

AIMS Mathematics, 8(10): 23016–23031. DOI: 10.3934/math.20231171 Received: 23 May 2023 Revised: 28 June 2023 Accepted: 04 July 2023 Published: 19 July 2023

http://www.aimspress.com/journal/Math

Research article

The solutions of two classes of dual matrix equations

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Abstract: The solvability conditions for the dual matrix equation AXB = D and a pair of dual matrix equations AX = C and XB = D are deduced by applying the singular value decomposition, and the expressions of the general solutions to these dual matrix equations are provided. Furthermore, the minimum-norm solutions of these dual matrix equations are provided. Finally, two numerical experiments are given to validate the accuracy of the results obtained.

Keywords: dual matrix equation; singular value decomposition; general solution; minimum-norm solution

Mathematics Subject Classification: 15A24, 65F05

1. Introduction

We will adopt the following terminology. $\mathbb{R}^{m \times n}$ and $\mathbb{OR}^{n \times n}$ denote the sets of all $m \times n$ real matrices and $n \times n$ orthogonal matrices, respectively. I_n denotes the identity matrix of size n. $A^{\top}, A^{\dagger}, \operatorname{tr}(A)$ and $||A||_{\mathrm{F}}$ represent the transpose, the Moore-Penrose inverse, the trace and the Frobenius norm of the matrix A, respectively. We use $A \otimes B$ to represent the Kronecker product of A and B, and $\operatorname{vec}(\cdot)$ to represent the vec operator, that is, $\operatorname{vec}(A) = [a_1^{\top}, \cdots, a_n^{\top}]^{\top}$, where $A = [a_1, \cdots, a_n] \in \mathbb{R}^{m \times n}, a_i \in \mathbb{R}^m, i = 1, \cdots, n$. Also, the symbols E_A and F_A stand for the two orthogonal projectors $E_A = I_m - AA^{\dagger}, F_A = I_n - A^{\dagger}A$ induced by $A \in \mathbb{R}^{m \times n}$.

As we know, the linear matrix equation

$$AXB = D, (1)$$

and the linear matrix equations

$$AX = C, XB = D \tag{2}$$

have been considered in real and complex matrix spaces by many scholars. For example, Penrose [1] acquired the consistency condition and the general solution of Eq (1) using the generalized inverse.

Dai [2] obtained the solvability conditions and the general symmetric solution to Eq (1) by utilizing the generalized singular value decomposition. Deng et al. [3] considered the symmetric, skew-symmetric and symmetric positive semidefinite solutions of Eq (1) by means of the quotient singular value decomposition. For Eq (2), Mitra [4] gave the solvability condition and the expression of the general common solution of Eq (2) by applying the generalized inverse. Dajić et al. [5] studied the positive solutions to Eq (2) for Hilbert space operators. Khatri et al. [6] provided the general expressions about the Hermitian solutions and Hermitian nonnegative definite solutions of Eqs (1) and (2).

In 1873, Clifford [7] introduced dual numbers to form dual quaternions for studying non-Euclidean geometry. Study [8] defined dual numbers as dual angles to specify the relation between two lines in Euclidean space. Due to the wide application in many engineering fields, the dual numbers and their algebra have attracted a large number of scholars in the past 30 years. For example, McAulay [9] used dual quaternions to describe finite displacement of rigid and deformable bodies. Dimentberg [10] pioneered kinematic analysis of spatial mechanisms through dual numbers. Pennock and Yang [11,12], Dooley and McCarthy [13], Ravani and Ge [14] studied the kinematics, dynamics and calibration of open-chain robot manipulators by applying dual numbers. The set of the dual numbers is usually denoted by

$$\mathbb{D} = \{a = a_1 + \varepsilon a_2 | a_1, a_2 \in \mathbb{R}, \varepsilon \neq 0, \varepsilon^2 = 0\}.$$

For any two dual numbers $a = a_1 + \varepsilon a_2$ and $b = b_1 + \varepsilon b_2$, the arithmetic operations for dual numbers are as follows:

- (1) Equality : $a = b \Leftrightarrow a_1 = b_1, a_2 = b_2$;
- (2) Addition : $a + b = (a_1 + b_1) + \varepsilon(a_2 + b_2)$;
- (3) Multiplication : $ab = a_1b_1 + \varepsilon(a_1b_2 + a_2b_1)$.

A matrix whose elements are dual numbers is called a dual matrix, namely, the set of all $m \times n$ real dual matrices is

$$\mathbb{D}^{m \times n} = \{ A = A_1 + \varepsilon A_2 | A_1, A_2 \in \mathbb{R}^{m \times n} \}.$$

The operational rules for dual matrices are similar to those of dual numbers. Dual matrices also have important applications in kinematic analysis and robotics [15–17]. The solution of systems of linear dual equations is a task often required in synthesis problems and sensor calibration problems. For instance, Condurache and Burlacu [18] solved AX = XB sensor calibration problems by applying the orthogonal dual tensor method. Condurache and Ciureanu [19] solved AX = YB sensor calibration problems by applying dual algebra. Udwadiae [20] dealt with properties of dual generalized inverses and then used them to solve the dual matrix equation Ax = b.

Inspired by the works of papers [18–20], in this paper, we will consider the general solutions and the minimum-norm solutions of matrix equations (1) and (2) in the dual matrix space, which can be formulated as follows:

Problem I(a). Given dual matrices $A = A_1 + \varepsilon A_2$, $B = B_1 + \varepsilon B_2$ and $D = D_1 + \varepsilon D_2$, where $A_i \in \mathbb{R}^{m \times p}$, $B_i \in \mathbb{R}^{q \times n}$ and $D_i \in \mathbb{R}^{m \times n}$ (i = 1, 2). Find a *p*-by-*q* dual matrix $X = X_1 + \varepsilon X_2$ such that the dual matrix equation (1) is satisfied.

Problem I(b). Let S_1 be the solution set of Problem I(a). Find $\hat{X} = \hat{X}_1 + \varepsilon \hat{X}_2 \in S_1$ such that $\|\hat{X}\|_D = \sqrt{\|\hat{X}_1\|_F^2 + \|\hat{X}_2\|_F^2} = \min$.

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Problem II(a). Given dual matrices $A = A_1 + \varepsilon A_2$, $B = B_1 + \varepsilon B_2$, $C = C_1 + \varepsilon C_2$ and $D = D_1 + \varepsilon D_2$, where $A_i \in \mathbb{R}^{m \times p}$, $B_i \in \mathbb{R}^{q \times n}$, $C_i \in \mathbb{R}^{m \times q}$ and $D_i \in \mathbb{R}^{p \times n}$ (i = 1, 2). Find a *p*-by-*q* dual matrix $X = X_1 + \varepsilon X_2$ such that the dual matrix equations (2) are satisfied.

Problem II(b). Supposed that S_2 is the solution set of Problem II(a). Find $\hat{X} = \hat{X}_1 + \varepsilon \hat{X}_2 \in S_2$ such that $\|\hat{X}\|_{\rm D} = \sqrt{\|\hat{X}_1\|_{\rm F}^2 + \|\hat{X}_2\|_{\rm F}^2} = \min$.

In order to solve Problems I and II, we first split the dual matrix equations (1) and (2) into the two real-valued matrix equations. Then, by applying the singular value decomposition (SVD), we obtain the solvability conditions and the expressions of the general solutions for Problems I(a) and II(a). Furthermore, we deduce the unique minimum-norm solutions of Problems I(b) and II(b) by using the Kronecker product and stretching function. Finally, we give two numerical experiments to validate the accuracy of the results.

2. The solutions to Problems I(a) and I(b)

To begin with, we introduce the following lemmas.

Lemma 2.1. [21] If $A \in \mathbb{R}^{m \times p}$, $B \in \mathbb{R}^{q \times n}$ and $D \in \mathbb{R}^{m \times n}$. Then the matrix equation (1) has a solution $X \in \mathbb{R}^{p \times q}$ if and only if $AA^{\dagger}DB^{\dagger}B = D$. In this case, the general solution is $X = A^{\dagger}DB^{\dagger} + F_AV_1 + V_2E_B$, where V_1, V_2 are arbitrary matrices.

Lemma 2.2. [22] Suppose that A, B are two real matrices, and X is an unknown variable matrix. Then 2tr(PV) = 2tr(VTPT) = 2tr(AVPV)

$$\frac{\partial \operatorname{tr}(BX)}{\partial X} = B^{\mathsf{T}}, \ \frac{\partial \operatorname{tr}(X^{\mathsf{T}}B^{\mathsf{T}})}{\partial X} = B^{\mathsf{T}}, \ \frac{\partial \operatorname{tr}(AXBX)}{\partial X} = (BXA + AXB)^{\mathsf{T}},$$
$$\frac{\partial \operatorname{tr}(AX^{\mathsf{T}}BX^{\mathsf{T}})}{\partial X} = BX^{\mathsf{T}}A + AX^{\mathsf{T}}B, \ \frac{\partial \operatorname{tr}(AXBX^{\mathsf{T}})}{\partial X} = AXB + A^{\mathsf{T}}XB^{\mathsf{T}}.$$

Lemma 2.3. [23] Let $A \in \mathbb{R}^{m \times n}$, $B \in \mathbb{R}^{n \times l}$, $C \in \mathbb{R}^{l \times s}$. Then $\operatorname{vec}(ABC) = (C^{\top} \otimes A) \operatorname{vec}(B)$.

By separating the dual equation of (1) into the real part and the dual part leads to the following two equations:

$$A_1 X_1 B_1 = D_1, \ A_1 X_2 B_1 + A_1 X_1 B_2 + A_2 X_1 B_1 = D_2.$$
(3)

Suppose that the SVDs of the matrices A_1 and B_1 are

$$A_1 = P \begin{bmatrix} \Sigma & 0 \\ 0 & 0 \end{bmatrix} Q^{\mathsf{T}}, \ B_1 = U \begin{bmatrix} \Omega & 0 \\ 0 & 0 \end{bmatrix} V^{\mathsf{T}}, \tag{4}$$

where $\Sigma = \text{diag}(\gamma_1, \dots, \gamma_{r_1}) > 0, r_1 = \text{rank}(A_1), P = [P_1, P_2] \in \mathbb{OR}^{m \times m}, Q = [Q_1, Q_2] \in \mathbb{OR}^{p \times p}, \Omega = \text{diag}(\beta_1, \dots, \beta_{r_2}) > 0, r_2 = \text{rank}(B_1), U = [U_1, U_2] \in \mathbb{OR}^{q \times q}, V = [V_1, V_2] \in \mathbb{OR}^{n \times n}$ with $P_1 \in \mathbb{R}^{m \times r_1}, Q_1 \in \mathbb{R}^{p \times r_1}, U_1 \in \mathbb{R}^{q \times r_2}$ and $V_1 \in \mathbb{R}^{n \times r_2}$. Let

$$Q^{\mathsf{T}}X_{1}U = \begin{bmatrix} Y_{1} & Y_{2} \\ Y_{3} & Y_{4} \end{bmatrix}, \ Q^{\mathsf{T}}X_{2}U = \begin{bmatrix} Z_{1} & Z_{2} \\ Z_{3} & Z_{4} \end{bmatrix}, \ P^{\mathsf{T}}D_{1}V = \begin{bmatrix} D_{11} & D_{12} \\ D_{13} & D_{14} \end{bmatrix},$$
(5)

$$U^{\mathsf{T}}B_{2}V = \begin{bmatrix} B_{21} & B_{22} \\ B_{23} & B_{24} \end{bmatrix}, \ P^{\mathsf{T}}A_{2}Q = \begin{bmatrix} A_{21} & A_{22} \\ A_{23} & A_{24} \end{bmatrix}, \ P^{\mathsf{T}}D_{2}V = \begin{bmatrix} D_{21} & D_{22} \\ D_{23} & D_{24} \end{bmatrix}.$$
(6)

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Then the matrix equations (3) are equivalent to

$$\begin{bmatrix} \Sigma Y_1 \Omega & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} D_{11} & D_{12} \\ D_{13} & D_{14} \end{bmatrix},$$

$$\begin{bmatrix} \Sigma Z_1 \Omega + \Sigma Y_1 B_{21} + \Sigma Y_2 B_{23} + A_{21} Y_1 \Omega + A_{22} Y_3 \Omega & \Sigma Y_1 B_{22} + \Sigma Y_2 B_{24} \\ A_{23} Y_1 \Omega + A_{24} Y_3 \Omega & 0 \end{bmatrix} = \begin{bmatrix} D_{21} & D_{22} \\ D_{23} & D_{24} \end{bmatrix}.$$
(8)

It follows from the equations of (7) and (8) that

$$D_{12} = 0, \ D_{13} = 0, \ D_{14} = 0, \ D_{24} = 0,$$
 (9)

$$Y_1 = \Sigma^{-1} D_{11} \Omega^{-1}, \tag{10}$$

$$\Sigma Y_1 B_{22} + \Sigma Y_2 B_{24} = D_{22}, \ A_{23} Y_1 \Omega + A_{24} Y_3 \Omega = D_{23}, \tag{11}$$

$$\Sigma Z_1 \Omega + \Sigma Y_1 B_{21} + \Sigma Y_2 B_{23} + A_{21} Y_1 \Omega + A_{22} Y_3 \Omega = D_{21}.$$
 (12)

We note that

$$D_{13} = 0, D_{14} = 0 \Leftrightarrow P_2^{\top} D_1 V = 0 \Leftrightarrow P_2 P_2^{\top} D_1 = 0 \Leftrightarrow E_{A_1} D_1 = 0;$$

$$D_{12} = 0 \Leftrightarrow P_1^{\top} D_1 V_2 = 0 \Leftrightarrow P_1 P_1^{\top} D_1 V_2 V_2^{\top} = 0 \Leftrightarrow A_1 A_1^{\dagger} D_1 F_{B_1} = 0 \Leftrightarrow D_1 F_{B_1} = 0;$$

$$D_{24} = 0 \Leftrightarrow P_2^{\top} D_2 V_2 = 0 \Leftrightarrow P_2 P_2^{\top} D_2 V_2 V_2^{\top} = 0 \Leftrightarrow E_{A_1} D_2 F_{B_1} = 0.$$

Plugging (10) into (11), we have

$$Y_2 B_{24} = J_1, (13)$$

$$A_{24}Y_3 = J_2, (14)$$

where $J_1 = \Sigma^{-1} D_{22} - \Sigma^{-1} D_{11} \Omega^{-1} B_{22}$, $J_2 = D_{23} \Omega^{-1} - A_{23} \Sigma^{-1} D_{11} \Omega^{-1}$. By Lemma 1, Eqs (13) and (14) with respect to Y_2 and Y_3 have solutions if and only if $J_1F_{B_{24}} = 0$, $E_{A_{24}}J_2 = 0$, and the general solutions are

$$Y_2 = J_1 B_{24}^{\dagger} + W_1 E_{B_{24}}, \ Y_3 = A_{24}^{\dagger} J_2 + F_{A_{24}} W_2, \tag{15}$$

where W_1 and W_2 are arbitrary matrices. Inserting (10) and (15) into (12) yields

$$Z_1 = J_3 - W_1 E_{B_{24}} B_{23} \Omega^{-1} - \Sigma^{-1} A_{22} F_{A_{24}} W_2,$$
(16)

where $J_3 = \Sigma^{-1} (D_{21} - D_{11} \Omega^{-1} B_{21} - A_{21} \Sigma^{-1} D_{11}) \Omega^{-1} - J_1 B_{24}^{\dagger} B_{23} \Omega^{-1} - \Sigma^{-1} A_{22} A_{24}^{\dagger} J_2.$

In summary, we have proven the following theorem.

Theorem 2.1. Given dual matrices $A = A_1 + \varepsilon A_2$, $B = B_1 + \varepsilon B_2$ and $D = D_1 + \varepsilon D_2$, where $A_i \in$ $\mathbb{R}^{m \times p}$, $B_i \in \mathbb{R}^{q \times n}$ and $D_i \in \mathbb{R}^{m \times n}$ (i = 1, 2). Let the SVDs of A_1, B_1 be given by (4) and the partitions of the matrices $U^{\mathsf{T}}B_2V$, $P^{\mathsf{T}}D_1V$, $P^{\mathsf{T}}D_2V$ and $P^{\mathsf{T}}A_2Q$ be given by (5) and (6). Then Eq (1) is solvable if and only if

$$J_1 F_{B_{24}} = 0, \ E_{A_{24}} J_2 = 0, \ E_{A_1} D_1 = 0, \ D_1 F_{B_1} = 0, \ E_{A_1} D_2 F_{B_1} = 0,$$
(17)

where $J_1 = \Sigma^{-1}D_{22} - \Sigma^{-1}D_{11}\Omega^{-1}B_{22}, J_2 = D_{23}\Omega^{-1} - A_{23}\Sigma^{-1}D_{11}\Omega^{-1}$. In this case, the solution set of Problem I(a) can be expressed as $S_1 = \{X = X_1 + \varepsilon X_2 | X_1, X_2 \in \mathbb{R}^{p \times q}\}$, where

$$X_{1} = Q \begin{bmatrix} \Sigma^{-1} D_{11} \Omega^{-1} & J_{1} B_{24}^{\dagger} + W_{1} E_{B_{24}} \\ A_{24}^{\dagger} J_{2} + F_{A_{24}} W_{2} & Y_{4} \end{bmatrix} U^{\top},$$
(18)

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$$X_{2} = Q \begin{bmatrix} J_{3} - W_{1} E_{B_{24}} B_{23} \Omega^{-1} - \Sigma^{-1} A_{22} F_{A_{24}} W_{2} & Z_{2} \\ Z_{3} & Z_{4} \end{bmatrix} U^{\mathsf{T}},$$
(19)

with

$$J_{1} = \Sigma^{-1} D_{22} - \Sigma^{-1} D_{11} \Omega^{-1} B_{22}, \quad J_{2} = D_{23} \Omega^{-1} - A_{23} \Sigma^{-1} D_{11} \Omega^{-1},$$

$$J_{3} = \Sigma^{-1} (D_{21} - D_{11} \Omega^{-1} B_{21} - A_{21} \Sigma^{-1} D_{11}) \Omega^{-1} - J_{1} B_{24}^{\dagger} B_{23} \Omega^{-1} - \Sigma^{-1} A_{22} A_{24}^{\dagger} J_{2},$$

and $W_1, W_2, Y_4, Z_2, Z_3, Z_4$ are arbitrary matrices.

Now, for $\forall X \in S_1$, we obtain

$$\begin{split} \|X\|_{\mathrm{D}}^{2} &= \|X_{1}\|_{\mathrm{F}}^{2} + \|X_{2}\|_{\mathrm{F}}^{2} = \|Q^{\top}X_{1}U\|_{\mathrm{F}}^{2} + \|Q^{\top}X_{2}U\|_{\mathrm{F}}^{2} \\ &= \|\Sigma^{-1}D_{11}\Omega^{-1}\|_{\mathrm{F}}^{2} + \|Y_{4}\|_{\mathrm{F}}^{2} + \|Z_{2}\|_{\mathrm{F}}^{2} + \|Z_{3}\|_{\mathrm{F}}^{2} + \|Z_{4}\|_{\mathrm{F}}^{2} + \|J_{1}B_{24}^{\dagger} + W_{1}E_{B_{24}}\|_{\mathrm{F}}^{2} \\ &+ \|A_{24}^{\dagger}J_{2} + F_{A_{24}}W_{2}\|_{\mathrm{F}}^{2} + \|J_{3} - W_{1}M_{2}^{\top}B_{23}\Omega^{-1} - \Sigma^{-1}A_{22}T_{2}W_{2}\|_{\mathrm{F}}^{2}. \end{split}$$

Then $||X||_D$ is minimized if and only if

$$Y_4 = 0, Z_2 = 0, Z_3 = 0, Z_4 = 0,$$
 (20)

$$\Phi(W_1, W_{12}) = \|J_1 B_{24}^{\dagger} + W_1 E_{B_{24}}\|_{\rm F}^2 + \|A_{24}^{\dagger} J_2 + F_{A_{24}} W_2\|_{\rm F}^2 + \|J_3 - W_1 E_{B_{24}} B_{23} \Omega^{-1} - \Sigma^{-1} A_{22} F_{A_{24}} W_2\|_{\rm F}^2 = \min.$$
(21)

Clearly, the minimization problem (21) is equivalent to

$$\Phi(W_1, W_2) = \operatorname{tr}[(B_{24}^{\dagger})^{\top} J_1^{\top} J_1 B_{24}^{\dagger} + E_{B_{24}} W_1^{\top} W_1 E_{B_{24}} + W_2^{\top} F_{A_{24}} W_2 + W_2^{\top} F_{A_{24}} A_{22}^{\top} \Sigma^{-1} \Sigma^{-1} A_{22} F_{A_{24}} W_2 + J_2^{\top} (A_{24}^{\dagger})^{\top} A_{24}^{\dagger} J_2 + J_3^{\top} J_3 + \Omega^{-1} B_{23}^{\top} E_{B_{24}} W_1^{\top} W_1 E_{B_{24}} B_{23} \Omega^{-1}] + 2 \operatorname{tr}(-J_3^{\top} W_1 E_{B_{24}} B_{23} \Omega^{-1} + \Omega^{-1} B_{23}^{\top} E_{B_{24}} W_1^{\top} \Sigma^{-1} A_{22} F_{A_{24}} W_2 - J_3^{\top} \Sigma^{-1} A_{22} F_{A_{24}} W_2) = \min .$$

Therefore, $\Phi(W_1, W_2)$ is minimized if and only if $\frac{\partial \Phi(W_1, W_2)}{\partial W_1} = 0$ and $\frac{\partial \Phi(W_1, W_2)}{\partial W_2} = 0$, which implies that

$$W_1(E_{B_{24}} + L_1^{\mathsf{T}}L_1) + L_2 W_2 L_1 = J_3 L_1,$$
(22)

$$L_2^{\mathsf{T}} W_1 L_1^{\mathsf{T}} + (F_{A_{24}} + L_2^{\mathsf{T}} L_2) W_2 = L_2^{\mathsf{T}} J_3,$$
(23)

where $L_1 = \Omega^{-1} B_{23}^{\top} E_{B_{24}}$, $L_2 = \Sigma^{-1} A_{22} F_{A_{24}}$. By applying the Kronecker product and stretching function, the matrix equations (22) and (23) can be equivalently written as

$$\Delta \begin{bmatrix} \operatorname{vec}(W_1) \\ \operatorname{vec}(W_2) \end{bmatrix} = \Gamma, \tag{24}$$

where

$$\Delta = \begin{bmatrix} (E_{B_{24}} + L_1^{\mathsf{T}}L_1) \otimes I_{r_1} & L_1^{\mathsf{T}} \otimes L_2 \\ L_1 \otimes L_2^{\mathsf{T}} & I_{r_2} \otimes (F_{A_{24}} + L_2^{\mathsf{T}}L_2) \end{bmatrix}, \Gamma = \begin{bmatrix} \operatorname{vec}(J_3L_1) \\ \operatorname{vec}(L_2^{\mathsf{T}}J_3) \end{bmatrix}.$$
(25)

Then, W_1 and W_2 are determined by solving the unique solution of the Eq (24). Inserting (20), W_1 and W_2 into (18)–(19), we have the following result.

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Theorem 2.2. If the conditions (17) are satisfied, then Problem I(b) has the unique solution \hat{X} , and \hat{X} admits the following representation: $\hat{X} = \hat{X}_1 + \varepsilon \hat{X}_2$, where

$$\hat{X}_{1} = Q \begin{bmatrix} \Sigma^{-1} D_{11} \Omega^{-1} & J_{1} B_{24}^{\dagger} + W_{1} E_{B_{24}} \\ A_{24}^{\dagger} J_{2} + F_{A_{24}} W_{2} & 0 \end{bmatrix} U^{\top},$$
(26)

$$\hat{X}_{2} = Q \begin{bmatrix} J_{3} - W_{1} E_{B_{24}} B_{23} \Omega^{-1} - \Sigma^{-1} A_{22} F_{A_{24}} W_{2} & 0\\ 0 & 0 \end{bmatrix} U^{\mathsf{T}},$$
(27)

with W_1 , W_2 being determined by (24).

Based on Theorems 2.1 and 2.2, we can describe an algorithm to solve Problems I(a) and I(b) as follows.

Algorithm 1

- 1: Input matrices A_i , B_i and D_i , i = 1, 2.
- 2: Compute the SVDs of the matrices A_1, B_1 as in (4).
- 3: Compute $J_1 = \Sigma^{-1} D_{22} \Sigma^{-1} D_{11} \Omega^{-1} B_{22}, J_2 = D_{23} \Omega^{-1} A_{23} \Sigma^{-1} D_{11} \Omega^{-1},$

$$J_3 = \Sigma^{-1} (D_{21} - D_{11} \Omega^{-1} B_{21} - A_{21} \Sigma^{-1} D_{11}) \Omega^{-1} - J_1 B_{24}^{\dagger} B_{23} \Omega^{-1} - \Sigma^{-1} A_{22} A_{24}^{\dagger} J_2.$$

- 4: Compute the partitions of the matrices $U^{\top}B_2V$, $P^{\top}D_1V$, $P^{\top}D_2V$ and $P^{\top}A_2Q$ as in (5) and (6).
- 5: If the conditions (17) are satisfied, then continue, otherwise, Problem I(a) has no solution, and stop.
- 6: Calculate Y_1 according to (10).
- 7: Calculate Y_4, Z_2, Z_3, Z_4 on the basis of (20).
- 8: Calculate $L_1 = \Omega^{-1} B_{23}^{\top} E_{B_{24}}, L_2 = \Sigma^{-1} A_{22} F_{A_{24}}.$
- 9: Compute Δ and Γ by (25).
- 10: Solve the equation of (24).
- 11: Compute *Y*₂, *Y*₃, *Z*₁ by (15) and (16).
- 12: Compute \hat{X}_1, \hat{X}_2 according to (26) and (27), then, compute the unique solution $\hat{X} = \hat{X}_1 + \varepsilon \hat{X}_2$ of Problem I(b).

Example 2.1. Let m = 8, n = 9, p = 10, q = 9, and the matrices $A_1, A_2, B_1, B_2, D_1, D_2$ be given by

0.2427 0.5962 0.7341 0.9667 1.3414 0.6568 0.7459 1.0132 0.7080 0.1301 0.0702 0.3974 0.5911 0.6660 0.8767 0.1408 0.2451 0.4824 0.5419 0.5388 $A_1 = \begin{bmatrix} 0.0695 & 0.3726 & 0.5316 & 0.6196 & 0.8258 & 0.1366 \\ 0.0589 & 0.3715 & 0.5941 & 0.6315 & 0.8125 & 0.1230 \\ 0.1363 & 0.6145 & 0.7427 & 0.9935 & 1.3850 & 0.2530 \\ \end{bmatrix}$ 0.2725 0.4556 0.5377 0.4903 0.1506 0.4450 0.4523 0.5310 0.7034 0.7709 1.0626 0.7206 0.1340 0.7068 0.9949 1.1725 1.5689 0.2620 1.0378 0.9212 0.5427 0.8662 0.2921 0.3424 0.4699 0.6601 0.1224 0.3544 0.3679 0.5189 0.3355 0.3061 0.3146 0.4831 0.6993 0.1374 0.4551 0.3920 0.0769 0.6018 0.3226

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$A_2 = $	1.0933	0.8244	0.6818	1.0802	0.8676	0.771	0.8561	0.3433	1.1049 ().4056]	
	0.6499	0.5789	0.3073	0.6518	0.6119	0.4878	8 0.6703	0.1769	0.7701 (0.4037	
	0.8323	0.9439	0.5413	0.7775	0.9802	0.9786	6 0.9816	0.3260	1.0896 (0.5079	
	0.6000	0.6703	0.4696	0.5449	0.6914	0.7547	7 0.6443	0.2677	0.7441 (0.2778	
	1.0163	1.0320	0.5939	0.9790	1.0786	1.0003	3 1.1115	0.3479	1.2602 (0.6043 '	
	0.5387	0.5719	0.3400	0.5104	0.5955	0.5792	0.5998	0.1997	0.6778 (0.3120	
	0.9388	1.1907	0.8139	0.8177	1.2209	1.4099	9 1.1104	0.4779	1.2467 (0.4531	
	1.2762	1.1933	0.7768	1.2358	1.2484	1.1587	7 1.2636	0.4326	1.4860 ().6551]	
ſ	0.8675	0.4523	0.8875	1.1213	0.9407	0.5982	2 0.6560	0.4344	0.3021		
	1.3819	0.7358	1.4222	1.7959	1.5165	0.9688	8 1.0587	0.6925	0.4697		
	1.3975	0.6758	1.2704	1.6948	1.3475	0.5510	0 1.0278	0.8306	0.6958		
	1.1300	0.5506	1.0451	1.3823	1.1071	0.4925	5 0.8326	0.6559	0.5393		
$B_1 =$	1.2322	0.6571	1.2424	1.5862	1.3321	0.7931	0.9486	0.6440	0.4518	,	
	1.2011	0.7269	1.0570	1.4796	1.2359	0.3112	2 1.0301	0.8344	0.6341		
	1.1470	0.7370	1.1095	1.4859	1.2928	0.5518	8 1.0113	0.7204	0.4738		
	0.5748	0.0891	0.6922	0.7423	0.5551	0.7782	2 0.2157	0.0595	0.0786		
ĺ	1.6950	0.8346	1.5239	2.0498	1.6313	0.6149	9 1.2635	1.0333	0.8650		
[0.2446	0.6215	0.3151	0.5083	0.1282	0.4160	0.3089	0.8484	0.3954		
	0.1970	0.5539	0.2235	0.2395	0.1191	0.4638	8 0.1467	0.6858	0.1861		
	0.1373	0.3891	0.1541	0.1577	0.0839	0.3304	4 0.0966	0.4783	0.1225		
	0.1673	0.4592	0.1961	0.2388	0.0978	0.3669	9 0.1458	0.5818	0.1856		
$B_2 =$	0.1324	0.3678	0.1528	0.1754	0.0787	0.3009	9 0.1072	0.4608	0.1363	,	
	0.0769	0.1862	0.1044	0.1894	0.0376	0.1084	4 0.1149	0.2664	0.1474		
	0.1138	0.3381	0.1188	0.0807	0.0742	0.3116	6 0.0500	0.3971	0.0626		
	0.3495	0.9051	0.4406	0.6721	0.1883	0.6355	5 0.4089	1.2131	0.5227		
	0.2584	0.7188	0.2976	0.3390	0.1539	0.5896	6 0.2073	0.8993	0.2634]		
	37.2772	19.549	35.8	150 46.	8226 3	8.6175	19.2825	28.6267	21.0905	15.9790	
	23.9336	12.547	6 22.9	702 30.	0461 2	4.7711	12.3149	18.3790	13.5635	10.2912	
	22.6423	11.871	6 21.7	366 28.	4287 2	3.4399	11.6655	17.3875	12.8265	9.7286	
$D_1 =$	22.0040	11.534	2 21.1	080 27.	6170 2	2.7643	11.2944	16.8970	12.4795	9.4750	
1	38.5481	20.216	8 37.0	394 48. -	4211 3	9.9373	19.9489	29.6028	21.8065	16.5194	Ĺ
	43.0737	22.584	5 41.3	541 54.	0838 4	4.5943	22.2009	33.0774	24.3976	18.5028	
	18.4160	9.658	9 17.6 5	978 23.	1345 1	9.0821	9.5372	14.1426	10.4156	7.8887	
	19.6970	10.332	5 18.9	394 24.	/505 2	0.4193	10.2287	15.1265	11.1304	8.4236]	J
	84.6062	55.853	0 82.8	008 107	.4806	84.9843	53.8552	65.0872	63.1407	39.8291	
	55.2478	36.365	2 54.0	591 70	0.1724	55.5620	35.0690	42.5217	41.0474	25.9362	
	67.2474	42.199	2 65.6	044 85	.2283	68.0915	41.1427	51.6625	47.2077	30.8779	
$D_2 =$	54.5082	35.341	4 53.2	854 69	0.1744	54.9224	34.2257	41.9065	39.7918	25.4096	
-	91.4598	59.847	3 89.4	671 116	0.1522	92.0215	57.8224	70.3578	67.5163	42.8523	
	75.5141	52.955	2 74.2	156 96	0.2120	/5.1667	50.4517	58.2216	60.4825	36.5619	
	70.9283	42.753	69.0	287 89	.7357	72.1844	42.0976	54.3926	47.4731	32.0002	
	78.0985	46.925	6 75.9	816 98	.7998	/9.5347	46.1771	59.9163	52.0616	35.1900]

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It is easy to verity that the conditions (17) hold:

$$||J_1F_{B_{24}}||_{\rm F} = 1.6016 \times 10^{-14}, \quad ||E_{A_{24}}J_2||_{\rm F} = 1.4097 \times 10^{-14}, \\ ||E_{A_1}D_1||_{\rm F} = 5.0053 \times 10^{-14}, \quad ||D_1F_{B_1}||_{\rm F} = 3.0245 \times 10^{-14}, \quad ||E_{A_1}D_2F_{B_1}||_{\rm F} = 4.6871 \times 10^{-14}.$$

By using Algorithm 1, we can obtain the unique solution $\hat{X} = \hat{X}_1 + \varepsilon \hat{X}_2$ of Problem I(b) as follows:

$$\hat{X}_1 = \begin{bmatrix} 0.3763 & 0.5895 & 0.3469 & 0.3112 & 0.4847 & 0.2216 & 0.3545 & 0.4573 & 0.3974 \\ 0.4878 & 0.7288 & 0.4641 & 0.4338 & 0.5808 & 0.4012 & 0.6001 & 0.4085 & 0.6289 \\ 0.3002 & 0.4251 & 0.4591 & 0.4264 & 0.3011 & 0.2937 & 0.4327 & 0.2866 & 0.7339 \\ 0.4595 & 0.6465 & 0.4887 & 0.4648 & 0.4980 & 0.4628 & 0.6505 & 0.2843 & 0.7430 \\ 0.5646 & 0.7673 & 0.5588 & 0.5374 & 0.5982 & 0.6037 & 0.8105 & 0.2761 & 0.8571 \\ 0.4124 & 0.6306 & 0.4015 & 0.3593 & 0.5174 & 0.2495 & 0.3785 & 0.5124 & 0.4744 \\ 0.4466 & 0.6045 & 0.2402 & 0.2273 & 0.5337 & 0.3848 & 0.4518 & 0.2738 & 0.2252 \\ 0.1781 & 0.1980 & 0.2181 & 0.2092 & 0.1482 & 0.2368 & 0.2677 & 0.0571 & 0.3676 \\ 0.5410 & 0.7325 & 0.4143 & 0.3931 & 0.6124 & 0.5088 & 0.6364 & 0.3260 & 0.5358 \\ 0.4019 & 0.6090 & 0.3849 & 0.3814 & 0.4576 & 0.4343 & 0.6872 & 0.1325 & 0.6204 \end{bmatrix}$$

$$\hat{X}_2 = \begin{bmatrix} 0.0594 & 0.0944 & 0.0812 & 0.0674 & 0.0811 & 0.0546 & 0.0604 & 0.0676 & 0.0954 \\ 0.2742 & 0.4398 & 0.3499 & 0.2935 & 0.3760 & 0.2580 & 0.3026 & 0.2770 & 0.4717 \\ 0.3404 & 0.5512 & 0.4009 & 0.3405 & 0.4687 & 0.3283 & 0.4070 & 0.2970 & 0.4726 \\ 0.4452 & 0.7152 & 0.5610 & 0.4714 & 0.6109 & 0.4206 & 0.4980 & 0.4398 & 0.6602 \\ 0.6164 & 0.9878 & 0.7924 & 0.6639 & 0.8450 & 0.5786 & 0.6749 & 0.6307 & 0.9321 \\ 0.1110 & 0.1770 & 0.1486 & 0.1238 & 0.1519 & 0.1028 & 0.1160 & 0.1218 & 0.1747 \\ 0.2968 & 0.4662 & 0.4423 & 0.3629 & 0.4033 & 0.2640 & 0.2681 & 0.3885 & 0.5185 \\ 0.3426 & 0.5488 & 0.4422 & 0.3702 & 0.4696 & 0.3212 & 0.3735 & 0.3529 & 0.5200 \\ 0.4623 & 0.7347 & 0.6340 & 0.5261 & 0.6314 & 0.4244 & 0.4690 & 0.5284 & 0.7446 \\ 0.3274 & 0.5287 & 0.3959 & 0.3348 & 0.4503 & 0.3134 & 0.3820 & 0.3000 & 0.4663 \end{bmatrix}$$

The absolute errors are estimated by

$$||A_1\hat{X}_1B_1 - D_1||_{\mathsf{F}} = 5.6258 \times 10^{-14},$$

$$||A_1\hat{X}_2B_1 + A_1\hat{X}_1B_2 + A_2\hat{X}_1B_1 - D_2||_{\mathsf{F}} = 1.7614 \times 10^{-13},$$

which implies that \hat{X} is the solution of Problem I(b).

3. The solutions to Problems II(a) and II(b)

We first introduce the following lemma.

Lemma 3.1. [24] If $A \in \mathbb{R}^{m \times p}$, $B \in \mathbb{R}^{q \times n}$, $C \in \mathbb{R}^{m \times q}$ and $D \in \mathbb{R}^{p \times n}$, then the matrix equations (2) have a common solution $X \in \mathbb{R}^{p \times q}$ if and only if $E_A C = 0$, $DF_B = 0$, AD = CB. In this case, the general solution is $X = A^{\dagger}C + F_A DB^{\dagger} + F_A VE_B$, where V is an arbitrary matrix.

Obviously, Eq (2) can be equivalently expressed as

$$A_1X_1 = C_1, X_1B_1 = D_1, A_2X_1 + A_1X_2 = C_2, X_1B_2 + X_2B_1 = D_2.$$
 (28)

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Let SVDs of the matrices A_1 and B_1 be given by (4), and let

$$P^{\mathsf{T}}C_{1}U = \begin{bmatrix} C_{11} & C_{12} \\ C_{13} & C_{14} \end{bmatrix}, P^{\mathsf{T}}C_{2}U = \begin{bmatrix} C_{21} & C_{22} \\ C_{23} & C_{24} \end{bmatrix},$$
(29)

$$Q^{\mathsf{T}}D_{1}V = \begin{bmatrix} \tilde{D}_{11} & \tilde{D}_{12} \\ \tilde{D}_{13} & \tilde{D}_{14} \end{bmatrix}, Q^{\mathsf{T}}D_{2}V = \begin{bmatrix} \tilde{D}_{21} & \tilde{D}_{22} \\ \tilde{D}_{23} & \tilde{D}_{24} \end{bmatrix}.$$
 (30)

Then, the matrix equations (28) are equivalent to

$$\begin{bmatrix} \Sigma Y_1 & \Sigma Y_2 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} C_{11} & C_{12} \\ C_{13} & C_{14} \end{bmatrix}, \begin{bmatrix} Y_1 \Omega & 0 \\ Y_3 \Omega & 0 \end{bmatrix} = \begin{bmatrix} \tilde{D}_{11} & \tilde{D}_{12} \\ \tilde{D}_{13} & \tilde{D}_{14} \end{bmatrix},$$
(31)

$$\begin{array}{c} A_{21}Y_1 + A_{22}Y_3 + \Sigma Z_1 & A_{21}Y_2 + A_{22}Y_4 + \Sigma Z_2 \\ A_{23}Y_1 + A_{24}Y_3 & A_{23}Y_2 + A_{24}Y_4 \end{array} \right] = \left[\begin{array}{c} C_{21} & C_{22} \\ C_{23} & C_{24} \end{array} \right],$$
(32)

$$\begin{bmatrix} Y_1 B_{21} + Y_2 B_{23} + Z_1 \Omega & Y_1 B_{22} + Y_2 B_{24} \\ Y_3 B_{21} + Y_4 B_{23} + Z_3 \Omega & Y_3 B_{22} + Y_4 B_{24} \end{bmatrix} = \begin{bmatrix} \tilde{D}_{21} & \tilde{D}_{22} \\ \tilde{D}_{23} & \tilde{D}_{24} \end{bmatrix}.$$
(33)

It follows from (31), (32) and (33) that

[

$$C_{13} = 0, \ C_{14} = 0, \ \tilde{D}_{12} = 0, \ \tilde{D}_{14} = 0,$$
 (34)

$$Y_1 = \Sigma^{-1} C_{11}, \ Y_2 = \Sigma^{-1} C_{12}, \ Y_3 = \tilde{D}_{13} \Omega^{-1},$$
(35)

$$A_{21}Y_1 + A_{22}Y_3 + \Sigma Z_1 = C_{21}, (36)$$

$$A_{23}Y_2 + A_{24}Y_4 = C_{24}, \ Y_3B_{22} + Y_4B_{24} = \tilde{D}_{24}, \tag{37}$$

$$A_{21}Y_2 + A_{22}Y_4 + \Sigma Z_2 = C_{22}, \ Y_3B_{21} + Y_4B_{23} + Z_3\Omega = \tilde{D}_{23},$$
(38)

$$Y_1\Omega = \tilde{D}_{11}, A_{23}Y_1 + A_{24}Y_3 = C_{23}, Y_1B_{22} + Y_2B_{24} = \tilde{D}_{22}, Y_1B_{21} + Y_2B_{23} + Z_1\Omega = \tilde{D}_{21}.$$
(39)

Plugging (35) into (36), we obtain

$$Z_1 = \Sigma^{-1} (C_{21} - A_{21} \Sigma^{-1} C_{11} - A_{22} \tilde{D}_{13} \Omega^{-1}).$$
(40)

Substituting (35) into (37) yields

$$A_{24}Y_4 = C_{24} - A_{23}\Sigma^{-1}C_{12}, \ Y_4B_{24} = \tilde{D}_{24} - \tilde{D}_{13}\Omega^{-1}B_{22}.$$
 (41)

By Lemma 3.1, Eq (41) with respect to Y_4 has a solution if and only if

$$E_{A_{24}}(C_{24} - A_{23}\Sigma^{-1}C_{12}) = 0, (\tilde{D}_{24} - \tilde{D}_{13}\Omega^{-1}B_{22})F_{B_{24}} = 0,$$

$$A_{24}(\tilde{D}_{24} - \tilde{D}_{13}\Omega^{-1}B_{22}) = (C_{24} - A_{23}\Sigma^{-1}C_{12})B_{24},$$
(42)

in this case, the general solution is

$$Y_4 = J_4 + F_{A_{24}} W_3 E_{B_{24}},\tag{43}$$

where W_3 is an arbitrary matrix and

$$J_4 = A_{24}^{\dagger} (C_{24} - A_{23} \Sigma^{-1} C_{12}) + F_{A_{24}} (\tilde{D}_{24} - \tilde{D}_{13} \Omega^{-1} B_{22}) B_{24}^{\dagger}.$$
 (44)

Inserting (35) and (43) into (38) yields

$$Z_2 = J_5 - \Sigma^{-1} A_{22} F_{A_{24}} W_3 E_{B_{24}}, \ Z_3 = J_6 - F_{A_{24}} W_3 E_{B_{24}} B_{23} \Omega^{-1},$$
(45)

where

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$$J_5 = \Sigma^{-1} C_{22} - \Sigma^{-1} A_{21} \Sigma^{-1} C_{12} - \Sigma^{-1} A_{22} J_4, J_6 = \tilde{D}_{23} \Omega^{-1} - \tilde{D}_{13} \Omega^{-1} B_{21} \Omega^{-1} - J_4 B_{23} \Omega^{-1}.$$
 (46)

Plugging (35) and (40) into (39), we arrive at

$$C_{11}\Omega = \Sigma \tilde{D}_{11}, \ A_{23}\Sigma^{-1}C_{11}\Omega + A_{24}\tilde{D}_{13} = C_{23}\Omega, \ C_{11}B_{22} + C_{12}B_{24} = \Sigma \tilde{D}_{22}, \tag{47}$$

$$C_{11}B_{21} + C_{12}B_{23} + (C_{21} - A_{21}\Sigma^{-1}C_{11} - A_{22}\tilde{D}_{13}\Omega^{-1})\Omega = \Sigma\tilde{D}_{21}.$$
(48)

We observe from (34), (47) and (48) that

$$\begin{split} &C_{13} = 0, C_{14} = 0 \Leftrightarrow P_2^{\mathsf{T}} C_1 U = 0 \Leftrightarrow P_2 P_2^{\mathsf{T}} C_1 = 0 \Leftrightarrow E_{A_1} C_1 = 0, \\ &\tilde{D}_{12} = 0, \tilde{D}_{14} = 0 \Leftrightarrow Q^{\mathsf{T}} D_1 V_2 = 0 \Leftrightarrow D_1 V_2 V_2^{\mathsf{T}} = 0 \Leftrightarrow D_1 F_{B_1} = 0, \\ &C_{11} \Omega = \Sigma \tilde{D}_{11} \Leftrightarrow P_1^{\mathsf{T}} C_1 U_1 \Omega = \Sigma Q_1^{\mathsf{T}} D_1 V_1 \Leftrightarrow P_1 P_1^{\mathsf{T}} C_1 U_1 \Omega V_1^{\mathsf{T}} = P_1 \Sigma Q_1^{\mathsf{T}} D_1 V_1 V_1^{\mathsf{T}} \Leftrightarrow C_1 B_1 = A_1 D_1, \\ &A_{23} \Sigma^{-1} C_{11} \Omega + A_{24} \tilde{D}_{13} = C_{23} \Omega \Leftrightarrow P_2^{\mathsf{T}} A_2 Q_2 Q_2^{\mathsf{T}} D_1 V_1 \Omega + P_2^{\mathsf{T}} A_2 Q_2 Q_2^{\mathsf{T}} D_1 V_1 = P_2^{\mathsf{T}} C_2 U_1 \Omega \\ \Leftrightarrow P_2 P_2^{\mathsf{T}} A_2 Q_1 \Sigma^{-1} P_1^{\mathsf{T}} C_1 U_1 \Omega V_1^{\mathsf{T}} + P_2 P_2^{\mathsf{T}} A_2 Q_2 Q_2^{\mathsf{T}} D_1 V_1 V_1^{\mathsf{T}} = P_2 P_2^{\mathsf{T}} C_2 U_1 \Omega V_1^{\mathsf{T}} \\ \Leftrightarrow E_{A_1} (A_2 D_1 - C_2 B_1) = 0, \\ &C_{11} B_{22} + C_{12} B_{24} = \Sigma \tilde{D}_{22} \Leftrightarrow P_1^{\mathsf{T}} C_1 U_1 U_1^{\mathsf{T}} B_2 V_2 + P_1^{\mathsf{T}} C_1 U_2 U_2^{\mathsf{T}} B_2 V_2 Z_2^{\mathsf{T}} \Theta (A_1 D_2 - C_1 B_2) F_{B_1} = 0, \\ &C_{11} B_{21} + C_{12} B_{23} + (C_{21} - A_{21} \Sigma^{-1} C_{11} - A_{22} \tilde{D}_{13} \Omega^{-1}) \Omega = \Sigma \tilde{D}_{21} \\ \Leftrightarrow P_1 P_1^{\mathsf{T}} C_1 U_1 U_1^{\mathsf{T}} B_2 V_1 + P_1^{\mathsf{T}} C_1 U_2 U_2^{\mathsf{T}} B_2 V_1 V_2^{\mathsf{T}} = P_1 \Sigma Q_1^{\mathsf{T}} D_2 V_2^{\mathsf{T}} O (A_1 D_2 - C_1 B_2) F_{B_1} = 0, \\ &C_{11} B_{21} + C_{12} B_{23} + (C_{21} - A_{21} \Sigma^{-1} C_{11} - A_{22} \tilde{D}_{13} \Omega^{-1}) \Omega = \Sigma \tilde{D}_{21} \\ \Leftrightarrow P_1^{\mathsf{T}} C_1 U_1 U_1^{\mathsf{T}} B_2 V_1 + P_1^{\mathsf{T}} C_1 U_2 U_2^{\mathsf{T}} B_2 V_1 + P_1^{\mathsf{T}} C_2 U_1 \Omega - P_1^{\mathsf{T}} A_2 Q_2 Q_2^{\mathsf{T}} D_1 V_1 + P_1^{\mathsf{T}} A_2 Q_1 \Sigma^{-1} P_1^{\mathsf{T}} C_1 U_1 \Omega \\ &= \Sigma Q_1^{\mathsf{T}} D_2 V_1 \\ \Leftrightarrow P_1 P_1^{\mathsf{T}} C_1 U_1 U_1^{\mathsf{T}} B_2 V_1 V_1^{\mathsf{T}} + P_1 P_1^{\mathsf{T}} C_1 U_2 U_2^{\mathsf{T}} B_2 V_1 V_1^{\mathsf{T}} \\ \Rightarrow P_1 P_1^{\mathsf{T}} A_2 Q_1 \Sigma^{-1} P_1^{\mathsf{T}} C_1 U_1 \Omega V_1^{\mathsf{T}} = P_1 \Sigma Q_1^{\mathsf{T}} D_2 V_1 V_1^{\mathsf{T}} \\ \Leftrightarrow C_2 B_1 - A_2 D_1 = A_1 D_2 - C_1 B_2, \\ &A_{24} (\tilde{D}_{24} - \tilde{D}_{13} \Omega^{-1} B_{22}) = (C_{24} - A_{23} \Sigma^{-1} C_{12}) B_{24} \\ \Leftrightarrow P_2^{\mathsf{T}} A_2 Q_2 (Q_2^{\mathsf{T}} D_2 V_2 - Q_2^{\mathsf{T}} D_1 V_1 \Omega^{-1} U_1^{\mathsf{T}} B_2 V_2) = (P_2^{\mathsf{T}} C_2 U_2 - P_2^{\mathsf{T}} A_2 Q_1 \Sigma^{-1} P_1^{\mathsf{T}} C_1 U_2) U_2^{\mathsf{T}} B_2 V_2 V_2^{\mathsf{T}} \\ \Leftrightarrow P_2 P_$$

Summing up the discussion above, we can reach a result as follows.

Theorem 3.1. Assume that dual matrices $A = A_1 + \varepsilon A_2$, $B = B_1 + \varepsilon B_2$, $C = C_1 + \varepsilon C_2$ and $D = D_1 + \varepsilon D_2$, where $A_i \in \mathbb{R}^{m \times p}$, $B_i \in \mathbb{R}^{q \times n}$, $C_i \in \mathbb{R}^{m \times q}$, $D_i \in \mathbb{R}^{p \times n}$ (i = 1, 2). Let the SVDs of A_1 , B_1 be given by (4) and the partitions of the matrices $U^{\top}B_2V$, $P^{\top}A_2Q$, $P^{\top}C_1U$, $P^{\top}C_2U$, $Q^{\top}D_1V$ and $Q^{\top}D_2V$ be given by (6), (29) and (30). Then Eq (2) is solvable if and only if

$$E_{A_{24}}(C_{24} - A_{23}\Sigma^{-1}C_{12}) = 0, (\tilde{D}_{24} - \tilde{D}_{13}\Omega^{-1}B_{22})F_{B_{24}} = 0, E_{A_1}(C_2B_2 - A_2D_2)F_{B_1} = 0,$$

$$E_{A_1}C_1 = 0, D_1F_{B_1} = 0, C_1B_1 = A_1D_1, C_2B_1 - A_2D_1 = A_1D_2 - C_1B_2,$$

$$E_{A_1}(C_2B_1 - A_2D_1) = 0, (C_1B_2 - A_1D_2)F_{B_1} = 0.$$
(49)

In this case, the solution set of Problem II(a) can be expressed as

$$X_{1} = Q \begin{bmatrix} \Sigma^{-1} C_{11} & \Sigma^{-1} C_{12} \\ \tilde{D}_{13} \Omega^{-1} & J_{4} + F_{A_{24}} W_{3} E_{B_{24}} \end{bmatrix} U^{\top},$$
(50)

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$$X_{2} = Q \begin{bmatrix} \Sigma^{-1}(C_{21} - A_{21}\Sigma^{-1}C_{11} - A_{22}\tilde{D}_{13}\Omega^{-1}) & J_{5} - \Sigma^{-1}A_{22}F_{A_{24}}W_{3}E_{B_{24}} \\ J_{6} - F_{A_{24}}W_{3}E_{B_{24}}B_{23}\Omega^{-1} & Z_{4} \end{bmatrix} U^{\top},$$
(51)

with J_4 , J_5 , J_6 being given by (44) and (46), and W_3 , Z_4 are arbitrary matrices.

Now, for $\forall X \in S_2$, we have

•

$$\begin{split} \|X\|_{\mathrm{D}}^{2} &= \|X_{1}\|_{\mathrm{F}}^{2} + \|X_{2}\|_{\mathrm{F}}^{2} = \|Q^{\top}X_{1}U\|_{\mathrm{F}}^{2} + \|Q^{\top}X_{2}U\|_{\mathrm{F}}^{2} \\ &= \|\Sigma^{-1}C_{12}\|_{\mathrm{F}}^{2} + \|\tilde{D}_{13}\Omega^{-1}\|_{\mathrm{F}}^{2} + \|J_{4} + F_{A_{24}}W_{3}E_{B_{24}}\|_{\mathrm{F}}^{2} + \|\Sigma^{-1}(C_{21} - A_{21}\Sigma^{-1}C_{11} - A_{22}\tilde{D}_{13}\Omega^{-1})\|_{\mathrm{F}}^{2} \\ &+ \|\Sigma^{-1}C_{11}\|_{\mathrm{F}}^{2} + \|Z_{4}\|_{\mathrm{F}}^{2} + \|J_{5} - \Sigma^{-1}A_{22}F_{A_{24}}W_{3}E_{B_{24}}\|_{\mathrm{F}}^{2} + \|J_{6} - F_{A_{24}}W_{3}E_{B_{24}}B_{23}\Omega^{-1}\|_{\mathrm{F}}^{2}. \end{split}$$

It is easily seen that $||X||_D$ is minimized if and only if $Z_4 = 0$, and

$$\Psi(W_3) = \|J_4 + F_{A_{24}}W_3E_{B_{24}}\|_F^2 + \|J_5 - \Sigma^{-1}A_{22}F_{A_{24}}W_3E_{B_{24}}\|_F^2 + \|J_6 - F_{A_{24}}W_3E_{B_{24}}B_{23}\Omega^{-1}\|_F^2 = \min.$$
(52)

The minimization problem (52) is equivalent to

$$\Psi(W_3) = \operatorname{tr}[J_4^{\mathsf{T}}J_4 + E_{B_{24}}W_3^{\mathsf{T}}F_{A_{24}}W_3E_{B_{24}} + J_6^{\mathsf{T}}J_6 + J_5^{\mathsf{T}}J_5 + 2J_4^{\mathsf{T}}F_{A_{24}}W_3E_{B_{24}} + \Omega^{-1}B_{23}^{\mathsf{T}}E_{B_{24}}W_3^{\mathsf{T}}F_{A_{24}}W_3E_{B_{24}}B_{23}\Omega^{-1} - 2J_5^{\mathsf{T}}\Sigma^{-1}A_{22}F_{A_{24}}W_3E_{B_{24}} + E_{B_{24}}W_3^{\mathsf{T}}F_{A_{24}}A_{22}^{\mathsf{T}}\Sigma^{-1}\Sigma^{-1}A_{22}F_{A_{24}}W_3E_{B_{24}} - 2J_6^{\mathsf{T}}F_{A_{24}}W_3E_{B_{24}}B_{23}\Omega^{-1}] = \min.$$

Therefore, $\Psi(W_3)$ is minimized if and only if $\frac{\partial \Psi(W_3)}{\partial W_3} = 0$, which implies that

$$(F_{A_{24}} + 2L_2^{\mathsf{T}}L_2)W_3E_{B_{24}} + F_{A_{24}}W_3(E_{B_{24}} + 2L_1^{\mathsf{T}}L_1) = J_7,$$
(53)

where

$$J_7 = 2F_{A_{24}}A_{22}^{\top}\Sigma^{-1}J_5E_{B_{24}} - 2F_{A_{24}}J_4E_{B_{24}} + 2F_{A_{24}}J_6\Omega^{-1}B_{23}^{\top}E_{B_{24}}.$$
 (54)

By applying the Kronecker product and stretching function, it follows from (53) that

$$\Pi \operatorname{vec}(W_3) = \operatorname{vec}(J_7), \tag{55}$$

where

$$\tilde{\Pi} = E_{B_{24}} \otimes (F_{A_{24}} + 2L_2^{\mathsf{T}}L_2) + (E_{B_{24}} + 2L_1^{\mathsf{T}}L_1) \otimes F_{A_{24}}.$$
(56)

Thus, by solving the equation of (55), we can get the solution of W_3 , and then, substituting $Z_4 = 0$ into (50) and (51), we have the following result.

Theorem 3.2. If the conditions (49) are satisfied, then the solution of Problem II(b) is given by

$$\hat{X} = \hat{X}_1 + \varepsilon \hat{X}_2,$$

where

$$\hat{X}_{1} = Q \begin{bmatrix} \Sigma^{-1} C_{11} & \Sigma^{-1} C_{12} \\ \tilde{D}_{13} \Omega^{-1} & J_{4} + F_{A_{24}} W_{3} E_{B_{24}} \end{bmatrix} U^{\mathsf{T}},$$
(57)

$$\hat{X}_2 = Q \begin{bmatrix} \Sigma^{-1} J_4 & J_5 - \Sigma^{-1} A_{22} F_{A_{24}} W_3 E_{B_{24}} \\ J_6 - F_{A_{24}} W_3 E_{B_{24}} B_{23} \Omega^{-1} & 0 \end{bmatrix} U^{\top},$$
(58)

with W_3 being given by (55).

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Based on Theorems 3.1 and 3.2, we can describe an algorithm to solve Problems II(a) and II(b) as follows.

Algorithm 2

- 1: Input matrices A_i , B_i , C_i and D_i , i = 1, 2.
- 2: Compute the SVDs of the matrices A_1 , B_1 on the basis of (4).
- 3: Compute the partitions of the matrices $U^{\top}B_2V$, $P^{\top}A_2Q$, $P^{\top}C_1U$, $P^{\top}C_2U$, $Q^{\top}D_1V$ and $Q^{\top}D_2V$ as in (6), (29) and (30).
- 4: If the conditions (49) are satisfied, then continue, otherwise, Problem II(a) has no solution, and stop.
- 5: Calculate Y_1 , Y_2 , Y_3 , Z_1 according to (35) and (40).
- 6: Compute J_4 , J_5 , J_6 in the light of (44) and (46), respectively.
- 7: Compute $L_1 = \Omega^{-1} B_{23}^{\top} E_{B_{24}}, L_2 = \Sigma^{-1} A_{22} F_{A_{24}}$ and compute $J_7, \tilde{\Pi}$ on the basis of (54) and (56). Then, calculate W_3 by the equation of (55).
- 8: Compute Y_4 , Z_2 , Z_3 by (43) and (45), respectively.
- 9: Compute \hat{X}_1, \hat{X}_2 as in (57) and (58), and compute the unique solution $\hat{X} = \hat{X}_1 + \varepsilon \hat{X}_2$ of Problem II(b).

Example 3.1. Let m = 8, n = 9, p = 6, q = 7, and the matrices $A_1, A_2, B_1, B_2, C_1, C_2, D_1, D_2$ be given by

$A_1 =$	1.5526	0.5212	0.3530	0.8404	0.8114	0.9951]			
	0.9989	0.4602	0.2003	0.7210	0.6974	0.8205				
	0.8074	0.2321	0.1919	0.3809	0.3673	0.4613				
	0.8202	0.4030	0.1591	0.6283	0.6080	0.7100				
	1.3044	0.4626	0.2913	0.7418	0.7164	0.8717	,			
	0.3896	0.1243	0.0900	0.2016	0.1945	0.2404				
	0.5914	0.1352	0.1481	0.2287	0.2201	0.2877				
İ	1.1147	0.3696	0.2544	0.5968	0.5761	0.7078				
	1.3349	1.3005	1.2315	1.6795	0.9943	0.6848]			
	0.2255	0.2339	0.1988	0.3186	0.2004	0.1629				
	1.1621	0.8659	0.7576	1.2040	0.6569	0.5779				
4 _	0.8167	0.5498	0.5099	0.7510	0.3773	0.3187				
$A_2 =$	1.0789	1.1427	1.0635	1.4789	0.9083	0.6311	,			
	0.8222	0.8648	0.8445	1.0876	0.6537	0.4073				
	0.8829	0.9214	0.8787	1.1768	0.7123	0.4716				
l	0.7993	0.8774	0.8067	1.1399	0.7115	0.5010	j			
ſ	0.0446	0.5327	0.2550	0.3699	0.1876	0.4685	0.3517	0.3635	0.4113	
	0.0686	0.8480	0.3537	0.5734	0.3122	0.7693	0.5629	0.5937	0.6660	
	0.0622	0.5497	0.6125	0.4845	0.1037	0.3269	0.3427	0.2741	0.3489	
$B_1 =$	0.0768	0.8412	0.5404	0.6246	0.2609	0.6782	0.5475	0.5342	0.6198	,
	0.0674	0.7740	0.4264	0.5538	0.2583	0.6557	0.5078	0.5120	0.5855	
	0.0836	0.7374	0.8254	0.6511	0.1381	0.4369	0.4596	0.3666	0.4673	
l	0.0751	0.7531	0.6204	0.5994	0.1984	0.5459	0.4822	0.4387	0.5253	

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<i>B</i> ₂ =	1.5570 0.6003 0.4270 0.9920 1.1495 0.8428 0.3822	1.3803 0.4898 0.3812 0.7537 0.8863 0.7449 0.3760	0.9873 0.6809 0.7336 0.7935 1.1764 1.0424 0.3597	1.2107 0.6524 0.8245 0.5618 0.9592 1.1799 0.5329	1.4629 0.7224 0.7820 0.8125 1.1759 1.1993 0.5291	0.7456 0.5833 0.6548 0.6460 0.9975 0.8982 0.2940	1.7241 0.8351 0.7463 1.1754 1.5157 1.2326 0.4984	1.0553 0.7540 0.7206 1.0193 1.4030 1.0525 0.3125	1.4645 0.7619 0.7354 0.9963 1.3448 1.1568 0.4601	,
<i>C</i> ₁ =	3.0365 2.2789 1.4777 1.9367 2.6154 0.7451 0.9917 2.1681	2.3514 1.9450 1.0882 1.6841 2.0611 0.5677 0.6766 1.6724	2.9403 2.1991 1.4332 1.8676 2.5310 0.7219 0.9641 2.0997	2.3560 1.9504 1.0897 1.6890 2.0654 0.5687 0.6771 1.6755	2.5632 1.9141 1.2503 1.6250 2.2058 0.6295 0.8420 1.8305	3.2087 2.3853 1.5686 2.0232 2.7592 0.7886 1.0596 2.2919	1.7223 1.3086 0.8332 1.1148 1.4866 0.4218 0.5544 1.2292],		
<i>C</i> ₂ =	6.1864 2.6188 4.0004 3.4300 5.3391 2.9339 3.3261 4.2214	6.6590 3.1458 4.0163 3.6096 5.8266 2.9822 3.4272 4.6594	6.3910 2.1945 4.1155 3.2798 5.5391 3.4557 3.7918 4.3246	6.6562 2.9235 3.9904 3.4675 5.8421 3.1650 3.5852 4.6529	7.0540 2.8930 4.3817 3.7839 6.0971 3.4627 3.8690 4.8070	4.9153 1.7231 3.3262 2.6145 4.2667 2.6087 2.8896 3.3348	5.8256 2.9235 3.4919 3.3363 5.0249 2.4604 2.8395 4.0138],		
<i>D</i> ₁ =	0.2502 0.2940 0.2053 0.2183 0.2591 0.2240	2.5297 3.1995 2.0830 2.3651 2.6380 2.4095	2.0412 2.0962 1.6636 1.5697 2.0893 1.6345	2.0009 2.3877 1.6426 1.7710 2.0750 1.8146	0.6774 0.9817 0.5622 0.7205 0.7165 0.7249	1.8530 2.5612 1.5335 1.8842 1.9500 1.9038	1.6223 2.0798 1.3368 1.5363 1.6940 1.5631	1.4861 2.0200 1.2287 1.4874 1.5611 1.5051	1.7738 2.3483 1.4643 1.7316 1.8583 1.7564	,
<i>D</i> ₂ =	3.7771 4.2620 2.6236 2.7051 3.2666 3.3398	6.1417 6.1961 3.9042 4.4944 4.7381 4.9858	5.4631 5.8150 3.8013 3.9692 4.5846 4.5069	6.0497 5.9103 3.9615 4.2023 4.7936 4.7309	5.0201 5.1259 3.5098 3.5063 4.2062 4.0859	5.2884 5.3279 3.8863 4.0012 4.3934 4.2395	6.5640 6.9569 4.6084 4.8152 5.4957 5.4946	5.5281 5.9395 4.2401 4.2480 4.8244 4.6454	6.36686.59894.48044.68455.27695.2375	

It is easy to verity that the conditions (49) hold:

$$\begin{split} \|E_{A_{24}}(C_{24}-A_{23}\Sigma^{-1}C_{12})\|_{\rm F} &= 1.9816\times 10^{-15}, \\ \|(\tilde{D}_{24}-\tilde{D}_{13}\Omega^{-1}B_{22})F_{B_{24}}\|_{\rm F} &= 1.9246\times 10^{-15}, \\ \|E_{A_1}C_1\|_{\rm F} &= 1.8511\times 10^{-15}, \\ \|D_1F_{B_1}\|_{\rm F} &= 1.8127\times 10^{-15}, \\ \|C_1B_1-A_1D_1\|_{\rm F} &= 5.2525\times 10^{-15}, \\ \|E_{A_1}(C_2B_2-A_2D_2)F_{B_1}\|_{\rm F} &= 1.8738\times 10^{-14}, \\ \|E_{A_1}(C_2B_1-A_2D_1)\|_{\rm F} &= 1.0763\times 10^{-14}, \\ \|(C_1B_2-A_1D_2)F_{B_1}\|_{\rm F} &= 1.2767\times 10^{-14}, \\ \|C_2B_1-A_2D_1-A_1D_2+C_1B_2\|_{\rm F} &= 2.0019\times 10^{-14}. \end{split}$$

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$\hat{X}_1 =$	[0.9156	0.1339	0.7289	0.0326	0.6173	0.8750	0.5012]	
	0.7731	0.7486	0.7637	0.7232	0.8637	0.1447	0.4641	
	0.0944	0.1119	0.8647	0.4961	0.8063	0.5116	0.0583	
	0.2780	0.6347	0.4317	0.4945	0.5737	0.0871	0.6869	,
	0.3531	0.4732	0.8058	0.9207	0.1796	0.8899	0.0477	
	0.6618	0.8004	0.0890	0.5937	0.2432	0.8028	0.0660]	
$\hat{X}_2 =$	[0.6273	0.7506	0.4251	0.6879	0.6432	0.2920	0.8319]	
	0.4378	0.5937	0.5012	0.6175	0.5620	0.5559	0.6663	
	0.4188	0.6570	0.0717	0.4749	0.4988	0.0348	0.3547	
	0.5093	0.6965	0.2578	0.5517	0.5655	0.1585	0.5424	•
	0.4335	0.5651	0.2867	0.4791	0.4770	0.2045	0.5139	
	0.4439	0.5362	0.3786	0.5012	0.4774	0.2950	0.6010	

The absolute errors are estimated by

$$\begin{aligned} \|A_1 \hat{X}_1 - C_1\|_{\rm F} &= 4.8343 \times 10^{-15}, \ \|A_2 \hat{X}_1 + A_1 \hat{X}_2 - C_2\|_{\rm F} &= 6.7203 \times 10^{-15}, \\ \|\hat{X}_1 B_1 - D_1\|_{\rm F} &= 1.0333 \times 10^{-14}, \ \|\hat{X}_1 B_2 + \hat{X}_2 B_1 - D_2\|_{\rm F} &= 2.8801 \times 10^{-14}, \end{aligned}$$

which implies that \hat{X} is the solution of Problem II(b).

4. Conclusions

Solving dual matrix equations is often required in kinematic analysis and sensor calibration. In the present paper, the dual matrix equations (1) and (2) are first factorized as two real matrix equations by separating them into the real parts and dual parts. By applying the SVDs of A_1 and B_1 , we have obtained the solvability conditions and the general solutions of Problems I(a) and II(a) (see Theorems 2.1 and 3.1). Further, based on the results of Theorems 2.1 and 3.1, the unique minimum-norm solutions $\hat{X} = \hat{X}_1 + \varepsilon \hat{X}_2$ of Problems I(b) and II(b) have achieved (see Theorems 2.2 and 3.2).

Acknowledgments

The authors would like to express their gratitude to the anonymous reviewers for their valuable suggestions and comments that improved the presentation of this manuscript.

Use of AI tools declaration

The authors declare that we have not used Artificial Intelligence (AI) tools in the creation of this article.

Conflict of interest

The authors declare that there is no conflict of interest.

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