

## Improving the Modified Gause – Seidel Method for M - Matrices

Nasser Mikaeilvand<sup>1</sup> and Zahra Lorkojori<sup>2</sup>

<sup>1</sup>Department of Mathematics, Ardabil branch, Islamic Azad University, Ardabil, Iran.

<sup>2</sup>Young Researchers Club, Ardabil branch, Islamic Azad University, Ardabil, Iran.

**Corresponding author:** Nasser Mikaeilvand, email: Mikaeilvand@IauArdabil.ac.ir

**Abstract:** In 1994, M. Usui et al. have reported the modified Gauss-Seidel method with a preconditioner  $(I + U)$ . The preconditioning effect is not observed on the  $n$ -th row. In this paper, to deal with this drawback, we propose a new preconditioner. In addition, the convergence and comparison theorems of the proposed method are established.

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### 1- Introduction:

We consider the following linear system:

$$AX = b, \quad (1)$$

where  $A \in R^{n \times n}$ ,  $b \in R^n$  are given and  $x \in R^n$  is unknown. For simplicity, let  $A = I - L - U$ , where  $I$  is the identity matrix,  $L$  and  $U$  are strictly lower and strictly upper triangular matrices, respectively.

Now, consider a preconditioned system of (1):

$$PAX = pb, \quad (2)$$

where  $P$  is a non-singular matrix. To effectively solve the preconditioned linear system (2), a variety of preconditioners have been proposed by several authors [1 – 8,11] and the references therein. The preconditioning effect is not observed on the last row of matrix  $A$ . For example, the preconditioner  $P_{U_1} = I + U$  In [11] where  $U$  is a strictly upper triangular part of  $-A$ .

In 2009, Zheng et al. [4] proposed the following two preconditioners:

$$P_{max} = I + S_{max} + R_{max}$$

and

$$P_R = I + S_{max} + R$$

where

$$S_{max} = \begin{cases} -a_{i,k_i} & i=1, \dots, n-1, j>i; \\ 0, & \text{Other Wise,} \end{cases}$$

$$K_i = \min\{j | \max_i |a_{i,j}| < n\}$$

and

$$(R_{max})_{i,j} = \begin{cases} -a_{n,k_n} & i = n, j = K_n \\ 0, & \text{Other Wise} \end{cases}$$

$$\text{with } K_n = \min\{j | |a_{n,j}| = \max\{|a_{n,l}|, l = 1, \dots, n-1\}\}$$

and

$$(R)_{i,j} = \begin{cases} -a_{i,j} & i = n, 1 \leq j \leq n-1, \\ 0, & \text{Other Wise} \end{cases}$$

The comparison result between the preconditioners  $P_{max}$  with  $P_R$  [4] shows that the preconditioner  $P_R$  is better than  $P_{max}$  for solving the preconditioned linear system (2).

In this paper, we propose the following a preconditioner:

(3)

(4)

$$P_U = (I + U + R)$$

$$= \begin{pmatrix} 1 & -a_{12} - a_{13} & \cdots & -a_{1n} \\ 0 & 1 & -a_{23} & -a_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & -a_{n-1n} \\ -a_{n1} - a_{n2} & \cdots & -a_{nn-1} & 1 \end{pmatrix}$$

Then  $AU$  can be written as follows:

$$AU = (I + U + R)A$$

$$= I - L - U + U - UL - U^2 + R - RL - RU = M_U - N_U,$$

where

$$M_U = (I - D - L - E + R - \hat{D} - \hat{E}), N_U = F + U^2$$

and  $D$ ,  $E$  and  $F$  are the diagonal, strictly lower triangular and strictly upper triangular parts of  $UL$ , while  $\hat{D}$  and  $\hat{E}$  are the diagonal, strictly lower triangular parts of  $R(L + U)$ , respectively. If  $M_U$  is nonsingular, the MGS iterative matrix is  $T_U = M_U^{-1}N_U$ .

## 2- Preliminaries:

In this section, we present some notation, definitions and lemmas.

For  $A = (a_{ij}), B = (b_{ij}) \in R^{n \times n}$ , we write  $A \geq B$  if  $a_{ij} \geq b_{ij}$  holds for all  $i, j = 1, 2, \dots, n$ . Calling  $A$  nonnegative if  $A \geq 0$  ( $a_{ij} \geq 0; i, j = 1, 2, \dots, n$ ).  $\rho(\cdot)$  denotes the spectral radius of a matrix.

**Definition 2.1.** A matrix  $A$  is a L-matrix if  $a_{ij} > 0; i = j = 1, \dots, n$  and  $a_{ij} \leq 0$  for all  $i, j = 1, 2, \dots, n; i \neq j$ . A nonsingular L-matrix  $A$  is a nonsingular M-matrix if  $A^{-1} \geq 0$ .

**Definition 2.2.** Let  $A$  be a real matrix. Then

$$A = M - N$$

is called a splitting of  $A$  if  $M$  is a nonsingular matrix. The splitting is called

- (a) regular if  $M^{-1} \geq 0$  and  $N \geq 0$ ;
- (b) weak regular if  $M^{-1} \geq 0$  and  $M^{-1}N \geq 0$ ;
- (c) nonnegative if  $M^{-1}N \geq 0$ ;

(d) M-splitting if  $M$  is a nonsingular M-matrix and  $N \geq 0$ .

**Lemma 2.1** ([14]). Let  $A \in R^{n \times n}$  be nonnegative  $n \times n$  matrix. Then

- (a)  $A$  has a positive real eigenvalue equal to its spectral radius  $\rho(A)$ ;
- (b) for  $(A)$ , there corresponds an eigenvector  $X > 0$ ;
- (c)  $\rho(A)$  is a simple eigenvalue of  $A$ ;
- (d)  $\rho(A)$  increases when any entry of  $A$  increases.

**Definition 2.3.** We call  $A = M - N$  the Gauss-Seidel splitting of  $A$ , if  $M = (I - L)$  is nonsingular and  $N =$ . In addition, the splitting is called

- (a) Gauss-Seidel convergent if  $(M^{-1}N) < 1$ ;
- (b) Gauss-Seidel regular if  $M^{-1} = (1 - L)^{-1} \geq 0$  and  $N = U \geq 0$ .

**Lemma 2.2** ([17]).  $A = M - N$  be an M-splitting of  $A$ . Then  $\rho(M^{-1}N) < 1$  if and only if  $A$  is a nonsingular M-matrix.

**Lemma 2.3** ([15]). Let  $A$  and  $B$  be  $n \times n$  matrices. Then  $AB$  and  $BA$  have the same eigenvalues, counting multiplicity.

**Lemma 2.4** ([10]). Let  $A$  be a nonsingular M-matrix, and let

$A = M_1 - N_1 = M_2 - N_2$  be two convergent splitting, the first one weak regular and the second one regular. If  $M_1^{-1} \geq M_2^{-1}$ , then

$$\rho(M_1^{-1}N_1) \leq \rho(M_2^{-1}N_2) < 1.$$

## 3. Comparison Theorems

In this section, we compare such MGS method with the classical Gauss-Seidel method and the MGS method with the preconditioner  $P_{U_1} = I + U$  ([11]), respectively.

To prove the theorems, we need some results.

We firstly prove that  $A_{U_1} = M_{U_1} - N_{U_1}$  and  $A_U = M_U - N_U$  are both regular and Gauss-Seidel convergent splitting.

For the preconditioner  $P_{U_1} = I + U$  the preconditioned matrix  $A_{U_1} = (I + U)A$  can be written as

$$A_{U_1} = M_{U_1} + N_{U_1} = (I - D - L - E) - (F + U^2).$$

In which  $D$ ,  $E$  and  $F$  are defined as in section 1. Hence, if  $\sum_{k=i+1}^n a_{i,k} a_{k,i} \neq 1$  ( $i = 1, 2, \dots, n-1$ ), Then, the MGS iterative matrix  $T_{U_1}$  for  $A_{U_1}$  can be defined by

$$T_{U_1} = M_{U_1}^{-1} N_{U_1} = (I - D - L - E)^{-1} (F + U^2)$$

as  $(I - D - L - E)^{-1}$  exists. There is the following result:

**Lemma 3.1.** Let  $A = I - L - U$  be a nonsingular M-matrix. Assume that  $0 \leq \sum_{k=i+1}^n a_{i,k} a_{k,i} < 1$ ,  $1 \leq i \leq n-1$ . Then  $A_{U_1} = M_{U_1} - N_{U_1}$  is regular and Gauss-Seidel convergent.

*Proof.* We observe that when  $0 \leq \sum_{k=i+1}^n a_{i,k} a_{k,i} < 1$ ,  $1 \leq i \leq n-1$ , the diagonal elements of  $A_{U_1}$  are positive and  $M_{U_1}^{-1}$  exists. It is known that (see ([18])) an L-matrix  $A$  is a nonsingular M-matrix if and only if there exists a positive vector  $y$  such that  $Ay > 0$ . By taking such  $y$ , the fact that  $I + U \geq 0$  implies  $A_{U_1}y = (I + U)Ay > 0$ . Consequently, the L-matrix  $A_{U_1}$  is a nonsingular M-matrix which means  $A_{U_1}^{-1} \geq 0$ . Since  $0 \leq \sum_{k=i+1}^n a_{i,k} a_{k,i} < 1$  we have  $(I - D)^{-1} \geq I$ .

As strictly lower triangular matrix  $L + E$  has nonnegative elements, by Neumann's series, the following inequality holds:

$$M_{U_1}^{-1} = [I + (I - D)^{-1}(L + E) + \{(I - D)^{-1}(L + E)\}^2 + \dots$$

$$+ \{(I - D)^{-1}(L + E)\}^{n-1}](I - D)^{-1} \geq 0$$

On the other hand, it is easy to see that  $N_{U_1} = F + U^2 \geq 0$ . Thus,

$A_{U_1} = M_{U_1} - N_{U_1}$  is a regular and Gauss-Seidel convergent splitting by Definition 2.3 And Lemma 2.2. ■

**Theorem 3.2.** Let  $A$  be a nonsingular M-matrix, assume that  $0 \leq \sum_{k=i+1}^n a_{i,k} a_{k,i} < 1$ ,  $1 \leq i \leq n-1$  and  $0 \leq \sum_{k=1}^{n-1} a_{n,k} a_{k,n} < 1$ , then  $A_U = M_U - N_U$  is regular and Gauss-Seidel convergent splitting.

*Proof.* We observe that when  $0 \leq \sum_{k=i+1}^n a_{i,k} a_{k,i} < 1$ ,  $1 \leq i \leq n-1$  and  $0 \leq \sum_{k=1}^{n-1} a_{n,k} a_{k,n} < 1$ , the

diagonal elements of  $A_U$  are positive and  $M_U^{-1}$  exists. Similar to the proof of Theorem 3.1, We can show that

$A_U = (I + U + R)A$  is a nonsingular M-matrix when  $A$  is a nonsingular M-matrix. Thus,  $A_U^{-1} \geq 0$ . When  $0 \leq \sum_{k=i+1}^n a_{i,k} a_{k,i} < 1$ ,  $1 \leq i \leq n-1$  and  $0 \leq \sum_{k=1}^{n-1} a_{n,k} a_{k,n} < 1$ , we have  $D + D \hat{<} I$ , so that  $(I - D - \hat{D}) \geq 0$ . Hence,

$$\begin{aligned} M_U^{-1} &= [(I - D - \hat{D}) - (L - R + E + \hat{E})]^{-1} \\ &= [I - (I - D - \hat{D})^{-1} (L - R + E + \hat{E})]^{-1} (I - D - \hat{D})^{-1} \\ &= \{I + (I - D - \hat{D})^{-1} (L - R + E + \hat{E}) + \\ &\quad [(I - D - \hat{D})^{-1} (L - R + E + \hat{E})]^2 + \dots \\ &\quad + [I - (I - D - \hat{D})^{-1} (L - R + E + \hat{E})]^{n-1}\} (I - D - \hat{D})^{-1} \geq 0 \end{aligned}$$

It is easy to see that  $N_U = F + U^2 \geq 0$ .

Therefore,  $A_U = M_U - N_U$  is a regular and Gauss-Seidel convergent splitting by Definition 2.3 And Lemma 2.2. ■

**Theorem 3.3.** Let  $A$  be a nonsingular M-matrix. Then under the assumptions of Theorem 3.2, the following inequality holds:

$$\rho(T_U) \leq \rho(T) < 1$$

where,  $T = (I - L)^{-1} U$  is the iterative matrix of the classical Gauss-Seidel method for  $A = I - L - U$ .

*Proof.* Since  $A$  is a nonsingular M-matrix, the classic Gauss-Seidel splitting  $A = (I - L) - U$  of  $A$  is clearly regular and convergent.

For  $M_U = I - D - L - E + R - \hat{D} - \hat{E}$  and  $N_U = F + U^2$  by Theorem 3.2 we know that  $A_U = M_U - N_U$  is a Gauss-Seidel convergent splitting.

To compare  $\rho(T_U)$  with  $\rho(T)$ , we have

$$A = (I + U + R)^{-1} M_U - (I + U + R)^{-1} N_U.$$

If we take  $M_1 = (I + U + R)^{-1}M_U$  and  $N_1 = (I + U + R)^{-1}N_U$ , then  $\rho(M_1^{-1}N_1) < 1$  since  $M_U^{-1}N_U = M_1^{-1}N_1$ . Also, we have

$$\begin{aligned} M_1^{-1} &= M_U^{-1}(I + U + R) \\ &= (I - D - L - E + R - \hat{D} - \hat{E})^{-1}(I + U + R) \\ &\geq (I - D - L - E + R - \hat{D} - \hat{E})^{-1} \\ &= [I - (I - D - \hat{D})^{-1}(L - R + E + \hat{E})]^{-1}(I - D - \hat{D})^{-1} \\ &\geq [I - (I - D - \hat{D})^{-1}(L - R + E + \hat{E})]^{-1} \\ &\geq (I - L)^{-1}, \end{aligned}$$

If follows from [Lemma 2.4](#) that  $\rho(M_1^{-1}N_1) \leq \rho(M^{-1}N) < 1$ . Hence,

$$\rho(M_U^{-1}N_U) \leq \rho(M^{-1}N) < 1, \text{ i.e., } \rho(T_U) \leq \rho(T) < 1.$$

Next, we give a comparison theorem between the MGS method with the preconditioners  $P_U$  and  $P_{U_1}$ , respectively.

**Theorem 3.4.** Let  $A$  be a nonsingular  $M$ -matrix. Then under the assumptions of [Theorem 3.2](#) and  $a_{nj} \sum_{k=1}^{n-1} a_{n,k} a_{k,n} \leq \sum_{k=1}^{n-1} a_{n,k} a_{k,j}$ ,  $1 \leq j \leq n-1$ , we have

$$\rho(T_U) \leq \rho(T_{U_1}) < 1$$

**Proof.** For the matrices  $M_{U_1}, M_U, N_{U_1}$  and  $N_U$  in the splitting of matrices  $P_{U_1}A = M_{U_1} - N_{U_1}$  and  $P_UA = M_U - N_U$ , they can be expressed in the partitioned forms as follows:

$$M_{U_1} = I - D - L - E = \begin{pmatrix} \hat{M} & \hat{q} \\ u^T & 1 \end{pmatrix},$$

$$M_U = M_{U_1} + RA = \begin{pmatrix} \hat{M} & 0 \\ v^T & v_n \end{pmatrix},$$

$$N_U = N_{U_1} = \begin{pmatrix} \hat{N} & w \\ 0 & 0 \end{pmatrix},$$

where

$$\begin{aligned} \hat{M} &= (\hat{m}_{ij}) \\ \hat{m}_{ij} &= \begin{cases} 0, & 1 \leq i < j \leq n-1, \\ 1 - \sum_{k=i+1}^n a_{i,k} a_{k,i}, & i = j, \\ a_{i,j} - \sum_{k=i+1}^n a_{i,k} a_{k,j}, & j < i \leq n-1, \end{cases} \end{aligned}$$

$$u^T = (a_{n,1}, \dots, a_{n,n-1}),$$

$$v^T = (v_1, \dots, v_{n-1})$$

$$v_j = a_{n,j} - \sum_{k=1}^{n-1} a_{n,k} a_{k,j} \quad (1 \leq j \leq n-1)$$

$$v_n = 1 - \sum_{k=1}^{n-1} a_{n,k} a_{k,n}$$

$$W = (\omega_1, \dots, \omega_{n-1})^T$$

$$\omega_i = -a_{i,n} + \sum_{k=i+1}^n a_{i,k} a_{k,n} \quad (1 \leq i \leq n-1)$$

and  $\hat{N} \geq 0$  is an  $(n-1) \times (n-1)$  strictly upper triangular matrix.

Direct computation yields

$$M_{U_1}^{-1} = \begin{pmatrix} \hat{M}^{-1} & 0 \\ -u^T \hat{M}^{-1} & 1 \end{pmatrix} \text{ and}$$

$$M_U^{-1} = \begin{pmatrix} \hat{M}^{-1} & 0 \\ -v_n^{-1} v^T \hat{M}^{-1} & v_n^{-1} \end{pmatrix}$$

therefore,

$$N_{U_1} M_{U_1}^{-1} = \begin{pmatrix} \hat{T}_{U_1} W \\ 0 \end{pmatrix} \geq 0$$

and

$$N_U M_U^{-1} = \begin{pmatrix} \bar{T}_U v_n^{-1} W \\ 0 \end{pmatrix} \geq 0$$

where  $\hat{T}_{U_1} = \hat{N} \hat{M}^{-1} - W u^T \hat{M}^{-1}$  and  $\bar{T}_U = \hat{N} \hat{M}^{-1} - W v_n^{-1} v^T \hat{M}^{-1}$ . Since both the lower-right corner of  $N_U M_U^{-1}$  and  $N_{U_1} M_{U_1}^{-1}$  have zeros,  $\rho(N_U M_U^{-1})$  and  $\rho(N_{U_1} M_{U_1}^{-1})$  exist in  $\bar{T}_U$  and  $\hat{T}_{U_1}$ , respectively. That is,  $\rho(N_U M_U^{-1}) = \rho(\bar{T}_U)$  and  $\rho(N_{U_1} M_{U_1}^{-1}) = \rho(\hat{T}_{U_1})$ . By simple computation, we know that  $\bar{T}_U \leq \hat{T}_{U_1}$  under the assumption  $a_{nj} - \sum_{k=1}^{n-1} a_{n,k} a_{k,n} \leq \sum_{k=1}^{n-1} a_{n,k} a_{k,j}$ ,  $1 \leq j \leq n-1$ . Hence by [Lemma 2.1](#), we have

$$\rho(N_U M_U^{-1}) = \rho(\bar{T}_U) \leq \rho(\hat{T}_{U_1}) = \rho(N_{U_1} M_{U_1}^{-1}).$$

Therefore, by [Lemma 2.3](#) we immediately know that

$$\rho(M_U^{-1}N_U) = \rho(N_U M_U^{-1}) \leq \rho(N_{U_1} M_{U_1}^{-1}) = \rho(M_{U_1}^{-1}N_{U_1}), \text{ which means that } \rho(T_U) \leq \rho(T_{U_1}).$$

#### 4. Comparison Theorems

In this section, we discuss a comparison with  $P_U$  and  $P_R$ . The comparison result show that the preconditioner  $P_U$  is better than  $P_R$  for sloving the preconditioned linear system (2).

**Theorem 4.1.** Let  $A$  be a nonsingular M-matrix. If  $0 \leq \sum_{k=i+1}^n a_{i,k} a_{k,i} < 1, 1 \leq i \leq n-1$  and  $0 \leq \sum_{k=1}^{n-1} a_{n,k} a_{k,n} < 1$  and  $0 \leq a_{i,k_i} a_{k_i,i} < 1, 1 \leq i \leq n-1$ , then  $\rho(T_U) \leq \rho(T_R) < 1$

**Proof.** For  $M_U = I - D - L - E + R - \hat{D} - \hat{E}$  and  $N_U = F + U^2$  by Theorem 3.2 we know that  $A_U = P_U A = M_U - N_U$  is a Gauss-Seidel convergent splitting. For  $M_R = I - \hat{D} - L - \hat{E} + R - \hat{D} - \hat{E}$  and  $N_R = U - S_{max} + \hat{F} + S_{max} U$  that  $\hat{D}, \hat{E}$  and  $\hat{F}$  are respectively the diagonal, strictly lower triangular and strictly upper triangular parts of  $S_{max} L$ , and  $\hat{D}$  and  $\hat{E}$  are the diagonal, strictly lower triangular parts of  $R(L + U)$ , respaectivly. From [4] we know that  $A_R = P_R A = M_R - N_R$  is a Gauss-Seidel convergant splitting. To compare  $\rho(T_U)$  with  $\rho(T_R)$ , we consider the following splitting of A:

$$A_U = P_U A = M_U - N_U$$

$$(I + U + R)A = M_U - N_U$$

$$A = (I + U + R)^{-1} M_U - (I + U + R)^{-1} N_U$$

that we take  $M_1 = (I + U + R)^{-1} M_U$  and  $N_1 = (I + U + R)^{-1} N_U$

and

$$A_R = P_R A = M_R - N_R$$

$$(I + S_{max} + R)A = M_R - N_R$$

$$A = (I + S_{max} + R)^{-1} M_R - (I + S_{max} + R)^{-1} N_R$$

If we take  $M_2 = (I + S_{max} + R)^{-1} M_R$  and  $N_2 = (I + S_{max} + R)^{-1} N_R$ , then  $\rho(M_1^{-1} N_1) < 1$  and  $\rho(M_2^{-1} N_2) < 1$  since  $M_U^{-1} N_U = M_1^{-1} N_1$  and  $M_R^{-1} N_R = M_2^{-1} N_2$ .

Then  $A = M_1 - N_1 = M_2 - N_2$  are two convergant splittings.

Since matrices  $L, D, \hat{D}, E, \hat{E}, R, \hat{D}$  and  $\hat{E}$  are positive and  $D \geq \hat{D}$  and  $E \geq \hat{E}$ , we have  $-D \leq -\hat{D}$  and  $-E \leq -\hat{E}$ . Then the following inequality holds:

$$I - D - L - E \leq I - \hat{D} - L - \hat{E}$$

and we have:

$$I - D - L - E + R - \hat{D} - \hat{E} \leq I - \hat{D} - L - \hat{E} + R - \hat{D} - \hat{E}.$$

Therefore

$$(I - D - L - E + R - \hat{D} - \hat{E})^{-1} \geq (I - \hat{D} - L - \hat{E} + R - \hat{D} - \hat{E})^{-1}$$

Also,  $P_U = I + U + R$  and  $P_R = I + S_{max} + R$  are positive matrices and we have

$$I + U + R \geq I + S_{max} + R \quad (6)$$

from (5) and (6) the following relation holds:

$$(I - D - L - E + R - \hat{D} - \hat{E})^{-1} (I + U + R) \geq (I - \hat{D} - L - \hat{E} + R - \hat{D} - \hat{E})^{-1} (I + S_{max} + R)$$

and we know that

$$M_1^{-1} = (I - D - L - E + R - \hat{D} - \hat{E})^{-1} (I + U + R)$$

and

$$M_2^{-1} = (I - \hat{D} - L - \hat{E} + R - \hat{D} - \hat{E})^{-1} (I + S_{max} + R)$$

Then, from (7),  $M_1^{-1} \geq M_2^{-1}$  it follows from Lemma 2.4 that

$$\rho(M_1^{-1} N_1) \leq \rho(M_2^{-1} N_2) < 1. \text{ Hence, } (M_U^{-1} N_U) \leq \rho(M_R^{-1} N_R) < 1, \text{ i.e., } \rho(T_U) \leq \rho(T_R) < 1.$$

■

## 5. Numerical Examples

**Example 5.1.** Consider the following matrix,

$$A = \begin{pmatrix} 1 & -0.2 & -0.3 & -0.1 & -0.2 \\ -0.1 & 1 & -0.1 & -0.3 & -0.1 \\ -0.2 & -0.1 & 1 & -0.1 & -0.2 \\ -0.2 & -0.1 & -0.1 & 1 & -0.3 \\ -0.1 & -0.2 & -0.2 & -0.1 & 1 \end{pmatrix}$$

by computation, we have

$$\rho(M^{-1} N) = 0.460779 > \rho(M_U^{-1} N_U) = 0.156956$$

and

$$\rho(M_{U_1}^{-1} N_{U_1}) = 0.186007 > \rho(M_U^{-1} N_U) = 0.156956 \text{ and}$$

$$\rho(M_R^{-1}N_R) = 0.257251 > \rho(M_U^{-1}N_U) = 0.156956.$$

**Example 5.2.** Let the coefficient matrix A given by

$$A = \begin{pmatrix} 1 & 0 & -0.1 & -0.2 & 0 & 0 & -0.4 & -0.1 & -0.1 \\ -0.1 & 1 & 0 & 0 & -0.3 & -0.1 & -0.1 & 0 & -0.2 \\ -0.1 & -0.2 & 1 & 0 & -0.1 & 0 & -0.3 & 0 & 0 \\ 0 & -0.1 & -0.1 & 1 & 0 & -0.1 & -0.4 & 0 & -0.1 \\ -0.2 & 0 & -0.1 & 0 & 1 & 0 & -0.4 & -0.1 & -0.1 \\ -0.1 & 0 & 0 & -0.1 & 0 & 1 & -0.3 & 0 & -0.2 \\ -0.2 & -0.2 & 0 & -0.1 & 0 & 0 & 1 & -0.2 & -0.1 \\ -0.1 & 0 & 0 & -0.2 & -0.2 & -0.1 & 0 & 1 & -0.3 \\ 0 & 0 & -0.1 & -0.2 & 0 & 0 & -0.1 & -0.3 & 1 \end{pmatrix}$$

Obviously, from numerical results, we have  $\rho(T_U) \leq \rho(T_R)$  and

$$\rho(T_U) \leq \rho(T_{U_1}) \leq \rho(T), \quad \text{we have} \quad \rho(T_U) = 0.414255, \quad \rho(T_R) = 0.478073, \quad \rho(T_{U_1}) = 0.421223 \quad \text{and} \quad \rho(T) = 0.670704.$$

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