

A Modified Gauss – Seidel Method for M - Matrices

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Abstract : in 1991, A. D. Gunawardena et al. proposed the modified Gauss-Seidel (MGS) method for solving the linear system with the preconditioned $= I + S$. The preconditioning Effect is not observed on the nth row. In the present paper, we suggest a new precondition. We get the convergence and comparison theorems for the proposed method. Numerical examples also given.

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1- Introduction:

We consider the following preconditioned linear system.

$$PAX = Pb, \quad (1.1)$$

where $A = (a_{ij})_{n \times n} \in R^{n \times n}$ is a known non singular M-matrix, $P \in R^{n \times n}$, called the preconditioned, is non singular, $b \in R(A)$ is known and $X \in R^{n \times 1}$ is unknown, (A) is the range of A. throughout this paper, without loss of generality, we always assume that the coefficient matrix A has a splitting of the form $A = I - L - U$, where I is the identity matrix, $-L$ and $-U$ are strictly lower triangular and strictly upper triangular parts of A, respectively.

To effectively solve the preconditioned linear system (1.1), a variety of preconditioners have been proposed by several authors, see [1 – 6] and the references there in. Since some preconditioned are constructed only from a part of upper triangular part of A, the preconditioning effect is not observed on the last row of matrix A. For example, the preconditioned $P_s = I + S$ presented in [1] and $P_{smax} = I + S_{max}$ in [7] are formed respectively by

$$S = (s_{ij}) = \begin{cases} -a_{i,i+1} & i = 1, 2, \dots, n-1; \\ 0, & \text{Other wise} \end{cases}$$

and

$$S_{max} = (s_{ij}^m) = \begin{cases} -a_{ij}, & i = 1, \dots, n-1, j > i; \\ 0, & \text{Other wise}, \end{cases}$$

$$K_i = \min\{j \mid \max_j |a_{ij}|, i < n\}$$

Motivated by their results, in this paper, we propose the following preconditioned:

$$P_m = I + S + R_{max} \quad (1.2)$$

where

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

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

For the preconditioned (1.2), the preconditioned matrix

$$A_m = (I + S + R_{max})A$$

Can be split as

$$A_m = M_m - N_m$$

$$= (I - D - L - E + R_{\max} - \hat{D} - \hat{E}) - (U - S + SU),$$

where D and E are respectively the diagonal, strictly lower triangular parts of SL , while \hat{D} and \hat{E} are the diagonal, strictly lower triangular, the MGS iterative matrix is $T_m = M_m^{-1}N_m$.

2. Preliminaries

For the convenience of the readers, we first give some of the notations, definitions and lemmas which will be used in what follows.

For $A = (a_{i,j}), B = (b_{i,j}) \in R^{n \times n}$, we write $A \geq B$ if $a_{i,j} \geq b_{i,j}$ holds for all $i, j = 1, 2, \dots, n$. $A \geq O$, called nonnegative, if $a_{i,j} \geq 0$ for all $i, j = 1, 2, \dots, n$, where O is a $n \times n$ zero matrix. For the vectors $a, b \in R^{n \times 1}$, $a \geq b$ and $a \geq o$ can be defined in the similar manner.

Definition 2.1 ([9]). A matrix A is a L -matrix if $a_{i,i} > 0, i = 1, \dots, n$ and $a_{i,j} \leq 0$ for all $i, j = 1, \dots, n, i \neq j$. A nonsingular L -matrix A is a nonsingular M -matrix if $A^{-1} \geq 0$.

Lemma 2.1 ([10]). Let A be a nonnegative nonzero matrix. Then

- (a) $\rho(A)$, the spectral radius of A , is an eigen value;
- (b) A has a nonnegative eigenvector corresponding to $\rho(A)$;
- (c) $\rho(A)$ is a simple eigen value of A ;
- (d) $\rho(A)$ increases when any entry of A increases.

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 $N \geq 0$ [10];
- (b) weak regular if $M^{-1} \geq 0$ and $M^{-1}N \geq 0$ [11];
- (c) nonnegative if $M^{-1}N \geq 0$ [12].
- (d) M -splitting if M is a nonsingular M -matrix and $N \geq 0$ [13].

Definition 2.3 ([8]). We call $A = M - N$ the Gauss-Seidel splitting of A , if $M = (I - L)$ is nonsingular and $N = V$. In addition, the splitting is called

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

Lemma 2.2 ([14]). Let $A = M - N$ be an M -Splitting Of A . Then $\rho(M^{-1}N) < 1$ if and 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 ([16]). 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. Convergence And Comparison Theorems

We begin this section with a lemma given in [7].

For the preconditioned $P_S = I + S$, the Preconditioned Matrix $A_S = (I + S)A$ can be written as

$$A_S = M_S - N_S = (I - D - L - E) - (U - S + SU).$$

In which D and E are defined as in section 1. If $a_{i,i+1}a_{i+1,i} \neq 1 (i = 1, 2, \dots, n-1)$, then the MGS iterative matrix T_S for A_S can be defined by $T_S = M_S^{-1}N_S = (I - D - L - E)^{-1}(U - S + SU)$ as $(I - D - L - E)^{-1}$ exists. In this case there is the following result:

Lemma 3.1 ([7]). Let $A = I - L - U$ be a nonsingular M -matrix. Assume that $0 \leq a_{i,i+1}a_{i+1,i} < 1, 1 \leq i \leq n-1$, then $A_S = M_S - N_S$ is regular and Gauss-Seidel convergent.

Theorem 3.2. Let A be a nonsingular M -matrix. Assume that $0 \leq a_{i,i+1}a_{i+1,i} < 1, 1 \leq i \leq n-1$ and $0 \leq a_{n,k_j}a_{k_j,n} < 1, k_j = 1, \dots, n-1$, then

$A_m = M_m - N_m$ regular and Gauss-Seidel convergent splitting.

Proof. We observe that when $0 \leq a_{i,i+1}a_{i+1,i} < 1, 1 \leq i \leq n-1$ and $0 \leq a_{n,k_j}a_{k_j,n} < 1, k_j = 1, \dots, n-1$, the diagonal elements of A_m are positive and M_m^{-1} exists. It is known that (see [11]) an L-matrix A is a non singular M-matrix if and only if there exists a positive vector y such that > 0 . By taking such, the fact that $I + S + R_{max} \geq 0$ implies $A_m y = (I + S + R_{max})A_y > 0$. Consequently, the L-matrix A_m is a nonsingular M-matrix, which means $A_m^{-1} \geq 0$.

We note that $L - R_{max} + E + \hat{E} \geq 0$ since $L \geq R_{max} \geq 0$.

When $0 \leq a_{i,i+1}a_{i+1,i} < 1, 1 \leq i \leq n-1$ and $0 \leq a_{n,k_j}a_{k_j,n} < 1, k_j = 1, \dots, n-1$, we have $D + \hat{D} < I$, so that $(I - D - \hat{D}) \geq 0$. Hence,

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

On the other hand, it is to see that $N_m = U - S + SU \geq 0$ since $U \geq S$ and $SU \geq 0$. Therefore, $A_m = M_m - N_m$ is a regular and Gauss-Seidel convergent splitting by definition 2.3 and lemma 2.2.

For the splitting $A = I - L - U$ of matrix A , the iteration matrix of the classical Gauss-Seidel method for A is $T = (I - L)^{-1}U$. Comparing $\rho(T)$ with $\rho(T_m)$, the spectral radius of the MGS with the preconditioned $P_m = I + S + R_{max}$, we have the following comparison theorem:

Theorem 3.3. Let A be a nonsingular M-matrix. Then under the assumptions of theorem 3.2, we have $\rho(T_m) \leq \rho(T) < 1$.

Proof. For $M_m = I - D - L - E + R_{max} - \hat{D} - \hat{E}$ and $N_m = U - S + SU$, by theorem 3.2, we know that $A_m = P_m A = M_m - N_m$ is a Gauss-Seidel convergent splitting. Since A is a nonsingular, the classic Gauss-Seidel splitting $A = (I - L) - U$ of A is clearly regular and convergent.

To compare $\rho(T_m)$ with $\rho(T)$, we consider the following splitting of A :

$$A = (I + S + R_{max})^{-1}M_m - (I + S + R_{max})^{-1}N_m$$

If we take $M_1 = (I + S + R_{max})^{-1}M_m$ and $N_1 = (I + S + R_{max})^{-1}N_m$, then $\rho(M_1^{-1}N_1) < 1$ since $M_1^{-1}N_m = M_1^{-1}N_1$.

Also note that

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

It follows from lemma 2.4 that $\rho(M_1^{-1}N_1) \leq \rho(M^{-1}N) < 1$.

Hence $\rho(M_m^{-1}N_m) \leq \rho(M^{-1}N) < 1$, i.e., $\rho(T_m) \leq \rho(T) < 1$.

Next, we give a comparison theorem between the MGS methods with the preconditioners P_m and P_s respectively.

Theorem 3.4. Let A be a nonsingular M-matrix. Then under the assumptions of theorem 3.2 and $a_{k_n,j} \leq a_{k_n}, n a_{n,j}, 1 \leq n-1$, we have $\rho(T_m) \leq \rho(T_s) < 1$.

Proof. For the matrices M_s, M_m, N_s and N_m in the splitting of matrices $P_s A = M_s - N_s$ and $P_m A = M_m - N_m$, they can be expressed in the partitioned forms as follows:

$$M_s = I - D - L - E = \left(\begin{array}{c|c} \hat{M} & 0 \\ \hline U^T & 1 \end{array} \right),$$

$$M_m = I - D - L - E + R_{max} - \hat{D} - \hat{E},$$

$$M_m = M_s + R_{max} \times A = \left(\begin{array}{c|c} \hat{M} & 0 \\ \hline V^T & U_n \end{array} \right),$$

$$N_m = N_s = \left(\begin{array}{c|c} \hat{N} & W \\ \hline 0 & 0 \end{array} \right),$$

where

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

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

$$V^T = (V_1, \dots, V_{n-1}), V_j = a_{n,j} - a_{n,kn}a_{kn,j} (1 \leq j \leq n-1),$$

$$V_n = 1 - a_{n,kn}a_{kn,n},$$

$$W = (W_1, \dots, W_{n-1})^T, W_i = -a_{i,n} + a_{i,i+1}a_{i+1,n} (1 \leq i \leq n-1),$$

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

Direct computation yields

$$M_s^{-1} = \left(\begin{array}{c|c} \hat{M}^{-1} & 0 \\ \hline U^T \hat{M}^{-1} & 1 \end{array} \right) \quad \text{and} \quad M_m^{-1} = \left(\begin{array}{c|c} \hat{M}^{-1} & 0 \\ \hline -V_n^{-1}V^T \hat{M}^{-1} & V_n^{-1} \end{array} \right).$$

Therefore,

$$N_s M_s^{-1} = \left(\begin{array}{c|c} \hat{T}_s & W \\ \hline 0 & 0 \end{array} \right) \geq 0$$

and

$$N_m M_m^{-1} = \left(\begin{array}{c|c} \bar{T}_m & U_n^{-1}W \\ \hline 0 & 0 \end{array} \right) \geq 0$$

where $\hat{T}_s = \hat{N}\hat{M}^{-1} - Wu^T \hat{M}^{-1}$ and $\bar{T}_m = \hat{N}\hat{M}^{-1} - WV_n^{-1}V^T \hat{M}^{-1}$.

Obviously, $\rho(N_s M_s^{-1}) = \rho(\hat{T}_s)$ and $\rho(N_m M_m^{-1}) = \rho(\bar{T}_m)$.

By simple computation, we know $\bar{T}_m \leq \hat{T}_s$ that under the assumption $a_{kn,j} \leq a_{kn,n}a_{n,j}$, $1 \leq j \leq n-1$. Hence by lemma 2.1, we have

$$\rho(N_m M_m^{-1}) = \rho(\bar{T}_m) \leq \rho(\hat{T}_s) = \rho(N_s M_s^{-1})$$

Therefore, by lemma 2.3, we immediately know that which means that $\rho(T_m) \leq \rho(T_s)$.

4. Numerical Examples

In this part, we give some examples to illustrate the theory in section 3.

Example 4.1. Let us consider the matrix A of (1.1), given by

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

We have $\rho(T_m) = 0.3114$ and $\rho(T_s) = 0.3384$. Clearly, $\rho(T_m) < \rho(T_s)$ holds.

Example 4.2. Let the coefficient matrix A of (1.1) be given by

$$A = \begin{pmatrix} 1 & -0.5 & -0.5 \\ -0.3 & 1 & -0.6 \\ -0.3 & -0.3 & 1 \end{pmatrix}.$$

We have $\rho(T_m) = 0.29167 < \rho(T_s) = 0.44763$

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