

# Parallel multisplitting methods for singular linear systems

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**Abstract**—In this paper, we discuss convergence of the extrapolated iterative methods for linear systems with the coefficient matrices are singular H-matrices. And we present the sufficient and necessary conditions for convergence of the extrapolated iterative methods. Moreover, we apply the results to the GMAOR methods. Finally, we give one numerical example.

**Keywords**—singular H-matrix, linear systems, extrapolated iterative method, GMAOR method, convergence.

## I. INTRODUCTION

LET us consider a system of  $n$  equations

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

where  $A \in C^{n \times n}$  is singular,  $b, x \in C^n$  with  $b$  known and  $x$  unknown. We assume that the system (1) is solvable, i.e., it has at least one solution. In order to solve the system (1) with parallel multi-splitting iterative methods, we assume that

$$(1) A = M_k - N_k, \quad k = 1, 2, \dots, \alpha,$$

where  $M_k$  is a nonsingular matrix;

$$(2) \sum_k E_k = I \quad (I \in R^{n \times n}),$$

where  $E_k$  are diagonal and  $E_k \geq 0$ .

Then a parallel multi-splitting iterative method for solving (1) can be described as follows

$$x^{m+1} = Tx^m + Sb, \quad m = 0, 1, 2, \dots, \quad (2)$$

where  $T = \sum_k E_k M_k^{-1} N_k$  is the iteration matrix,  $S = \sum_k E_k M_k^{-1}$ .

It is well known that for singular systems the iterative method (2) is convergent if and only if the associated convergence factor

$$\vartheta(T) \equiv \max\{|\mu|, \mu \in (\sigma(T) \setminus \{1\})\} < 1$$

and the elementary divisors associated with  $\mu = 1 \in \sigma(T)$  are linear, i.e.,

$$\text{index}(I - T) = 1,$$

where  $\sigma(T)$  denotes the spectrum of  $T$  and  $\text{index}(B)$  denotes the index of the matrix  $B$ , i.e., the smallest nonnegative integer

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$k$  such that  $\text{rank}(B^{k+1}) = \text{rank}(B^k)$ . In this case,  $T$  is called a semi-convergent matrix. In the extrapolated case, the method (2) can be defined by

$$x^{k+1} = T_\omega x^k + \omega M^{-1}b, \quad k = 0, 1, 2, \dots, \quad (3)$$

where

$$T_\omega = (1 - \omega)I + \omega T, \quad (4)$$

is the iteration matrix and  $\omega \in R$  is called the extrapolated parameter ([1]). Clearly, if  $\omega = 0$  then  $T_0 = I$  and the extrapolated method (3) becomes

$$x^{k+1} = x^k, \quad k = 0, 1, 2, \dots$$

Thus we assume that  $\omega \neq 0$  in further considerations.

Now we assume that

(1)  $A = D - L_k - U_k$ ,  $k = 1, 2, \dots, \alpha$ , where  $\text{diag}(D) = \text{diag}(A)$ ,  $D$  is a nonsingular matrix,  $L_k$  and  $U_k$  are matrices with zeros in the diagonal, where  $L_k = (l_{ij})_k$ ,  $U_k = (u_{ij})_k$ . In general, we don't assume that  $L_k$  and  $U_k$  are triangular matrices;

(2)  $\sum_k E_k = I$ , where  $E_k$  are diagonal and  $E_k \geq 0$ .

Then the collection of triples  $(D - L_k, U_k, E_k)$ ,  $k = 1, 2, \dots, \alpha$ , is called a multi-splitting of  $A$ .

We introduce the operators  $F_k$  by

$$F_k(\gamma, \omega, x) = (D - \gamma L_k)^{-1} \times [(1 - \omega)D + (\omega - \gamma)L_k + \omega U_k]x + \omega b, \\ \gamma \geq 0, \quad \omega > 0, \quad k = 1, 2, \dots, \alpha.$$

Algorithm: Choose  $x^0 \in R^n$  arbitrarily. For  $m = 0, 1, 2, \dots$  until convergence,

$$x^{m+1} = \sum_k E_k F_k(\gamma, \omega, x^m),$$

If we define the matrix

$$\ell_{GMAOR}(\gamma, \omega) = \sum_k E_k (D - \gamma L_k)^{-1} \times [(1 - \omega)D + (\omega - \gamma)L_k + \omega U_k]$$

and the vector

$$b_{GMAOR}(\gamma, \omega) = \sum_k E_k (D - \gamma L_k)^{-1} (\omega b),$$

then from Algorithm we get

$$x^{m+1} = \ell_{GMAOR}(\gamma, \omega) x^m + b_{GMAOR}(\gamma, \omega), \quad m = 0, 1, \dots$$

Obviously, if  $D$  is diagonal,  $L_k$  are strictly lower triangular matrices and  $U_k$  are matrices with zeros in the diagonal, then

the above algorithm will reduce to the well-known MAOR algorithm (parallel multi-splitting AOR algorithm [2]). Hence we call Algorithm a parallel generalized multi-splitting AOR algorithm (GMAOR). Furthermore, we observe that when  $(\gamma, \omega)$  is equal to  $(\omega, \omega), (1, 1), (0, \omega)$  and  $(0, 1)$  the GMAOR method reduces to the GMSOR, GMGS, GMJOR and GMJ iterative methods, respectively, with the iteration matrices  $\ell_{GMSOR}(\gamma), \ell_{GMGS}, \ell_{GMJOR}$  and  $\ell_{GMJ}$ .

It should be noted that, if  $\gamma \neq 0$ , the GMAOR method is an extrapolated method of the GMSOR method with the relaxation factor  $\gamma$  and the extrapolated parameter  $\frac{\omega}{\gamma}$ , namely

$$\ell_{GMAOR}(\gamma, \omega) = (1 - \frac{\omega}{\gamma})I + \frac{\omega}{\gamma}\ell_{GMSOR}(\gamma).$$

In this paper, we discuss convergence of the extrapolated iterative methods for solving singular linear systems with the coefficient matrices are singular H-matrices. In Section 2 some sufficient and necessary conditions for convergence of the extrapolated iterative methods are presented. In Section 3 we apply the results of Section 2 to the GMAOR method, which are the extrapolated methods of the GMSOR method.

**Definition 1.1**([3]) A matrix  $A = (a_{ij}) \in R^{n \times n}$  is called a singular M-matrix if  $A$  can be expressed in the form

$$A = sI - B, \quad s > 0, B \geq 0, \quad (5)$$

and

$$s = \rho(B).$$

**Definition 1.2**([4]) A matrix  $A = (a_{ij}) \in C^{n \times n}$  is called a singular H-matrix if its comparison matrix  $M(A) = (\tilde{a}_{ij})$  is a singular M-matrix, where

$$\tilde{a}_{ij} = \begin{cases} |a_{ii}|, & i = j \\ -|a_{ij}|, & i \neq j \end{cases}$$

**Definition 1.3**([5]) Let  $A \in R^{n \times n}$ .  $A = M - N (M, N \in R^{n \times n})$  is called as an H-splitting if  $M(M) - |N|$  is an M-matrix. If  $M(A) = M(M) - |N|$ , then  $A = M - N$  is called as an H-compatible splitting.

## II. SUFFICIENT AND NECESSARY CONDITIONS FOR CONVERGENCE

**Lemma 2.1**([6]) The extrapolated method

$$x^{(m+1)} = [(1 - \omega)I + \omega M^{-1}N]x^{(m)} + \omega M^{-1}b,$$

$m = 0, 1, 2, \dots$ , is convergent if and only if  $\text{index}(I - T) = 1$  and one of the following conditions is satisfied.

(1)  $Re\mu < 1$ , for all  $\mu \in \sigma(T) \setminus \{1\}$ , and

$$0 < \omega < \min_{\mu \in \sigma(T) \setminus \{1\}} \frac{2(1 - Re\mu)}{1 - 2Re\mu + |\mu|^2};$$

(2)  $Re\mu > 1$ , for all  $\mu \in \sigma(T) \setminus \{1\}$ , and

$$0 > \omega > \max_{\mu \in \sigma(T) \setminus \{1\}} \frac{2(1 - Re\mu)}{1 - 2Re\mu + |\mu|^2}.$$

**Theorem 2.1** The extrapolated method (??) is convergent if and only if  $\text{index}(I - T) = 1$  and one of the following conditions is satisfied.

(1)  $Re\mu < 1$ , for all  $\mu \in \sigma(T) \setminus \{1\}$ , and

$$0 < \omega < \min_{\mu \in \sigma(T) \setminus \{1\}} \frac{2(1 - Re\mu)}{1 - 2Re\mu + |\mu|^2};$$

(2)  $Re\mu > 1$ , for all  $\mu \in \sigma(T) \setminus \{1\}$ , and

$$0 > \omega > \max_{\mu \in \sigma(T) \setminus \{1\}} \frac{2(1 - Re\mu)}{1 - 2Re\mu + |\mu|^2}.$$

**Proof:** By Lemma 1 we know that Theorem 1 holds obviously.

Now we denote

$$\tau(T) = \min_{\mu \in \sigma(T) \setminus \{1\}} \frac{2(1 - Re\mu)}{1 - 2Re\mu + |\mu|^2}.$$

**Corollary 2.1** If  $\rho(T) = 1$ , then the following statements are true.

(1) The extrapolated iterative method is convergent if and only if  $\text{index}(I - T) = 1$  and  $0 < \omega < \tau(T)$ .

(2) The inequalities

$$\tau(T) \geq \frac{2}{1 + \vartheta(T)} \geq 1$$

hold.

**Proof:** (1) Since  $\rho(T) = 1$ , we have  $Re\mu < 1$  for  $\mu \in \sigma(T) \setminus \{1\}$ . Thus by Theorem 1 it follows that (??) is convergent if and only if  $\text{index}(I - T) = 1$  and  $0 < \omega < \tau(T)$ .

(2) For  $\mu \in \sigma(T) \setminus \{1\}$ , we have

$$\frac{(1 - Re\mu)}{1 - 2Re\mu + |\mu|^2} \geq \frac{1}{1 + |\mu|},$$

if  $|\mu| < 1$ . And if  $|\mu| = 1$  then  $Re\mu < 1$ , hence

$$\frac{(1 - Re\mu)}{1 - 2Re\mu + |\mu|^2} = \frac{1}{1 + |\mu|} = \frac{1}{2}.$$

Correspondingly, we have

$$\tau(T) \geq \min_{\mu \in \sigma(T) \setminus \{1\}} \frac{2}{1 + |\mu|} = \frac{2}{1 + \vartheta(T)},$$

thus (2) follows immediately.

**Lemma 2.2**([7]) Let  $A \in R^{n \times n}$  be an irreducible singular H-matrix. Further, assume that the splitting  $A = M_k - N_k (k = 1, 2, \dots, \alpha)$  is an H-compatible splitting, then  $\rho(T) = 1$  and  $\text{index}(I - T) = 1$ .

**Lemma 2.3**([7]) Let  $A \in R^{n \times n}$  be a singular H-matrix. Further, assume that the splitting  $A = M_k - N_k (k = 1, 2, \dots, \alpha)$  is an H-compatible splitting and  $\text{ind}_0(A) = \inf\{k : \ker(A^k) = \ker(A^{k+1})\} = 1$ , then  $\rho(T) = 1$  and  $\text{index}(I - T) = 1$ , where  $\ker A$  is the kernel of the linear transformation  $A$ .

**Theorem 2.2** Let  $A \in R^{n \times n}$  be an irreducible singular H-matrix. Further, assume that the splitting  $A = M_k - N_k, k = 1, 2, \dots, \alpha$  is an H-compatible splitting. Then the extrapolated method (??) is convergent if and only if  $0 < \omega < \tau(T)$ .

**Proof:** From Lemma 2 we know that  $\rho(T) = 1$  and  $\text{index}(I - T) = 1$ . From Corollary 1 we know that the extrapolated method (??) is convergent if and only if  $0 < \omega < \tau(T)$ .

**Theorem 2.3** Let  $A \in R^{n \times n}$  be a singular H-matrix. Further, assume that the splitting  $A = M_k - N_k, k = 1, 2, \dots, \alpha$  is an

H-compatible splitting and  $ind_0(A) = 1$ . Then the extrapolated method (??) is convergent if and only if  $0 < \omega < \tau(T)$ .

Proof: From Lemma 3 we know that  $\rho(T) = 1$  and  $index(I - T) = 1$ . From Corollary 1 we know that the extrapolated method (??) is convergent if and only if  $0 < \omega < \tau(T)$ .

III. APPLICATIONS

**Theorem 3.1** Let  $A \in R^{n \times n}$  be a irreducible singular H-matrix,  $d_{ij} - \gamma(l_{ij})_k \geq 0, (1 - \gamma)d_{ij} + \gamma(u_{ij})_k \geq 0$  or  $d_{ij} - \gamma(l_{ij})_k \leq 0, (1 - \gamma)d_{ij} + \gamma(u_{ij})_k \leq 0$ . If  $0 < \gamma \leq 1$ , then the GMAOR method is convergent if and only if  $0 < \frac{\omega}{\gamma} < \tau(\ell_{GMSOR}(\gamma))$ .

Proof: Let  $M_k = \frac{1}{\gamma}(D - \gamma L_k)$  and  $N_k = \frac{1}{\gamma}[(1 - \gamma)D + \gamma U_k]$ , so the iterative matrix of GMSOR method is  $\ell_{GMSOR}(\gamma) = \sum_k E_k M_k^{-1} N_k$ .

From hypothesis we have

$$M(A) = M\left(\frac{1}{\gamma}(D - \gamma L_k)\right) - \left|\frac{1}{\gamma}[(1 - \gamma)D + \gamma U_k]\right|,$$

so  $A = \frac{1}{\gamma}(D - \gamma L_k) - \frac{1}{\gamma}[(1 - \gamma)D + \gamma U_k]$  is an H-compatible splitting. Hence Theorem 3.1 follows by Theorem 2.2 immediately.

**Corollary 3.1** Let  $A \in R^{n \times n}$  be a irreducible singular H-matrix,  $(l_{ij})_k \geq 0, (u_{ij})_k \geq 0$  or  $(l_{ij})_k \leq 0, (u_{ij})_k \leq 0$ . If  $0 < \gamma \leq 1$ , then the MAOR method is convergent if and only if  $0 < \frac{\omega}{\gamma} < \tau(\ell_{MSOR}(\gamma))$ .

**Theorem 3.2** Let  $A \in R^{n \times n}$  be a singular H-matrix,  $d_{ij} - \gamma(l_{ij})_k \geq 0, (1 - \gamma)d_{ij} + \gamma(u_{ij})_k \geq 0$  or  $d_{ij} - \gamma(l_{ij})_k \leq 0, (1 - \gamma)d_{ij} + \gamma(u_{ij})_k \leq 0$ . Further, assume that  $ind_0(A) = 1, 0 < \gamma \leq 1$ , then the GMAOR method is convergent if and only if  $0 < \frac{\omega}{\gamma} < \tau(\ell_{GMSOR}(\gamma))$ .

Proof: Let  $M_k = \frac{1}{\gamma}(D - \gamma L_k)$  and  $N_k = \frac{1}{\gamma}[(1 - \gamma)D + \gamma U_k]$ , so the iterative matrix of GMSOR method is  $\ell_{GMSOR}(\gamma) = \sum_k E_k M_k^{-1} N_k$ .

From hypothesis we have

$$M(A) = M\left(\frac{1}{\gamma}(D - \gamma L_k)\right) - \left|\frac{1}{\gamma}[(1 - \gamma)D + \gamma U_k]\right|,$$

so  $A = \frac{1}{\gamma}(D - \gamma L_k) - \frac{1}{\gamma}[(1 - \gamma)D + \gamma U_k]$  is an H-compatible splitting. Hence Theorem 3.2 follows by Theorem 2.3 immediately.

**Corollary 3.2** Let  $A \in R^{n \times n}$  be a singular H-matrix,  $(l_{ij})_k \geq 0, (u_{ij})_k \geq 0$  or  $(l_{ij})_k \leq 0, (u_{ij})_k \leq 0$ . Further, assume that  $ind_0(A) = 1$ . If  $0 < \gamma \leq 1$ , then the MAOR method is convergent if and only if  $0 < \frac{\omega}{\gamma} < \tau(\ell_{MSOR}(\gamma))$ .

IV. NUMERICAL EXAMPLE

Consider  $Ax = b$ , where

$$A = \begin{bmatrix} 3 & -1 & 0 & -1 & 0 & -1 \\ 0 & 2 & -1 & 0 & -1 & 0 \\ -1 & 0 & 3 & -1 & 0 & -1 \\ 0 & -1 & 0 & 2 & -1 & 0 \\ -1 & 0 & -1 & 0 & 3 & -1 \\ 0 & -1 & 0 & -1 & 0 & 2 \end{bmatrix}$$

We choose  $D = diag(A), n = 6, p = 3, \gamma = 0.5$ ,

$$L_1 = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \end{bmatrix}$$

$$L_2 = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 1/2 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \end{bmatrix}$$

$$L_3 = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 1/4 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \end{bmatrix}$$

$A = D - L_i - U_i, i = 1, 2, 3$ .

It's easy to know that  $A$  is an irreducible H-matrix and  $D, L_i, U_i$  satisfy the conditions of Theorem 1.

We choose  $E_1 = diag(1, 1, 0, 0, 0, 0), E_2 = diag(0, 0, 1, 1, 0, 0), E_3 = diag(0, 0, 0, 0, 1, 1)$ , then

$$\ell_{GMSOR}(0.5) = \begin{bmatrix} 1/2 & 1/6 & 0 & 1/6 & 0 & 1/6 \\ 0 & 1/2 & 1/4 & 0 & 1/4 & 0 \\ 1/12 & \frac{23}{768} & \frac{2309}{4608} & \frac{113}{576} & \frac{1}{512} & 3/16 \\ 0 & 1/8 & 1/16 & 1/2 & \frac{5}{16} & 0 \\ \frac{7}{72} & \frac{931}{27648} & \frac{4643}{55296} & \frac{47}{768} & \frac{3079}{6144} & \frac{385}{1728} \\ 0 & \frac{5}{32} & \frac{5}{64} & 1/8 & \frac{9}{64} & 1/2 \end{bmatrix}$$

$$\tau(\ell_{GMSOR}(0.5)) = \frac{5902206365182033}{2251799813685248}$$

From Theorem 1 we know that the GMAOR method is convergent when

$$0 < \omega < 0.5\tau(\ell_{GMSOR}(0.5)) = \frac{5902206365182033}{4503599627370496}$$

For example we choose  $\omega = 1.25 < \frac{5902206365182033}{4503599627370496}$ , then

$$\ell_{GMAOR}(0.5, 1.25) =$$

$$\begin{bmatrix} -1/4 & \frac{5}{12} & 0 & \frac{5}{12} & 0 & \frac{5}{12} \\ 0 & -1/4 & 5/8 & 0 & 5/8 & 0 \\ \frac{5}{24} & \frac{115}{1536} & -\frac{2279}{9216} & \frac{565}{1152} & \frac{5}{1024} & \frac{15}{32} \\ 0 & \frac{5}{16} & \frac{5}{32} & -1/4 & \frac{25}{32} & 0 \\ \frac{35}{144} & \frac{4655}{55296} & 0.2099 & \frac{235}{1536} & -\frac{3037}{12288} & \frac{1925}{3456} \\ 0 & \frac{25}{64} & \frac{25}{128} & \frac{5}{16} & \frac{45}{128} & -1/4 \end{bmatrix}$$

$\vartheta(\ell_{GMAOR}(0.5, 1.25)) = \frac{4136045821846107}{4503599627370496} < 1$ ,  $index(I - \ell_{GMAOR}(0.5, 1.25)) = 1$ . That's the GMAOR method is convergent.

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