tilde{z})(\tilde{y}^T \otimes \tilde{z}^T)}{\|\tilde{x}\|_2^4\|\tilde{z}\|_2^4},
\end{aligned}$$

和

$$e = \frac{\theta_2^2(r_g \otimes \tilde{z})}{\|\tilde{x}\|_2^2} + \frac{2\theta_4^2(\tilde{y} \otimes r_h)}{\|\tilde{z}\|_2^2} - \left( \frac{\theta_1^2(r_f^T \tilde{x})}{\|\tilde{x}\|_2^4} - \frac{\theta_1^2(r_g^T \tilde{y})}{\|\tilde{x}\|_2^4} + \frac{\theta_4^2(r_h^T \tilde{z})}{\|\tilde{z}\|_2^4} \right) (\tilde{y} \otimes \tilde{z}),$$

利用(2.3), 并注意到  $(I_l - \tilde{z}\tilde{z}^\dagger)^2 = (I_l - \tilde{z}\tilde{z}^\dagger)$  和  $\tilde{z}^\dagger = \tilde{z}^T/\|\tilde{z}\|_2^2$ , 可以得出

$$\begin{aligned}
E &= \theta_3^2 I_{ml} + \frac{\theta_2^2(I_m \otimes \tilde{z})(I_m \otimes \tilde{z}^T)}{\|\tilde{x}\|_2^2} + \frac{\theta_4^2(\tilde{y} \otimes I_l)(\tilde{y}^T \otimes I_l)}{\|\tilde{z}\|_2^2} + \frac{\theta_1^2(\tilde{y} \otimes \tilde{z})(\tilde{y}^T \otimes \tilde{z}^T)}{\|\tilde{x}\|_2^4} \\
&\quad + \frac{\theta_4^2[\tilde{y} \otimes (I_l - \tilde{z}\tilde{z}^\dagger)] [\tilde{y}^T \otimes (I_l - \tilde{z}\tilde{z}^\dagger)]}{\|\tilde{z}\|_2^2}.
\end{aligned}$$

明显地,  $E$  是一个对称正定矩阵. 从而当  $s = E^{-1}e$  时,  $H(s)$  能取得最小值. 相应的,

$$\begin{aligned} [\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z})]^2 &= \frac{2\theta_1^2\theta_2^2\|r_f\|_2^2}{\theta_2^2\|\tilde{x}\|_2^2 + 2\theta_1^2\|\tilde{y}\|_2^2} + \frac{\theta_2^2\|r_g\|_2^2}{\|\tilde{x}\|_2^2} + \frac{2\theta_4^2\|r_h\|_2^2}{\|\tilde{z}\|_2^2} - \frac{2\theta_1^2(r_f^T\tilde{x})(r_g^T\tilde{y})}{\|\tilde{x}\|_2^4} \\ &+ \frac{\theta_1^2(2\theta_1^2\|\tilde{y}\|_2^2 - \theta_2^2\|\tilde{x}\|_2^2)(r_f^T\tilde{x})^2}{\|\tilde{x}\|_2^4(\theta_2^2\|\tilde{x}\|_2^2 + 2\theta_1^2\|\tilde{y}\|_2^2)} + \frac{\theta_1^2(r_g^T\tilde{y})^2}{\|\tilde{x}\|_2^4} - \frac{\theta_4^2(r_h^T\tilde{z})^2}{\|\tilde{z}\|_2^4} \\ &- e^T E^{-1}e. \end{aligned} \quad (3.15)$$

多次利用 Sherman-Morrison-Woodbury 公式(见文献 [18]), 经过一些初等计算可得

$$\begin{aligned} E^{-1} &= \frac{1}{\theta_3^2} I_{ml} - \frac{\theta_2^2(I_m \otimes \tilde{z}\tilde{z}^T)}{\theta_3^2(\theta_3^2\|\tilde{x}\|_2^2 + \theta_2^2\|\tilde{z}\|_2^2)} - \frac{2\theta_4^2(\tilde{y}\tilde{y}^T \otimes I_l)}{\theta_3^2(\theta_3^2\|\tilde{z}\|_2^2 + 2\theta_4^2\|\tilde{y}\|_2^2)} \\ &+ \frac{\omega(\tilde{y}\tilde{y}^T \otimes \tilde{z}\tilde{z}^T)}{\theta_3^2(\theta_3^2\|\tilde{z}\|_2^2 + 2\theta_4^2\|\tilde{y}\|_2^2)} \end{aligned}$$

其中

$$\omega = \frac{\|\tilde{x}\|_2^2(2\theta_2^2\theta_4^2\|\tilde{x}\|_2^2\|\tilde{z}\|_2^2 - \theta_1^2\theta_3^2\|\tilde{z}\|_2^4 + \theta_3^2\theta_4^2\|\tilde{x}\|_2^4)}{\gamma_2\gamma_3\gamma_4} + \frac{2\theta_2^2\theta_4^2}{\theta_3^2\gamma_2\gamma_3}.$$

经过一些繁琐的计算, 可以得出

$$\begin{aligned} e^T E^{-1}e &= \frac{\theta_4^2\|\tilde{z}\|_2^2}{\|\tilde{x}\|_2^2\gamma_2}\|r_g\|_2^2 + \frac{4\theta_4^4\|\tilde{y}\|_2^2}{\|\tilde{z}\|_2^2\gamma_3}\|r_h\|_2^2 + \frac{\theta_1^4\|\tilde{y}\|_2^2\|\tilde{z}\|_2^4}{\|\tilde{x}\|_2^4\gamma_4}(r_f^T\tilde{x})^2 + \frac{\|\tilde{z}\|_2^2\gamma_{10}}{\|\tilde{x}\|_2^4\gamma_2\gamma_4}(r_g^T\tilde{y})^2 \\ &- \frac{\theta_4^4\|\tilde{y}\|_2^2\gamma_{11}}{\|\tilde{z}\|_2^4\gamma_3\gamma_4}(r_h^T\tilde{z})^2 - \frac{2\theta_1^2\|\tilde{z}\|_2^4\gamma_9}{\|\tilde{x}\|_2^4\gamma_4}(r_f^T\tilde{x})(r_g^T\tilde{y}) - \frac{2\theta_1^2\theta_4^2\|\tilde{y}\|_2^2}{\gamma_4}(r_f^T\tilde{x})(r_h^T\tilde{z}) \\ &+ \frac{2\theta_4^2\gamma_9}{\gamma_4}(r_g^T\tilde{y})(r_h^T\tilde{z}), \end{aligned}$$

其中

$$\gamma_{10} = \theta_1^2\theta_4^2\|\tilde{x}\|_2^2\|\tilde{z}\|_2^4 + \theta_1^4\theta_2^2\|\tilde{y}\|_2^2\|\tilde{z}\|_2^4 + \theta_1^4\theta_3^2\|\tilde{x}\|_2^2\|\tilde{y}\|_2^2\|\tilde{z}\|_2^2 + 2\theta_1^2\theta_2^2\theta_3^2\|\tilde{x}\|_2^4\|\tilde{z}\|_2^2 - \theta_2^4\theta_4^2\|\tilde{x}\|_2^6,$$

$$\gamma_{11} = 3\theta_3^2\|\tilde{x}\|_2^4\|\tilde{z}\|_2^2 + 2\theta_4^2\|\tilde{x}\|_2^4\|\tilde{y}\|_2^2 + 4\theta_1^2\|\tilde{y}\|_2^2\|\tilde{z}\|_2^4 + 4\theta_2^2\|\tilde{x}\|_2^2\|\tilde{z}\|_2^4.$$

其中  $\gamma_2, \gamma_3, \gamma_4$  和  $\gamma_9$  在定理 3.1 的前面部分定义过, 将上式带入到 (3.15) 中, 推出 (3.3). 证明完毕.

下面, 利用定理 3.1 中部分结构向后误差  $\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z})$  的表达式推出结构向后误差  $\eta^{(\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2, \lambda_3)}(\tilde{x}, \tilde{y}, \tilde{z})$  的具体表达式.

**定理 3.2** 假设  $(\tilde{x}^T, \tilde{y}^T, \tilde{z}^T)^T$  满足  $\tilde{x} \neq 0$  和  $\tilde{z} \neq 0$  为系统 (1.1) 的一个计算解. 且

$\eta^{(\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2, \lambda_3)}(\tilde{x}, \tilde{y}, \tilde{z})$  的定义由 (1.3) 和 (1.4) 给出. 则

$$\begin{aligned} & [\eta^{(\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2, \lambda_3)}(\tilde{x}, \tilde{y}, \tilde{z})]^2 \\ &= \frac{2\theta_1^2\theta_2^2(\mu_1 - 2\theta_1^2\theta_2^2)}{\gamma_1\mu_1} \|r_f\|_2^2 + \frac{\theta_2^2\theta_3^2(\mu_2 - \theta_2^2\theta_3^2)}{\gamma_2\mu_2} \|r_g\|_2^2 + \frac{2\theta_3^2\theta_4^2(\mu_3 - 2\theta_4^2)}{\gamma_3\mu_3} \|r_h\|_2^2 \\ &+ (r_f^T \tilde{x})^2 k_1 + (r_g^T \tilde{y})^2 k_2 + (r_h^T \tilde{z})^2 k_3 + (r_f^T \tilde{x})(r_g^T \tilde{y}) k_4 \\ &+ (r_f^T \tilde{x})(r_h^T \tilde{z}) k_5 + (r_g^T \tilde{y})(r_h^T \tilde{z}) k_6, \end{aligned} \quad (3.16)$$

其中

$$\begin{aligned} \mu_1 &= \gamma_1\lambda_1^2 + 2\theta_1^2\theta_2^2, \\ \mu_2 &= \gamma_2\lambda_2^2 + \theta_2^2\theta_3^2, \\ \mu_3 &= \gamma_3\lambda_3^2 + 2\theta_3^2\theta_4^2, \\ \mu_4 &= \theta_1^2\|\tilde{x}\|_2^2\gamma_5 + \gamma_1\gamma_4\lambda_1^2 + 2\theta_1^2\theta_2^2\gamma_4, \\ \mu_5 &= \theta_1^2\|\tilde{x}\|_2^2\gamma_5\gamma_6 + \gamma_1\gamma_4\gamma_6\lambda_1^2 + 2\theta_1^2\theta_2^2\gamma_4\gamma_6 - \theta_1^4\|\tilde{x}\|_2^2\gamma_1\gamma_2\gamma_8^2, \\ \mu_6 &= \theta_1^4\|\tilde{x}\|_2^2\|\tilde{y}\|_2^2\gamma_1\gamma_8 - \theta_1^2\|\tilde{x}\|_2^2\gamma_5\gamma_9 - \gamma_1\gamma_4\gamma_9\lambda_1^2 - 2\theta_1^2\theta_2^2\gamma_4\gamma_9, \\ \mu_7 &= 2\theta_1^4\theta_4^4\|\tilde{x}\|_2^2\|\tilde{y}\|_2^4\gamma_1\gamma_2\gamma_3\gamma_8\gamma_9 - \theta_1^2\theta_4^4\|\tilde{x}\|_2^2\|\tilde{y}\|_2^2\gamma_2\gamma_3\gamma_5\gamma_9^2 - \theta_4^4\|\tilde{y}\|_2^2\gamma_1\gamma_2\gamma_3\gamma_4\gamma_9^2\lambda_1^2, \\ &- 2\theta_1^2\theta_2^2\theta_4^4\|\tilde{y}\|_2^2\gamma_2\gamma_3\gamma_4\gamma_9^2 - \theta_1^4\theta_4^4\|\tilde{x}\|_2^2\|\tilde{y}\|_2^4\gamma_1\gamma_2\gamma_3\gamma_4\lambda_2^2 - \theta_1^4\theta_2^2\theta_3^2\theta_4^4\|\tilde{x}\|_2^2\|\tilde{y}\|_2^4\gamma_1\gamma_3\gamma_4, \\ &- \theta_1^4\theta_4^4\|\tilde{x}\|_2^2\|\tilde{y}\|_2^6\gamma_1\gamma_3\gamma_6, \end{aligned}$$

和

$$\begin{aligned} \xi_1 &= \gamma_4\mu_3^2(\gamma_4\mu_2\mu_4 + \|\tilde{y}\|_2^2\mu_5) + \|\tilde{z}\|_2^2\mu_3[\mu_7 + \theta_4^2\gamma_7(\gamma_4\mu_2\mu_4 + \|\tilde{y}\|_2^2\mu_5)], \\ \xi_2 &= \gamma_2\gamma_8\mu_6 + \gamma_4\mu_2\mu_4 + \|\tilde{y}\|_2^2\mu_5, \\ \xi_3 &= \theta_1^2\|\tilde{y}\|_2^2\gamma_1\gamma_2\gamma_4\gamma_8^2\mu_1 - \gamma_4\gamma_5\mu_2\mu_4 - \|\tilde{y}\|_2^2\gamma_5\mu_5, \end{aligned}$$

和

$$\begin{aligned} \Omega_1 &= \frac{\theta_1^2\gamma_1\mu_4\xi_1\xi_3 + \theta_1^4\theta_4^4\|\tilde{y}\|_2^4\|\tilde{z}\|_2^2\gamma_1^2\gamma_3\gamma_4\mu_1\mu_3\xi_2^2}{\mu_1\mu_4^2\xi_1(\gamma_4\mu_2\mu_4 + \|\tilde{y}\|_2^2\mu_5)}, \\ \Omega_4 &= \frac{\theta_4^4\|\tilde{z}\|_2^2\gamma_2^2\gamma_3\gamma_4\mu_2\mu_3\mu_6^2 - \gamma_2\mu_5\xi_1}{\mu_2\xi_1(\gamma_4\mu_2\mu_4 + \|\tilde{y}\|_2^2\mu_5)}, \\ \Omega_2 &= \frac{\theta_1^2\gamma_1\gamma_2\gamma_4\gamma_8\mu_4\xi_1 + \theta_1^2\theta_4^4\|\tilde{y}\|_2^2\|\tilde{z}\|_2^2\gamma_1\gamma_2\gamma_3\gamma_4\mu_3\mu_6\xi_2}{\mu_4\xi_1(\gamma_4\mu_2\mu_4 + \|\tilde{y}\|_2^2\mu_5)}, \\ \Omega_3 &= \frac{\theta_1^2\theta_4^2\|\tilde{y}\|_2^2\gamma_1\gamma_3\gamma_4\mu_3\xi_2}{\mu_4\xi_1}, \quad \Omega_5 = \frac{\theta_4^2\gamma_2\gamma_3\gamma_4\mu_3\mu_6}{\xi_1}, \\ \Omega_6 &= \frac{\gamma_3[\mu_7 + \theta_4^2\gamma_7(\gamma_4\mu_2\mu_4 + \|\tilde{y}\|_2^2\mu_5)]}{\xi_1}, \end{aligned}$$

和

$$\begin{aligned}
k_1 = & -\frac{\theta_1^4 \|\tilde{y}\|_2^2 \gamma_8 (2\theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \Omega_5 + \|\tilde{y}\|_2^2 \gamma_1 \gamma_8 \Omega_4 - 4\theta_2^2 \gamma_4 \Omega_2 - 2\|\tilde{x}\|_2^2 \gamma_5 \Omega_2)}{\gamma_1 \gamma_4^2} \\
& + \frac{\theta_1^4 \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 (4\theta_2^2 \gamma_4 \Omega_3 + 2\|\tilde{x}\|_2^2 \gamma_5 \Omega_3 + \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \Omega_6)}{\gamma_1 \gamma_4^2} \\
& + \frac{\theta_1^2 (\gamma_1 \gamma_4 \gamma_5 - 4\theta_1^2 \theta_2^4 \gamma_4^2 \Omega_1 - 4\theta_1^2 \theta_2^2 \|\tilde{x}\|_2^2 \gamma_4 \gamma_5 \Omega_1 - \theta_1^2 \|\tilde{x}\|_2^4 \gamma_5^2 \Omega_1)}{\gamma_1^2 \gamma_4^2} \\
& - \frac{\theta_1^4 (4\theta_2^2 \gamma_4 \gamma_5 \mu_2 \mu_3 + \|\tilde{x}\|_2^2 \gamma_5^2 \mu_2 \mu_3 + \|\tilde{y}\|_2^2 \gamma_1 \gamma_2 \gamma_8^2 \mu_1 \mu_3 + \theta_4^4 \|\tilde{y}\|_2^4 \|\tilde{z}\|_2^2 \gamma_1 \gamma_3 \mu_1 \mu_2)}{\gamma_1 \gamma_4^2 \mu_1 \mu_2 \mu_3}, \\
k_2 = & -\frac{\theta_1^2 \|\tilde{x}\|_2^2 \gamma_8 (\theta_1^2 \|\tilde{x}\|_2^2 \gamma_2 \gamma_8 \Omega_1 - 2\theta_4^2 \|\tilde{z}\|_2^2 \gamma_2 \gamma_9 \Omega_3 - 2\theta_2^2 \theta_3^2 \gamma_4 \Omega_2 - 2\|\tilde{y}\|_2^2 \gamma_6 \Omega_2)}{\gamma_2 \gamma_4^2} \\
& - \frac{\theta_4^2 \|\tilde{z}\|_2^2 \gamma_9 (2\theta_2^2 \theta_3^2 \gamma_4 \Omega_5 + 2\|\tilde{y}\|_2^2 \gamma_6 \Omega_5 - \theta_4^2 \|\tilde{z}\|_2^2 \gamma_2 \gamma_9 \Omega_6)}{\gamma_2 \gamma_4^2} \\
& + \frac{\gamma_6 (\gamma_2 \gamma_4 \mu_2 - 2\theta_2^2 \theta_3^2 \|\tilde{y}\|_2^2 \gamma_4 \mu_2 \Omega_4 - \|\tilde{y}\|_2^2 \gamma_2 \gamma_6 - \|\tilde{y}\|_2^4 \gamma_6 \mu_2 \Omega_4)}{\gamma_2^2 \gamma_4^2 \mu_2} \\
& - \frac{2\theta_2^2 \theta_3^2 \gamma_2 \gamma_4 \gamma_6 \mu_1 \mu_3 + \theta_1^4 \|\tilde{x}\|_2^2 \gamma_1 \gamma_2^2 \gamma_8^2 \mu_2 \mu_3 + \theta_2^4 \theta_3^4 \gamma_4^2 \mu_1 \mu_2 \mu_3 \Omega_4 + \theta_4^4 \|\tilde{z}\|_2^2 \gamma_2^2 \gamma_3 \gamma_9^2 \mu_1 \mu_2}{\gamma_2^2 \gamma_4^2 \mu_1 \mu_2 \mu_3}, \\
k_3 = & -\frac{\theta_4^4 \|\tilde{y}\|_2^2 \gamma_9 (\|\tilde{y}\|_2^2 \gamma_3 \gamma_9 \Omega_4 - 2\theta_1^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \gamma_3 \Omega_2 + 4\theta_3^2 \gamma_4 \Omega_5 + 2\|\tilde{z}\|_2^2 \gamma_7 \Omega_5)}{\gamma_3 \gamma_4^2} \\
& - \frac{\theta_1^2 \theta_4^4 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 (\theta_1^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \gamma_3 \Omega_1 - 4\theta_3^2 \gamma_4 \Omega_3 - 2\|\tilde{z}\|_2^2 \gamma_7 \Omega_3)}{\gamma_3 \gamma_4^2} \\
& + \frac{\theta_4^2 \gamma_7 (\gamma_3 \gamma_4 \mu_3 - \theta_4^2 \|\tilde{z}\|_2^2 \gamma_3 \gamma_7 + 4\theta_3^2 \theta_4^2 \|\tilde{z}\|_2^2 \gamma_4 \mu_3 \Omega_6 + \theta_4^2 \|\tilde{z}\|_2^4 \gamma_7 \mu_3 \Omega_6)}{\gamma_3^2 \gamma_4^2 \mu_3} \\
& - \frac{\theta_4^4 (4\theta_3^2 \gamma_3 \gamma_4 \gamma_7 \mu_1 \mu_2 + \theta_1^4 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^4 \gamma_1 \gamma_3^2 \mu_2 \mu_3 + \|\tilde{y}\|_2^2 \gamma_2 \gamma_3^2 \gamma_9^2 \mu_1 \mu_3 - 4\theta_3^4 \gamma_4^2 \mu_1 \mu_2 \mu_3 \Omega_6)}{\gamma_3^2 \gamma_4^2 \mu_1 \mu_2 \mu_3}, \\
k_4 = & -\frac{2\theta_1^4 \|\tilde{x}\|_2^2 \gamma_8 (\|\tilde{y}\|_2^2 \gamma_1 \gamma_8 \Omega_2 + \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \Omega_3 - 2\theta_2^2 \gamma_4 \Omega_1 - \|\tilde{x}\|_2^2 \gamma_5 \Omega_1)}{\gamma_1 \gamma_4^2} \\
& - \frac{2\theta_1^2 \theta_4^2 \|\tilde{z}\|_2^2 \gamma_9 (2\theta_2^2 \gamma_4 \Omega_3 + \|\tilde{x}\|_2^2 \gamma_5 \Omega_3 - \|\tilde{y}\|_2^2 \gamma_1 \gamma_8 \Omega_5 + \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \Omega_6)}{\gamma_1 \gamma_4^2} \\
& + \frac{2\theta_1^2 \gamma_8 (2\theta_1^2 \theta_2^2 \gamma_4 \mu_2 + \theta_2^2 \theta_3^2 \gamma_4 \mu_1 + \theta_1^2 \|\tilde{x}\|_2^2 \gamma_5 \mu_2 + \|\tilde{y}\|_2^2 \gamma_6 \mu_1)}{\gamma_4^2 \mu_1 \mu_2} \\
& + \frac{2\theta_1^2 \|\tilde{y}\|_2^2 (\theta_2^2 \theta_3^2 \theta_4^2 \|\tilde{z}\|_2^2 \gamma_4 \Omega_5 + \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_6 \Omega_5 + \theta_2^2 \theta_3^2 \gamma_4 \gamma_8 \Omega_4 + \|\tilde{y}\|_2^2 \gamma_6 \gamma_8 \Omega_4)}{\gamma_2 \gamma_4^2} \\
& - \frac{2\theta_1^2 \theta_2^2 \Omega_2 (2\theta_2^2 \theta_3^2 \gamma_4 + \theta_3^2 \|\tilde{x}\|_2^2 \gamma_5 + 2\|\tilde{y}\|_2^2 \gamma_6)}{\gamma_1 \gamma_2 \gamma_4} \\
& + \frac{2\theta_1^2 (\theta_4^4 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \gamma_2 \gamma_3 \gamma_9 - \gamma_1 \gamma_2 \gamma_4 \gamma_8 \mu_3 - \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \gamma_5 \gamma_6 \mu_3 \Omega_2)}{\gamma_1 \gamma_2 \gamma_4^2 \mu_3},
\end{aligned}$$

$$\begin{aligned}
k_5 = & - \frac{2\theta_1^2\theta_4^2\|\tilde{y}\|_2^2(\theta_4^2\|\tilde{y}\|_2^2\|\tilde{z}\|_2^2\gamma_9\Omega_5 - \theta_1^2\|\tilde{x}\|_2^2\|\tilde{y}\|_2^2\gamma_8\Omega_2 - \theta_1^2\theta_4^2\|\tilde{x}\|_2^2\|\tilde{y}\|_2^2\|\tilde{z}\|_2^2\Omega_3 + \|\tilde{y}\|_2^2\gamma_8\gamma_9\Omega_4)}{\gamma_4^2} \\
& - \frac{4\theta_1^2\theta_4^2\|\tilde{y}\|_2^2(\theta_1^2\theta_2^2\gamma_1\mu_3 + \theta_3^2\theta_4^2\gamma_1\mu_1 - \theta_2^2\gamma_9\mu_1\mu_3\Omega_2 + \theta_1^2\theta_2^2\|\tilde{x}\|_2^2\mu_1\mu_3\Omega_1)}{\gamma_1\gamma_4\mu_1\mu_3} \\
& - \frac{4\theta_1^2\theta_4^2(\theta_3^2\|\tilde{y}\|_2^2\gamma_1\gamma_8\Omega_5 - \theta_2^2\|\tilde{z}\|_2^2\gamma_7\Omega_3 - \theta_3^2\|\tilde{x}\|_2^2\gamma_5\Omega_3 - \theta_3^2\theta_4^2\|\tilde{y}\|_2^2\|\tilde{z}\|_2^2\gamma_1\Omega_6)}{\gamma_1\gamma_3\gamma_4} \\
& - \frac{2\theta_1^2\theta_4^2(\|\tilde{y}\|_2^2\|\tilde{z}\|_2^2\gamma_1\gamma_7\gamma_8\Omega_5 - 4\theta_2^2\theta_3^2\gamma_4^2\Omega_3 - \|\tilde{x}\|_2^2\|\tilde{z}\|_2^2\gamma_5\gamma_7\Omega_3 - \theta_4^2\|\tilde{y}\|_2^2\|\tilde{z}\|_2^4\gamma_1\gamma_7\Omega_6)}{\gamma_1\gamma_3\gamma_4^2} \\
& - \frac{2\theta_1^2\theta_4^2\|\tilde{x}\|_2^2\|\tilde{y}\|_2^2\gamma_5(\theta_1^2\gamma_1 - \gamma_9\mu_1\Omega_2 + \theta_1^2\|\tilde{x}\|_2^2\mu_1\Omega_1)}{\gamma_1\gamma_4^2\mu_1} \\
& + \frac{2\theta_1^2\theta_4^2\|\tilde{y}\|_2^2(\gamma_4\mu_2\mu_3 - \gamma_2\gamma_8\gamma_9\mu_3 - \theta_4^2\|\tilde{z}\|_2^2\gamma_7\mu_2)}{\gamma_4^2\mu_2\mu_3}, \\
k_6 = & - \frac{2\theta_4^4\|\tilde{z}\|_2^2\gamma_9(\theta_1^2\|\tilde{x}\|_2^2\|\tilde{y}\|_2^2\gamma_3\Omega_3 - \|\tilde{y}\|_2^2\gamma_3\gamma_9\Omega_5 + 2\theta_3^2\gamma_4\Omega_6 + \|\tilde{z}\|_2^2\gamma_7\Omega_6)}{\gamma_3\gamma_4^2} \\
& - \frac{2\theta_3^2\theta_4^2(\theta_1^2\theta_2^2\|\tilde{x}\|_2^2\|\tilde{y}\|_2^2\mu_2\mu_3\Omega_2 - \theta_2^2\gamma_2\gamma_9\mu_3 - \theta_2^2\|\tilde{y}\|_2^2\gamma_9\mu_2\mu_3\Omega_4 - 2\theta_4^2\gamma_2\gamma_9\mu_2)}{\gamma_2\gamma_4\mu_2\mu_3} \\
& - \frac{2\theta_1^2\theta_4^2\|\tilde{x}\|_2^2\|\tilde{y}\|_2^2\gamma_8(\gamma_9\mu_1\Omega_2 - \theta_1^2\gamma_1 - \theta_1^2\|\tilde{x}\|_2^2\mu_1\Omega_1)}{\gamma_4^2\mu_1} \\
& - \frac{2\theta_4^2(\gamma_3\gamma_4\gamma_9 + 2\theta_1^2\theta_3^2\|\tilde{x}\|_2^2\gamma_4\gamma_8\Omega_3 + \theta_1^2\|\tilde{x}\|_2^2\|\tilde{z}\|_2^2\gamma_7\gamma_8\Omega_3)}{\gamma_3\gamma_4^2} \\
& - \frac{2\theta_4^2\|\tilde{y}\|_2^2\gamma_6(\theta_1^2\|\tilde{x}\|_2^2\|\tilde{y}\|_2^2\gamma_3\Omega_2 - \|\tilde{y}\|_2^2\gamma_3\gamma_9\Omega_4 - 2\theta_3^2\gamma_4\Omega_5 - \|\tilde{z}\|_2^2\gamma_7\Omega_5)}{\gamma_2\gamma_3\gamma_4^2} \\
& + \frac{2\theta_4^2(\|\tilde{y}\|_2^2\gamma_2\gamma_3\gamma_6\gamma_9\mu_3 + 2\theta_2^2\theta_3^4\gamma_4^2\mu_2\mu_3\Omega_5 + \theta_2^2\theta_3^2\|\tilde{z}\|_2^2\gamma_4\gamma_7\mu_2\mu_3\Omega_5 + \theta_4^2\|\tilde{z}\|_2^2\gamma_2\gamma_3\gamma_7\gamma_9\mu_2)}{\gamma_2\gamma_3\gamma_4^2\mu_2\mu_3}.
\end{aligned}$$

**证明** 根据  $\eta^{(\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2, \lambda_3)}(\tilde{x}, \tilde{y}, \tilde{z})$  的定义 (1.4) 和部分结构向后误差  $\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z})$  的表达式 (3.3), 可以推出

$$\left[\eta^{(\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2, \lambda_3)}(\tilde{x}, \tilde{y}, \tilde{z})\right]^2 = \min_{\Delta f \in \mathbb{R}^n, \Delta g \in \mathbb{R}^m, \Delta h \in \mathbb{R}^l} \mathcal{X}(\Delta f, \Delta g, \Delta h),$$

其中

$$\begin{aligned}
& \mathcal{X}(\Delta f, \Delta g, \Delta h) \\
= & \lambda_1^2 \|\Delta f\|_2^2 + \lambda_2^2 \|\Delta g\|_2^2 + \lambda_3^2 \|\Delta h\|_2^2 + \frac{2\theta_1^2\theta_2^2}{\gamma_1} \|r_f + \Delta f\|_2^2 + \frac{\theta_2^2\theta_3^2}{\gamma_2} \|r_g + \Delta g\|_2^2 \\
& + \frac{2\theta_3^2\theta_4^2}{\gamma_3} \|r_h + \Delta h\|_2^2 + \frac{\theta_1^2\gamma_5}{\gamma_1\gamma_4} \left[(r_f + \Delta f)^T \tilde{x}\right]^2 + \frac{\gamma_6}{\gamma_2\gamma_4} \left[(r_g + \Delta g)^T \tilde{y}\right]^2 \\
& + \frac{\theta_4^2\gamma_7}{\gamma_3\gamma_4} \left[(r_h + \Delta h)^T \tilde{z}\right]^2 - \frac{2\theta_1^2\gamma_8}{\gamma_4} \left[(r_f + \Delta f)^T \tilde{x}\right] \left[(r_g + \Delta g)^T \tilde{y}\right] \\
& + \frac{2\theta_1^2\theta_4^2\|\tilde{y}\|_2^2}{\gamma_4} \left[(r_f + \Delta f)^T \tilde{x}\right] \left[(r_h + \Delta h)^T \tilde{z}\right] - \frac{2\theta_4^2\gamma_9}{\gamma_4} \left[(r_g + \Delta g)^T \tilde{y}\right] \left[(r_h + \Delta h)^T \tilde{z}\right].
\end{aligned}$$

经过一系列初级运算, 可以推出

$$\mathcal{X}(\Delta f, \Delta g, \Delta h) = [\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z})]^2 + \begin{bmatrix} \Delta f \\ \Delta g \\ \Delta h \end{bmatrix}^T \Phi \begin{bmatrix} \Delta f \\ \Delta g \\ \Delta h \end{bmatrix} + 2 \begin{bmatrix} \Delta f \\ \Delta g \\ \Delta h \end{bmatrix}^T q,$$

且

$$\Phi = \begin{bmatrix} \frac{\gamma_1 \lambda_1^2 + 2\theta_1^2 \theta_2^2}{\gamma_1} I_n + \frac{\theta_1^2 \gamma_5 \tilde{x} \tilde{x}^T}{\gamma_1 \gamma_4} & -\frac{\theta_1^2 \gamma_8 \tilde{x} \tilde{y}^T}{\gamma_4} & \frac{\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^2 \tilde{x} \tilde{z}^T}{\gamma_4} \\ -\frac{\theta_1^2 \gamma_8 \tilde{y} \tilde{x}^T}{\gamma_4} & \frac{\gamma_2 \lambda_2^2 + \theta_2^2 \theta_3^2}{\gamma_2} I_m + \frac{\gamma_6 \tilde{y} \tilde{y}^T}{\gamma_2 \gamma_4} & -\frac{\theta_4^2 \gamma_9 \tilde{y} \tilde{z}^T}{\gamma_4} \\ \frac{\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^2 \tilde{z} \tilde{x}^T}{\gamma_4} & -\frac{\theta_4^2 \gamma_9 \tilde{z} \tilde{y}^T}{\gamma_4} & \frac{\gamma_3 \lambda_3^2 + 2\theta_3^2 \theta_4^2}{\gamma_3} I_l + \frac{\theta_4^2 \gamma_7 \tilde{z} \tilde{z}^T}{\gamma_3 \gamma_4} \end{bmatrix},$$

和

$$q = \begin{pmatrix} \frac{2\theta_1^2 \theta_2^2}{\gamma_1} r_f + \frac{\theta_1^2 \gamma_5}{\gamma_1 \gamma_4} (r_f^T \tilde{x}) \tilde{x} - \frac{\theta_1^2 \gamma_8}{\gamma_4} (r_g^T \tilde{y}) \tilde{x} + \frac{\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^2}{\gamma_4} (r_h^T \tilde{z}) \tilde{x} \\ \frac{\theta_2^2 \theta_3^2}{\gamma_2} r_g + \frac{\gamma_6}{\gamma_2 \gamma_4} (r_g^T \tilde{y}) \tilde{y} - \frac{\theta_1^2 \gamma_8}{\gamma_4} (r_f^T \tilde{x}) \tilde{y} - \frac{\theta_4^2 \gamma_9}{\gamma_4} (r_h^T \tilde{z}) \tilde{y} \\ \frac{2\theta_3^2 \theta_4^2}{\gamma_3} r_h + \frac{\theta_4^2 \gamma_7}{\gamma_3 \gamma_4} (r_h^T \tilde{z}) \tilde{z} + \frac{\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^2}{\gamma_4} (r_f^T \tilde{x}) \tilde{z} - \frac{\theta_4^2 \gamma_9}{\gamma_4} (r_g^T \tilde{y}) \tilde{z} \end{pmatrix}.$$

通过繁琐的计算, 对于任意非零向量  $t = (a^T, b^T, c^T)^T$ , 其中  $a \in \mathbb{R}^n$ ,  $b \in \mathbb{R}^m$ , 和  $c \in \mathbb{R}^l$ , 可得

$$\begin{aligned} t^T \Phi t &= \frac{\gamma_1 \lambda_1^2 + 2\theta_1^2 \theta_2^2}{\gamma_1} a^T a + \frac{\theta_1^2 \gamma_5}{\gamma_1 \gamma_4} a^T \tilde{x} \tilde{x}^T a - \frac{2\theta_1^2 \gamma_8}{\gamma_4} a^T \tilde{x} \tilde{y}^T b + \frac{2\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^2}{\gamma_4} a^T \tilde{x} \tilde{z}^T c + \frac{\theta_4^2 \gamma_7}{\gamma_3 \gamma_4} c^T \tilde{z} \tilde{z}^T c \\ &+ \frac{\gamma_2 \lambda_2^2 + \theta_2^2 \theta_3^2}{\gamma_2} b^T b + \frac{\gamma_6}{\gamma_2 \gamma_4} b^T \tilde{y} \tilde{y}^T b - \frac{2\theta_4^2 \gamma_9}{\gamma_4} b^T \tilde{y} \tilde{z}^T c + \frac{\gamma_3 \lambda_3^2 + 2\theta_3^2 \theta_4^2}{\gamma_3} c^T c \\ &= \frac{\gamma_4 \lambda_1^2 + \theta_1^2 \theta_2^2 \|\tilde{z}\|_2^4}{\gamma_4} a^T a + \frac{\gamma_4 \lambda_2^2 + \theta_2^2 \theta_3^2 \|\tilde{x}\|_2^2 \|\tilde{z}\|_2^2}{\gamma_4} b^T b + \frac{\gamma_4 \lambda_3^2 + \theta_3^2 \theta_4^2 \|\tilde{x}\|_2^4}{\gamma_4} c^T c \\ &+ \frac{\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^2}{\gamma_4} (a^T \tilde{x} - b^T \tilde{y} + c^T \tilde{z})^2 + \frac{\theta_1^2 \theta_3^2 \|\tilde{z}\|_2^2}{\gamma_4} (a^T \tilde{x} - b^T \tilde{y})^2 + \frac{\theta_2^2 \theta_4^2 \|\tilde{x}\|_2^2}{\gamma_4} (b^T \tilde{y} - c^T \tilde{z})^2 \\ &+ \frac{\theta_1^2 (\gamma_1 \gamma_8 - \gamma_5)}{\gamma_1 \gamma_4} a^T (\|\tilde{x}\|_2^2 I_n - \tilde{x} \tilde{x}^T) a + \frac{\theta_2^2 \theta_3^2 (\theta_4^2 \|\tilde{x}\|_2^4 + \theta_1^2 \|\tilde{z}\|_2^4)}{\gamma_2 \gamma_4} b^T (\|\tilde{y}\|_2^2 I_m - \tilde{y} \tilde{y}^T) b \\ &+ \frac{\theta_4^2 (\gamma_3 \gamma_9 - \gamma_7)}{\gamma_3 \gamma_4} c^T (\|\tilde{z}\|_2^2 I_l - \tilde{z} \tilde{z}^T) c \\ &> 0, \end{aligned}$$

这表明  $\Phi$  是一个对称正定矩阵. 因此,  $\mathcal{X}(\Delta f, \Delta g, \Delta h)$  的最小值点为

$$(\Delta f^T, \Delta g^T, \Delta h^T)^T = -\Phi^{-1} q.$$

相应的,

$$[\eta^{(\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2, \lambda_3)}(\tilde{x}, \tilde{y}, \tilde{z})]^2 = [\eta^{(\theta_1, \theta_2, \theta_3, \theta_4)}(\tilde{x}, \tilde{y}, \tilde{z})]^2 - q^T \Phi^{-1} q. \quad (3.17)$$

令

$$\Phi^{-1} = \begin{bmatrix} \Psi_{11} & \Psi_{12} & \Psi_{13} \\ \Psi_{12}^T & \Psi_{22} & \Psi_{23} \\ \Psi_{13}^T & \Psi_{23}^T & \Psi_{33} \end{bmatrix}, \Psi_{11} \in \mathbb{R}^{n \times n}, \Psi_{22} \in \mathbb{R}^{m \times m}, \Psi_{33} \in \mathbb{R}^{l \times l}.$$

经过一系列初等行变换, 有

$$\begin{aligned} \Psi_{11} &= \frac{\gamma_1}{\mu_1} I_n + \frac{\theta_1^2 \gamma_1 \mu_4 \xi_1 \xi_3 + \theta_1^4 \theta_4^4 \|\tilde{y}\|_2^4 \|\tilde{z}\|_2^2 \gamma_1^2 \gamma_3 \gamma_4 \mu_1 \mu_3 \xi_2^2}{\mu_1 \mu_4^2 \xi_1 (\gamma_4 \mu_2 \mu_4 + \|\tilde{y}\|_2^2 \mu_5)} \tilde{x} \tilde{x}^T, \\ \Psi_{12} &= \frac{\theta_1^2 \gamma_1 \gamma_2 \gamma_4 \gamma_8 \mu_4 \xi_1 + \theta_1^2 \theta_4^4 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \gamma_2 \gamma_3 \gamma_4 \mu_3 \mu_6 \xi_2}{\mu_4 \xi_1 (\gamma_4 \mu_2 \mu_4 + \|\tilde{y}\|_2^2 \mu_5)} \tilde{x} \tilde{y}^T, \\ \Psi_{13} &= -\frac{\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^2 \gamma_1 \gamma_3 \gamma_4 \mu_3 \xi_2}{\mu_4 \xi_1} \tilde{x} \tilde{z}^T, \Psi_{22} = \frac{\gamma_2}{\mu_2} I_m + \frac{\theta_4^4 \|\tilde{z}\|_2^2 \gamma_2^2 \gamma_3 \gamma_4 \mu_2 \mu_3 \mu_6^2 - \gamma_2 \mu_5 \xi_1}{\mu_2 \xi_1 (\gamma_4 \mu_2 \mu_4 + \|\tilde{y}\|_2^2 \mu_5)} \tilde{y} \tilde{y}^T, \\ \Psi_{23} &= -\frac{\theta_4^2 \gamma_2 \gamma_3 \gamma_4 \mu_3 \mu_6}{\xi_1} \tilde{y} \tilde{z}^T, \Psi_{33} = \frac{\gamma_3}{\mu_3} I_l - \frac{\gamma_3 [\mu_7 + \theta_4^2 \gamma_7 (\gamma_4 \mu_2 \mu_4 + \|\tilde{y}\|_2^2 \mu_5)]}{\xi_1} \tilde{z} \tilde{z}^T. \end{aligned}$$

经过一系列繁琐的初等代数计算, 可以得到

$$\begin{aligned} q^T \Phi^{-1} q &= \frac{4\theta_1^4 \theta_2^4 \|r_f\|_2^2}{\gamma_1 \mu_1} + \frac{\theta_2^4 \theta_3^4 \|r_g\|_2^2}{\gamma_2 \mu_2} + \frac{4\theta_3^2 \theta_4^4 \|r_h\|_2^2}{\gamma_3 \mu_3} + (r_f^T \tilde{x})^2 l_1 + (r_g^T \tilde{y})^2 l_2 \\ &\quad + (r_h^T \tilde{z})^2 l_3 + (r_f^T \tilde{x}) (r_g^T \tilde{y}) l_4 + (r_f^T \tilde{x}) (r_h^T \tilde{z}) l_5 + (r_g^T \tilde{y}) (r_h^T \tilde{z}) l_6, \end{aligned} \quad (3.18)$$

这里

$$\begin{aligned} l_1 &= \frac{\theta_1^4 \|\tilde{y}\|_2^2 \gamma_8 (2\theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \Omega_5 + \|\tilde{y}\|_2^2 \gamma_1 \gamma_8 \Omega_4 - 4\theta_2^2 \gamma_4 \Omega_2 - 2\|\tilde{x}\|_2^2 \gamma_5 \Omega_2)}{\gamma_1 \gamma_4^2} \\ &\quad - \frac{\theta_1^4 \theta_2^4 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 (4\theta_2^2 \gamma_4 \Omega_3 + 2\|\tilde{x}\|_2^2 \gamma_5 \Omega_3 + \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \Omega_6)}{\gamma_1 \gamma_4^2} \\ &\quad + \frac{\theta_1^4 \Omega_1 (4\theta_2^4 \gamma_4^2 + 4\theta_2^2 \|\tilde{x}\|_2^2 \gamma_4 \gamma_5 + \|\tilde{x}\|_2^4 \gamma_5^2)}{\gamma_1^2 \gamma_4^2} \\ &\quad + \frac{\theta_1^4 (4\theta_2^2 \gamma_4 \gamma_5 \mu_2 \mu_3 + \|\tilde{x}\|_2^2 \gamma_5^2 \mu_2 \mu_3 + \|\tilde{y}\|_2^2 \gamma_1 \gamma_2 \gamma_8^2 \mu_1 \mu_3 + \theta_4^4 \|\tilde{y}\|_2^4 \|\tilde{z}\|_2^2 \gamma_1 \gamma_3 \mu_1 \mu_2)}{\gamma_1 \gamma_4^2 \mu_1 \mu_2 \mu_3}, \\ l_2 &= \frac{\theta_1^2 \|\tilde{x}\|_2^2 \gamma_8 (\theta_1^2 \|\tilde{x}\|_2^2 \gamma_2 \gamma_8 \Omega_1 - 2\theta_4^2 \|\tilde{z}\|_2^2 \gamma_2 \gamma_9 \Omega_3 - 2\theta_2^2 \theta_3^2 \gamma_4 \Omega_2 - 2\|\tilde{y}\|_2^2 \gamma_6 \Omega_2)}{\gamma_2 \gamma_4^2} \\ &\quad + \frac{\theta_2^4 \|\tilde{z}\|_2^2 \gamma_9 (2\theta_2^2 \theta_3^2 \gamma_4 \Omega_5 + 2\|\tilde{y}\|_2^2 \gamma_6 \Omega_5 - \theta_4^2 \|\tilde{z}\|_2^2 \gamma_2 \gamma_9 \Omega_6)}{\gamma_2 \gamma_4^2} \\ &\quad + \frac{\|\tilde{y}\|_2^2 \gamma_6 (2\theta_2^2 \theta_3^2 \gamma_4 \mu_2 \Omega_4 + \gamma_2 \gamma_6 + \|\tilde{y}\|_2^2 \gamma_6 \mu_2 \Omega_4)}{\gamma_2^2 \gamma_4^2 \mu_2} \\ &\quad + \frac{2\theta_2^2 \theta_3^2 \gamma_2 \gamma_4 \gamma_6 \mu_1 \mu_3 + \theta_1^4 \|\tilde{x}\|_2^2 \gamma_1 \gamma_2^2 \gamma_8^2 \mu_2 \mu_3 + \theta_2^4 \theta_3^4 \gamma_4^2 \mu_1 \mu_2 \mu_3 \Omega_4 + \theta_4^4 \|\tilde{z}\|_2^2 \gamma_2^2 \gamma_3 \gamma_9^2 \mu_1 \mu_2}{\gamma_2^2 \gamma_4^2 \mu_1 \mu_2 \mu_3}, \end{aligned}$$

$$\begin{aligned}
l_3 = & \frac{\theta_4^4 \|\tilde{y}\|_2^2 \gamma_9 (\|\tilde{y}\|_2^2 \gamma_3 \gamma_9 \Omega_4 - 2\theta_1^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \gamma_3 \Omega_2 + 4\theta_3^2 \gamma_4 \Omega_5 + 2\|\tilde{z}\|_2^2 \gamma_7 \Omega_5)}{\gamma_3 \gamma_4^2} \\
& + \frac{\theta_1^2 \theta_4^4 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 (\theta_1^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \gamma_3 \Omega_1 - 4\theta_3^2 \gamma_4 \Omega_3 - 2\|\tilde{z}\|_2^2 \gamma_7 \Omega_3)}{\gamma_3 \gamma_4^2} \\
& + \frac{\theta_4^4 \|\tilde{z}\|_2^2 \gamma_7 (\gamma_3 \gamma_7 - 4\theta_3^2 \gamma_4 \mu_3 \Omega_6 - \|\tilde{z}\|_2^2 \gamma_7 \mu_3 \Omega_6)}{\gamma_3^2 \gamma_4^2 \mu_3} \\
& + \frac{\theta_4^4 (4\theta_3^2 \gamma_3 \gamma_4 \gamma_7 \mu_1 \mu_2 + \theta_1^4 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^4 \gamma_1 \gamma_3^2 \mu_2 \mu_3 + \|\tilde{y}\|_2^2 \gamma_2 \gamma_3^2 \gamma_9^2 \mu_1 \mu_3 - 4\theta_3^4 \gamma_4^2 \mu_1 \mu_2 \mu_3 \Omega_6)}{\gamma_3^2 \gamma_4^2 \mu_1 \mu_2 \mu_3}, \\
l_4 = & \frac{2\theta_1^4 \|\tilde{x}\|_2^2 \gamma_8 (\|\tilde{y}\|_2^2 \gamma_1 \gamma_8 \Omega_2 + \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \Omega_3 - 2\theta_2^2 \gamma_4 \Omega_1 - \|\tilde{x}\|_2^2 \gamma_5 \Omega_1)}{\gamma_1 \gamma_4^2} \\
& + \frac{2\theta_1^2 \theta_4^2 \|\tilde{z}\|_2^2 \gamma_9 (2\theta_2^2 \gamma_4 \Omega_3 + \|\tilde{x}\|_2^2 \gamma_5 \Omega_3 - \|\tilde{y}\|_2^2 \gamma_1 \gamma_8 \Omega_5 + \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \Omega_6)}{\gamma_1 \gamma_4^2} \\
& - \frac{2\theta_1^2 \gamma_8 (2\theta_1^2 \theta_2^2 \gamma_4 \mu_2 + \theta_2^2 \theta_3^2 \gamma_4 \mu_1 + \theta_1^2 \|\tilde{x}\|_2^2 \gamma_5 \mu_2 + \|\tilde{y}\|_2^2 \gamma_6 \mu_1)}{\gamma_4^2 \mu_1 \mu_2} \\
& - \frac{2\theta_1^2 \|\tilde{y}\|_2^2 (\theta_2^2 \theta_3^2 \theta_4^2 \|\tilde{z}\|_2^2 \gamma_4 \Omega_5 + \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_6 \Omega_5 + \theta_2^2 \theta_3^2 \gamma_4 \gamma_8 \Omega_4 + \|\tilde{y}\|_2^2 \gamma_6 \gamma_8 \Omega_4)}{\gamma_2 \gamma_4^2} \\
& + \frac{2\theta_1^2 \theta_2^2 \Omega_2 (2\theta_2^2 \theta_3^2 \gamma_4 + \theta_3^2 \|\tilde{x}\|_2^2 \gamma_5 + 2\|\tilde{y}\|_2^2 \gamma_6)}{\gamma_1 \gamma_2 \gamma_4} \\
& + \frac{2\theta_1^2 \|\tilde{y}\|_2^2 (\|\tilde{x}\|_2^2 \gamma_5 \gamma_6 \mu_3 \Omega_2 - \theta_4^4 \|\tilde{z}\|_2^2 \gamma_1 \gamma_2 \gamma_3 \gamma_9)}{\gamma_1 \gamma_2 \gamma_4^2 \mu_3}, \\
l_5 = & \frac{2\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^2 (\theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_9 \Omega_5 - \theta_1^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \gamma_8 \Omega_2 - \theta_1^2 \theta_4^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \Omega_3 + \|\tilde{y}\|_2^2 \gamma_8 \gamma_9 \Omega_4)}{\gamma_4^2} \\
& + \frac{4\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^2 (\theta_1^2 \theta_2^2 \gamma_1 \mu_3 + \theta_3^2 \theta_4^2 \gamma_1 \mu_1 - \theta_2^2 \gamma_9 \mu_1 \mu_3 \Omega_2 + \theta_1^2 \theta_2^2 \|\tilde{x}\|_2^2 \mu_1 \mu_3 \Omega_1)}{\gamma_1 \gamma_4 \mu_1 \mu_3} \\
& + \frac{4\theta_1^2 \theta_4^2 (\theta_3^2 \|\tilde{y}\|_2^2 \gamma_1 \gamma_8 \Omega_5 - \theta_2^2 \|\tilde{z}\|_2^2 \gamma_7 \Omega_3 - \theta_3^2 \|\tilde{x}\|_2^2 \gamma_5 \Omega_3 - \theta_3^2 \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \Omega_6)}{\gamma_1 \gamma_3 \gamma_4} \\
& + \frac{2\theta_1^2 \theta_4^2 (\|\tilde{y}\|_2^2 \|\tilde{z}\|_2^2 \gamma_1 \gamma_7 \gamma_8 \Omega_5 - 4\theta_2^2 \theta_3^2 \gamma_4^2 \Omega_3 - \|\tilde{x}\|_2^2 \|\tilde{z}\|_2^2 \gamma_5 \gamma_7 \Omega_3 - \theta_4^2 \|\tilde{y}\|_2^2 \|\tilde{z}\|_2^4 \gamma_1 \gamma_7 \Omega_6)}{\gamma_1 \gamma_3 \gamma_4^2} \\
& + \frac{2\theta_1^2 \theta_4^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \gamma_5 (\theta_1^2 \gamma_1 - \gamma_9 \mu_1 \Omega_2 + \theta_1^2 \|\tilde{x}\|_2^2 \mu_1 \Omega_1)}{\gamma_1 \gamma_4^2 \mu_1} \\
& + \frac{2\theta_1^2 \theta_4^2 \|\tilde{y}\|_2^2 (\gamma_2 \gamma_8 \gamma_9 \mu_3 + \theta_4^2 \|\tilde{z}\|_2^2 \gamma_7 \mu_2)}{\gamma_4^2 \mu_2 \mu_3},
\end{aligned}$$



$$\begin{aligned}
l_6 = & \frac{2\theta_4^4 \|\tilde{z}\|_2^2 \gamma_9 (\theta_1^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \gamma_3 \Omega_3 - \|\tilde{y}\|_2^2 \gamma_3 \gamma_9 \Omega_5 + 2\theta_3^2 \gamma_4 \Omega_6 + \|\tilde{z}\|_2^2 \gamma_7 \Omega_6)}{\gamma_3 \gamma_4^2} \\
& + \frac{2\theta_3^2 \theta_4^2 (\theta_1^2 \theta_2^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \mu_2 \mu_3 \Omega_2 - \theta_2^2 \gamma_2 \gamma_9 \mu_3 - \theta_2^2 \|\tilde{y}\|_2^2 \gamma_9 \mu_2 \mu_3 \Omega_4 - 2\theta_4^2 \gamma_2 \gamma_9 \mu_2)}{\gamma_2 \gamma_4 \mu_2 \mu_3} \\
& + \frac{2\theta_1^2 \theta_4^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \gamma_8 (\gamma_9 \mu_1 \Omega_2 - \theta_1^2 \gamma_1 - \theta_1^2 \|\tilde{x}\|_2^2 \mu_1 \Omega_1)}{\gamma_4^2 \mu_1} \\
& + \frac{2\theta_1^2 \theta_4^2 \|\tilde{x}\|_2^2 \gamma_8 \Omega_3 (2\theta_3^2 \gamma_4 + \|\tilde{z}\|_2^2 \gamma_7)}{\gamma_3 \gamma_4^2} \\
& + \frac{2\theta_4^2 \|\tilde{y}\|_2^2 \gamma_6 (\theta_1^2 \|\tilde{x}\|_2^2 \|\tilde{y}\|_2^2 \gamma_3 \Omega_2 - \|\tilde{y}\|_2^2 \gamma_3 \gamma_9 \Omega_4 - 2\theta_3^2 \gamma_4 \Omega_5 - \|\tilde{z}\|_2^2 \gamma_7 \Omega_5)}{\gamma_2 \gamma_3 \gamma_4^2} \\
& - \frac{2\theta_4^2 (\|\tilde{y}\|_2^2 \gamma_2 \gamma_3 \gamma_6 \gamma_9 \mu_3 + 2\theta_2^2 \theta_3^4 \gamma_4^2 \mu_2 \mu_3 \Omega_5 + \theta_2^2 \theta_3^2 \|\tilde{z}\|_2^2 \gamma_4 \gamma_7 \mu_2 \mu_3 \Omega_5 + \theta_4^2 \|\tilde{z}\|_2^2 \gamma_2 \gamma_3 \gamma_7 \gamma_9 \mu_2)}{\gamma_2 \gamma_3 \gamma_4^2 \mu_2 \mu_3}.
\end{aligned}$$

其中  $\gamma_1, \gamma_2, \gamma_3, \gamma_4, \gamma_5, \gamma_6, \gamma_7, \gamma_8$  和  $\gamma_9$  在定理 3.1 的前面部分定义过,  $\mu_1, \mu_2, \mu_3, \Omega_1, \Omega_2, \Omega_3, \Omega_4, \Omega_5$  和  $\Omega_6$  在定理 3.2 的前面部分定义过, 最后将 (3.18) 和 (3.3) 带入 (3.17) 中得到期望的表达式 (3.16).

结构向后误差  $\eta^{(\theta_1, \theta_2, \theta_3, \theta_4, \lambda_1, \lambda_2, \lambda_3)}(\tilde{x}, \tilde{y}, \tilde{z})$  的表达式 (3.16) 虽然看起来很复杂, 但是从计算的角度来看, 表达式 (3.16) 是容易计算的, 因为它只涉及加、减、乘、除这样基本的运算.

## 4. 数值实验

本节将给出一个数值例子来比较第 3 节中推出的结构向后误差  $\eta_S(\tilde{x}, \tilde{y}, \tilde{z})$  和相应的无结构向后误差  $\eta(\tilde{t})$ . 数值实验在 MATLAB R2015b 中进行, 机器精度为  $2.2204 \times 10^{-16}$ .

考虑线性系统 (1.1) 满足

$$A = MPM, \quad B = \begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 10^{-3} & 0 & 0 & 0 & 0 & 0 \end{bmatrix}, \quad C = \begin{bmatrix} 1 & -2 & 1 \\ -2 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

和

$$D = -I_6, \quad f = [10^8, 10, 0, 0, 0, 0]^T, \quad g = [10^{-8}, 0, 0]^T, \quad h = [10^{-8}, 0, 0]^T$$

其中

$$M = \text{diag}(1, 5, 10, 50, 100, 10000), \quad P = (p_{ij}), \quad p_{ij} = \frac{(i+j-2)!}{(i-1)!(j-1)!}$$

这个问题是由文 [11] 中的 Example 5.2 修改而来. 很显然系数矩阵是非奇异的. 使用列选主元的高斯消去法可以得到一个计算解  $\tilde{t} = (\tilde{x}^T, \tilde{y}^T, \tilde{z}^T)^T$ , 其中

$$\tilde{x} = \begin{pmatrix} 2.0264 \times 10^8 \\ 3.0780 \times 10^6 \\ -4.1382 \times 10^7 \\ 1.0019 \times 10^7 \\ -2.5379 \times 10^6 \\ 4.8888 \times 10^3 \end{pmatrix}, \tilde{y} = \begin{pmatrix} 4.8552 \times 10^7 \\ -1.0059 \times 10^8 \\ -2.4993 \times 10^8 \end{pmatrix} \text{ 和 } \tilde{z} = \begin{pmatrix} -2.0264 \times 10^5 \\ 3.4833 \times 10^6 \\ 4.8552 \times 10^7 \\ 0.0000 \\ 0.0000 \\ 0.0000 \end{pmatrix}.$$

由 (1.2) 得, 无结构向后误差

$$\eta(\tilde{t}) = 9.7551 \times 10^{-22}.$$

由 (1.5) 和 (3.16) 得, 结构向后误差

$$\eta_S(\tilde{x}, \tilde{y}, \tilde{z}) = 3.8498 \times 10^{-6}.$$

## 5. 总结

本文首先利用文 [13, 19] 中的技巧和 Kronecker 积的性质, 将矩阵优化问题转化为正定二次型的最小值问题, 然后利用 Sherman-Morrison-Woodbury 公式, 最终获得结构向后误差的可计算的具体表达式. 最后我们给出了一个数值实验, 以证明我们的结果可以很容易地用来测试实际数值算法的稳定性.

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