Refinement of Extended Accelerated Over-Relaxation Method for Solution of Linear Systems

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Abstract
Given any linear stationary iterative methods in the form $z^{(i+1)} = Jz^{(i)} + f$, where $J$ is the iteration matrix, a significant improvements of the iteration matrix will decrease the spectral radius and enhances the rate of convergence of the particular method while solving system of linear equations in the form $A\mathbf{z} = \mathbf{b}$. This motivates us to refine the Extended Accelerated Over-Relaxation (EAOR) method called Refinement of Extended Accelerated Over-Relaxation (REAOR) so as to accelerate the convergence rate of the method. In this paper, a refinement of Extended Accelerated Over-Relaxation method that would minimize the spectral radius, when compared to EAOR method, is proposed. The method is a 3-parameter generalization of the refinement of Accelerated Over-Relaxation (RAOR) method, refinement of Successive Over-Relaxation (RSOR) method, refinement of Gauss-Seidel (RGS) method and refinement of Jacobi (RJ) method. We investigated the convergence of the method for weak irreducible diagonally dominant matrix, matrix or matrix and presented some numerical examples to check the performance of the method. The results indicate the superiority of the method over some existing methods.

Keywords: EAOR, Refinement of EAOR, iterative method, convergence rate
Introduction

Solutions of linear systems is basically one of the most important aspect of solving physical problems encountered in fields like science, engineering, biological sciences and so on. This is due to the fact that, it is virtually impossible to perform any sort of numerical analysis without coming across linear algebraic equations (Laskar and Behera, 2014). Moreso, sets of linear equations generated from partial differential equations are mostly sparse and large, consumes computation time and resources. The Direct and iterative methods are usually employed to solve such linear systems and utilizing some properties of the coefficient matrices like sparseness for example usually makes it possible for storage reduction, run time and cost (Kiusalaas, 2005). Iterative methods are quite effective and usually preferable over direct methods in finding solutions to sparse/large linear systems. Modification of existing iterative methods in order to achieve higher rate of convergence led to several refinement methods. Dafchahi (2008) introduced the refinement of Jacobi (RJ) method and proved that the method is superior to Jacobi method. Vatti and Eneyew (2011) and Genanew (2016) enhanced the convergence rate of the Gauss-Seidel method by developing the refinement of Gauss-Seidel (RGS) method. Kyurkhiev and Iliev (2013) modified the SOR and SSOR schemes by proposing the refinement of Successive Over-Relaxation (SOR) and refinement of Symmetric Successive Over-Relaxation (SSOR) methods. Vatti et al. (2015) improved on the SOR scheme and proposed the RSOR method. Vatti et al. (2018) attempted to magnify convergence rate of the AOR method by the method called refinement of Accelerated Over-Relaxation (AOOR) method. Muleta et al. (2018) developed a Refinement of generalized Accelerated Over Relaxation (GAOR) method. Recently, Audu et al. (2021) introduced the Extended Accelerated Over-Relaxation method, an efficient three-parameter method that generalizes the AOR, SOR, Jacobi and Gauss-Seidel methods and superior to the above mentioned methods.

This research work is aimed at increasing the rate of convergence of the Extended Accelerated Over-Relaxation (EAOR) method by refining the method into a more efficient method that would hasten the numerical solution of linear systems.

We consider the linear system expressed as

$$Az = b$$

where $\tilde{A} = [a_{ij}]$ is the coefficient matrix, $z$ is the unknowns and $b$ is the constant at the right hand side. Many methods are employed in solving (1) through the splitting of

$$A = \tilde{D} - \tilde{L} - \tilde{U}$$

where $-\tilde{U}, -\tilde{L}$ and $\tilde{D}$ are strictly upper, strictly lower and diagonal part of $A$ respectively. Furthermore, (1) can be written as

$$(\tilde{D} - \tilde{L} - \tilde{U})z = b$$

Or

$$(I - \tilde{L} - \tilde{U})z = b$$

Where $\tilde{D}^{-1}L = L$, $\tilde{D}^{-1}U = U$, $\tilde{D}^{-1}\tilde{D} = I$ and $\tilde{D}^{-1}b = \tilde{b}$. Hence the system of equation in (1) is transformed into

$$\tilde{A}z = \tilde{b}$$

where $\tilde{A} = \tilde{D}^{-1}A$. The popular AOR method by Hadjidimos (1978) for numerical solution of (1) is expressed as

$$z^{(i+1)} = E_{\omega,r}z^{(i)} + [\tilde{D} - r\tilde{L}]^{-1}\omega\tilde{b}, \quad i = 0, 1, 2$$

Here $E_{\omega,r} = [I - rL]^{-1}[(1 - \omega)I + (\omega - r)L + \omega U]$ is the iteration matrix of the AOR method. The Extended Accelerated Over-Relaxation method for numerical solution of (1) is given as

$$z^{(i+1)} = E_{\omega,v,r}z^{(i)} + [I - (r + v)L]^{-1}\omega\tilde{b}, i = 0, 1, 2$$

Where $E_{\omega,v,r} = [I - (r + v)L]^{-1}[(1 - \omega)I + (\omega - (r + v))L + \omega U]$ represents the EAOR iterative matrix and the spectral radius of the iterative matrix is denoted as $\rho(E_{\omega,v,r})$. We shall consider weak irreducible diagonally dominant and $L$ (or $M$) matrices for the coefficient matrix $A$ in (1).
The REAOR Method

Considering the regular splitting of
\[
\tilde{A} = \frac{1}{\omega} (P - Q) \quad \text{yields} \quad \omega \tilde{A} = P - Q
\]
(8)

Where the choices for \(P\) and \(Q\) of the EAOR method are represented as
\[
P = \begin{bmatrix} I - (v + r)L \end{bmatrix}
\]
\[
Q = \begin{bmatrix} (1 - \omega)I + [(\omega - (v + r)L + \omega U) \end{bmatrix}
\]
(9)

to solve (1) by a refinement iterative method, we multiply the linear system \(\tilde{A}z = \tilde{b}\) by the over-relaxation parameter \(\omega\) to obtain
\[
\omega(\tilde{A}z) = \omega(\tilde{b})
\]
(10)

Or
\[
(P - Q)z = \omega\tilde{b}
\]
(11)

Which gives
\[
[I - (v + r)L] - [[1 - \omega]I + [\omega - (v + r)L + \omega U]z = \omega\tilde{b}
\]
(12)
\[
[I - (v + r)L]z = [(1 - \omega)I + [\omega - (v + r)L + \omega U]z + \omega\tilde{b}
\]
(13)
\[
[I - (v + r)L]z = [I - (v + r)L]z + \omega(\tilde{b} - \tilde{A}z)
\]
(14)
\[
z = z + \omega[I - (v + r)L]^{-1}(\tilde{b} - \tilde{A}z)
\]
(15)

From (15), the REAOR formula takes the form:
\[
\tilde{z}^{(i+1)} = \tilde{z}^{(i+1)} + \omega[I - (v + r)L]^{-1}(\tilde{b} - \tilde{A}z^{(i+1)})
\]
(16)

Where \(z^{(i+1)}\) appearing in the right hand side, is \((i + 1)^{th}\) estimation of the EAOR iterative method. The EAOR method in (7) is inserted into (16) to obtain
\[
\tilde{z}^{(i+1)} = [I - (v + r)L]^{-1}[(1 - \omega)I + [\omega - (v + r)L + \omega U]z^{(i)} + [I - (v + r)L]^{-1}\omega\tilde{b}
\]
\[+ [I - (v + r)L]^{-1}\omega\tilde{b} - [I - (v + r)L]^{-1}\omega A
\]
\[\times [(I - (v + r)L)^{-1}[(1 - \omega)I + [\omega - (v + r)L + \omega U]z^{(i)} + [I - (v + r)L]^{-1}\omega\tilde{b}
\]
\[+ [I - (v + r)L]^{-1}\omega\tilde{b}]
\]
(17)

We substituted the value of \(\omega A = (P - Q)\) into (17) to get
\[
\tilde{z}^{(i+1)} = [I - (v + r)L]^{-1}[(1 - \omega)I + [\omega - (v + r)L + \omega U]z^{(i)} + [I - (v + r)L]^{-1}\omega\tilde{b}
\]
\[+ [I - (v + r)L]^{-1}\omega\tilde{b} - [I - (v + r)L]^{-1}\omega A
\]
\[\times [(I - (v + r)L)^{-1}[(1 - \omega)I + [\omega - (v + r)L + \omega U]z^{(i)} + [I - (v + r)L]^{-1}\omega\tilde{b}
\]
\[+ [I - (v + r)L]^{-1}\omega\tilde{b}]
\]
(18)

After several simplifications, we obtain
\[
\tilde{z}^{(i+1)} = [(I - (v + r)L)^{-1}[(1 - \omega)I + [\omega - (v + r)L + \omega U])z^{(i)}
\]
\[+(I + [(1 - \omega)I + [\omega - (v + r)L + \omega U][I - (v + r)L]^{-1}\omega\tilde{b}
\]
(19)

The method in (19) is called the REAOR method, or can be represented as
\[
\tilde{z}^{(i+1)} = Hz^{(i)} + K,
\]
(20)

Where \(H = [(I - (v + r)L)^{-1}[(1 - \omega)I + [\omega - (v + r)L + \omega U])z^{(i)} \quad (i = 0,1,2,\ldots)\) is the iteration matrix of the REAOR method.

The spectral radius of the REAOR method is the largest eigenvalue of its iteration matrix represented as \(\rho(\tilde{R}E_{0,0,1})\). With some specific values of the parameters \(\omega, r, v\), some refinement methods can be obtained such as
\[
RE_{0,0,1} = \text{Refinement of Jacobi (RJ) method}
\]
\[
RE_{0,1,1} = \text{Refinement of Gauss-Seidel (RGS) method}
\]
\[
RE_{0,\omega,\omega} = \text{Refinement of Successive Over-Relaxation (RSOR) method}
\]
\[
RE_{0,\omega,\omega} = \text{Refinement of Accelerated Over-Relaxation (RAOR) method}
\]

Convergence Analysis of the Method

Definition 1: A square matrix \(A\) is considered as a weak diagonally dominant if and only if
\[
|a_{ii}| \geq \sum_{j=1, j \neq i}^{n} |a_{ij}|, \quad i = 1, 2, \ldots, n
\]
Definition 2: A square matrix \((A_{ij})_{m\times n}\) is considered as an \(L\) matrix if \(a_{ij} \leq 0\) (\(i \neq j\)) and \(a_{ii} > 0\), \(i = 1,2,\ldots,n\).

Definition 3: Spectral radius is the absolute value of the greatest eigenvalues of a \(m \times m\) matrix denoted as \(\rho(A) = \max_{\lambda \in \Lambda} |\lambda|\).

Definition 4: A square matrix \((A_{ij})_{m\times n}\) is considered as an \(M\) matrix if \(a_{ij} \leq 0\), \(a_{ii} > 0\), \(i = 1,2,\ldots,n\), \(A\) is non-singular and \(A^{-1} \geq 0\).

Definition 5: A stationary method \(z^{(i+1)} = Jz^{(i)} + f\) converges if the spectral radius of its iteration matrix \((f)\) is less than \(1\), that is to say if \(\rho(f) < 1\).

Corollary 1. Suppose \(A\) is an \(L\) or \(M\) matrix with the range of \(0 < v + r \leq \omega \leq 1\) for the parameters, then the EAOR method is convergent for any arbitrary initial approximation \(z^{(0)}\).

Proof: see Audu et al. (2021).

Corollary 2. If \(A\) is an irreducible weak diagonally dominant matrix for \(0 < v + r < 1\) and \(0 < \omega \leq 1\). Then the EAOR method is convergent for any arbitrary initial estimation \(z^{(0)}\).

Proof: see Audu et al. (2021).

Theorem 1: If matrix \(A\) is an \(L\) or \(M\) matrix, then the Refinement of the Extended Accelerated Over-Relaxation (REAOR) method converges to the exact solution for any initial guess \(z^{(0)}\).

Proof: If \(z\) is the true solution of \(Az = b\), since the coefficient matrix \(A\) is an \(L\) matrix, it follows from corollary 1 that the EAOR method is convergent. This makes it possible to arbitrarily close \(z^{(k+1)}\) to \(z\), hence we get

\[
\|z^{(k+1)} - z\|_\infty = \|z^{(k+1)} + \omega[(I - (v + r)L)^{-1}(b - Az^{(k+1)})] - z\|_\infty
\]

\[
\leq \|z^{(k+1)} - z\|_\infty + |\omega|\|[(I - (v + r)L)^{-1}(b - Az^{(k+1)})]\|_\infty
\]

From the point that \(\|z^{(k+1)} - z\|_\infty \rightarrow 0\), we get \(\|(b - Az^{(k+1)})\|_\infty \rightarrow 0\). This signifies that \(z^{(i+1)}\) tends to zero as \(k \rightarrow \infty\)

\[
\|z^{(i+1)} - z\|_\infty \rightarrow 0
\]

Thus \(\rho(RE_{v,r,\omega}) < 1\) or equivalently \(\rho \left((I - (v + r)L)^{-1}((1 - \omega)(I + [\omega - (v + r)]L + \omega U)\right)^2 < 1\). This shows that the REAOR is a convergent method and the proof is completed.

Theorem 2: Let \(A\) be weak irreducible diagonally dominant. Then for any choice of initial guess \(z^{(0)}\), the REAOR method converges to the true solution.

Proof: Let \(z\) be the true solution of \((1)\) and \(A = (a_{ij})\) is weak irreducible diagonally dominant. Then in view of corollary 2, the EAOR method is convergent and applying similar procedure of theorem 1, it is obvious that the REAOR method is convergent for irreducible weak diagonally dominant matrix and this completes the proof.

Theorem 3: For any initial estimation \(z^{(0)}\), the refinement of Extended Accelerated Over Relaxation method converges to the true solution twice as fast as the EAOR method.

Proof: Let \(\|\| < 1\), we consider the EAOR method \(z^{(i+1)} = Jz^{(i)} + E\) and the REAOR method \(\tilde{z}^{(i+1)} = J\tilde{z}^{(i)} + K\) where
\[ J = [I - rL - vL]^{-1}[(1 - \omega)I + (\omega - r - v)L + \omega U] \]
\[ E = [I - (r + v)L]^{-1}\omega \tilde{b} \]
\[ K = (I + J)[I - (r + v)L]^{-1}\omega \tilde{b} \] (23)

Let \( Z \) be the true solution of (1) that satisfies \( z^{(i+1)} = Jz^{(i)} + E \), then it implies that \( Z = JZ + E \) with respect to EAOR method and similarly \( Z = j^2Z + K \) also satisfies the equation \( z^{(i+1)} = j^2z^{(i)} + K \) with respect to REAOR method.

Considering the EAOR method, we have \( z^{(i+1)} = Jz^{(i)} + E \) and \( Z = JZ + E \), we obtain
\[
\begin{align*}
    z^{(i+1)} - Z &= J(z^{(i)} - Z) + E - Z \\
    &= J(z^{(i)} - Z) + JZ + E - Z \\
    &= J(z^{(i)} - Z) + JZ + E - (JZ + E)
\end{align*}
\] (24)

Taking norm of \( z^{(i+1)} - Z = J(z^{(i)} - Z) \) gives
\[
\begin{align*}
    \|z^{(i+1)} - Z\|_\infty &= \|J(z^{(i)} - Z)\|_\infty \\
    \|z^{(i+1)} - Z\|_\infty &\leq \|J\| \|z^{(i)} - Z\|_\infty \\
    \|z^{(i+1)} - Z\|_\infty &\leq \|J\| \|z^{(i-1)} - Z\|_\infty \\
    \|z^{(i+1)} - Z\|_\infty &\leq \|J\| \|z^{(0)} - Z\|_\infty 
\end{align*}
\] (25)

Then \( z^{(i+1)} \to Z \) as \( i \to \infty \) if \( \|J\| < 1 \). Now considering the REAOR method, we have
\[
\begin{align*}
    \tilde{z}^{(i+1)} - Z &= j^2z^{(i)} - Z + K \\
    &= j^2(z^{(i)} - Z) + j^2Z + K - Z \\
    &= j^2(z^{(i)} - Z) + j^2Z + E - (jZ + E) \\
    \tilde{z}^{(i+1)} - Z &= j^2(z^{(i)} - Z)
\end{align*}
\] (26)

Taking norm of \( \tilde{z}^{(i+1)} - Z = j^2(z^{(i)} - Z) \) gives;
\[
\begin{align*}
    \|\tilde{z}^{(i+1)} - Z\|_\infty &= \|j^2(z^{(i)} - Z)\|_\infty \\
    \|\tilde{z}^{(i+1)} - Z\|_\infty &\leq \|j^2\| \|z^{(i)} - Z\|_\infty \\
    \|\tilde{z}^{(i+1)} - Z\|_\infty &\leq \|j^2\| \|z^{(i-1)} - Z\|_\infty \\
    \|\tilde{z}^{(i+1)} - Z\|_\infty &\leq \|j^2\| \|z^{(0)} - Z\|_\infty 
\end{align*}
\] (27)

Comparing coefficients of \( \|J\|_\infty \|z^{(i)} - Z\|_\infty \) and \( \|j^2\| \|z^{(i)} - Z\|_\infty \), clearly \( \|j^2\| < \|J\| \) since \( \|J\| < 1 \) and this tells us that the REAOR method converges faster than the EAOR method and the proof is completed.

**Numerical Experiment**

In this section we present some numerical tests with respect to the new method. Specifically, we computed the spectral radius of the REAOR iteration matrix and further obtained its convergence results. We compared the results obtained with those of Accelerated Over-Relaxation method, Refinement of Accelerated Over-Relaxation method and Extended Accelerated Over-Relaxation to test the efficiency of the new method. The computations were carried out using Maple 2017 software with accuracy of 10 decimal places and the results are presented in tables 1 to 4.

**Problem 1:** Solve the linear system \( A_1 \) whose coefficient matrix is an Irreducible diagonally dominant matrix by AOR, RAOR, EAOR and REAOR methods.
Problem 2: Solve the linear system $A_2$, whose coefficient matrix is an $L$ matrix by AOR, RAOR, EAOR and REAOR methods.

$$A_2 = \begin{pmatrix} 1 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & 0 \\ -\frac{1}{7} & 1 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 \\ 0 & -\frac{1}{7} & 1 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} \\ -\frac{1}{7} & 0 & -\frac{1}{7} & 1 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 \\ 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 1 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} \\ -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 1 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 \\ 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 1 & -\frac{1}{7} & 0 & -\frac{1}{7} \\ -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 1 & -\frac{1}{7} & 0 \\ 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 1 & -\frac{1}{7} \\ 0 & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 0 & -\frac{1}{7} & 1 \end{pmatrix} \begin{pmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \\ z_5 \\ z_6 \\ z_7 \\ z_8 \\ z_9 \\ z_{10} \end{pmatrix} = \begin{pmatrix} 11.90 \\ 9.32 \\ 8.09 \\ 9.32 \\ 8.09 \\ 8.32 \\ 8.09 \\ 8.32 \\ 8.09 \\ 8.32 \end{pmatrix}$
### Table 1: Comparison Results for Problem 1

| ω  | r   | $\rho$(AOR)  | $\rho$(RAOR)  | v   | $\rho$(EAOR)  | $\rho$(REAOR)  |
|----|-----|--------------|--------------|-----|--------------|--------------|
| 0.1| 0.04| 0.9702892488| 0.9414612263| 0.05| 0.9697412031| 0.9403980010 |
| 0.2| 0.08| 0.9397055636| 0.8830465463| 0.10| 0.9373783471| 0.8786781656 |
| 0.3| 0.12| 0.9082009036| 0.8248288813| 0.15| 0.902634150  | 0.8147290293 |
| 0.4| 0.16| 0.8757229293| 0.7668906489| 0.20| 0.8651200559| 0.7484327111 |
| 0.5| 0.20| 0.8422144385| 0.7093251604| 0.25| 0.8244179744| 0.6796649965 |
| 0.6| 0.24| 0.8076127016| 0.652382758 | 0.30| 0.779331029  | 0.6082956450 |
| 0.7| 0.28| 0.7718486727| 0.5957503735| 0.35| 0.730882855  | 0.5341889152 |
| 0.8| 0.32| 0.734860470 | 0.539987128 | 0.40| 0.6761686319| 0.4572040187 |
| 0.9| 0.36| 0.6965201257| 0.4851402855| 0.45| 0.6141612268| 0.3771940125 |

### Table 2: Comparison Results for Problem 2

| ω  | r   | $\rho$(AOR)  | $\rho$(RAOR)  | v   | $\rho$(EAOR)  | $\rho$(REAOR)  |
|----|-----|--------------|--------------|-----|--------------|--------------|
| 0.1| 0.04| 0.9695659299| 0.9400580924| 0.05| 0.9690220033| 0.9390036428 |
| 0.2| 0.08| 0.9382651880| 0.8803415630| 0.10| 0.9359663473| 0.8760330033 |
| 0.3| 0.12| 0.9060538059| 0.8209349992| 0.15| 0.9005738634| 0.8110332834 |
| 0.4| 0.16| 0.8728842724| 0.7619269530| 0.20| 0.8625309526| 0.7439596442 |
| 0.5| 0.20| 0.8387051160| 0.7034262716| 0.25| 0.8214516534| 0.6747821899 |
| 0.6| 0.24| 0.8034602420| 0.6455486468| 0.30| 0.7768590636| 0.6034974196 |
| 0.7| 0.28| 0.7670892570| 0.5884259282| 0.35| 0.7281033573| 0.5301344989 |
| 0.8| 0.32| 0.7295250213| 0.5322067567| 0.40| 0.6743746009| 0.4547811023 |
| 0.9| 0.36| 0.6906946434| 0.4770590904| 0.45| 0.6145013535| 0.3776119135 |

### Table 3: Convergence Result for Problem 1

| Numerical Methods | Number of Iterations | CPU Time (seconds) |
|-------------------|----------------------|--------------------|
| AOR               | 81                   | 0.296              |
| EAOR              | 56                   | 0.266              |
| RAOR              | 41                   | 0.125              |
| REAOR             | 29                   | 0.094              |

### Table 4: Convergence Result for Problem 2

| Numerical Methods | Number of Iterations | CPU Time (seconds) |
|-------------------|----------------------|--------------------|
| AOR               | 60                   | 0.109              |
| EAOR              | 43                   | 0.031              |
| RAOR              | 31                   | 0.062              |
| REAOR             | 22                   | 0.016              |
Discussion of Results

The rate of convergence is strongly connected to the spectral radius. The closer the spectral radius is to zero the faster the convergence. Comparison results of spectral radius of the new REAOR method and other existing methods for problem 1 and 2 are displayed in tables 1 and 2 respectively. It is observed that the spectral radius of the new method is lower than the spectral radii of EAOR, RAOR and AOR methods due to the fact that $\rho(J_{REAOR}) < \rho(J_{RAOR}) < \rho(J_{EAO}) < \rho(J_{AOR}) < 1$. This implies that the REAOR method will converge faster to the true solution than EAOR, AOR and RAOR methods. Figures 1 and 2 are graphical illustrations of the results in Tables 1 and 2 for easy understanding and comprehension where $J_{V,2018}$ in figure 2 represents the RAOR method.

To validate the spectral radius result, we checked the convergence results for problem 1 and 2 given in tables 3 and 4. The REAOR method
converges at 29\textsuperscript{th} and 22\textsuperscript{nd} iterations in comparison to 56\textsuperscript{th} and 43\textsuperscript{rd} iterations of the EAOR method, 41\textsuperscript{st} and 31\textsuperscript{st} iterations of the RAOR method and 81\textsuperscript{st} and 60\textsuperscript{th} iterations of the AOR method with respect to problems 1 and 2 respectively. The result shows that the new REAOR method converges faster to the true solution and took shorter time to achieve convergence when compared to similar existing methods.

Conclusion

Iterative methods are often employed in solving sparsely large linear systems. Although convergence is a major requirement of such methods, but how fast they converge is equally important. In this study, a new numerical iterative method for solving linear systems Refinement of Extended Accelerated Over-Relaxation method has been proposed. Convergence criteria for some special matrices were examined and two numerical examples of 9 × 9 and 10 × 10 linear systems were presented. Comparison of results of the new method with other existing methods in literature by their spectral radii and number of iterations, indicates that the REAOR method surpasses the other methods. Also, the study has shown that the Refinement method increases the convergence rate of the Extended Accelerated Over-Relaxation method.

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