

RANKS AND INDEPENDENCE OF SOLUTIONS OF THE MATRIX EQUATION

$$AXB + CYD = M$$

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ABSTRACT. Suppose $AXB + CYD = M$ is a consistent matrix equation. In this paper, we give some formulas for the maximal and minimal ranks of two solutions X and Y to the equation. In addition, we investigate the independence of solutions X and Y to this equation.

1. INTRODUCTION

Throughout this paper, the notation A^T , A^* , $r(A)$ and $\mathcal{R}(A)$ stand for the transpose, conjugate transpose, rank and range (column space) of a matrix A over the field \mathbb{C} of complex numbers, respectively. A matrix X is called a generalized inverse of A , denoted by A^- , if it satisfies $AXA = A$. In addition, E_A and F_A stand for the two oblique projectors $E_A = I - AA^-$ and $F_A = I - A^-A$ induced by A and A^- .

Linear matrix equations have been the objects of many studies in matrix theory and its applications. The primary work in the investigation of a matrix equation is to give its solvability conditions and general solutions. In additions to these two problems, many other topics can be investigated for a matrix equation. For example, the uniqueness of solution, minimal norm solutions, least-squares solutions, Hermitian solutions, and skew-Hermitian solutions to the equation. For some simplest matrix equations, it is easy to characterize the solvability and to give general solutions by generalized inverses. For instance, the matrix equation $AXB = C$, where A , B and C

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are $m \times p$, $q \times n$ and $m \times n$ matrices, respectively, is consistent if and only if $AA^{-1}CB^{-1}B = C$. In this case, the general solution of $AXB = C$ can be written as $X = A^{-1}CB^{-1} + (I_p - A^{-1}A)U + V(I_q - BB^{-1})$, where U and V are arbitrary. Many problems can be considered for solutions of $AXB = C$, one of which is to determine the maximal and minimal possible ranks of solutions. The present author has shown in [10] that

$$\begin{aligned} \max_{AXB=C} r(X) &= \min \{ p, q, p + q + r(C) - r(A) - r(B) \}, \\ \min_{AXB=C} r(X) &= r(C). \end{aligned}$$

Write complex solution of $AXB = C$ as $X = X_0 + X_1i$, where X_0 and X_1 are both real. The present author also gives in [10] the maximal and minimal ranks of X_0 and X_1 . In addition to $AXB = C$, another well-known matrix equation is

$$(1.1) \quad AXB + CYD = M,$$

where $A \in \mathbb{C}^{m \times p}$, $B \in \mathbb{C}^{q \times n}$, $C \in \mathbb{C}^{m \times s}$, $D \in \mathbb{C}^{t \times n}$, $M \in \mathbb{C}^{m \times n}$. Equation (1.1) and its applications have been investigated extensively, see, e.g., [1, 3, 4, 6, 7, 13, 15]. A regression model related to (1.1) is

$$M = AXB + CYD + \varepsilon,$$

where both X and Y are unknown parameter matrices and ε is a random error matrix. This model is also called the nested growth curve model in the literature, see, e.g., [5, 14].

The rank of a matrix A , a key concept in linear algebra, is the dimension of the vector space generated by the columns or rows of A , that is, the maximum number of linearly independent columns or rows of A . Equivalently, the rank of a matrix A is the largest order of square submatrix of A which determinant is nonzero. If a matrix has some variant entries, the rank of the matrix is also variant with respect to the entries.

A general method for solving linear matrix equations is the vec operation of a matrix $Z = (z_{ij}) \in \mathbb{C}^{m \times n}$ defined by

$$\text{vec } Z = [z_{11}, \dots, z_{m1}, z_{12}, \dots, z_{m2}, \dots, z_{1n}, \dots, z_{mn}]^T.$$

Applying the well-known formula $\text{vec}(AXB) = (B^T \otimes A)\text{vec}X$, where $B^T \otimes A$ is the Kronecker product of B^T and A , to (1.1) gives

$$(1.2) \quad [B^T \otimes A, D^T \otimes C] \begin{bmatrix} \text{vec}X \\ \text{vec}Y \end{bmatrix} = \text{vec}M,$$

where $[B^T \otimes A, D^T \otimes C]$ is a row block matrix. Hence (1.2) is solvable if and only if $[B^T \otimes A, D^T \otimes C][B^T \otimes A, D^T \otimes C]^- \text{vec}M = \text{vec}M$. In such a case, the general solution of (1.2) can be written as

$$(1.3) \quad \begin{bmatrix} \text{vec}X \\ \text{vec}Y \end{bmatrix} = [B^T \otimes A, D^T \otimes C]^- \text{vec}M \\ + (I - [B^T \otimes A, D^T \otimes C]^- [B^T \otimes A, D^T \otimes C])V,$$

where V is an arbitrary column vector. Result (1.3) implies that the general solutions X and Y of (1.1) are in fact two linear matrix expressions involving variant entries.

Since the two matrices X and Y satisfying (1.1) are not necessarily unique, it is of interest to find the maximal and minimal possible ranks of X , Y , AXB and CYD in (1.1). Another problem on a pair solutions X and Y to (1.1) is concerned with their independence, where the independence means that for any two pairs of solutions X_1, Y_1 and X_2, Y_2 of (1.1), the two new pairs X_1, Y_2 and X_2, Y_1 are also solutions to (1.1). This problem can also be solved through some rank formulas associated with (1.1).

Some useful rank formulas for partitioned matrices are given in the following lemma.

Lemma 1.1 ([2]). *Let $A \in \mathbb{C}^{m \times n}$, $B \in \mathbb{C}^{m \times k}$ and $C \in \mathbb{C}^{l \times n}$. Then:*

- (a) $r[A, B] = r(A) + r(E_A B) = r(B) + r(E_B A)$.
- (b) $r \begin{bmatrix} A \\ C \end{bmatrix} = r(A) + r(CF_A) = r(C) + r(AF_C)$.
- (c) $r \begin{bmatrix} A & B \\ C & 0 \end{bmatrix} = r(B) + r(C) + r(E_B AF_C)$.

The formulas in Lemma 1.1 can be used to simplify various matrix expressions involving generalized inverses of matrices. For example,

$$(1.4) \quad r \begin{bmatrix} E_{A_1} B_1 \\ E_{A_2} B_2 \end{bmatrix} = r \begin{bmatrix} A_1 & 0 & B_1 \\ 0 & A_2 & B_2 \end{bmatrix} - r(A_1) - r(A_2),$$

$$(1.5) \quad r[D_1 F_{C_1}, D_2 F_{C_2}] = r \begin{bmatrix} D_1 & D_2 \\ C_1 & 0 \\ 0 & C_2 \end{bmatrix} - r(C_1) - r(C_2),$$

$$(1.6) \quad r \begin{bmatrix} A & B F_{B_1} \\ E_{C_1} C & 0 \end{bmatrix} = r \begin{bmatrix} A & B & 0 \\ C & 0 & C_1 \\ 0 & B_1 & 0 \end{bmatrix} - r(B_1) - r(C_1).$$

Lemma 1.2. *Let $A \in \mathbb{C}^{m \times n}$, $B \in \mathbb{C}^{m \times k}$, $C \in \mathbb{C}^{l \times n}$, $B_1 \in \mathbb{C}^{m \times p}$ and $C_1 \in \mathbb{C}^{q \times n}$ be given, $X \in \mathbb{C}^{k \times l}$, $Y \in \mathbb{C}^{k \times n}$, $Z \in \mathbb{C}^{m \times l}$ and $U \in \mathbb{C}^{p \times q}$ be variant matrices. Then*

$$(1.7) \quad \max_X r(A - BXC) = \min \left\{ r[A, B], \quad r \begin{bmatrix} A \\ C \end{bmatrix} \right\},$$

$$(1.8) \quad \min_X r(A - BXC) = r[A, B] + r \begin{bmatrix} A \\ C \end{bmatrix} - r \begin{bmatrix} A & B \\ C & 0 \end{bmatrix},$$

$$(1.9) \quad \max_{Y, Z} r(A - BY - ZC) = \min \left\{ m, \quad n, \quad r \begin{bmatrix} A & B \\ C & 0 \end{bmatrix} \right\},$$

$$(1.10) \quad \min_{Y, Z} r(A - BY - ZC) = r \begin{bmatrix} A & B \\ C & 0 \end{bmatrix} - r(B) - r(C),$$

$$(1.11) \quad \begin{aligned} & \max_{Y, Z, U} r(A - BY - ZC - B_1UC_1) \\ & = \min \left\{ m, \quad n, \quad r \begin{bmatrix} A & B & B_1 \\ C & 0 & 0 \end{bmatrix}, \quad r \begin{bmatrix} A & B \\ C & 0 \\ C_1 & 0 \end{bmatrix} \right\}, \end{aligned}$$

$$(1.12) \quad \begin{aligned} & \min_{Y, Z, U} r(A - BY - ZC - B_1UC_1) = r \begin{bmatrix} A & B & B_1 \\ C & 0 & 0 \end{bmatrix} \\ & + r \begin{bmatrix} A & B \\ C & 0 \\ C_1 & 0 \end{bmatrix} - r \begin{bmatrix} A & B & B_1 \\ C & 0 & 0 \\ C_1 & 0 & 0 \end{bmatrix} - r(B) - r(C). \end{aligned}$$

Results (1.7) and (1.8) are shown in [12]; (1.9) and (1.10) are shown in [8, 9]. The general expressions of X and Y satisfying (1.7)–(1.10) are given in [8, 9, 12]. Combining (1.7) and (1.9), (1.8) and (1.10) yields (1.11) and (1.12), respectively.

2. RANKS OF SOLUTIONS TO $AXB + CYD = M$

Concerning the solvability conditions and general solutions of (1.1), the following results have been shown.

Lemma 2.1.

(a) [3] *There are X and Y that satisfy (1.1) if and only if*

$$(2.1) \quad r[A, C, M] = r[A, C], \quad r \begin{bmatrix} B \\ D \\ M \end{bmatrix} = r \begin{bmatrix} B \\ D \end{bmatrix},$$

$$(2.2) \quad r \begin{bmatrix} M & A \\ D & 0 \end{bmatrix} = r(A) + r(D), \quad r \begin{bmatrix} M & C \\ B & 0 \end{bmatrix} = r(B) + r(C),$$

or equivalently,

$$[A, C][A, C]^{-}M = M, \quad M \begin{bmatrix} B \\ D \end{bmatrix}^{-} \begin{bmatrix} B \\ D \end{bmatrix} = M,$$

$$(I_m - AA^{-})M(I_n - D^{-}D) = 0, \quad (I_m - CC^{-})M(I_n - B^{-}B) = 0.$$

(b) [6, 7] Under (2.1) and (2.2), the general solutions of X and Y to (1.1) can be decomposed as

$$X = X_0 + X_1X_2 + X_3, \quad Y = Y_0 + Y_1Y_2 + Y_3,$$

where X_0 and Y_0 are a pair of special solutions of (1.1), X_1, X_2, X_3 and Y_1, Y_2, Y_3 are the general solutions of the following four homogeneous matrix equations

$$AX_1 = -CY_1, \quad X_2B = Y_2D, \quad AX_3B = 0, \quad CY_3D = 0,$$

or explicitly,

$$(2.3) \quad X = X_0 + S_1F_GUE_HT_1 + F_AV_1 + V_2E_B,$$

$$(2.4) \quad Y = Y_0 + S_2F_GUE_HT_2 + F_CW_1 + W_2E_D,$$

where

$$S_1 = [I_p, 0], \quad S_2 = [0, I_s], \quad T_1 = \begin{bmatrix} I_q \\ 0 \end{bmatrix}, \quad T_2 = \begin{bmatrix} 0 \\ I_t \end{bmatrix}, \quad G = [A, C], \quad H = \begin{bmatrix} B \\ -D \end{bmatrix};$$

the matrices U, V_1, V_2, W_1 and W_2 are arbitrary.

For convenience, we adopt the following notation

$$(2.5) \quad J_1 = \{ X \in \mathbb{C}^{p \times q} \mid AXB + CYD = M \},$$

$$(2.6) \quad J_2 = \{ Y \in \mathbb{C}^{s \times t} \mid AXB + CYD = M \}.$$

Results (2.3) and (2.4) imply that the general solutions of (1.1) are in fact two linear matrix expressions, each of them involves three independent variant matrices. Applying Lemma 1.2 to (2.3) and (2.4) gives the following result.

Theorem 2.2. *Suppose that the matrix equation (1.1) is solvable, and let J_1 and J_2 be defined in (2.5) and (2.6). Then:*

(a) *The maximal and minimal ranks of solution X of (1.1) are given by*

$$\begin{aligned} \max_{X \in J_1} r(X) &= \min \left\{ p, \quad q, \quad p + q + r[M, C] - r[A, C] - r(B), \right. \\ &\quad \left. p + q + r \begin{bmatrix} M \\ D \end{bmatrix} - r \begin{bmatrix} B \\ D \end{bmatrix} - r(A) \right\}, \\ \min_{X \in J_1} r(X) &= r[M, C] + r \begin{bmatrix} M \\ D \end{bmatrix} - r \begin{bmatrix} M & C \\ D & 0 \end{bmatrix}. \end{aligned}$$

(b) *The maximal and minimal ranks of solution Y of (1.1) are given by*

$$\begin{aligned} \max_{Y \in J_2} r(Y) &= \min \left\{ s, \quad t, \quad s + t + r[M, A] - r[C, A] - r(D), \right. \\ &\quad \left. s + t + r \begin{bmatrix} M \\ B \end{bmatrix} - r \begin{bmatrix} D \\ B \end{bmatrix} - r(C) \right\}, \\ \min_{Y \in J_2} r(Y) &= r[M, A] + r \begin{bmatrix} M \\ B \end{bmatrix} - r \begin{bmatrix} M & A \\ B & 0 \end{bmatrix}. \end{aligned}$$

Proof. Applying (1.7) and (1.8) to (2.3) yields

$$\max_{X \in J_1} r(X) = \max_{U, V_1, V_2} r(X_0 + S_1 F_G U E_H T_1 + F_A V_1 + V_2 E_B) = \min \left\{ p, q, r \begin{bmatrix} X_0 & F_A & S_1 F_G \\ E_B & 0 & 0 \end{bmatrix}, r \begin{bmatrix} X_0 & F_A \\ E_B & 0 \\ E_H T_1 & 0 \end{bmatrix} \right\},$$

$$\begin{aligned} \min_{X \in J_1} r(X) &= \min_{U, V_1, V_2} r(X_0 + S_1 F_G U E_H T_1 + F_A V_1 + V_2 E_B) \\ &= r \begin{bmatrix} X_0 & F_A & S_1 F_G \\ E_B & 0 & 0 \end{bmatrix} + r \begin{bmatrix} X_0 & F_A \\ E_B & 0 \\ E_H T_1 & 0 \end{bmatrix} - r \begin{bmatrix} X_0 & F_A & S_1 F_G \\ E_B & 0 & 0 \\ E_H T_1 & 0 & 0 \end{bmatrix} - r(F_A) - r(E_B), \end{aligned}$$

where $r(F_A) = p - r(A)$ and $r(E_B) = q - r(B)$. As shown in (1.4), (1.5) and (1.6), the ranks of the block matrices in these two formulas can be simplified further by Lemma 1.1, as well as the equality $AX_0B + CY_0D = M$ and elementary block matrix operations

$$\begin{aligned} &r \begin{bmatrix} X_0 & F_A & S_1 F_G \\ E_B & 0 & 0 \end{bmatrix} \\ &= r \begin{bmatrix} X_0 & I_p & S_1 & 0 \\ I_q & 0 & 0 & B \\ 0 & A & 0 & 0 \\ 0 & 0 & G & 0 \end{bmatrix} - r(A) - r(B) - r(G) = r \begin{bmatrix} 0 & I_p & 0 & 0 \\ I_q & 0 & 0 & 0 \\ 0 & 0 & -AS_1 & AX_0B \\ 0 & 0 & G & 0 \end{bmatrix} - r(A) - r(B) - r(G) \\ &= r \begin{bmatrix} -A & 0 & AX_0B \\ A & C & 0 \end{bmatrix} + p + q - r(A) - r(B) - r(G) = r[C, AX_0B] + p + q - r(B) - r(G) \\ &= r[C, M] + p + q - r(B) - r(G), \end{aligned}$$

$$\begin{aligned}
& r \begin{bmatrix} X_0 & F_A \\ E_B & 0 \\ E_H T_1 & 0 \end{bmatrix} \\
&= r \begin{bmatrix} X_0 & I_p & 0 & 0 \\ I_q & 0 & B & 0 \\ T_1 & 0 & 0 & H \\ 0 & A & 0 & 0 \end{bmatrix} - r(A) - r(B) - r(H) = r \begin{bmatrix} 0 & I_p & 0 & 0 \\ I_q & 0 & 0 & 0 \\ 0 & 0 & -T_1 B & H \\ 0 & 0 & A X_0 B & 0 \end{bmatrix} - r(A) - r(B) - r(H) \\
&= r \begin{bmatrix} B & B \\ 0 & D \\ A X_0 B & 0 \end{bmatrix} + p + q - r(A) - r(B) - r(H) = r \begin{bmatrix} D \\ A X_0 B \end{bmatrix} + p + q - r(A) - r(H) \\
&= r \begin{bmatrix} D \\ M \end{bmatrix} + p + q - r(A) - r(H),
\end{aligned}$$

$$\begin{aligned}
& r \begin{bmatrix} X_0 & F_A & S_1 F_G \\ E_B & 0 & 0 \\ E_H T_1 & 0 & 0 \end{bmatrix} \\
&= r \begin{bmatrix} X_0 & I_p & S_1 & 0 & 0 \\ I_q & 0 & 0 & B & 0 \\ T_1 & 0 & 0 & 0 & H \\ 0 & A & 0 & 0 & 0 \\ 0 & 0 & G & 0 & 0 \end{bmatrix} - r(A) - r(B) - r(G) - r(H) \\
&= r \begin{bmatrix} 0 & I_p & 0 & 0 & 0 \\ I_q & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & -T_1 B & H \\ 0 & 0 & -A S_1 & A X_0 B & 0 \\ 0 & 0 & G & 0 & 0 \end{bmatrix} - r(A) - r(B) - r(G) - r(H)
\end{aligned}$$

$$\begin{aligned}
&= r \begin{bmatrix} 0 & 0 & -B & B \\ 0 & 0 & 0 & -D \\ -A & 0 & AX_0B & 0 \\ A & C & 0 & 0 \end{bmatrix} + p + q - r(A) - r(B) - r(G) - r(H) \\
&= r \begin{bmatrix} 0 & 0 & -B & 0 \\ 0 & 0 & 0 & D \\ -A & 0 & 0 & 0 \\ 0 & C & 0 & M \end{bmatrix} + p + q - r(A) - r(B) - r(G) - r(H) \\
&= r \begin{bmatrix} M & C \\ D & 0 \end{bmatrix} + p + q - r(G) - r(H).
\end{aligned}$$

Thus, we have (a). Similarly, we can show (b). □

Furthermore, we can give the formulas for the maximal and minimal ranks of AXB and CYD in (1.1) when it is solvable.

Theorem 2.3. *Suppose that there are X and Y that satisfy (1.1), and let J_1 and J_2 be defined in (2.5) and (2.6). Then*

$$(2.7) \quad \max_{X \in J_1} r(AXB) = \min \left\{ r[M, C] - r[A, C] + r(A), \quad r \begin{bmatrix} M \\ D \end{bmatrix} - r \begin{bmatrix} B \\ D \end{bmatrix} + r(B) \right\},$$

$$(2.8) \quad \min_{X \in J_1} r(AXB) = r[M, C] + r \begin{bmatrix} M \\ D \end{bmatrix} - r \begin{bmatrix} M & C \\ D & 0 \end{bmatrix},$$

$$(2.9) \quad \max_{Y \in J_2} r(CYD) = \min \left\{ r[M, A] - r[C, A] + r(C), r \begin{bmatrix} M \\ B \end{bmatrix} - r \begin{bmatrix} D \\ B \end{bmatrix} + r(D) \right\},$$

$$(2.10) \quad \min_{Y \in J_2} r(CYD) = r[M, A] + r \begin{bmatrix} M \\ B \end{bmatrix} - r \begin{bmatrix} M & A \\ B & 0 \end{bmatrix}.$$

Proof. Applying (1.7) and (1.8) to $AXB = AX_0B + AS_1F_GUE_HT_1B$ yields

$$\begin{aligned} \max_{X \in J_1} r(AXB) &= \max_U r(AX_0B + AS_1F_GUE_HT_1B) \\ &= \min \left\{ r[AX_0B, AS_1F_G], r \begin{bmatrix} AX_0B \\ E_HT_1B \end{bmatrix} \right\}, \\ \min_{X \in J_1} r(AXB) &= \min_U r(AX_0B + AS_1F_GUE_HT_1B) \\ &= r[AX_0B, AS_1F_G] + r \begin{bmatrix} AX_0B \\ E_HT_1B \end{bmatrix} - r \begin{bmatrix} AX_0B & AS_1F_G \\ E_HT_1B & 0 \end{bmatrix}. \end{aligned}$$

Also find by Lemma 1.1, $AX_0B + CY_0D = M$ and elementary block matrix operations that

$$\begin{aligned} r[AX_0B, AS_1F_G] &= r \begin{bmatrix} AX_0B & AS_1 \\ 0 & G \end{bmatrix} - r(G) = r \begin{bmatrix} AX_0B & A & 0 \\ 0 & A & C \end{bmatrix} - r(G) \\ &= r[AX_0B, C] + r(A) - r(G) = r[M, C] + r(A) - r(G), \end{aligned}$$

$$\begin{aligned} r \begin{bmatrix} AX_0B \\ P_HT_1B \end{bmatrix} &= r \begin{bmatrix} AX_0B & 0 \\ T_1B & H \end{bmatrix} - r(H) = r \begin{bmatrix} AX_0B & 0 \\ B & B \\ 0 & -D \end{bmatrix} - r(H) \\ &= r \begin{bmatrix} AX_0B \\ D \end{bmatrix} + r(B) - r(H) = r \begin{bmatrix} M \\ D \end{bmatrix} + r(B) - r(H), \end{aligned}$$

$$\begin{aligned}
r \begin{bmatrix} AX_0B & AS_1F_G \\ E_H T_1 B & 0 \end{bmatrix} &= r \begin{bmatrix} AX_0B & AS_1 & 0 \\ T_1 B & 0 & H \\ 0 & G & 0 \end{bmatrix} - r(G) - r(H) = r \begin{bmatrix} AX_0B & A & 0 & 0 \\ B & 0 & 0 & B \\ 0 & 0 & 0 & -D \\ 0 & A & C & 0 \end{bmatrix} - r(G) - r(H) \\
&= r \begin{bmatrix} 0 & A & 0 & 0 \\ B & 0 & 0 & 0 \\ 0 & 0 & 0 & D \\ 0 & 0 & C & AX_0B \end{bmatrix} - r(G) - r(H) = r \begin{bmatrix} M & C \\ D & 0 \end{bmatrix} + r(A) + r(B) - r(G) - r(H).
\end{aligned}$$

Therefore, we have (2.7) and (2.8). In the same manner, one can show (2.9) and (2.10). \square

3. INDEPENDENCE OF SOLUTIONS X AND Y TO $AXB + CYD = M$

The independence of the two matrices X_1 and X_2 that satisfy the matrix equation $A_1X_1 + A_2X_2 = B$ is investigated in the author's recent paper [11]. In this section, we consider the independence of X and Y that satisfy (1.1).

Consider J_1 and J_2 in (2.5) and (2.6) as two independent matrix sets. If for any given $X \in J_1$ and $Y \in J_2$, the pair satisfy (1.1), X and Y for (1.1) are said to be independent. The independence of solutions X and Y for (1.1) can also be examined through the rank formulas in Lemma 1.2.

Theorem 3.1. *Suppose that the matrix equation (1.1) is solvable. Moreover, let J_1 and J_2 in (2.5) and (2.6) as two independent matrix sets. Then*

$$(3.1) \quad \max_{X \in J_1, Y \in J_2} r(M - AXB - CYD) = \min \left\{ r(A) + r(C) - r[A, C], \quad r(B) + r(D) - r \begin{bmatrix} B \\ D \end{bmatrix} \right\}.$$

In particular,

(a) *Solutions X and Y of (1.1) are independent if and only if*

$$(3.2) \quad \mathcal{R}(A) \cap \mathcal{R}(C) = \{0\} \quad \text{or} \quad \mathcal{R}(B^*) \cap \mathcal{R}(D^*) = \{0\}.$$

(b) If (3.2) holds, the general solution of (1.1) can be written as the two independent forms

$$(3.3) \quad X = X_0 + S_1 F_G U_1 E_H T_1 + F_A V_1 + V_2 E_B,$$

$$(3.4) \quad Y = Y_0 + S_2 F_G U_2 E_H T_2 + F_C W_1 + W_2 E_D,$$

where X_0 and Y_0 are a pair of special solutions to (1.1), U_1, U_2, V_1, V_2, W_1 and W_2 are arbitrary.

Proof. Writing (2.3) and (2.4) as two independent matrix expressions, substituting them into $M - AXB - CYD$ and observing $AS_1 F_G = -CS_2 F_G$ and $E_H T_1 B = E_H T_2 D$ gives

$$\begin{aligned} M - AXB - CYD &= M - AX_0 B - CY_0 D - AS_1 F_G U_1 E_H T_1 B - CS_2 F_G U_2 E_H T_2 D \\ &= -AS_1 F_G U_1 E_H T_1 B - CS_2 F_G U_2 E_H T_2 D \\ &= -AS_1 F_G U_1 E_H T_1 B + AS_1 F_G U_2 E_H T_1 B \\ &= AS_1 F_G (-U_1 + U_2) E_H T_1 B, \end{aligned}$$

where U_1 and U_2 are arbitrary. Then it follows by (1.3) that

$$\begin{aligned} \max_{X \in J_1, Y \in J_2} r(M - AXB - CYD) &= \max_{U_1, U_2} r[AS_1 F_G (-U_1 + U_2) E_H T_1 B] \\ &= \min \{ r(AS_1 F_G), \quad r(E_H T_1 B) \}, \end{aligned}$$

where by Lemma 1.1

$$\begin{aligned} r(AS_1 F_G) &= r \begin{bmatrix} AS_1 \\ G \end{bmatrix} - r(G) = r \begin{bmatrix} A & 0 \\ A & C \end{bmatrix} - r(G) = r(A) + r(C) - r(G), \\ r(E_H T_1 B) &= r[T_1 B, H] - r(H) = r \begin{bmatrix} B & B \\ 0 & -D \end{bmatrix} - r(H) = r(B) + r(D) - r(H). \end{aligned}$$

Therefore, (3.1) follows. Result (3.2) follows from (3.1); the solutions in (3.3) and (3.4) follow from (2.3) and (2.4). \square

Remark 3.2. The matrix equation (1.1) is one of the basic linear matrix equations. Many other types of matrix equations can be solved through (1.1). For example, From Lemma 1.2, one can derive necessary and sufficient conditions for the matrix equation $AXA^* + BYB^* = C$ to have Hermitian and skew-Hermitian solutions. From Lemma 2.1, one can also give necessary and sufficient conditions for the two matrix equations $AXB + (AXB)^* = C$ and $AXB - (AXB)^* = C$ to be solvable.

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