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## Article

# The General Solution to a System of Tensor Equations over the Split Quaternion Algebra with Applications

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**Abstract:** In this paper, considering a general solution to a system of tensor equations within the framework of the split quaternion algebra, we establish the necessary and sufficient conditions for the existence of the general solution. Furthermore, we derive the general solution for the system of tensor equations when it is solvable. In addition, we provide algorithms and numerical examples to illustrate the results presented in this study. To demonstrate the applicability of our findings, we give the algorithms and examples for the encryption and decryption of color videos. Finally, we make a conclusion.

**Keywords:** split quaternion algebra; tensor equations; color images encryption and decryption

**MSC:** 15A09; 15A24; 15B33

## 1. Introduction

In 1843, William Rowan Hamilton introduced the concept of real quaternions [1] defined by

$$\mathbb{H} = \{q = q_0 + q_1i + q_2j + q_3k : i^2 = j^2 = k^2 = -1, ijk = -1, q_0, q_1, q_2, q_3 \in \mathbb{R}\},$$

where  $\mathbb{H}$  and  $\mathbb{R}$  separately denoted the set of real quaternions and the set of real numbers. Quaternions have been utilized in numerous areas, including the analysis of quaternion random signals [2], color image processing [3,4] and face recognition [5]. As theory of quaternion evolved, James Cockle introduced the concept of split quaternions [6] in 1849:

$$\mathbb{Q}_S = \{q = q_0 + q_1i + q_2j + q_3k : q_0, q_1, q_2, q_3 \in \mathbb{R}\},$$

where the imaginary identities  $i, j, k$  satisfied

$$i^2 = -1, j^2 = k^2 = 1, ijk = 1,$$

$$ij = -ji = k, jk = -kj = -i, ki = -ik = j.$$

Split quaternions have diverse applications across various fields. In physics, they are used to model spacetime symmetries and rotations in relativity [7]. In computer graphics, split quaternions assist in efficiently handling 3D rotations and transformations [8]. Split quaternions also have applications in control theory, signal processing, and robotics, where they help in representing and solving complex spatial relationships and transformations [9].

Tensors, as mathematical constructs that generalize scalars, vectors, and matrices to higher dimensions, are extensively applied across a wide range of scientific and engineering domains due to their capability to manage complex data structures. In computer vision and image processing, tensors are used to represent multi-dimensional image data, with convolution operations in neural networks being performed as tensor operations [10]. In physics, tensors are fundamental in general relativity

for describing spacetime geometry [11] and in electromagnetism for unifying electric and magnetic fields [12]. In machine learning, tensors are core data structures in frameworks like TensorFlow and PyTorch, playing a crucial role in optimization and model training [13,14]. They are also essential in data analysis for dimensionality reduction and feature extraction, particularly within recommendation systems [15]. The multi-dimensional nature of tensors makes them powerful tools for handling complex data, and their significance continues to grow with advancements in computational power and algorithms.

$\mathcal{A} = (a_{i_1 \dots i_N})(1 \leq i_j \leq I_j, j = 1, \dots, N)$  presents an  $N$ -dimensional tensor, which is a natural extension of the matrix. For  $N = 1$ ,  $\mathcal{A}$  is a vector. For  $N = 2$ ,  $\mathcal{A}$  is a matrix. In recent decades, numerous researchers have made significant contributions to the study of tensors and quaternions. For example, Qi et al. [16–18] researched the spectral theory of tensor and tensor eigenvalues. Wang et al. [19–21] examined systems of tensor equations within the framework of quaternion algebra. Chen et al. [22,23] derived some solvability conditions and formulated expressions of the general  $\eta$ -(anti)-Hermitian solution to constrained Sylvester-type matrix equations over the generalized commutative quaternion algebra  $\mathbb{S}_g$ . Gao et al. [24] established the necessary and sufficient conditions for the existence of (anti)- $\eta$ -Hermitian solutions to a system of matrix equations over split quaternion algebra. Xie et al. [25] proposed the BiCG algorithm for solving the minimal Frobenius norm solution of generalized Sylvester tensor equation over the quaternions. Yang et al. [26] researched a system of tensor equations over the dual split quaternion algebra with an application.

To the best of our knowledge, there is relatively little research on tensor equations over split quaternions at present. Motivated by the aforementioned research and the existing applications of tensor equation systems over split quaternion algebra, we aim to study the solvability conditions and find the general solution for the following system of split quaternion tensor equations

$$\begin{cases} \mathcal{A}_1 *_{\mathcal{N}} \mathcal{X}_1 = \mathcal{P}_1, \\ \mathcal{X}_1 *_{\mathcal{N}} \mathcal{B}_1 = \mathcal{Q}_1, \\ \mathcal{A}_2 *_{\mathcal{N}} \mathcal{X}_2 = \mathcal{P}_2, \\ \mathcal{X}_2 *_{\mathcal{N}} \mathcal{B}_2 = \mathcal{Q}_2, \\ \dots \\ \mathcal{A}_n *_{\mathcal{N}} \mathcal{X}_n = \mathcal{P}_n, \\ \mathcal{X}_n *_{\mathcal{N}} \mathcal{B}_n = \mathcal{Q}_n, \\ \sum_{t=1}^n \mathcal{E}_t *_{\mathcal{N}} \mathcal{X}_t *_{\mathcal{N}} \mathcal{F}_t = \mathcal{H}, \end{cases} \quad (1)$$

where  $\mathcal{A}_t, \mathcal{P}_t, \mathcal{B}_t, \mathcal{Q}_t, \mathcal{E}_t, \mathcal{F}_t (t = 1, \dots, n)$  and  $\mathcal{H}$  are given quaternion tensors and  $\mathcal{X}_t (t = 1, \dots, n)$  are unknown split quaternion tensors.

This paper is organized as follows: In Section 2, some basic theories and relevant properties of quaternion tensors are reviewed. In Section 3, we derive some practical necessary and sufficient conditions for the existence of the general solution to the system (1) over  $\mathbb{Q}_S$ . Furthermore, the formula of the general solution to the system (1) is also given when it is solvable. Especially, the unique solution is given when the uniqueness condition is satisfied. In Section 4, two algorithms and one numerical example are given in order to illustrate our results. In Section 5, we provide four algorithms and two examples for color videos encryption and decryption. In Section 6, we make a conclusion to summarize this paper.

In subsequent chapters, the symbols  $\mathbb{R}^{m \times n}$ ,  $\mathbb{C}^{m \times n}$ ,  $\mathbb{Q}_S^n$ ,  $\mathbb{Q}_S^{m \times n}$  are used to represent the set of all  $m \times n$  real matrices, all  $m \times n$  complex matrices, all  $n$  dimensional split quaternion column vectors and all  $m \times n$  split quaternion matrices. The symbol  $r(A)$  denotes the rank of the matrix  $A$ . Denote  $I$  as the identity matrix and  $O$  as the zero matrix with an appropriate size.  $\bar{A}$ ,  $A^T$ ,  $A^*$ ,  $A^\dagger$  denote the conjugate, the transpose, the conjugate transpose and the Moore-Penrose inverse of the matrix  $A$ , respectively. Let  $\mathbb{R}^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_N}$ ,  $\mathbb{C}^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_N}$ , and  $\mathbb{Q}_S^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_N}$  represent the set of all corresponding dimensional real tensors, complex tensors, and split quaternion tensors, respectively.

## 2. Preliminaries

This chapter primarily introduces split quaternion tensors, the Einstein product of split quaternion tensors, the complex representation and related properties of split quaternion matrices and so on. These concepts lay the groundwork for deriving the main conclusions presented in the subsequent chapter.

**Definition 1.** [27] For two split quaternion tensors  $\mathcal{A} = (a_{i_1 \dots i_N j_1 \dots j_L}) \in \mathbb{Q}_S^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_L}$  and  $\mathcal{B} = (b_{j_1 \dots j_L k_1 \dots k_M}) \in \mathbb{Q}_S^{J_1 \times \dots \times J_L \times K_1 \times \dots \times K_M}$ , the Einstein product of  $\mathcal{A}$  and  $\mathcal{B}$  is defined by the operation  $*_L$  via

$$(\mathcal{A} *_L \mathcal{B})_{i_1 \dots i_N k_1 \dots k_M} = \sum_{j_1 \dots j_L} a_{i_1 \dots i_N j_1 \dots j_L} b_{j_1 \dots j_L k_1 \dots k_M},$$

where  $\mathcal{A} *_L \mathcal{B} \in \mathbb{Q}_S^{I_1 \times \dots \times I_N \times K_1 \times \dots \times K_M}$ .

The Einstein product of tensors is closely related to matrix multiplication. We introduce a mapping that establishes an isomorphic relationship between tensors with respect to the Einstein product and matrices with respect to the general matrix product.

**Definition 2.** [28] There exists an isomorphic mapping  $f$  from  $\mathbb{Q}_S^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_N}$  to  $\mathbb{Q}_S^{I \times J}$  denoted as

$$\begin{aligned} f : \mathbb{Q}_S^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_N} &\rightarrow \mathbb{Q}_S^{I \times J} \\ a_{i_1 \dots i_N j_1 \dots j_N} &\mapsto a_{[i_N + \sum_{k=1}^{N-1} (i_k - 1) \prod_{s=k+1}^N I_s] [j_N + \sum_{k=1}^{N-1} (j_k - 1) \prod_{s=k+1}^N J_s]} \end{aligned}$$

where  $I = I_1 \cdots I_N$  and  $J = J_1 \cdots J_N$ .

It is apparent that  $f$  is a one-to-one mapping. Therefore, the inverse mapping  $f^{-1} : \mathbb{Q}_S^{I \times J} \rightarrow \mathbb{Q}_S^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_N}$  exists. Subsequently, we prove that  $f$  is an isomorphic mapping.

**Proof.** For  $\mathcal{A} = (a_{i_1 \dots i_N j_1 \dots j_L}) \in \mathbb{Q}_S^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_L}$  and  $\mathcal{B} = (b_{j_1 \dots j_L k_1 \dots k_M}) \in \mathbb{Q}_S^{J_1 \times \dots \times J_L \times K_1 \times \dots \times K_M}$ , we need to prove  $f(\mathcal{A} *_L \mathcal{B}) = f(\mathcal{A}) \cdot f(\mathcal{B})$ , where  $*_L$  and  $\cdot$  denote the Einstein product and matrix product, respectively.

Through the definition of the Einstein product, then

$$(\mathcal{A} *_L \mathcal{B})_{i_1 \dots i_N k_1 \dots k_M} = \sum_{j_1 \dots j_L} a_{i_1 \dots i_N j_1 \dots j_L} b_{j_1 \dots j_L k_1 \dots k_M}.$$

By direct computation, we obtain

$$\begin{aligned} f(\mathcal{A} *_L \mathcal{B}) &= f\left(\sum_{j_1 \dots j_L} a_{i_1 \dots i_N j_1 \dots j_L} b_{j_1 \dots j_L k_1 \dots k_M}\right) \\ &= \sum_{j_1 \dots j_L} a_{[i_N + \sum_{k=1}^{N-1} (i_k - 1) \prod_{s=k+1}^N I_s] [j_L + \sum_{k=1}^{L-1} (j_k - 1) \prod_{s=k+1}^L J_s]} b_{[j_L + \sum_{k=1}^{L-1} (j_k - 1) \prod_{s=k+1}^L J_s] [k_M + \sum_{k=1}^{M-1} (k_k - 1) \prod_{s=k+1}^M K_s]} \\ &= \sum_{j=1}^{J_1 \dots J_N} f(\mathcal{A})_{ij} f(\mathcal{B})_{jk} \\ &= (f(\mathcal{A}) \cdot f(\mathcal{B})). \end{aligned}$$

Therefore,  $f$  is isomorphic.  $\square$

As  $f$  is an isomorphic mapping, utilizing  $f$  can convert split quaternion tensors into matrices during operations. In the following, we present the complex representation of split quaternions and split quaternion matrices.

For any split quaternion  $q = q_0 + q_1 i + q_2 j + q_3 k \in \mathbb{Q}_S$ , it can be uniquely expressed as  $q = c_1 + c_2 j$  where  $c_1 = q_0 + q_1 i$  and  $c_2 = q_2 + q_3 i$  are complex numbers, which is known as the complex



representation of split quaternion. Noting that  $c_1j = j\bar{c}_1$  and  $c_2j = j\bar{c}_2$ , we know the algebra of split quaternion is a noncommutative algebra.

**Definition 3.** The complex representation of a split quaternion is  $q = d_1 + d_2j \in \mathbb{Q}_S$  where  $d_1 = q_0 + q_1i, d_2 = q_2 + q_3i \in \mathbb{C}$ . Use the symbol  $g$  to denote it as

$$g(q) = \begin{pmatrix} d_1 & d_2 \\ \bar{d}_2 & \bar{d}_1 \end{pmatrix} \in \mathbb{C}^{2 \times 2}.$$

Similarly, for any given split quaternion matrix  $A$ , the complex representation of matrix  $A = A_1 + A_2j \in \mathbb{Q}_S^{m \times n}$  where  $A_1, A_2 \in \mathbb{C}^{m \times n}$ , is denoted by  $G$  as

$$G(A) = \begin{pmatrix} A_1 & A_2 \\ \bar{A}_2 & \bar{A}_1 \end{pmatrix} \in \mathbb{C}^{2m \times 2n}. \quad (2)$$

Obviously,  $G(A)$  and  $A$  correspond one to one. Moreover, considering about its properties, we can summarise this as the following proposition.

**Proposition 1.** If  $A, B \in \mathbb{Q}_S^{n \times n}$  and  $G$  is defined in Definition 3, then

- (a)  $A = B$  if and only if  $G(A) = G(B)$ ;
- (b)  $G(A + B) = G(A) + G(B)$ ;
- (c)  $G(I_n) = I_{2n}$ ;
- (d)  $G(AB) = G(A)G(B)$ .

**Proof.** (a), (b) and (c) hold obviously so we just need to prove (d). By computation, on the one hand

$$AB = (A_1 + A_2j)(B_1 + B_2j) = A_1B_1 + A_2\bar{B}_2 + (A_1B_2 + A_2\bar{B}_1)j.$$

On the other hand

$$\begin{aligned} G(AB) &= \begin{pmatrix} A_1B_1 + A_2\bar{B}_2 & A_1B_2 + A_2\bar{B}_1 \\ \bar{A}_1\bar{B}_2 + \bar{A}_2B_1 & \bar{A}_1B_1 + \bar{A}_2B_2 \end{pmatrix} \\ &= \begin{pmatrix} A_1 & A_2 \\ \bar{A}_2 & \bar{A}_1 \end{pmatrix} \begin{pmatrix} B_1 & B_2 \\ \bar{B}_2 & \bar{B}_1 \end{pmatrix} \\ &= G(A)G(B). \end{aligned}$$

□

In the following, we explore the structure of the operator  $\text{vec}(UXV)$ .

**Definition 4.** [29] For any  $A \in \mathbb{Q}_S^{m \times n}$ ,  $A$  can be uniquely expressed as  $A = A_{11} + A_{12}i + A_{21}j + A_{22}k$  where  $A_{11}, A_{12}, A_{21}, A_{22} \in \mathbb{R}^{m \times n}$ . It can also be uniquely expressed as  $A = A_1 + A_2j$  where  $A_1 = A_{11} + A_{12}i, A_2 = A_{21} + A_{22}i \in \mathbb{C}^{m \times n}$ . According to this, we define

$$A = A_1 + A_2j \cong \Phi_A = [A_1, A_2] \quad (3)$$

where the symbol  $\cong$  denotes an equivalence relation. Define that

$$\hat{A}_1 = \begin{bmatrix} \text{Re}(A_1) & \text{Im}(A_1) \end{bmatrix}, \hat{A} = \begin{bmatrix} \text{Re}(A_1) & \text{Im}(A_1) & \text{Re}(A_2) & \text{Im}(A_2) \end{bmatrix}.$$

Obviously the following is true.

$$\text{vec}(\hat{A}_1) = \begin{bmatrix} \text{vec}(\text{Re}(A_1)) \\ \text{vec}(\text{Im}(A_1)) \end{bmatrix}, \text{vec}(\hat{A}) = \begin{bmatrix} \text{vec}(\text{Re}(A_1)) \\ \text{vec}(\text{Im}(A_1)) \\ \text{vec}(\text{Re}(A_2)) \\ \text{vec}(\text{Im}(A_2)) \end{bmatrix}.$$

Then we focus on main properties related to  $\Phi_A$  as follows.

**Proposition 2.** Let  $E, F \in \mathbb{Q}_S^{m \times n}$ ,  $H \in \mathbb{Q}_S^{n \times s}$  and  $p \in \mathbb{R}$ .  $\Phi$  is defined in Definition 4. Then

- (a)  $E = F$  if and only if  $\Phi_E = \Phi_F$ ,
- (b)  $\Phi_{E+F} = \Phi_E + \Phi_F$ ,  $\Phi_{pE} = p\Phi_E$ ,
- (c)  $\Phi_{EH} = \Phi_E G(H)$ .

**Proof.** Obviously, (a) and (b) hold so we only need to prove (c). By calculating, we have

$$EH = (E_1 + E_2j)(H_1 + H_2j) = E_1H_1 + E_2\overline{H_2} + (E_1H_2 + E_2\overline{H_1})j.$$

Then

$$\begin{aligned} \Phi_{EH} &= [E_1H_1 + E_2\overline{H_2}, E_1H_2 + E_2\overline{H_1}] \\ &= [E_1, E_2] \begin{bmatrix} H_1 & H_2 \\ \overline{H_2} & \overline{H_1} \end{bmatrix} \\ &= \Phi_E G(H). \end{aligned}$$

□

Based on this proposition, we then investigate the structure of the operator  $\text{vec}(UXV)$ .

**Theorem 1.** [29] Suppose that  $U = U_1 + U_2j \in \mathbb{Q}_S^{m \times n}$ ,  $V = V_1 + V_2j \in \mathbb{Q}_S^{s \times t}$  and  $X = X_1 + X_2j \in \mathbb{Q}_S^{n \times s}$ , where  $U_1, U_2 \in \mathbb{C}^{m \times n}$ ,  $V_1, V_2 \in \mathbb{C}^{s \times t}$  and  $X_1, X_2 \in \mathbb{C}^{n \times s}$ . Then

$$\text{vec}(\Phi_{UXV}) = \left[ G(V)^T \otimes U_1, G(Vj)^* \otimes U_2 \right] \begin{bmatrix} \text{vec}(\Phi_X) \\ \text{vec}(\Phi_{jXj}) \end{bmatrix}. \quad (4)$$

The result on  $\text{vec}(\Phi_{UXV})$  in Theorem 1 about the complex representation of split quaternion matrices product  $UXV$ , is an essential tool for transforming the split quaternion matrix equation into the complex matrix equation.

**Lemma 1.** For  $B = B_1 + B_2j \in \mathbb{Q}_S^{r \times t}$ ,  $B_1, B_2 \in \mathbb{C}^{r \times t}$ . Let

$$K_t = \begin{bmatrix} I_{rt} & iI_{rt} & 0 & 0 \\ 0 & 0 & I_{rt} & iI_{rt} \\ I_{rt} & -iI_{rt} & 0 & 0 \\ 0 & 0 & I_{rt} & -iI_{rt} \end{bmatrix},$$

then

$$\begin{bmatrix} \text{vec}(\Phi_B) \\ \text{vec}(\Phi_{jBj}) \end{bmatrix} = K_t \text{vec}(\hat{B}).$$

**Proof.** According to the given conditions, it follows that

$$\begin{aligned} \begin{bmatrix} \text{vec}(\Phi_B) \\ \text{vec}(\Phi_{jBj}) \end{bmatrix} &= \begin{bmatrix} \text{vec}(B_1) \\ \text{vec}(B_2) \\ \text{vec}(\overline{B_1}) \\ \text{vec}(\overline{B_2}) \end{bmatrix} \\ &= \begin{bmatrix} I_{rt} & iI_{rt} & 0 & 0 \\ 0 & 0 & I_{rt} & iI_{rt} \\ I_{rt} & -iI_{rt} & 0 & 0 \\ 0 & 0 & I_{rt} & -iI_{rt} \end{bmatrix} \begin{bmatrix} \text{vec}(\text{Re}(B_1)) \\ \text{vec}(\text{Im}(B_1)) \\ \text{vec}(\text{Re}(B_2)) \\ \text{vec}(\text{Im}(B_2)) \end{bmatrix} \\ &= K_t \text{vec}(\hat{B}). \end{aligned}$$

□

By Theorem 1 and Lemma 1, we can obtain the following corollary.

**Corollary 1.** Suppose  $U = U_1 + U_2j \in \mathbb{Q}_S^{m \times n}$ ,  $V = V_1 + V_2j \in \mathbb{Q}_S^{s \times t}$  and  $X = X_1 + X_2j \in \mathbb{Q}_S^{n \times s}$  where  $U_1, U_2 \in \mathbb{C}^{m \times n}$ ,  $V_1, V_2 \in \mathbb{C}^{s \times t}$  and  $X_1, X_2 \in \mathbb{C}^{n \times s}$ . Then

$$\text{vec}(\Phi_{UXV}) = [G(V)^T \otimes U_1, G(Vj)^* \otimes U_2] K_s \text{vec}(\hat{X}) \quad (5)$$

**Lemma 2.** [30] With  $A \in \mathbb{R}^{m \times n}$  and  $b \in \mathbb{R}^m$ , the matrix equation  $Ax = b$  has a solution  $x \in \mathbb{R}^n$  if and only if

$$AA^\dagger b = b. \quad (6)$$

In that case, it has the general solution given by

$$x = A^\dagger b + (I_n - A^\dagger A)y, \quad (7)$$

where  $y \in \mathbb{R}^n$  is an arbitrary vector. Especially, it has a unique solution  $x = A^\dagger b$  if  $\text{rank}(A) = n$ .

### 3. Solution of System (1)

From the previous discussion, we now turn our attention to solving the system of split quaternion matrix equations (1). For convenience, we define some useful notations that will be used in the following section. For  $t = 1, \dots, n$ , we denote  $A_t = f(\mathcal{A}_t)$ ,  $B_t = f(\mathcal{B}_t)$ ,  $P_t = f(\mathcal{P}_t)$ ,  $Q_t = f(\mathcal{Q}_t)$ ,  $E_t = f(\mathcal{E}_t)$ ,  $F_t = f(\mathcal{F}_t)$ ,  $H = f(\mathcal{H})$ , then  $A_t = A_{t1} + A_{t2}j$ ,  $P_t = P_{t1} + P_{t2}j \in \mathbb{Q}_S^{m \times r}$ ,  $B_t = B_{t1} + B_{t2}j$ ,  $Q_t = Q_{t1} + Q_{t2}j \in \mathbb{Q}_S^{r \times k}$ ,  $E_t = E_{t1} + E_{t2}j \in \mathbb{Q}_S^{l \times r}$ ,  $F_t = F_{t1} + F_{t2}j \in \mathbb{Q}_S^{r \times w}$  and  $H \in \mathbb{Q}_S^{l \times w}$ , where  $m = I_1 \cdots I_N$ ,  $r = J_1 \cdots J_N$ ,  $k = K_1 \cdots K_N$ ,  $l = L_1 \cdots L_N$ ,  $w = W_1 \cdots W_N$ . Set

$$L_1 = \begin{bmatrix} G(F_1)^T \otimes E_{11} & G(F_1j)^* \otimes E_{12} \\ G(I)^T \otimes A_{11} & G(Ij)^* \otimes A_{12} \\ G(B_1)^T \otimes I & G(B_1j)^* \otimes I \\ 0 & 0 \\ 0 & 0 \\ 0 & 0 \\ \vdots & \vdots \\ 0 & 0 \end{bmatrix} K_r, L_2 = \begin{bmatrix} G(F_2)^T \otimes E_{21} & G(F_2j)^* \otimes E_{22} \\ 0 & 0 \\ 0 & 0 \\ G(I)^T \otimes A_{21} & G(Ij)^* \otimes A_{22} \\ G(B_2)^T \otimes I & G(B_2j)^* \otimes I \\ 0 & 0 \\ \vdots & \vdots \\ 0 & 0 \end{bmatrix} K_r,$$

$$\cdots, L_n = \begin{bmatrix} G(F_n)^T \otimes E_{n1} & G(F_n)^* \otimes E_{n2} \\ 0 & 0 \\ 0 & 0 \\ 0 & 0 \\ 0 & 0 \\ \vdots & \vdots \\ 0 & 0 \\ 0 & 0 \\ G(I)^T \otimes A_{n1} & G(Ij)^* \otimes A_{n2} \\ G(B_n)^T \otimes I & G(B_nj)^* \otimes I \end{bmatrix} K_r, Z = \begin{bmatrix} \text{vec}(\Phi_H) \\ \text{vec}(\Phi_{P_1}) \\ \text{vec}(\Phi_{Q_1}) \\ \text{vec}(\Phi_{P_2}) \\ \text{vec}(\Phi_{Q_2}) \\ \vdots \\ \text{vec}(\Phi_{P_{n-1}}) \\ \text{vec}(\Phi_{Q_{n-1}}) \\ \text{vec}(\Phi_{P_n}) \\ \text{vec}(\Phi_{Q_n}) \end{bmatrix}.$$

For  $t = 1, \dots, n$ , we suppose

$$L_{t1} = \text{Re}(L_t), L_{t2} = \text{Im}(L_t), V_1 = [L_{11}, L_{21}, \dots, L_{n1}], V_2 = [L_{12}, L_{22}, \dots, L_{n2}] \quad (8)$$

and

$$Z_1 = \begin{bmatrix} \text{vec}(\text{Re}(\Phi_H)) \\ \text{vec}(\text{Re}(\Phi_{P_1})) \\ \text{vec}(\text{Re}(\Phi_{Q_1})) \\ \text{vec}(\text{Re}(\Phi_{P_2})) \\ \text{vec}(\text{Re}(\Phi_{Q_2})) \\ \vdots \\ \text{vec}(\text{Re}(\Phi_{P_n})) \\ \text{vec}(\text{Re}(\Phi_{Q_n})) \end{bmatrix}, Z_2 = \begin{bmatrix} \text{vec}(\text{Im}(\Phi_H)) \\ \text{vec}(\text{Im}(\Phi_{P_1})) \\ \text{vec}(\text{Im}(\Phi_{Q_1})) \\ \text{vec}(\text{Im}(\Phi_{P_2})) \\ \text{vec}(\text{Im}(\Phi_{Q_2})) \\ \vdots \\ \text{vec}(\text{Im}(\Phi_{P_n})) \\ \text{vec}(\text{Im}(\Phi_{Q_n})) \end{bmatrix}, z = \begin{bmatrix} Z_1 \\ Z_2 \end{bmatrix}. \quad (9)$$

Evidently, the following equation

$$\begin{bmatrix} V_1 \\ V_2 \end{bmatrix} \begin{bmatrix} \text{vec}(\hat{X}_1) \\ \text{vec}(\hat{X}_2) \\ \vdots \\ \text{vec}(\hat{X}_n) \end{bmatrix} = z \quad (10)$$

holds. Now according to Lemma 2 and equation (10), we can give the expression of general solution to the system (1).

**Theorem 2.** For  $t = 1, \dots, n$ , we suppose  $\mathcal{A}_t, \mathcal{P}_t \in \mathbb{Q}_S^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_N}$ ,  $\mathcal{B}_t, \mathcal{Q}_t \in \mathbb{Q}_S^{J_1 \times \dots \times J_N \times K_1 \times \dots \times K_N}$ ,  $\mathcal{E}_t \in \mathbb{Q}_S^{L_1 \times \dots \times L_N \times I_1 \times \dots \times I_N}$ ,  $\mathcal{F}_t \in \mathbb{Q}_S^{J_1 \times \dots \times J_N \times W_1 \times \dots \times W_N}$  and  $\mathcal{H} \in \mathbb{Q}_S^{L_1 \times \dots \times L_N \times W_1 \times \dots \times W_N}$ . For  $t = 1, \dots, n$ , we denote  $A_t = f(\mathcal{A}_t)$ ,  $B_t = f(\mathcal{B}_t)$ ,  $P_t = f(\mathcal{P}_t)$ ,  $Q_t = f(\mathcal{Q}_t)$ ,  $E_t = f(\mathcal{E}_t)$ ,  $F_t = f(\mathcal{F}_t)$ ,  $H = f(\mathcal{H})$ , then  $A_t = A_{t1} + A_{t2}j$ ,  $P_t = P_{t1} + P_{t2}j \in \mathbb{Q}_S^{m \times r}$ ,  $B_t = B_{t1} + B_{t2}j$ ,  $Q_t = Q_{t1} + Q_{t2}j \in \mathbb{Q}_S^{r \times k}$ ,  $E_t = E_{t1} + E_{t2}j \in \mathbb{Q}_S^{l \times r}$ ,  $F_t = F_{t1} + F_{t2}j \in \mathbb{Q}_S^{r \times w}$  and  $H \in \mathbb{Q}_S^{l \times w}$ , where  $m = I_1 \cdots I_N$ ,  $r = J_1 \cdots J_N$ ,  $k = K_1 \cdots K_N$ ,  $l = L_1 \cdots L_N$ ,  $w = W_1 \cdots W_N$ . Meanwhile,  $V_1, V_2, Z_1, Z_2, z$  are defined in equation (8) and equation (9). Then the system (1) has a set of solutions  $\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n \in \mathbb{Q}_S^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_N}$  if and only if

$$\begin{bmatrix} V_1 \\ V_2 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}^\dagger z = z. \quad (11)$$

In this case, the set of general solutions can be expressed as

$$\mathcal{X} = \left\{ [f^{-1}(X_1), f^{-1}(X_2), \dots, f^{-1}(X_n)] \begin{bmatrix} \text{vec}(\hat{X}_1) \\ \text{vec}(\hat{X}_2) \\ \vdots \\ \text{vec}(\hat{X}_n) \end{bmatrix} = \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}^\dagger z + \left[ I_{4nr^2} - \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}^\dagger \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} \right] y \right\}, \quad (12)$$

where  $y$  is an arbitrary vector with appropriate size. Furthermore, the system of split quaternion matrix equations (1) has a unique solution if and only if

$$\text{rank} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = 4nr^2. \quad (13)$$

After calculation, we have

$$\mathcal{X} = \left\{ [f^{-1}(X_1), f^{-1}(X_2), \dots, f^{-1}(X_n)] \middle| \begin{bmatrix} \text{vec}(\hat{X}_1) \\ \text{vec}(\hat{X}_2) \\ \vdots \\ \text{vec}(\hat{X}_n) \end{bmatrix} = \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}^{\dagger} z \right\}. \quad (14)$$

**Proof.** According to the Corollary 1 and Theorem 2, it follows that

$$\begin{aligned} \text{system(1)} &\Leftrightarrow \begin{cases} \sum_{t=1}^n f(\mathcal{E}_t *_{\mathcal{N}} \mathcal{X}_t *_{\mathcal{N}} \mathcal{F}_t) = f(\mathcal{H}), \\ f(\mathcal{A}_1 *_{\mathcal{N}} \mathcal{X}_1) = f(\mathcal{P}_1), f(\mathcal{X}_1 *_{\mathcal{N}} \mathcal{B}_1) = f(\mathcal{Q}_1), \\ f(\mathcal{A}_2 *_{\mathcal{N}} \mathcal{X}_2) = f(\mathcal{P}_2), f(\mathcal{X}_2 *_{\mathcal{N}} \mathcal{B}_2) = f(\mathcal{Q}_2), \\ \dots, \dots \\ f(\mathcal{A}_n *_{\mathcal{N}} \mathcal{X}_n) = f(\mathcal{P}_n), f(\mathcal{X}_n *_{\mathcal{N}} \mathcal{B}_n) = f(\mathcal{Q}_n), \end{cases} \\ &\Leftrightarrow \begin{cases} \sum_{t=1}^n E_t X_t F_t = H, \\ A_1 X_1 = P_1, X_1 B_1 = Q_1, \\ A_2 X_2 = P_2, X_2 B_2 = Q_2, \\ \dots \\ A_n X_n = P_n, X_n B_n = Q_n, \end{cases} \\ &\Leftrightarrow \begin{cases} \sum_{t=1}^n \Phi_{E_t X_t F_t} = \Phi_H, \\ \Phi_{A_1 X_1} = \Phi_{P_1}, \Phi_{X_1 B_1} = \Phi_{Q_1}, \\ \Phi_{A_2 X_2} = \Phi_{P_2}, \Phi_{X_2 B_2} = \Phi_{Q_2}, \\ \dots, \dots \\ \Phi_{A_n X_n} = \Phi_{P_n}, \Phi_{X_n B_n} = \Phi_{Q_n}, \end{cases} \\ &\Leftrightarrow L_1 \text{vec}(\hat{X}_1) + L_2 \text{vec}(\hat{X}_2) + \dots + L_n \text{vec}(\hat{X}_n) = Z, \\ &\Leftrightarrow (\text{Re}(L_1) + i \text{Im}(L_1)) \text{vec}(\hat{X}_1) + (\text{Re}(L_2) + i \text{Im}(L_2)) \text{vec}(\hat{X}_2) + \\ &\quad \dots + (\text{Re}(L_n) + i \text{Im}(L_n)) \text{vec}(\hat{X}_n) = Z_1 + i Z_2, \\ &\Leftrightarrow \begin{bmatrix} \text{Re}(L_1) & \text{Re}(L_2) & \dots & \text{Re}(L_n) \\ \text{Im}(L_1) & \text{Im}(L_2) & \dots & \text{Im}(L_n) \end{bmatrix} \begin{bmatrix} \text{vec}(\hat{X}_1) \\ \text{vec}(\hat{X}_2) \\ \vdots \\ \text{vec}(\hat{X}_n) \end{bmatrix} = \begin{bmatrix} Z_1 \\ Z_2 \end{bmatrix}, \\ &\Leftrightarrow \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} \begin{bmatrix} \text{vec}(\hat{X}_1) \\ \text{vec}(\hat{X}_2) \\ \vdots \\ \text{vec}(\hat{X}_n) \end{bmatrix} = z, \\ &\Leftrightarrow [\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n] = [f^{-1}(X_1), f^{-1}(X_2), \dots, f^{-1}(X_n)]. \end{aligned}$$



By Lemma 2, we can obtain that the system (1) has a solution if and only if the equation (11) holds. That is to say

$$\begin{bmatrix} \text{vec}(\hat{X}_1) \\ \text{vec}(\hat{X}_2) \\ \vdots \\ \text{vec}(\hat{X}_n) \end{bmatrix} = \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}^+ z + \left[ I_{4nr^2} - \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}^+ \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} \right] y,$$

which is expressed as the set (12). Moreover, if the equation (11) holds, then the system (1) has a unique solution if and only if

$$\begin{bmatrix} V_1 \\ V_2 \end{bmatrix}^+ \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = I_{4nr^2}.$$

Equivalently, the equation (13) holds and the set (14) is the unique solution.  $\square$

In the next part, we consider the Moore-Penrose generalized inverse of the column block matrix. Let

$$\begin{aligned} p &= 2wl + 2nmr + 2nkr, \\ K &= (I_{4nr^2} - V_1^+ V_1) V_2^T, \\ W &= (I_p + (I_p - K^+ K) V_2 V_1^+ V_1^{+T} V_2^T (I_p - K^+ K))^{-1}, \\ J &= K^+ + (I_p - K^+ K) W V_2 V_1^+ V_1^{+T} (I_{4nr^2} - V_2^T K^+), \\ \theta_1 &= I_p - V_1 V_1^+ + V_1^{+T} V_2^T W (I_p - K^+ K) V_2 V_1^+, \\ \theta_2 &= -V_1^{+T} V_2^T (I_p - K^+ K) W, \\ \theta_3 &= (I_p - K^+ K) W, \end{aligned} \quad (15)$$

where  $n$  is same as the system (1) and  $m, k, r, l, w$  are same in Theorem 2. From the results in [31], we have

$$\begin{bmatrix} V_1 \\ V_2 \end{bmatrix}^+ = \begin{bmatrix} V_1^+ - J^T V_2 V_1^+ & J^T \end{bmatrix}, \quad \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}^+ \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} = V_1^+ V_1 + K^+ K. \quad (16)$$

$$I_{2p} - \begin{bmatrix} V_1 \\ V_2 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}^+ = \begin{bmatrix} \theta_1 & \theta_2 \\ \theta_2^T & \theta_3 \end{bmatrix}. \quad (17)$$

**Corollary 2.** [32] The system of split quaternion matrix equations (1) has a solution  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n]$  if and only if

$$\begin{bmatrix} \theta_1 & \theta_2 \\ \theta_2^T & \theta_3 \end{bmatrix} z = 0. \quad (18)$$

In this case, the set of general solution of system (1) can be expressed as  $\mathcal{X} =$

$$\left\{ [f^{-1}(X_1), f^{-1}(X_2), \dots, f^{-1}(X_n)] \left| \begin{bmatrix} \text{vec}(\hat{X}_1) \\ \text{vec}(\hat{X}_2) \\ \vdots \\ \text{vec}(\hat{X}_n) \end{bmatrix} = \begin{bmatrix} V_1^+ - J^T V_2 V_1^+ & J^T \end{bmatrix} z + (I_{4nr^2} - V_1^+ V_1 - K^+ K) y \right. \right\}, \quad (19)$$

where  $X_1, X_2, \dots, X_n \in \mathbb{Q}_S^{r \times r}$  and  $y$  is an arbitrary vector with appropriate size. Furthermore, if the equation (18) holds, then the system (1) has a unique solution  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n] \in \mathcal{X}$  if and only if the equation (13) holds. In this case,

$$\mathcal{X} = \left\{ [f^{-1}(X_1), f^{-1}(X_2), \dots, f^{-1}(X_n)] \left| \begin{bmatrix} \text{vec}(\hat{X}_1) \\ \text{vec}(\hat{X}_2) \\ \vdots \\ \text{vec}(\hat{X}_n) \end{bmatrix} = \begin{bmatrix} V_1^+ - J^T V_2 V_1^+ & J^T \end{bmatrix} z \right. \right\}, \quad (20)$$

**Corollary 3.** Let the condition be satisfied in Corollary 2. Then the optimization problem

$$\min_{[X_1, X_2, \dots, X_n] \in X} (\|\Phi_{X_1}\|^2 + \|\Phi_{X_2}\|^2 + \dots + \|\Phi_{X_n}\|^2)$$

has a unique minimizer  $[X_{1w}, X_{2w}, \dots, X_{nw}] = [f(\mathcal{X}_{1w}), f(\mathcal{X}_{2w}), \dots, f(\mathcal{X}_{nw})]$  that satisfies

$$\begin{bmatrix} \text{vec}(\hat{X}_{1w}) \\ \text{vec}(\hat{X}_{2w}) \\ \vdots \\ \text{vec}(\hat{X}_{nw}) \end{bmatrix} = \begin{bmatrix} V_1^\dagger - J^T V_2 V_1^\dagger & J^T \end{bmatrix} z. \quad (21)$$

**Proof.** From the solution set (19), we can see that the solution set  $\mathcal{X}$  is a nonempty closed convex set. Hence,

$$\begin{aligned} & \min_{[X_1, X_2, \dots, X_n] \in X} (\|\Phi_{X_1}\|^2 + \|\Phi_{X_2}\|^2 + \dots + \|\Phi_{X_n}\|^2) \\ &= \min_{[X_1, X_2, \dots, X_n] \in X} (\|\hat{X}_1\|^2 + \|\hat{X}_2\|^2 + \dots + \|\hat{X}_n\|^2) \\ &= \min_{[X_1, X_2, \dots, X_n] \in X} (\|\text{vec}(\hat{X}_1)\|^2 + \|\text{vec}(\hat{X}_2)\|^2 + \dots + \|\text{vec}(\hat{X}_n)\|^2) \\ &= \min_{[X_1, X_2, \dots, X_n] \in X} \left\| \begin{bmatrix} \text{vec}(\hat{X}_1) \\ \text{vec}(\hat{X}_2) \\ \vdots \\ \text{vec}(\hat{X}_n) \end{bmatrix} \right\|^2. \end{aligned}$$

By Corollary 2, we have  $\begin{bmatrix} \text{vec}(\hat{X}_{1w}) \\ \text{vec}(\hat{X}_{2w}) \\ \vdots \\ \text{vec}(\hat{X}_{nw}) \end{bmatrix}$  is in the form (21).  $\square$

#### 4. Numerical Exemplification

In this section, we give two numerical algorithms and two numerical examples to solve the system (1).

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##### Algorithm 1 General Solution of System (1)

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(1) **Input the factors:**  $n$ .

(2) **Input the tensors:**  $A_1, A_2, \dots, A_n, P_1, P_2, \dots, P_n \in \mathbb{Q}_S^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_N}$ ,  
 $B_1, B_2, \dots, B_n, Q_1, Q_2, \dots, Q_n \in \mathbb{Q}_S^{J_1 \times \dots \times J_N \times K_1 \times \dots \times K_N}$ ,  $\mathcal{E}_1, \mathcal{E}_2, \dots, \mathcal{E}_n \in \mathbb{Q}_S^{L_1 \times \dots \times L_N \times J_1 \times \dots \times J_N}$ ,  
 $\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_n \in \mathbb{Q}_S^{J_1 \times \dots \times J_N \times W_1 \times \dots \times W_N}$  and  $\mathcal{H} \in \mathbb{Q}_S^{L_1 \times \dots \times L_N \times W_1 \times \dots \times W_N}$ .

(3) Calculate the marices  $A_1, A_2, \dots, A_n, P_1, P_2, \dots, P_n \in \mathbb{Q}_S^{m \times r}$ ,  $B_1, B_2, \dots, B_n, Q_1, Q_2, \dots, Q_n \in \mathbb{Q}_S^{r \times k}$ ,  $E_1, E_2, \dots, E_n \in \mathbb{Q}_S^{l \times r}$ ,  $F_1, F_2, \dots, F_n \in \mathbb{Q}_S^{r \times w}$  and  $H \in \mathbb{Q}_S^{l \times w}$ , where  $m = I_1 \dots I_N$ ,  $r = J_1 \dots J_N$ ,  $k = K_1 \dots K_N$ ,  $l = L_1 \dots L_N$ ,  $w = W_1 \dots W_N$ .

(4) Calculate  $V_1, V_2$  by (8) and  $z$  by (9).

(5) If both the equation (11) and the equation (13) hold, then calculate the unique solution  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n] \in \mathcal{X}$  by the equation (14).

(6) If the equation (11) holds and the equation (13) falses, then calculate  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n] \in \mathcal{X}$  according to (12).

(7) **Output:**  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n]$ .

---

**Algorithm 2** General Solution of System (1)(1) **Input the factors:**  $n$ .(2) **Input the tensors:**  $\mathcal{A}_1, \mathcal{A}_2, \dots, \mathcal{A}_n, \mathcal{P}_1, \mathcal{P}_2, \dots, \mathcal{P}_n \in \mathbb{Q}_S^{I_1 \times \dots \times I_N \times J_1 \times \dots \times J_N}$ ,  
 $\mathcal{B}_1, \mathcal{B}_2, \dots, \mathcal{B}_n, \mathcal{Q}_1, \mathcal{Q}_2, \dots, \mathcal{Q}_n \in \mathbb{Q}_S^{J_1 \times \dots \times J_N \times K_1 \times \dots \times K_N}$ ,  $\mathcal{E}_1, \mathcal{E}_2, \dots, \mathcal{E}_n \in \mathbb{Q}_S^{L_1 \times \dots \times L_N \times J_1 \times \dots \times J_N}$ ,  
 $\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_n \in \mathbb{Q}_S^{J_1 \times \dots \times J_N \times W_1 \times \dots \times W_N}$  and  $\mathcal{H} \in \mathbb{Q}_S^{L_1 \times \dots \times L_N \times W_1 \times \dots \times W_N}$ .(3) Calculate the matrices  $A_1, A_2, \dots, A_n, P_1, P_2, \dots, P_n \in \mathbb{Q}_S^{m \times r}$ ,  $B_1, B_2, \dots, B_n, Q_1, Q_2, \dots, Q_n \in \mathbb{Q}_S^{r \times k}$ ,  $E_1, E_2, \dots, E_n \in \mathbb{Q}_S^{l \times r}$ ,  $F_1, F_2, \dots, F_n \in \mathbb{Q}_S^{r \times w}$  and  $H \in \mathbb{Q}_S^{l \times w}$ , where  $m = I_1 \dots I_N$ ,  $r = J_1 \dots J_N$ ,  $k = K_1 \dots K_N$ ,  $l = L_1 \dots L_N$ ,  $w = W_1 \dots W_N$ .(4) Calculate  $V_1, V_2$  by (8),  $z$  by (9) and  $p, K, W, J, \theta_1, \theta_2, \theta_3$  by (15).(5) If both the equation (18) and the equation (13) hold, then calculate the unique solution  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n] \in \mathcal{X}$  by the equation (21).(6) If the equation (18) holds and the equation (13) fails, then calculate  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n] \in \mathcal{X}$  according to (19).(7) **Output:**  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n]$ .**Example 1.** Given the split quaternion tensors:

$$\begin{aligned}
\mathcal{A}_1(:, :, 1, 1) &= \begin{bmatrix} i+k & 1 \\ 0 & 0.5i+j+2k \\ i+2j+k & i \end{bmatrix}, \mathcal{A}_1(:, :, 1, 2) = \begin{bmatrix} i & 2i+j+2k \\ 2i+j+k & 4k \\ 1 & j+k \end{bmatrix}, \\
\mathcal{A}_1(:, :, 2, 1) &= \begin{bmatrix} i & i+j+4k \\ i+k & 2j \\ 0 & i+2j+2k \end{bmatrix}, \mathcal{A}_1(:, :, 2, 2) = \begin{bmatrix} 1 & j+5k \\ 2i+j+3k & i \\ 0 & i+j+k \end{bmatrix}, \\
\mathcal{A}_1(:, :, 3, 1) &= \begin{bmatrix} 0.5i & 0 \\ i+2j+k & 3i \\ 0 & 3i+j+k \end{bmatrix}, \mathcal{A}_1(:, :, 3, 2) = \begin{bmatrix} 2i+3k & 0 \\ 2j & i+0.5j+4k \\ 1 & i \end{bmatrix}, \\
\mathcal{P}_1(:, :, 1, 1) &= \begin{bmatrix} 7+4.25i-3.5j+3k & 11+3.25i+2.75j+8.5k \\ 6-2.75i+2.5j+1.75k & 9.5+4.5i-j+8.5k \\ 2+2j+2k & -2+4.75i-3j+4k \end{bmatrix}, \\
\mathcal{P}_1(:, :, 1, 2) &= \begin{bmatrix} -1.25+7.5i-2.5j+8.5k & 12.5+i+8j+5.75k \\ 7.5-2.25j+4.25k & 13.5+11.5i+1j \\ 5+1.25i+0.5j+4k & 4+3i+5.25j+3.25k \end{bmatrix}, \\
\mathcal{P}_1(:, :, 2, 1) &= \begin{bmatrix} -0.5+9.5i+8k & 11+11i+0.5k \\ 13+5i+0.5j+8.5k & 20.5+22i+10j+5.5k \\ 1+3.5i+6j+1k & 5+9i+2.5j+12.5k \end{bmatrix}, \\
\mathcal{P}_1(:, :, 2, 2) &= \begin{bmatrix} -2+8.5i-4.5j+4k & 24+8i+2j+3.5k \\ 9+8i+8k & 21.5+10.75i+4.25j+8.5k \\ 2.5+i+3.5j+6.5k & 10+2i+6j+3k \end{bmatrix}, \\
\mathcal{P}_1(:, :, 3, 1) &= \begin{bmatrix} 7i+7.5k & 1+4i+12j+3k \\ 11j+3k & 9+11i+9j+4.5k \\ i+3j-k & -3+10i+7j+2k \end{bmatrix}, \\
\mathcal{P}_1(:, :, 3, 2) &= \begin{bmatrix} -1.5+9.5i-2j+7k & 13+7i+4j+7k \\ 2+4i & 9.5+7i+16j+5k \\ 6+7i+3j+6k & 6+2i+4j+8k \end{bmatrix}, \\
\mathcal{B}_1(1, 1, :, :) &= \begin{bmatrix} i & i+4j+k \\ 0 & i+3j+k \end{bmatrix}, \mathcal{B}_1(1, 2, :, :) = \begin{bmatrix} k & 5i+j+k \\ 0 & i+j+0.5k \end{bmatrix},
\end{aligned}$$

$$\begin{aligned}
\mathcal{B}_1(2,1,::) &= \begin{bmatrix} i+3j & 1 \\ j+k & i+j+3k \end{bmatrix}, \mathcal{B}_1(2,2,::) = \begin{bmatrix} 2i+k & 1+k \\ 0 & 2+k \end{bmatrix}, \\
\mathcal{B}_1(3,1,::) &= \begin{bmatrix} 2j & i+k \\ 1+k & 2i+j \end{bmatrix}, \mathcal{B}_1(3,2,::) = \begin{bmatrix} 2i+k & 2i+3k \\ j & i \end{bmatrix}, \\
\mathcal{Q}_1(:, :, 1, 1) &= \begin{bmatrix} 3+2j+7k & 8.5+16i+12.75j+0.5k \\ 1+5.25i+11j+8k & 6-2.5i+j-5.5k \\ 5+13i+13j+3k & 1.5+7i+10j-0.5k \end{bmatrix}, \\
\mathcal{Q}_1(:, :, 1, 2) &= \begin{bmatrix} 2-0.5i+6.5j-6.5k & 7.75+2.5i+16.75j+11.25k \\ 1+9.25i+2j+7.25k & 3+9i+16.5j-k \\ 4+14i+8j+14k & 5.5+13.5i+4.5j+2k \end{bmatrix}, \\
\mathcal{Q}_1(:, :, 2, 1) &= \begin{bmatrix} 5-4j+3k & 5+2i+0.5j+3.5k \\ 3+6i-j+2k & 3-i-j \\ 4-3i+5j+6k & 9+2i-2j+6k \end{bmatrix}, \\
\mathcal{Q}_1(:, :, 2, 2) &= \begin{bmatrix} -0.25-1.5i-5.5j+4k & 18.25+4i+14.375j+11.75k \\ 9.75i+0.75j+9.75k & 3+3i+9j-1.25k \\ 8+8i+12j+13.5k & 7.5+12i+9.5j+2.25k \end{bmatrix}, \\
\mathcal{A}_2(:, :, 1, 1) &= \begin{bmatrix} i+k & 1 \\ 0 & 0.5i+j+2k \\ i+2j+k & i \end{bmatrix}, \mathcal{A}_2(:, :, 1, 2) = \begin{bmatrix} 2+3i & 2i+2k \\ 1+2i+k & 3k \\ 1+3k & j+0.5k \end{bmatrix}, \\
\mathcal{A}_2(:, :, 2, 1) &= \begin{bmatrix} j+2k & i+j+5k \\ 3i+k & 2j+0.25k \\ 0.25+j & 1+i+2j \end{bmatrix}, \mathcal{A}_2(:, :, 2, 2) = \begin{bmatrix} 1+2i & 0.5j+5k \\ 2i+3k & i+3j \\ 0 & 2+i+j \end{bmatrix}, \\
\mathcal{A}_2(:, :, 3, 1) &= \begin{bmatrix} 3+i & 0.5 \\ i+j+2k & 2i+3k \\ 0.25+j & 3i+j \end{bmatrix}, \mathcal{A}_2(:, :, 3, 2) = \begin{bmatrix} 0 & 2i+j \\ 2j & 0.5+i+j+k \\ 1+j & i+2k \end{bmatrix}, \\
\mathcal{P}_2(:, :, 1, 1) &= \begin{bmatrix} -1+3.5i-0.75j+9.5k & -0.5+8.25i+2.75j+2.5k \\ 3+3.75i+1.5j-2.75k & 4-0.5i+4.5j-4.4375k \\ 1.0625+1.25i+0.25j-5.75k & -5.25-0.25i+2.5j+1.5k \end{bmatrix}, \\
\mathcal{P}_2(:, :, 1, 2) &= \begin{bmatrix} 2.5+8.5i-12j+9k & 26.25+7.5i+7j+4k \\ 11+1.5i-2.5j+7k & 11.25-3.75i-1.5j+3k \\ 15.125+3.75j+8.25k & 3+2.5i-2.5j+13.5k \end{bmatrix}, \\
\mathcal{P}_2(:, :, 2, 1) &= \begin{bmatrix} 3+7i-3j+10k & 17+3i+6j+3k \\ 5+8i-7j+6k & 10+4i+4.75j+3k \\ 16+0.25i+4.5j+7.5k & -1+2j+6k \end{bmatrix}, \\
\mathcal{P}_2(:, :, 2, 2) &= \begin{bmatrix} 3+2i-j+6.5k & 24+2.5i+5.5j+4k \\ 10+1.5i-8.5j+6k & 8.25-4.75i-1.5j+2.5k \\ 7+i+5j+1.25k & 4+1.25i-4j+9k \end{bmatrix}, \\
\mathcal{P}_2(:, :, 3, 1) &= \begin{bmatrix} 3+3i-2.5j+8k & 12.5-4i+6.5j+4k \\ 4-2.5i-4j+0.5k & 9.5-6i-0.75j+1.125k \\ 2.375-4.75i+6.5j+2k & 1.5+2.5i-7.5j+k \end{bmatrix}, \\
\mathcal{P}_2(:, :, 3, 2) &= \begin{bmatrix} -0.5+19i-5j+8k & 15.5+13.5i+2.75j+6.5k \\ 2+6i-1j+13.5k & 13.75+0.75i+11j+10k \\ 13.25+1.25i+13j+5.25k & 9.5+0.5i+11k \end{bmatrix},
\end{aligned}$$

$$\begin{aligned}
\mathcal{B}_2(1,1,::) &= \begin{bmatrix} 5i+j & 1 \\ 2k & j+4k \end{bmatrix}, \mathcal{B}_2(1,2,::) = \begin{bmatrix} 0 & 0.5j+k \\ i+k & 2i \end{bmatrix}, \\
\mathcal{B}_2(2,1,::) &= \begin{bmatrix} 2i+j & 2k \\ 0 & 2i \end{bmatrix}, \mathcal{B}_2(2,2,::) = \begin{bmatrix} 3i & i+j+2k \\ 0 & i+0.5j+k \end{bmatrix}, \\
\mathcal{B}_2(3,1,::) &= \begin{bmatrix} j & 1+j \\ 2 & 2i+k \end{bmatrix}, \mathcal{B}_2(3,2,::) = \begin{bmatrix} 2k & i+k \\ 0 & 1+2k \end{bmatrix}, \\
\mathcal{Q}_2(:, :, 1, 1) &= \begin{bmatrix} -5+4i+12j+4k & 3+8i+7j-5.5k \\ -6-0.75i+7.75j-k & -1.5+2i+14j-0.5k \\ -3+i+4j-4k & -11+2i+3j-k \end{bmatrix}, \\
\mathcal{Q}_2(:, :, 1, 2) &= \begin{bmatrix} 3+j+6k & 10.5+2i+0.5j+5.5k \\ 8.25-1.5i-3.5j+k & 11+7.5i+4.5k \\ 1-3i+0.25j+2.5k & 8.5+1.5i+3j+4k \end{bmatrix}, \\
\mathcal{Q}_2(:, :, 2, 1) &= \begin{bmatrix} 6-2i-2j & 2+i+2j+k \\ 2+i+j+3.5k & 3+2k \\ 4+0.5i-2j+4.5k & 1+3i-3j+5k \end{bmatrix}, \\
\mathcal{Q}_2(:, :, 2, 2) &= \begin{bmatrix} 10+3i+0.5j+2k & 2.25+5i+6j+0.5k \\ -0.5-5.5i+11.25j+0.5k & 3.5-1.5i+8j+3.5k \\ 7.5+2i-j-k & 5+1.5i-6j-k \end{bmatrix}, \\
\mathcal{E}_1(:, :, 1, 1) &= \begin{bmatrix} i+j+k & 0 & 3 \\ i+0.5j+2k & i & 2i+k \end{bmatrix}, \mathcal{E}_1(:, :, 1, 2) = \begin{bmatrix} 1 & i+j+2k & 0.5i \\ i+j+5k & 3j & j+2k \end{bmatrix}, \\
\mathcal{E}_1(:, :, 2, 1) &= \begin{bmatrix} 3k & j+2k & j \\ 0 & i+j & 2i \end{bmatrix}, \mathcal{E}_1(:, :, 2, 2) = \begin{bmatrix} j+2k & 4i+j+k & k \\ 0 & 3i & 1 \end{bmatrix}, \\
\mathcal{E}_1(:, :, 3, 1) &= \begin{bmatrix} k & 3i+j & 0 \\ i+5j+k & i & 2j \end{bmatrix}, \mathcal{E}_1(:, :, 3, 2) = \begin{bmatrix} i & 2i+3k & k \\ 0 & i & 5j \end{bmatrix}, \\
\mathcal{F}_1(1,1,::) &= \begin{bmatrix} i & i+4j+k \\ 0 & i+3j+k \end{bmatrix}, \mathcal{F}_1(1,2,::) = \begin{bmatrix} k & 5i+j+k \\ 0 & i+j+0.5k \end{bmatrix}, \\
\mathcal{F}_1(2,1,::) &= \begin{bmatrix} i+j & 2+3k \\ 0 & i+k \end{bmatrix}, \mathcal{F}_1(2,2,::) = \begin{bmatrix} 3+k & 1 \\ j & 2j+k \end{bmatrix}, \\
\mathcal{F}_1(3,1,::) &= \begin{bmatrix} 1+i+j & 2k \\ 2i & 2i+j+k \end{bmatrix}, \mathcal{F}_1(3,2,::) = \begin{bmatrix} 2+2k & i \\ 1+3j+k & 0.5+k \end{bmatrix}, \\
\mathcal{E}_2(:, :, 1, 1) &= \begin{bmatrix} i+2k & i+0.5j+2k & 2k \\ 0 & i+0.25k & i+2j \end{bmatrix}, \mathcal{E}_2(:, :, 1, 2) = \begin{bmatrix} 1+j+k & 2i & 3i+j \\ 3+i & 3j+k & i+k \end{bmatrix}, \\
\mathcal{E}_2(:, :, 2, 1) &= \begin{bmatrix} 2i+3k & 0 & 2j \\ 2 & i+j & i+k \end{bmatrix}, \mathcal{E}_2(:, :, 2, 2) = \begin{bmatrix} 2+i+2k & 0 & k \\ i+j+k & 3j & 1+k \end{bmatrix}, \\
\mathcal{E}_2(:, :, 3, 1) &= \begin{bmatrix} i+0.25j & 3k & i+2j+0.5k \\ 0 & i+k & 2j+3k \end{bmatrix}, \mathcal{E}_2(:, :, 3, 2) = \begin{bmatrix} i+0.5j & 0 & 3+k \\ 2i+k & i+2k & 3j+0.5k \end{bmatrix}, \\
\mathcal{F}_2(1,1,::) &= \begin{bmatrix} 3+i+j & 1+i+2k \\ 3 & i+3j+k \end{bmatrix}, \mathcal{F}_2(1,2,::) = \begin{bmatrix} 2j+k & 2+j+k \\ 0 & i+j+0.5k \end{bmatrix}, \\
\mathcal{F}_2(2,1,::) &= \begin{bmatrix} 2+j+k & 1+i \\ j+k & 1+k \end{bmatrix}, \mathcal{F}_2(2,2,::) = \begin{bmatrix} 2i+j+k & 1+0.5i \\ 2j+k & i+k \end{bmatrix},
\end{aligned}$$



$$\begin{aligned}
\mathcal{F}_2(3,1, :, :) &= \begin{bmatrix} i+k & j \\ 1+0.5i+k & 1+j+2k \end{bmatrix}, \mathcal{F}_2(3,2, :, :) = \begin{bmatrix} 1+j & 1+k \\ 3i+j+2k & 2j+k \end{bmatrix}, \\
\mathcal{H}(:, :, 1, 1)^T &= \begin{bmatrix} 172 + 157.63i + 101.75j + 218k & 147.75 + 120.5i + 144j + 145.75k \\ 71.25 + 156i + 58.5j + 174k & 34 - 48.5i + 99.5j - 20k \\ 58.75 + 63.25i + 101.75j + 68.625k & 196.5 + 42.5i + 99.25j + 0.5k \end{bmatrix}, \\
\mathcal{H}(:, :, 1, 2)^T &= \begin{bmatrix} 106.88 + 136.75i + 89.5j + 101.5k & 166.75 + 54i + 51j + 138.5k \\ 39.75 + 144.25i - 21j + 209.25k & 40 - 66.75i + 44.875j + 57.5k \\ 72 + 72.125i + 3.125j + 42.875k & 115.75 - 26.75i + 163.75j - 11.75k \end{bmatrix}, \\
\mathcal{H}(:, :, 2, 1)^T &= \begin{bmatrix} 49 + 158.25i + 79.75j + 133.38k & 41 + 79.5i + 66.5j + 44.25k \\ 35.75 + 51.5i + 50j + 59.75k & 21.5 + 18.375i + 42.625j - 33.25k \\ 66 + 44i + 95.75j + 38k & 87 - 2.75i + 104.25j - 13.5k \end{bmatrix}, \\
\mathcal{H}(:, :, 2, 2)^T &= \begin{bmatrix} 110.88 + 109.69i + 82.125j + 141.88k & 143.63 + 98.125i + 60.125j + 78.875k \\ 18.875 + 129.25i + 48.5j + 153.13k & 29.125 - 2.75i + 71.5j - 24.5k \\ 31.625 + 37.625i + 73.625j + 89.438k & 145.63 - 20.5i + 115.5j - 41.5k \end{bmatrix}.
\end{aligned}$$

By calculation of MATLAB, the output is: Exist a unique solution and the unique numerical solution is

$$\begin{aligned}
\mathcal{X}_1(:, :, 1, 1) &= \begin{bmatrix} 1 + 1i - 9.2704e^{-15}j + 1k & 2 + 1.088e^{-14}i + 8.88e^{-16}j - 3.33e^{-15}k \\ 0.25 - 6.1062e^{-15}i + 1.5848e^{-14}j - 3.6082e^{-15}k & 0.5 - 3.1086e^{-15}i - 1.1324e^{-14}j + 2k \\ 2 + 1i - 3.7192e^{-15}j + 1k & 4.885e^{-15} + 1i + 5.4609e^{-15}j + 3k \end{bmatrix}, \\
\mathcal{X}_1(:, :, 1, 2) &= \begin{bmatrix} 1.4821e^{-14} + 1.1324e^{-14}i + 2j + 0.5k & -1.9651e^{-14} + 0.25i + 1.1879e^{-14}j + 2k \\ 1 + 2i + 1j + 1.36e^{-15}k & 0.5 - 6.9389e^{-16}i + 6.4393e^{-15}j + 2k \\ 1 - 5.107e^{-15}i + 8.9928e^{-15}j + 5.5511e^{-15}k & 0.5 + 1.1435e^{-14}i + 1j - 3.2196e^{-15}k \end{bmatrix}, \\
\mathcal{X}_1(:, :, 2, 1) &= \begin{bmatrix} -1.138e^{-14} + 3i + 2j - 8.7846e^{-15}k & -1.3496e^{-14} + 0.5i + 1j + 3k \\ 5.3187e^{-15} + 1i + 9.6867e^{-15}j + 1k & 6.5781e^{-15} + 7.7716e^{-16}i + 2j + 1.2434e^{-14}k \\ 3 + 1.2879e^{-14}i + 2j + 9.2079e^{-15}k & 1.0042e^{-14} + 1i + 2j + 2k \end{bmatrix}, \\
\mathcal{X}_1(:, :, 2, 2) &= \begin{bmatrix} 1 + 6.9944e^{-15}i - 2.387e^{-15}j + 0.5k & 2 + 8.2157e^{-15}i + 1j + 5k \\ 6.2728e^{-15} + 2i + 5.4401e^{-15}j + 3k & -1.2434e^{-14} + 1i + 2j + 3.6152e^{-15}k \\ 3 - 1.4849e^{-14}i + 5.4956e^{-15}j - 5.3985e^{-15}k & -5.4956e^{-15} + 1i + 1j + 0.5k \end{bmatrix}, \\
\mathcal{X}_1(:, :, 3, 1) &= \begin{bmatrix} 7.0499e^{-15} + 2i - 5.3291e^{-15}j + 3.1752e^{-14}k & 8.6042e^{-15} - 7.1054e^{-15}i + 1j + 1k \\ 1 + 2i - 1.5418e^{-14}j - 9.4369e^{-16}k & -1.571e^{-14} + 1i + 3.8858e^{-15}j + 7.77e^{-15}k \\ 2 - 2.7756e^{-15}i + 1j - 3.7748e^{-15}k & 2 + 1i - 4.774e^{-15}j + 2k \end{bmatrix}, \\
\mathcal{X}_1(:, :, 3, 2) &= \begin{bmatrix} 1 - 1.1102e^{-15}i + 3j - 6.5226e^{-15}k & 5.5511e^{-15} + 2i - 3.7748e^{-15}j + 1k \\ 2 + 7.9242e^{-15}i + 1j + 3k & -1.0103e^{-14} - 1.1102e^{-16}i + 1j + 1.2768e^{-14}k \\ 1 + 1i - 1.1324e^{-14}j - 9.4369e^{-16}k & 5.7732e^{-15} + 1i + 1j + 1k \end{bmatrix},
\end{aligned}$$

$$\begin{aligned}
& \mathcal{X}_2(:, :, 1, 1) \\
&= \begin{bmatrix} 1.07e^{-14} + 1i - 1.55e^{-15}j + 2k & -8.33e^{-15} - 2.78e^{-15}i + 1j + 2.36e^{-15}k \\ 0.25 - 2.22e^{-16}i - 5.7732e^{-15}j - 8.9581e^{-15}k & 7.3275e^{-15} + 0.5i + 1j + 1k \\ -1.44e^{-15} + 1i + 4.16e^{-15}j + 1k & 5.41e^{-15} + 3i - 2.83e^{-15}j + 5.61e^{-15}k \end{bmatrix}, \\
& \mathcal{X}_2(:, :, 1, 2) \\
&= \begin{bmatrix} 5.7732e^{-15} - 1.0214e^{-14}i + 2j + 5.7732e^{-15}k & 1 - 2.6645e^{-15}i - 3.3307e^{-15}j + 2k \\ 1.4433e^{-14} - 2.5535e^{-15}i + 1j + 1k & -3.9968e^{-15} + 2i + 1.249e^{-14}j + 3k \\ 0.5 + 1.4058e^{-14}i - 2.0539e^{-15}j - 8.4377e^{-15}k & 2 - 1.5543e^{-15}i + 1j + 1k \end{bmatrix}, \\
& \mathcal{X}_2(:, :, 2, 1) \\
&= \begin{bmatrix} 1 + 1i + 6.4393e^{-15}j - 4.4409e^{-15}k & 2 + 2.9976e^{-15}i + 2.7478e^{-15}j + 4k \\ 2.4425e^{-15} + 1i - 1.0769e^{-14}j + 2k & -4.996e^{-16} + 2i + 9.8047e^{-15}j - 4.996e^{-15}k \\ 9.992e^{-16} + 2.3315e^{-15}i + 2j + 1.0214e^{-14}k & 5.5511e^{-15} + 1i + 1j - 5.6066e^{-15}k \end{bmatrix}, \\
& \mathcal{X}_2(:, :, 2, 2) \\
&= \begin{bmatrix} 1 + 6.8834e^{-15}i - 1.8319e^{-15}j + 3.5527e^{-15}k & 1.3545e^{-14} + 5.218e^{-15}i + 0.5j + 1k \\ -3.3307e^{-15} + 2i + 1j + 1k & -4.55e^{-15} + 1.11e^{-16}i - 1.06e^{-14}j + 3k \\ -1.6653e^{-15} - 8.9928e^{-15}i + 1j + 1.1047e^{-14}k & -4.3299e^{-15} + 1.3212e^{-14}i + 2j + 1k \end{bmatrix}, \\
& \mathcal{X}_2(:, :, 3, 1) \\
&= \begin{bmatrix} 1 - 3.9135e^{-15}i + 1.8874e^{-15}j + 1k & -3.8511e^{-15} + 1i - 3.858e^{-15}j - 7.4662e^{-15}k \\ 0.5 + 1i + 1.1435e^{-14}j + 2k & 5.1625e^{-15} + 1i + 5.2909e^{-15}j + 1k \\ 1 + 7.2164e^{-16}i + 1.4988e^{-15}j + 2k & 3.1086e^{-15} + 1i + 1j + 2k \end{bmatrix}, \\
& \mathcal{X}_2(:, :, 3, 2) = \begin{bmatrix} 1 + 8.3822e^{-15}i + 2j + 1k & 2.9976e^{-15} + 3i + 9.992e^{-16}j + 1k \\ 1.5543e^{-15} - 1.1768e^{-14}i + 3j + 1k & 0.5 + 6.9944e^{-15}i - 1.131e^{-15}j + 2k \\ 1 + 1i + 1j - 2.2204e^{-16}k & 1 + 3.1086e^{-15}i - 2.1094e^{-15}j + 2k \end{bmatrix},
\end{aligned}$$

The true solution is

$$\begin{aligned}
& \mathcal{X}_1(:, :, 1, 1) = \begin{bmatrix} 1 + i + k & 2 \\ 0.25 & 0.5 + 2k \\ 2 + i + k & i + 3k \end{bmatrix}, \mathcal{X}_1(:, :, 1, 2) = \begin{bmatrix} 2j + 0.5k & 0.25i + 2k \\ 1 + 2i + j & 0.5 + 2k \\ 1 & 0.5 + j \end{bmatrix}, \\
& \mathcal{X}_1(:, :, 2, 1) = \begin{bmatrix} 3i + 2j & 0.5i + j + 3k \\ i + k & 2j \\ 3 + 2j & i + 2j + 2k \end{bmatrix}, \mathcal{X}_1(:, :, 2, 2) = \begin{bmatrix} 1 + 0.5k & 2 + j + 5k \\ 2i + 3k & i + 2j \\ 3 & i + j + 0.5k \end{bmatrix}, \\
& \mathcal{X}_1(:, :, 3, 1) = \begin{bmatrix} 2i & j + k \\ 1 + 2i & i \\ 2 + j & 2 + i + 2k \end{bmatrix}, \mathcal{X}_1(:, :, 3, 2) = \begin{bmatrix} 1 + 3j & 2i + k \\ 2 + j + 3k & j \\ 1 + i & i + j + k \end{bmatrix}, \\
& \mathcal{X}_2(:, :, 1, 1) = \begin{bmatrix} i + 2k & j \\ 0.25 & 0.5i + j + k \\ i + k & 3i \end{bmatrix}, \mathcal{X}_2(:, :, 1, 2) = \begin{bmatrix} 2j & 1 + 2k \\ j + k & 2i + 3k \\ 0.5 & 2 + j + k \end{bmatrix}, \\
& \mathcal{X}_2(:, :, 2, 1) = \begin{bmatrix} 1 + i & 2 + 4k \\ i + 2k & 2i \\ 2j & i + j \end{bmatrix}, \mathcal{X}_2(:, :, 2, 2) = \begin{bmatrix} 1 & 0.5j + k \\ 2i + j + k & 3k \\ j & 2j + k \end{bmatrix}, \\
& \mathcal{X}_2(:, :, 3, 1) = \begin{bmatrix} 1 + k & i \\ 0.5 + i + 2k & i + k \\ 1 + 2k & i + j + 2k \end{bmatrix}, \mathcal{X}_2(:, :, 3, 2) = \begin{bmatrix} 1 + 2j + k & 3i + k \\ 3j + k & 0.5 + 2k \\ 1 + i + j & 1 + 2k \end{bmatrix}.
\end{aligned}$$

It is worth noting that the error between the numerical solution and the true solution is  $4.5778 \times 10^{-13}$ , which undoubtedly demonstrates the accuracy of our results.

## 5. Color Videos Encryption and Decryption Based on (1)

Similar to quaternions, it has been demonstrated that split quaternions can represent color images. Split quaternion tensors can represent color videos, when the video is segmented into multiple image slices. In this section, we propose a method for  $n$  videos encryption and decryption by using Theorem 2. In the following, we provide two encryption algorithms, two decryption algorithms and two examples to demonstrate our result.

---

### Algorithm 3 Encryption Process of Videos

---

- (1) **Input:**  $n$  original videos and system coefficients  $\mathcal{A}_1, \mathcal{A}_2, \dots, \mathcal{A}_n, \mathcal{B}_1, \mathcal{B}_2, \dots, \mathcal{B}_n, \mathcal{E}_1, \mathcal{E}_2, \dots, \mathcal{E}_n$  and  $\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_n$ .
  - (2) **Parse the videos:** The  $\mathcal{X}_i$  represents  $i$ -th video where  $i = 1, \dots, n$  and  $\mathcal{X}_i(:, :, j)$  represents  $j$ -th frame of  $i$ -th video.
  - (3) Calculate the tensors  $\mathcal{P}_1, \mathcal{P}_2, \dots, \mathcal{P}_n, \mathcal{Q}_1, \mathcal{Q}_2, \dots, \mathcal{Q}_n$  and  $\mathcal{H}$  by system (1).
  - (4) Encrypt  $i$ -th video  $\mathcal{X}_i$  by  $\mathcal{P}_i - \mathcal{Q}_i$  where  $i = 1, \dots, n$ .  $\mathcal{P}_i, \mathcal{Q}_i$  and  $\mathcal{H}$  is the key of  $i$ -th video.
  - (5) **Output:** Encrypted video.
- 

---

### Algorithm 4 Decryption Process of Videos

---

- (1) **Input:** Encrypted videos  $\mathcal{X}_i$ , keys  $\mathcal{P}_i, \mathcal{Q}_i, \mathcal{H}$  and system coefficients  $\mathcal{A}_i, \mathcal{B}_i, \mathcal{E}_i, \mathcal{F}_i$  where  $i = 1, \dots, n$ .
  - (2) Calculate the numerical tensors  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n]$  by Algorithm 1.
  - (3) **Recovered the videos:** The  $\mathcal{X}_i$  represents  $i$ -th recovered video where  $i = 1, \dots, n$  and  $\mathcal{X}_i(:, :, j)$  represents  $j$ -th recovered frame of  $i$ -th video.
  - (4) Calculate the error norm for  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n]$  and  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n]$ .
  - (5) **Output:** Decrypted video.
- 

---

### Algorithm 5 Block Encryption Process of Frame

---

- (1) **Input:** one original frame and system coefficients  $\mathcal{A}_1, \mathcal{A}_2, \dots, \mathcal{A}_n, \mathcal{B}_1, \mathcal{B}_2, \dots, \mathcal{B}_n, \mathcal{E}_1, \mathcal{E}_2, \dots, \mathcal{E}_n$  and  $\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_n \in \mathbb{Q}_S^{3 \times 3}$ .
  - (2) **Parse the frames:** The  $X \in \mathbb{Q}_S^{480 \times 480}$  represents the frame and  $X_i \in \mathbb{Q}_S^{3 \times 3}$  represents  $i$ -th sub-frame of the frame where  $i = 1, \dots, n$ .
  - (3) Calculate the matrices  $\mathcal{P}_1, \mathcal{P}_2, \dots, \mathcal{P}_n, \mathcal{Q}_1, \mathcal{Q}_2, \dots, \mathcal{Q}_n \in \mathbb{Q}_S^{3 \times 3}$  and  $H \in \mathbb{Q}_S^{3 \times 3}$  by system (1).
  - (4) Assemble the  $i$ -th sub-frame  $X_i, \mathcal{P}_i, \mathcal{Q}_i$  where  $i = 1, \dots, n$  back into the frame  $X, P, Q \in \mathbb{Q}_S^{480 \times 480}$ . Encrypt the frame  $X$  by  $(P - Q) \in \mathbb{Q}_S^{480 \times 480}$ .
  - (5) **Output:** Encrypted frame.
- 

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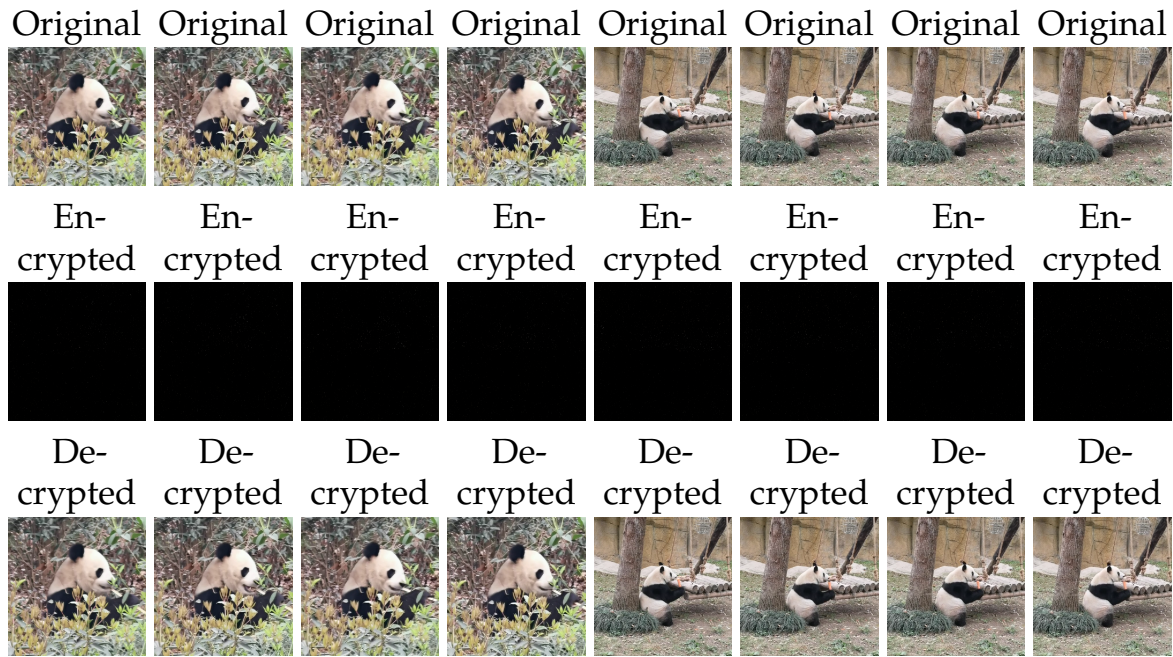
### Algorithm 6 Block Decryption Process of Frame

---

- (1) **Input:** Encrypted frame  $X \in \mathbb{Q}_S^{480 \times 480}$ , keys  $P, Q, H \in \mathbb{Q}_S^{480 \times 480}$  and system coefficients  $\mathcal{A}_i, \mathcal{B}_i, \mathcal{E}_i, \mathcal{F}_i \in \mathbb{Q}_S^{3 \times 3}$  where  $i = 1, \dots, n$ .
  - (2) Calculate the matrices  $[\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n]$  by Algorithm 1.
  - (3) **Recovered the frame:** The  $\mathcal{X}_i \in \mathbb{Q}_S^{3 \times 3}$  represents  $i$ -th recovered sub-frame where  $i = 1, \dots, n$  and assemble the  $\mathcal{X}_i$  back into the frame  $\hat{X} \in \mathbb{Q}_S^{480 \times 480}$ .
  - (4) Calculate the error norm between  $X$  and  $\hat{X}$ .
  - (5) **Output:** Decrypted frame.
- 

**Remark 1.** As the dimensions of the decrypted video increase, the time and space complexity of algorithms 3 and 4 increase exponentially. Therefore, we propose a block encryption and decryption algorithm in algorithms 5 and 6. By reusing algorithms 5 and 6, the memory usage can be reduced and the operational speed can be significantly enhanced. It is worth mentioning that algorithms 5 and 6 are also suitable for local encryption and decryption tasks.

**Example 2.** Let’s input two videos and execute the above algorithms. The following is the result by MATLAB:



In the result, the average error between the numerical solution and the true solution is  $2.7106 \times 10^{-9}$ . The error of each frame is shown in the Table 1 as follows.

**Table 1.** Error of Numerical Solution and True Solution.

Video 1	Error	Video 2	Error
1-frame	$2.6377 \times 10^{-9}$	1-frame	$2.7776 \times 10^{-9}$
10-frame	$2.6527 \times 10^{-9}$	10-frame	$2.7644 \times 10^{-9}$
16-frame	$2.6655 \times 10^{-9}$	16-frame	$2.7569 \times 10^{-9}$
25-frame	$2.6595 \times 10^{-9}$	25-frame	$2.7707 \times 10^{-9}$

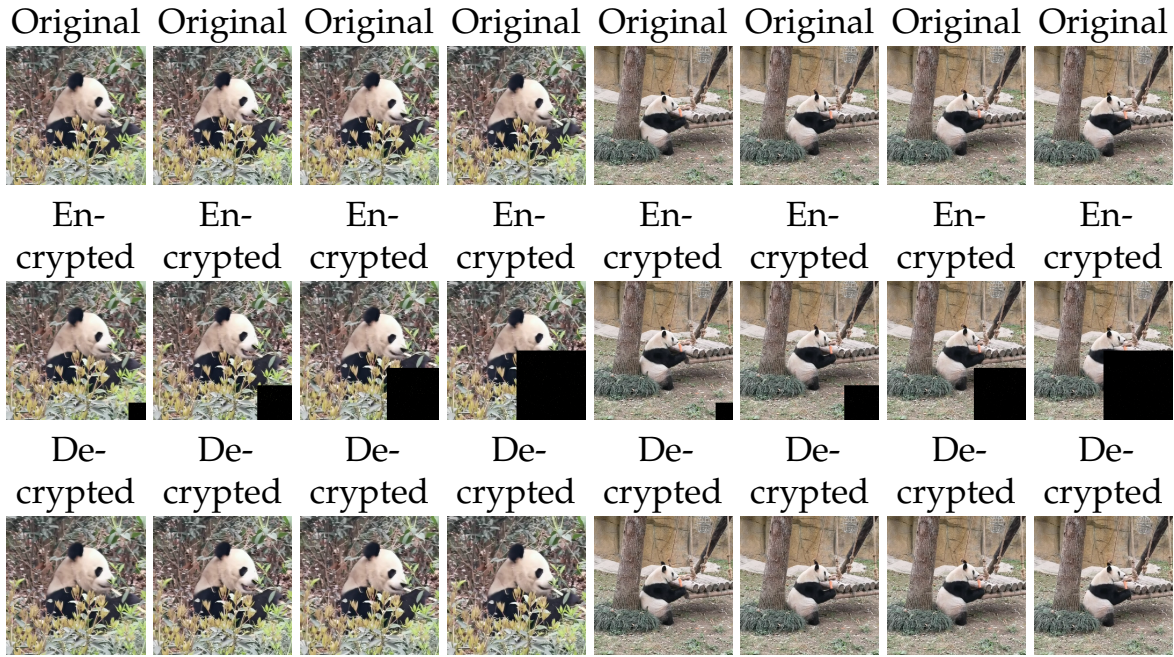
To further evaluate the quality of decrypted video, we use the Peak Signal-to-Noise Ratio (PSNR), Structural Similarity Index (SSIM) and Feature Similarity Index (FSIM) [33]. The results are summarized in Table 2. All PSNR values exceed 50, while both SSIM and FSIM values are 1, demonstrating the outstanding quality of our decrypted videos.

**Table 2.** PSNR, SSIM and FSIM.

Video 1	PSNR	SSIM	FSIM	Video 2	PSNR	SSIM	FSIM
1-frame	245.4833	1	1	1-frame	244.6598	1	1
10-frame	246.1960	1	1	10-frame	245.8679	1	1
16-frame	246.0469	1	1	16-frame	245.6220	1	1
25-frame	246.0457	1	1	25-frame	245.4920	1	1
(a) PSNR, SSIM and FSIM of the video 1				(b) PSNR, SSIM and FSIM of the video 2			

**Example 3.** Let’s input two videos and finish local encryption and decryption using algorithms 5 and 6 by MATLAB:





In the result, the average error between the numerical solution and the true solution is  $8.9857 \times 10^{-10}$ . The error of each frame is listed in the Table 3 as follows.

Table 3. Error of Numerical Solution and True Solution.

Video 1	Error	Video 2	Error
3-frame	$4.0604 \times 10^{-10}$	3-frame	$4.1622 \times 10^{-10}$
12-frame	$8.4663 \times 10^{-10}$	12-frame	$7.7403 \times 10^{-10}$
18-frame	$1.1171 \times 10^{-9}$	18-frame	$9.9505 \times 10^{-10}$
27-frame	$1.4076 \times 10^{-9}$	27-frame	$1.2259 \times 10^{-9}$

The results of PSNR, SSIM and FSIM are shown in Table 4. PSNR values all exceed 50, while both SSIM and FSIM values reach 1, indicating that the quality of local decryption is excellent.

Table 4. PSNR, SSIM and FSIM.

Video 1	PSNR	SSIM	FSIM	Video 2	PSNR	SSIM	FSIM
3-frame	245.0190	1	1	3-frame	244.3713	1	1
12-frame	244.5530	1	1	12-frame	244.8359	1	1
18-frame	244.6391	1	1	18-frame	246.7069	1	1
27-frame	245.3403	1	1	27-frame	246.3910	1	1
(a) PSNR, SSIM and FSIM of the video 1				(b) PSNR, SSIM and FSIM of the video 2			

6. Conclusion

The system of quaternion tensor equations (1) is universal, encompassing many other systems. This kind of system of quaternion tensor equations has practical applications in many areas. In this paper, we first transformed the quaternion tensor equations into quaternion matrix equations through the function  $f$ . By using the complex representation of split quaternion matrices and Moore–Penrose inverse, we presented a specific necessary and sufficient condition for the existence of a general solution to the system (1). When the solvability condition holds, we also derived the expression of the general solution to system (1). In particular, the unique solution was also given when the uniqueness condition held. Then we converted the obtained quaternion matrix solution back into a quaternion tensor solution using the function  $f^{-1}$ . Additionally, we established two algorithms to solve the system (1) and one example was given to prove the correctness of our results. In the last part, we provided



the algorithms and two examples for color videos encryption and decryption using the system (1). In the future, we will focus on other systems of quaternion tensor equations over the split quaternion algebra.

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