# PERTURBATION ANAYSIS FOR THE MATRIX EQUATION $X = I - A^*X^{-1}A + B^*X^{-1}B$

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ABSTRACT. The purpose of this paper is to study the perturbation analysis of the matrix equation  $X = I - A^*X^{-1}A + B^*X^{-1}B$ . Based on the matrix differentiation, we give a precise perturbation bound for the positive definite solution. A numerical example is presented to illustrate the shrpness of the perturbation bound.

### 1. Introduction

We consider the matrix equation

(1.1) 
$$X = Q - A^*X^{-1}A + B^*X^{-1}B,$$

where A, B are arbitrary  $n \times n$  matrices. Some special cases of Equation (1.1) are problems of practical importance, such as the matrix equation  $X + M^*X^{-1}M = Q$  that arises in the control theory, ladder networks, dynamic programming, stochastic filtering, statistics, and so on [5, 8, 10]. The matrix equation  $X - M^*X^{-1}M = Q$  arises in the analysis of stationary Gaussian reciprocal processes over a finite interval [1, 7].

In [2], Berzig, Duan and Samet established the existence and uniqueness of a positive definite solution of (1.1) via the Bhaskar-Lakshkanthan coupled fixed point theorem([3]).

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Theorem 1.1 ([2]). If there exist a, b > 0 satisfying following conditions

- (i)  $a^{-1}A^*A + aI \le Q \le bI$ ,
- (ii)  $bA^*A aB^*B \le ab(Q aI)$ ,
- (iii)  $bB^*B aA^*A \le ab(bI Q),$
- (iv)  $A^*A < \frac{a^2}{2}I, B^*B < \frac{a^2}{2}I$

then (1.1) has a unique solution  $X \in [aI, \infty)$  and

$$X \in [Q + b^{-1}B^*B - a^{-1}A^*A, \ Q + a^{-1}B^*B - b^{-1}A^*A].$$

The following result is immediate consequence of Theorem 1.1.

Theorem 1.2. If there exist  $0 < a \le \frac{2}{3}$  such that

$$A^*A \le \frac{a^2}{2}I, \quad B^*B \le \frac{a^2}{2}I,$$

then the matrix equation

$$(1.2) X = I - A^* X^{-1} A + B^* X^{-1} B.$$

has a unique solution  $X_U \in [aI, \infty)$  and

$$(1.3) X_U \in \left[ I + \frac{2}{2+a} B^* B - \frac{1}{a} A^* A, I + \frac{1}{a} B^* B - \frac{2}{2+a} A^* A \right].$$

In this paper, we study the perturbation analysis of the matrix equation (1.2). Based on the matrix differentiation, we firstly give a differential bound for the unique solution of (1.2) in certain set, and then use it to derive a precise perturbation bound. A numerical example is used to show that the perturbation bound is very sharp.

Throughout this paper, we write B > 0  $(B \ge 0)$  if the matrix B is positive definite (semidefinite). If B - C is positive definite (semidefinite), then we write B > C  $(B \ge C)$ . If a positive definite matrix X satisfies  $B \le X \le C$ , we denote that  $X \in [B, C]$ . The symbols  $\lambda_1(B)$  and  $\lambda_n(B)$  denote the maximal and minimal eigenvalues of an  $n \times n$  Hermitian matrix B, respectively. The symbol ||B|| denotes the spectral norm of the matrix B.

## 2. Perturbation Analysis for the Matrix equation (1.2)

Based on the matrix differentiation, we firstly give a differential bound for the unique positive definite solution  $X_U$  of (1.2), and then use it to derive a precise perturbation bound for  $X_U$  in this section.

DEFINITION 2.1. ([6, 9]) Let  $F = (f_{ij})_{mn}$ , then the matrix differentiation of F is  $dF = (df_{ij})_{mn}$ . For example, let

$$F = \left(\begin{array}{cc} s+t & s^2-2t \\ 2s+t^3 & t^2 \end{array}\right).$$

Then

$$dF = \begin{pmatrix} ds + dt & 2sds - 2dt \\ 2ds + 3t^2dt & 2tdt \end{pmatrix}.$$

LEMMA 2.2 ([6, 9]). The matrix differentiation has the following properties:

- (1) dA = 0 for a constant matrix A;
- (2)  $d(\alpha X) = \alpha(dX)$ , where  $\alpha$  is a complex number;
- (3) d(X + Y) = dX + dY;
- $(4) \ d(XY) = (dX)Y + X(DY);$
- (5)  $d(X^*) = (dX)^*;$ (6)  $d(X^{-1}) = -X^{-1}(DX)X^{-1}.$

THEOREM 2.3. If there exist  $0 < a \le \frac{1}{2}$  such that

(2.4) 
$$||A||^2 \le \frac{a^2}{2}, \quad ||B||^2 \le \frac{a^2}{2},$$

then then (1.2) has a unique solution  $X_U \in [aI, \infty)$ , and it satisfies

(2.5) 
$$||dX_U|| \le \frac{2a(||A|| ||dA|| + ||B|| ||dB||)}{a^2 - ||A||^2 - ||B||^2}.$$

*Proof.* Since

$$\lambda_1(A^*A) \le ||A^*A|| \le ||A||^2,$$
  
 $\lambda_1(B^*B) \le ||B^*B|| \le ||B||^2$ 

then

(2.6) 
$$A^*A \le \lambda_1(A^*A)I \le ||A^*A||I \le ||A||^2I, B^*B \le \lambda_1(B^*B)I \le ||B^*B||I \le ||B||^2I.$$

Combining (2.4) and (2.6) we haver

$$A^*A \le \frac{a^2}{2}I, \quad B^*B \le \frac{a^2}{2}I.$$

Then by Theorem 1.2 we have that (1.2) has a unique solution  $X_U$  in  $[aI, \infty)$ , which satisfies

(2.7) 
$$X_U \in \left[ I + \frac{2}{2+a} B^* B - \frac{1}{a} A^* A, I + \frac{1}{a} B^* B - \frac{2}{2+a} A^* A \right].$$

Since  $X_U$  is the unique solution of (1.2) in  $[aI, \infty)$ ,

$$(2.8) X_U + A^* X_U A - B^* X_U B = I.$$

It is known that the elements of  $X_U$  are differentiable functions of the elements of A and B. Differentianting (2.8), and by Lemma 2.2, we have

$$dX_U + (dA^*)X_U^{-1}A - A^*X_U^{-1}(dX_U)X_U^{-1}A + A^*X_U^{-1}(dA)$$
$$-(dB^*)X_U^{-1}B + B^*X_U^{-1}(dX_U)X_U^{-1}B - B^*X_U^{-1}(dB) = 0,$$

which implies that

(2.9)

$$dX_{U} - A^{*}X_{U}^{-1}(dX_{U})X_{U}^{-1}A + B^{*}X_{U}^{-1}(dX_{U})X_{U}^{-1}B$$
  
=  $-(dA^{*})X_{U}^{-1}A - A^{*}X_{U}^{-1}(dA) + (dB^{*})X_{U}^{-1}B + B^{*}X_{U}^{-1}(dB).$ 

By taking spectral norm for both sides of (2.9), we have that (2.10)

$$\begin{aligned} & \| - (dA^*)X_U^{-1}A - A^*X_U^{-1}(dA) + (dB^*)X_U^{-1}B + B^*X_U^{-1}(dB) \| \\ & \leq \| (dA^*)X_U^{-1}A\| + \|A^*X_U^{-1}(dA)\| + \|(dB^*)X_U^{-1}B\| + \|B^*X_U^{-1}(dB)\| \\ & \leq \| dA^*\| \|X_U^{-1}\| \|A\| + \|A^*\| \|X_U^{-1}\| \|dA\| + \|dB^*\| \|X_U^{-1}\| \|B\| \\ & + \|B^*\| \|X_U^{-1}\| \|dB\| \\ & = 2\|X_U^{-1}\| (\|dA\| \|A\| + \|dB\| \|B\|) \\ & \leq \frac{2}{a} (\|dA\| \|A\| + \|dB\| \|B\|) \end{aligned}$$

and

$$(2.11) \begin{array}{l} \|dX_{U} - A^{*}X_{U}^{-1}(dX_{U})X_{U}^{-1}A + B^{*}X_{U}^{-1}(dX_{U})X_{U}^{-1}B\| \\ \geq \|dX_{U}\| - \|A^{*}X_{U}^{-1}(dX_{U})X_{U}^{-1}A\| - \|B^{*}X_{U}^{-1}(dX_{U})X_{U}^{-1}B\| \\ \geq \|dX_{U}\| - \|A^{*}\|\|X_{U}^{-1}\|\|dX_{U}\|\|X_{U}^{-1}\|\|A\| \\ - \|B^{*}\|\|X_{U}^{-1}\|\|dX_{U}\|\|X_{U}^{-1}\|\|B\| \\ = (1 - \|A\|^{2}\|X_{U}^{-1}\|^{2} - \|B\|^{2}\|X_{U}^{-1}\|^{2})\|dX_{U}\| \\ \geq \left(1 - \frac{\|A\|^{2}}{a^{2}} - \frac{\|B\|^{2}}{a^{2}}\right)\|dX_{U}\|. \end{array}$$

Due to (2.4) we have

(2.12) 
$$1 - \frac{\|A\|^2}{a^2} - \frac{\|B\|^2}{a^2} > 0.$$

Combination (2.10), (2.11) and noting (2.12), we have

$$\left(1 - \frac{\|A\|^2}{a^2} - \frac{\|B\|^2}{a^2}\right) \|dX_U\| \le \frac{2}{a} (\|dA\| \|A\| + \|dB\| \|B\|)$$

which implies to

$$||dX_U|| \le \frac{2a(||A|| ||dA|| + ||B|| ||dB||)}{a^2 - ||A||^2 - ||B||^2}.$$

THEOREM 2.4. Let  $\tilde{A}$ ,  $\tilde{B}$  be perturbed matrices of A, B in (1.2) and  $\Delta A = \tilde{A} - A$ ,  $\Delta B = \tilde{B} - B$ . If there exist  $0 < a \le \frac{1}{2}$  such that

(2.13) 
$$||A||^2 \le \frac{a^2}{2}, \quad ||B||^2 \le \frac{a^2}{2},$$

(2.14) 
$$2\|A\|\|\Delta A\| + \|\Delta A\|^2 < \frac{a^2}{2} - \|A\|^2,$$

(2.15) 
$$2\|B\|\|\Delta B\| + \|\Delta B\|^2 < \frac{a^2}{2} - \|B\|^2,$$

then then (1.2) and its perturbed equation

(2.16) 
$$\tilde{X} = I - \tilde{A}^* \tilde{X}^{-1} \tilde{A} + \tilde{B}^* \tilde{X}^{-1} \tilde{B}$$

have a unique solutions  $X_U$  and  $\tilde{X}_U$  in  $[aI, \infty)$ , respectively, which satisfy

$$\left\|\tilde{X}_U - X_U\right\| \le S_{err}$$

where

$$S_{err} = \frac{2a(\|A\|\|\Delta A\| + \|\Delta A\|^2 + \|B\|\|\Delta B\| + \|\Delta B\|^2)}{a^2 - (\|A\| + \|\Delta A\|)^2 - (\|B\| + \|\Delta B\|)^2}.$$

*Proof.* Set  $A(t) = A + t\Delta A$  and  $B(t) = B + t\Delta B$ ,  $t \in [0, 1]$  then by (2.14)

(2.17) 
$$\begin{aligned} \|A(t)\|^2 &= \|A + t\Delta A\|^2 \le (\|A\| + t\|\Delta A\|)^2 \\ &= \|A\|^2 + 2t\|A\|\|\Delta A\| + t^2\|\Delta A\|^2 \\ &\le \|A\|^2 + 2\|A\|\|\Delta A\| + \|\Delta A\|^2 \\ &< \|A\|^2 + \frac{a^2}{2} - \|A\|^2 = \frac{a^2}{2}, \end{aligned}$$

similarly, by (2.15) we have

$$(2.18) ||B(t)||^2 < \frac{a^2}{2}.$$

By (2.17), (2.18) and Theorem 2.3 we derive that for arbitrary  $t \in [0, 1]$ , the matrix equation

$$X = I - A(t)^* X^{-1} A(t) + B(t)^* X^{-1} B(t)$$

has a unique solution  $X_U(t)$  in  $[aI, \infty)$ , especially,

$$X_U(0) = X_U, \quad X_U(1) = (\tilde{X})_U,$$

where  $X_U$  and  $\tilde{X}_U$  are the unique solutions of (1.2) and (2.16), respectively.

From Theorem 2.3 it follows that

$$\left\| \tilde{X}_{U} - X_{U} \right\| = \left\| X_{U}(1) - X_{U}(0) \right\| = \left\| \int_{0}^{1} dX_{U}(t) \right\| \leq \int_{0}^{1} \left\| dX_{U}(t) \right\|$$

$$\leq \int_{0}^{1} \frac{2a(\|A(t)\| \|dA(t)\| + \|B(t)\| \|dB(t)\|)}{a^{2} - \|A(t)\|^{2} - \|B(t)\|^{2}}$$

$$\leq \int_{0}^{1} \frac{2a(\|A(t)\| \|\Delta A\|dt + \|B(t)\| \|\Delta B\|dt)}{a^{2} - \|A(t)\|^{2} - \|B(t)\|^{2}}$$

$$\leq \int_{0}^{1} \frac{2a(\|A(t)\| \|\Delta A\| + \|B(t)\| \|\Delta B\|)}{a^{2} - \|A(t)\|^{2} - \|B(t)\|^{2}} dt.$$

Noting that

$$||A(t)|| = ||A + t\Delta A|| \le ||A|| + t||\Delta A||,$$

$$||B(t)|| = ||B + t\Delta B|| \le ||B|| + t||\Delta B||,$$

and combining Mean Value Theorem of Integration, we have

$$\begin{split} & \left\| \tilde{X}_{U} - X_{U} \right\| \\ \leq & \int_{0}^{1} \frac{2a(\|A(t)\| \|\Delta A\| + \|B(t)\| \|\Delta B\|)}{a^{2} - \|A(t)\|^{2} - \|B(t)\|^{2}} dt \\ \leq & \int_{0}^{1} \frac{2a((\|A\| + t\|\Delta A\|) \|\Delta A\| + (\|B\| + t\|\Delta B\|) \|\Delta B\|)}{a^{2} - (\|A\| + t\|\Delta A\|)^{2} - (\|B\| + t\|\Delta B\|)^{2}} dt \\ \leq & \frac{2a((\|A\| + \xi\|\Delta A\|) \|\Delta A\| + (\|B\| + \xi\|\Delta B\|) \|\Delta B\|)}{a^{2} - (\|A\| + \xi\|\Delta A\|)^{2} - (\|B\| + \xi\|\Delta B\|)^{2}} \\ & \times (1 - 0) \quad (0 < \xi < 1) \\ \leq & \frac{2a((\|A\| + \|\Delta A\|) \|\Delta A\| + (\|B\| + \|\Delta B\|) \|\Delta B\|)}{a^{2} - (\|A\| + \|\Delta A\|)^{2} - (\|B\| + \|\Delta B\|)^{2}} = S_{err}. \end{split}$$

## 3. Numerical Experiments

In this section, we use a numerical example to confirm the correctness of Theorem 2.4 and the precision of the perturbation bound for the unique positive definite solution  $X_U$  of (1.2).

Example 3.1. Consider the matrix equation

$$X = I - A^*X^{-1}A + B^*X^{-1}B,$$

and its pertubed equation

(3.19) 
$$\tilde{X} = I - \tilde{A}^* \tilde{X}^{-1} \tilde{A} + \tilde{B}^* \tilde{X}^{-1} \tilde{B},$$

where

$$\begin{split} A &= \left( \begin{array}{ccc} 0.1 & 0.05 & 0 \\ 0 & 0.05 & -0.02 \\ 0.05 & -0.05 & -0.05 \end{array} \right), \tilde{A} = A + \left( \begin{array}{ccc} 0.5 & 0.1 & -0.1 \\ -0.1 & 0.5 & 0.5 \\ -0.2 & 0.1 & -0.1 \end{array} \right) \times 10^{-j}, \\ B &= \left( \begin{array}{ccc} -0.05 & 0.1 & 0 \\ -0.05 & 0 & -0.05 \\ 0.05 & 0 & -0.1 \end{array} \right), \tilde{B} = B + \left( \begin{array}{ccc} 0.1 & 0.02 & 0.05 \\ -0.2 & 0.12 & 0.14 \\ -0.25 & 0.2 & 0.26 \end{array} \right) \times 10^{-j}, \\ j \in \mathbb{N}. \end{split}$$

It is easy to verify that the conditions (2.13)-(2.15) are satisfied with a = 0.5, then (1.2) and its perturbed equation (3.19) have unique positive definite solutions  $X_U$  and  $\tilde{X}_U$ , respectively. From Berzig, Duan and Samet [2] it follows that the sequence  $\{X_k\}$  and  $\{Y_k\}$  generated by the iterative method

$$\begin{split} X_0 &= 0.5I, \quad Y_0 = 5I, \\ X_{k+1} &= I - A^* X_k^{-1} A + B^* Y_k^{-1} B \\ Y_{k+1} &= I - A^* Y_k^{-1} A + B^* X_k^{-1} B, \quad k = 0, 1, 2, \ldots. \end{split}$$

both convege to  $X_U$ . Choose  $\tau = 1.0 \times 10^{-15}$  as the termination scalar, that is,

$$R(X_k) = ||X_k + A^* X_k^{-1} A - B^* X_k^{-1} B - I||$$
  

$$R(Y_k) = ||Y_k + A^* Y_k^{-1} A - B^* Y_k^{-1} B - I||$$

and

$$R(X) = \max\{R(X_k), R(Y_K)\} \le \tau = 1.0 \times 10^{-15}.$$

By using the iterative method we can get the computed solution X of (1.2). Since  $R(X) < 1.0 \times 10^{-15}$ , then the computed solution X has a very high precision. For simplicity, we write the computed solution as

the unique positive definite solution  $X_U$ . Similarly, we can also get the unique positive definite solution  $\tilde{X}_U$  of the perturbed equation (3.19).

Some numerical results on the perturbation bounds for the unique positive definite solution  $X_U$  are listed in Table 1. From Table 1, we see that Theorem 2.4 gives a precise perturbation bound for the unique positive definite solution of (1.2).

Table 1. Numerical results for the different value of j

j	2	3	4	5	6
$\ \tilde{X}_U - X_U\ /\ X_U\ $					
$S_{err}/\ X_U\ $	$7.867 \times 10^{-3}$	$7.356 \times 10^{-4}$	$7.305 \times 10^{-5}$	$7.300 \times 10^{-6}$	$7.299 \times 10^{-7}$

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