

## MULTIPLICATIVE PERTURBATION ANALYSIS FOR THE GENERALIZED CHOLESKY BLOCK DOWNDATING PROBLEM

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(Communicated by T. Burić)

*Abstract.* This article is devoted to the multiplicative perturbation analysis of the generalized Cholesky block downdating problem. The strong rigorous multiplicative perturbation bounds are first presented by bringing together the modified matrix-vector equation approach with the technique of Lyapunov majorant function and the Banach fixed point theorem. Then, the weak rigorous multiplicative bounds are developed by using the matrix-equation approach. Numerical results demonstrate that these bounds are constantly tighter than the additive perturbation bounds.

### 1. Introduction

Suppose that  $A \in \mathbb{R}_m^{m \times m}$  is symmetric positive definite,  $B \in \mathbb{R}_n^{n \times m}$ , and  $C \in \mathbb{R}^{n \times n}$  are symmetric positive and semi-definite, then the symmetric quasi-definite matrix  $K \in \mathbb{R}^{(m+n) \times (m+n)}$  can be expressed as

$$K = \begin{bmatrix} A & B^T \\ B & -C \end{bmatrix}. \quad (1.1)$$

The matrix  $K$  always has the generalized Cholesky factorization

$$K = LJ_{m+n}L^T, \quad (1.2)$$

where

$$L = \begin{bmatrix} L_{11} & 0 \\ L_{21} & L_{22} \end{bmatrix}, \quad J_{m+n} = \begin{bmatrix} I_m & 0 \\ 0 & -I_n \end{bmatrix},$$

$L_{11} \in \mathbb{R}_m^{m \times m}$ ,  $L_{22} \in \mathbb{R}_n^{n \times n}$  are lower triangular and  $L_{21} \in \mathbb{R}_n^{n \times m}$ . From (1.2), it can be simply verify that

$$A = L_{11}L_{11}^T, \quad B = L_{21}L_{11}^T, \quad C + L_{21}L_{21}^T = L_{22}L_{22}^T.$$

*Mathematics subject classification* (2020): 15A45, 15A23.

*Keywords and phrases:* Generalized Cholesky block downdating problem, multiplicative perturbation, rigorous perturbation bounds, Lyapunov majorant function, Banach fixed point theorem, matrix-equation approach.

This work was supported by the Zhejiang Normal University Postdoctoral Research Fund (Grant No. ZC304022938), the Natural Science Foundation of China (Project No. 61976196) and the Zhejiang Provincial Natural Science Foundation of China under Grant No. LZ22F030003.

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If the diagonal elements of the lower triangular matrices,  $L_{11}$  and  $L_{22}$  are positive, the factorization (1.2) is unique and  $L$  is known as the generalized Cholesky factor [1].

In this paper, we consider the generalized Cholesky block downdating problem (GCBD)

$$LJ_{m+n}L^T - YY^T = VJ_{m+n}V^T. \tag{1.3}$$

Given that  $K$  is the same as in (1.1) and  $Y = (Y_m^T, Y_n^T) \in \mathbb{R}^{m+n \times k}$ , find a lower triangular matrix

$$V = \begin{bmatrix} V_{11} & 0 \\ V_{21} & V_{22} \end{bmatrix} \in \mathbb{R}^{(m+n) \times (m+n)},$$

where  $V_{11} \in \mathbb{R}_m^{m \times m}$ ,  $V_{22} \in \mathbb{R}_n^{n \times n}$  are lower triangular with positive elements and  $V_{21} \in \mathbb{R}_n^{n \times m}$ . From [2, Corollary 1], it is simply to show that when  $\|L^{-1}Y\|_2 < 1$  the GCBD problem (1.3) is always exists and the matrix  $V$  is known as the GCBD factor. Moreover, in this case

$$A - Y_m Y_m^T = V_{11} V_{11}^T, \quad B - Y_n Y_n^T = V_{21} V_{11}^T, \quad C + V_{21} V_{21}^T + Y_n Y_n^T = V_{22} V_{22}^T.$$

This problem is reduced to the Cholesky block downdating problem if we choose  $K = A$ , i.e.,  $B$  and  $C$  are nonexistent. The Cholesky block downdating problem has acquired remarkable consideration, and its special case, i.e., single downdating ( $Y \in \mathbb{R}^{m \times 1}$ ), has been extensively studied in the literature; see [3, 4, 5, 6, 7, 8, 9, 10] for details. Additive perturbation and multiplicative perturbation are the two distinctive types of perturbation models of matrix factorizations. The additive perturbation analysis for the GCBD problem has been considered in [11, 12].

Obviously, an additive perturbation can be obtained from the multiplicative perturbation, but the derived additive perturbation bounds will destroy the unique structures of multiplicative perturbations and lose their tendency [13]. Since the matrix scaling technique is often employed to provide better-conditioned problems [13], the multiplicative perturbation has received significant attention, and has some elements of interest compared with the additive perturbation; see [13, 14, 15, 16, 17] and references therein. Until now, there has been no work on the multiplicative perturbation bounds for the GCBD problem. So, it is interesting to introduce the multiplicative perturbation bounds for the GCBD problem.

Particularly, for (1.3), we assume two types of multiplicative perturbation matrices,  $W = 1_{m+n} + N$  on  $L$ : (1)  $W = I_{m+n} + N$  is a general matrix; (2)  $W = I_{m+n} + N$  is a lower triangular matrix. The rest of the paper is organized as follows: In Section 3, we will develop the strong rigorous multiplicative perturbation bounds by using the modified matrix-vector equation approach [13, 18, 19, 20], the Lyapunov majorant function (e.g., [11, Chapter 5]), and the Banach fixed point theorem (e.g., [11, Appendix 5]). Moreover, we will use the matrix-equation approach [21] to derive the weak rigorous multiplicative bounds in Section 4. These bounds will be less expensive to compute as compared to the strong, rigorous multiplicative perturbation bounds. Section 2 provides some useful notation and preliminary knowledge. Finally, we provide some numerical experiments to verify the theoretical results, and we show the numerical comparison between the multiplicative rigorous perturbation bounds and the additive rigorous perturbation bounds [11, 12] for the GCBD problem in Section 5.

### 2. Preliminaries

Some notations can be endorsed in [21] to make the presentation apparent. We still illustrate them here to make easier for readers.

The Frobenius norm and spectral norm for a given matrix  $Z = (z_{ij}) \in \mathbb{R}^{m \times n}$  are denoted by  $\|Z\|_F$  and  $\|Z\|_2$ , respectively. The following inequalities hold for these two matrix norms; see [22] for details.

$$\|QRS\|_2 \leq \|Q\|_2 \|R\|_2 \|S\|_2, \quad \|QRS\|_F \leq \|Q\|_2 \|R\|_F \|S\|_2, \tag{2.1}$$

whenever the matrix product  $QRS$  is well-defined.

For any matrix  $Z = [z_1, z_2, \dots, z_n] = (z_{ij}) \in \mathbb{R}^{n \times n}$ , denote the vector of the last  $i$  elements of  $z_j$  by  $z_j^{(i)}$  and define

$$\text{lvec}(Z) := \begin{bmatrix} z_1^{(n)} \\ z_2^{(n-1)} \\ \vdots \\ z_n^{(1)} \end{bmatrix} \in \mathbb{R}^{v_1}, \quad \text{vec}(Z) := \begin{bmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{bmatrix} \in \mathbb{R}^{n^2}, \quad \text{slt}(Z) := \begin{bmatrix} 0 & 0 & 0 & \cdots & 0 \\ z_{21} & 0 & 0 & \cdots & 0 \\ z_{31} & z_{32} & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ z_{n1} & z_{n2} & \cdots & z_{n,n-1} & 0 \end{bmatrix},$$

$$\text{low}(Z) := \begin{bmatrix} \frac{1}{2}z_{11} & 0 & \cdots & 0 \\ z_{21} & \frac{1}{2}z_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ z_{n1} & z_{n2} & \cdots & \frac{1}{2}z_{nn} \end{bmatrix}, \quad \text{lt}(Z) := \begin{bmatrix} z_{11} & 0 & \cdots & 0 \\ z_{21} & z_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ z_{n1} & z_{n2} & \cdots & z_{nn} \end{bmatrix},$$

$\text{sut}(Z) = \text{slt}(Z^T)^T$ , and  $\text{diag}(Z) = \text{diag}(z_{11}, z_{22}, \dots, z_{nn})$ , where  $v_1 = \frac{n(n+1)}{2}$ . Using the structures of these operators, we have

$$\text{lvec}(Z) = \mathfrak{K}_{\text{lvec}} \text{vec}(Z), \quad \text{vec}(\text{lt}(Z)) = \mathfrak{K}_{\text{lt}} \text{vec}(Z), \quad \text{vec}(\text{low}(Z)) = \mathfrak{K}_{\text{low}} \text{vec}(Z), \tag{2.2}$$

where

$$\begin{aligned} \mathfrak{K}_{\text{lvec}} &= \text{diag}(G_1, G_2, \dots, G_n) \in \mathbb{R}^{v_1 \times n^2}, \quad G_i = [0_{n-(i-1) \times (i-1)}, I_{n-(i-1)}] \in \mathbb{R}^{n-(i-1) \times n}, \\ \mathfrak{K}_{\text{lt}} &= \text{diag}(\hat{G}_1, \hat{G}_2, \dots, \hat{G}_n) \in \mathbb{R}^{n^2 \times n^2}, \quad \hat{G}_i = \text{diag}(0_{(i-1) \times (i-1)}, I_{n-(i-1)}) \in \mathbb{R}^{n \times n}, \\ \mathfrak{K}_{\text{low}} &= \text{diag}(\tilde{G}_1, \tilde{G}_2, \dots, \tilde{G}_n) \in \mathbb{R}^{n^2 \times n^2}, \quad \tilde{G}_i = \text{diag}(0_{(i-1) \times (i-1)}, 1/2, I_{n-i}) \in \mathbb{R}^{n \times n}. \end{aligned}$$

Moreover,

$$\mathfrak{K}_{\text{lvec}} \mathfrak{K}_{\text{lvec}}^T = I_{v_1}, \quad \mathfrak{K}_{\text{lvec}}^T \mathfrak{K}_{\text{lvec}} = \mathfrak{K}_{\text{lt}}. \tag{2.3}$$

Let  $\text{lvec}^\dagger : \mathbb{R}^{v_1} \rightarrow \mathbb{R}^{n \times n}$  be the right inverse of the operator ‘lvec’ such that  $\text{lvec} \cdot \text{lvec}^\dagger = I_{v_1 \times v_1}$  and  $\text{lvec}^\dagger \cdot \text{lvec} = \text{lt}$ . Then the matrix of the operator ‘lvec’ is  $\mathfrak{K}_{\text{lvec}}^T$ . That is,  $\text{lvec}^\dagger(Z) = \mathfrak{K}_{\text{lvec}}^T \text{vec}(Z)$ .

Let  $\mathbb{D}_n \in \mathbb{R}^{n \times n}$  be the set of diagonal matrices with positive diagonal elements. Then, for any  $D_n = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_n) \in \mathbb{D}_n$ , it follows that

$$\text{low}(ZD_n) = \text{low}(Z)D_n, \quad \text{low}(D_nZ) = D_n\text{low}(Z). \tag{2.4}$$

Moreover, from [23, Lemma 5.1], we have

$$\|\text{low}(Z) + D_n\text{low}(Z^T)D_n^{-1}\|_F \leq \sqrt{1 + \zeta_{D_n}^2} \|Z\|_F, \tag{2.5}$$

where  $\zeta_{D_n} = \max_{1 \leq i < j \leq n} \{\sigma_j / \sigma_i\}$ . From [21], we have

$$\|\text{low}(Z)\|_F \leq \|Z\|_F. \tag{2.6}$$

If  $Z$  is symmetric, then

$$\|\text{low}(Z)\|_F \leq \frac{1}{\sqrt{2}} \|Z\|_F. \tag{2.7}$$

Also, we have

$$\|\text{low}(Z + Z^T)\|_F \leq \sqrt{2} \|Z\|_F. \tag{2.8}$$

The Kronecker product between  $Z = (z_{ij}) \in \mathbb{R}^{m \times n}$  and  $X \in \mathbb{R}^{p \times q}$  is defined as  $Z \otimes X = [z_{ij}X] \in \mathbb{R}^{mp \times nq}$ . Some useful results of the Kronecker product are listed below [24]:

$$\text{vec}(ZCX) = (X^T \otimes Z)\text{vec}(C), \tag{2.9}$$

$$\Pi \text{vec}(Z) = \text{vec}(Z^T), \tag{2.10}$$

$$\|X \otimes Z\|_2 = \|X\|_2 \|Z\|_2, \tag{2.11}$$

$$(X \otimes Z)(B \otimes C) = (XB \otimes ZC), \tag{2.12}$$

$$(X \otimes Z)^{-1} = X^{-1} \otimes Z^{-1}, \text{ if } X \text{ and } Z \text{ are nonsingular,} \tag{2.13}$$

where  $B$  and  $C$  are of suitable orders.

### 3. Strong rigorous multiplicative perturbation bounds

Let us consider that the matrices  $L$ ,  $Y$  and  $V$  in (1.3) are perturbed as

$$L \rightarrow LW, \quad Y \rightarrow YU, \quad V \rightarrow \Delta V,$$

where  $W = I_{m+n} + N \in \mathbb{R}^{(m+n) \times (m+n)}$ ,  $U = I_{m+n} + M \in \mathbb{R}^{(m+n) \times (m+n)}$  and  $\Delta V \in \mathbb{R}^{(m+n) \times (m+n)}$  is a lower triangular matrix with positive diagonal elements. Therefore, the perturbed form of (1.3) is

$$\begin{aligned} (V + \Delta V)J_{m+n}(V + \Delta V)^T &= (I_{m+n} + N)LJ_{m+n}L^T(I_{m+n} + N)^T \\ &\quad - (I_{m+n} + M)YY^T(I_{m+n} + M)^T. \end{aligned} \tag{3.1}$$

Extending (3.1) and using (1.3), we have

$$VJ_{m+n}(\Delta V)^T + (\Delta V)J_{m+n}V^T = N(LJ_{m+n}L^T) + (LJ_{m+n}L^T)N^T - M(Y Y^T) - (Y Y^T)M^T + N(LJ_{m+n}L^T)N^T - M(Y Y^T)M^T - \Delta VJ_{m+n}(\Delta V)^T. \quad (3.2)$$

Premultiplying (3.2) by  $V^{-1}$  and right postmultiplying it by  $V^{-T}$  lead to

$$\begin{aligned} & J_{m+n}(\Delta V)^T V^{-T} + V^{-1}(\Delta V)J_{m+n} \\ &= V^{-1}(N(LJ_{m+n}L^T) + (LJ_{m+n}L^T)N^T)V^{-T} - V^{-1}(M(Y Y^T) + (Y Y^T)M^T)V^{-T} \\ & \quad + V^{-1}(N(LJ_{m+n}L^T)N^T - M(Y Y^T)M^T - \Delta VJ_{m+n}(\Delta V)^T)V^{-T}. \end{aligned} \quad (3.3)$$

As performed in [8, 9, 11, 14], from (3.3), we have

$$\begin{aligned} V^{-1}\Delta VJ_{m+n} &= \text{low}(V^{-1}(N(LJ_{m+n}L^T) + (LJ_{m+n}L^T)N^T)V^{-T}) \\ & \quad - \text{low}(V^{-1}(M(Y Y^T) + (Y Y^T)M^T)V^{-T}) \\ & \quad + \text{low}(V^{-1}(N(LJ_{m+n}L^T)N^T - M(Y Y^T)M^T - \Delta VJ_{m+n}(\Delta V)^T)V^{-T}). \end{aligned} \quad (3.4)$$

Applying the operator ‘vec’ to (3.4), and noting (2.2), (2.9) and (2.10), we get

$$\begin{aligned} & (J_{m+n}^T \otimes V^{-1}) \text{vec}(\Delta V) \\ &= \mathfrak{K}_{\text{low}}((V^{-1}(LJ_{m+n}L^T) \otimes V^{-1}) + (V^{-1} \otimes V^{-1}(LJ_{m+n}L^T))\Pi) \text{vec}(N) \\ & \quad - \mathfrak{K}_{\text{low}}((V^{-1}(Y Y^T)^T \otimes V^{-1}) + (V^{-1} \otimes V^{-1}(Y Y^T))\Pi) \text{vec}(M) \\ & \quad + \mathfrak{K}_{\text{low}}(V^{-1} \otimes V^{-1}) \text{vec}(N(LJ_{m+n}L^T)N^T - M(Y Y^T)M^T - \Delta VJ_{m+n}(\Delta V)^T). \end{aligned}$$

As performed in [8, 9, 11, 14], we can obtain

$$\begin{aligned} \text{vec}(\Delta V) &= (J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}}((V^{-1}(LJ_{m+n}L^T) \otimes V^{-1}) \\ & \quad + (V^{-1} \otimes V^{-1}(LJ_{m+n}L^T))\Pi) \text{vec}(N) \\ & \quad - (J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}}((V^{-1}(Y Y^T)^T \otimes V^{-1}) + (V^{-1} \otimes V^{-1}(Y Y^T))\Pi) \text{vec}(M) \\ & \quad + (J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}}(V^{-1} \otimes V^{-1}) \\ & \quad \times \text{vec}(N(LJ_{m+n}L^T)N^T - M(Y Y^T)M^T - \Delta VJ_{m+n}(\Delta V)^T), \end{aligned} \quad (3.5)$$

and show that (3.5) is equivalent to

$$\begin{aligned} \text{lvec}(\Delta V) &= \mathfrak{K}_{\text{lvec}}(J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}}((V^{-1}(LJ_{m+n}L^T) \otimes V^{-1}) \\ & \quad + (V^{-1} \otimes V^{-1}(LJ_{m+n}L^T))\Pi) \text{vec}(N) \\ & \quad - \mathfrak{K}_{\text{lvec}}(J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}}((V^{-1}(Y Y^T)^T \otimes V^{-1}) \\ & \quad + (V^{-1} \otimes V^{-1}(Y Y^T))\Pi) \text{vec}(M) \\ & \quad + \mathfrak{K}_{\text{lvec}}(J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}}(V^{-1} \otimes V^{-1}) \text{vec}(N(LJ_{m+n}L^T)N^T \\ & \quad - M(Y Y^T)M^T - \Delta VJ_{m+n}(\Delta V)^T). \end{aligned} \quad (3.6)$$

As a matter of convenience, suppose

$$\begin{aligned}
 Q_{LM} &= \mathfrak{K}_{\text{lvec}}(J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}}((V^{-1}(LJ_{m+n}L^T) \otimes V^{-1}) + (V^{-1} \otimes V^{-1}(LJ_{m+n}L^T)) \Pi) \\
 Q_Y &= \mathfrak{K}_{\text{lvec}}(J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}}((V^{-1}(YY^T)^T \otimes V^{-1}) + (V^{-1} \otimes V^{-1}(YY^T)) \Pi) \\
 R_V &= \cdot \mathfrak{K}_{\text{lvec}}(J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}}(V^{-1} \otimes V^{-1}).
 \end{aligned} \tag{3.7}$$

Then, (3.6) becomes

$$\begin{aligned}
 \text{lvec}(\Delta V) &= (Q_{LM} \text{vec}(N) - Q_Y \text{vec}(M) \\
 &\quad + R_V \text{vec}(N(LJ_{m+n}L^T)N^T - M(YY^T)M^T - \Delta V J_{m+n}(\Delta V)^T)).
 \end{aligned} \tag{3.8}$$

Thus, applying the operator ‘lvec<sup>†</sup>’ to (3.8) yields

$$\begin{aligned}
 \Delta V &= \text{lvec}^\dagger(Q_{LM} \text{vec}(N) - Q_Y \text{vec}(M) \\
 &\quad + R_V \text{vec}(N(LJ_{m+n}L^T)N^T - M(YY^T)M^T - \Delta V J_{m+n}(\Delta V)^T)).
 \end{aligned} \tag{3.9}$$

We will derive the rigorous multiplicative perturbation bound for  $\Delta V$  using the method of the Lyapunov majorant function and the Banach fixed point theorem based on the operator equation (3.10) as shown in [8, 9, 11, 14]. The equation (3.9) can be written as an operator equation for  $\Delta V$ :

$$\begin{aligned}
 \Delta V &= \Phi(\Delta V, N, M) \\
 &= \text{lvec}^\dagger(Q_{LM} \text{vec}(N) - Q_Y \text{vec}(M) \\
 &\quad + R_V \text{vec}(N(LJ_{m+n}L^T)N^T - M(YY^T)M^T - \Delta V J_{m+n}(\Delta V)^T)).
 \end{aligned} \tag{3.10}$$

Suppose that  $H \in \mathbb{R}^{(m+n) \times (m+n)}$  is lower triangular with positive diagonal elements and has the same structure as that of  $\Delta V$ ,  $\|H\|_F \leq \eta$  for some  $\eta \geq 0$ ,  $\|N\|_F = \sigma_1$  and  $\|M\|_F = \sigma_2$ . Then it follows from the definition of the operator ‘lvec<sup>†</sup>’ and (2.1) that

$$\|\Phi(H, N, M)\|_F \leq \|Q_{LM}\|_2 \sigma_1 + \|Q_Y\|_2 \sigma_2 + \|R_V\|_2 (\|L\|_2^2 \sigma_1^2 + \|Y\|_2^2 \sigma_2^2 + \eta^2). \tag{3.11}$$

Using (3.11), we get the Lyapunov majorant function of the operator equation (3.10)

$$q(\eta, \sigma_1, \sigma_2) = \|Q_{LM}\|_2 \sigma_1 + \|Q_Y\|_2 \sigma_2 + \|R_V\|_2 (\|L\|_2^2 \sigma_1^2 + \|Y\|_2^2 \sigma_2^2 + \eta^2)$$

and the Lyapunov majorant equation

$$\begin{aligned}
 q(\eta, \sigma_1, \sigma_2) &= \eta, \quad \text{i.e.} \\
 \|Q_{LM}\|_2 \sigma_1 + \|Q_Y\|_2 \sigma_2 + \|R_V\|_2 (\|L\|_2^2 \sigma_1^2 + \|Y\|_2^2 \sigma_2^2 + \eta^2) &= \eta.
 \end{aligned} \tag{3.12}$$

Suppose that  $\sigma_1$ ,

$$\begin{aligned}
 \sigma_2 \in \Omega &= \{ \sigma_1, \sigma_2 \geq 0 : 1 - 4 \|R_V\|_2 (\|Q_{LM}\|_2 \sigma_1 + \|Q_Y\|_2 \sigma_2 \\
 &\quad + \|R_V\|_2 (\|L\|_2^2 \sigma_1^2 + \|Y\|_2^2 \sigma_2^2)) \geq 0 \}.
 \end{aligned}$$

Then, the equation (3.12) has two nonnegative roots:  $\eta_1(\sigma_1, \sigma_2) \leq \eta_2(\sigma_1, \sigma_2)$  with

$$\eta_1(\sigma_1, \sigma_2) = f(\sigma_1, \sigma_2) = \frac{2(\|Q_{LM}\|_2 \sigma_1 + \|Q_Y\|_2 \sigma_2 + \|R_V\|_2 (\|L\|_2^2 \sigma_1^2 + \|Y\|_2^2 \sigma_2^2))}{1 + \sqrt{1 - 4\|R_V\|_2 (\|Q_{LM}\|_2 \sigma_1 + \|Q_Y\|_2 \sigma_2 + \|R_V\|_2 (\|L\|_2^2 \sigma_1^2 + \|Y\|_2^2 \sigma_2^2))}}$$

Let the set  $A(\sigma_1, \sigma_2)$  be  $A(\sigma_1, \sigma_2) = \{H \in \mathbb{R}^{(m+n) \times (m+n)} : H \text{ has the same structure as that of } \Delta V \text{ and } \|H\|_F \leq \eta\}$ , which is closed and convex. Furthermore, we can simply verify that the operator  $\Phi(\cdot, N, M)$  maps the set  $A(\sigma_1, \sigma_2)$  into itself for  $H, \tilde{H} \in A(\sigma_1, \sigma_2)$ ,

$$\|\Phi(H, N, M) - \Phi(\tilde{H}, N, M)\|_F \leq q'_\eta(f(\sigma_1, \sigma_2), \sigma_1, \sigma_2) \|H - \tilde{H}\|_F.$$

Since the derivative of the function  $q(\eta, \sigma_1, \sigma_2)$  relative to  $\eta$  at  $f(\sigma_1, \sigma_2)$  satisfies

$$q'_\eta(f(\sigma_1, \sigma_2), \sigma_1, \sigma_2) = 1 - \sqrt{1 - 4\|R_V\|_2 (\|Q_{LM}\|_2 \sigma_1 + \|Q_Y\|_2 \sigma_2 + \|R_V\|_2 (\|L\|_2^2 \sigma_1^2 + \|Y\|_2^2 \sigma_2^2))} < 1,$$

when  $\sigma_1, \sigma_2 \in \Omega_1 = \{\sigma_1, \sigma_2 \geq 0 : 1 - 4\|R_V\|_2 (\|Q_{LM}\|_2 \sigma_1 + \|Q_Y\|_2 \sigma_2 + \|R_V\|_2 (\|L\|_2^2 \sigma_1^2 + \|Y\|_2^2 \sigma_2^2)) > 0\}$ . Then the operator  $\Phi(\cdot, N, M)$  is contractive on the set  $A(\sigma_1, \sigma_2)$  for  $\sigma_1, \sigma_2 \in \Omega_1$ . Thus, from the Banach fixed point theorem, we have that the operator Equation (3.10), i.e. the matrix equation (3.2), has a unique solution in the set  $A(\sigma_1, \sigma_2)$ . As a result,  $\|\Delta V\|_F \leq f(\sigma_1, \sigma_2)$  for  $\sigma_1, \sigma_2 \in \Omega_1$ . We summarize these results in the following theorem.

**THEOREM 3.1.** *Given a lower triangular matrix  $L \in \mathbb{R}^{(m+n) \times (m+n)}$  with positive diagonal elements and a matrix  $Y \in \mathbb{R}^{(m+n) \times k}$  such that the generalized Cholesky factorization  $VJ_{m+n}V^T = LJ_{m+n}L^T - YY^T$  holds. Suppose  $W = I_{m+n} + N \in \mathbb{R}^{(m+n) \times (m+n)}$  and  $U = I_{m+n} + M \in \mathbb{R}^{(m+n) \times (m+n)}$ . If*

$$\|R_V\|_2 (\|Q_{LM}\|_2 \|N\|_F + \|Q_Y\|_2 \|M\|_F + \|R_V\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2)) < \frac{1}{4}, \tag{3.13}$$

then the following generalized Cholesky factorization holds:

$$\begin{aligned} & (V + \Delta V)J_{m+n}(V + \Delta V)^T \\ &= LW(LW)^T - YU(YU)^T, \\ &= (I_{m+n} + N)LJ_{m+n}L^T(I_{m+n} + N)^T - (I_{m+n} + M)YY^T(I_{m+n} + M)^T, \end{aligned}$$

and

$$\begin{aligned} & \|\Delta V\|_F \\ & \leq \frac{2(\|Q_{LM}\|_2 \|N\|_F + \|Q_Y\|_2 \|M\|_F + \|R_V\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2))}{1 + \sqrt{1 - 4\|R_V\|_2 (\|Q_{LM}\|_2 \|N\|_F + \|Q_Y\|_2 \|M\|_F + \|R_V\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2))}} \end{aligned} \tag{3.14}$$

$$\leq 2(\|Q_{LM}\|_2 \|N\|_F + \|Q_Y\|_2 \|M\|_F + \|R_V\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2)). \tag{3.15}$$

*Proof.* It is absolutely not difficult to check that the condition (3.13) is the same as the one in  $\Omega_1$ . So, (3.14) and thus (3.15) hold.  $\square$

REMARK 3.2. The resulting first order multiplicative perturbation bound can be obtained from (3.15) by neglecting high-order terms

$$\begin{aligned} \|\Delta V\|_F &\leq \|Q_{LM}\|_2 \|N\|_F + \|Q_Y\|_2 \|M\|_F \\ &= \|Q_{LM}\|_2 \|W - I_{m+n}\|_F + \|Q_Y\|_2 \|U - I_{m+n}\|_F. \end{aligned} \tag{3.16}$$

We used  $W = I_{m+n} + N$  as a general matrix in the previous analysis. Next, we choose  $W = I_{m+n} + N$  as a lower triangular matrix. Thus, from (3.10), (2.2) and (2.3), it follows that

$$\begin{aligned} \text{lvec}(\Delta V) &= (Q_{LT} \text{lvec}(N) - Q_Y \text{vec}(\Delta Y) \\ &\quad + R_V \text{vec}(N(LJ_{m+n}L^T)N^T - M(YY^T)M^T - \Delta VJ_{m+n}(\Delta V)^T)). \end{aligned} \tag{3.17}$$

where

$$\begin{aligned} Q_{LT} &= \mathfrak{K}_{\text{lvec}}(J_{m+n}^T \otimes V) \mathfrak{K}_{\text{low}}((V^{-1}(LJ_{m+n}L^T) \otimes V^{-1}) \\ &\quad + (V^{-1} \otimes V^{-1}(LJ_{m+n}L^T)) \Pi) \mathfrak{K}_{\text{lvec}}^T. \end{aligned} \tag{3.18}$$

Based on the results of Theorem 3.1, we have the following theorem.

THEOREM 3.3. *Given a lower triangular matrix  $L \in \mathbb{R}^{(m+n) \times (m+n)}$  with positive diagonal elements and a matrix  $Y \in \mathbb{R}^{(m+n) \times k}$  such that the generalized Cholesky factorization  $VJ_{m+n}V^T = LJ_{m+n}L^T - YY^T$  holds. Suppose  $W = I_{m+n} + N \in \mathbb{R}^{(m+n) \times (m+n)}$  and  $U = I_{m+n} + M \in \mathbb{R}^{(m+n) \times (m+n)}$ . If*

$$\|R_V\|_2 (\|Q_{LT}\|_2 \|N\|_F + \|Q_Y\|_2 \|M\|_F + \|R_V\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2)) < \frac{1}{4}, \tag{3.19}$$

then the following generalized Cholesky factorization holds:

$$\begin{aligned} &(V + \Delta V)J_{m+n}(V + \Delta V)^T \\ &= LW(LW)^T - YU(YU)^T, \\ &= (I_{m+n} + N)LJ_{m+n}L^T(I_{m+n} + N)^T - (I_{m+n} + M)YY^T(I_{m+n} + M)^T, \end{aligned}$$

and

$$\begin{aligned} &\|\Delta V\|_F \\ &\leq \frac{2(\|Q_{LT}\|_2 \|N\|_F + \|Q_Y\|_2 \|\Delta Y\|_F + \|R_V\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2))}{1 + \sqrt{1 - 4\|R_V\|_2 (\|Q_{LT}\|_2 \|N\|_F + \|Q_Y\|_2 \|M\|_F + \|R_V\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2))}}, \end{aligned} \tag{3.20}$$

$$\leq 2(\|Q_{LT}\|_2 \|N\|_F + \|Q_Y\|_2 \|M\|_F + \|R_V\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2)). \tag{3.21}$$



REMARK 3.4. By ignoring the high-order terms, we can investigate a first-order multiplicative perturbation bound from (3.21):

$$\begin{aligned} \|\Delta V\|_F &\leq \|Q_{LT}\|_2 \|N\|_F + \|Q_Y\|_2 \|M\|_F \\ &= \|Q_{LT}\|_2 \|W - I_{m+n}\|_F + \|Q_Y\|_2 \|U - I_{m+n}\|_F. \end{aligned} \quad (3.22)$$

#### 4. Weak rigorous multiplicative perturbation bounds

To obtain the explicit expression of weak rigorous multiplicative perturbation bounds for the GCBD problem, we will adopt the matrix-equation approach originated by Chang [21] in this section.

THEOREM 4.1. Given a lower triangular matrix  $L \in \mathbb{R}^{(m+n) \times (m+n)}$  with positive diagonal elements and a matrix  $Y \in \mathbb{R}^{(m+n) \times k}$  such that the generalized Cholesky factorization  $VJ_{m+n}V^T = LJ_{m+n}L^T - YY^T$  holds. Suppose  $S = (S_m^T, S_n^T)^T \in \mathbb{R}^{(m+n) \times k}$  and  $H \in \mathbb{R}^{(m+n) \times (m+n)}$  be lower triangular of the following form

$$H = \begin{bmatrix} H_{11} & 0 \\ H_{21} & H_{22} \end{bmatrix},$$

where  $H_{11} \in \mathbb{R}^{m \times m}$  and  $H_{22} \in \mathbb{R}^{n \times n}$  are lower triangular,  $H_{21} \in \mathbb{R}^{n \times m}$ . Define  $N = \varepsilon H$  and  $M = \varepsilon S$  for some  $\varepsilon \geq 0$ . If

$$\|N\|_2 < 1, \quad \frac{\|L^{-1}\|_2 (\|Y\|_2 + \|YM\|_2)}{1 - \|N\|_2} < 1, \quad (4.1)$$

then  $(L(I_{m+n} + N))J_{m+n}(L(I_{m+n} + N))^T - (Y(I_{m+n} + M))(Y(I_{m+n} + M))^T$  has the generalized Cholesky factorization

$$\begin{aligned} &(V + \Delta V)J_{m+n}(V + \Delta V)^T \\ &= (I_{m+n} + N)LJ_{m+n}L^T(I_{m+n} + N)^T - (I_{m+n} + M)YY^T(I_{m+n} + M)^T. \end{aligned} \quad (4.2)$$

*Proof.* For any  $|t| \leq \varepsilon$  from the first condition of (4.1), it follows that

$$\|tH\|_2 \leq \|\varepsilon H\|_2 = \|N\|_2 < 1. \quad (4.3)$$

Then  $L(I_{m+n} + tH)$  is nonsingular. Thus, we have

$$\begin{aligned} &(L(I_{m+n} + tH))J_{m+n}(L(I_{m+n} + tH))^T - (Y(I_{m+n} + tS))(Y(I_{m+n} + tS))^T \\ &= (L(I_{m+n} + tH))(J_{m+n} - ZZ^T)(L(I_{m+n} + tH))^T, \end{aligned}$$

where  $Z = (L(I_{m+n} + tH))^{-1}(Y(I_{m+n} + tS)) = (I_{m+n} + tH)^{-1}L^{-1}(Y(I_{m+n} + tS))$ . From (4.3) and noting (2.1), we get

$$\left\| (I_{m+n} + tH)^{-1} \right\|_2 \leq \frac{1}{1 - \|tH\|_2}.$$

Furthermore,

$$\|Z\|_2 \leq \frac{\|L^{-1}\|_2 (\|Y\|_2 + \|tSY\|_2)}{1 - \|tH\|_2} \leq \mu < 1. \tag{4.4}$$

Using Zeyl’s theorem on eigenvalues of Nermitian matrices (e.g., [25], pp. 181), for the  $i$ -th eigenvalue of  $J_{m+n} - ZZ^T$ ,

$$\lambda_i(J_{m+n}) + \lambda_{\min}(-ZZ^T) \leq \lambda_i(J_{m+n} - ZZ^T) \leq \lambda_i(J_{m+n}) + \lambda_{\max}(-ZZ^T).$$

Note that (4.4) implies  $\lambda_{\min}(-ZZ^T) \geq -\mu^2$ . Assume the eigenvalues of a matrix be ordered in non-decreasing order. Thus, for  $0 < i \leq n$ ,

$$-1 - \mu^2 \leq \lambda_i(J_{m+n} - ZZ^T) \leq -1,$$

and for  $n < i \leq m+n$ ,

$$1 - \mu^2 \leq \lambda_i(J_{m+n} - ZZ^T) \leq 1.$$

That is,  $J_{m+n} - ZZ^T$  is nonsingular and has  $m$  positive eigenvalues and  $n$  negative eigenvalues, so is  $(L(I_{m+n} + tH))J_{m+n}(L(I_{m+n} + tH))^T - (Y(I_{m+n} + tS))(Y(I_{m+n} + tS))^T$ . In addition, we obtain

$$(L(I_{m+n} + tH))J_{m+n}(L(I_{m+n} + tH))^T - (Y(I_{m+n} + tS))(Y(I_{m+n} + tS))^T = \begin{bmatrix} G_{11} & G_{21}^T \\ G_{21} & -G_{22} \end{bmatrix},$$

where

$$\begin{aligned} G_{11} &= (L_{11}(I_m + tH_{11}))(L_{11}(I_m + tH_{11}))^T - (Y_m(I_m + tS_m))(Y_m(I_m + tS_m))^T \\ G_{22} &= (L_{22}(I_n + tH_{22}))(L_{22}(I_n + tH_{22}))^T - (L_{21}(I_n + tH_{21}))(L_{21}(I_n + tH_{21}))^T \\ &\quad + (Y_n(I_n + tS_n))(Y_n(I_n + tS_n))^T. \end{aligned}$$

Note that

$$tH = \begin{bmatrix} tH_{11} & 0 \\ * & * \end{bmatrix}. \tag{4.5}$$

Then  $\|tH_{11}\|_2 \leq \|tH\|_2 < 1$ , which implies that  $(L_{11}(I_m + tH_{11}))$  is nonsingular. Thus,  $G_{11}$  can be rewritten as

$$G_{11} = (L_{11}(I_m + tH_{11}))(I_m - Z_m Z_m^T)(L_{11}(I_m + tH_{11}))^T,$$

where  $Z_m = (L_{11}(I_m + tH_{11}))^{-1}(Y_m(I_m + tS_m))$ . Similar to (4.5), we can note that  $Z_m$  is the principal submatrix with order  $m$  of  $Z$ . Then, using (4.4),  $\|Z_m\|_2 \leq \|Z\|_2 \leq \mu < 1$ . Accordingly,  $I_m - Z_m Z_m^T$  is symmetric positive definite, so is  $G_{11}$ . Furthermore,  $G_{22}$  is clearly symmetric positive semi-definite. Thus, for each  $|t| \leq \varepsilon$ ,  $(L(I_{m+n} + tH))J_{m+n}(L(I_{m+n} + tH))^T - (Y(I_{m+n} + tS))(Y(I_{m+n} + tS))^T$  has the generalized Cholesky factorization

$$\begin{aligned} V(t)J_{m+n}V^T(t) &= (L(I_{m+n} + tH))J_{m+n}(L(I_{m+n} + tH))^T \\ &\quad - (Y(I_{m+n} + tS))(Y(I_{m+n} + tS))^T. \end{aligned}$$

Note that  $V(0) = V$  and  $V(\varepsilon) = V + \Delta V$ . Then (4.2) holds.  $\square$

**THEOREM 4.2.** *Given a lower triangular matrix  $L \in \mathbb{R}^{(m+n) \times (m+n)}$  with positive diagonal elements and a matrix  $Y \in \mathbb{R}^{(m+n) \times k}$  such that the generalized Cholesky factorization  $VJ_{m+n}V^T = LJ_{m+n}L^T - YY^T$  holds. Suppose  $W = I_{m+n} + N \in \mathbb{R}^{(m+n) \times (m+n)}$  and  $U = I_{m+n} + M \in \mathbb{R}^{(m+n) \times (m+n)}$ . If*

$$\|N\|_2 < 1, \quad \frac{\|L^{-1}\|_2 (\|Y\|_2 + \|YM\|_2)}{1 - \|N\|_2} < 1, \quad (4.6)$$

and

$$\begin{aligned} & \left( \|(LJ_{m+n}L^T)V^{-T}\|_2 \|V^{-1}N\|_F + \|(YY^T)V^{-T}\|_2 \|V^{-1}M\|_F \right. \\ & \quad \left. + \|V^{-1}NL\|_F^2 + \|V^{-1}MY\|_F^2 \right) < \frac{1}{2}, \end{aligned} \quad (4.7)$$

then the generalized Cholesky factorization (4.2) always exists and

$$\begin{aligned} \|\Delta V\|_F & \leq (2 + \sqrt{2}) \inf_{D_{m+n} \in \mathbb{D}_{m+n}} \left( \sqrt{1 + \zeta_D^2 \kappa(VD_{m+n}^{-1})} \right) \|V^{-1}\|_2 (\|L\|_2^2 \|N\|_F + \|Y\|_2^2 \|M\|_F) \\ & \quad + (2 + \sqrt{2}) \inf_{D_{m+n} \in \mathbb{D}_{m+n}} (\kappa(VD_{m+n}^{-1})) \|V^{-1}\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2). \end{aligned} \quad (4.8)$$

*Proof.* Let  $W(t) = I_{m+n} + tN$  and  $U(t) = I_{m+n} + tM$ , for any  $|t| \leq \varepsilon$ . Using (1.3) leads to

$$\begin{aligned} & (L(I_{m+n} + tN))J_{m+n}(L(I_{m+n} + tN))^T - (Y(I_{m+n} + tM))(Y(I_{m+n} + tM))^T \\ & = (L(I_{m+n} + tN))(J_{m+n} - (L(I_{m+n} + tN))^{-1} \left( (Y(I_{m+n} + tM))(Y(I_{m+n} + tM))^T \right) \\ & \quad \times (L(I_{m+n} + tN))^{-T}) (L(I_{m+n} + tN))^T, \end{aligned}$$

and

$$\left\| (L(I_{m+n} + tN))^{-1} (Y(I_{m+n} + tM)) \right\|_2 \leq \frac{\|L^{-1}\|_2 (\|Y\|_2 + \|YM\|_2)}{1 - \|N\|_2},$$

and considering the results of Theorem 4.1, it follows that (4.6) holds,

$$(L(I_{m+n} + tN))J_{m+n}(L(I_{m+n} + tN))^T - (Y(I_{m+n} + tM))(Y(I_{m+n} + tM))^T$$

is positive definite and has the unique generalized Cholesky factorization, i.e.

$$\begin{aligned} & (V + \Delta V(t))J_{m+n}(V + \Delta V(t))^T \\ & = (L(I_{m+n} + tN))J_{m+n}(L(I_{m+n} + tN))^T - (Y(I_{m+n} + tM))(Y(I_{m+n} + tM))^T, \end{aligned} \quad (4.9)$$

which, with  $\Delta V(0) = 0$  and  $\Delta V(\varepsilon) = \Delta V$ , implies (4.2). Using (4.9) and (1.3) leads to

$$\begin{aligned} & (tN(LJ_{m+n}L^T) + t(LJ_{m+n}L^T)N^T + t^2N(LJ_{m+n}L^T)N^T) - (tM(YY^T) \\ & \quad + tM(YY^T)M^T + t^2M(YY^T)M^T) \\ & = VJ_{m+n}(\Delta V(t))^T + \Delta V(t)J_{m+n}V^T + \Delta V(t)J_{m+n}(\Delta V(t))^T. \end{aligned}$$

Pre-multiplying the above equation by  $V^{-1}$  and post-multiplying by  $V^{-T}$  lead to

$$\begin{aligned}
 & J_{m+n}(\Delta V(t))^T V^{-T} + V^{-1} \Delta V(t) J_{m+n} \\
 &= tV^{-1}N(LJ_{m+n}L^T)V^{-T} + tV^{-1}(LJ_{m+n}L^T)N^T V^{-T} + t^2V^{-1}N(LJ_{m+n}L^T)N^T V^{-T} \\
 &\quad - \left( tV^{-1}M(YY^T)V^{-T} + tV^{-1}(YY^T)M^T V^{-T} + t^2V^{-1}M(YY^T)M^T V^{-T} \right) \\
 &\quad - V^{-1} \Delta V(t) J_{m+n} (\Delta V(t))^T V^{-T}.
 \end{aligned} \tag{4.10}$$

Since  $V^{-1} \Delta V(t)$  is a lower triangular matrix. Using the symbol ‘low’ and noting (4.10), we get

$$\begin{aligned}
 V^{-1} \Delta V(t) J_{m+n} &= \text{low} \left( tV^{-1}N(LJ_{m+n}L^T)V^{-T} + tV^{-1}(LJ_{m+n}L^T)N^T V^{-T} \right) \\
 &\quad + \text{low} \left( t^2V^{-1}N(LJ_{m+n}L^T)N^T V^{-T} \right) \\
 &\quad - \text{low} \left( tV^{-1}M(YY^T)V^{-T} + tV^{-1}(YY^T)M^T V^{-T} \right) \\
 &\quad - \text{low} \left( t^2V^{-1}M(YY^T)M^T V^{-T} \right) \\
 &\quad - \text{low} \left( V^{-1} \Delta V(t) J_{m+n} (\Delta V(t))^T V^{-T} \right).
 \end{aligned} \tag{4.11}$$

Applying the Frobenius norm to (4.11) and considering (2.1), (2.7) and (2.8) yields

$$\begin{aligned}
 & \|V^{-1} \Delta V(t) J_{m+n}\|_F \\
 &\leq \sqrt{2} \left( \|tV^{-1}N(LJ_{m+n}L^T)V^{-T}\|_F + \|tV^{-1}M(YY^T)V^{-T}\|_F \right) \\
 &\quad + \frac{1}{\sqrt{2}} \| (V^{-1} \Delta V(t) J_{m+n} (\Delta V(t))^T V^{-T} ) \|_F \\
 &\quad + \frac{1}{\sqrt{2}} \left( \|t^2V^{-1}N(LJ_{m+n}L^T)N^T V^{-T}\|_F + \|t^2V^{-1}M(YY^T)M^T V^{-T}\|_F \right) \\
 &\leq \sqrt{2} \left( \|tV^{-1}N\|_F \| (LJ_{m+n}L^T) V^{-T} \|_2 + \|tV^{-1}M\|_2 \| (YY^T) V^{-T} \|_F \right) \\
 &\quad + \frac{1}{\sqrt{2}} \left( \|V^{-1} \Delta V(t)\|_F^2 + \|tV^{-1}NL\|_F^2 + \|tV^{-1}MY\|_F^2 \right).
 \end{aligned} \tag{4.12}$$

Assume  $p(t) = \|V^{-1} \Delta V(t) J_{m+n}\|_F$  and

$$\begin{aligned}
 q(t) &= 2 \left( \|tV^{-1}N\|_F \| (LJ_{m+n}L^T) V^{-T} \|_2 + \|tV^{-1}M\|_2 \| (YY^T) V^{-T} \|_F \right) \\
 &\quad + \left( \|tV^{-1}NL\|_F^2 + \|tV^{-1}MY\|_F^2 \right).
 \end{aligned}$$

Then, the above equation can be rewritten as

$$p(t)^2 - \sqrt{2}p(t) + q(t) \geq 0.$$

Equation (4.7), clearly shows that  $q(t) < 1/2$ , for  $|t| \leq \varepsilon$ . Therefore,  $p(t) \leq p_1(t)$  or  $p(t) \geq p_2(t)$ , where

$$p_1(t) = \frac{1}{\sqrt{2}}(1 - \sqrt{1 - 2q(t)}) < p_2(t) = \frac{1}{\sqrt{2}}(1 + \sqrt{1 - 2q(t)}). \quad (4.13)$$

Note that  $p(t)$  is continuous and  $p(0) = 0 = p_1(0) < p_2(0) = \sqrt{2}$ . Then  $p(t) \leq p_1(t)$ , for any  $|t| \leq \varepsilon$ . Consequently,  $p(\varepsilon) \leq p_1(\varepsilon)$ , i.e.

$$\|V^{-1}\Delta V(t)J_{m+n}\|_F \leq \frac{1}{\sqrt{2}}(1 - \sqrt{1 - 2q(t)}) < \frac{1}{\sqrt{2}}. \quad (4.14)$$

Putting  $t = \varepsilon$  in (4.11)

$$\begin{aligned} V^{-1}\Delta VJ_{m+n} = & \text{low} \left( V^{-1}N(LJ_{m+n}L^T)V^{-T} + V^{-1}(LJ_{m+n}L^T)N^TV^{-T} \right) \\ & + \text{low} \left( V^{-1}N(LJ_{m+n}L^T)N^TV^{-T} \right) - \text{low} \left( V^{-1}M(YY^T)M^TV^{-T} \right) \\ & - \text{low} \left( V^{-1}M(YY^T)V^{-T} + V^{-1}(YY^T)M^TV^{-T} \right) \\ & - \text{low} \left( V^{-1}\Delta VJ_{m+n}(\Delta V)^TV^{-T} \right). \end{aligned} \quad (4.15)$$

Right multiplying by  $D_{m+n} \in \mathbb{D}_{m+n}$ , and noting  $V = D_{m+n}\check{V}$  and (2.4) yield

$$\begin{aligned} \check{V}^{-1}\Delta VJ_{m+n} = & \text{low} \left( \check{V}^{-1}N(LJ_{m+n}L^T)V^{-T} + D_{m+n}(\check{V}^{-1}N(LJ_{m+n}L^T)V^{-T})^TD_{m+n}^{-1} \right) \\ & + \text{low} \left( \check{V}^{-1}N(LJ_{m+n}L^T)N^TV^{-T} \right) - \text{low} \left( \check{V}^{-1}M(YY^T)M^TV^{-T} \right) \\ & - \text{low} \left( \check{V}^{-1}M(YY^T)V^{-T} + D_{m+n}(\check{V}^{-1}M(YY^T)V^{-T})^TD_{m+n}^{-1} \right) \\ & - \text{low} \left( \check{V}^{-1}\Delta VJ_{m+n}(\Delta V)^TV^{-T} \right). \end{aligned} \quad (4.16)$$

Implementing the Frobenius norm on (4.16) and utilizing (2.5), (2.6), and (2.1) lead to

$$\begin{aligned} \|\check{V}^{-1}\Delta VJ_{m+n}\|_F & \leq \sqrt{1 + \zeta_{D_{m+n}}^2} (\|\check{V}^{-1}N(LJ_{m+n}L^T)V^{-T}\|_F \\ & \quad + \|\check{V}^{-1}M(YY^T)V^{-T}\|_F) \\ & \quad + \|\check{V}^{-1}N(LJ_{m+n}L^T)N^TV^{-T}\|_F + \|\check{V}^{-1}M(YY^T)M^TV^{-T}\|_F \\ & \quad + \|\check{V}^{-1}\Delta VJ_{m+n}(\Delta V)^TV^{-T}\|_F \\ & \leq \sqrt{1 + \zeta_{D_{m+n}}^2} \|\check{V}^{-1}\|_2 \|V^{-1}\|_2 (\|L\|_2^2\|N\|_F + \|Y\|_2^2\|M\|_F) \\ & \quad + \|\check{V}^{-1}\Delta V\|_2 \|V^{-1}\Delta V\|_F \\ & \quad + \|\check{V}^{-1}\|_2 \|V^{-1}\|_2 (\|L\|_2^2\|N\|_F^2 + \|Y\|_2^2\|M\|_F^2). \end{aligned} \quad (4.17)$$

Considering (4.14), (2.1), (4.6) and (4.7), we obtain

$$\begin{aligned} \|\check{V}^{-1}\Delta V J_{m+n}\|_F &\leq (2 + \sqrt{2})\sqrt{1 + \zeta_D^2} \|\check{V}^{-1}\|_2 \|V^{-1}\|_2 (\|L\|_2^2 \|N\|_F + \|Y\|_2^2 \|M\|_F) \\ &\quad + (2 + \sqrt{2}) \|\check{V}^{-1}\|_2 \|V^{-1}\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2), \end{aligned} \quad (4.18)$$

which, along with the fact that  $\|J_{m+n}\|_2 = 1$  and the supporting proof

$$\|\Delta V\|_F \leq \|\check{V}\check{V}^{-1}\Delta V J_{m+n}\|_F \leq \|\check{V}\|_2 \|\check{V}^{-1}\Delta V\|_F \|J_{m+n}\|_2, \quad \text{by (2.1)} \quad (4.19)$$

shows the bound (4.8).  $\square$

To check the efficiency of bounds, let us prove that the bound (3.15) is significantly sharper than (4.8). Taking into consideration [26, Corollary 3.4], for any  $D_{m+n} \in \mathbb{D}_{m+n}$  and  $X \in \mathbb{R}^{(m+n) \times (m+n)}$ , using (2.11) and (2.12), we obtain

$$\begin{aligned} \|QLM\|_2 &= \left\| \mathfrak{K}_{\text{Ivec}}(J_{m+n}^{-T} \otimes V)(I_{m+n} \otimes D_{m+n}^{-1})(I_{m+n} \otimes D_{m+n}) \right. \\ &\quad \times \left. \mathfrak{K}_{\text{low}}\left((V^{-1}(LJ_{m+n}L^T) \otimes V^{-1}) + (V^{-1} \otimes V^{-1}(LJ_{m+n}L^T))\Pi\right) \right\|_2 \\ &= \left\| \mathfrak{K}_{\text{Ivec}}(J_{m+n} \otimes VD_{m+n}^{-1}) \mathfrak{K}_{\text{low}}\left((V^{-1}(LJ_{m+n}L^T) \otimes D_{m+n}V^{-1}) \right. \right. \\ &\quad \left. \left. + (V^{-1} \otimes D_{m+n}V^{-1}(LJ_{m+n}L^T))\Pi\right) \right\|_2 \quad \text{by (2.12)} \\ &\leq \|VD_{m+n}^{-1}\|_2 \left\| \mathfrak{K}_{\text{low}}\left((V^{-1}(LJ_{m+n}L^T) \otimes D_{m+n}V^{-1}) \right. \right. \\ &\quad \left. \left. + (V^{-1} \otimes D_{m+n}V^{-1}(LJ_{m+n}L^T))\Pi\right) \right\|_2 \quad \text{by (2.11)} \\ &= \|VD_{m+n}^{-1}\|_2 \max_{\|\text{vec}(X)\|_2=1} \left\| \mathfrak{K}_{\text{low}}\left((V^{-1}(LJ_{m+n}L^T) \otimes D_{m+n}V^{-1}) \right. \right. \\ &\quad \left. \left. + (V^{-1} \otimes D_{m+n}V^{-1}(LJ_{m+n}L^T))\Pi\right) \text{vec}(X) \right\|_2. \end{aligned} \quad (4.20)$$

Taking into account (2.9), (2.2), (2.4), (2.5) and (2.1) yields

$$\begin{aligned} &\max_{\|\text{vec}(X)\|_2=1} \left\| \mathfrak{K}_{\text{low}}\left((V^{-1}(LJ_{m+n}L^T) \otimes D_{m+n}V^{-1}) \right. \right. \\ &\quad \left. \left. + (V^{-1} \otimes D_{m+n}V^{-1}(LJ_{m+n}L^T))\Pi\right) \text{vec}(X) \right\|_2 \\ &= \max_{\|\text{vec}(X)\|_2=1} \left\| \mathfrak{K}_{\text{low}} \text{vec}\left(D_{m+n}V^{-1}X(LJ_{m+n}L^T)V^{-T} \right. \right. \\ &\quad \left. \left. + D_{m+n}V^{-1}(LJ_{m+n}L^T)X^T V^{-T}\right) \right\|_2 \quad \text{by (2.9)} \end{aligned}$$

$$\begin{aligned}
&= \max_{\|\text{vec}(X)\|_2=1} \left\| \text{vec} \left( \text{low} \left( D_{m+n} V^{-1} X (LJ_{m+n} L^T) V^{-T} \right. \right. \right. \\
&\quad \left. \left. \left. + D_{m+n} (D_{m+n} V^{-1} X (LJ_{m+n} L^T) V^{-T})^T D_{m+n}^{-1} \right) \right) \right\|_2 \quad \text{by (2.2)} \\
&= \max_{\|X\|_F=1} \left\| \text{low} \left( D_{m+n} V^{-1} X (LJ_{m+n} L^T) V^{-T} \right) \right. \\
&\quad \left. + D_{m+n} \text{low} \left( D_{m+n} V^{-1} X (LJ_{m+n} L^T) V^{-T} \right)^T D_{m+n}^{-1} \right\|_F \quad \text{by (2.4)} \\
&\leq \max_{\|X\|_F=1} \sqrt{1 + \zeta_{D_{m+n}}^2} \|D_{m+n} V^{-1} X (LJ_{m+n} L^T) V^{-T}\|_F \quad \text{by (2.5)} \\
&\leq \sqrt{1 + \zeta_{D_{m+n}}^2} \|D_{m+n} V^{-1}\|_2 \|(LJ_{m+n} L^T) V^{-T}\|_2. \quad \text{by (2.1)} \tag{4.21}
\end{aligned}$$

Hence, putting (4.21) into (4.20) gives

$$\|Q_{LM}\|_2 \leq \left( \inf_{D_{m+n} \in \mathbb{D}_{m+n}} \sqrt{1 + \zeta_{D_{m+n}}^2} \kappa(VD_{m+n}^{-1}) \right) \|L\|_2^2 \|V^{-1}\|_2. \tag{4.22}$$

Therefore, we can demonstrate that,

$$\|Q_Y\|_2 \leq \left( \inf_{D_{m+n} \in \mathbb{D}_{m+n}} \sqrt{1 + \zeta_{D_{m+n}}^2} \kappa(VD_{m+n}^{-1}) \right) \|Y\|_2^2 \|V^{-1}\|_2, \tag{4.23}$$

$$\|R_V\|_2 \leq \left( \inf_{D_{m+n} \in \mathbb{D}_{m+n}} \kappa(VD_{m+n}^{-1}) \right) \|V^{-1}\|_2, \tag{4.24}$$

(4.24) together with the fact  $\kappa(VD_{m+n}^{-1}) \geq 1$  and (4.22) indicates that the bound (3.15) is absolutely tighter than (4.8).

REMARK 4.3. By ignoring the high-order terms, we can obtain a first-order multiplicative bound from (4.8):

$$\|\Delta V\|_F \leq \inf_{D_{m+n} \in \mathbb{D}_{m+n}} \left( \sqrt{1 + \zeta_{D_{m+n}}^2} \kappa(VD_{m+n}^{-1}) \right) \|V^{-1}\|_2 (\|L\|_2^2 \|N\|_F + \|Y\|_2^2 \|M\|_F). \tag{4.25}$$

Obviously, from (4.22) and (4.23), we can check that the bound in Remark 3.2 is always tighter than the (4.25).

THEOREM 4.4. *With the same assumptions as in Theorem 4.1, we have*

$$\begin{aligned}
\|\Delta V\|_F &\leq (2 + \sqrt{2}) \inf_{D_{m+n} \in \mathbb{D}_{m+n}} \left( \sqrt{1 + \zeta_{D_{m+n}}^2} \kappa(VD_{m+n}^{-1}) \right) \\
&\quad \times \left( \|\text{sut}((LJ_{m+n} L^T) V^{-T})\|_2 \|N\|_F + \|V^{-1}\|_2 \|Y\|_2^2 \|M\|_F \right) \\
&\quad + (2 + \sqrt{2}) \inf_{D_{m+n} \in \mathbb{D}_{m+n}} \left( \kappa(VD_{m+n}^{-1}) \right) \\
&\quad \times \left( \|\text{diag}((LJ_{m+n} L^T) V^{-T})\|_2 \|N\|_F + \|V^{-1}\|_2 (\|L\|_2^2 \|N\|_F^2 + \|Y\|_2^2 \|M\|_F^2) \right). \tag{4.26}
\end{aligned}$$

*Proof.* Applying the symbols 'sut', 'diag' and 'slt',  $\text{low}(V^{-1}N(LJ_{m+n}L^T)V^{-T} + V^{-1}(LJ_{m+n}L^T)N^TV^{-T})$  can be rewritten as

$$\text{low}(V^{-1}N(LJ_{m+n}L^T)V^{-T} + V^{-1}(LJ_{m+n}L^T)N^TV^{-T}) \tag{4.27}$$

$$\begin{aligned} &= V^{-1}N \text{diag}((LJ_{m+n}L^T)V^{-T}) + \text{low}(V^{-1}N \text{sut}((LJ_{m+n}L^T)V^{-T}) \\ &\quad + \text{slt}(V^{-1}(LJ_{m+n}L^T))N^TV^{-T}). \end{aligned} \tag{4.28}$$

Substituting (4.27) into (4.15) and then pre-multiplying it by  $D_{m+n} \in \mathbb{D}_{m+n}$  and using  $V = D_{m+n}\check{V}$ , we have

$$\begin{aligned} &\check{V}^{-1}\Delta V J_{m+n} \\ &= V^{-1}N \text{diag}((LJ_{m+n}L^T)V^{-T}) \\ &\quad + \text{low}\left(V^{-1}N \text{sut}((LJ_{m+n}L^T)V^{-T}) + D_{m+n}(\text{slt}(V^{-1}(LJ_{m+n}L^T))N^TV^{-T})D_{m+n}^{-1}\right) \\ &\quad + \text{low}\left(\check{V}^{-1}N(LJ_{m+n}L^T)N^TV^{-T}\right) - \text{low}\left(\check{V}^{-1}M(YY^T)M^TV^{-T}\right) \\ &\quad - \text{low}\left(\check{V}^{-1}M(YY^T)V^{-T} + D_{m+n}(\check{V}^{-1}M(YY^T)V^{-T})^TD_{m+n}^{-1}\right) \\ &\quad - \text{low}\left(\check{V}^{-1}\Delta V J_{m+n}(\Delta V)^TV^{-T}\right). \end{aligned} \tag{4.29}$$

Hence, we get the bound (4.26) by applying the Frobenius norm to (4.29) and utilizing (2.5),(2.6), (4.14), and (4.19).

Moving forward, we will show that the bound (3.21) is tighter than (4.26). By using (3.18), we get

$$\begin{aligned} \|Q_{LT}\|_2 &\leq \left\| (J_{m+n}^{-T} \otimes V) (I_{m+n} \otimes D_{m+n}^{-1}) (I_{m+n} \otimes D_{m+n}) \mathfrak{K}_{\text{low}} \left( (V^{-1}(LJ_{m+n}L^T) \otimes V^{-1}) \right. \right. \\ &\quad \left. \left. + (V^{-1} \otimes V^{-1}(LJ_{m+n}L^T)) \Pi \right) \mathfrak{K}_{\text{lvec}}^T \right\| \\ &= \left\| (J_{m+n}^{-T} \otimes VD_{m+n}^{-1}) \mathfrak{K}_{\text{low}} \left( (V^{-1}(LJ_{m+n}L^T) \otimes D_{m+n}V^{-1}) \right. \right. \\ &\quad \left. \left. + (V^{-1} \otimes D_{m+n}V^{-1}(LJ_{m+n}L^T)) \Pi \right) \mathfrak{K}_{\text{lvec}}^T \right\| \quad \text{by (2.12)} \\ &\leq \|VD_{m+n}^{-1}\|_2 \max_{\|\text{lvec}(X)\|_2=1} \left\| \mathfrak{K}_{\text{low}} \left( (V^{-1}(LJ_{m+n}L^T) \otimes D_{m+n}V^{-1}) \right. \right. \\ &\quad \left. \left. + (V^{-1} \otimes D_{m+n}V^{-1}(LJ_{m+n}L^T)) \Pi \right) \mathfrak{K}_{\text{lvec}}^T \text{lvec}(X) \right\|_2. \end{aligned} \tag{4.30}$$



We assume  $X$  is a lower triangular matrix. Obviously, we only take the lower triangular part of  $X$ . As a result of using (2.2), (2.3), and (2.9), we get

$$\begin{aligned}
& \max_{\|\text{lvec}(X)\|_2=1} \left\| \mathfrak{K}_{\text{low}} \left( (V^{-1}(LJ_{m+n}L^T) \otimes D_{m+n}V^{-1}) \right. \right. \\
& \quad \left. \left. + (V^{-1} \otimes D_{m+n}V^{-1}(LJ_{m+n}L^T)) \Pi \right) \mathfrak{K}_{\text{lvec}}^T \mathfrak{K}_{\text{lvec}} \text{vec}(X) \right\|_2 \\
&= \max_{\|\text{lvec}(X)\|_2=1} \left\| \mathfrak{K}_{\text{low}} \left( (V^{-1}(LJ_{m+n}L^T) \otimes D_{m+n}V^{-1}) \right. \right. \\
& \quad \left. \left. + (V^{-1} \otimes D_{m+n}V^{-1}(LJ_{m+n}L^T)) \Pi \right) \mathfrak{K}_{\text{lt}} \text{vec}(X) \right\|_2 \quad \text{by (2.3)} \\
&= \max_{\|\text{lvec}(X)\|_2=1} \left\| \mathfrak{K}_{\text{low}} \left( (V^{-1}(LJ_{m+n}L^T) \otimes D_{m+n}V^{-1}) \right. \right. \\
& \quad \left. \left. + (V^{-1} \otimes D_{m+n}V^{-1}(LJ_{m+n}L^T)) \Pi \right) \text{vec}(\text{lt}(X)) \right\|_2 \quad \text{by (2.2)} \\
&= \max_{\|\text{lvec}(X)\|_2=1} \left\| \mathfrak{K}_{\text{low}} \left( (V^{-1}(LJ_{m+n}L^T) \otimes D_{m+n}V^{-1}) \right. \right. \\
& \quad \left. \left. + (V^{-1} \otimes D_{m+n}V^{-1}(LJ_{m+n}L^T)) \Pi \right) \text{vec}(X) \right\|_2 \\
&= \max_{\|\text{lvec}(X)\|_2=1} \left\| \mathfrak{K}_{\text{low}} \text{vec} \left( D_{m+n}V^{-1}X (V^{-1}(LJ_{m+n}L^T))^T \right. \right. \\
& \quad \left. \left. + D_{m+n}V^{-1}(LJ_{m+n}L^T)X^T V^{-T} \right) \right\|_2 \quad \text{by (2.9)} \\
&= \max_{\|\text{lvec}(X)\|_2=1} \left\| \text{vec} \left( \text{low} \left( D_{m+n}V^{-1}X (V^{-1}(LJ_{m+n}L^T))^T \right. \right. \right. \\
& \quad \left. \left. + D_{m+n}V^{-1}(LJ_{m+n}L^T)X^T V^{-T} \right) \right\|_2 \quad \text{by (2.2)} \\
&= \max_{\|\text{lvec}(X)\|_2=1} \left\| \text{vec} \left( \text{low} \left( D_{m+n}V^{-1}X \left( \text{sut} (V^{-1}(LJ_{m+n}L^T))^T \right. \right. \right. \right. \\
& \quad \left. \left. + D_{m+n} \left( \text{slt} (V^{-1}(LJ_{m+n}L^T)) + \text{diag} (V^{-1}(LJ_{m+n}L^T))^T \right) \right. \right. \\
& \quad \left. \left. \left. + \text{diag} (V^{-1}(LJ_{m+n}L^T)) \right) X^T V^{-T} \right) \right\|_2. \tag{4.31}
\end{aligned}$$

For any lower triangular matrix  $K$ , we have

$$\text{low}(K) + \text{low}(K^T) = K.$$

Therefore, if we set  $K \equiv D_{m+n}V^{-1}X \text{diag}((LJ_{m+n}L^T)V^{-T})$ , then

$$\begin{aligned}
& D_{m+n}V^{-1}X \text{diag}((LJ_{m+n}L^T)V^{-T}) \\
&= \text{low} \left( D_{m+n}V^{-1}X \text{diag} (V^{-1}(LJ_{m+n}L^T))^T + D_{m+n} \text{diag} (V^{-1}(LJ_{m+n}L^T)) X^T V^{-T} \right),
\end{aligned}$$

which together with (4.31) and (4.30) implies

$$\begin{aligned}
\|Q_{LT}\|_2 &\leq \|VD_{m+n}^{-1}\|_2 \max_{\|\text{vec}(X)\|_2=1} \left\| \text{vec} \left( D_{m+n} V^{-1} X \text{diag} \left( (LJ_{m+n} L^T) V^{-T} \right) \right. \right. \\
&\quad \left. \left. + \text{low} \left( D_{m+n} V^{-1} X \left( \text{sut} \left( V^{-1} (LJ_{m+n} L^T) \right) \right)^T \right) \right. \right. \\
&\quad \left. \left. + D_{m+n} \left( \text{slt} \left( V^{-1} (LJ_{m+n} L^T) \right) \right) X^T V^{-T} \right) \right\|_2 \\
&\leq \|VD_{m+n}^{-1}\|_2 \max_{\|X\|_F=1} \|D_{m+n} V^{-1} X \text{diag} \left( (LJ_{m+n} L^T) V^{-T} \right)\|_F \\
&\quad + \|VD_{m+n}^{-1}\|_2 \max_{\|X\|_F=1} \left\| \text{low} \left( D_{m+n} V^{-1} X \left( \text{sut} \left( V^{-1} (LJ_{m+n} L^T) \right) \right)^T \right) \right. \\
&\quad \left. + D_{m+n} \left( \text{slt} \left( V^{-1} (LJ_{m+n} L^T) \right) \right) X^T V^{-T} \right\|_F.
\end{aligned}$$

With the help of (2.5), we get

$$\begin{aligned}
\|Q_{LT}\|_2 &\leq \kappa(VD_{m+n}^{-1}) \left\| \text{diag} \left( (LJ_{m+n} L^T) V^{-T} \right) \right\|_2 \\
&\quad + \max_{\|X\|_F=1} \left\| \text{low} \left( D_{m+n} V^{-1} X \left( \text{sut} \left( (LJ_{m+n} L^T) V^{-T} \right) \right) \right) \right. \\
&\quad \left. + D_{m+n} \left( \text{low} \left( D_{m+n} V^{-1} X \left( \text{sut} \left( (LJ_{m+n} L^T) V^{-T} \right) \right) \right) \right)^T D_{m+n}^{-1} \right\|_F \\
&\leq \kappa(VD_{m+n}^{-1}) \left\| \text{diag} \left( (LJ_{m+n} L^T) V^{-T} \right) \right\|_2 \\
&\quad + \sqrt{1 + \zeta_{D_{m+n}}^2} \|VD_{m+n}^{-1}\|_2 \max_{\|X\|_F=1} \left\| \left( D_{m+n} V^{-1} X \left( \text{sut} \left( (LJ_{m+n} L^T) V^{-T} \right) \right) \right) \right\|_F \\
&\leq \kappa(VD_{m+n}^{-1}) \left\| \text{diag} \left( (LJ_{m+n} L^T) V^{-T} \right) \right\|_2 \\
&\quad + \sqrt{1 + \zeta_{D_{m+n}}^2} \kappa(VD_{m+n}^{-1}) \left\| \left( \text{sut} \left( (LJ_{m+n} L^T) V^{-T} \right) \right) \right\|_2. \tag{4.32}
\end{aligned}$$

This result together with the fact  $\kappa(VD_{m+n}^{-1}) \geq 1$  and (4.23) illustrates that the bound (3.21) is significantly smaller than (4.26).  $\square$

REMARK 4.5. By neglecting the high-order terms, we can derive a first-order multiplicative bound from (4.26):

$$\begin{aligned}
\|\Delta V\|_F &\leq \inf_{D_{m+n} \in \mathbb{D}_{m+n}} \left( \sqrt{1 + \zeta_{D_{m+n}}^2} \kappa(VD_{m+n}^{-1}) \right) \left( \left\| \text{sut} \left( (LJ_{m+n} L^T) V^{-T} \right) \right\|_2 \|N\|_F \right. \\
&\quad \left. + \|V^{-1}\|_2 \|Y\|_2^2 \|M\|_F \right) \\
&\quad + \inf_{D_{m+n} \in \mathbb{D}_{m+n}} \left( \kappa(VD_{m+n}^{-1}) \right) \left( \left\| \text{diag} \left( (LJ_{m+n} L^T) V^{-T} \right) \right\|_2 \|N\|_F \right). \tag{4.33}
\end{aligned}$$

From (4.32) and (4.23), it is easy to check that the bound in Remark 3.4 is tighter than (4.33), when  $W = I_{m+n} + N$  is assumed to be a lower triangular matrix.

### 5. Numerical results

In this section, we provide three numerical examples to illustrate the results derived in Sections 3–4. We use an algorithm from [1] to obtain the generalized Cholesky factors  $L$  and  $V$  in (1.3). All numerical experiments are performed by using Matlab 2018a.

EXAMPLE 5.1. In first example, we compare the strong and weak multiplicative perturbation bounds. Let  $A = [a_{ij}] = P_m + I_m \in \mathbb{R}^{m \times m}$  where  $P_m = [p_{ij}]$  is the Pascal matrix, (i.e.,  $p_{1j} = p_{i1} = 1, p_{ij} = p_{i(j-1)} + p_{(i-1)j}$ ),  $B = [b_{ij}] = 0.7 * [\max(i, j)] \in \mathbb{R}^{n \times m}$  and  $C = (c_{ij}) \in \mathbb{R}^{n \times n}$  is the Lehmar matrix, (i.e.,  $c_{ij} = i/j$  for  $j \geq i$ ),  $J_{m+n} = \text{diag}(I_m, -I_n)$  and  $Y = P\Lambda P^T$ , where  $P \in \mathbb{R}^{(m+n) \times (m+n)}$  is an orthogonal matrix taken from the QR factorization,  $\Lambda = \text{diag}(d, d, \dots, d) \in \mathbb{R}^{(m+n) \times (m+n)}$  with  $d = 0.02$ . From [11], the scaled matrix  $D_{m+n}$  is denoted as follows: suppose  $\zeta_1 = \sqrt{\sum_{j=1}^{m+n} f_{1j}^2}$ ,  $\zeta_i = \sqrt{\sum_{j=i}^{m+n} f_{ij}^2}$  if  $\sqrt{\sum_{j=1}^{m+n} f_{ij}^2} \leq \zeta_{i-1}$  otherwise  $\zeta_i = \zeta_{i-1}$ , for  $i = 2, \dots, m+n$ ,  $F = (f_{ij}) = V^T$ .

Table 1: Results for various values of  $n, m$  and  $k$

$n, m, k$	5, 5, 10	10, 10, 20	13, 12, 25	18, 12, 30	20, 20, 40	40, 10, 50
$\ Q_{LM}\ _2$	4.1503	13.0342	26.3812	39.3403	52.5054	87.4032
$t_{\ Q_{LM}\ _2}$	0.011385	0.014349	0.039375	0.497504	2.567943	9.750072
$\gamma_{11}$	32.9031	190.1043	530.2156	6.5214e+02	2.5475e+03	5.6403e+03
$t_{\gamma_{11}}$	0.014312	0.015310	0.016385	0.018066	0.015910	0.019883
$\ Q_Y\ _2$	0.1832	0.2072	0.3702	0.5361	1.0945	1.7304
$t_{\ Q_Y\ _2}$	0.014291	0.021533	0.028610	0.169868	3.543310	13.918751
$\gamma_{12}$	0.4732	3.1903	9.0704	39.1204	67.0123	124.2617
$t_{\gamma_{12}}$	0.011736	0.014821	0.015570	0.016765	0.015092	0.019160
$\ Q_{LT}\ _2$	2.0145	3.1026	3.5017	3.7102	3.9143	4.9813
$t_{\ Q_{LT}\ _2}$	0.015054	0.019183	0.029361	0.182677	0.577617	3.769436
$\gamma_{13}$	8.0537	21.7241	33.2653	59.4293	83.3427	162.3076
$t_{\gamma_{13}}$	0.015381	0.017907	0.017463	0.015684	0.018650	0.019702
$\ R_V\ _2$	11.3821	26.4376	51.0436	132.3054	174.3076	294.2643
$t_{\ R_V\ _2}$	0.016202	0.015268	0.022487	0.014396	0.015743	0.017608
$\gamma_{14}$	23.0543	146.0354	397.0548	3.4256e+02	7.5473e+02	3.7354e+03
$t_{\gamma_{14}}$	0.014631	0.017183	0.016836	0.016179	0.015437	0.019326
$\ R_Y\ _2$	0.2537	0.3741	0.7451	1.5803	3.5071	4.9644
$t_{\ R_Y\ _2}$	0.015494	0.015406	0.013707	0.016474	0.014801	0.015409
$\gamma_{15}$	0.4322	1.3605	6.8065	31.7402	59.4223	95.7733
$t_{\gamma_{15}}$	0.014988	0.015179	0.014046	0.019076	0.016228	0.019694

Moreover, in Table 1, we denote

$$\gamma_{11} = \sqrt{1 + \zeta_{D_{m+n}}^2} \kappa(VD_{m+n}^{-1}) \|L\|_2 \|V^{-1}\|_2, \quad \gamma_{12} = \sqrt{1 + \zeta_{D_{m+n}}^2} \kappa(VD_{m+n}^{-1}) \|Y\|_2 \|V^{-1}\|_2,$$

$$\gamma_{13} = \kappa(VD_{m+n}^{-1}) \left( \|\text{diag}((LJ_{m+n}L^T)V^{-T})\|_2 + \sqrt{1 + \zeta_{D_{m+n}}^2} \|\text{slt}((LJ_{m+n}L^T)V^{-T})\|_2 \right),$$

$$\gamma_{14} = \kappa(VD_{m+n}^{-1}) \|L\|_2^2 \|V^{-1}\|_2, \quad \gamma_{15} = \kappa(VD_{m+n}^{-1}) \|Y\|_2^2 \|V^{-1}\|_2,$$

and  $t_{(\cdot)}$  denote the time cost in seconds for computing the bounds (3.15), (3.21), (4.8), (4.26).

From Table 1, we can see that the strong rigorous multiplicative perturbation bounds (3.15) and (3.22), the rows marked by  $\|Q_{LM}\|_2$ ,  $\|Q_Y\|_2$ ,  $\|Q_{LT}\|_2$ ,  $\|R_V\|_2 \|L\|_2^2$  and  $\|R_V\|_2 \|Y\|_2^2$  are always tighter than the weak rigorous multiplicative perturbation bounds (4.8) and (4.26), the rows marked by  $\gamma_{11}, \gamma_{12}, \gamma_{13}, \gamma_{14}$  and  $\gamma_{15}$ . In addition, we can also observe that it is indeed more expensive to estimate the bounds (3.15) and (3.22); compare the rows marked by  $t_{\gamma_{11}}, t_{\gamma_{12}}, t_{\gamma_{13}}, t_{\gamma_{14}}$  and  $t_{\gamma_{15}}$ .

EXAMPLE 5.2. The test matrix  $K_1$  and  $K_2$  are set to be

$$K_1 = \begin{bmatrix} 70 & 70 & 70 & 3 & 3 \\ 70 & 90 & 100 & 7 & 9 \\ 70 & 100 & 200 & 13 & 17 \\ 3 & 7 & 13 & -20 & -56 \\ 3 & 9 & 17 & -56 & -80 \end{bmatrix}, \quad K_2 = \begin{bmatrix} 2.4957 & -0.9922 & 0.9008 & -1.0195 & -0.3655 \\ -0.9922 & 2.4957 & -0.9922 & 0.9008 & -1.0195 \\ 0.9008 & -0.9922 & 2.4957 & -0.9922 & 0.9008 \\ -1.0195 & 0.9008 & -0.9922 & -2.4957 & 0.9922 \\ -0.3655 & -1.0195 & 0.9008 & 0.9922 & -2.4957 \end{bmatrix},$$

here  $Y = \mu [0.240 \ -0.899 \ 0.899 \ 1.560 \ -2.390]^T$  and  $J_{3+2} = \text{diag}(I_3, -I_2)$ .

Table 2: Results for various values of  $\mu$

$\omega$	Test matrices	$\mu_1$	$\mu_2$	$\mu_3$	$\mu_4$	$\mu_5$	$\mu_6$
$\ Q_{LM}\ _2$	$K_1$	1.2903	1.2903	1.2903	1.2904	1.2904	1.2904
	$K_2$	1.5214	1.5214	1.5214	1.5215	1.5215	1.5215
$\gamma_{11}$	$K_1$	1.2643e+02	1.2643e+02	1.2872e+02	1.7642e+02	1.7642e+02	1.7642e+02
	$K_2$	1.4453e+02	1.4453e+02	1.4467e+02	1.9362e+02	1.9362e+02	1.9362e+02
$\ Q_Y\ _2$	$K_1$	2.2632	2.2580	2.2531	0.0232	3.3227e-04	4.2827e-06
	$K_2$	2.5214	2.5112	2.5034	0.3204	5.3097e-04	7.3437e-06
$\gamma_{12}$	$K_1$	3.9126	3.9042	2.9959	1.4420	4.8028e-02	4.7028e-04
	$K_2$	5.8632	5.8436	5.6209	1.8062	9.9201e-02	8.7403e-03
$\ Q_{LT}\ _2$	$K_1$	1.1357	1.1357	1.1357	1.1357	1.1357	1.1357
	$K_2$	1.4792	1.4792	1.4792	1.4792	1.4792	1.4792
$\gamma_{13}$	$K_1$	76.1082	76.3042	76.6201	77.3321	77.4508	77.6902
	$K_2$	84.1376	84.4201	84.4853	85.0675	85.3065	85.3457
$\ R_V\ _2$	$K_1$	7.3214e+01	7.3214e+01	7.5326e+01	7.5422e+01	7.6402e+01	7.6402e+01
	$K_2$	9.3053e+01	9.3053e+01	9.4467e+01	9.4761e+01	9.9502e+01	9.9502e+01
$\gamma_{14}$	$K_1$	1.1643e+02	1.1643e+02	1.1872e+02	1.6642e+02	1.6642e+02	1.6642e+02
	$K_2$	1.3453e+02	1.3453e+02	1.3467e+02	1.8362e+02	1.8362e+02	1.8362e+02
$\ R_Y\ _2$	$K_1$	0.7861	0.7835	0.6328	0.3116	1.2092e-03	1.0453e-05
	$K_2$	0.9861	0.9835	0.8328	0.5116	1.1654e-02	1.1921e-04
$\gamma_{15}$	$K_1$	3.6126	3.6042	2.9059	1.0420	3.2028e-02	3.2028e-04
	$K_2$	5.5632	5.5436	5.3209	1.2062	5.9201e-02	5.7403e-03

Table 2 provides the numerical results for various values of  $\mu$ 's, as follows:

$$\begin{aligned} \mu_1 &= 1.004015006005433e-2, & \mu_2 &= 1.003021021209640e-2, \\ \mu_3 &= 9.036225416303058e-3, & \mu_4 &= \mu_3e-1, & \mu_5 &= \mu_3e-3, & \mu_6 &= \mu_3e-5. \end{aligned}$$

From Table 2, we can find that the strong multiplicative perturbation bounds (3.15) and (3.21) are always less than the weak multiplicative perturbation bounds (4.8) and (4.26), where  $\gamma_{11}, \gamma_{12}, \gamma_{13}, \gamma_{14}$  and  $\gamma_{15}$  are given in the previous example. Furthermore, Tables 1 and 2 show that when the multiplicative perturbation matrix  $W = I_{m+n} + N$  is set to be a lower triangular matrix, the strong and weak multiplicative perturbation bounds for the GCBD problem are smaller than the corresponding ones when the matrix  $W = I_{m+n} + N$  is set to be a general matrix.

EXAMPLE 5.3. Let

$$L = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ -e & 1 & 0 & 0 & 0 & 0 & 0 \\ -e & -e & 1 & 0 & 0 & 0 & 0 \\ -e & -e & -e & 1 & 0 & 0 & 0 \\ -e & -e & -e & -e & 1 & 0 & 0 \\ -e & -e & -e & -e & -e & 1 & 0 \\ -e & -e & -e & -e & -e & -e & 1 \end{bmatrix} \text{diag} \left( 1, \delta, \delta^2, \delta^3, \delta^4, \delta^5, \delta^6 \right),$$

$$Y = \mu \begin{bmatrix} 0.240 \\ -0.899 \\ 0.899 \\ 1.360 \\ -2.190 \\ 1.560 \\ 2.301 \end{bmatrix}, \quad J_{4+3} = \text{diag} (I_4, -I_3),$$

where  $e = 0.98$  and  $\delta = \sqrt{1 - e^2}$ . From [11, 12], we have the following strong and weak additive rigorous perturbation bounds respectively:

$$\begin{aligned} G_{LG} &= \mathfrak{K}_{\text{lvec}} (J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}} ((V^{-1} \otimes V^{-1} L J_{m+n}) \Pi + (V^{-1} L J_{m+n}^T \otimes V^{-1})), \\ G_Y &= \mathfrak{K}_{\text{lvec}} (J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}} ((V^{-1} \otimes V^{-1} Y) \Pi + (V^{-1} Y \otimes V^{-1})), \\ H_V &= \mathfrak{K}_{\text{lvec}} (J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}} (V^{-1} \otimes V^{-1}), \\ G_{LT} &= (\mathfrak{K}_{\text{lvec}} (J_{m+n}^{-T} \otimes V) \mathfrak{K}_{\text{low}} ((V^{-1} \otimes V^{-1} L J_{m+n}) \Pi + (V^{-1} L J_{m+n}^T \otimes V^{-1}))) \mathfrak{K}_{\text{lvec}}^T, \end{aligned}$$

and

$$\begin{aligned} \beta_{11} &= \sqrt{1 + \zeta_{D_{m+n}}^2} \kappa (V D_{m+n}^{-1}) \|V^{-1} L\|_2, \quad \beta_{12} = \sqrt{1 + \zeta_{D_{m+n}}^2} \kappa (V D_{m+n}^{-1}) \|V^{-1} Y\|_2, \\ \beta_{13} &= \kappa (V D_{m+n}^{-1}) \left( \|\text{diag} (V^{-1} L)\|_2 + \sqrt{1 + \zeta_{D_{m+n}}^2} \|\text{slt} (V^{-1} L)\|_2 \right), \\ \beta_{14} &= \kappa (V D_{m+n}^{-1}) \|V^{-1}\|_2. \end{aligned}$$

In this example, we compare the additive and multiplicative rigorous perturbation bounds for different values of  $\mu$ 's, which are given above. From Table 3, we can clearly observe that the strong and weak multiplicative bounds  $\|Q_{LM}\|_2, \|Q_Y\|_2, \|Q_{LT}\|_2, \|R_V\|_2$ , and  $\gamma_{11}, \gamma_{12}, \gamma_{13}, \gamma_{14}$  are tighter than the strong and weak additive perturbation

Table 3: Comparison of additive and multiplicative rigorous perturbation bounds

$\omega$	$\mu_1$	$\mu_2$	$\mu_3$	$\mu_4$	$\mu_5$	$\mu_6$
$\ Q_{LM}\ _2$	2.2997	2.2996	2.2983	2.2981	2.2981	2.2981
$\ G_{LG}\ _2$	12.6871	12.6871	12.6871	12.6871	12.6871	12.6871
$\ Q_Y\ _2$	21.4490	21.0468	9.0424	0.0592	5.4072e-06	5.9131e-10
$\ G_Y\ _2$	43.2302	43.6391	26.3015	1.8391	6.3261e-02	7.2604e-03
$\ Q_{LT}\ _2$	1.3628	1.3628	1.3628	1.3628	1.3628	1.3628
$\ G_{LT}\ _2$	2.5943	2.5943	2.5943	2.5943	2.5943	2.5943
$\ R_V\ _2$	46.3491	46.3491	48.3607	48.3607	57.5203	57.7287
$\ H_V\ _2$	89.6343	89.6343	92.7343	92.9807	98.8408	98.8626
$\beta_{11}$	8.0674e+02	8.2665e+02	8.5674e+02	8.5832e+02	8.7609e+02	8.9612e+02
$\beta_{12}$	1.4553e+03	1.4597e+03	1.6138e+03	1.6234e+03	1.9721e+03	1.9860e+03
$\beta_{13}$	46.5435	45.6645	19.4237	1.1456	1.2537e-03	1.2537e-06
$\beta_{14}$	1.4806e+03	1.4540e+03	686.5176	44.3787	0.4431	0.1434
$\gamma_{13}$	35.0578	35.0389	35.0104	17.1122	17.1307	17.1307
$\beta_{13}$	106.3003	104.3310	47.1390	37.2814	37.1612	37.1612
$\gamma_{14}$	5.3607e+02	5.3607e+02	5.4241e+02	5.4241e+02	5.7437e+02	5.8327e+02
$\beta_{14}$	1.2062e+03	1.2062e+03	1.2938e+03	1.2938e+03	1.6072e+03	1.6849e+03

bounds  $\|G_{LG}\|_2, \|G_Y\|_2, \|G_{LT}\|_2, \|H_V\|_2$ , and  $\beta_{11}, \beta_{12}, \beta_{13}, \beta_{14}$  respectively. The multiplicative perturbation bounds are useful to estimate the tighter bounds since the matrix is significantly ill-conditioned.

*Acknowledgements.* We acknowledge the principal editor and anonymous referees for their useful comments and constructive suggestions which helped considerably to improve the quality of the paper.

## REFERENCES

- [1] J. X. ZHAO, *The generalized Cholesky factorization method for saddle point problems*, Appl. Math. Comput. 92 (1998) 49–58.
- [2] X. M. FANG, *The generalized Cholesky factorization and perturbation of the real symmetric matrices*, J. Zhaoqing. Univ. 29 (2008) 14–16.
- [3] Å. BJÖRCK, H. PARK, L. ELDÉN, *Accurate downdating of least squares solutions*, SIAM J. Matrix Anal. Appl. 15 (1994) 549–568.
- [4] A. W. BOJANCZYK, R. P. BRENT, V. P. DOOREN, F. R. DE HOOG, *A note on downdating the Cholesky factorization*, SIAM J. Sci. Statist. Comput. 8 (1987) 210–221.
- [5] X. W. CHANG, C. C. PAIGE, *Perturbation analyses for the Cholesky downdating problem*, SIAM J. Matrix Anal. Appl. 19 (1998) 429–443.
- [6] L. ELDÉN, H. PARK, *Perturbation analysis for block downdating of a Cholesky decomposition* Numerische Mathematik. 68 (1994) 457–467.
- [7] C. PAN, *A perturbation analysis of the problem of downdating a Cholesky factorization*, Linear Algebra Appl. 183 (1993) 103–115.
- [8] G. STEWART, *The effects of rounding error on an algorithm for downdating a Cholesky factorization*, J. Inst. Math. Appl. 23 (1979) 203–213.
- [9] J. SUN, *Perturbation analysis of the Cholesky downdating and QR updating problems*, SIAM J. Matrix Anal. Appl. 16 (1995) 760–775.
- [10] A. FAROOQ, M. SAMAR, H. LI, C. MU, *Sensitivity analysis for the block Cholesky downdating problem*, Int. J. Comput. Math. 97 (2020) 1234–1253.
- [11] H. LI, H. YANG, H. SHAO, *Perturbation analysis for block downdating of the generalized cholesky factorization*, Appl. Math. Comput. 218 (2012) 9451–9461.

- [12] A. FAROOQ, M. SAMAR, *Sensitivity analysis for the generalized Cholesky block downdating problem*, *Linear and Multilinear Algebra* 70 (2022), 997–1022.
- [13] X. W. CHANG, R. C. LI, *Multiplicative perturbation analysis for QR factorizations*, *Num. Algebra Control Optim.* 1 (2011) 301–316.
- [14] A. FAROOQ, M. SAMAR, H. LI AND C. MU, *Improved rigorous multiplicative perturbation bounds for the generalized Cholesky factorization and the Cholesky-like factorization*, *Math. Inequal. Appl.* 22 (2019) 133–149.
- [15] A. FAROOQ, M. SAMAR, *Multiplicative perturbation bounds for the block Cholesky downdating problem*, *Int. J. Comput. Math.* 97 (2020) 2421–2435.
- [16] H.-Y. LI, Y. YANG, *Rigorous multiplicative perturbation bounds for the generalized Cholesky factorization and the Cholesky-like factorization*, *J. Math. Inequal.* 8 (2014) 925–937.
- [17] Y. YANG, H. LI, *Multiplicative perturbation bounds for the SR factorization*, *J. Math. Res. Appl.* 34 (2014) 423–434.
- [18] H.-Y. LI, Y. WEI, *Improved rigorous perturbation bounds for the LU and QR factorizations*, *Numer. Linear Algebra Appl.* 22 (2015) 1115–1130.
- [19] M. SAMAR, A. FAROOQ, H. LI, C. MU, *Sensitivity analysis for the generalized Cholesky factorization*, *App. Math. Comput.* 362 (2019) 124556.
- [20] H.-Y. LI, Y. M. WEI, Y. YANG, *New rigorous perturbation bounds for the Cholesky-like factorization of skew-symmetric matrix*, *Linear Algebra Appl.* 491 (2016) 83–100.
- [21] X.-W. CHANG, *Perturbation analysis of Some Matrix factorization*, PhD thesis, School of Computer Science, McGill University, Montreal, 1997.
- [22] G. W. STEWART, J.-G. SUN, *Matrix Perturbation Theory*, Academic Press, Boston, 1990.
- [23] X.-W. CHANG, C. C. PAIGE, G. W. STEWART, *Perturbation analyses for the QR factorization*, *SIAM J. Matrix Anal. Appl.* 18 (1997) 775–791.
- [24] R. A. HORN, C. R. JOHNSON, *Topics in matrix analysis*, Cambridge University Press, 1991.
- [25] R. A. HORN, C. R. JOHNSON, *Matrix Analysis*, Cambridge University Press, Cambridge, UK, 1985.
- [26] W. LI, Z. XIE, S. VONG, *Sensitivity analysis for the symplectic QR factorization*, *J. Franklin. Inst.* 353 (2016) 1186–1205.

(Received April 19, 2022)

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