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Solutions to matrix equations $X - AXB = CY + R$ and $X - A\hat{X}B = CY + R$ *

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Abstract

The present work proposed an alternative approach to find the closed-form solutions of the non-homogeneous Yakubovich matrix equation $X - AXB = CY + R$. Based on the derived closed-form solution to the nonhomogeneous Yakubovich matrix equation, the solutions to the nonhomogeneous Yakubovich quaternion j-conjugate matrix equation $X - A\hat{X}B = CY + R$ are obtained by the use of the real representation of a quaternion matrix. The existing complex representation method requires the coefficient matrix A to be a block diagonal matrix over complex field. In contrast in this publication we allow a quaternion matrix of any dimension. As an application, eigenstructure assignment problem for descriptor linear systems is considered.

Keywords: Closed-form solution; quaternion matrix equation; real representation

1. Introduction

The generalized Sylvester matrix equation

$$AX - EXF = CY, \tag{1.1}$$

is closely related with many problems in control theory, such as pole/eigenstructure assignment

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design [1-2], Luenberger-type observer design [3-5], robust fault detection [6-11] and so on. Finding the complete parametric solutions of the matrix equation (1.1), i.e. parametric solutions consisting of the maximum number of free parameters, allows for a robust control system design for a larger class of design problems. The generalized Sylvester matrix equation (1.1) is equivalent to the generalized Sylvester matrix equation

$$X - AXB = CY, \quad (1.2)$$

with their coefficient matrices satisfying some relations. Due to the above mentioned applications, the generalized Sylvester matrix equation (1.1) and (1.2) has been studied by many authors. Assuming F in matrix equation (1.1) is in Jordan form, Duan [12, 13] has obtained a parametric solution of the generalized Sylvester matrix equation through the use of the right coprime factorization of the input-state transfer function $(sE - A)^{-1}B$. However, this solution is not in a direct, explicit form but in a recursive form. In [14-16], an explicit, analytical and complete solution to the matrix equation (1.2) is obtained. In [14], the explicit solution can be obtained by the use of the Kronecker map [52]. In [15], the proposed solution is in a very compact form and can be immediately obtained with a series of matrices $D_i, i = 0, 1, \dots, \omega$. Obviously, the following nonhomogeneous Yakubovich matrix equation

$$X - AXB = CY + R, \quad (1.3)$$

where $A \in R^{n \times n}$, $B \in R^{p \times p}$ and $C \in R^{n \times r}$ are the given real matrices, $X \in R^{n \times p}$ and $Y \in R^{r \times p}$ are the matrices to be determined, is the generalized form of the matrix equation (1.2). This nonhomogeneous Yakubovich matrix equation plays an important role in output regulator design for time-invariant linear systems and eigenstructure assignment for matrix second-order linear systems [17-20]. However, there is little in the literature considering such type of matrix equation [21].

For two complex square matrices A and B , there exists a nonsingular complex matrix P such that $A = P^{-1}B\bar{P}$ [22]. Then A and B is called consimilarity. This is another type of similarity in the field of linear algebra. The consimilarity theory of complex matrices plays an important role in the research of modern quantum theory [23]. For this reason, the matrix equation $X - AXB = CY$ in complex field has been extended to complex conjugate matrix equation $X - A\bar{X}B = CY$ by an application of the consimilarity concept [18]. Moreover, the consimilarity of quaternion matrix is also defined in [24]. Similarly, the quaternion matrix

equation $X - AXB = CY$ has also been extended to the quaternion j-conjugate matrix equation $X - A\hat{X}B = CY$ by an application of the consimilarity concept over quaternion field [25]. Due to its compact notation, moderate computational requirements and avoidance of singularities associated to 3×3 rotation matrices [26], these quaternion matrix equations have been extensively used in computer graphics, vector-sensor processing, and aerospace problems [27-30].

Let us point out that the least squares solution of the quaternion j-conjugate matrix equation $X - A\hat{X}B = C$ (where \hat{X} denotes the j-conjugate of quaternion matrix X) with the least squares norm [31] has been obtained by the use of the complex representation of quaternion matrix, the Moore-Penrose generalized inverse and the Kronecker product. Some necessary and sufficient conditions for the existence of the solution to the quaternion matrix equations $AXB + CYD = E$, $(AX, XC) = (B, D)$, $AXB = C$, have been proposed and the solutions have been derived in the references [32-34]. The explicit solutions to the quaternion j-conjugate matrix equation $X - A\hat{X}B = C$ and $X - A\hat{X}B = CY$ [25, 35, 36], where the coefficient quaternion matrix A is a block diagonal form over complex field, have been studied. The general solution, parameterize general solution and explicit solution to some quaternion matrix equations are obtained in [37-41]. Some necessary and sufficient conditions, the maximal and minimal ranks of the solutions are presented in these paper.

In this paper, we are concerned with the explicit solutions to two kinds of nonhomogeneous Yakubovich matrix equations, which are considered in real field and in quaternion field. The class of matrix equations which are studied in the current paper include many linear matrix equations as special cases, such as discrete Sylvester matrix equation, generalized Sylvester matrix equation, discrete Lyapunov matrix equation, and so on [18-22, 31, 35-36]. The motivation for this paper comes from the references [15-18, 20, 21, 25, 35, 36, 39-46, 48-55]. This study mainly includes two parts. Firstly, we present an alternative approach to obtain the explicit solution of the nonhomogeneous Yakubovich matrix equation in real field. It is also shown that the explicit solution to the considered matrix equation is expressed as the coefficient matrices A, B, C , the free parameter matrix Z and the coefficient matrix of polynomial matrix $adj(I - sA)$. Secondly, based on the derived solutions to the nonhomogeneous Yakubovich matrix equation in real field, the explicit solutions to the j-conjugate matrix equation in quaternion field are discussed by means of real representation of a quaternion matrix.

The rest of this paper is organized as follows. In Section 2, the real matrix equation is studied. First of all, a method is presented for obtaining the solution to the nonhomogeneous Yakubovich matrix equation. Section 3 is devoted to introducing the real representation of a quaternion matrix and studying the solution to the quaternion j -conjugate matrix equation. In Section 4, some numerical examples and application examples are provided to show the effectiveness of the obtained results. Finally, the paper is ended with a brief conclusion in Section 5.

Throughout this paper, we use the following notations. Let R, C and $Q = R \oplus Ri \oplus Rj \oplus Rk$ denote the real number field, the complex number field and the quaternion field, respectively, where $i^2 = j^2 = k^2 = -1, ij = -ji = k$. $R^{m \times n} (C^{m \times n} \text{ or } Q^{m \times n})$ denotes the set of all $m \times n$ matrices on R (C or Q). For any matrix $A \in C^{m \times n}$, $A^T, \bar{A}, A^H, \det A$ and $\text{adj}(A)$ represent the transpose, conjugate, conjugate transpose, determinant and adjoint of A , respectively. In addition, symbol A_σ is the real representation of a quaternion matrix A . $A \otimes B = (a_{ij}B)$ denotes the Kronecker product of two matrices A and B . If $A \in Q^{m \times n}$, let $A = A_1 + A_2i + A_3j + A_4k$, where $A_t \in R^{m \times n}, t = 1, \dots, 4$, and define $\hat{A} = A_1 - A_2i + A_3j - A_4k$ to be the j -conjugate of the quaternion matrix A . For $A \in C^{m \times n}$, $\text{vec}(A)$ is defined as $\text{vec}(A) = [a_1^T \ a_2^T \ \dots \ a_n^T]^T$, where $a_i (i = 1, 2, \dots, n)$ are some complex numbers. We denote the $n \times n$ identity matrix by I_n . We also write it as I . Furthermore, let $A \in R^{n \times n}, B \in R^{n \times r}$, and $C \in R^{m \times n}$, we have the following notations associated with these matrices:

$$\begin{aligned} \text{Ctr}_n(A, B) &= \begin{bmatrix} B & AB & \dots & A^{n-1}B \end{bmatrix}, \text{Obs}_k(A, C) = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{k-1} \end{bmatrix}, \\ f_{(I,A)}(s) &= \det(I - sA) = \alpha_n s^n + \alpha_{n-1} s^{n-1} + \dots + \alpha_1 s + 1, \\ S_r(I, A) &= \begin{bmatrix} I_r & \alpha_1 I_r & \alpha_2 I_r & \dots & \alpha_{n-1} I_r \\ & I_r & \alpha_1 I_r & \dots & \alpha_{n-2} I_r \\ & & & \dots & \dots \\ & & & I_r & \alpha_1 I_r \\ & & & & I_r \end{bmatrix}. \end{aligned}$$

Thus, $\text{Ctr}_n(A, B)$ is the controllability matrix of the matrix pair (A, B) , $\text{Obs}_k(A, C)$ is the observability matrix of the matrix pair (A, C) , $S_r(I, A)$ is a symmetric operator matrix.

2. Real matrix equation $X - AXB = CY + R$

In this section, we will propose an alternative approach to solve the nonhomogeneous Yakubovich matrix equation (1.3). Before proceeding, we need the following lemma on the matrix equation

$$X_0 - AX_0B = R, \quad (2.1)$$

where $A \in R^{n \times n}$, $B \in R^{p \times p}$ and $R \in R^{n \times p}$ are known matrices, and X_0 is the matrix to be determined.

Lemma 1 [47]. Given matrices $A \in R^{n \times n}$, $B \in R^{p \times p}$ and $R \in R^{n \times p}$, suppose that $\{s | \det(I - sA) = 0\} \cap \lambda(B) = \emptyset$, and let

$$f_{(I,A)}(s) = \det(I - sA) = \alpha_n s^n + \cdots + \alpha_1 s + \alpha_0, \alpha_0 = 1,$$

and

$$\text{adj}(I - sA) = \sum_{i=0}^{n-1} R_i s^i. \quad (2.2)$$

Then the unique solution to the matrix equation (2.1) can be characterized by

$$X_0 = \left(\sum_{i=0}^{n-1} R_i R B^i \right) [f_{(I,A)}(B)]^{-1}.$$

Theorem 1. Given matrices $A \in R^{n \times n}$, $B \in R^{p \times p}$, $C \in R^{n \times r}$ and $R \in R^{n \times p}$, suppose that $\{s | \det(I - sA) = 0\} \cap \lambda(B) = \emptyset$, and (2.2) holds. Then all solutions to the nonhomogeneous Yakubovich matrix equation (1.3) can be expressed as

$$\begin{cases} X = \sum_{i=0}^{n-1} R_i C Z B^i + \left(\sum_{i=0}^{n-1} R_i R B^i \right) [f_{(I,A)}(B)]^{-1}, \\ Y = Z f_{(I,A)}(B), \end{cases} \quad (2.3)$$

where $Z \in R^{r \times p}$ is an arbitrary chosen free parameter matrix.

Proof. Regarding the solution of $\alpha_i, i \in I[0, n]$, and $R_i, i \in I[0, n-1]$ in Lemma 1, the so-called generalized Leverrier algorithm [45] can be stated as the following relation:

$$\begin{cases} R_i = AR_{i-1} + \alpha_i I, & R_0 = I, i \in I[1, n]. \\ \alpha_i = \frac{\text{trace}(AR_{i-1})}{i}, & i \in I[1, n]. \end{cases}$$

Suppose that a solution to the nonhomogeneous Yakubovich matrix equation (1.3) can be expressed as $X = T + \tilde{X}, Y = \tilde{Y}$ with (\tilde{X}, \tilde{Y}) being a solution to matrix equation (1.2). Then one has

$$(T + \tilde{X}) - A(T + \tilde{X})B - C\tilde{Y} - R = \tilde{X} - A\tilde{X}B + T - ATB - C\tilde{Y} - R = T - ATB - R.$$

Under this condition, when $X_0 = T$ is a solution to (2.1), $X = T + \tilde{X}, Y = \tilde{Y}$ is a solution to (1.3). Combining this fact and Theorem 2 in [14], now we can obtain the solution to the nonhomogeneous Yakubovich matrix equation (1.3).

Corollary 1. Given matrices $A \in R^{n \times n}$, $B \in R^{p \times p}$, $C \in R^{n \times r}$ and $R \in R^{n \times p}$, suppose that $\{s | \det(I - sA) = 0\} \cap \lambda(B) = \emptyset$, and (2.2) holds, then the relation

$$\begin{cases} X = \left(\sum_{i=0}^{n-1} R_i R B^i \right) [f_{(I,A)}(B)]^{-1}, \\ Y = 0, \end{cases} \quad (2.4)$$

is also a special solution to the matrix equation (1.3), where $Z \in R^{r \times p}$ is an arbitrary chosen free parameter matrix.

Proof. If we let $Z = 0$ in relation (2.3), we can obtain the relation (2.4). This complete the proof of this theorem.

Next we have the following equivalent forms of the solution in Theorem 1.

Theorem 2. Given matrices $A \in R^{n \times n}$, $B \in R^{p \times p}$, $C \in R^{n \times r}$ and $R \in R^{n \times p}$, suppose that $\{s | \det(I - sA) = 0\} \cap \lambda(B) = \emptyset$, and (2.2) holds. Then all solutions to the nonhomogeneous Yakubovich matrix equation (1.3) can be expressed as

$$\begin{cases} X f_{(I,A)}(B) = \sum_{k=0}^{n-1} \sum_{i=0}^{n-1} \alpha_k A^{i-k} C Z B^i f_{(I,A)}(B) + \sum_{k=0}^{n-1} \sum_{i=0}^{n-1} \alpha_k A^{i-k} R B^i, \\ Y = Z f_{(I,A)}(B), \end{cases} \quad (2.5)$$

where $Z \in R^{r \times p}$ is an arbitrary chosen free parameter matrix.

Proof. According to the relation (2.3), we have

$$\begin{cases} R_0 = I_n, \\ R_1 = \alpha_1 I_n + A, \\ R_2 = \alpha_2 I_n + \alpha_1 A + A^2, \\ \dots \\ R_{n-1} = \alpha_{n-1} I_n + \alpha_{n-2} A + \dots + A^{n-1}. \end{cases}$$

This relation can be compactly expressed as

$$R_i = \sum_{k=0}^i \alpha_k A^{i-k}, \alpha_0 = 1, i = 1, 2, \dots, n-1.$$

Hence, we can obtain

$$\sum_{i=0}^{n-1} R_i C Z B^i = \sum_{i=0}^{n-1} \sum_{k=0}^i \alpha_k A^{i-k} C Z B^i = \sum_{k=0}^{n-1} \sum_{i=0}^{n-1} \alpha_k A^{i-k} C Z B^i,$$

and

$$\sum_{i=0}^{n-1} R_i R B^i = \sum_{i=0}^{n-1} \sum_{k=0}^i \alpha_k A^{i-k} R B^i = \sum_{k=0}^{n-1} \sum_{i=0}^{n-1} \alpha_k A^{i-k} R B^i.$$

By Theorem 1, we can derive the conclusion. \square

Corollary 2. Let $A \in R^{n \times n}$, $B \in R^{p \times p}$, $C \in R^{n \times r}$ and $R \in R^{n \times p}$, suppose that $\{s | \det(I - sA) = 0\} \cap \lambda(B) = \emptyset$, and (2.2) holds. Then all solutions to the nonhomogeneous Yakubovich matrix equation (1.3) can be expressed as

$$\begin{cases} X = \sum_{i=0}^{n-1} R_i [C Z f_{(I,A)}(B) + R] B^i [f_{(I,A)}(B)]^{-1}, \\ Y = Z f_{(I,A)}(B), \end{cases}$$

where $Z \in R^{r \times p}$ is an arbitrary chosen free parameter matrix.

Proof. By considering (2.3), we can get

$$\begin{aligned} X f_{(I,A)}(B) &= \sum_{i=0}^{n-1} R_i C Z B^i f_{(I,A)}(B) + \sum_{i=0}^{n-1} R_i R B^i \\ &= \sum_{i=0}^{n-1} R_i C Z f_{(I,A)}(B) B^i + \sum_{i=0}^{n-1} R_i R B^i \\ &= \sum_{i=0}^{n-1} R_i (C Z f_{(I,A)}(B) + R) B^i \end{aligned}$$

Now Post-multiplying the two sides of the above relation by $(f_{(I,A)}(B))^{-1}$, we can obtain

$$X = \sum_{i=0}^{n-1} R_i (CZ f_{(I,A)}(B) + R) B^i [f_{(I,A)}(B)]^{-1}$$

The proof is finished. \square

Corollary 3. Given matrices $A \in R^{n \times n}$, $B \in R^{p \times p}$, $C \in R^{n \times r}$ and $R \in R^{n \times p}$, suppose that $\{s | \det(I - sA) = 0\} \cap \lambda(B) = \emptyset$, and (2.2) holds. Then all the solutions to the nonhomogeneous Yakubovich matrix equation (1.3) can be characterized by

$$\begin{cases} X = Ctr_n(A, CZ)S_r(I, A)Obs_n(B, f_{(I,A)}(B)) + Ctr_n(A, R)S_r(I, A)Obs_n(B, I_n) \\ Y = Zf_{(I,A)}(B), \end{cases}$$

where $Z \in R^{r \times p}$ is an arbitrary chosen free parameter matrix.

Proof. By the direct computation, it follows from Theorem 2 that we can get

$$\begin{aligned} \sum_{k=0}^{n-1} \sum_{i=0}^{n-1} \alpha_k(A)^{i-k} CZ B^i f_{(I,A)}(B) &= \sum_{k=0}^{n-1} \sum_{i=0}^{n-1} \alpha_k(A)^{i-k} CZ f_{(I,A)}(B) B^i \\ &= Ctr_n(A, CZ)S_r(I, A)Obs_n(B, f_{(I,A)}(B)) \end{aligned},$$

and

$$\sum_{k=0}^{n-1} \sum_{i=0}^{n-1} \alpha_k A^{i-k} R B^i = Ctr_n(A, R)S_r(I, A)Obs_n(B, I_n).$$

Hence, we have

$$\begin{aligned} \sum_{k=0}^{n-1} \sum_{i=0}^{n-1} \alpha_k(A)^{i-k} CZ B^i f_{(I,A)}(B) + \sum_{k=0}^{n-1} \sum_{i=0}^{n-1} \alpha_k A^{i-k} R B^i \\ = Ctr_n(A, CZ)S_r(I, A)Obs_n(B, f_{(I,A)}(B)) + Ctr_n(A, R)S_r(I, A)Obs_n(B, I_n) \end{aligned}.$$

The proof is finished. \square

3. Quaternion j-conjugate matrix equation $X - A\hat{X}B = CY + R$

In this section, we consider the solution to the quaternion j-conjugate matrix equation by applying of the real representation of a quaternion matrix. First of all, we introduce the definition and some properties of the real representation of a quaternion matrix.

3.1. Real representation of a quaternion matrix

For any quaternion matrix $A = A_1 + A_2i + A_3j + A_4k \in Q^{m \times n}$, $A_l \in R^{m \times n} (l = 1, 2, 3, 4)$, the real representation matrix of quaternion matrix A can be defined as

$$A_\sigma = \begin{bmatrix} A_1 & A_2 & -A_3 & A_4 \\ A_2 & -A_1 & -A_4 & -A_3 \\ A_3 & -A_4 & A_1 & A_2 \\ A_4 & A_3 & A_2 & -A_1 \end{bmatrix} \in R^{4m \times 4n}.$$

For a $m \times n$ quaternion matrix A , we define $A_\sigma^t = (A_\sigma)^t$. In addition, if we let

$$P_t = \begin{bmatrix} I_t & 0 & 0 & 0 \\ 0 & -I_t & 0 & 0 \\ 0 & 0 & I_t & 0 \\ 0 & 0 & 0 & -I_t \end{bmatrix}, Q_t = \begin{bmatrix} 0 & -I_t & 0 & 0 \\ I_t & 0 & 0 & 0 \\ 0 & 0 & 0 & I_t \\ 0 & 0 & -I_t & 0 \end{bmatrix},$$

$$S_t = \begin{bmatrix} 0 & 0 & 0 & -I_t \\ 0 & 0 & I_t & 0 \\ 0 & -I_t & 0 & 0 \\ I_t & 0 & 0 & 0 \end{bmatrix}, R_t = \begin{bmatrix} 0 & 0 & I_t & 0 \\ 0 & 0 & 0 & I_t \\ -I_t & 0 & 0 & 0 \\ 0 & -I_t & 0 & 0 \end{bmatrix}.$$

in which I_t is a $t \times t$ identity matrix, then P_t, Q_t, S_t, R_t are unitary matrices. Next, the properties of the real representation can be expressed as the following, which are given in [43].

Proposition 1 [43]. Assume that the quaternion matrices $A, B \in Q^{m \times n}, C \in Q^{n \times s}, a \in R$, then

- (1) $(A + B)_\sigma = A_\sigma + B_\sigma, (aA)_\sigma = aA_\sigma, (AC)_\sigma = A_\sigma P_n C_\sigma = A_\sigma (\hat{C})_\sigma P_s$;
- (2) $A = B \Leftrightarrow A_\sigma = B_\sigma$;
- (3) $Q_m^{-1} A_\sigma Q_n = -A_\sigma, R_m^{-1} A_\sigma R_n = A_\sigma, S_m^{-1} A_\sigma S_n = -A_\sigma, P_m^{-1} A_\sigma P_n = (\hat{A})_\sigma$;
- (4) The quaternion matrix A is nonsingular if and only if A_σ is nonsingular, and the quaternion matrix A is an unitary matrix if and only if A_σ is an orthogonal matrix;
- (5) If $A \in Q^{m \times m}$, then $A_\sigma^{2k} = ((A\hat{A})^k)_\sigma P_m$;
- (6) $A \in Q^{m \times m}, B \in Q^{n \times n}, C \in Q^{m \times n}$ and $k + l$ is even, then

$$A_\sigma^k C_\sigma B_\sigma^l = \begin{cases} ((A\hat{A})^s (A\hat{C}B) (\hat{B}B)^t)_\sigma, & k = 2s + 1, l = 2t + 1, \\ ((A\hat{A})^s C (\hat{B}B)^t)_\sigma, & k = 2s, l = 2t. \end{cases}$$

Proposition 2 [43]. If λ is a characteristic value of A_σ , then so are $\pm\lambda, \pm\bar{\lambda}$.

For any $A \in Q^{m \times m}$, let the characteristic polynomial of the real representation matrix A_σ be $f_{(I, A_\sigma)}(\lambda) = \det(I_{4m} - \lambda A_\sigma) = \sum_{k=0}^{2m} a_{2k} \lambda^{2k}$, and define $h_{A_\sigma}(\lambda) = \lambda^{4m} f_{(I, A_\sigma)}(\lambda^{-1}) = \sum_{k=0}^{2m} a_{2k} \lambda^{2(2m-k)}$. So by Proposition 1 and Proposition 2 we have the following Proposition 3.

Proposition 3. Let $A \in Q^{m \times m}, B \in Q^{n \times n}$. Then

- (1) $f_{(I, A_\sigma)}(\lambda)$ is a real polynomial, and $f_{(I, A_\sigma)}(\lambda) = \sum_{k=0}^{2m} a_{2k} \lambda^{2k}$;
- (2) $h_{A_\sigma}(\lambda)$ is a real polynomial, and $h_{A_\sigma}(\lambda) = \sum_{k=0}^{2m} a_{2k} \lambda^{2(2m-k)}$;
- (3) $h_{A_\sigma}(B_\sigma) = (g_{A_\sigma}(B\tilde{B}))_\sigma P_n, f_{(I, A_\sigma)}(B_\sigma) = (p_{A_\sigma}(B\tilde{B}))_\sigma P_n$,

in which $g_{A_\sigma}(\lambda) = \sum_{k=0}^{2m} a_{2k} \lambda^{2m-k}, p_{A_\sigma}(\lambda) = \sum_{k=0}^{2m} a_{2k} \lambda^k$ are real polynomials.

Proof. By Proposition 2, we easily know that a_k is a real number, and $a_{2k+1} = 0$. For any k , by Proposition 1, we have $B_\sigma^{2k} = ((B\tilde{B})^k)_\sigma P_n$, so we can obtain the result (3).

3.2. On solutions to the quaternion j-conjugate matrix equation $X - A\hat{X}B = CY + R$

In this subsection, the solution to the quaternion j-conjugate matrix equation

$$X - A\hat{X}B = CY + R, \quad (3.1)$$

is presented by the use of real representation of a quaternion matrix, in which $A \in Q^{n \times n}, B \in Q^{p \times p}, C \in Q^{n \times r}$ and $R \in Q^{n \times p}$ are given quaternion matrices, $X \in Q^{n \times p}$ and $Y \in Q^{r \times p}$ are the determined quaternion matrices. Moreover, the real representation matrix equation of quaternion matrix equation (3.1) is defined as

$$V - A_\sigma V B_\sigma = C_\sigma P_r W + R_\sigma. \quad (3.2)$$

According to (1) in Proposition 1, the quaternion matrix equation (3.1) is equivalent to the matrix equation $(X - A\hat{X}B)_\sigma = X_\sigma - A_\sigma X_\sigma B_\sigma$. Hence, the matrix equation (3.1) can be converted into $X_\sigma - A_\sigma X_\sigma B_\sigma = C_\sigma P_r Y_\sigma + R_\sigma$. Therefore, we have the following conclusions.

Theorem 3. Suppose that $A \in Q^{n \times n}, B \in Q^{p \times p}, C \in Q^{n \times r}$ and $R \in Q^{n \times p}$, the quaternion matrix equation (3.1) has a solution (X, Y) if and only if the real representation matrix

equation (3.2) has a solution $(V, W) = (X_\sigma, Y_\sigma)$. Furthermore, if (V, W) is a solution to (3.2), the following quaternion matrices are solutions to the quaternion matrix equation (3.1)

$$\begin{cases} X = \frac{1}{16} \begin{bmatrix} I_n & iI_n & jI_n & kI_n \end{bmatrix} (V - Q_n^{-1}VQ_p + R_n^{-1}VR_p - S_n^{-1}VS_p) \begin{bmatrix} I_p \\ -iI_p \\ -jI_p \\ -kI_p \end{bmatrix}, \\ Y = \frac{1}{16} \begin{bmatrix} I_r & iI_r & jI_r & kI_r \end{bmatrix} (W - Q_n^{-1}WQ_p + R_n^{-1}WR_p - S_n^{-1}WS_p) \begin{bmatrix} I_p \\ -iI_p \\ -jI_p \\ -kI_p \end{bmatrix}. \end{cases} \quad (3.3)$$

Proof. By (3) of Proposition 1, the matrix equation (3.2) is equivalent to

$$V - R_n^{-1}A_\sigma R_n V R_p^{-1} B_\sigma R_p = R_n^{-1}C_\sigma R_r P_r W + R_\sigma. \quad (3.4)$$

Post-multiplying the two sides of the quaternion matrix equation (3.4) by R_p^{-1} , we can obtain

$$V R_p^{-1} - R_n^{-1}A_\sigma R_n V R_p^{-1} B_\sigma = R_n^{-1}C_\sigma R_r P_r W R_p^{-1} + R_\sigma R_p^{-1}. \quad (3.5)$$

Pre-multiplying the two sides of (3.5) by R_n and $R_p^{-1} = -R_p$, $R_r P_r = P_r R_r$, we can get

$$R_n^{-1}V R_p - A_\sigma R_n^{-1}V R_p B_\sigma = C_\sigma P_r R_r^{-1}W R_p + R_\sigma. \quad (3.6)$$

This shows that if (V, W) is a real solution of matrix equation (3.2), $(R_n^{-1}V R_p, R_r^{-1}W R_p)$ is also a real solution of matrix equation (3.2). In addition, according to (3) of Proposition 1, the matrix equation (3.2) is also equivalent to

$$V - Q_n A_\sigma Q_n V Q_p B_\sigma Q_p = Q_n C_\sigma Q_r P_r W + Q_n R_\sigma Q_p. \quad (3.7)$$

Post-multiplying the two sides of the quaternion matrix equation (3.7) by Q_p^{-1} , we have

$$V Q_p^{-1} - Q_n A_\sigma Q_n V Q_p B_\sigma = Q_n C_\sigma Q_r P_r W Q_p^{-1} + Q_n R_\sigma. \quad (3.8)$$

Note $Q_p^{-1} = -Q_p$, $Q_r P_r = -P_r Q_r$, pre-multiplying the two sides of the quaternion matrix equation (3.8) by Q_n^{-1} , we conclude that

$$(-Q_n^{-1}V Q_p) - A_\sigma(-Q_n^{-1}V Q_p) B_\sigma = C_\sigma P_r(-Q_r^{-1}W Q_p) + R_\sigma.$$

This means $(-Q_n^{-1}VQ_p, -Q_r^{-1}WQ_p)$ is also a real solution of matrix equation (3.2) if (V, W) is a real solution of matrix equation (3.2). Similarly, we can prove $(-S_n^{-1}VS_p, -S_r^{-1}WS_p)$ is also a real solution of quaternion matrix equation (3.2). In this case, the conclusion can be obtained along the line of the proof of Theorem 4.2 in [43]. \square

Theorem 4. Given quaternion matrices $A \in Q^{n \times n}$, $B \in Q^{p \times p}$, $C \in Q^{n \times r}$, $R \in Q^{n \times p}$, let

$$f_{(I, A_\sigma)}(s) = \det(I_{4n} - sA_\sigma) = \sum_{k=0}^{2n} a_{2k} s^{2k}, \quad p_{A_\sigma}(s) = \sum_{k=0}^{2n} a_{2k} s^k,$$

then the solutions to the quaternion j-conjugate matrix equation (3.1) can be characterized by

$$\begin{cases} X p_{A_\sigma}(\widehat{B}B) = \sum_{k=0}^{2n-1} \sum_{s=0}^{2n-1} \alpha_{2k} (A\widehat{A})^{s-k} (CZ + A\widehat{C}\widehat{Z}B) (\widehat{B}B)^s p_{A_\sigma}(\widehat{B}B) \\ \quad + \sum_{k=0}^{2n-1} \sum_{s=0}^{2n-1} \alpha_{2k} (A\widehat{A})^{s-k} (R + A\widehat{R}B) (\widehat{B}B)^s, \\ Y = Z p_{A_\sigma}(\widehat{B}B), \end{cases}$$

in which Z is an arbitrary quaternion matrix.

Proof. If the Yakubovich quaternion j-conjugate matrix equation (3.1) has solution (X, Y) , then its real representation matrix equation (3.2) has solution $(V, W) = (X_\sigma, Y_\sigma)$ with the free parameter Z_σ . Now according to Theorem 2 and 3, we can conclude that

$$X_\sigma f_{(I, A_\sigma)}(B_\sigma) = \sum_{k=0}^{2n-1} \sum_{j=0}^{4n-1} \alpha_{2k} A_\sigma^{j-2k} C_\sigma P_r Z_\sigma B_\sigma^j f_{(I, A_\sigma)}(B_\sigma) + \sum_{k=0}^{2n-1} \sum_{j=0}^{4n-1} \alpha_{2k} A_\sigma^{j-2k} R_\sigma B_\sigma^j,$$

and

$$Y_\sigma = Z_\sigma f_{(I, A_\sigma)}(B_\sigma).$$

Moreover, by Proposition 1 and Proposition 3, we can get

$$f_{(I, A_\sigma)}(B_\sigma) = (p_{A_\sigma}(B\widehat{B}))_\sigma P_p.$$

Hence, we can obtain

$$X_\sigma f_{(I, A_\sigma)}(B_\sigma) = X_\sigma [p_{A_\sigma}(B\widehat{B})]_\sigma P_p = [X p_{A_\sigma}(\widehat{B}B)]_\sigma,$$

and

$$Y_\sigma = Z_\sigma f_{(I, A_\sigma)}(B_\sigma) = Z_\sigma(p_{A_\sigma}(\widehat{B}\widehat{B}))_\sigma P_p = (Zp_{A_\sigma}(\widehat{B}\widehat{B}))_\sigma.$$

Furthermore, by the use of Proposition 1 we have

$$\begin{aligned} & \sum_{k=0}^{2n-1} \sum_{j=0}^{4n-1} \alpha_{2k} A_\sigma^{j-2k} C_\sigma P_r Z_\sigma B_\sigma^j f_{(I, A_\sigma)}(B_\sigma) \\ &= \sum_{k=0}^{2n-1} \alpha_{2k} \left[\sum_{s=0}^{2n-1} A_\sigma^{2s-2k} C_\sigma P_r Z_\sigma B_\sigma^{2s} + \sum_{s=1}^{2n} A_\sigma^{2s-2k+1} C_\sigma P_r Z_\sigma B_\sigma^{2s+1} \right] f_{(I, A_\sigma)}(B_\sigma) \\ &= \sum_{k=0}^{2n-1} \sum_{s=0}^{2n-1} \alpha_{2k} \left[((A\widehat{A})^{s-k})_\sigma P_n C_\sigma P_r Z_\sigma ((B\widehat{B})^s)_\sigma P_p + ((A\widehat{A})^{s-k})_\sigma P_n A_\sigma C_\sigma P_r Z_\sigma B_\sigma ((B\widehat{B})^s)_\sigma P_p \right] f_{(I, A_\sigma)}(B_\sigma) \\ &= \sum_{k=0}^{2n-1} \sum_{s=0}^{2n-1} \alpha_{2k} \left[((A\widehat{A})^{s-k} C Z (\widehat{B}\widehat{B})^s)_\sigma + ((A\widehat{A})^{s-k} A \widehat{C} \widehat{Z} B (\widehat{B}\widehat{B})^s)_\sigma \right] f_{(I, A_\sigma)}(B_\sigma) \\ &= \sum_{k=0}^{2n-1} \sum_{s=0}^{2n-1} \alpha_{2k} \left[((A\widehat{A})^{s-k} C Z (\widehat{B}\widehat{B})^s p_{A_\sigma}(\widehat{B}\widehat{B}))_\sigma + ((A\widehat{A})^{s-k} A \widehat{C} \widehat{Z} B (\widehat{B}\widehat{B})^s p_{A_\sigma}(\widehat{B}\widehat{B}))_\sigma \right], \end{aligned}$$

and

$$\begin{aligned} & \sum_{k=0}^{2n-1} \sum_{j=0}^{4n-1} \alpha_{2k} A_\sigma^{j-2k} R_\sigma B_\sigma^j = \sum_{k=0}^{2n-1} \alpha_{2k} \left[\sum_{s=0}^{2n-1} A_\sigma^{2s-2k} R_\sigma B_\sigma^{2s} + \sum_{s=1}^{2n} A_\sigma^{2s+1-2k} R_\sigma B_\sigma^{2s+1} \right] \\ &= \sum_{k=0}^{2n-1} \alpha_{2k} \left[\sum_{s=0}^{2n-1} ((A\widehat{A})^{s-k})_\sigma P_n R_\sigma ((\widehat{B}\widehat{B})^s)_\sigma P_p + \sum_{s=1}^{2n} ((A\widehat{A})^{s-k})_\sigma P_n A_\sigma R_\sigma B_\sigma ((\widehat{B}\widehat{B})^s)_\sigma P_p \right] \\ &= \sum_{k=0}^{2n-1} \alpha_{2k} \left[\sum_{s=0}^{2n-1} ((A\widehat{A})^{s-k} R (\widehat{B}\widehat{B})^s)_\sigma + \sum_{s=0}^{2n-1} ((A\widehat{A})^{s-k} A \widehat{R} B (\widehat{B}\widehat{B})^s)_\sigma \right] \\ &= \sum_{k=0}^{2n-1} \sum_{s=0}^{2n-1} \alpha_{2k} \left[((A\widehat{A})^{s-k} R (\widehat{B}\widehat{B})^s)_\sigma + ((A\widehat{A})^{s-k} A \widehat{R} B (\widehat{B}\widehat{B})^s)_\sigma \right]. \end{aligned}$$

The proof is finished. \square

In the following, we provide an equivalent statement of the above Theorem 4.

Theorem 5. Given quaternion matrices $A \in Q^{n \times n}$, $B \in Q^{p \times p}$, $C \in Q^{n \times r}$ and $R \in Q^{n \times p}$, let

$$f_{(I, A_\sigma)}(s) = \det(I_{4n} - sA_\sigma) = \sum_{k=0}^{2n} a_{2k} s^{2k}, \quad p_{A_\sigma}(s) = \sum_{k=0}^{2n} a_{2k} s^k,$$

then the solutions to the quaternion j -conjugate matrix equation (3.1) can be characterized

as

$$\left\{ \begin{array}{l} Xp_{A_\sigma}(\hat{B}B) = Ctr_{2n}(A\hat{A}, C)S_r(I, A_\sigma)Obs_{2n}(\hat{B}B, Zp_{A_\sigma}(\hat{B}B)) + Ctr_{2n}(A\hat{A}, A\hat{C})S_r(I, A_\sigma) \\ \quad Obs_{2n}(\hat{B}B, \hat{Z}Bp_{A_\sigma}(\hat{B}B)) + Ctr_{2n}(A\hat{A}, R)S_r(I, A_\sigma)Obs_{2n}(\hat{B}B, I_r) \\ \quad + Ctr_{2n}(A\hat{A}, A, 2n)S_r(I, A_\sigma)Obs_{2n}(\hat{B}B, \hat{R}B), \\ Y = Zp_{A_\sigma}(\hat{B}B), \end{array} \right.$$

in which Z is an arbitrary quaternion matrix.

Proof. By the direct computation, we can get

$$\sum_{k=0}^{2n-1} \sum_{s=0}^{2n-1} \alpha_{2k}(A\hat{A})^{s-k} CZ(\hat{B}B)^s = Ctr_{2n}(A\hat{A}, C)S_r(I, A_\sigma)Obs_{2n}(\hat{B}B, Zp_{A_\sigma}(\hat{B}B)),$$

and

$$\sum_{k=0}^{2n-1} \sum_{s=0}^{2n-1} \alpha_{2k}(A\hat{A})^{s-k} A\hat{C}\hat{Z}B(\hat{B}B)^s = Ctr_{2n}(A\hat{A}, A\hat{C})S_r(I, A_\sigma)Obs_{2n}(\hat{B}B, \hat{Z}Bp_{A_\sigma}(\hat{B}B)).$$

Moreover, we can obtain

$$\sum_{k=0}^{2n-1} \sum_{s=0}^{2n-1} \alpha_{2k}(A\hat{A})^{s-k} R(\hat{B}B)^s = Ctr_{2n}(A\hat{A}, R)S_r(I, A_\sigma)Obs_{2n}(\hat{B}B, I_r),$$

and

$$\sum_{k=0}^{2n-1} \sum_{s=0}^{2n-1} \alpha_{2k}(A\hat{A})^{s-k} A\hat{R}B(\hat{B}B)^s = Ctr_{2n}(A\hat{A}, A)S_r(I, A_\sigma)Obs_{2n}(\hat{B}B, \hat{R}B).$$

Thus, the first conclusion has been proved. With this the second conclusion is obviously true. \square

4. Examples

4.1. Numerical Examples

In the following, we give two numerical examples and application examples to show the effectiveness of the obtained results.

Example 1 . Consider the nonhomogeneous Yakubovich matrix equation in the form of (1.3) with the following parameters:

$$A = \begin{pmatrix} 1 & 2 & 4 & -6 & -1 & -2 \\ -8 & -1 & 0 & -8 & 0 & 1 \\ 1 & 9 & 6 & -12 & 4 & 0 \\ -5 & -2 & 1 & -9 & 0 & 0 \\ 0 & 1 & 2 & -3 & -1 & 0 \\ 1 & 0 & 0 & 1 & -2 & 3 \end{pmatrix}, B = \begin{pmatrix} -11 & 1 & 0 & 7 & 2 & 1 & 0 & -11 \\ -9 & -8 & 1 & 19 & 21 & 23 & 24 & 2 \\ 1 & -2 & 3 & 9 & 11 & 5 & -4 & 3 \\ 18 & 1 & 4 & 6 & 7 & 9 & -1 & 0 \\ 9 & 8 & -7 & 6 & 5 & 4 & 3 & -2 \\ 0 & 1 & 11 & -2 & -3 & -4 & 5 & 4 \\ 11 & -1 & -2 & -3 & 4 & 5 & 0 & -5 \\ -4 & 2 & 3 & 6 & -7 & 9 & -19 & 0 \end{pmatrix},$$

$$C = \begin{pmatrix} -6.8367 & 6.9493 & -9.0781 & 2.5301 & -3.9528 & -5.5645 & -1.0500 & 2.2414 \\ 19.3891 & -7.2607 & 13.7305 & -8.0050 & 6.7793 & 7.7253 & -9.2575 & -1.0242 \\ -12.8797 & -8.9994 & -17.5988 & 9.5575 & 2.0448 & -11.2606 & 5.3451 & 11.1267 \\ 14.6108 & -3.0322 & 10.1333 & -5.9000 & 1.7579 & 7.7094 & -7.7738 & -1.9275 \\ 0.5285 & -1.7823 & -2.0056 & 0.7015 & -1.2497 & -0.5900 & -0.8977 & 0.7600 \\ 9.4432 & -17.0285 & 5.1795 & 0.1731 & -0.6505 & 6.7242 & 0.9935 & -1.7142 \end{pmatrix},$$

$$R = \begin{pmatrix} 1 & 2 & 3 & -11 & 4 & -3 & 9 & 9 \\ 4 & 3 & 7 & -12 & 3 & 4 & 10 & 7 \\ -1 & 4 & 6 & 13 & 2 & 5 & 11 & 6 \\ -12 & -9 & 5 & 14 & -1 & 6 & 12 & 5 \\ 1 & 2 & -4 & 6 & -1 & 7 & 11 & 4 \\ 6 & 7 & 3 & 7 & -2 & 8 & 13 & 3 \end{pmatrix}.$$

It is easy to check that $\{s | \det(I - sA) = 0\} \cap \lambda(B) = \emptyset$. By simple computations, we have

$$f_{(I,A)}(s) = \det(I - sA) = 752s^6 - 2648s^5 + 110s^4 + 529s^3 - 92s^2 + s + 1,$$

$$\alpha_6 = 752, \alpha_5 = -2648, \alpha_4 = 110, \alpha_3 = 529, \alpha_2 = -92, \alpha_1 = 1, \alpha_0 = 1,$$

$$[f_{(I,A)}(B)]^{-1} = 10^{-4} \times \begin{pmatrix} -0.0117 & 0.0064 & 0.0121 & -0.0416 & 0.0177 & 0.0315 & 0.0466 & 0.0111 \\ 0.0261 & -0.0143 & -0.0269 & 0.0928 & -0.0396 & -0.0702 & -0.1040 & -0.0247 \\ 0.0046 & -0.0025 & -0.0047 & 0.0162 & -0.0069 & -0.0123 & -0.0182 & -0.0043 \\ -0.0223 & 0.0122 & 0.0230 & -0.0791 & 0.0337 & 0.0599 & 0.0887 & 0.0210 \\ 0.0154 & -0.0084 & -0.0159 & 0.0549 & -0.0234 & -0.0415 & -0.0615 & -0.0146 \\ 0.0133 & -0.0073 & -0.0137 & 0.0473 & -0.0202 & -0.0358 & -0.0531 & -0.0126 \\ -0.0017 & 0.0009 & 0.0017 & -0.0060 & 0.0026 & 0.0045 & 0.0067 & 0.0016 \\ 0.0073 & -0.0040 & -0.0075 & 0.0259 & -0.0110 & -0.0196 & -0.0290 & -0.0069 \end{pmatrix},$$

$$\begin{aligned}
 R_4 &= \begin{pmatrix} 71511 & -250742 & -151984 & 219259 & -510594 & -884097 \\ -310094 & 995853 & 901646 & -452672 & 1219872 & 1634013 \\ -617957 & 1092862 & 1659610 & 306572 & -67287 & -2376366 \\ 51196 & 319077 & 658978 & 689135 & -879342 & -2518257 \\ -15284 & 57010 & 50659 & 28685 & 24113 & -166172 \\ -189958 & 122836 & 126710 & -397128 & 742797 & 1698921 \end{pmatrix}, \\
 R_3 &= \begin{pmatrix} 40 & -4938 & -7886 & -4554 & 6482 & 13963 \\ 1005 & 4570 & 17506 & 20766 & -31423 & -73745 \\ 13188 & -15327 & -15015 & 29613 & -57452 & -134461 \\ 578 & -29546 & -15099 & 10990 & -37314 & -40491 \\ 3721 & 504 & -1626 & 1667 & -5308 & -5002 \\ -8426 & 7928 & 15041 & 222 & 2923 & -12609 \end{pmatrix}, \\
 R_2 &= \begin{pmatrix} 26 & 289 & 38 & -125 & 207 & 444 \\ -128 & -934 & -430 & 190 & -747 & -969 \\ 714 & -414 & -1265 & -136 & -72 & 1257 \\ -510 & -319 & -364 & -638 & 1227 & 1350 \\ -81 & -119 & -22 & 15 & 115 & 227 \\ 68 & -46 & -79 & 211 & -405 & -1278 \end{pmatrix}, \\
 R_1 &= \begin{pmatrix} 2 & 2 & 4 & 6 & -1 & -2 \\ -8 & 0 & -11 & -18 & 29 & 31 \\ 1 & 9 & 7 & -12 & 34 & 98 \\ 25 & 42 & 1 & -8 & 0 & 0 \\ 0 & 1 & 2 & -3 & -10 & 0 \\ 11 & 0 & -12 & 1 & -2 & 4 \end{pmatrix}, R_0 = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}, \\
 R_5 &= \begin{pmatrix} -2320777 & 7724192 & 11939537 & 5440584 & -5125757 & -25462747 \\ -717929 & -11296227 & -24405967 & -28557225 & 43159177 & 124754978 \\ -26176969 & 25416451 & 24150092 & -48228231 & 94230584 & 190627318 \\ -12314894 & 33778445 & 29798340 & -19429040 & 46316565 & 66814068 \\ -1531472 & 1597236 & 1686683 & -2222468 & 3455433 & 6263944 \\ 7713995 & -15298941 & -20649354 & -1826634 & -3508267 & 21699527 \end{pmatrix}, \\
 R_6 &= 10^9 \begin{pmatrix} -0.1962 & 0.3185 & 0.2781 & -0.3553 & 0.7396 & 1.3378 \\ 0.7236 & -1.3660 & -1.4643 & 0.7442 & -1.8809 & -2.3663 \\ 0.6858 & -1.7918 & -2.3867 & -0.5624 & 0.1666 & 3.7789 \\ -0.0035 & -0.5599 & -0.9706 & -0.9368 & 1.3619 & 4.1924 \\ 0.0007 & -0.0794 & -0.0841 & -0.0423 & 0.0547 & 0.2367 \\ 0.3025 & -0.2353 & -0.1940 & 0.6181 & -1.1583 & -2.4482 \end{pmatrix},
 \end{aligned}$$

Choose $Z = \begin{pmatrix} 1 & 2 & 19 & 2 & 2 & 21 & 0 & -1 \\ 4 & 3 & 21 & 0 & 3 & 12 & 2 & 0 \\ -1 & -13 & 22 & 1 & 4 & -13 & 3 & 1 \\ 1 & -14 & 23 & -1 & -5 & 6 & 4 & -9 \\ 9 & 1 & 2 & 3 & 4 & 0 & 1 & -1 \\ 2 & 3 & -1 & 0 & 0 & 9 & 6 & 2 \\ 9 & -1 & 2 & -9 & 8 & 7 & 2 & -1 \\ -10 & -9 & 8 & 7 & 0 & 10 & 11 & 13 \end{pmatrix}$. So from Theorem 1 we have

$$X = \begin{pmatrix} 2 & 3 & 36 & 16 & 9 & 4 & 1 & 9 \\ 3 & 4 & 48 & 1 & 7 & 9 & 2 & -1 \\ 4 & 2 & 1 & 2 & 6 & 2 & -3 & -2 \\ -5 & 0 & -1 & 3 & -2 & 7 & 0 & 0 \\ -6 & 11 & 0 & 4 & -3 & 8 & -9 & -3 \\ 7 & 29 & 15 & -4 & -2 & 9 & -8 & 0 \end{pmatrix},$$

$$Y = 10^{12} \times \begin{pmatrix} 0.9967 & 0.1219 & 0.1325 & 0.7507 & 1.2366 & 0.7164 & 1.2608 & -0.2614 \\ 1.4717 & 0.1579 & 0.1926 & 0.4968 & 1.2267 & 0.5219 & 1.5579 & 0.0043 \\ 0.1493 & 0.2854 & -0.1894 & 0.1609 & -0.1677 & 0.2949 & -0.9352 & -0.5724 \\ -0.6657 & 0.0236 & -0.2209 & 0.2364 & -0.2277 & 0.2602 & -0.9383 & -0.5034 \\ 0.3622 & 0.1181 & 0.0215 & 0.1922 & 0.3992 & 0.1250 & 0.6353 & -0.1945 \\ 0.2367 & 0.0378 & 0.0426 & 0.1431 & 0.2769 & 0.1123 & 0.3676 & -0.0511 \\ -0.7892 & -0.0015 & -0.0096 & 0.2866 & 0.0100 & -0.0059 & 0.1746 & -0.3507 \\ -0.2435 & 0.3674 & -0.2141 & 0.6507 & -0.0918 & 0.7113 & -1.5539 & -1.0505 \end{pmatrix}.$$

4.2. Application Examples

Lemma 2 [15]. The generalized Sylvester matrix equation (1.1), where $A, E \in R^{n \times n}$, $B \in R^{n \times r}$, and $F \in R^{p \times p}$ are known and the matrix pair (E, A) is regular, is equivalent to the generalized discrete Sylvester matrix equation (1.2) with

$$M = (\gamma E - A)^{-1}E, \quad N = \gamma I - F, \quad T = (\gamma E - A)^{-1}B,$$

where γ is an arbitrary scalar such that $(\gamma E - A)$ is nonsingular.

Proof. Since the matrix pair (E, A) is regular, there exists a scalar γ such that $(\gamma E - A)$ is nonsingular. Premultiplying (1.2) by $(\gamma E - A)^{-1}$ produces

$$(\gamma E - A)^{-1}AX - (\gamma E - A)^{-1}EXF = (\gamma E - A)^{-1}BY, \quad (4.1)$$

Let $M = (\gamma E - A)^{-1}E$ and note that

$$\begin{aligned} \gamma M - (\gamma E - A)^{-1}A &= \gamma(\gamma E - A)^{-1}E - (\gamma E - A)^{-1}A \\ &= (\gamma E - A)^{-1}(\gamma E - A) \\ &= I, \end{aligned}$$

We have $(\gamma E - A)^{-1}A = \gamma M - I$. So (4.1) is equivalent to

$$(\gamma M - I)X - MXF = TY$$

or

$$MX(\gamma I - F)X - X = TY.$$

Let $N = \gamma I - F$, then the above equation is reduced to (1.1).

Example 2. Consider the following linear system:

$$E\dot{x} = Ax + Bu, \quad (4.2)$$

where $A, E \in R^{n \times n}$ and $B \in R^{n \times r}$ are known coefficient matrices. If the following state feedback controller

$$u = -K_p x - K_d \dot{x}, \quad (4.3)$$

is applied to the above system (4.2), the closed-loop system becomes

$$(E + BK_d)\dot{x} = (A - BK_p)x, \quad (4.4)$$

The eigenstructure assignment problem is to determine the matrix K_p such that the closed-loop system matrix $A - BK_p$ has desired eigenvalues and struture, i.e., determine the matrix K_p such that

$$A - BK_p = EXFX^{-1}, \quad (4.5)$$

where F is the desired Jordan form of the closed-loop system and X is the corresponding eigenvector matrix. If we choose $K_p X + K_d X F = Y$, the equation (4.5) is equivalent to the generalized Sylvester matrix equation $AX - EXF = BY$. By Lemma 2, the generalized Sylvester matrix equation $AX - EXF = BY$ is equivalent into the Yakubovich matrix equation $X - MXF = NY$. By Corollary 2, we can obtain the solution to Yakubovich matrix equation $X - MXF = NY$.

Example 3 . Consider the following linear system:

$$A\dot{x} = x + Bu, \quad (4.6)$$

where $A \in R^{n \times n}$ and $B \in R^{n \times r}$ are known coefficient matrices. If the following state feedback controller

$$u = -K_p x, \quad (4.7)$$

is applied to the above system (4.6), the closed-loop system becomes

$$A\dot{x} = (I - BK_p)x, \quad (4.8)$$

The eigenstructure assignment problem is to determine the matrix K_p such that the closed-loop system matrix $I - BK_p$ has desired eigenvalues and struture, i.e., determine the matrix K_p such that

$$AXF = X - BK_pX, \quad (4.9)$$

where F is the desired Jordan form of the closed-loop system and X is the corresponding eigenvector matrix. If we choose $K_pX = Y$, the equation (4.9) is equivalent to the generalized Sylvester matrix equation $X - AXF = BY$. We choose the following parametric matrices.

$$A = \begin{pmatrix} 1 & 21 & 3 & 4 \\ 2 & 3 & 11 & 132 \\ 2 & 3 & 4 & 0 \\ 3 & 4 & 0 & 45 \end{pmatrix}, B = \begin{pmatrix} 10 & 0 & 9 & 19 \\ 21 & 2 & 7 & 20 \\ 2 & 33 & 8 & 10 \\ 3 & 4 & 1 & 0 \end{pmatrix}.$$

$$F = \begin{pmatrix} 6 & -11 & 14 & 12 & 1 & -29 \\ 1 & 129 & 118 & 11 & -2 & -8 \\ -6 & -4 & -2 & -131 & 2 & -6 \\ 9 & 7 & -6 & 5 & 14 & -3 \\ 21 & 23 & 24 & 126 & 115 & 119 \\ 26 & -11 & 16 & -71 & -9 & -112 \end{pmatrix},$$

By computation, one can obtain

$$f_A(s) = 43069s^4 - 3107s^3 - 242s^2 - 53s + 1,$$

$$\alpha_4 = 43069, \alpha_3 = -3107, \alpha_2 = -242, \alpha_1 = -53, \alpha_0 = 1.$$

By (2.10), we have

$$R_3 = \begin{pmatrix} 3057 & 3311 & -11398 & -9984 \\ -2214 & 138 & 1281 & -208 \\ 132 & -1759 & -6029 & 5148 \\ -7 & -233 & 646 & -273 \end{pmatrix}, R_2 = \begin{pmatrix} -234 & -1004 & 87 & 2744 \\ 320 & 211 & -500 & -652 \\ -90 & -96 & -399 & 404 \\ -13 & 43 & 53 & -62 \end{pmatrix},$$

$$R_1 = \begin{pmatrix} -52 & 21 & 3 & 4 \\ 2 & -50 & 11 & 132 \\ 2 & 3 & -49 & 0 \\ 3 & 4 & 0 & -8 \end{pmatrix}, R_0 = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}.$$

If we choose $Z = 10^{-3} * \begin{pmatrix} 0.1078 & 0.0030 & 0.0277 & 0.1043 & -0.0177 & -0.0525 \\ -0.0108 & -0.0003 & -0.0028 & -0.0104 & 0.0018 & 0.0053 \\ 0.2041 & 0.0057 & 0.0528 & 0.1972 & -0.0335 & -0.0996 \\ -0.1479 & -0.0041 & -0.0382 & -0.1429 & 0.0243 & 0.0721 \end{pmatrix}$,
then by (2.9) we can derive

$$Y = \begin{pmatrix} -83150 & 19690 & -57530 & 286620 & 59350 & 400540 \\ -5200 & 880 & -3340 & 17640 & 3060 & 23820 \\ -26860 & 10120 & -20370 & 97890 & 25940 & 142270 \\ 55140 & -14770 & 38330 & -19173 & -41460 & -269840 \end{pmatrix},$$

$$X = \begin{pmatrix} 1 & 0 & 7 & -11 & -25 & 129 \\ -2 & 2 & -6 & -29 & -2 & 11 \\ -21 & 4 & -5 & -39 & -1 & -21 \\ -98 & 9 & -9 & -48 & -9 & 291 \end{pmatrix}.$$

5. Conclusions

In this paper we mainly consider the explicit solutions to two types of matrix equations. Firstly, we provide an alternative approach to solve the explicit solution to the nonhomogeneous Yakubovich matrix equation. These obtained solutions can provide all the degrees of freedom, which is represented by an arbitrarily chosen parameter matrix Z . All the coefficient matrices are not restricted to be in any canonical form. The matrix B explicitly appears in the solutions, thus can be unknown a priori. Secondly, we present the solutions to the nonhomogeneous Yakubovich j -conjugate matrix equation in quaternion field by the real representation of a quaternion matrix. We generalize our previous results in [25, 35, 32].

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