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# Solving systems of nonlinear matrix equations involving Lipschitzian mappings

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## Abstract

In this study, both theoretical results and numerical methods are derived for solving different classes of systems of nonlinear matrix equations involving Lipschitzian mappings.

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## 1 Introduction

Fixed point theory is a very attractive subject, which has recently drawn much attention from the communities of physics, engineering, mathematics, etc. The Banach contraction principle [1] is one of the most important theorems in fixed point theory. It has applications in many diverse areas.

**Definition 1.1** Let  $M$  be a nonempty set and  $f: M \rightarrow M$  be a given mapping. We say that  $x^* \in M$  is a fixed point of  $f$  if  $fx^* = x^*$ .

**Theorem 1.1** (Banach contraction principle [1]). Let  $(M, d)$  be a complete metric space and  $f: M \rightarrow M$  be a contractive mapping, i.e., there exists  $\lambda \in [0, 1)$  such that for all  $x, y \in M$ ,

$$d(fx, fy) \leq \lambda d(x, y). \quad (1)$$

Then the mapping  $f$  has a unique fixed point  $x^* \in M$ . Moreover, for every  $x_0 \in M$ , the sequence  $(x_k)$  defined by:  $x_{k+1} = fx_k$  for all  $k = 0, 1, 2, \dots$  converges to  $x^*$ , and the error estimate is given by:

$$d(x_k, x^*) \leq \frac{\lambda^k}{1 - \lambda} d(x_0, x_1), \quad \text{for all } k = 0, 1, 2, \dots$$

Many generalizations of Banach contraction principle exists in the literature. For more details, we refer the reader to [2-4].

To apply the Banach fixed point theorem, the choice of the metric plays a crucial role. In this study, we use the Thompson metric introduced by Thompson [5] for the study of solutions to systems of nonlinear matrix equations involving contractive mappings.

We first review the Thompson metric on the open convex cone  $P(n)$  ( $n \geq 2$ ), the set of all  $n \times n$  Hermitian positive definite matrices. We endow  $P(n)$  with the Thompson

metric defined by:

$$d(A, B) = \max \{ \log M(A/B), \log M(B/A) \},$$

where  $M(A/B) = \inf \{ \lambda > 0 : A \leq \lambda B \} = \lambda^+(B^{-1/2}AB^{-1/2})$ , the maximal eigenvalue of  $B^{-1/2}AB^{-1/2}$ . Here,  $X \leq Y$  means that  $Y - X$  is positive semidefinite and  $X < Y$  means that  $Y - X$  is positive definite. Thompson [5] (cf. [6,7]) has proved that  $P(n)$  is a complete metric space with respect to the Thompson metric  $d$  and  $d(A, B) = \| \log(A^{-1/2}BA^{-1/2}) \|$ , where  $\| \cdot \|$  stands for the spectral norm. The Thompson metric exists on any open normal convex cones of real Banach spaces [5,6]; in particular, the open convex cone of positive definite operators of a Hilbert space. It is invariant under the matrix inversion and congruence transformations, that is,

$$d(A, B) = d(A^{-1}, B^{-1}) = d(MAM^*, MBM^*) \tag{2}$$

for any nonsingular matrix  $M$ . The other useful result is the nonpositive curvature property of the Thompson metric, that is,

$$d(X^r, Y^r) \leq r d(X, Y), \quad r \in [0, 1]. \tag{3}$$

By the invariant properties of the metric, we then have

$$d(MX^rM^*, MY^rM^*) \leq |r|d(X, Y), \quad r \in [-1, 1] \tag{4}$$

for any  $X, Y \in P(n)$  and nonsingular matrix  $M$ .

**Lemma 1.1** (see [8]). *For all  $A, B, C, D \in P(n)$ , we have*

$$d(A + B, C + D) \leq \max \{ d(A, C), d(B, D) \}.$$

*In particular,*

$$d(A + B, A + C) \leq d(B, C).$$

## 2 Main result

In the last few years, there has been a constantly increasing interest in developing the theory and numerical approaches for HPD (Hermitian positive definite) solutions to different classes of nonlinear matrix equations (see [8-21]). In this study, we consider the following problem: Find  $(X_1, X_2, \dots, X_m) \in (P(n))^m$  solution to the following system of nonlinear matrix equations:

$$X_i^{r_i} = Q_i + \sum_{j=1}^m \left( A_j^* F_{ij}(X_j) A_j \right)^{\alpha_{ij}}, \quad i = 1, 2, \dots, m, \tag{5}$$

where  $r_i \geq 1$ ,  $0 < |\alpha_{ij}| \leq 1$ ,  $Q_i \geq 0$ ,  $A_i$  are nonsingular matrices, and  $F_{ij}: P(n) \rightarrow P(n)$  are Lipschitzian mappings, that is,

$$\sup_{X, Y \in P(n), X \neq Y} \frac{d(F_{ij}(X), F_{ij}(Y))}{d(X, Y)} = k_{ij} < \infty. \tag{6}$$

If  $m = 1$  and  $\alpha_{11} = 1$ , then (5) reduces to find  $X \in P(n)$  solution to  $X^r = Q + A^*F(X)A$ . Such problem was studied by Liao et al. [15]. Now, we introduce the following definition.

**Definition 2.1** We say that Problem (5) is Banach admissible if the following hypothesis is satisfied:

$$\max_{1 \leq i \leq m} \left\{ \max_{1 \leq j \leq m} \{ |\alpha_{ij}| k_{ij} / r_i \} \right\} < 1.$$

Our main result is the following.

**Theorem 2.1** If Problem (5) is Banach admissible, then it has one and only one solution  $(X_1^*, X_2^*, \dots, X_m^*) \in (P(n))^m$ . Moreover, for any  $(X_1(0), X_2(0), \dots, X_m(0)) \in (P(n))^m$ , the sequences  $(X_i(k))_{k \geq 0}$ ,  $1 \leq i \leq m$ , defined by:

$$X_i(k+1) = \left( Q_i + \sum_{j=1}^m (A_j^* F_{ij}(X_j(k)) A_j)^{\alpha_{ij}} \right)^{1/r_i}, \tag{7}$$

converge respectively to  $X_1^*, X_2^*, \dots, X_m^*$ , and the error estimation is

$$\begin{aligned} & \max\{d(X_1(k), X_1^*), d(X_2(k), X_2^*), \dots, d(X_m(k), X_m^*)\} \\ & \leq \frac{q_m^k}{1 - q_m} \max\{d(X_1(1), X_1(0)), d(X_2(1), X_2(0)), \dots, d(X_m(1), X_m(0))\}, \end{aligned} \tag{8}$$

where

$$q_m = \max_{1 \leq i \leq m} \left\{ \max_{1 \leq j \leq m} \{ |\alpha_{ij}| k_{ij} / r_i \} \right\}.$$

**Proof.** Define the mapping  $G: (P(n))^m \rightarrow (P(n))^m$  by:

$$G(X_1, X_2, \dots, X_m) = (G_1(X_1, X_2, \dots, X_m), G_2(X_1, X_2, \dots, X_m), \dots, G_m(X_1, X_2, \dots, X_m)),$$

for all  $X = (X_1, X_2, \dots, X_m) \in (P(n))^m$ , where

$$G_i(X) = \left( Q_i + \sum_{j=1}^m (A_j^* F_{ij}(X_j) A_j)^{\alpha_{ij}} \right)^{1/r_i},$$

for all  $i = 1, 2, \dots, m$ . We endow  $(P(n))^m$  with the metric  $d_m$  defined by:

$$d_m((X_1, X_2, \dots, X_m), (Y_1, Y_2, \dots, Y_m)) = \max \{d(X_1, Y_1), d(X_2, Y_2), \dots, d(X_m, Y_m)\},$$

for all  $X = (X_1, X_2, \dots, X_m)$ ,  $Y = (Y_1, Y_2, \dots, Y_m) \in (P(n))^m$ . Obviously,  $((P(n))^m, d_m)$  is a complete metric space.

We claim that

$$d_m(G(X), G(Y)) \leq q_m d_m(X, Y), \quad \text{for all } X, Y \in (P(n))^m. \tag{9}$$

For all  $X, Y \in (P(n))^m$ , We have

$$d_m(G(X), G(Y)) = \max_{1 \leq i \leq m} \{d(G_i(X), G_i(Y))\}. \tag{10}$$

On the other hand, using the properties of the Thompson metric (see Section 1), for all  $i = 1, 2, \dots, m$ , we have

$$\begin{aligned}
 d(G_i(X), G_i(Y)) &= d\left(\left(Q_i + \sum_{j=1}^m (A_j^* F_{ij}(X_j) A_j)^{\alpha_{ij}}\right)^{1/r_i}, \left(Q_i + \sum_{j=1}^m (A_j^* F_{ij}(Y_j) A_j)^{\alpha_{ij}}\right)^{1/r_i}\right) \\
 &\leq \frac{1}{r_i} d\left(Q_i + \sum_{j=1}^m (A_j^* F_{ij}(X_j) A_j)^{\alpha_{ij}}, Q_i + \sum_{j=1}^m (A_j^* F_{ij}(Y_j) A_j)^{\alpha_{ij}}\right) \\
 &\leq \frac{1}{r_i} d\left(\sum_{j=1}^m (A_j^* F_{ij}(X_j) A_j)^{\alpha_{ij}}, \sum_{j=1}^m (A_j^* F_{ij}(Y_j) A_j)^{\alpha_{ij}}\right) \\
 &\leq \frac{1}{r_i} d\left((A_1^* F_{i1}(X_1) A_1)^{\alpha_{i1}} + \sum_{j=2}^m (A_j^* F_{ij}(X_j) A_j)^{\alpha_{ij}}, (A_1^* F_{i1}(Y_1) A_1)^{\alpha_{i1}} + \sum_{j=2}^m (A_j^* F_{ij}(Y_j) A_j)^{\alpha_{ij}}\right) \\
 &\leq \frac{1}{r_i} \max \left\{ d((A_1^* F_{i1}(X_1) A_1)^{\alpha_{i1}}, (A_1^* F_{i1}(Y_1) A_1)^{\alpha_{i1}}), d\left(\sum_{j=2}^m (A_j^* F_{ij}(X_j) A_j)^{\alpha_{ij}}, \sum_{j=2}^m (A_j^* F_{ij}(Y_j) A_j)^{\alpha_{ij}}\right) \right\} \\
 &\leq \dots \\
 &\leq \frac{1}{r_i} \max \{d((A_1^* F_{i1}(X_1) A_1)^{\alpha_{i1}}, (A_1^* F_{i1}(Y_1) A_1)^{\alpha_{i1}}), \dots, d((A_m^* F_{im}(X_m) A_m)^{\alpha_{im}}, (A_m^* F_{im}(Y_m) A_m)^{\alpha_{im}})\} \\
 &\leq \frac{1}{r_i} \max \{|\alpha_{i1}| d(A_1^* F_{i1}(X_1) A_1, A_1^* F_{i1}(Y_1) A_1), \dots, |\alpha_{im}| d(A_m^* F_{im}(X_m) A_m, A_m^* F_{im}(Y_m) A_m)\} \\
 &\leq \frac{1}{r_i} \max \{|\alpha_{i1}| d(F_{i1}(X_1), F_{i1}(Y_1)), \dots, |\alpha_{im}| d(F_{im}(X_m), F_{im}(Y_m))\} \\
 &\leq \frac{1}{r_i} \max \{|\alpha_{i1}| k_{i1} d(X_1, Y_1), \dots, |\alpha_{im}| k_{im} d(X_m, Y_m)\} \\
 &\leq \frac{\max_{1 \leq j \leq m} \{|\alpha_{ij}| k_{ij}\}}{r_i} \max \{d(X_1, Y_1), \dots, d(X_m, Y_m)\} \\
 &\leq \max_{1 \leq j \leq m} \{|\alpha_{ij}| k_{ij} / r_i\} d_m(X, Y).
 \end{aligned}$$

Thus, we proved that for all  $i = 1, 2, \dots, m$ , we have

$$d(G_i(X), G_i(Y)) \leq \max_{1 \leq j \leq m} \{|\alpha_{ij}| k_{ij} / r_i\} d_m(X, Y). \tag{11}$$

Now, (9) holds immediately from (10) and (11). Applying the Banach contraction principle (see Theorem 1.1) to the mapping  $G$ , we get the desired result.  $\square$

### 3 Examples and numerical results

#### 3.1 The matrix equation: $X = \left(\left((X^{1/2} + B_1)^{-1/2} + B_2\right)^{1/3} + B_3\right)^{1/2}$

We consider the problem: Find  $X \in P(n)$  solution to

$$X = \left(\left((X^{1/2} + B_1)^{-1/2} + B_2\right)^{1/3} + B_3\right)^{1/2}, \tag{12}$$

where  $B_i \geq 0$  for all  $i = 1, 2, 3$ .

Problem (12) is equivalent to: Find  $X_1 \in P(n)$  solution to

$$X_1^{r_1} = Q_1 + (A_1^* F_{11}(X_1) A_1)^{\alpha_{11}}, \tag{13}$$

where  $r_1 = 2$ ,  $Q_1 = B_3$ ,  $A_1 = I_n$  (the identity matrix),  $\alpha_{11} = 1/3$  and  $F_{11} : P(n) \rightarrow P(n)$  is given by:

$$F_{11}(X) = (X^{1/2} + B_1)^{-1/2} + B_2.$$

**Proposition 3.1**  $F_{11}$  is a Lipschitzian mapping with  $k_{11} \leq 1/4$ .

**Proof.** Using the properties of the Thompson metric, for all  $X, Y \in P(n)$ , we have

$$\begin{aligned} d(F_{11}(X), F_{11}(Y)) &= d((X^{1/2} + B_1)^{-1/2} + B_2, (Y^{1/2} + B_1)^{-1/2} + B_2) \\ &\leq d((X^{1/2} + B_1)^{-1/2}, (Y^{1/2} + B_1)^{-1/2}) \\ &\leq \frac{1}{2} d(X^{1/2} + B_1, Y^{1/2} + B_1) \\ &\leq \frac{1}{2} d(X^{1/2}, Y^{1/2}) \leq \frac{1}{4} d(X, Y). \end{aligned}$$

Thus, we have  $k_{11} \leq 1/4$ .  $\square$

**Proposition 3.2** *Problem (13) is Banach admissible.*

**Proof.** We have

$$\frac{|\alpha_{11}|k_{11}}{r_1} \leq \frac{\frac{1}{3} \frac{1}{4}}{2} = \frac{1}{24} < 1.$$

This implies that Problem (13) is Banach admissible.  $\square$

**Theorem 3.1** *Problem (13) has one and only one solution  $X_1^* \in P(n)$ . Moreover, for any  $X_1(0) \in P(n)$ , the sequence  $(X_1(k))_{k \geq 0}$  defined by:*

$$X_1(k+1) = \left( \left( (X_1(k)^{1/2} + B_1)^{-1/2} + B_2 \right)^{1/3} + B_3 \right)^{1/2}, \tag{14}$$

converges to  $X_1^*$ , and the error estimation is

$$d(X_1(k), X_1^*) \leq \frac{q_1^k}{1 - q_1} d(X_1(1), X_1(0)), \tag{15}$$

where  $q_1 = 1/4$ .

**Proof.** Follows from Propositions 3.1, 3.2 and Theorem 2.1.  $\square$

Now, we give a numerical example to illustrate our result given by Theorem 3.1.

We consider the  $5 \times 5$  positive matrices  $B_1, B_2$ , and  $B_3$  given by:

$$B_1 = \begin{pmatrix} 1.0000 & 0.5000 & 0.3333 & 0.2500 & 0 \\ 0.5000 & 1.0000 & 0.6667 & 0.5000 & 0 \\ 0.3333 & 0.6667 & 1.0000 & 0.7500 & 0 \\ 0.2500 & 0.5000 & 0.7500 & 1.0000 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}, \quad B_2 = \begin{pmatrix} 1.4236 & 1.3472 & 1.1875 & 1.0000 & 0 \\ 1.3472 & 1.9444 & 1.8750 & 1.6250 & 0 \\ 1.1875 & 1.8750 & 2.1181 & 1.9167 & 0 \\ 1.0000 & 1.6250 & 1.9167 & 1.8750 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

and

$$B_3 = \begin{pmatrix} 2.7431 & 3.3507 & 3.3102 & 2.9201 & 0 \\ 3.3507 & 4.6806 & 4.8391 & 4.3403 & 0 \\ 3.3102 & 4.8391 & 5.2014 & 4.7396 & 0 \\ 2.9201 & 4.3403 & 4.7396 & 4.3750 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}.$$

We use the iterative algorithm (14) to solve (12) for different values of  $X_1(0)$ :

$$X_1(0) = M_1 = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 3 & 0 & 0 \\ 0 & 0 & 0 & 4 & 0 \\ 0 & 0 & 0 & 0 & 5 \end{pmatrix}, \quad X_1(0) = M_2 = \begin{pmatrix} 0.02 & 0.01 & 0 & 0 & 0 \\ 0.01 & 0.02 & 0.01 & 0 & 0 \\ 0 & 0.01 & 0.02 & 0.01 & 0 \\ 0 & 0 & 0.01 & 0.02 & 0.01 \\ 0 & 0 & 0 & 0.01 & 0.02 \end{pmatrix}$$

and

$$X_1(0) = M_3 = \begin{pmatrix} 30 & 15 & 10 & 7.5 & 6 \\ 15 & 30 & 20 & 15 & 12 \\ 10 & 20 & 30 & 22.5 & 18 \\ 7.5 & 15 & 22.5 & 30 & 24 \\ 6 & 12 & 18 & 24 & 30 \end{pmatrix}.$$

For  $X_1(0) = M_1$ , after 9 iterations, we get the unique positive definite solution

$$X_1(9) = \begin{pmatrix} 1.6819 & 0.69442 & 0.61478 & 0.51591 & 0 \\ 0.69442 & 1.9552 & 0.96059 & 0.84385 & 0 \\ 0.61478 & 0.96059 & 2.0567 & 0.9785 & 0 \\ 0.51591 & 0.84385 & 0.9785 & 1.9227 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

and its residual error

$$R(X_1(9)) = \left\| X_1(9) - \left( \left( (X_1(9))^{1/2} + B_1 \right)^{-1/2} + B_2 \right)^{1/3} + B_3 \right\|^{1/2} = 6.346 \times 10^{-13}.$$

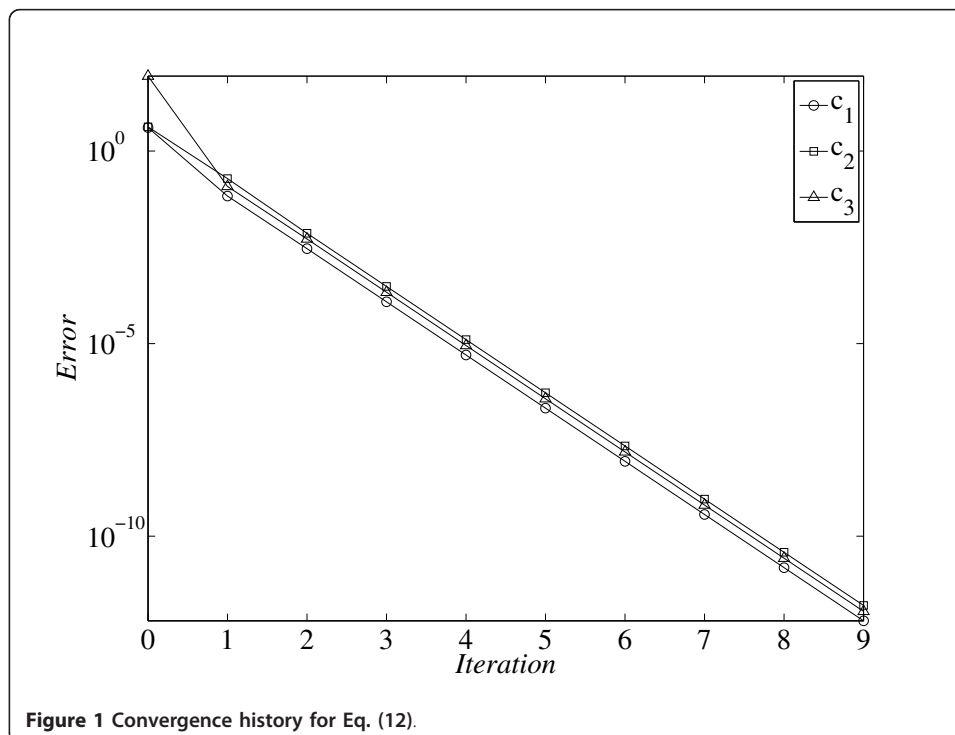
For  $X_1(0) = M_2$ , after 9 iterations, the residual error

$$R(X_1(9)) = 1.5884 \times 10^{-12}.$$

For  $X_1(0) = M_3$ , after 9 iterations, the residual error

$$R(X_1(9)) = 1.1123 \times 10^{-12}.$$

The convergence history of the algorithm for different values of  $X_1(0)$  is given by Figure 1, where  $c_1$  corresponds to  $X_1(0) = M_1$ ,  $c_2$  corresponds to  $X_1(0) = M_2$ , and  $c_3$  corresponds to  $X_1(0) = M_3$ .



### 3.2 System of three nonlinear matrix equations

We consider the problem: Find  $(X_1, X_2, X_3) \in (P(n))^3$  solution to

$$\begin{cases} X_1 = I_n + A_1^*(X_1^{1/3} + B_1)^{1/2}A_1 + A_2^*(X_2^{1/4} + B_2)^{1/3}A_2 + A_3^*(X_3^{1/5} + B_3)^{1/4}A_3, \\ X_2 = I_n + A_1^*(X_1^{1/5} + B_1)^{1/4}A_1 + A_2^*(X_2^{1/3} + B_2)^{1/2}A_2 + A_3^*(X_3^{1/4} + B_3)^{1/3}A_3, \\ X_3 = I_n + A_1^*(X_1^{1/4} + B_1)^{1/3}A_1 + A_2^*(X_2^{1/5} + B_2)^{1/4}A_2 + A_3^*(X_3^{1/3} + B_3)^{1/2}A_3, \end{cases} \quad (16)$$

where  $A_i$  are  $n \times n$  singular matrices.

Problem (16) is equivalent to: Find  $(X_1, X_2, X_3) \in (P(n))^3$  solution to

$$X_i^{r_i} = Q_i + \sum_{j=1}^3 (A_j^* F_{ij}(X_j) A_j)^{\alpha_{ij}}, \quad i = 1, 2, 3, \quad (17)$$

where  $r_1 = r_2 = r_3 = 1$ ,  $Q_1 = Q_2 = Q_3 = I_n$  and for all  $i, j \in \{1, 2, 3\}$ ,  $\alpha_{ij} = 1$ ,

$$F_{ij}(X_j) = (X_j^{\theta_{ij}} + B_j)^{\gamma_{ij}}, \quad \theta = (\theta_{ij}) = \begin{pmatrix} 1/3 & 1/4 & 1/5 \\ 1/5 & 1/3 & 1/4 \\ 1/4 & 1/5 & 1/3 \end{pmatrix}, \quad \gamma = (\gamma_{ij}) = \begin{pmatrix} 1/2 & 1/3 & 1/4 \\ 1/4 & 1/2 & 1/3 \\ 1/3 & 1/4 & 1/2 \end{pmatrix}.$$

**Proposition 3.3** For all  $i, j \in \{1, 2, 3\}$ ,  $F_{ij}: P(n) \rightarrow P(n)$  is a Lipschitzian mapping with  $k_{ij} \leq \gamma_{ij}\theta_{ij}$ .

**Proof.** For all  $X, Y \in P(n)$ , since  $\theta_{ij}, \gamma_{ij} \in (0, 1)$ , we have

$$\begin{aligned} d(F_{ij}(X), F_{ij}(Y)) &= d((X^{\theta_{ij}} + B_j)^{\gamma_{ij}}, (Y^{\theta_{ij}} + B_j)^{\gamma_{ij}}) \\ &\leq \gamma_{ij}d(X^{\theta_{ij}} + B_j, Y^{\theta_{ij}} + B_j) \\ &\leq \gamma_{ij}d(X^{\theta_{ij}}, Y^{\theta_{ij}}) \\ &\leq \gamma_{ij}\theta_{ij}d(X, Y). \end{aligned}$$

Then,  $F_{ij}$  is a Lipschitzian mapping with  $k_{ij} \leq \gamma_{ij}\theta_{ij}$ .  $\square$

**Proposition 3.4** Problem (17) is Banach admissible.

**Proof.** We have

$$\begin{aligned} \max_{1 \leq i \leq 3} \left\{ \max_{1 \leq j \leq 3} \{|\alpha_{ij}|k_{ij}/r_i\} \right\} &= \max_{1 \leq i, j \leq 3} k_{ij} \\ &\leq \max_{1 \leq i, j \leq 3} \gamma_{ij}\theta_{ij} \\ &= 1/6 < 1. \end{aligned}$$

This implies that Problem (17) is Banach admissible.  $\square$

**Theorem 3.2** Problem (16) has one and only one solution  $(X_1^*, X_2^*, X_3^*) \in (P(n))^3$ . Moreover, for any  $(X_1(0), X_2(0), X_3(0)) \in (P(n))^3$ , the sequences  $(X_i(k))_{k \geq 0}$ ,  $1 \leq i \leq 3$ , defined by:

$$X_i(k+1) = I_n + \sum_{j=1}^3 A_j^* F_{ij}(X_j(k)) A_j, \quad (18)$$

converge respectively to  $X_1^*, X_2^*, X_3^*$ , and the error estimation is

$$\begin{aligned} &\max\{d(X_1(k), X_1^*), d(X_2(k), X_2^*), d(X_3(k), X_3^*)\} \\ &\leq \frac{q_3^k}{1 - q_3} \max\{d(X_1(1), X_1(0)), d(X_2(1), X_2(0)), d(X_3(1), X_3(0))\}, \end{aligned} \quad (19)$$

where  $q_3 = 1/6$ .

**Proof.** Follows from Propositions 3.3, 3.4 and Theorem 2.1.  $\square$

Now, we give a numerical example to illustrate our obtained result given by Theorem 3.2.

We consider the  $3 \times 3$  positive matrices  $B_1, B_2$  and  $B_3$  given by:

$$B_1 = \begin{pmatrix} 1. & 0.5 & 0 \\ 0.5 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \quad B_2 = \begin{pmatrix} 1.25 & 1 & 0 \\ 1 & 1.25 & 0 \\ 0 & 0 & 0 \end{pmatrix} \quad \text{and} \quad B_3 = \begin{pmatrix} 1.75 & 1.625 & 0 \\ 1.625 & 1.75 & 0 \\ 0 & 0 & 0 \end{pmatrix}.$$

We consider the  $3 \times 3$  nonsingular matrices  $A_1, A_2$  and  $A_3$  given by:

$$A_1 = \begin{pmatrix} 0.3107 & -0.5972 & 0.7395 \\ 0.9505 & 0.1952 & -0.2417 \\ 0 & -0.7780 & -0.6282 \end{pmatrix}, \quad A_2 = \begin{pmatrix} 1.5 & -2 & 0.5 \\ 0.5 & 0 & -0.5 \\ -0.5 & 2 & -1.5 \end{pmatrix}$$

and

$$A_3 = \begin{pmatrix} -1 & -1 & 1 \\ 1 & -1 & 1 \\ -1 & -1 & -1 \end{pmatrix}.$$

We use the iterative algorithm (18) to solve Problem (16) for different values of  $(X_1(0), X_2(0), X_3(0))$ :

$$X_1(0) = X_2(0) = X_3(0) = M_1 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix},$$

$$X_1(0) = X_2(0) = X_3(0) = M_2 = \begin{pmatrix} 0.02 & 0.01 & 0 \\ 0.01 & 0.02 & 0.01 \\ 0 & 0.01 & 0.02 \end{pmatrix}$$

and

$$X_1(0) = X_2(0) = X_3(0) = M_3 = \begin{pmatrix} 30 & 15 & 10 \\ 15 & 30 & 20 \\ 10 & 20 & 30 \end{pmatrix}.$$

The error at the iteration  $k$  is given by:

$$R(X_1(k), X_2(k), X_3(k)) = \max_{1 \leq i \leq 3} \left\| X_i(k) - I_3 - \sum_{j=1}^3 A_j^* F_{ij}(X_j(k)) A_j \right\|.$$

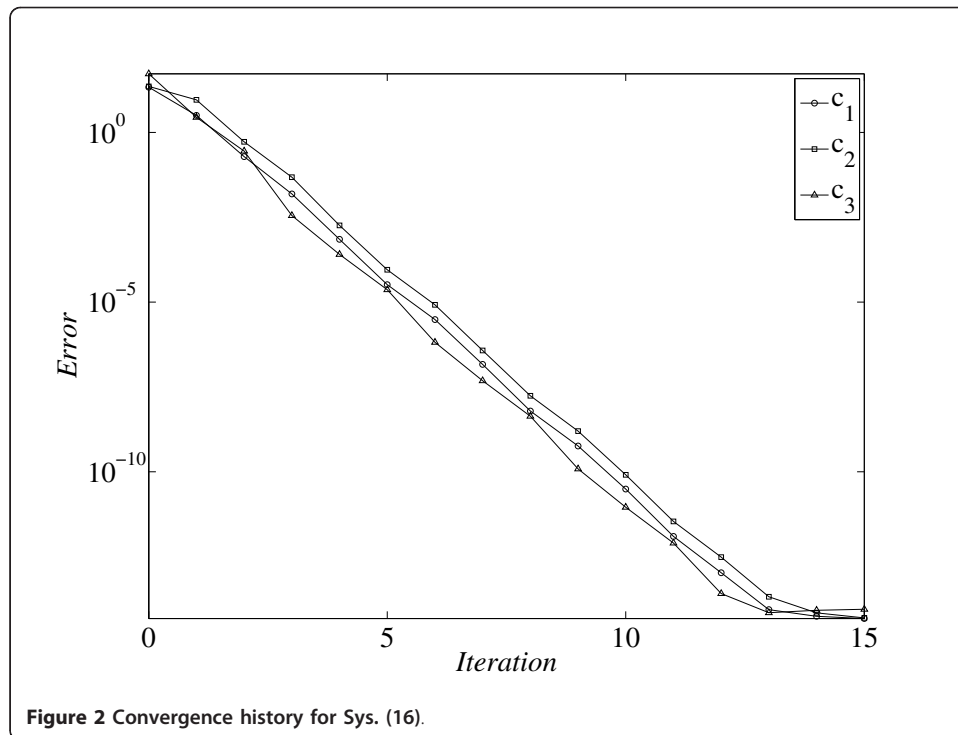
For  $X_1(0) = X_2(0) = X_3(0) = M_1$ , after 15 iterations, we obtain

$$X_1(15) = \begin{pmatrix} 10.565 & -4.4081 & 2.7937 \\ -4.4081 & 16.883 & -6.6118 \\ 2.7937 & -6.6118 & 9.7152 \end{pmatrix}, \quad X_2(15) = \begin{pmatrix} 11.512 & -5.8429 & 3.1922 \\ -5.8429 & 19.485 & -7.9308 \\ 3.1922 & -7.9308 & 10.68 \end{pmatrix}$$

and

$$X_3(15) = \begin{pmatrix} 11.235 & -3.5241 & 3.2712 \\ -3.5241 & 17.839 & -7.8035 \\ 3.2712 & -7.8035 & 11.618 \end{pmatrix}.$$





The residual error is given by:

$$R(X_1(15), X_2(15), X_3(15)) = 4.722 \times 10^{-15}.$$

For  $X_1(0) = X_2(0) = X_3(0) = M_2$ , after 15 iterations, the residual error is given by:

$$R(X_1(15), X_2(15), X_3(15)) = 4.911 \times 10^{-15}.$$

For  $X_1(0) = X_2(0) = X_3(0) = M_3$ , after 15 iterations, the residual error is given by:

$$R(X_1(15), X_2(15), X_3(15)) = 8.869 \times 10^{-15}.$$

The convergence history of the algorithm for different values of  $X_1(0)$ ,  $X_2(0)$ , and  $X_3(0)$  is given by Figure 2, where  $c_1$  corresponds to  $X_1(0) = X_2(0) = X_3(0) = M_1$ ,  $c_2$  corresponds to  $X_1(0) = X_2(0) = X_3(0) = M_2$  and  $c_3$  corresponds to  $X_1(0) = X_2(0) = X_3(0) = M_3$ .

#### Authors' contributions

All authors contributed equally and significantly in writing this paper. All authors read and approved the final manuscript.

#### Competing interests

The authors declare that they have no competing interests.

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