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Efficient Families of Multi-Point Iterative Methods and Their Self-Acceleration with Memory for Solving Nonlinear Equations

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Abstract: In this paper, we have constructed new families of derivative-free three- and four-parametric methods with and without memory for finding the roots of nonlinear equations. Error analysis verifies that the without-memory methods are optimal as per Kung–Traub’s conjecture, with orders of convergence of 4 and 8, respectively. To further enhance their convergence capabilities, the with-memory methods incorporate accelerating parameters, elevating their convergence orders to 7.5311 and 15.5156, respectively, without introducing extra function evaluations. As such, they exhibit exceptional efficiency indices of 1.9601 and 1.9847, respectively, nearing the maximum efficiency index of 2. The convergence domains are also analysed using the basins of attraction, which exhibit symmetrical patterns and shed light on the fascinating interplay between symmetry, dynamic behaviour, the number of diverging points, and efficient root-finding methods for nonlinear equations. Numerical experiments and comparison with existing methods are carried out on some nonlinear functions, including real-world chemical engineering problems, to demonstrate the effectiveness of the new proposed methods and confirm the theoretical results. Notably, our numerical experiments reveal that the proposed methods outperform their existing counterparts, offering superior precision in computation.

Keywords: with-memory method; simple roots; nonlinear equation; R-order of convergence; Newton interpolating polynomial; chemical engineering applications



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1. Introduction

Iterative methods play a crucial role in solving complex nonlinear equations of the form

$$\Omega(s) = 0, \quad (1)$$

where $\Omega : D \subseteq \mathbb{R} \rightarrow \mathbb{R}$ represents a real function defined on an open interval D . With diverse applications spanning scientific and engineering domains, the computation of nonlinear Equation (1) remains a common yet formidable challenge due to the lack of analytical methods. However, iterative methods provide approximate solutions to nonlinear Equation (1) with high accuracy.

Let $\alpha \in D \subseteq \mathbb{R}$ be a simple root of (1) and s_0 be an initial approximation to α . Then, $\Omega(\alpha) = 0$ and $\Omega'(\alpha) \neq 0$. The most widely used iterative method for finding the simple root of (1) is given below.

$$s_{n+1} = s_n - \frac{\Omega(s_n)}{\Omega'(s_n)}, \quad n = 0, 1, 2, \dots, \quad (2)$$

which is the well-known Newton method [1]. It is a one-point without-memory method with a quadratic order of convergence.

In the past few decades, various optimal multi-point without-memory methods have been developed for the computation of approximated simple roots [2–6]. The concept of an optimal without-memory method is rooted in Kung–Traub’s conjecture [7]. This conjecture proposes that a multi-point without-memory iterative method, which requires q function evaluations per iteration, achieves optimality when its convergence order is precisely 2^{q-1} . The Newton method (2) is optimal for $q = 2$. However, the evaluation of the derivative in the Newton method is a setback for many problems where derivative evaluation is complicated or even does not exist.

To obtain a derivative-free variant of the Newton method (2), the first derivative $\Omega'(s_n)$ in (2) is approximated using the first-order Newton divided difference

$$\Omega'(s_n) \approx \Omega[s_n, w_n] = \frac{\Omega(s_n) - \Omega(w_n)}{s_n - w_n}, \quad (3)$$

where $w_n = s_n + \gamma\Omega(s_n)$, $\gamma \neq 0$ is any real parameter, Equation (2) can be expressed as

$$s_{n+1} = s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n]}, \quad (4)$$

which corresponds to the Traub–Steffensen method [1]. By setting $\gamma = 1$, we obtain the well-known Steffensen method [8]. To assess the efficiency of an iterative method, Ostrowski [9] introduced the efficiency index $(EI) = p^{\frac{1}{q}}$, where q represents the number of function evaluations per iteration and p denotes the order of convergence.

The extension of without-memory methods into with-memory methods using accelerating parameters have gained much attention in recent years [10–13]. In multi-point with memory iterative methods, the order of convergence is significantly increased without any additional function evaluation by using information from the current as well as the previous iterations. In this paper, we introduce new derivative-free families of three-parametric three-point and four-parametric four-point with and without-memory methods for finding simple roots of nonlinear equations. The formulation of the methods is based on the derivative-free biparametric families of without-memory methods developed in [14] and by using accelerating parameters without any additional function evaluations for the with-memory methods. As a result, the orders of convergence of the with-memory methods increase from 4 to 7.5311 and 8 to 15.5156. The accelerating parameters are approximated using Newton’s interpolating polynomials so as to obtain highly efficient with-memory methods.

The subsequent sections of this paper are organised as follows. Section 2 provides the development of modified derivative-free families of without-memory methods, including an in-depth analysis of their theoretical convergence properties. In Section 3, we delve into the derivation and convergence analysis of the derivative-free families of with-memory methods. The numerical experiments and comparative study of the proposed with and without-memory methods against existing approaches on various test functions, including real-world problems, are presented in Section 4 to assess the effectiveness and applicability of our proposed methods. In this section, we also explore the dynamical properties of the methods through the study of basins of attraction, revealing the presence of reflection symmetry in all provided basins of attraction. Finally, Section 5 concludes this paper with key remarks and observations.

2. Modified Families of Three- and Four-Parametric Without-Memory Methods

In this section, we present the new modified derivative-free families of three- and four-parametric multi-point without-memory methods of optimal order in two separate subsections.

2.1. Modified Families of Three-Parametric Three-Point Without-Memory Methods

First, let us consider the following two derivative-free families of biparametric three-point without-memory methods [14].

$$\begin{aligned}
 w_n &= s_n + \gamma\Omega(s_n), \\
 y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta\Omega(w_n)}, \\
 s_{n+1} &= y_n - \frac{\Omega(y_n)}{\zeta(y_n)} \left[1 + \frac{1}{2} \left(\frac{\Omega(y_n)}{\Omega[y_n, w_n] + \beta\Omega(w_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right].
 \end{aligned}
 \tag{5}$$

$$\begin{aligned}
 w_n &= s_n + \gamma\Omega(s_n), \\
 y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta\Omega(w_n)}, \\
 s_{n+1} &= y_n - \frac{\Omega(y_n)}{\zeta(y_n)} \left[1 + \frac{1}{2} \left(\frac{2\Omega(y_n)}{\Omega[y_n, w_n] + \zeta(y_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right],
 \end{aligned}
 \tag{6}$$

where $\zeta(y_n) = \frac{\Omega[s_n, y_n]\Omega[y_n, w_n]}{\Omega[s_n, w_n]}$ and $\rho(y_n) = 2 \frac{\Omega[s_n, y_n]\Omega[y_n, w_n]\Omega[s_n, y_n, w_n]}{(\Omega[s_n, w_n])^2}$.

From here, we introduce a new parameter $\lambda \in \mathbb{R} - \{0\}$ through the modification of $\zeta(y_n)$ as follows:

$$M(y_n) = \frac{\Omega[s_n, y_n]\Omega[y_n, w_n]}{\Omega[s_n, w_n]} + \lambda(y_n - s_n)(y_n - w_n) = \zeta(y_n) + \lambda(y_n - s_n)(y_n - w_n). \tag{7}$$

Now, substituting the above Equation (7) in Equations (5) and (6), we get two new three-parametric families of without memory iterative methods. These modified methods, denoted as Modified Methods (MM), are defined as follows:

Modified Method 4a (MM₄^a):

$$\begin{aligned}
 w_n &= s_n + \gamma\Omega(s_n), \\
 y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta\Omega(w_n)}, \\
 s_{n+1} &= y_n - \frac{\Omega(y_n)}{M(y_n)} \left[1 + \frac{1}{2} \left(\frac{\Omega(y_n)}{\Omega[y_n, w_n] + \beta\Omega(w_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right].
 \end{aligned}
 \tag{8}$$

Modified Method 4b (MM₄^b):

$$\begin{aligned}
 w_n &= s_n + \gamma\Omega(s_n), \\
 y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta\Omega(w_n)}, \\
 s_{n+1} &= y_n - \frac{\Omega(y_n)}{M(y_n)} \left[1 + \frac{1}{2} \left(\frac{2\Omega(y_n)}{\Omega[y_n, w_n] + \zeta(y_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right].
 \end{aligned}
 \tag{9}$$

These modified methods adhere to Kung–Traub’s conjecture, requiring three evaluations of the function at each iteration and exhibit an efficiency index of $4^{1/3} \approx 1.587$.

Next, we explore the theoretical convergence analysis of the newly introduced modified methods, specifically MM₄^a and MM₄^b, as outlined in the following theorem.

Theorem 1. *Let an initial approximation s_0 be close enough to the root α of a sufficiently differentiable real function $\Omega : D \subseteq \mathbb{R} \rightarrow \mathbb{R}$, where D is an open interval. Then, the modified methods MM₄^a (8) and MM₄^b (9) exhibit a fourth order of convergence for any $\beta, \gamma, \lambda \in \mathbb{R} - \{0\}$. Additionally, both the methods have the same error equation given by*

$$\varepsilon_{n+1} = \frac{(1 + \Omega'(\alpha)\gamma)^2(\beta + d_2)(\lambda + \Omega'(\alpha)d_2^2 - \Omega'(\alpha)d_3)}{\Omega'(\alpha)} \varepsilon_n^4 + O(\varepsilon_n^5), \tag{10}$$

where $d_j = \frac{1}{j!} \frac{\Omega^{(j)}(\alpha)}{\Omega'(\alpha)}$, $j = 2, 3, \dots$, and $\varepsilon_n = s_n - \alpha$ is the error at n^{th} iteration.

Proof. The proof of Modified Method 4a (MM₄^a):

Let the error at n^{th} iteration be $\epsilon_n = s_n - \alpha$. Then, employing the Taylor’s series expansion in the vicinity of $s = \alpha$, we obtain

$$\Omega(s_n) = \Omega'(\alpha) \left[\epsilon_n + d_2 \epsilon_n^2 + d_3 \epsilon_n^3 + d_4 \epsilon_n^4 + O(\epsilon_n^5) \right], \tag{11}$$

where $d_j = \frac{1}{j!} \frac{\Omega^{(j)}(\alpha)}{\Omega'(\alpha)}$, $j = 2, 3, \dots$. By using $w_n = s_n + \gamma \Omega(s_n)$, $\gamma \in \mathbb{R} - \{0\}$, expanding $\Omega(w_n)$ using Taylor’s series yields

$$\Omega(w_n) = \Omega'(\alpha) \left[(1 + \Omega'(\alpha)\gamma)\epsilon_n + (1 + \Omega'(\alpha)\gamma(3 + \Omega'(\alpha)\gamma))d_2 \epsilon_n^2 + \sum_{i=1}^2 A_i \epsilon_n^{i+2} + O(\epsilon_n^5) \right], \tag{12}$$

where $A_i, i = 1, 2$ are functions of $\gamma, \Omega'(\alpha), d_2, d_3, d_4$, i.e., $A_1 = 2\Omega'(\alpha)\gamma(1 + \Omega'(\alpha)\gamma)d_2^2 + \Omega'(\alpha)\gamma d_3 + (1 + \Omega'(\alpha)\gamma)^3 d_3$, etc.

Then, using (11) and (12), we have

$$\Omega[s_n, w_n] = \Omega'(\alpha) \left[1 + (2 + \Omega'(\alpha)\gamma)d_2 \epsilon_n + \sum_{i=1}^3 B_i \epsilon_n^{i+1} + O(\epsilon_n^5) \right], \tag{13}$$

where $B_i, i = 1, 2, 3$ are functions of $\gamma, \Omega'(\alpha), d_2, d_3, d_4$, i.e., $B_1 = \Omega'(\alpha)\gamma d_2^2 + (3 + \Omega'(\alpha)\gamma)d_3$, $B_2 = (2 + \Omega'(\alpha)\gamma)(2\Omega'(\alpha)\gamma d_2 d_3 + (2 + \Omega'(\alpha)\gamma(2 + \Omega'(\alpha)\gamma))d_4)$, etc.

Using (11), (12) and (13), we can write

$$y_n - \alpha = (1 + \Omega'(\alpha)\gamma) (\beta + d_2) \epsilon_n^2 + \sum_{i=1}^2 C_i \epsilon_n^{i+2} + O(\epsilon_n^5), \tag{14}$$

where $C_i, i = 1, 2$ are functions of $\beta, \gamma, \Omega'(\alpha), d_2, d_3, d_4$.

Then, by employing (14), $\Omega(y_n)$ is obtained as follows:

$$\Omega(y_n) = \Omega'(\alpha) \left[(1 + \Omega'(\alpha)\gamma) (\beta + d_2) \epsilon_n^2 + \sum_{i=1}^2 C_i \epsilon_n^{i+2} + O(\epsilon_n^5) \right]. \tag{15}$$

With the help of (11)–(15), we obtain

$$M(y_n) = \Omega'(\alpha) + (1 + \Omega'(\alpha)\gamma) (\lambda + 2\Omega'(\alpha)\beta d_2 + 3\Omega'(\alpha)d_2^2 - \Omega'(\alpha)d_3) \epsilon_n^2 + \sum_{i=1}^2 D_i \epsilon_n^{i+2} + O(\epsilon_n^5), \tag{16}$$

where $D_i, i = 1, 2$ are functions of $\beta, \lambda, \gamma, \Omega'(\alpha), d_2, d_3, d_4$.

Now, putting the values of Equations (11)–(16) into the final step of Modified Method 4a (MM₄^a) (8), we get the following expression for the error equation:

$$\epsilon_{n+1} = \frac{(1 + \Omega'(\alpha)\gamma)^2 (\beta + d_2) (\lambda + \Omega'(\alpha)d_2^2 - \Omega'(\alpha)d_3)}{\Omega'(\alpha)} \epsilon_n^4 + O(\epsilon_n^5), \tag{17}$$

which confirms the optimal fourth order for the Modified Method 4a (MM₄^a) (8). Similarly, we can prove the optimal fourth order convergence for the Modified Method 4b (MM₄^b) (9). The proof of the theorem is completed. \square

2.2. Modified Families of Four-Parametric Four-Point Without-Memory Methods

Here, we examine the derivative-free families of biparametric four-point without-memory methods proposed in [14].

$$\begin{aligned}
 w_n &= s_n + \gamma\Omega(s_n), \\
 y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta\Omega(w_n)}, \\
 z_n &= y_n - \frac{\Omega(y_n)}{\xi(y_n)} \left[1 + \frac{1}{2} \left(\frac{\Omega(y_n)}{\Omega[y_n, w_n] + \beta\Omega(w_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right], \\
 s_{n+1} &= z_n - \frac{\Omega(z_n)}{\eta(z_n)} \left[1 + \frac{1}{2} \left(\frac{\Omega(z_n)}{\Omega[z_n, w_n] + \beta\Omega(w_n)} \right)^2 \frac{\psi(z_n)}{\Omega(z_n)} \right].
 \end{aligned} \tag{18}$$

$$\begin{aligned}
 w_n &= s_n + \gamma\Omega(s_n), \\
 y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta\Omega(w_n)}, \\
 z_n &= y_n - \frac{\Omega(y_n)}{\xi(y_n)} \left[1 + \frac{1}{2} \left(\frac{2\Omega(y_n)}{\Omega[y_n, w_n] + \xi(y_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right], \\
 s_{n+1} &= z_n - \frac{\Omega(z_n)}{\eta(z_n)} \left[1 + \frac{1}{2} \left(\frac{2\Omega(z_n)}{\Omega[z_n, w_n] + \xi(y_n)} \right)^2 \frac{\psi(z_n)}{\Omega(z_n)} \right],
 \end{aligned} \tag{19}$$

where $\eta(z_n) = \Omega[s_n, z_n] + (\Omega[s_n, y_n, w_n] - \Omega[s_n, z_n, w_n] - \Omega[s_n, y_n, z_n])(s_n - z_n)$ and $\psi(z_n) = 2\Omega[s_n, y_n, w_n]$.

Now, we introduce a new parameter $\theta \in \mathbb{R} - \{0\}$ through the modification of $\eta(z_n)$ as follows:

$$N(z_n) = \eta(z_n) + \theta(z_n - s_n)(z_n - y_n)(z_n - w_n). \tag{20}$$

Then, substituting the above Equation (20) as well as Equation (7) into Equations (18) and (19) yields two new four-parametric families of without-memory iterative methods. These modified methods, denoted as Modified Methods (MM), are defined as follows:

Modified Method 8a (MM^a):

$$\begin{aligned}
 w_n &= s_n + \gamma\Omega(s_n), \\
 y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta\Omega(w_n)}, \\
 z_n &= y_n - \frac{\Omega(y_n)}{M(y_n)} \left[1 + \frac{1}{2} \left(\frac{\Omega(y_n)}{\Omega[y_n, w_n] + \beta\Omega(w_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right], \\
 s_{n+1} &= z_n - \frac{\Omega(z_n)}{N(z_n)} \left[1 + \frac{1}{2} \left(\frac{\Omega(z_n)}{\Omega[z_n, w_n] + \beta\Omega(w_n)} \right)^2 \frac{\psi(z_n)}{\Omega(z_n)} \right].
 \end{aligned} \tag{21}$$

Modified Method 8b (MM^b):

$$\begin{aligned}
 w_n &= s_n + \gamma\Omega(s_n), \\
 y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta\Omega(w_n)}, \\
 z_n &= y_n - \frac{\Omega(y_n)}{M(y_n)} \left[1 + \frac{1}{2} \left(\frac{2\Omega(y_n)}{\Omega[y_n, w_n] + \xi(y_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right], \\
 s_{n+1} &= z_n - \frac{\Omega(z_n)}{N(z_n)} \left[1 + \frac{1}{2} \left(\frac{2\Omega(z_n)}{\Omega[z_n, w_n] + \xi(y_n)} \right)^2 \frac{\psi(z_n)}{\Omega(z_n)} \right].
 \end{aligned} \tag{22}$$

Modified methods MM^a and MM^b are optimal as per Kung–Traub’s conjecture, require four function evaluations per iteration, and exhibit an efficiency index of $8^{1/4} \approx 1.682$.

Next, we delve into the theoretical convergence analysis of the newly introduced modified methods, namely MM^a and MM^b, as outlined in the following theorem.

Theorem 2. Let an initial approximation s_0 be close enough to the root α of a sufficiently differentiable real function $\Omega : D \subseteq \mathbb{R} \rightarrow \mathbb{R}$, where D is an open interval. Then, the modified methods MM_8^a (21) and MM_8^b (22) exhibit the eighth order of convergence for any $\beta, \gamma, \lambda, \theta \in \mathbb{R} - \{0\}$. In addition, the methods MM_8^a and MM_8^b have the same error equation given by

$$\varepsilon_{n+1} = \frac{(1 + \Omega'(\alpha)\gamma)^4(\beta + d_2)^2(\lambda + \Omega'(\alpha)d_2^2 - \Omega'(\alpha)d_3)(-\theta + \Omega'(\alpha)d_4)}{\Omega'(\alpha)^2} \varepsilon_n^8 + O(\varepsilon_n^9), \tag{23}$$

where $d_j = \frac{1}{j!} \frac{\Omega^{(j)}(\alpha)}{\Omega'(\alpha)}$, $j = 2, 3, \dots$, and $\varepsilon_n = s_n - \alpha$ is the error at n^{th} iteration.

Proof. The proof of Modified Method 8a (MM_8^a):

Considering all the assumptions made in Theorems 1, from Equation (17), we have

$$z_n - \alpha = \frac{(1 + \Omega'(\alpha)\gamma)^2(\beta + d_2)(\lambda + \Omega'(\alpha)d_2^2 - \Omega'(\alpha)d_3)}{\Omega'(\alpha)} \varepsilon_n^4 + \sum_{i=5}^8 F_i \varepsilon_n^i + O(\varepsilon_n^9), \tag{24}$$

where $F_i, i = 5, 6, \dots, 8$ are functions of $\beta, \gamma, \Omega'(\alpha), d_2, d_3, \dots, d_8$.

Using the above Equation (24), we have

$$\Omega(z_n) = (1 + \Omega'(\alpha)\gamma)^2(\beta + d_2)(\lambda + \Omega'(\alpha)d_2^2 - \Omega'(\alpha)d_3) \varepsilon_n^4 + \sum_{i=5}^8 F_i \varepsilon_n^i + O(\varepsilon_n^9). \tag{25}$$

Applying Equations (11), (12), (15), (24) and (25), the approximation of $\Omega'(z_n)$ is obtained as follows:

$$N(z_n) = \Omega'(\alpha) + (1 + \Omega'(\alpha)\gamma)^2(\beta + d_2)(-\theta + 2\Omega'(\alpha)d_2^2 + 2d_2(\lambda - \Omega'(\alpha)d_3) + \Omega'(\alpha)d_4) \varepsilon_n^4 + \sum_{i=5}^8 G_i \varepsilon_n^i + O(\varepsilon_n^9), \tag{26}$$

where $G_i, i = 5, 6, \dots, 8$ are functions of $\beta, \gamma, \lambda, \theta, \Omega'(\alpha), d_2, d_3, \dots, d_8$.

Now, substituting the values of Equations (12), (24)–(26) in the last step of Equation (21), we obtain the error equation as follows:

$$\varepsilon_{n+1} = \frac{1}{\Omega'(\alpha)^2} (1 + \Omega'(\alpha)\gamma)^4(\beta + d_2)^2(\lambda + \Omega'(\alpha)d_2^2 - \Omega'(\alpha)d_3)(-\theta + \Omega'(\alpha)d_4) \varepsilon_n^8 + O(\varepsilon_n^9), \tag{27}$$

which confirms the optimal eight order for the Modified Method 8a (MM_8^a) (21). Similarly, we can prove the optimal eighth order convergence for the modified method MM_8^b (22). This completes the proof of the theorem. \square

Remark 1. From Theorems 1 and 2, the analysis of the error Equations (10) and (23) shows that the convergence order of the new modified derivative-free families of without-memory methods (MM_4^a and MM_4^b, MM_8^a and MM_8^b) can be increased significantly without any additional function evaluations using the free parameters γ, β, λ and θ , i.e., by putting $\gamma = -\frac{1}{\Omega'(\alpha)}, \beta = -d_2, \lambda = \Omega'(\alpha)d_3 - \Omega'(\alpha)d_2^2$ and $\theta = \Omega'(\alpha)d_4$.

However, the exact values of $\Omega'(\alpha), d_2, d_3$, and d_4 are not known to us. So, the parameters γ, β, λ , and θ have to be approximated using known information available from the current as well as the previous iterations. This will be the basis for extending the modified derivative-free families of without-memory methods into derivative-free with-memory methods.

3. New Families of Three- and Four-Parametric With-Memory Methods

In this section, we shall discuss the extension of the new modified derivative-free families of without-memory methods presented in Section 2 into their respective with-memory versions under two separate subsections. Using the available free parameters as accelerating parameters, we aim to increase the convergence order without any additional function evaluations per iteration thereby obtaining highly efficient multi-point with-memory methods.

Let us now discuss in detail the formulation of the methods, the approximations of the accelerating parameters, and the convergence analysis of the with-memory methods in the following subsections.

3.1. Three-Parametric Three-Point With-Memory Methods

Here, we introduce new derivative-free with-memory methods based on the newly suggested modified fourth order derivative-free families of without-memory methods MM_4^a (8) and MM_4^b (9).

From error Equation (10), the convergence order of the methods MM_4^a (8), and MM_4^b (9) can be increased from 4 to 8 without any additional function evaluation if we take $\gamma = -\frac{1}{\Omega'(\alpha)}$, $\beta = -d_2$ and $\lambda = \Omega'(\alpha)d_3 - \Omega'(\alpha)d_2^2$, where $d_2 = \frac{\Omega''(\alpha)}{2\Omega'(\alpha)}$, $d_3 = \frac{\Omega'''(\alpha)}{6\Omega'(\alpha)}$. However, the problem is that the exact values of $\Omega'(\alpha)$, $\Omega''(\alpha)$ and $\Omega'''(\alpha)$ are not available to us. So, we use the approximations $\gamma = \gamma_n$, $\beta = \beta_n$ and $\lambda = \lambda_n$, where γ_n , β_n and λ_n are the accelerating parameters computed using the available information from the current as well as the previous iterations such that the following conditions are satisfied:

$$\lim_{n \rightarrow \infty} \gamma_n = -\frac{1}{\Omega'(\alpha)}, \lim_{n \rightarrow \infty} \beta_n = -\frac{\Omega''(\alpha)}{2\Omega'(\alpha)} \text{ and } \lim_{n \rightarrow \infty} \lambda_n = \frac{\Omega'''(\alpha)}{6} - \frac{\Omega''(\alpha)^2}{4\Omega'(\alpha)}.$$

Now, we consider the following approximations for the accelerating parameters γ_n , β_n and λ_n .

$$\gamma_n = -\frac{1}{N_3'(s_n)}, \beta_n = -\frac{N_4''(w_n)}{2N_4'(w_n)}, \lambda_n = \frac{N_5'''(y_n)}{6} - \frac{N_5''(y_n)^2}{4N_5'(y_n)}, n = 0, 1, 2, \dots, \tag{28}$$

where $N_3(t)$, $N_4(t)$ and $N_5(t)$ are the respective Newton’s interpolating polynomials of third, fourth, and fifth degrees passing through the best saved points, i.e.,

$$\begin{aligned} N_3(t) &= N_3(t; s_n, y_{n-1}, w_{n-1}, s_{n-1}); \\ N_4(t) &= N_4(t; w_n, s_n, y_{n-1}, w_{n-1}, s_{n-1}); \\ N_5(t) &= N_5(t; y_n, w_n, s_n, y_{n-1}, w_{n-1}, s_{n-1}). \end{aligned}$$

Now, applying the approximations of the three accelerating parameters β_n , γ_n and λ_n from (28) in the methods MM_4^a (8) and MM_4^b (9), we obtain the following new derivative-free with-memory methods.

New With-Memory Method 4a (NWMM₄^a): For a given $s_0, \gamma_0, \beta_0, \lambda_0$, we have $w_0 = s_0 + \gamma_0\Omega(s_0)$. Then,

$$\begin{aligned} \gamma_n &= -\frac{1}{N_3'(s_n)}, \beta_n = -\frac{N_4''(w_n)}{2N_4'(w_n)}, \lambda_n = \frac{N_5'''(y_n)}{6} - \frac{N_5''(y_n)^2}{4N_5'(y_n)}, \\ w_n &= s_n + \gamma_n\Omega(s_n), \\ y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta_n\Omega(w_n)}, \\ s_{n+1} &= y_n - \frac{\Omega(y_n)}{M(y_n)} \left[1 + \frac{1}{2} \left(\frac{\Omega(y_n)}{\Omega[y_n, w_n] + \beta_n\Omega(w_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right]. \end{aligned} \tag{29}$$

New With-Memory Method 4b (NWMM₄^b): For a given $s_0, \gamma_0, \beta_0, \lambda_0$, we have $w_0 = s_0 + \gamma_0\Omega(s_0)$. Then,

$$\begin{aligned}
 \gamma_n &= -\frac{1}{\mathbb{N}'_3(s_n)}, \beta_n = -\frac{\mathbb{N}'_4(w_n)}{2\mathbb{N}'_4(w_n)}, \lambda_n = \frac{N'''_5(y_n)}{6} - \frac{N''_5(y_n)^2}{4N'_5(y_n)}, \\
 w_n &= s_n + \gamma_n \Omega(s_n), \\
 y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta_n \Omega(w_n)}, \\
 s_{n+1} &= y_n - \frac{\Omega(y_n)}{M(y_n)} \left[1 + \frac{1}{2} \left(\frac{2\Omega(y_n)}{\Omega[y_n, w_n] + \zeta(y_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right].
 \end{aligned} \tag{30}$$

In order to prove the convergence order of methods NWMM₄^a (29) and NWMM₄^b (30), we first present the following lemma.

Lemma 1. *If $\gamma_n = -\frac{1}{\mathbb{N}'_3(s_n)}$, $\beta_n = -\frac{1}{2} \frac{\mathbb{N}'_4(w_n)}{\mathbb{N}'_4(w_n)}$ and $\lambda_n = \frac{1}{6} N'''_5(y_n) - \frac{1}{4} \frac{N''_5(y_n)^2}{N'_5(y_n)}$, $n = 0, 1, 2, \dots$, then the following estimates*

$$1 + \gamma_n \Omega'(\alpha) \sim P_1 \varepsilon_{n-1,y} \varepsilon_{n-1,w} \varepsilon_{n-1} \sim \varepsilon_{n-1,y} \varepsilon_{n-1,w} \varepsilon_{n-1}, \tag{31}$$

$$\beta_n + d_2 \sim P_2 \varepsilon_{n-1,y} \varepsilon_{n-1,w} \varepsilon_{n-1} \sim \varepsilon_{n-1,y} \varepsilon_{n-1,w} \varepsilon_{n-1}, \tag{32}$$

$$\lambda_n + \Omega'(\alpha) d_2^2 - \Omega'(\alpha) d_3 \sim P_3 \varepsilon_{n-1,y} \varepsilon_{n-1,w} \varepsilon_{n-1} \sim \varepsilon_{n-1,y} \varepsilon_{n-1,w} \varepsilon_{n-1} \tag{33}$$

hold, where $\varepsilon_n = s_n - \alpha$, $\varepsilon_{n,y} = y_n - \alpha$, $\varepsilon_{n,w} = w_n - \alpha$, and P_1, P_2, P_3 are some asymptotic constants.

Proof. The proof is similar to Lemma 1 of [12]. □

Now, we state and prove the following theorem for obtaining the R-order of convergence [8] of the new three-point with-memory methods NWMM₄^a (29) and NWMM₄^b (30).

Theorem 3. *If an initial approximation s_0 is sufficiently close to the root α of $\Omega(s) = 0$, the parameters γ_n, β_n and λ_n are calculated by the expressions (28), then the R-order of convergence of the methods NWMM₄^a (29) and NWMM₄^b (30) is at least 7.5311.*

Proof. Let the sequence of approximations $\{s_n\}$ produced by the method NWMM₄^a (29) converges to the root α with order r . Then, we can write

$$\varepsilon_{n+1} \sim \varepsilon_n^r, \tag{34}$$

where $\varepsilon_n = s_n - \alpha$.

Then,

$$\varepsilon_n \sim \varepsilon_{n-1}^r. \tag{35}$$

Thus,

$$\varepsilon_{n+1} \sim \varepsilon_n^r = (\varepsilon_{n-1}^r)^r = \varepsilon_{n-1}^{r^2}. \tag{36}$$

Assuming the iterative sequences $\{w_n\}, \{y_n\}$ have orders r_1, r_2 , respectively, then using (34) and (35) gives

$$\varepsilon_{n,w} \sim \varepsilon_n^{r_1} = (\varepsilon_{n-1}^r)^{r_1} = \varepsilon_{n-1}^{rr_1}, \tag{37}$$

$$\varepsilon_{n,y} \sim \varepsilon_n^{r_2} = (\varepsilon_{n-1}^r)^{r_2} = \varepsilon_{n-1}^{rr_2}. \tag{38}$$

Using Theorem 1 and Lemma 1, we get

$$\varepsilon_{n,w} \sim (1 + \gamma_n \Omega'(\alpha)) \varepsilon_n = \varepsilon_{n-1}^{r+r_1+r_2+1}, \tag{39}$$

$$\varepsilon_{n,y} \sim (1 + \gamma_n \Omega'(\alpha)) (\beta_n + d_2) \varepsilon_n^2 = \varepsilon_{n-1}^{2r+2r_1+2r_2+2}, \tag{40}$$

$$\varepsilon_{n+1} \sim (1 + \gamma_n \Omega'(\alpha))^2 (\beta_n + d_2) (\lambda_n + \Omega'(\alpha) d_2^2 - \Omega'(\alpha) d_3) \varepsilon_n^4 = \varepsilon_{n-1}^{4r+4r_1+4r_2+4}. \tag{41}$$

Now, comparing the corresponding powers of ε_{n-1} on the right hand sides of (37), and (39), (38) and (40), (36) and (41), we get

$$\begin{aligned}
 rr_1 - r - r_1 - r_2 - 1 &= 0, \\
 rr_2 - 2r - 2r_1 - 2r_2 - 2 &= 0, \\
 r^2 - 4r - 4r_1 - 4r_2 - 4 &= 0.
 \end{aligned}
 \tag{42}$$

This system of equations has the non-trivial solution $r_1 = 1.8828, r_2 = 3.7656$ and $r = 7.5311$. Hence, the R-order of convergence of the method $NWMM_4^a$ (29) is at least 7.5311. The R-order of convergence for the methods $NWMM_4^b$ (30) can be proved in a similar manner. The proof is complete. \square

3.2. Four-Parametric Four-Point With-Memory Methods

Here, we introduce new derivative-free with-memory methods which are extensions of the newly suggested modified eighth order derivative-free families of without-memory methods MM_8^a (21) and MM_8^b (22).

It is evident from error Equation (23) that the convergence order of the methods MM_8^a (21), and MM_8^b (22) can be increased from 8 to 16 if we take $\gamma = -\frac{1}{\Omega'(\alpha)}, \beta = -d_2, \lambda = \Omega'(\alpha)d_3 - \Omega'(\alpha)d_2^2$ and $\theta = \Omega'(\alpha)d_4$, where $d_2 = \frac{\Omega''(\alpha)}{2\Omega'(\alpha)}, d_3 = \frac{\Omega'''(\alpha)}{6\Omega'(\alpha)}, d_4 = \frac{\Omega^{iv}(\alpha)}{24}$. In a similar manner to the previous subsection, we use the approximations $\gamma = \gamma_n, \beta = \beta_n, \lambda = \lambda_n$, and $\theta = \theta_n$, where $\gamma_n, \beta_n, \lambda_n$, and θ_n are the accelerating parameters computed using the available information from the current as well as the previous iterations such that the following conditions are satisfied:

$$\lim_{n \rightarrow \infty} \gamma_n = -\frac{1}{\Omega'(\alpha)}, \lim_{n \rightarrow \infty} \beta_n = -\frac{\Omega''(\alpha)}{2\Omega'(\alpha)}, \lim_{n \rightarrow \infty} \lambda_n = \frac{\Omega'''(\alpha)}{6} - \frac{\Omega''(\alpha)^2}{4\Omega'(\alpha)} \text{ and } \lim_{n \rightarrow \infty} \theta_n = \frac{\Omega^{iv}(\alpha)}{24}.$$

Now, we consider the following approximations for the accelerating parameters $\gamma_n, \beta_n, \lambda_n$ and θ_n .

$$\gamma_n = -\frac{1}{\mathbb{N}'_4(s_n)}, \beta_n = -\frac{\mathbb{N}''_5(w_n)}{2\mathbb{N}'_5(w_n)}, \lambda_n = \frac{\mathbb{N}'''_6(y_n)}{6} - \frac{\mathbb{N}''_6(y_n)^2}{4\mathbb{N}'_6(y_n)}, \text{ and } \theta_n = \frac{\mathbb{N}^{iv}_7(z_n)}{24}, n = 0, 1, 2, \dots,
 \tag{43}$$

where $\mathbb{N}_4(t), \mathbb{N}_5(t), \mathbb{N}_6(t)$, and $\mathbb{N}_7(t)$ are the respective Newton’s interpolating polynomials of fourth, fifth, sixth, and seventh degrees passing through the best saved points, i.e.,

$$\begin{aligned}
 \mathbb{N}_4(t) &= \mathbb{N}_4(t; s_n, z_{n-1}, y_{n-1}, w_{n-1}, s_{n-1}); \\
 \mathbb{N}_5(t) &= \mathbb{N}_5(t; w_n, s_n, z_{n-1}, y_{n-1}, w_{n-1}, s_{n-1}) \\
 \mathbb{N}_6(t) &= \mathbb{N}_6(t; y_n, w_n, s_n, z_{n-1}, y_{n-1}, w_{n-1}, s_{n-1}) \\
 \mathbb{N}_7(t) &= \mathbb{N}_7(t; z_n, y_n, w_n, s_n, z_{n-1}, y_{n-1}, w_{n-1}, s_{n-1}).
 \end{aligned}$$

Now, applying the approximations of the four accelerating parameters $\gamma_n, \beta_n, \lambda_n$, and θ_n from (43) in the modified methods MM_8^a (21) and MM_8^b (22), we obtain the following new derivative-free with-memory methods.

New With-Memory Method 8a (NWMM₈^a): For a given $s_0, \gamma_0, \beta_0, \lambda_0, \theta_0$, we have $w_0 = s_0 + \gamma_0\Omega(s_0)$. Then,

$$\begin{aligned}
 \gamma_n &= -\frac{1}{\mathbb{N}'_4(s_n)}, \beta_n = -\frac{\mathbb{N}''_5(w_n)}{2\mathbb{N}'_5(w_n)}, \lambda_n = \frac{\mathbb{N}'''_6(y_n)}{6} - \frac{\mathbb{N}''_6(y_n)^2}{4\mathbb{N}'_6(y_n)}, \theta_n = \frac{\mathbb{N}^{iv}_7(z_n)}{24}, \\
 w_n &= s_n + \gamma_n\Omega(s_n), \\
 y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta_n\Omega(w_n)}, \\
 z_n &= y_n - \frac{\Omega(y_n)}{M(y_n)} \left[1 + \frac{1}{2} \left(\frac{\Omega(y_n)}{\Omega[y_n, w_n] + \beta_n\Omega(w_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right] \\
 s_{n+1} &= z_n - \frac{\Omega(z_n)}{N(z_n)} \left[1 + \frac{1}{2} \left(\frac{\Omega(z_n)}{\Omega[z_n, w_n] + \beta_n\Omega(w_n)} \right)^2 \frac{\psi(z_n)}{\Omega(z_n)} \right].
 \end{aligned}
 \tag{44}$$

New With-Memory Method 8b (NWMM₈^b): For a given $s_0, \gamma_0, \beta_0, \lambda_0, \theta_0$, we have $w_0 = s_0 + \gamma_0\Omega(s_0)$. Then,

$$\begin{aligned} \gamma_n &= -\frac{1}{N'_4(s_n)}, \beta_n = -\frac{N''_5(w_n)}{2N'_5(w_n)}, \lambda_n = \frac{N'''_6(y_n)}{6} - \frac{N''_6(y_n)^2}{4N'_6(y_n)}, \theta_n = \frac{N^{iv}_7(z_n)}{24}, \\ w_n &= s_n + \gamma_n\Omega(s_n), \\ y_n &= s_n - \frac{\Omega(s_n)}{\Omega[s_n, w_n] + \beta_n\Omega(w_n)}, \\ z_n &= y_n - \frac{\Omega(y_n)}{M(y_n)} \left[1 + \frac{1}{2} \left(\frac{2\Omega(y_n)}{\Omega[y_n, w_n] + \xi(y_n)} \right)^2 \frac{\rho(y_n)}{\Omega(y_n)} \right], \\ s_{n+1} &= z_n - \frac{\Omega(z_n)}{N(z_n)} \left[1 + \frac{1}{2} \left(\frac{2\Omega(z_n)}{\Omega[z_n, w_n] + \xi(y_n)} \right)^2 \frac{\psi(z_n)}{\Omega(z_n)} \right]. \end{aligned} \tag{45}$$

In order to prove the convergence order of methods NWMM₈^a (44) and NWMM₈^b (45), we first present the following lemma.

Lemma 2. If $\gamma_n = -\frac{1}{N'_4(s_n)}, \beta_n = -\frac{N''_5(w_n)}{2N'_5(w_n)}, \lambda_n = \frac{N'''_6(y_n)}{6} - \frac{N''_6(y_n)^2}{4N'_6(y_n)}$, and $\theta_n = \frac{N^{iv}_7(z_n)}{24}$, $n = 0, 1, 2, \dots$, then the following estimates

$$\begin{aligned} 1 + \gamma_n\Omega'(\alpha) &\sim Q_1\varepsilon_{n-1,z}\varepsilon_{n-1,y}\varepsilon_{n-1,w}\varepsilon_{n-1} \sim \varepsilon_{n-1,z}\varepsilon_{n-1,y}\varepsilon_{n-1,w}\varepsilon_{n-1}, \tag{46} \\ \beta_n + d_2 &\sim Q_2\varepsilon_{n-1,z}\varepsilon_{n-1,y}\varepsilon_{n-1,w}\varepsilon_{n-1} \sim \varepsilon_{n-1,z}\varepsilon_{n-1,y}\varepsilon_{n-1,w}\varepsilon_{n-1}, \tag{47} \\ \lambda_n + \Omega'(\alpha)d_2^2 - \Omega'(\alpha)d_3 &\sim Q_3\varepsilon_{n-1,z}\varepsilon_{n-1,y}\varepsilon_{n-1,w}\varepsilon_{n-1} \sim \varepsilon_{n-1,z}\varepsilon_{n-1,y}\varepsilon_{n-1,w}\varepsilon_{n-1}, \tag{48} \\ \theta_n - \Omega'(\alpha)d_4 &\sim Q_4\varepsilon_{n-1,z}\varepsilon_{n-1,y}\varepsilon_{n-1,w}\varepsilon_{n-1} \sim \varepsilon_{n-1,z}\varepsilon_{n-1,y}\varepsilon_{n-1,w}\varepsilon_{n-1} \tag{49} \end{aligned}$$

hold, where $\varepsilon_n = s_n - \alpha$, $\varepsilon_{n,y} = y_n - \alpha$, $\varepsilon_{n,w} = w_n - \alpha$, and Q_1, Q_2, Q_3, Q_4 are some asymptotic constants.

Proof. The proof is similar to Lemma 1 of [12]. □

Now, we state and prove the following theorem for obtaining the R-order of convergence [8] of the new four-point with-memory methods NWMM₈^a (44) and NWMM₈^b (45).

Theorem 4. If an initial approximation s_0 is sufficiently close to the root α of $\Omega(s) = 0$, the parameters $\gamma_n, \beta_n, \lambda_n$, and θ_n are calculated by the expressions (43), then the R-order of convergence of the methods NWMM₈^a (44) and NWMM₈^b (45) is at least 15.5156.

Proof. Let the sequence of approximations $\{s_n\}$ produced by the method NWMM₈^a (44) converges to the root α with order r . Then, we can write

$$\varepsilon_{n+1} \sim \varepsilon_n^r, \tag{50}$$

where $\varepsilon_n = s_n - \alpha$.

Then,

$$\varepsilon_n \sim \varepsilon_{n-1}^r. \tag{51}$$

Thus,

$$\varepsilon_{n+1} \sim \varepsilon_n^r = (\varepsilon_{n-1}^r)^r = \varepsilon_{n-1}^{r^2}. \tag{52}$$

Assuming the iterative sequences $\{w_n\}$, $\{y_n\}$ and $\{z_n\}$ have orders r_1, r_2 , and r_3 , respectively, then using (50) and (51) gives

$$\varepsilon_{n,w} \sim \varepsilon_n^{r_1} = (\varepsilon_{n-1}^r)^{r_1} = \varepsilon_{n-1}^{rr_1}, \tag{53}$$

$$\varepsilon_{n,y} \sim \varepsilon_n^{r_2} = (\varepsilon_{n-1}^r)^{r_2} = \varepsilon_{n-1}^{rr_2}, \tag{54}$$

$$\varepsilon_{n,z} \sim \varepsilon_n^{r_3} = (\varepsilon_{n-1}^r)^{r_3} = \varepsilon_{n-1}^{rr_3}. \tag{55}$$

Using Theorem 2 and Lemma 2, we get

$$\varepsilon_{n,w} \sim (1 + \gamma_n \Omega'(\alpha)) \varepsilon_n = \varepsilon_{n-1}^{r+r_1+r_2+r_3+1}, \tag{56}$$

$$\varepsilon_{n,y} \sim (1 + \gamma_n \Omega'(\alpha)) (\beta + d_2) \varepsilon_n^2 = \varepsilon_{n-1}^{2r+2r_1+2r_2+2r_3+2}, \tag{57}$$

$$\varepsilon_{n,z} \sim (1 + \gamma_n \Omega'(\alpha))^2 (\beta_n + d_2) (\lambda_n + \Omega'(\alpha) d_2^2 - \Omega'(\alpha) d_3) \varepsilon_n^4 = \varepsilon_{n-1}^{4r+4r_1+4r_2+4r_3+4}, \tag{58}$$

$$\varepsilon_{n+1} \sim (1 + \gamma_n \Omega'(\alpha))^4 (\beta + d_2)^2 (\lambda_n + \Omega'(\alpha) d_2^2 - \Omega'(\alpha) d_3) (\theta_n - \Omega'(\alpha) d_4) \varepsilon_n^8 = \varepsilon_{n-1}^{8r+8r_1+8r_2+8r_3+8}. \tag{59}$$

Now, comparing the corresponding powers of ε_{n-1} on the right hand sides of (53) and (56), (54) and (57), (55) and (58), (52) and (59), we get

$$\begin{aligned} rr_1 - r - r_1 - r_2 - r_3 - 1 &= 0, \\ rr_2 - 2r - 2r_1 - 2r_2 - 2r_3 - 2 &= 0, \\ rr_3 - 4r - 4r_1 - 4r_2 - 4r_3 - 4 &= 0, \\ r^2 - 8r - 8r_1 - 8r_2 - 8r_3 - 8 &= 0. \end{aligned} \tag{60}$$

This system of equations has the non-trivial solution $r_1 = 1.9394, r_2 = 3.8789, r_3 = 7.7578$ and $r = 15.5156$. Hence, the R-order of convergence of the method $NWMM_8^a$ (44) is at least 15.5156. The R-order of convergence for the methods $NWMM_8^b$ can be proved in similar manner. The proof is complete. \square

4. Numerical Experiments

In this section, we examine the performance and the computational efficiency of the newly developed with and without-memory methods discussed in Sections 2 and 3 and compare with some methods of similar nature available in the literature. In particular, we have considered for the comparison the following derivative-free three-parametric methods: $FZM_4(4.1)$ [15], $VTM_4(28)$ [16], and $SM_4(4.1)$ [17], and the following four-parametric methods: AJM_8 [13], ZM_8 (ZR1 from [18]), and ACM_8 (M1 from [19]).

All numerical tests have been executed using the multi-precision arithmetic programming software Mathematica 12.2. For all methods, we have chosen the same values of the parameters $\gamma_0 = \beta_0 = \lambda_0 = \theta_0 = -1$ in all the test functions in order to start the initial iteration. These same values are used for the corresponding parameters of all the compared methods in order to have uniform and fair comparison in all the test functions.

Numerical test functions which comprise a standard academic example and some real-life chemical engineering problems along with their simple roots (α) and initial guesses (s_0) are presented below.

Example 1. A standard academic test function given by

$$\Omega_1(s) = e^{-s^2} (1 + s^3 + s^6)(s - 2). \tag{61}$$

It has a simple root $\alpha = 2$. We use $s_0 = 2.3$ as the initial guess and the results are displayed in Table 1.

Table 1. Comparison of test function $\Omega_1(s)$.

Methods	n	$ s_1 - s_0 $	$ s_2 - s_1 $	$ s_3 - s_2 $	$ \Omega_1(s_3) $	COC
Without memory						
FZM ₄	5	0.33009	0.030086	9.2612×10^{-7}	9.2120×10^{-25}	4.0000
VTM ₄	5	0.31989	0.019888	2.0505×10^{-7}	2.6989×10^{-27}	4.0000
SM ₄	5	0.33009	0.030086	9.2612×10^{-7}	9.2120×10^{-25}	4.0000
MM ₄ ^a	5	0.29425	0.0057493	4.0008×10^{-10}	1.2989×10^{-38}	4.0000
MM ₄ ^b	5	0.30077	0.00076940	1.3367×10^{-13}	1.6187×10^{-52}	4.0000
ZM ₈	4	0.29937	0.00062516	6.3288×10^{-26}	9.4210×10^{-202}	8.0000
ACM ₈	4	0.30459	0.0045854	1.0110×10^{-20}	7.6376×10^{-162}	8.0000
AJM ₈	4	0.28708	0.012920	7.4714×10^{-15}	1.4745×10^{-112}	8.0000
MM ₈ ^a	4	0.30000	1.2807×10^{-6}	1.6291×10^{-48}	1.4926×10^{-383}	8.0000
MM ₈ ^b	4	0.30000	2.2530×10^{-6}	1.4939×10^{-46}	7.4639×10^{-368}	8.0000
With memory						
FZM ₄	4	0.33009	0.030087	9.0157×10^{-15}	2.2995×10^{-100}	7.6456
VTM ₄	4	0.31989	0.019888	1.3343×10^{-15}	1.0917×10^{-108}	7.6031
SM ₄	4	0.33009	0.030087	1.4778×10^{-13}	5.4594×10^{-90}	7.2871
NWMM ₄ ^a	4	0.29425	0.0057493	6.0134×10^{-18}	2.7847×10^{-131}	7.5271
NWMM ₄ ^b	4	0.30077	0.00076940	4.1128×10^{-26}	2.1806×10^{-189}	7.5603
ZM ₈	4	0.29937	0.00062516	3.6733×10^{-50}	4.7977×10^{-749}	15.145
ACM ₈	4	0.30459	0.0045854	1.6369×10^{-39}	1.3451×10^{-598}	15.541
AJM ₈	4	0.28708	0.012920	8.6225×10^{-26}	1.4868×10^{-350}	14.000
NWMM ₈ ^a	3	0.30000	1.2807×10^{-6}	4.4603×10^{-96}	1.7660×10^{-1486}	15.544
NWMM ₈ ^b	3	0.30000	2.2530×10^{-6}	6.8171×10^{-91}	3.0895×10^{-1398}	15.470

Example 2. The Michaelis–Menten model [20] describes the kinetics of enzyme-mediated reactions and has the following expression:

$$\frac{dS}{dt} = -v_m \frac{S}{K_s + S}, \tag{62}$$

where S is the substrate concentration (moles/L), v_m is the maximum uptake rate (moles/L/d), and K_s is the half-saturation constant, which is the substrate level at which uptake is half of the maximum (moles/L).

If S_0 is the initial substrate level at $t = 0$, then the above equation can be solved for S as follows:

$$S = S_0 - v_m t + K_s \log(S_0/S). \tag{63}$$

For a particular case where $t = 10$, $S_0 = 8$ moles/L, $v_m = 0.7$ moles/L/d, and $K_s = 2.5$ moles/L, the above equation reduces to the following nonlinear function.

$$\Omega_2(s) = s - 2.5 \log\left(\frac{8}{s}\right) - 1, \tag{64}$$

where s denotes the substrate concentration S to be determined. The nonlinear equation $\Omega_2(s) = 0$ has a simple root $\alpha \approx 3.2511115053800575$. We use $s_0 = 3.8$ as the initial guess and the results are displayed in Table 2.

Example 3. Let us consider the conversion of the fraction of Nitrogen–Hydrogen feed into Ammonia, called fractional conversion, at a pressure of 250 atm and temperature of 500 °C (see [21] for details). When reduced to the polynomial form, the problem has the following expression:

$$\Omega_3(s) = s^4 - 7.79075s^3 + 14.7445s^2 + 2.511s - 1.674. \tag{65}$$

The nonlinear equation $\Omega_3(s) = 0$ has a simple root $\alpha \approx 0.27775954284172066$. We take $s_0 = 0.6$ as the initial guess and the results are displayed in Table 3.

Table 2. Comparison of test function $\Omega_2(s)$.

Methods	n	$ s_1 - s_0 $	$ s_2 - s_1 $	$ s_3 - s_2 $	$ \Omega_2(s_3) $	COC
Without memory						
FZM ₄	5	0.52278	0.026103	3.6376×10^{-7}	2.5627×10^{-26}	4.0000
VTM ₄	5	0.52808	0.020806	1.0009×10^{-7}	9.7869×10^{-29}	4.0000
SM ₄	5	0.52278	0.026103	3.6376×10^{-7}	2.5627×10^{-26}	4.0000
MM ₄ ^a	5	0.53110	0.017791	3.5680×10^{-8}	1.0391×10^{-30}	4.0000
MM ₄ ^b	5	0.53031	0.018583	4.2589×10^{-8}	2.1093×10^{-30}	4.0000
ZM ₈	4	0.54884	0.000043546	2.1311×10^{-36}	1.2407×10^{-286}	8.0000
ACM ₈	4	0.54976	0.00086944	3.0625×10^{-25}	1.2791×10^{-196}	8.0000
AJM ₈	4	0.54818	0.00070892	1.1467×10^{-26}	9.5186×10^{-209}	8.0000
MM ₈ ^a	4	0.54923	0.00034385	2.5135×10^{-29}	3.6221×10^{-230}	8.0000
MM ₈ ^b	4	0.54924	0.00034685	2.6942×10^{-29}	6.3116×10^{-230}	8.0000
With memory						
FZM ₄	4	0.52278	0.026104	1.8327×10^{-20}	7.9524×10^{-153}	7.5500
VTM ₄	4	0.52808	0.020806	3.0179×10^{-21}	9.7288×10^{-159}	7.5491
SM ₄	4	0.52278	0.026104	8.1149×10^{-20}	5.7816×10^{-140}	7.2839
NWMM ₄ ^a	4	0.53110	0.017791	1.1641×10^{-21}	5.3718×10^{-162}	7.5518
NWMM ₄ ^b	4	0.53031	0.018583	1.6423×10^{-21}	7.3086×10^{-161}	7.5520
ZM ₈	3	0.54884	0.000043546	1.4128×10^{-80}	1.4045×10^{-1223}	15.145
ACM ₈	3	0.54976	0.00086944	3.5527×10^{-64}	4.1590×10^{-995}	15.420
AJM ₈	4	0.54818	0.00070892	2.1494×10^{-58}	4.0336×10^{-814}	14.000
NWMM ₈ ^a	3	0.54923	0.00034385	1.7197×10^{-70}	5.4863×10^{-1093}	15.426
NWMM ₈ ^b	3	0.54924	0.00034685	1.9715×10^{-70}	4.5987×10^{-1092}	15.426

Table 3. Comparison of test function $\Omega_3(s)$.

Methods	n	$ s_1 - s_0 $	$ s_2 - s_1 $	$ s_3 - s_2 $	$ \Omega_3(s_3) $	COC
Without memory						
FZM ₄	-	<i>Divergent</i>	<i>Divergent</i>	<i>Divergent</i>	<i>Divergent</i>	-
VTM ₄	-	<i>Divergent</i>	<i>Divergent</i>	<i>Divergent</i>	<i>Divergent</i>	-
SM ₄	-	<i>Divergent</i>	<i>Divergent</i>	<i>Divergent</i>	<i>Divergent</i>	-
MM ₄ ^a	6	0.28422	0.037789	2.2754×10^{-4}	6.5335×10^{-14}	4.0000
MM ₄ ^b	6	0.28258	0.039391	2.7277×10^{-4}	1.3208×10^{-13}	4.0000
ZM ₈	-	<i>Divergent</i>	<i>Divergent</i>	<i>Divergent</i>	<i>Divergent</i>	-
ACM ₈	5	0.28268	0.039558	1.7645×10^{-6}	1.2870×10^{-45}	8.0000
AJM ₈	4	0.32808	0.0058362	4.7720×10^{-16}	8.2224×10^{-121}	8.0000
MM ₈ ^a	4	0.32322	0.00098388	2.3084×10^{-24}	9.2316×10^{-189}	8.0000
MM ₈ ^b	4	0.32335	0.0011099	6.9239×10^{-24}	6.0483×10^{-185}	8.0000
With memory						
FZM ₄	5	0.021086	0.34332	5.6284×10^{-6}	2.5093×10^{-42}	7.7424
VTM ₄	5	0.024773	0.34701	5.9869×10^{-6}	4.1983×10^{-42}	7.7424
SM ₄	5	0.021086	0.34332	5.1371×10^{-6}	9.9324×10^{-43}	7.7423
NWMM ₄ ^a	4	0.28422	0.038016	8.7889×10^{-14}	3.8698×10^{-105}	7.7335
NWMM ₄ ^b	4	0.28258	0.039664	2.1998×10^{-13}	3.8297×10^{-102}	7.7325
ZM ₈	5	0.59976	0.28631	8.7926×10^{-3}	9.9619×10^{-33}	16.000
ACM ₈	4	0.28268	0.039560	1.2422×10^{-23}	3.3101×10^{-366}	16.000
AJM ₈	4	0.32808	0.0058362	2.8373×10^{-31}	9.0156×10^{-427}	14.000
NWMM ₈ ^a	4	0.32322	0.00098388	1.7162×10^{-51}	1.9333×10^{-862}	17.000
NWMM ₈ ^b	4	0.32335	0.0011099	1.3326×10^{-50}	2.6215×10^{-847}	17.000

Example 4. The equation of state for a van der Waals fluid takes the following form [22]:

$$\left(P + \frac{a}{V^2}\right)(V - b) = RT, \tag{66}$$

where a, b , and R are positive constants, P is the pressure, T is the absolute temperature, and V is the molar volume.

Now, let us substitute $p = \frac{P}{P_c} = \frac{27b^2P}{a}$, $t = \frac{T}{T_c} = \frac{27RbT}{8a}$ and $v = \frac{V}{V_c} = \frac{v}{3b}$, where $P_c = \frac{a}{27b^2}$ is the critical pressure, $T_c = \frac{8a}{27Rb}$ is the critical temperature, and $V_c = 3b$ is the critical molar volume.

Then, the above Equation (66), in which the pressure, temperature, and volume are expressed in terms of their critical values, becomes

$$\left(p + \frac{3}{v^2}\right)(3v - 1) = 8t, \tag{67}$$

where p, t , and v are called the reduced pressure, temperature, and volume, respectively. For particular values of $p = 6$ and $t = 2$, Equation (67) reduces to the following nonlinear equation.

$$\Omega_4(s) = 18s^3 - 22s^2 + 9s - 3 = 0, \tag{68}$$

where s represents the reduced volume v to be determined. Equation (68) has a simple root $\alpha \approx 0.86728815393727851$. We use $s_0 = 1.2$ as the initial guess and the results are displayed in Table 4.

Table 4. Comparison of test function $\Omega_4(s)$.

Methods	n	$ s_1 - s_0 $	$ s_2 - s_1 $	$ s_3 - s_2 $	$ \Omega_4(s_3) $	COC
Without memory						
FZM ₄	—	Divergent	Divergent	Divergent	Divergent	—
VTM ₄	—	Divergent	Divergent	Divergent	Divergent	—
SM ₄	—	Divergent	Divergent	Divergent	Divergent	—
MM ₄ ^a	6	0.21298	0.10244	0.017273	1.6018×10^{-4}	4.0000
MM ₄ ^b	6	0.21292	0.10228	0.017481	3.0615×10^{-4}	4.0000
ZM ₈	5	0.42172	0.088533	4.7739×10^{-4}	8.4089×10^{-20}	8.0000
ACM ₈	5	0.21344	0.11153	7.7426×10^{-3}	2.1579×10^{-8}	8.0000
AJM ₈	4	0.31961	0.013099	3.3791×10^{-9}	2.5912×10^{-61}	8.0000
MM ₈ ^a	4	0.31144	0.021272	8.9028×10^{-10}	1.9552×10^{-68}	8.0000
MM ₈ ^b	4	0.31130	0.021408	7.1073×10^{-10}	3.2256×10^{-69}	8.0000
With memory						
FZM ₄	5	0.0039485	0.32818	5.8627×10^{-4}	8.0061×10^{-24}	8.0000
VTM ₄	5	0.0045984	0.32754	5.7705×10^{-4}	7.0536×10^{-24}	8.0000
SM ₄	5	0.0039485	0.32818	5.8627×10^{-4}	8.0061×10^{-24}	8.0000
NWMM ₄ ^a	5	0.21298	0.11973	4.4869×10^{-7}	4.6843×10^{-49}	8.0000
NWMM ₄ ^b	5	0.21292	0.11979	4.2775×10^{-7}	3.1960×10^{-49}	8.0000
ZM ₈	4	0.42172	0.089010	2.2383×10^{-13}	2.4905×10^{-198}	16.0000
ACM ₈	4	0.21344	0.11927	1.7133×10^{-12}	3.4570×10^{-184}	16.0000
AJM ₈	4	0.31961	0.013099	1.0088×10^{-23}	3.6015×10^{-318}	14.0000
NWMM ₈ ^a	4	0.31144	0.021272	2.9242×10^{-26}	9.3185×10^{-431}	17.0000
NWMM ₈ ^b	4	0.31130	0.021408	2.8163×10^{-26}	4.9186×10^{-431}	17.0000

All the results and analysis of the numerical computations are displayed in Tables 1–4. The aim of these tables is to showcase the performance of the iterative methods in terms of their convergence speed and accuracy. We measure the convergence by tracking the number of iterations (n) required to satisfy the stopping criterion:

$$|s_n - s_{n-1}| + |\Omega(s_n)| < 10^{-60}, \tag{69}$$

where s_n represents the current iterate and $|\Omega(s_n)|$ denotes the absolute residual error of the function. To provide further insights, we also include in the tables the estimated error

in consecutive iterations, $|s_n - s_{n-1}|$, for the initial three iterations. Moreover, we calculate the computational order of convergence (COC) using the formula [23]:

$$\text{COC} = \frac{\log|\Omega(s_n)/\Omega(s_{n-1})|}{\log|\Omega(s_{n-1})/\Omega(s_{n-2})|}. \tag{70}$$

From Tables 1–4, the numerical results reveal the good performance and better efficiency of the proposed with and without-memory methods, thus confirming their theoretical results. The proposed methods show better accuracy with high efficiency in terms of minimal errors after three iterations as compared to the existing methods in comparison. Tables 2–4 confirm the applicability of the proposed families of methods when applied to some real world chemical problems. In addition, some of the compared methods fail to converge to the required roots and diverge away from the roots, which is not the case for the proposed families of methods, as can be observed from Figures 1–4. Further, the numerical test results reveal that the COC supports the theoretical convergence order of the new proposed with and without-memory methods in the test functions.

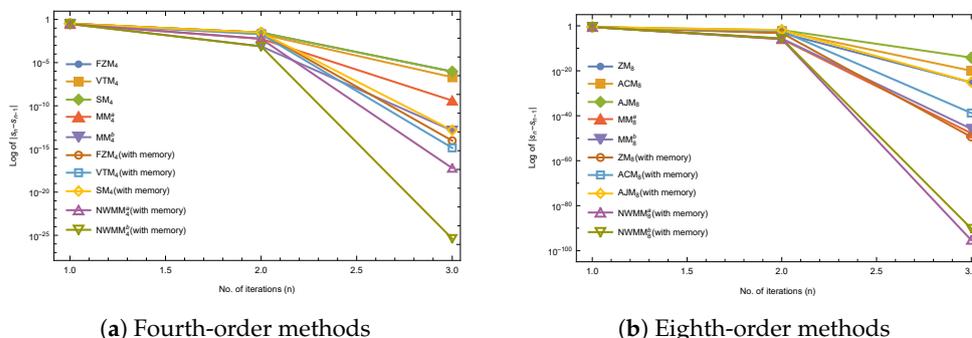


Figure 1. Graphical comparison based on Log of $|s_n - s_{n-1}|$ for $\Omega_1(s)$.

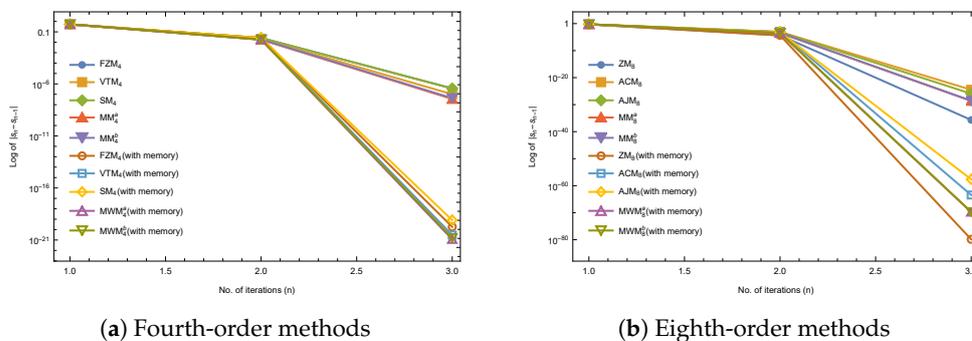


Figure 2. Graphical comparison based on Log of $|s_n - s_{n-1}|$ for $\Omega_2(s)$.

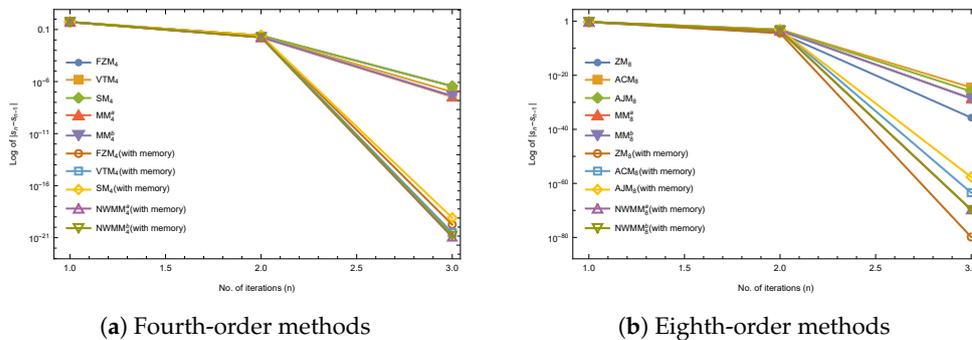


Figure 3. Graphical comparison based on Log of $|s_n - s_{n-1}|$ for $\Omega_3(s)$.

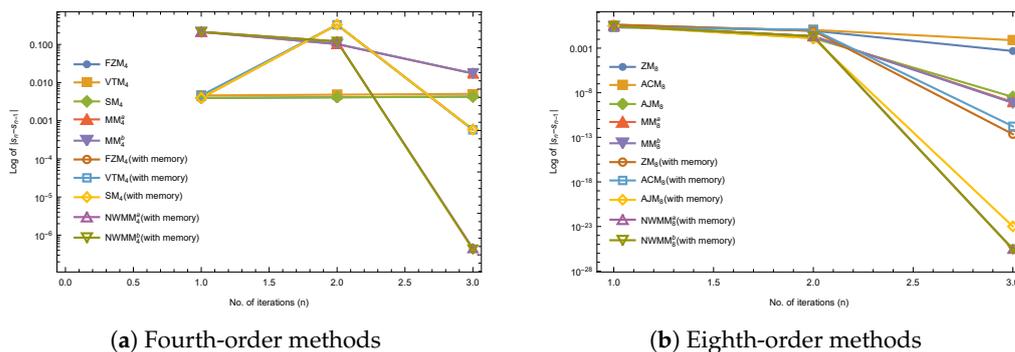


Figure 4. Graphical comparison based on Log of $|s_n - s_{n-1}|$ for $\Omega_4(s)$.

Comparison by Basins of Attraction

In this section, we explore the dynamical properties of the proposed methods discussed in Section 2. To analyse their behaviour in the complex plane, we examine the basins of attraction associated with each method. Specifically, we compare MM_4^a with Method (5), MM_4^b with Method (6), MM_8^a with Method (18), and MM_8^b with Method (19), respectively.

We used a 401×401 grid to represent the complex plane region $R = [-2, 2] \times [-2, 2]$. Each point z_0 in R was assigned a colour based on the root it converged to using an iterative method. Divergent points were marked in black if they failed to converge within 100 iterations or within a tolerance of 10^{-4} . Simple roots were represented by white circles. Brighter colours indicated faster convergence, while darker colours indicated slower convergence. In Figure 5, we illustrate the basins of attraction obtained by applying the fourth and eighth-order methods to the function $p(z) = z^3 + z$. To have a fair comparison, we take the same values of the parameters $\gamma = \beta = 0.001$ for all compared methods.

In Figure 5, it is evident that all the compared methods exhibit large basins of attraction with only a few divergent points. However, the proposed modified methods outperform the biparametric methods due to the inclusion of additional parameters. In fact, the proposed methods MM_4^b and MM_8^b are the best with no divergent points.

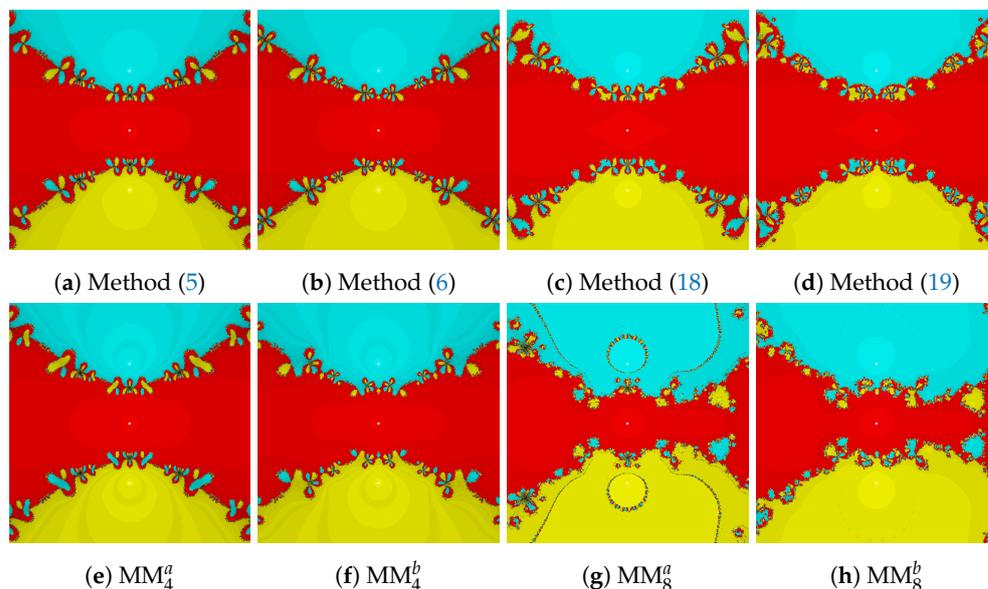


Figure 5. Basins of attraction for the proposed methods and the biparametric methods applied to the function $p(z) = z^3 + z$.

Moreover, we can observe from Tables 5 and 6 that each of the proposed methods show significant improvements in terms of fewer divergent points. In particular, MM_4^a shows an improvement of 71.4% over method (5), and 100% improvement for method MM_4^b over method (6). Similarly, MM_8^a and MM_8^b show 48.1% and 100% improvements

over methods (18) and (19), respectively. Notably, both methods MM_4^a and MM_4^b show no divergent points, as observed from Tables 5 and 6, respectively. This underscores the crucial role of the extra parameters in enhancing stability and reducing divergent points in the proposed methods.

Table 5. Number of divergent points on the function $p(z)$ for the fourth-order methods.

Methods	Method (5) and MM_4^a			Method (6) and MM_4^b		
	Method (5)	MM_4^a	Improv. (%)	Method (6)	MM_4^b	Improv. (%)
Parameters	γ, β	$\gamma, \beta, \lambda = 0.3$		γ, β	$\gamma, \beta, \lambda = -1$	
No. of div. pts.	105	30	71.4%	14	0	100%
% of div. pts.	0.065%	0.019%		0.0087%	0%	

Abbreviations used: div. = divergent, pts. = points, Improv. = Improvement.

Table 6. Number of divergent points on the function $p(z)$ for the eighth-order methods.

Methods	Method (18) and MM_8^a			Method (19) and MM_8^b		
	Method (18)	MM_8^a	Improv. (%)	Method (19)	MM_8^b	Improv. (%)
Parameters	γ, β	$\gamma, \beta, \lambda = -4, \theta = 0.4$		γ, β	$\gamma, \beta, \lambda = -4, \theta = 0.4$	
No. of div. pts.	79	41	48.1%	14	0	100%
% of div. pts.	0.049%	0.025%		0.0087%	0%	

5. Concluding Remarks

In this paper, we have presented new derivative-free three- and four-parametric with and without-memory methods for finding simple roots of nonlinear equations. The methods are based on the modifications of the derivative-free without-memory methods developed in [14]. The use of accelerating parameters in the with-memory methods has enabled us to increase the convergence order of the without-memory methods and obtain very high computational efficiency index of $7.5311^{1/3} \approx 1.9601$ for the three-point and $15.5156^{1/4} \approx 1.9847$ for the four-point with-memory methods. The numerical test results have demonstrated the good performance and applicability of the proposed with and without-memory methods. They are found to have better accuracy and efficiency as compared to the existing methods in the comparison in terms of minimal residual errors and errors in consecutive iterations for convergence towards the required simple roots in minimal number of iterations. Moreover, the study of the dynamical aspects through the basins of attraction further confirms the crucial role of the extra parameters in enhancing stability and reducing divergent points in the proposed methods.

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