

On Positive Definite Solutions of the Linear Matrix Equation $X + A^*XA = I$

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Abstract—Two effective iterative methods are constructed to solve the linear matrix equation of the form $X + A^*XA = I$. Some properties of a positive definite solution of the linear matrix equation are discussed. Necessary and sufficient conditions for existence of a positive definite solution are derived for $\|A\| < 1$ and $\|A\| > 1$. Several numerical examples are given to show the efficiency of the presented iterative methods.

Index Terms—Algorithm, Fixed-Point-Iteration, linear-Matrix-Equation, Numerical-Analysis, Positive-Definite -Solutions, Two-Sided-Iteration.

I. INTRODUCTION

Considering the linear matrix equation $X + A^*XA = I$, (1) with unknown matrix X , where $A \in C^{n \times n}$, I is the identity matrix of order n . The equation (1) could be viewed as a special case of the symmetric matrix equations $X \pm A_1^*XA_1 \pm \dots \pm A_m^*XA_m = Q$. (2)

Where Q is a positive definite matrix [15]. There are many linear matrix equations which were studied by some authors [2],[3],[8]-[11],[13]-[15],[18]-[21],[24]. Two effective iterative methods for computing a positive definite solution of this equation are proposed. The first one is fixed point iteration method and the second one is two sided iteration method of the fixed point iteration method. These two iterative methods are used for computing a positive definite solution of nonlinear matrix equations, see [1],[4]-[7],[12],[16],[17],[22],[23].

This paper aims to find the positive definite solution of the matrix equation (1) for all values of $\|A\| \neq 1$, for this purpose we investigated two iterative methods, the first one is based on fixed point iteration and the second is based on two sided iteration method, also to derive necessary and sufficient conditions for the existence of the solution of equation (1).

Section II describes some properties of positive definite solutions of the equation (1). Section III, presents a first iterative method (Fixed point iteration method) for obtaining the solution of our problem. Also, it presents theorems for obtaining the necessary and sufficient conditions for the existence of a solution of matrix equation (1). Section IV represents the second iterative method (Two sided iteration method of the fixed point iteration method) for obtaining the solution of the problem and theorems for the sufficient conditions for the existence of a positive definite solution of (1). Numerical examples in Section V illustrate the effectiveness of these methods. Conclusion drawn from the results obtained in this paper are in section VI.

The notation $X > 0$ means that X is a positive definite Hermitian matrix and $A > B$ is used to indicate that $A - B > 0$. A^* denotes the complex conjugate transpose of A . Finally, throughout the paper, $\|\cdot\|$ will be the spectral norm for square matrices unless otherwise noted.

II. SOME PROPERTIES OF THE SOLUTIONS

This section discusses some properties of positive definite solutions of the matrix equation (1).

1) Theorem

If m and M are the smallest and the largest eigenvalues of a solution X of (1), respectively, and λ is an eigenvalue of A , then $\sqrt{\frac{1-M}{M}} \leq |\lambda| \leq \sqrt{\frac{1-m}{m}}$.

Proof

Let v be an eigenvector corresponding to an eigenvalue λ of the matrix A and $\|v\| = 1$. Since the solution X of (1) is a positive definite matrix, then $0 < m \leq M < 1$,

$$\langle (X + A^*XA)v, v \rangle = \langle Av, v \rangle = \langle v, v \rangle, \langle Xv, v \rangle + \langle A^*XAv, v \rangle = \langle v, v \rangle, \langle Xv, v \rangle + \langle XAv, Av \rangle = 1 \text{ and } \langle Xv, v \rangle + |\lambda|^2 \langle Xv, v \rangle = 1.$$

Consequently, $\sqrt{\frac{1-M}{M}} \leq |\lambda| \leq \sqrt{\frac{1-m}{m}}$.

2) Theorem

If (1) has a positive definite solution X , then $A^*A + (AA^*)^{-1} > I$

Proof

Since X is a positive definite solution of (1), then $X < I, A^*XA < I$, i.e. $X < (AA^*)^{-1}$.

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Thus we have $(AA^*)^{-1} > X = I - A^*XA > I - A^*A$.

Consequently, $A^*A + (AA^*)^{-1} > I$.

III. THE FIRST ITERATION METHOD (FIXED POINT ITERATION METHOD)

This section establishes the first iterative method which is suitable for obtaining a positive definite solution of (1) when $\|A\| < 1$.

A. Algorithm

Take $X_0 = \alpha I$. For $k = 0, 1, 2, \dots$, compute

$$X_{k+1} = I - A^*X_kA. \quad (3)$$

Our theorems give necessary and sufficient conditions for the existence of a positive definite solution of (1).

1) Theorem

Let the sequence $\{X_k\}$ be determined by the Algorithm

$$A \text{ and } \|A\|^2 < 1 \quad (4)$$

If (1) has a positive definite solution, then $\{X_k\}$ converges to X , which is a solution of (1) for all numbers $\alpha > 1$. Moreover, if $X_k > 0$ for every k , then (1) has a positive definite solution.

Proof.

Let (1) has a positive definite solution. From Algorithm A, we have

$$X_0 = \alpha I > I > X_1 = I - \alpha A^*A,$$

$$X_0 = \alpha I > I > X_2 = I - A^*X_1A > I - \alpha A^*A = X_1, \text{ i.e.,}$$

$X_0 > X_2 > X_1$. To prove $X_s > X_l$ if $X_{s-1} < X_0$ for all s , we have $X_s = I - A^*X_{s-1}A > I - A^*X_0A = I - \alpha A^*A = X_1$. We

will find the relation between X_2, X_3, X_4, X_5 . Since

$$X_1 < X_2, \text{ then } X_2 = I - A^*X_1A > I - A^*X_2A = X_3 \text{ and}$$

$$X_4 = I - A^*X_3A > I - A^*X_2A = X_3, \text{ since } X_3 > X_1, \text{ then}$$

$$X_4 = I - A^*X_3A < I - A^*X_1A = X_2 \text{ and}$$

$$X_5 = I - A^*X_4A > I - A^*X_2A = X_3.$$

Also, since $X_4 > X_3$, then

$$X_5 = I - A^*X_4A < I - A^*X_3A = X_4. \text{ Thus, we get}$$

$$X_0 = \alpha I > X_2 > X_4 > X_5 > X_3 > X_1 = I - \alpha A^*A.$$

We will prove that $X_0 > X_{s+1} > X_s$, if we have

$$X_0 > X_{s-1} > X_s, \text{ thus}$$

$$X_0 = \alpha I > I > I - A^*X_{s-1}A < I - A^*X_sA \Rightarrow X_0 > X_{s+1} > X_s.$$

Also, we can prove that $X_0 > X_s > X_{s+1}$, if we have

$$X_0 > X_s > X_{s-1}, X_0 = \alpha I > I - A^*X_{s-1}A > I - A^*X_sA \Rightarrow$$

$$X_0 > X_s > X_{s+1}. \text{ Therefore, we have}$$

$$X_0 = \alpha I > X_{2r} > X_{2r+2} > X_{2s+3} > X_{2s+1} > X_1 = I - \alpha A^*A, \text{ for}$$

every positive integers r, s . Consequently, the subsequences $\{X_{2r}\}, \{X_{2s+1}\}$ are monotonic and bounded, and $\lim_{s \rightarrow \infty} X_{2s}$,

$$\lim_{s \rightarrow \infty} X_{2s+1} \text{ exist.}$$

To prove these sequences have a common limit, we have

$$\begin{aligned} \|X_{2s} - X_{2s+1}\| &= \|I - A^*X_{2s-1}A - I + A^*X_{2s}A\| \\ &= \|A^*(X_{2s} - X_{2s-1})A\| \leq \|A\|^2 \|X_{2s} - X_{2s-1}\|. \end{aligned}$$

Let $q = \|A\|^2 < 1$, and we get

$$\begin{aligned} \|X_{2s} - X_{2s+1}\| &\leq q \|X_{2s} - X_{2s-1}\| \leq q^2 \|X_{2s-2} - X_{2s-1}\| \\ &\leq \dots \leq q^{2s} \|X_0 - X_1\| \leq q^{2s} (2\alpha - 1). \end{aligned}$$

Since $q < 1$ and $(2\alpha - 1) > 0$, $\|X_{2s} - X_{2s+1}\| \rightarrow 0$ as $s \rightarrow \infty$, that is, $\{X_{2r}\}$ and $\{X_{2s+1}\}$ have the same limit X and $X_{2s} > X > X_{2s+1}, s = 1, 2, \dots$. Taking the limit of the sequence $\{X_k\}$ generated by Algorithm A leads to $X = I - A^*XA$, which is a solution of (1). Assuming that $X_k > 0$ for every k . We proved that the sequences have a common limit X . Since $X_{k+1} = I - A^*X_kA > 0$, taking the limits of both sides as k approaches to ∞ , we get $X = I - A^*XA > 0$. Hence (1) has a positive definite solution.

2) Theorem

Let X_k be the iterates in Algorithm A. If $q = \|A\|^2 < 1$.

Then $\|X_k - X\| < q^k (2\alpha - 1)$, for all real number $\alpha > 1$,

where X is a positive definite solution of (1).

Proof

From previous Theorem it follows that the sequence (3) is convergent to a positive definite solution X of (1). From (3) and $X = I - A^*XA$, we have

$$X_k - X = A^*(X - X_{k-1})A. \text{ Thus,}$$

$$\begin{aligned} \|X_k - X\| &= \|A^*(X - X_{k-1})A\| \leq \|A\|^2 \|X - X_{k-1}\| \leq q \|X - X_{k-1}\| \\ &\leq q^2 \|X - X_{k-2}\|, \text{ after } k - \text{ steps} \\ &\leq q^k \|X_0 - X\|. \end{aligned}$$

From Theorem 1, we have

$$\|X_k - X\| \leq q^k \|X_0 - X\| < q^k (2\alpha - 1).$$

3) Corollary

Suppose that (1) has a solution. If $q = \|A\|^2 < 1$, then $\{X_k\}$ converges to X with at least the linear convergence rate.

Proof

We have $\|X_{k+1} - X\| \leq \|A\|^2 \|X_k - X\|$. Choose a real number satisfying $\|A\| < \theta < 1$. Since $X_k \rightarrow X$, there exists

a N such that for any $k \geq N$, $\|A\| \leq \theta$. Hence

$$\|X_{k+1} - X\| \leq \theta^2 \|X_k - X\|.$$

4) Theorem

If (1) has a positive definite solution and after k iterative steps of Algorithm A, we have $\|I - X_k^{-1}X_{k-1}\| < \varepsilon$, then

$$\|X_k + A^*X_kA - I\| < \alpha \varepsilon \|A\|^2, \text{ where } X_k \text{ is the iterates in Algorithm A.}$$

Proof

$$X_k + A^*X_k A - I = X_k + A^*X_k A - X_k - A^*X_{k-1}A$$

$$= A^*(X_k - X_{k-1})A. \quad (5)$$

Take the norms of both sides of (5)

$$\|X_k + A^*X_k A - I\| \leq \|A\|^2 \|X_k\| \|I - X_k^{-1}X_{k-1}\| < \alpha\varepsilon \|A\|^2. \text{ For}$$

$$\|I - X_k^{-1}X_{k-1}\|, X_k \rightarrow X \text{ as } k \rightarrow \infty.$$

Consequently, $\|I - X_k^{-1}X_{k-1}\| \rightarrow 0$ as $k \rightarrow \infty$, that is,
 $\|I - X_k^{-1}X_{k-1}\| < \varepsilon$, for $\varepsilon > 0$ and from *theorem 1*, $\|X_k\| \leq \alpha$
 for every k , thus $\|X_k + A^*X_k A - I\| < \alpha\varepsilon \|A\|^2$.

IV. THE SECOND ITERATION METHOD (TWO SIDED ITERATION METHOD OF THE FIXED POINT ITERATION METHOD)

This section establishes the second iterative method which is suitable for obtaining a positive definite solution of (1).

B. Algorithm

Take $X_0 = \alpha I, Y_0 = \beta I$. For $k = 0, 1, 2, \dots$, compute
 $X_{k+1} = I - A^*X_k A$ and $Y_{k+1} = I - A^*Y_k A$. (6)

Next theorems provide necessary and sufficient conditions for the existence of a solution of (1) when $\|A\| < 1$.

1) Theorem

If (1) has a positive definite solution, the sequences $\{X_k\}$ and $\{Y_k\}$ are determined by *Algorithm B* and $\|A\|^2 < 1$,
 then the two sequences $\{X_k\}, \{Y_k\}$ converge to the positive definite solution X for all real numbers α, β such that $\beta > \alpha > 0$. On the other hand, if $X_k, Y_k > 0$ for every k , $\|A\|^2 < 1$ and $\beta > \alpha > 0$, then (1) has a positive definite solution.

Proof

First, considering sequence (6), for X_1, X_2 we have
 $X_0 = \alpha I > I > X_1 = I - \alpha A^* A = \alpha I > I > X_2 = I - A^* X_1 A >$
 $I - \alpha A^* A = X_1$, i.e., $X_0 > X_2 > X_1$. To prove $X_s > X_1$ if $X_{s-1} < X_0$ for all s , hence
 $X_s = I - A^* X_{s-1} A > I - A^* X_0 A = I - \alpha A^* A = X_1$. We will find the relation between X_2, X_3, X_4, X_5 . Since $X_1 < X_2$, then $X_2 = I - A^* X_1 A > I - A^* X_2 A = X_3$ and
 $X_4 = I - A^* X_3 A > I - A^* X_2 A = X_3$, since $X_3 > X_1$, then
 $X_4 = I - A^* X_3 A < I - A^* X_1 A = X_2$ and
 $X_5 = I - A^* X_4 A > I - A^* X_2 A = X_3$. Also, since $X_4 > X_3$, then $X_5 = I - A^* X_4 A < I - A^* X_3 A = X_4$. Thus, we get
 $X_0 = \alpha I > X_2 > X_4 > X_5 > X_3 > X_1 = I - \alpha A^* A$. We will prove that $X_0 > X_{s+1} > X_s$, if we have $X_0 > X_{s-1} > X_s$, thus
 $X_0 = \alpha I > I > I - A^* X_{s-1} A < I - A^* X_s A \Rightarrow X_0 > X_{s+1} > X_s$.
 Also, we can prove that $X_0 > X_s > X_{s+1}$, if we have
 $X_0 > X_s > X_{s-1}$, thus
 $X_0 = \alpha I > I - A^* X_{s-1} A > I - A^* X_s A \Rightarrow X_0 > X_s > X_{s+1}$.

Therefore, we have

$X_0 = \alpha I > X_{2r} > X_{2r+2} > X_{2s+3} > X_{2s+1} > X_1 = I - \alpha A^* A$, for every positive integers r, s . Consequently, the subsequences $\{X_{2r}\}, \{X_{2s+1}\}$ are monotonic and bounded, and $\lim_{s \rightarrow \infty} X_{2s}$ and $\lim_{s \rightarrow \infty} X_{2s+1}$ exist.

For the sequence $\{Y_k\}$, similarly, we can prove that
 $Y_0 = \beta I > Y_{2r} > Y_{2r+2} > Y_{2s+3} > Y_{2s+1} > Y_1 = I - \beta A^* A$, for every positive integers r, s .
 Consequently, the subsequences $\{Y_{2r}\}, \{Y_{2s+1}\}$ are monotonic and bounded, and $\lim_{s \rightarrow \infty} Y_{2s}$ and $\lim_{s \rightarrow \infty} Y_{2s+1}$ exist.

Finally, from (6) we have $Y_0 = \beta I > \alpha I = X_0$ and
 $Y_1 = I - A^* Y_0 A < I - A^* X_0 A = X_1$, i.e., $Y_0 > X_0 > X_1 > Y_1$.
 Also, $Y_2 = I - A^* Y_1 A > I - A^* X_1 A = X_2$.

Similarly, $Y_3 < X_3, Y_4 > X_4$. Since $Y_0 > X_1, X_0 > Y_1$, then
 $X_1 = I - A^* X_0 A < I - A^* Y_1 A = Y_2$ and
 $Y_1 = I - A^* Y_0 A < I - A^* X_1 A = X_2$. Also, since $Y_2 > X_1$, then
 $Y_3 = I - A^* Y_2 A < I - A^* X_1 A = X_2$. From (6) we have
 $X_0 = \alpha I > I > I - A^* Y_1 A = Y_2$. Therefore,
 $X_1 = I - A^* X_0 A < I - A^* Y_2 A = Y_3$ and
 $X_2 = I - A^* X_1 A > I - A^* Y_3 A = Y_4$. Consequently,
 $X_1 = I - A^* X_0 A < I - A^* Y_2 A = Y_3$ and
 $X_2 = I - A^* X_1 A > I - A^* Y_3 A = Y_4$. Consequently,
 $Y_0 > X_0 > Y_2 > X_2 > Y_4 > X_4 > X_3 > Y_3 > X_1 > Y_1$.

We will prove that $Y_0 > X_{s+1} > Y_s$, if we have $Y_0 > Y_{s-1} > X_s$, thus
 $Y_s = I - A^* Y_{s-1} A < I - A^* X_s A = X_{s+1} < I < \beta I = Y_0$.

Also, we can prove that $Y_0 > X_s > Y_{s+1}$, if we have
 $Y_0 > Y_s > X_{s-1}$, thus
 $Y_{s+1} = I - A^* Y_s A < I - A^* X_{s-1} A = X_s < I < \beta I = Y_0$.
 Therefore, we have
 $Y_0 > X_0 > Y_{2r} > X_{2r} > Y_{2r+2} > X_{2s+3} > Y_{2s+3} > X_{2s+1} > Y_{2s+1}$
 $> X_1 > Y_1$, for every positive integers r, s .

Finally, to prove that the subsequences $\{Y_{2r}\}, \{Y_{2s+1}\}$ have the same limit, we have

$$\|Y_{2s} - Y_{2s+1}\| = \|I - A^* Y_{2s-1} A - I + A^* Y_{2s} A\| =$$

$$\|A^*(Y_{2s} - Y_{2s-1})A\| \leq \|A\|^2 \|Y_{2s} - Y_{2s-1}\|,$$

from (7), it follows that $q = \|A\|^2 < 1$, and we get

$$\|Y_{2s} - Y_{2s+1}\| \leq q \|Y_{2s-1} - Y_{2s}\| \leq \dots \leq q^{2s} \|Y_0 - Y_1\| < q^{2s} (2\beta - I).$$

Therefore, we have for the limit X of the subsequences $\{Y_{2r}\}, \{Y_{2s+1}\}$, $Y_{2s} > X > Y_{2s+1}$, $s = 1, 2, \dots$.

Therefore, also we have for the limit X of the subsequences $\{X_{2r}\}, \{X_{2s+1}\}$

$$Y_{2s} > X_{2s} > X > X_{2s+1} > Y_{2s+1}, \quad s = 1, 2, \dots$$

Taking limit in the *Algorithm B* leads to $X = I - A^* X A$.

If $X_k, Y_k > 0$ for every k . We proved that the sequences have the same limit X . Since $X_{k+1} = I - A^* X_k A > 0$ and $Y_{k+1} = I - A^* Y_k A > 0$, hence $X = I - A^* X A$ and (1) has a positive definite solution.

2) *Theorem.*

For the *Algorithm B*, if there exist a positive real numbers α and β such that $\beta > \alpha$ and $q = \|A\|^2 < 1$, then $\|X_k - X\| < q^k(2\alpha - 1)$, $\|Y_k - X\| < q^k(2\beta - 1)$ and $\|X_k - X\| < \|Y_k - X\| < q^k(2\beta - 1)$, where X is a positive definite solution of (1) and $X_k, Y_k, k = 0, 1, 2, \dots$ are defined in (6).

Proof.

From previous Theorem it follows that the sequence (6) is convergent to a positive definite solution X of (1). We will compute the norm of the matrices $X_k - X$ and $Y_k - X$,

$$\begin{aligned} \|X_k - X\| &= \|I - A^*X_{k-1}A - I + A^*XA\| = \|A^*(X - X_{k-1})A\| \\ &\leq \|A\|^2 \|X - X_{k-1}\| \leq q \|X - X_{k-1}\| \leq \dots \leq q^k \|X_0 - X\|. \end{aligned}$$

From previous Theorem, we have

$$\|X_k - X\| \leq q^k \|X_0 - X\| < q^k(2\alpha - 1).$$

Similarly, $\|Y_k - X\| \leq q^k \|Y_0 - X\| < q^k(2\beta - 1)$. Also, we have

$$\begin{aligned} \|X_k - X\| &< \|Y_k - X\|. \text{ Therefore,} \\ \|X_k - X\| &< \|Y_k - X\| < q^k(2\beta - 1). \end{aligned}$$

3) *Corollary*

Suppose that (1) has a solution. If $q = \|A\|^2 < 1$, then $\{X_k\}$ and $\{Y_k\}$ converge to X with at least the linear convergence rate.

4) *Theorem*

If the (1) has a solution and after k -iterative steps of the *Algorithm B*, we have $\|I - X_k^{-1}X_{k-1}\| < \varepsilon$ and

$$\begin{aligned} \|I - Y_k^{-1}Y_{k-1}\| &< \varepsilon. \text{ Then } \|X_k + A^*X_kA - I\| < \alpha\varepsilon\|A\|^2 \text{ and} \\ \|Y_k + A^*Y_kA - I\| &< \beta\varepsilon\|A\|^2. \end{aligned}$$

$$\text{Also, } \|X_k + A^*X_kA - I\| < \|Y_k + A^*Y_kA - I\| < \beta\varepsilon\|A\|^2,$$

where $X_k, Y_k, k = 0, 1, 2, \dots$ are defined in (6) and $\varepsilon > 0$.

Proof

Since, $X_k + A^*X_kA - I = X_k + A^*X_kA - X_k - A^*X_{k-1}A = A^*(X_k - X_{k-1})A$. Take the norms of both sides

$$\begin{aligned} \|X_k + A^*X_kA - I\| &\leq \|A\|^2 \|X_k - X_{k-1}\| \\ &\leq \|A\|^2 \|X_k\| \|I - X_k^{-1}X_{k-1}\| < \alpha\varepsilon\|A\|^2. \end{aligned}$$

Similarly, $\|A^*Y_kA - Y_k - I\| < \beta\varepsilon\|A\|^2$. From *Theorem I*, we

have $\|X_{k+1} - X_k\| < \|Y_{k+1} - Y_k\|$ and since

$$\|X_{k+1} - X_k\| = \|I - A^*X_kA - X_k\| = \|X_k + A^*X_kA - I\| \text{ and}$$

$$\|Y_{k+1} - Y_k\| = \|I - A^*Y_kA - Y_k\| = \|Y_k + A^*Y_kA - I\|, \text{ then}$$

$$\|X_k + A^*X_kA - I\| < \|Y_k + A^*Y_kA - I\| < \beta\varepsilon\|A\|^2.$$

C. Algorithm

Take $X_0 = \alpha I, Y_0 = \beta I$. For $k = 0, 1, 2, \dots$, compute

$$X_{k+1} = B^*(I - X_k)B \text{ and } Y_{k+1} = B^*(I - Y_k)B. \quad (8)$$

where $B = A^{-1}, B^* = A^{*-1}$.

Next theorems provide necessary and sufficient conditions for the existence of a solution of (1) when $\|A\| > 1$.

1) *Theorem*

If (1) has a positive definite solution, the sequences $\{X_k\}$ and $\{Y_k\}$ are determined by the *Algorithm C* and the inequalities

$$\begin{aligned} \text{(i) } B^*B &< \alpha I \text{ and } B^*B < \beta I, \quad 0 < \alpha < \beta, \\ \text{(ii) } q &= \|B\|^2 < 1, \end{aligned} \quad (9)$$

are satisfied, then $\{X_k\}, \{Y_k\}$ converge to a positive definite solution X . Moreover, if $X_k > 0$ and $Y_k > 0$ for every k , $B^*B < \alpha I$, $B^*B < \beta I$ and $\alpha, \beta > 0$, then (1) has a positive definite solution.

Proof

First, from *Algorithm C*, we have

$$X_0 = \alpha I > B^*B > X_1 = B^*B - \alpha B^*B \text{ and}$$

$$X_0 = \alpha I > B^*B > X_2 = B^*B - B^*X_1B > B^*B - \alpha B^*B = X_1. \text{ i.e.,}$$

$X_0 > X_2 > X_1$. To prove $X_s > X_1$ if $X_{s-1} < X_0$ for all s ,

from *Algorithm C*.

$$X_s = B^*B - B^*X_{s-1}B > B^*B - B^*X_0B = B^*B - \alpha B^*B = X_1. \text{ We}$$

will find the relation between X_2, X_3, X_4, X_5 . Since

$$X_1 < X_2, \text{ then } X_2 = B^*B - B^*X_1B > B^*B - B^*X_2B = X_3 \text{ and}$$

$$X_4 = B^*B - B^*X_3B > B^*B - B^*X_2B = X_3, \text{ since } X_3 > X_1,$$

$$\text{then } X_4 = B^*B - B^*X_3B < B^*B - B^*X_1B = X_2 \text{ and}$$

$$X_5 = B^*B - B^*X_4B > B^*B - B^*X_2B = X_3. \text{ Also since}$$

$$X_4 > X_3, \text{ then } X_5 = B^*B - B^*X_4B < B^*B - B^*X_3B = X_4. \text{ Thus,}$$

$$\text{we get } X_0 = \alpha I > X_2 > X_4 > X_5 > X_3 > X_1 = B^*B - \alpha B^*B.$$

We will prove that $X_0 > X_{s+1} > X_s$ if we have

$$X_0 > X_{s-1} > X_s, \text{ thus}$$

$$X_0 = \alpha I > B^*B > B^*B - B^*X_{s-1}B < B^*B - B^*X_sB. \text{ i.e.,}$$

$$X_0 > X_{s+1} > X_s. \text{ Also, we will prove that } X_0 > X_s > X_{s+1} \text{ if}$$

$$\text{we have } X_0 > X_s > X_{s-1}, \text{ thus}$$

$$X_0 = \alpha I > B^*B - B^*X_{s-1}B > B^*B - B^*X_sB. \text{ i.e.,}$$

$$X_0 > X_s > X_{s+1}. \text{ Therefore, we have}$$

$$\begin{aligned} X_0 = \alpha I > X_{2r} > X_{2r+2} > X_{2s+3} > X_{2s+1} > X_1 = B^*B - \alpha B^*B \\ &= (1 - \alpha)B^*B, \end{aligned}$$

for every positive integers r, s . Consequently, the subsequences $\{X_{2r}\}, \{X_{2s+1}\}$ are monotonic and bounded, and $\lim_{s \rightarrow \infty} X_{2s}, \lim_{s \rightarrow \infty} X_{2s+1}$ exist.

For the sequence $\{Y_k\}$, similarly,

$$Y_0 = \beta I > Y_{2r} > Y_{2r+2} > Y_{2s+3} > Y_{2s+1} > Y_1 = (1 - \beta)B^*B, \text{ for every}$$

positive integers r, s . Consequently, the subsequences $\{Y_{2r}\}, \{Y_{2s+1}\}$ are monotonic and bounded, and $\lim_{s \rightarrow \infty} Y_{2s},$

$$\lim_{s \rightarrow \infty} Y_{2s+1} \text{ exist.}$$

Finally, from (8) we have $Y_0 = \beta I > \alpha I = X_0$ and
 $Y_1 = B^*B - B^*Y_0B < B^*B - B^*X_0B = X_1$,
 i.e. $Y_0 > X_0 > X_1 > Y_1$. Also,
 $Y_2 = B^*B - B^*Y_1B > B^*B - B^*X_1B = X_2$. Similarly,
 $Y_3 < X_3, Y_4 > X_4$. Since $Y_0 > X_1, X_0 > Y_1$, then
 $X_1 = B^*B - B^*X_0B < B^*B - B^*Y_1B = Y_2$ and
 $Y_1 = B^*B - B^*Y_0B < B^*B - B^*X_1B = X_2$.
 Also, since $Y_2 > X_1$, then
 $Y_3 = B^*B - B^*Y_2B < B^*B - B^*X_1B = X_2$. From (8) we have
 $X_0 = \alpha I > B^*B > B^*B - B^*Y_1B = Y_2$. Therefore,
 $X_1 = B^*B - B^*X_0B < B^*B - B^*Y_2B = Y_3$ and
 $X_2 = B^*B - B^*X_1B > B^*B - B^*Y_3B = Y_4$. Consequently,
 $Y_0 > X_0 > Y_2 > X_2 > Y_4 > X_4 > X_3 > Y_3 > X_1 > Y_1$.

We will prove that $Y_0 > X_{s+1} > Y_s$, if we have
 $Y_0 > Y_{s-1} > X_s$,
 $Y_s = B^*B - B^*Y_{s-1}B < B^*B - B^*X_sB = X_{s+1} < B^*B < \beta I = Y_0$.
 Also, we can prove that $Y_0 > X_s > Y_{s+1}$. if we have
 $Y_0 > Y_s > X_{s-1}$,
 $Y_{s+1} = B^*B - B^*Y_sB < B^*B - B^*X_{s-1}B = X_s < B^*B < \beta I = Y_0$.

Therefore, we have
 $Y_0 > X_0 > Y_{2r} > X_{2r} > Y_{2r+2} > X_{2r+2} > Y_{2s+3} > X_{2s+3} > Y_{2s+1} > X_1 > Y_1$,
 for every positive integers r, s .

Finally, to prove that the subsequences $\{Y_{2r}\}, \{Y_{2s+1}\}$ have the
 same limit, we have
 $\|Y_{2s} - Y_{2s+1}\| = \|B^*B - B^*Y_{2s-1}B - B^*B + B^*Y_{2s}B\| = \|B^*(Y_{2s} - Y_{2s-1})B\| \leq \|B\|^2 \|Y_{2s} - Y_{2s-1}\|$.

From (9), let $q = \|B\|^2 < 1$, and we get
 $\|Y_{2s} - Y_{2s+1}\| \leq q \|Y_{2s} - Y_{2s-1}\| \leq \dots \leq q^{2s} \|Y_0 - Y_1\| < q^{2s} (2\beta - 1)$.
 Therefore, we have for the limit X of the subsequences
 $\{Y_{2r}\}, \{Y_{2s+1}\}$,
 $Y_{2s} > X > Y_{2s+1}, \quad s = 1, 2, \dots$.

Therefore, also we have for the limit X of the
 subsequences $\{X_{2r}\}, \{X_{2s+1}\}$
 $Y_{2s} > X_{2s} > X > X_{2s+1} > Y_{2s+1}, \quad s = 1, 2, \dots$.
 Taking the limit of (8) as $s \rightarrow \infty$, we have
 $X = B^*(I - X)B$.

If $X_k > 0$ and $Y_k > 0$ for every k . We proved that the
 sequences have the same limit X . Since
 $X_{k+1} = B^*(I - X_k)B > 0$ and $Y_{k+1} = B^*(I - Y_k)B > 0$ and
 hence $X = B^*(I - X)B$, equation (1) has a positive definite
 solution.

2) Theorem

For the Algorithm C, if there exist positive numbers α
 and β such that $0 < \alpha < \beta$ and the following two
 conditions are hold

(i) $B^*B < \alpha I$ and $B^*B < \beta I$,

(ii) $q = \|B\|^2 < 1$,

then $\|X_k - X\| < q^k (2\alpha - 1)$, $\|Y_k - X\| < q^k (2\beta - 1)$ and
 $\|X_k - X\| < \|Y_k - X\| < q^k (2\beta - 1)$, where X is a positive

definite solution of (1) and $X_k, Y_k, k = 0, 1, 2, \dots$ is defined in
 Algorithm C.

Proof. From Theorem 1, it follows that the sequence (8)
 is convergent to a positive definite solution X of (1). We
 compute the norms of the matrix $X_k - X$ and $Y_k - X$. We
 obtain

$$\|X_k - X\| = \|B^*B - B^*X_{k-1}B - B^*B + B^*XB\| = \|B^*(X - X_{k-1})B\|$$

$$\leq \|B\|^2 \|X - X_{k-1}\| \leq q \|X - X_{k-1}\| \leq \dots \leq q^k \|X_0 - X\| < q^k (2\alpha - 1).$$

Similarly, $\|Y_k - X\| \leq q^k \|Y_0 - X\| < q^k (2\beta - 1)$. Also, we have

$$\|X_k - X\| < \|Y_k - X\|.$$

$$\|X_k - X\| < \|Y_k - X\| < q^k (2\beta - 1).$$

3) Corollary

Suppose that (1) has a solution. If $q = \|B\|^2 < 1$, then
 $\{X_k\}$ and $\{Y_k\}$ converge to X with at least the linear
 convergence rate.

4) Theorem

If (1) has a positive definite Solution and after k
 iterative steps of the Algorithm C, we have

$$\|I - X_k^{-1}X_{k-1}\| < \varepsilon \text{ and } \|I - Y_k^{-1}Y_{k-1}\| < \varepsilon, \text{ then}$$

$$\|X_k + B^*X_kB - B^*B\| < \alpha\varepsilon\|B\|^2 \text{ and}$$

$$\|Y_k + B^*Y_kB - B^*B\| < \beta\varepsilon\|B\|^2.$$

$$\|X_k + B^*X_kB - B^*B\| < \|Y_k + B^*Y_kB - B^*B\|$$

$$< \beta\varepsilon\|B\|^2.$$

Where $X_k, Y_k, k = 0, 1, 2, \dots$ are the iterates generated by
 Algorithm C and $\varepsilon > 0$.

Proof

(i) Since,

$$X_k + B^*X_kB - B^*B = X_k + B^*X_kB - X_k - B^*X_{k-1}B$$

$$= B^*(X_k - X_{k-1})B.$$

Take the norms of both sides,

$$\|X_k + B^*X_kB - B^*B\| \leq \|B\|^2 \|X_k - X_{k-1}\|$$

$$\leq \|B\|^2 \|X_k\| \|I - X_k^{-1}X_{k-1}\| < \alpha\varepsilon\|B\|^2.$$

$$\text{Similarly, } \|Y_k + B^*Y_kB - B^*B\| < \beta\varepsilon\|B\|^2.$$

$$\text{(ii) From Theorem 1, we have } \|X_{k+1} - X_k\| < \|Y_{k+1} - Y_k\|.$$

Since

$$\|X_{k+1} - X_k\| = \|B^*B - B^*X_kB - X_k\| = \|X_k + B^*X_kB - B^*B\|$$

$$\|Y_{k+1} - Y_k\| = \|B^*B - B^*Y_kB - Y_k\| = \|Y_k + B^*Y_kB - B^*B\|,$$

$$\text{thus } \|X_k + B^*X_kB - B^*B\| < \|Y_k + B^*Y_kB - B^*B\| < \beta\varepsilon\|B\|^2.$$

V. NUMERICAL EXPERIMENTS

In this section the numerical experiments are used to
 display the flexibility of the methods. The solutions are
 computed for some different matrices A with different
 sizes n . For the following examples, practical stopping

criterion $\|X - X_k\| \leq 10^{-9}$ and obtains the maximal solution $X = X_{500}$.

A. Numerical experiments for the first method (Algorithm A)

In the following tables we denote

$q = \|A\|^2$, $\varepsilon_1(X) = \|X - X_k\|$, $\varepsilon_2(X) = \|X_k + A^* X_k A - I\|$, where X the solution which is obtained by the iterative method (Algorithm A)

I. Example

Let $\alpha = 10$ and

$$A = \frac{1}{100} \begin{pmatrix} 1 & 0.5542 & 0.6684 & 9.9700 & 0.2218 & 0.1120 \\ 0.5542 & 1 & 1.1270 & 3.2320 & 1.1440 & 1.1880 \\ 0.6684 & 1.1270 & 1 & 1.1570 & 2.3180 & 0.1430 \\ 9.9700 & 3.2320 & 1.1570 & 1 & 0.9990 & 0.8855 \\ 0.2218 & 1.1440 & 2.3180 & 0.9990 & 1 & 0.1287 \\ 0.1120 & 1.1880 & 0.1430 & 0.8855 & 0.1287 & 1 \end{pmatrix}$$

$\|A\| = 0.121701$, $q = 0.014811 < 1$, see Table I.

B. Numerical experiments for the second method (Algorithm B):

The following tables denotes

$q = \|A\|^2$, $\varepsilon_1(X) = \|X - X_k\|$, $\varepsilon_2(X) = \|Y - Y_k\|$, $\varepsilon_3(X) = \|X_k - Y_k\|$, $\varepsilon_4(X) = \|X_k + A^* X_k A - I\|$, $\varepsilon_5(X) = \|Y_k + A^* Y_k A - I\|$, X and

Y are the solutions which are obtained by the iterative method (Algorithm B).

II. Example

Let $\alpha = 2$, $\beta = 3$ and

$$A = \frac{1}{1000} \begin{pmatrix} -1.274 & -0.5755 & 2.384 & 4.118 & -0.1482 \\ -0.5755 & -3.221 & -0.9663 & 5.737 & 6.286 \\ -0.8774 & 6.286 & -0.9663 & 5.737 & -3.221 \\ 2.384 & -0.5755 & -0.1482 & 4.118 & 1.274 \\ 0.8774 & 5.737 & -0.9663 & -0.5755 & -2.384 \end{pmatrix}$$

$\|A\| = 0.0112205$, $q = 0.00012589 < 1$, see Table II.

C. Numerical experiments for the second method (Algorithm C)

The following tables denotes

$q = \|B\|^2$, $\varepsilon_1(X) = \|X - X_k\|$, $\varepsilon_2(X) = \|Y - Y_k\|$, $\varepsilon_3(X) = \|X_k - Y_k\|$, $\varepsilon_4(X) = \|X_k + B^* X_k B - B^* B\|$, $\varepsilon_5(X) = \|Y_k + B^* Y_k B - B^* B\|$, X and

Y are the solutions obtained by the iterative method (Algorithm C).

III. Example

Let $\alpha = 5$, $\beta = 7$ and

$$A = \begin{pmatrix} -12.74 & 5.755 \\ -5.755 & -32.21 \end{pmatrix},$$

$\|A\| = 32.9351$, $\|B\| = 0.0726747$ and $q = 0.00528 < 1$, see Table III.

VI. CONCLUSIONS

In this paper, the positive definite solution of the linear matrix equation $X + A^* X A = I$, which is a special case of

the symmetric matrix equations (2) for $\|A\| \neq 1$ was obtained. Two effective iterative methods for computing a positive definite solution of this equation were proposed. The first one is fixed point iteration method when $\|A\| < 1$ and the second one is two sided iteration method of the fixed point iteration method when $\|A\| < 1$ and $\|A\| > 1$. By Algorithm A, for initial matrix $X_0 = \alpha I$ and Algorithms B and C, for initial matrices $X_0 = \alpha I, Y_0 = \beta I$ satisfying the hypothesis of theorems (A.1) in chapter III, (B.1) and (C.1) in chapter IV, a positive definite solution X can be obtained in finite iteration, with at least the linear convergence rate. The given numerical examples show that the proposed iterative algorithms are efficient.

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TABLE I
EXAMPLE I

k	ϵ_1	q^k	$q^k(2\alpha - 1)$	ϵ_2
0	9.01459	1.	19.	9.11397
1	1.33516E-01	1.48111E-02	2.81411E-01	1.04486E-01
2	1.97752E-03	2.19369E-04	4.16801E-03	1.35647E-03
3	2.92893E-05	3.2491E-06	6.17329E-05	1.90955E-05
4	4.33808E-07	4.81228E-08	9.14333E-07	2.77617E-07
5	6.42517E-09	7.12752E-10	1.35423E-08	4.08379E-09
6	9.5164E-11	1.05567E-11	2.00577E-10	6.03296E-11

TABLE II
EXAMPLE II

k	ϵ_1	q^k	$q^k(2\alpha - 1)$	ϵ_4
0	1.00013	1.0	3.0	1.00025
1	1.25911E-04	1.25899E-04	3.77698E-04	1.25922E-04
2	1.39267E-08	1.58507E-08	4.7552E-08	1.39278E-08
3	1.07697E-12	1.99559E-12	5.98677E-12	1.07704E-12
k	ϵ_2	ϵ_3	$q^k(2\beta - 1)$	ϵ_5
0	2.00013	1.0	5.0	2.00038
1	2.5181E-04	1.25899E-04	6.29497E-04	2.51832E-04
2	2.78524E-08	1.39257E-08	7.92533E-08	2.78545E-08
3	2.15383E-12	1.07685E-12	9.97795E-12	2.15399E-12

TABLE III
EXAMPLE III

k	ϵ_1	q^k	$q^k(2\alpha - 1)$	ϵ_4
0	4.99918	1.0	9.0	5.02206
1	2.63485E-02	5.28161E-03	4.75344E-02	2.6672E-02
2	1.38995E-04	2.78954E-05	2.51058E-04	6.65329E-04
3	7.33231E-07	1.47332E-07	1.32599E-06	7.33359E-04
4	3.86797E-09	7.78151E-10	7.00336E-09	7.32986E-04
5	2.04045E-11	4.10989E-12	3.6989E-11	7.32988E-04
k	ϵ_2	ϵ_3	$q^k(2\beta - 1)$	ϵ_5
0	6.99918	2.0	13.0	7.03309
1	3.6899E-02	1.05505E-02	6.86609E-02	3.72743E-02
2	1.94651E-04	5.56563E-05	3.6264E-04	6.40508E-04
3	1.02683E-06	2.93601E-07	1.91532E-06	7.33508E-04
4	5.41679E-09	1.54881E-09	1.0116E-08	7.32985E-04
5	2.85749E-11	8.17037E-12	5.34285E-11	7.32988E-04

1) Date of modification: 14/03/2014

2) Brief description of the changes

Page	Column	Line	Change: From	Change: To
1	1	22	[16]	[15]
1	1	24	[2],[3],[8]- [11],[13]- [16],[18]- [21],[24].	[2],[3],[8]- [11],[13]- [15],[18]- [21],[24].
1	1	31	[7],[12],[17],[22],[23].	[7],[12],[16],[17],[22],[23].
2	1	10	<i>B. Algorithm</i>	<i>A. Algorithm</i>
3	1	7	<i>theorem B</i>	<i>theorem 1</i>
3	1	20	1) <i>Theorem</i>	1) <i>Theorem</i> (<i>Shifting</i>)
4	2	10	$B^* B < \alpha I, \beta I$	$B^* B < \alpha I$