

## Research Article

# Polynomial Solutions to the Matrix Equation $X - AX^T B = C$

Caiqin Song<sup>1,2</sup> and Jun-e Feng<sup>1</sup>

<sup>1</sup> School of Mathematics, Shandong University, Jinan 250100, China

<sup>2</sup> College of Information Science and Engineering, Shandong University of Science and Technology, Qingdao 266590, China

Correspondence should be addressed to Caiqin Song; songcaiqin@gmail.com and Jun-e Feng; fengjune@sdu.edu.cn

Received 31 October 2013; Accepted 27 December 2013; Published 16 January 2014

Academic Editor: Yongkun Li

Copyright © 2014 C. Song and J.-e. Feng. This is an open access article distributed under the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Solutions are constructed for the Kalman-Yakubovich-transpose equation  $X - AX^T B = C$ . The solutions are stated as a polynomial of parameters of the matrix equation. One of the polynomial solutions is expressed by the symmetric operator matrix, controllability matrix, and observability matrix. Moreover, the explicit solution is proposed when the Kalman-Yakubovich-transpose matrix equation has a unique solution. The provided approach does not require the coefficient matrices to be in canonical form. In addition, the numerical example is given to illustrate the effectiveness of the derived method. Some applications in control theory are discussed at the end of this paper.

## 1. Introduction

In the control area, the Kalman-Yakubovich-transpose matrix equation  $X - AX^T B = C$  occurs in fault detection [1], control with constrains systems [2], eigenstructure assignment [3], and observer design [4]. In order to obtain explicit solutions, many researchers have made much efforts. Braden [5] studies the Lyapunov-transpose matrix equation  $A^T X \pm X^T A = B$  via matrix decomposition. Liao et al. [6] propose an effective method to obtain the least square solution of the matrix equation  $A^T X B + B^T X^T A = D$  using GSVD, CCD and projective theorem. Piao et al. [7] investigate the matrix equation  $AX + X^T B = C$  by the Moore-Penrose generalized inverse and give the explicit solutions for the Sylvester-transpose matrix equation. Song et al. [8, 9] establish the explicit solution of the quaternion matrix equation  $XF - AX = C$  and  $X - A\bar{X}F = C$ , where  $\bar{X}$  denotes the  $j$ -conjugate of the quaternion matrix. Moreover, other matrix equations such as the coupled Sylvester matrix equations and the Riccati equations have also been found numerous applications in control theory. For more related introduction, see [10, 11] and the references therein. The matrix equation  $AXB + CX^T D = E$

is considered by the iterative algorithm [12, 13]. In [14, 15], the following linear equation

$$\sum_{i=1}^r A_i X B_i + \sum_{j=1}^s C_j X^T D_j = E \quad (1)$$

is considered, where  $A_i, B_i, C_j, D_j$  ( $i = 1, \dots, r; j = 1, \dots, s$ ), and  $E$  are some known constant matrices of appropriate dimensions and  $X$  is a matrix to be determined. And the least squares solutions and least square solutions with the minimal-norm have been obtained. In [16], using the hierarchical identification principle, authors consider the following more general coupled Sylvester-transpose matrix equation:

$$\sum_{\eta=1}^p (A_{i\eta} X_{\eta} B_{i\eta} + C_{i\eta} X_{\eta}^T D_{i\eta}) = F_i, \quad i \in I[1, N], \quad (2)$$

where  $A_{i\eta} \in R^{m_i \times l_{\eta}}, B_{i\eta} \in R^{n_{\eta} \times p_i}, C_{i\eta} \in R^{m_i \times n_{\eta}}, D_{i\eta} \in R^{l_{\eta} \times p_i}, F_i \in R^{m_i \times n_i}, i \in I[1, N]$ , and  $\eta \in I[1, p]$  are the given known matrices and  $X_{\eta} \in R^{l_{\eta} \times n_{\eta}}$  and  $\eta \in I[1, p]$  are the matrices to be determined. In addition, the generalized discrete Yakubovich-transpose matrix equation  $X - AX^T B = CY$

has important applications in dealing with complicated linear systems, such as large scale systems with interconnections, linear systems with certain partitioned structures or extended models, and second order or higher order linear systems [17, 18]. Song et al. [19] constructed the complete parametric solutions to the generalized discrete Yakubovich-transpose matrix equation. And one of the parametric solutions has a neat and elegant form in terms of the Krylov matrix, a block Hankel matrix, and an observability matrix. Other matrix equations such as descriptor linear systems and quaternion matrix equations have also been investigated, see [20–25] and the references therein. Wu et al. [26] have discussed the closed-form solutions to the generalized Sylvester-conjugate matrix equation and the proposed solution can provide all the degrees of freedom which are represented by a free parameter matrix. Wu et al. [27] proposed a general, complete parametric solution to the nonhomogeneous generalized Sylvester matrix equation  $AV + BW = EVF + R$ . One advantage of the proposed solution is that the matrices  $F$  and  $R$  are in an arbitrary form and can be set undetermined. This may give great convenience to many problems in descriptor linear systems, such as observer design and model reference control. Zhou et al. [28] investigated the problem of parameterizing all solutions to the polynomial Diophantine matrix equation and the generalized Sylvester matrix equation by using the so-called generalized Sylvester mapping, right coprime factorization, and Bezout identity associated with certain polynomial matrix. It is shown that the provided solutions can be parameterized as soon as two pairs of polynomial matrices satisfying the right coprime factorization and Bezout identity are obtained.

The rest of this paper is outlined as follows. In Section 2, the polynomial solutions to the Karm-Yakubovich-transpose matrix equation  $X - AX^T B = C$  are given. One of the polynomial solutions has a neat and elegant form in terms of symmetric operator matrix, controllability matrix and observability matrix. The polynomial solution to the Karm-Yakubovich-transpose is also proposed by the generalized Leverrier algorithm. Numerical example is given to show the efficiency of the proposed algorithm in Section 3. Some applications of the Karm-Yakubovich-transpose have been mentioned in Section 4 to end this paper.

Throughout this paper, we use  $R$  and  $C$  to denote the real number field and the complex number field.  $A^T$ ,  $\bar{A}$ ,  $A^H$ , and  $A^*$  refer to transpose, conjugate, conjugate transpose, and the adjoint matrix of  $A$ , respectively.  $\sigma(A)$  and  $\lambda(B)$  are the sets of characteristic eigenvalue of matrices  $A$  and  $B$ , respectively.  $I$  represents the identity matrix with appropriate dimensions. Moreover, for  $A \in C^{n \times n}$ ,  $B \in C^{n \times r}$ , and  $C \in C^{m \times n}$ , we have the following notations:

$$Q_c(A, B, n) = [B \ AB \ \cdots \ A^{n-1}B],$$

$$Q_o(A, C, k) = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{k-1} \end{bmatrix},$$

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

$$S_r(I, A) = \begin{bmatrix} I_r & \alpha_1 I_r & \alpha_2 I_r & \cdots & \alpha_{n-1} I_r \\ & I_r & \alpha_1 I_r & \cdots & \alpha_{n-2} I_r \\ & & \cdots & \cdots & \\ & & & I_r & \alpha_1 I_r \\ & & & & I_r \end{bmatrix}. \quad (3)$$

In this case,  $Q_c(A, B, n)$ ,  $Q_o(A, C, k)$ , and  $S_r(I, A)$  are named as the controllability matrix, the observability matrix, and a symmetric operator matrix, respectively.

## 2. Kalman-Yakubovich-Transpose Matrix Equation $X - AX^T B = C$

In this section, we consider the following matrix equation

$$X - AX^T B = C, \quad (4)$$

where  $A \in C^{n \times p}$ ,  $B \in C^{n \times p}$ ,  $C \in C^{n \times p}$ .

**Lemma 1.** *Given the matrices  $A \in C^{n \times p}$ ,  $B \in C^{n \times p}$ , and  $C \in C^{n \times p}$ , if  $X$  is a solution of (4), then for any integer  $n-1 \geq k \geq 0$ , the following conclusion holds:*

$$\begin{aligned} X(A^T B)^k - (AB^T)^{n-k} X(A^T B)^n \\ = \sum_{j=k}^{n-1} (AB^T)^{j-k} C(A^T B)^j \\ + \sum_{j=k}^{n-1} (AB^T)^{j-k} (AC^T B)(A^T B)^j. \end{aligned} \quad (5)$$

*Proof.* We prove this conclusion by mathematical induction. Postmultiplying both sides of (4) by  $(A^T B)^{n-1}$ , we have

$$X(A^T B)^{n-1} - AX^T B(A^T B)^{n-1} = C(A^T B)^{n-1}. \quad (6)$$

Using (4), we get

$$X^T - B^T X A^T = C^T. \quad (7)$$

Postmultiplying and premultiplying both sides of (7) by  $B(A^T B)^{n-1}$  and  $A$ , respectively, we have

$$AX^T B(A^T B)^{n-1} - AB^T X A^T B(A^T B)^{n-1} = AC^T B(A^T B)^{n-1}. \quad (8)$$

Combining (6) with (8), we obtain

$$\begin{aligned} X(A^T B)^{n-1} - AB^T X A^T B(A^T B)^{n-1} \\ = C(A^T B)^{n-1} + AC^T B(A^T B)^{n-1}. \end{aligned} \quad (9)$$

This implies that the relation (5) holds for  $k = n - 1$ .

Now assume that the relation (5) holds for  $k = N$ , ( $n-1 \geq N \geq 1$ ); that is,

$$\begin{aligned} X(A^T B)^N - (AB^T)^{n-N} X(A^T B)^n \\ = \sum_{j=N}^{n-1} (AB^T)^{j-N} (C + AC^T B) (A^T B)^j. \end{aligned} \quad (10)$$

Premultiplying both sides of (10) by  $AB^T$ , we have

$$\begin{aligned} AB^T X(A^T B)^N - (AB^T)^{n-N+1} X(A^T B)^n \\ = \sum_{j=N}^{n-1} (AB^T)^{j-N+1} C(A^T B)^j \\ + \sum_{j=N}^{n-1} (AB^T)^{j-N+1} (AC^T B) (A^T B)^j. \end{aligned} \quad (11)$$

Postmultiplying and premultiplying both sides of (7) by  $B$  and  $A$ , respectively, we have

$$AX^T B - AB^T XA^T B = AC^T B. \quad (12)$$

Combining (12) with (4), we derive

$$X - AB^T XA^T B = C + AC^T B. \quad (13)$$

Postmultiplying both sides of (13) by  $(A^T B)^{N-1}$  yields

$$X(A^T B)^{N-1} - AB^T X(A^T B)^N = (C + AC^T B) (A^T B)^{N-1}. \quad (14)$$

Combining (11) with (14), we get

$$\begin{aligned} X(A^T B)^{N-1} - (AB^T)^{n-N+1} X(A^T B)^n \\ = \sum_{j=N-1}^{n-1} (AB^T)^{j-(N-1)} C(A^T B)^j \\ + \sum_{j=N-1}^{n-1} (AB^T)^{j-(N-1)} (AC^T B) (A^T B)^j, \end{aligned} \quad (15)$$

which implies that the relation (5) holds for  $k = N - 1$ . So the conclusion is true.  $\square$

Define

$$\begin{aligned} \prod_k (A, C, B) = \sum_{j=k}^{n-1} (AB^T)^{j-k} C(A^T B)^j \\ + \sum_{j=k}^{n-1} (AB^T)^{j-k} (AC^T B) (A^T B)^j. \end{aligned} \quad (16)$$

Then the equality (5) can be rewritten as

$$X(A^T B)^k - (AB^T)^{n-k} X(A^T B)^n = \prod_k (A, C, B). \quad (17)$$

Let

$$f_{AB^T}(s) = \det(sI - AB^T) = \alpha_n + \alpha_{n-1}s + \cdots + \alpha_0 s^n. \quad (18)$$

Then

$$\tilde{f}_{AB^T}(s) = \det(I - sAB^T) = \alpha_0 + \alpha_1 s + \cdots + \alpha_n s^n. \quad (19)$$

Following (17), (18), and (19), we have

$$\begin{aligned} \sum_{k=0}^{n-1} \alpha_k \left[ X(A^T B)^k - (AB^T)^{n-k} X(A^T B)^n \right] \\ = \sum_{k=0}^{n-1} \left[ \alpha_k X(A^T B)^k - \alpha_k (AB^T)^{n-k} X(A^T B)^n \right] \\ + \alpha_n \left[ X(A^T B)^n - X(A^T B)^n \right] \\ = \sum_{k=0}^n \alpha_k X(A^T B)^k - \sum_{k=0}^n \alpha_k (AB^T)^{n-k} X(A^T B)^n \\ = X \tilde{f}_{AB^T}(A^T B) - f_{AB^T}(AB^T) X(A^T B)^n \\ = X \tilde{f}_{AB^T}(A^T B). \end{aligned} \quad (20)$$

On the other hand,

$$\begin{aligned} \sum_{k=0}^{n-1} \alpha_k \left[ X(A^T B)^k - (AB^T)^{n-k} X(A^T B)^n \right] \\ = \sum_{k=0}^{n-1} \alpha_k \prod_k (A, C, B). \end{aligned} \quad (21)$$

Define

$$\prod (A, C, B) = \sum_{k=0}^{n-1} \alpha_k \prod_k (A, C, B). \quad (22)$$

Hence, for each Kalman-Yakubovich-transpose equation of form (4), there is a uniquely determined polynomial  $\prod(A, C, B)$  of its coefficients. Therefore, the following equation is obtained:

$$X \tilde{f}_{AB^T}(A^T B) = \prod (A, C, B). \quad (23)$$

**Theorem 2.** For any  $\lambda, \mu \in \sigma(AB^T)$ , if  $\lambda\mu \neq 1$ , then (23) is equivalent to (4).

*Proof.* According to the argument above, it is shown that (4) implies (23). Now we prove that (23) implies (4) when  $\lambda\mu \neq 1$  for any  $\lambda, \mu \in \sigma(AB^T)$ . Suppose that  $X$  is a solution of (23); we have

$$\begin{aligned} X \tilde{f}_{AB^T}(A^T B) - AB^T XA^T B \tilde{f}_{AB^T}(A^T B) \\ = (X - AB^T XA^T B) \tilde{f}_{AB^T}(A^T B) \\ = \prod (A, C, B) - AB^T \prod (A, C, B) A^T B. \end{aligned} \quad (24)$$

In addition, we have

$$\begin{aligned}
& \prod (A, C, B) - AB^T \prod (A, C, B) A^T B \\
&= \sum_{k=0}^{n-1} \alpha_k \left[ \sum_{j=k}^{n-1} (AB^T)^{j-k} C(A^T B)^j \right. \\
&\quad \left. + \sum_{j=k}^{n-1} (AB^T)^{j-k} (AC^T B) (A^T B)^j \right] \\
&\quad - \sum_{k=0}^{n-1} \alpha_k \left[ \sum_{j=k}^{n-1} (AB^T)^{j-k+1} C(A^T B)^{j+1} \right. \\
&\quad \left. + \sum_{j=k}^{n-1} (AB^T)^{j-k+1} (AC^T B) (A^T B)^{j+1} \right] \\
&= \sum_{k=0}^{n-1} \alpha_k \left\{ \sum_{j=k}^{n-1} \left[ (AB^T)^{j-k} C(A^T B)^j \right. \right. \\
&\quad \left. \left. - (AB^T)^{j-k+1} C(A^T B)^{j+1} \right] \right. \\
&\quad \left. + \left[ (AB^T)^{j-k} (AC^T B) (A^T B)^j \right. \right. \\
&\quad \left. \left. - (AB^T)^{j-k+1} (AC^T B) (A^T B)^{j+1} \right] \right\} \\
&= \sum_{k=0}^{n-1} \alpha_k \left\{ \left[ C(A^T B)^k - (AB^T)^{n-k} C(A^T B)^n \right] \right. \\
&\quad \left. + \left[ (AC^T B) (A^T B)^k \right. \right. \\
&\quad \left. \left. - (AB^T)^{n-k} (AC^T B) (A^T B)^n \right] \right\} \\
&= C \tilde{f}_{AB^T} (A^T B) - f_{AB^T} (AB^T) C (A^T B)^n \\
&\quad + (AC^T B) \tilde{f}_{AB^T} (A^T B) \\
&\quad - f_{AB^T} (AB^T) (AC^T B) (A^T B)^n \\
&= C \tilde{f}_{AB^T} (A^T B) + (AC^T B) \tilde{f}_{AB^T} (A^T B). \tag{25}
\end{aligned}$$

Combining this with (24) gives

$$(X - AB^T X A^T B) \tilde{f}_{AB^T} (A^T B) = (C + AC^T B) \tilde{f}_{AB^T} (A^T B). \tag{26}$$

Because of  $\lambda\mu \neq 1$  for any  $\lambda, \mu \in \sigma(AB^T)$ , the matrix  $\tilde{f}_{AB^T}(A^T B)$  is nonsingular. Thus, it is obtained from (26) that (23) implies (4). With the two aspects above, the conclusion has been proved.  $\square$

The following theorem presents a result on the unique solution of the Kalman-Yakubovich-transpose matrix equation.

**Theorem 3.** *If  $\lambda\mu \neq 1$  for any  $\lambda, \mu \in \sigma(AB^T)$ , the solution to matrix equation (4) is*

$$X = \prod (A, C, B) [\tilde{f}_{AB^T} (A^T B)]^{-1}, \tag{27}$$

which is a polynomial of matrices  $A$ ,  $B$ , and  $C$ .

*Proof.* Assume that the characteristic polynomial of  $\tilde{f}_{AB^T}(A^T B)$  is  $f_{\tilde{f}_{AB^T}(A^T B)}(s) = \sum_{i=0}^p \gamma_i s^i$ . Since  $\tilde{f}_{AB^T}(A^T B)$  is nonsingular, it is shown that  $\gamma_0 \neq 0$ . It follows from Cayley-Hamilton Theorem that

$$f_{\tilde{f}_{AB^T}(A^T B)}(\tilde{f}_{AB^T}(A^T B)) = \sum_{i=1}^p \gamma_i [\tilde{f}_{AB^T}(A^T B)]^i + \gamma_0 I = 0, \tag{28}$$

which implies that

$$\tilde{f}_{AB^T}(A^T B) \left\{ \sum_{i=1}^p \gamma_i [\tilde{f}_{AB^T}(A^T B)]^{i-1} \right\} = -\gamma_0 I. \tag{29}$$

Therefore, it is derived that

$$[\tilde{f}_{AB^T}(A^T B)]^{-1} = -\frac{1}{\gamma_0} \sum_{i=1}^p \gamma_i [\tilde{f}_{AB^T}(A^T B)]^{i-1}, \tag{30}$$

which is a polynomial of  $\tilde{f}_{AB^T}(A^T B)$ . Because  $\tilde{f}_{AB^T}(A^T B)$  is a polynomial of  $A^T B$ , it is easy to know that  $[\tilde{f}_{AB^T}(A^T B)]^{-1}$  is a polynomial of  $A^T B$ . So we can see that  $\prod(A, C, B)[\tilde{f}_{AB^T}(A^T B)]^{-1}$  is a polynomial of matrices  $A$ ,  $B$  and  $C$ . Thus the conclusion can be proved.  $\square$

*Remark 4.* From Theorem 2 and its proof, it is shown that the solution  $X$  of matrix equation (4) is completely dependent on the coefficient matrices of transpose matrix equation (4). If  $\lambda\mu \neq 1$  for any  $\lambda, \mu \in \sigma(AB^T)$ , the matrix equation (4) has a unique solution. Otherwise, the above mentioned matrix equation (4) has no unique solution.

Next, we provide two equivalent forms of the solution to matrix equation (4). In order to obtain the unique solution of matrix equation (4), only the coefficients of characteristic polynomial of  $AB^T$  are required. Firstly, the so-called generalized Faddeev-Leverrier algorithm [29] is stated as the following iterative relations:

$$\begin{aligned}
U_k &= U_{k-1} (AB^T) + \alpha_k I, \quad U_0 = I, \quad k = 1, 2, \dots, n, \\
\alpha_k &= \frac{\text{trace}(U_{k-1} AB^T)}{k}, \quad \alpha_0 = 1, \quad k = 1, 2, \dots, n,
\end{aligned} \tag{31}$$

where  $\alpha_i$ ,  $i = 0, 1, 2, \dots, n-1$ , are the coefficients of the characteristic polynomial of the matrix  $AB^T$  and  $U_i$ ,  $i = 0, 1, \dots, n-1$ , are the coefficient matrices of the adjoint matrix  $(sI_n - AB^T)^*$ .

So we have the following theorems.

**Theorem 5.** Given the matrices  $A \in C^{n \times p}$ ,  $B \in C^{n \times p}$ , and  $C \in C^{n \times p}$ , let

$$\begin{aligned} \tilde{f}_{AB^T}(s) &= \det(I - sAB^T) = \alpha_n s^n + \dots + \alpha_1 s + \alpha_0, \alpha_0 = 1, \\ (I - sAB^T)^* &= U_{n-1} s^{n-1} + \dots + U_1 s + U_0. \end{aligned} \tag{32}$$

(1) If  $X$  is a solution to (4), then

$$X \tilde{f}_{AB^T}(A^T B) = \sum_{j=0}^{n-1} U_j C (A^T B)^j + \sum_{j=0}^{n-1} U_j (AC^T B) (A^T B)^j. \tag{33}$$

(2) If  $\lambda\mu \neq 1$  for any  $\lambda, \mu \in \sigma(AB^T)$ , then the matrix equation (4) has the unique solution; that is,

$$\begin{aligned} X &= \left[ \sum_{j=0}^{n-1} U_j C (A^T B)^j + \sum_{j=0}^{n-1} U_j (AC^T B) (A^T B)^j \right] \\ &\times [\tilde{f}_{AB^T}(A^T B)]^{-1}. \end{aligned} \tag{34}$$

*Proof.* Applying (31), we can obtain the following expression:

$$\begin{aligned} U_0 &= I, \\ U_1 &= \alpha_1 I + AB^T, \\ U_2 &= \alpha_2 I + \alpha_1 AB^T + (AB^T)^2, \\ &\vdots \\ U_{n-1} &= \alpha_{n-1} I + \alpha_{n-2} AB^T + \dots + (AB^T)^{n-1}. \end{aligned} \tag{35}$$

Meanwhile, the above formula can be stated as

$$U_j = \sum_{k=0}^j \alpha_k (AB^T)^{j-k}, \quad \alpha_0 = 1, \quad j = 1, 2, \dots, n-1. \tag{36}$$

Thus, it is easy to know

$$\begin{aligned} &\sum_{k=0}^{n-1} \sum_{j=k}^{n-1} \alpha_k (AB^T)^{j-k} C (A^T B)^j \\ &+ \sum_{k=0}^{n-1} \sum_{j=k}^{n-1} \alpha_k (AB^T)^{j-k} (AC^T B) (A^T B)^j \\ &= \sum_{j=0}^{n-1} \left[ \sum_{k=0}^j \alpha_k (AB^T)^{j-k} \right] C (A^T B)^j \\ &+ \sum_{j=0}^{n-1} \left[ \sum_{k=0}^j \alpha_k (AB^T)^{j-k} \right] (AC^T B) (A^T B)^j \\ &= \sum_{j=0}^{n-1} U_j C (A^T B)^j + \sum_{j=0}^{n-1} U_j (AC^T B) (A^T B)^j. \end{aligned} \tag{37}$$

Thus, we can easily obtain the conclusions.  $\square$

**Theorem 6.** Suppose the matrices  $A \in C^{n \times p}$ ,  $B \in C^{n \times p}$ , and  $C \in C^{n \times p}$ .

(1) Let  $X$  be a solution of (4); thus,

$$\begin{aligned} X \tilde{f}_{AB^T}(A^T B) &= Q_c(AB^T, C, n) S_p(I, AB^T) Q_o(A^T B, I, n) \\ &+ Q_c(AB^T, A, n) S_p(I, AB^T) Q_o(A^T B, C^T B, n). \end{aligned} \tag{38}$$

(2) Let  $\lambda\mu \neq 1$  for any  $\lambda, \mu \in \sigma(AB^T)$ ; thus, the matrix equation (4) has a unique solution

$$\begin{aligned} X &= [Q_c(AB^T, C, n) S_p(I, AB^T) Q_o(A^T B, I, n) \\ &+ Q_c(AB^T, A, n) S_p(I, AB^T) Q_o(A^T B, C^T B, n)] \\ &\times [\tilde{f}_{AB^T}(A^T B)]^{-1}. \end{aligned} \tag{39}$$

*Proof.* In view of the relation (35), it is obvious that

$$\begin{aligned} [U_0 C \quad U_1 C \quad \dots \quad U_{n-1} C] &= Q_c(AB^T, C, n) S_p(I, AB^T), \\ [U_0 A \quad U_1 A \quad \dots \quad U_{n-1} A] &= Q_c(AB^T, A, n) S_p(I, AB^T). \end{aligned} \tag{40}$$

Then it is easy to obtain that

$$\begin{aligned} &\sum_{j=0}^{n-1} U_j C (A^T B)^j \\ &= [U_0 C \quad U_1 C \quad \dots \quad U_{n-1} C] Q_o(A^T B, I, n) \\ &= Q_c(AB^T, C, n) S_p(I, AB^T) Q_o(A^T B, I, n), \end{aligned} \tag{41}$$

$$\begin{aligned} &\sum_{j=0}^{n-1} U_j (AC^T B) (A^T B)^j \\ &= [U_0 A \quad U_1 A \quad \dots \quad U_{n-1} A] Q_o(A^T B, C^T B, n) \\ &= Q_c(AB^T, A, n) S_p(I, AB^T) Q_o(A^T B, C^T B, n). \end{aligned}$$

Combining this with Theorem 5, we complete the proof.  $\square$

On the basis of the results above, we have the following corollary on the solution of the Stein-conjugate matrix equation  $X - AX^T A^T = C$ .

**Corollary 7.** Given the matrices  $A \in C^{n \times n}$  and  $C \in C^{n \times n}$ . If the matrix  $A^2$  is Schur stable, then the unique solution of matrix equation  $X - AX^T A^T = C$  is expressed as

$$X = \prod (A, C, A^T) [\tilde{f}_{A^2}((A^T)^2)]^{-1}. \tag{42}$$

### 3. Numerical Example

*Example 1.* Now we propose an example to compute the solution of matrix equation (4).

The parameters are stated as follows:

$$\begin{aligned} A &= \begin{bmatrix} 3 & 4 & 1 \\ 2 & 0 & 9 \end{bmatrix}, & B &= \begin{bmatrix} 1 & 10 & 8 \\ 2 & 11 & 6 \end{bmatrix}, \\ C &= \begin{bmatrix} -107 & -655 & -391 \\ -227 & -1370 & -815 \end{bmatrix}. \end{aligned} \quad (43)$$

It is easy to check that  $\alpha\beta \neq 1$ , for any  $\alpha, \beta \in \lambda(AB^T)$ . Therefore, the above matrix equation has a unique solution.

Via simple computation, we get

$$\begin{aligned} f_{(AB^T)}(s) &= s^2 - 109s - 1186, \\ \tilde{f}_{(AB^T)}(s) &= -1186s^2 - 109s + 1. \end{aligned} \quad (44)$$

Thus  $\alpha_0 = 1$ ,  $\alpha_1 = -109$ . In addition, we have

$$\begin{aligned} f_{(AB^T)}(A^T B) &= \begin{bmatrix} -1116788 & -7558116 & -4923452 \\ -944492 & -6285415 & -4045376 \\ -2074013 & -14369597 & -9513733 \end{bmatrix}, \\ [f_{(AB^T)}(A^T B)]^{-1} &= \begin{bmatrix} -0.8499 & -0.5902 & -0.1889 \\ -0.3035 & 0.2108 & 0.0675 \\ 0.2732 & -0.1897 & -0.0607 \end{bmatrix}. \end{aligned} \quad (45)$$

According to the following matrix equation expressions

$$\begin{aligned} \prod_0(A, C, B) &= C + AC^T B + (AB^T)(C + AC^T B)(A^T B), \\ \prod_1(A, C, B) &= (C + AC^T B)(A^T B), \end{aligned} \quad (46)$$

we have

$$\begin{aligned} \alpha_0 \prod_0(A, C, B) + \alpha_1 \prod_1(A, C, B) \\ = \begin{bmatrix} -9227811 & -63237737 & -41555403 \\ -31314571 & -214634302 & -141059447 \end{bmatrix}. \end{aligned} \quad (47)$$

Thus, it follows from Theorem 3 that the unique solution of equation (4) is

$$\begin{aligned} X &= \left[ \alpha_0 \prod_0(A, C, B) + \alpha_1 \prod_1(A, C, B) \right] [f_{(AB^T)}(A^T B)]^{-1} \\ &= \begin{bmatrix} 1.0000 & 2.0000 & 3.0000 \\ -0.0000 & 9.0000 & 11.0000 \end{bmatrix}. \end{aligned} \quad (48)$$

### 4. Applications

A number of control problems are related to the SCA (state covariance assignment) problem [30], for example, controllability/observability Gramian assignment [31], a certain class of  $H_\infty$  control problems [32]. From a mathematical viewpoint, we point out that both the continuous-time and discrete-time SCA problems can be reduced to solving a symmetric matrix equation. Let us expand the continuous-time Lyapunov equation:

$$A_K X + X A_K^T + B_K B_K^T = 0 \quad (49)$$

as

$$\begin{aligned} (AX + XA^T + B_1 B_1^T) + B_2 K (C_2 X + D_{21} B_1^T) \\ + (C_2 X + D_{21} B_1^T)^T K^T B_2^T + B_2 K (D_{21} D_{21}^T) K^T B_2^T = 0. \end{aligned} \quad (50)$$

We also expand the discrete-time Lyapunov equation:

$$A_K X A_K^T - X + B_K B_K^T = 0, \quad (51)$$

as

$$\begin{aligned} (AXA^T - X + B_1 B_1^T) + B_2 K (C_2 X A^T + D_{21} B_1^T) \\ + (C_2 X A^T + D_{21} B_1^T)^T K^T B_2^T \\ + B_2 K (C_2 X C_2^T + D_{21} D_{21}^T) K^T B_2^T = 0. \end{aligned} \quad (52)$$

It is easy to see that both the continuous-time and discrete-time SCA problems are essentially the same from the viewpoint of mathematics.

We also consider the relationship between the matrix equation and some important special cases as a control problem. For example, let matrices  $B \in R^{n \times m}$ ,  $Q \in R^{n \times n}$ ,  $R \in R^{p \times p}$ , and  $S \in R^{p \times n}$  be given, where  $Q = Q^T$ ,  $R = R^T \geq 0$ , and  $\text{rank } R = r$ . We focus on a symmetric matrix equation of quadratic type [31]:

$$Q + BKS + (BKS)^T + BKRK^T B^T = 0. \quad (53)$$

The solution of the symmetric matrix equation (53) is equivalent to the SCA problem with proper definitions of  $B$ ,  $Q$ ,  $R$ , and  $S$ . When  $R = 0$ , the special case of matrix equation (53) is reduced to the Karm-Yakubovich-transpose matrix equation.

### 5. Conclusions

The well-known Karm-Yakubovich-transpose matrix equation has many important applications in control system theory, such as fault detection, control with constrains systems, eigenstructure assignment, and observer design. In this paper we have proposed polynomial solutions to the Karm-Yakubovich-transpose matrix equation. The solutions are stated as a polynomial of parameters of the matrix equation. All the coefficient matrices are not restricted to be in any

canonical form. Meanwhile, an equivalent form of the solutions to the Karm-Yakubovich-transpose matrix equation has been expressed in terms of controllability matrix associated with  $A$ ,  $B$ , and  $C$  and observability matrix associated with  $A$  and  $B$ . Such a feature may bring greater convenience and advantages to some problems related to the Karm-Yakubovich-transpose matrix equation. From the discussion in our paper, one can observe that the solutions to the Karm-Yakubovich-transpose matrix equation are crucial as the theoretical basis of the development of many kinds of other matrix equations and are deserved further investigation in the future. In addition, as the theoretical generalization of the well-known Karm-Yakubovich-transpose matrix equation, it may be helpful for future control applications.

### Conflict of Interests

The authors declare that there is no conflict of interests regarding the publication of this paper.

### Acknowledgments

The authors are very grateful to the anonymous reviewers and the editor for their helpful comments and suggestions which have helped us in improving the quality of this paper. This project is granted financial support from NNSF 61174141 of China, Postdoctoral Science Foundation of China (no. 2013M541900), Research Awards Young and Middle-Aged Scientists of Shandong Province (BS2011SF009 and BS2011DX019), and Excellent Youth Foundation of Shandong's Natural Scientific Committee (JQ201219).

### References

- [1] P. M. Frank, "Fault diagnosis in dynamic systems using analytical and knowledge-based redundancy—a survey and some new results," *Automatica*, vol. 26, no. 3, pp. 459–474, 1990.
- [2] G. R. Duan, "The solution to the matrix equation  $AV + BW = EVJ + R$ ," *Applied Mathematics Letters*, vol. 17, no. 10, pp. 1197–1202, 2004.
- [3] L. R. Fletcher, J. Kuatslcy, and N. K. Nichols, "Eigenstructure assignment in descriptor systems," *IEEE Transactions on Automatic Control*, vol. 31, no. 12, pp. 1138–1141, 1986.
- [4] L. Dai, *Singular Control Systems*, vol. 118 of *Lecture Notes in Control and Information Sciences*, Springer, Berlin, Germany, 1989.
- [5] H. W. Braden, "The equations  $A^T X \pm X^T A = B$ ," *SIAM Journal on Matrix Analysis and Applications*, vol. 20, no. 2, pp. 295–302, 1999.
- [6] A. P. Liao, Y. Lei, and X. Y. Hu, "Least-squares solution with the minimum-norm for the matrix equation  $A^T X B + B^T X^T A = D$  and its applications," *Acta Mathematicae Applicatae Sinica*, vol. 23, no. 2, pp. 269–280, 2007.
- [7] F. X. Piao, Q. Zhang, and Z. Wang, "The solution to matrix equation  $AX + X^T C = B$ ," *Journal of the Franklin Institute*, vol. 344, no. 8, pp. 1056–1062, 2007.
- [8] C. Q. Song and G. L. Chen, "On solutions of matrix equation  $XF - AX = C$  and  $XF - A\bar{X} = C$  over quaternion field," *Journal of Applied Mathematics and Computing*, vol. 37, no. 1-2, pp. 57–68, 2011.
- [9] C. Q. Song, G. L. Chen, and Q. B. Liu, "Explicit solutions to the quaternion matrix equations  $X - AXF = C$  and  $X - A\bar{X}F = C$ ," *International Journal of Computer Mathematics*, vol. 89, no. 7, pp. 890–900, 2012.
- [10] Q. W. Wang, H. S. Zhang, and S. W. Yu, "On solutions to the quaternion matrix equation  $AXB + CYD = E$ ," *Electronic Journal of Linear Algebra*, vol. 17, pp. 343–358, 2008.
- [11] Q. W. Wang, H. S. Zhang, and G. J. Song, "A new solvable condition for a pair of generalized Sylvester equations," *Electronic Journal of Linear Algebra*, vol. 18, pp. 289–301, 2009.
- [12] M. H. Wang, X. H. Cheng, and M. S. Wei, "Iterative algorithms for solving the matrix equation  $AXB + CX^T D = E$ ," *Applied Mathematics and Computation*, vol. 187, no. 2, pp. 622–629, 2007.
- [13] L. Xie, Y. J. Liu, and H. Z. Yang, "Gradient based and least squares based iterative algorithms for matrix equations  $AXB + CX^T D = F$ ," *Applied Mathematics and Computation*, vol. 217, no. 5, pp. 2191–2199, 2010.
- [14] Z. Y. Li, Y. Wang, B. Zhou, and G. R. Duan, "Least squares solution with the minimum-norm to general matrix equations via iteration," *Applied Mathematics and Computation*, vol. 215, no. 10, pp. 3547–3562, 2010.
- [15] L. Xie, J. Ding, and F. Ding, "Gradient based iterative solutions for general linear matrix equations," *Computers & Mathematics with Applications*, vol. 58, no. 7, pp. 1441–1448, 2009.
- [16] C. Q. Song, G. L. Chen, and L. L. Zhao, "Iterative solutions to coupled Sylvester-transpose matrix equations," *Applied Mathematical Modelling*, vol. 35, no. 10, pp. 4675–4683, 2011.
- [17] G. R. Duan and G. P. Liu, "Complete parametric approach for eigenstructure assignment in a class of second-order linear systems," *Automatica*, vol. 38, no. 4, pp. 725–729, 2002.
- [18] G. R. Duan, "Solutions of the equation  $AV + BW = VF$  and their application to eigenstructure assignment in linear systems," *IEEE Transactions on Automatic Control*, vol. 38, no. 2, pp. 276–280, 1993.
- [19] C. Q. Song, J. Feng, X. D. Wang, and J. L. Zhao, "Parametric solutions to the generalized discrete Yakubovich-transpose matrix equation," *Asian Journal of Control*, vol. 16, no. 4, pp. 1–8, 2014.
- [20] X. P. Sheng and G. L. Chen, "An iterative method for the symmetric and skew symmetric solutions of a linear matrix equation  $AXB + CYD = E$ ," *Journal of Computational and Applied Mathematics*, vol. 233, no. 11, pp. 3030–3040, 2010.
- [21] F. Ding, "Hierarchical multi-innovation stochastic gradient algorithm for Hammerstein nonlinear system modeling," *Applied Mathematical Modelling*, vol. 37, no. 4, pp. 1694–1704, 2013.
- [22] F. Ding and T. Chen, "A gradient based adaptive control algorithm for dual-rate systems," *Asian Journal of Control*, vol. 8, no. 4, pp. 314–323, 2006.
- [23] Y. Lin and Q. W. Wang, "Completing a  $2 \times 2$  block matrix of real quaternions with a partial specified inverse," *Journal of Applied Mathematics*, vol. 2013, Article ID 271978, 5 pages, 2013.
- [24] C. Z. Dong, Q. W. Wang, and Y. P. Zhang, "On the Hermitian  $R$ -conjugate solution of a system of matrix equations," *Journal of Applied Mathematics*, vol. 2012, Article ID 398085, 14 pages, 2012.
- [25] X. F. Duan and Q. W. Wang, "Perturbation analysis for the matrix equation  $X - \sum_{i=1}^m A_i^* X A_i + \sum_{j=1}^n B_j^* X B_j = I$ ," *Journal of Applied Mathematics*, vol. 2012, Article ID 784620, 13 pages, 2012.

- [26] A. G. Wu, E. Z. Zhang, and F. C. Liu, "On closed-form solutions to the generalized Sylvester-conjugate matrix equation," *Applied Mathematics and Computation*, vol. 218, no. 19, pp. 9730–9741, 2012.
- [27] A. G. Wu, Y. Sun, and G. Feng, "Closed-form solution to the non-homogeneous generalised Sylvester matrix equation," *IET Control Theory & Applications*, vol. 4, no. 10, pp. 1914–1921, 2010.
- [28] B. Zhou, Z. B. Yan, and G. R. Duan, "Unified parametrization for the solutions to the polynomial diophantine matrix equation and the generalized Sylvester matrix equation," *International Journal of Control, Automation, and Systems*, vol. 8, no. 1, pp. 29–35, 2010.
- [29] B. Hanzon and R. L. M. Peeters, "A Faddeev sequence method for solving Lyapunov and Sylvester equations," *Linear Algebra and Its Applications*, vol. 241–243, pp. 401–430, 1996.
- [30] H. Fujioka and S. Hara, "State covariance assignment problem with measurement noise: a unified approach based on a symmetric matrix equation," *Linear Algebra and Its Applications*, vol. 203–204, pp. 579–605, 1994.
- [31] K. Yasuda and R. E. Skelton, "Assigning controllability and observability Gramians in feedback control," *Journal of Guidance, Control, and Dynamics*, vol. 14, no. 5, pp. 878–885, 1991.
- [32] R. E. Skelton and T. Iwasaki, "Liapunov and covariance controllers," *International Journal of Control*, vol. 57, no. 3, pp. 519–536, 1993.