

**High-order Newton-type iterative methods with memory for solving nonlinear equations\***XIAOFENG WANG<sup>1,†</sup> AND TIE ZHANG<sup>2</sup><sup>1</sup> *School of Mathematics and Physics, Bohai University, Jinzhou 121 013, Liaoning, China*<sup>2</sup> *College of Sciences, Northeastern University, Shenyang 110 819 Liaoning, China*

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**Abstract.** In this paper, we present a new family of two-step Newton-type iterative methods with memory for solving nonlinear equations. In order to obtain a Newton-type method with memory, we first present an optimal two-parameter fourth-order Newton-type method without memory. Then, based on the two-parameter method without memory, we present a new two-parameter Newton-type method with memory. Using two self-correcting parameters calculated by Hermite interpolatory polynomials, the  $R$ -order of convergence of a new Newton-type method with memory is increased from 4 to 5.7016 without any additional calculations. Numerical comparisons are made with some known methods by using the basins of attraction and through numerical computations to demonstrate the efficiency and the performance of the presented methods.

**AMS subject classifications:** 65H05, 65B99**Key words:** Newton-type iterative method with memory, nonlinear equations,  $R$ -order convergence, root-finding**1. Introduction**

Finding the root of a nonlinear equation  $f(x) = 0$  is a classical problem in scientific computation. Recently, many iterative methods have been proposed for solving nonlinear equations, see [1-14] and the references therein. In these methods, the iterative methods with memory are worth studying. The iterative method with memory can improve the order of convergence of the method without memory without any additional calculations and it has a very high computational efficiency. There are two kinds of iterative methods with memory, i.e. Steffensen-type method and Newton-type method. In this paper, we only consider the Newton-type method with memory. For example, in [9] Petković et al. proposed a two-step Newton-type iterative method with memory of the  $R$ -order 4.562 by using inverse interpolation, which is written as

$$\begin{cases} N(x_n) = x_n - \frac{f(x_n)}{f'(x_n)}, \quad \varphi(t) = \frac{1}{f(t) - f(x_n)} \left[ \frac{t - x_n}{f(t) - f(x_n)} - \frac{1}{f'(x_n)} \right], \\ y_n = N(x_n) + f(x_n)^2 \varphi(y_{n-1}), \\ x_{n+1} = N(x_n) + f(x_n)^2 \varphi(y_n), \end{cases} \quad (1)$$

\*For interpretation of color in all figures, the reader is referred to the web version of this article available at [www.mathos.hr/mc](http://www.mathos.hr/mc).†Corresponding author. *Email addresses:* [w200888w@163.com](mailto:w200888w@163.com) (X. Wang), [ztmath@163.com](mailto:ztmath@163.com) (T. Zhang)

where  $y_{-1} = N(x_0)$ . The basic idea of (1) comes from Neta who in [5] derived the following three-step Newton-type method with memory of the  $R$ -order 10.815:

$$\left\{ \begin{array}{l} N(x_n) = x_n - \frac{f(x_n)}{f'(x_n)}, \quad \psi(t) = \frac{1}{f(t) - f(x_n)} \left[ \frac{t - x_n}{f(t) - f(x_n)} - \frac{1}{f'(x_n)} \right], \\ y_n = N(x_n) + [f(y_{n-1})\psi(z_{n-1}) - f(z_{n-1})\psi(y_{n-1})] \frac{f(x_n)^2}{f(y_{n-1}) - f(z_{n-1})}, \\ z_n = N(x_n) + [f(y_n)\psi(z_{n-1}) - f(z_{n-1})\psi(y_n)] \frac{f(x_n)^2}{f(y_n) - f(z_{n-1})}, \\ x_{n+1} = N(x_n) + [f(y_n)\psi(z_n) - f(z_n)\psi(y_n)] \frac{f(x_n)^2}{f(y_n) - f(z_n)}, \end{array} \right. \quad (2)$$

where  $y_{-1} = N(x_0)$  and  $z_{-1} = y_{-1} + |f(x_0)|/10$ . The efficiency indices of methods (1) and (2) are low. The reason is that iterative methods (1) and (2) require four and six functional evaluations in their first iteration, respectively. In order to accelerate the convergence or improve the computational efficiency of the Newton-type method with memory, in [14] Wang and Zhang developed a two-step Newton-type method with memory of  $R$ -order 5, which is given by

$$\left\{ \begin{array}{l} \lambda_n = -\frac{H_4''(x_n)}{2f'(x_n)}, \\ y_n = x_n - \frac{f(x_n)}{\lambda_n f(x_n) + f'(x_n)}, \\ x_{n+1} = y_n - \frac{f(y_n)}{2\lambda_n f(x_n) + f'(x_n)} \left( \frac{f(x_n) + (2 + \beta)f(y_n)}{f(x_n) + \beta f(y_n)} \right), \end{array} \right. \quad (3)$$

where  $\beta \in R$ ,  $H_4''(x_n) = 2f[x_n, x_n, y_{n-1}] + (4f[x_n, x_n, y_{n-1}, x_{n-1}] - 2f[x_n, y_{n-1}, x_{n-1}, x_{n-1}]) (x_n - y_{n-1})$  and  $H_4(x) = H_4(x; x_n, x_n, y_{n-1}, x_{n-1}, x_{n-1})$  is a Hermite interpolatory polynomial of fourth degree. The efficiency index of method (3) is  $5^{1/3} \approx 1.7100$ , which is higher than the efficiency indices of methods (1) and (2), see [14].

The purpose of this paper is to improve further the  $R$ -order of convergence and the efficiency index of the two-step Newton-type iterative method and give the convergence analysis. This paper is organized as follows. In Section 2, we derive a family of optimal fourth-order iterative methods without memory for solving nonlinear equations. Further accelerations of convergence speed are attained in Section 3. We obtain a new family of two-point Newton-type iterative methods with memory by varying two free parameters in per full iteration. The two self-accelerating parameters are calculated using information available from the current and previous iterations. The corresponding  $R$ -order of convergence is increased from 4 to 5.7016. The maximal efficiency index of the new method with memory is  $(5.7016)^{1/3} \approx 1.7865$ , which is higher than the efficiency indices of the existing Newton-type methods. Numerical examples are given in Section 4 to illustrate convergence behavior of our methods for simple roots. In Section 5, some dynamical aspects associated to the presented methods are studied. Section 6 gives a short conclusion.

## 2. The two-step Newton-type method without memory

In order to get a two-step Newton-type iterative method with memory based on the well known Ostrowski's method [7, 10], we consider the following two-step iterative method without memory by using the weight function method.

$$\begin{cases} y_n = x_n - \frac{f(x_n)}{f'(x_n)}G(s_n), \\ x_{n+1} = y_n - \frac{f(y_n)}{2f[x_n, y_n] - f'(x_n)}H(s_n), \end{cases} \quad (4)$$

where  $G(s_n)$  and  $H(s_n)$  are two weight functions with  $s_n = f(x_n)/f'(x_n)$ . The functions  $G(s_n)$  and  $H(s_n)$  should be determined so that the iterative method (4) is of order four. For the iterative method defined by (4) we have the following result.

**Theorem 1.** *Let  $a \in I$  be a simple zero of a sufficiently differentiable function  $f : I \subset \mathbb{R} \rightarrow \mathbb{R}$  for an open interval  $I$ . Then the iterative method defined by (4) is of fourth-order convergence when*

$$G(0) = 1, \quad G'(0) = \gamma, \quad H(0) = 1, \quad H'(0) = 0, \quad H''(0) = 2\beta \text{ and } \gamma, \beta \in \mathbb{R}, \quad (5)$$

and it satisfies the error equation below

$$e_{n+1} = (c_2 - \gamma)(c_2^2 - c_3 - \beta - c_2\gamma)e_n^4 + O(e_n^5). \quad (6)$$

**Proof.** Let  $e_n = x_n - a$ ,  $c_n = (1/n!)f^{(n)}(a)/f'(a)$ ,  $n = 2, 3, \dots$ . Using the Taylor expansion and taking into account  $f(a) = 0$  we have

$$f(x_n) = f'(a)[e_n + c_2e_n^2 + c_3e_n^3 + c_4e_n^4 + c_5e_n^5 + O(e_n^6)], \quad (7)$$

$$f'(x_n) = f'(a)[1 + 2c_2e_n + 3c_3e_n^2 + 4c_4e_n^3 + 5c_5e_n^4 + O(e_n^5)]. \quad (8)$$

Dividing (7) by (8) we obtain

$$s_n = \frac{f(x_n)}{f'(x_n)} = e_n - c_2e_n^2 + (2c_2^2 - 2c_3)e_n^3 + (-4c_2^3 + 7c_2c_3 - 3c_4)e_n^4 + O(e_n^5). \quad (9)$$

Using (5), (9) and Taylor expansion  $G(s_n) = G(0) + G'(0)s_n + O(s_n^2)$  we have

$$\begin{aligned} G(s_n) &= 1 + \gamma s_n \\ &= 1 + \gamma(e_n - c_2e_n^2 + (2c_2^2 - 2c_3)e_n^3 + (-4c_2^3 + 7c_2c_3 - 3c_4)e_n^4 + O(e_n^5)). \end{aligned} \quad (10)$$

From (9) and (10) we get

$$\begin{aligned} e_{n,y} = y_n - a &= x_n - a - G(s_n) \frac{f(x_n)}{f'(x_n)} \\ &= (c_2 - \gamma)e_n^2 + 2(-c_2^2 + c_3 + c_2\gamma)e_n^3 \\ &\quad + (4c_2^3 - 7c_2c_3 + 3c_4 - 5c_2^2\gamma + 4c_3\gamma)e_n^4 + O(e_n^5). \end{aligned} \quad (11)$$

By an argument similar to that of (7), we have

$$\begin{aligned} f(y_n) = & f'(a)[(c_2 - \gamma)e_n^2 - 2(c_2^2 - c_3 - c_2\gamma)e_n^3 \\ & + (5c_2^3 - 7c_2^2\gamma + 4c_4\gamma + c_2(-7c_3 - \gamma^2))e_n^4 + O(e_n^5)]. \end{aligned} \quad (12)$$

From (7), (11) and (12) we obtain

$$\begin{aligned} f[x_n, y_n] = & f'(a)(1 + c_2e_n + (c_2^2 + c_3 - c_2\gamma)e_n^2 \\ & + (-2c_2^3 + 3c_2c_3 + c_4 + 2c_2^2\gamma - c_3\gamma)e_n^3 + O(e_n^4)), \quad (13) \\ f(y_n)(2f[x_n, y_n] - f'(x_n))^{-1} = & (c_2 - \gamma)e_n^2 + 2(-c_2^2 + c_3 + c_2\gamma)e_n^3 \\ & + (3c_2^3 - 3c_2^2\gamma + 3(c_4 + c_3\gamma) \\ & - c_2(6c_3 + \gamma^2))e_n^4 + O(e_n^5). \end{aligned} \quad (14)$$

Using (5), (9) and Taylor expansion  $H(s_n) = H(0) + H'(0)s_n + H''(0)s_n^2/2 + O(s_n^3)$  we get

$$H(s_n) = 1 + \beta e_n^2 - 2\beta c_2 e_n^3 + \beta(5c_2^2 - 4c_3)e_n^4 + O(e_n^5). \quad (15)$$

Hence, together with (11), (14) and (15), we obtain the error equation

$$\begin{aligned} e_{n+1} = & x_{n+1} - a = y_n - a - f(y_n)H(s_n)(2f[x_n, y_n] - f'(x_n))^{-1} \\ = & (c_2 - \gamma)(c_2^2 - c_3 - \beta - \gamma c_2)e_n^4 + O(e_n^5). \end{aligned} \quad (16)$$

The proof is completed.  $\square$

It is obvious that the iterative method (4) requires two functions and one derivative per iteration, thus it is an optimal scheme. Let  $p^{1/n}$  be the efficiency index (see [7]), where  $p$  is the order of the method and  $n$  is the number of functional evaluations per iteration required by the method. We see that the new method (4) without memory has the efficiency index of  $\sqrt[3]{4} \approx 1.587$ .

**Remark 1.** *Based on Theorem 1, we obtain a new iterative method without memory as follows:*

$$\begin{cases} y_n = x_n - \frac{f(x_n)}{f'(x_n)}G(s_n), \\ x_{n+1} = y_n - \frac{f(y_n)}{2f[x_n, y_n] - f'(x_n)}H(s_n), \end{cases} \quad (17)$$

where  $s_n = f(x_n)/f'(x_n)$ .  $G(s_n)$  and  $H(s_n)$  in (17) are two weight functions satisfying condition (5). Functions  $G(s_n)$  and  $H(s_n)$  in (17) can take many forms. For example, taking the functions  $G(s_n) = (1 - \gamma s_n)^{-1}$  and  $H(s_n) = (1 - \beta s_n^2)^{-1}$ , we can give the following fourth-order iterative method without memory:

$$\begin{cases} y_n = x_n - \frac{f(x_n)}{f'(x_n) - \gamma f(x_n)}, \\ x_{n+1} = y_n - \frac{f(y_n)}{2f[x_n, y_n] - f'(x_n)} \frac{f'(x_n)^2}{f'(x_n)^2 - \beta f(x_n)^2}, \end{cases} \quad (\gamma, \beta \in R). \quad (18)$$

Taking the functions  $G(s_n) = (1 - \gamma s_n)^{-1}$  and  $H(s_n) = 1 + \beta s_n^2$ , we obtain the following fourth-order iterative method without memory:

$$\begin{cases} y_n = x_n - \frac{f(x_n)}{f'(x_n) - \gamma f(x_n)}, \\ x_{n+1} = y_n - \frac{f(y_n)}{2f[x_n, y_n] - f'(x_n)} \left( 1 + \beta \left( \frac{f(x_n)}{f'(x_n)} \right)^2 \right), \end{cases} \quad (\gamma, \beta \in \mathbb{R}). \quad (19)$$

**Remark 2.** It is noteworthy that the error equation (6) can be written as the following scheme

$$\begin{aligned} e_{n+1} &= (c_2 - \gamma)(c_2^2 - c_3 - \beta - c_2\gamma)e_n^4 + O(e_n^5) \\ &= (c_2 - \gamma)[c_2(c_2 - \gamma) - (c_3 + \beta)]e_n^4 + O(e_n^5). \end{aligned} \quad (20)$$

We can improve the order of convergence of method (17) by taking parameters  $\gamma = c_2$  and  $\beta = -c_3$  in (20). Therefore, the new method with memory can be obtained by method (17) without memory.

### 3. New Newton-type iterative methods with memory

In this section we will improve the convergence rate of method (17) by varying the parameters  $\gamma$  and  $\beta$  in per full iteration. Taking  $\gamma = c_2$  and  $\beta = -c_3$  in (20), we can improve the order of convergence of method (17). However, the exact values of  $f'(a)$ ,  $f''(a)$  and  $f'''(a)$  are not available in practice and such acceleration of convergence cannot be realized. But we could approximate the parameter  $\gamma$  by  $\gamma_n$  and approximate the parameter  $\beta$  by  $\beta_n$ . Parameters  $\gamma_n$  and  $\beta_n$  can be computed by using information available from the current and previous iterations and satisfy

$$\lim_{n \rightarrow \infty} \gamma_n = c_2 = \frac{f''(a)}{2f'(a)} \quad \text{and} \quad \lim_{n \rightarrow \infty} \beta_n = -c_3 = -\frac{f'''(a)}{6f'(a)}$$

such that the fourth-order asymptotic convergence constant is zero in (20). In this paper, the self-accelerating parameter  $\gamma_n$  is equal to the parameter  $-\lambda_n$  of Newton-type method (3). We consider the following four methods for  $\beta_n$ :

**Method 1:**

$$\gamma_n = \frac{H_4''(x_n)}{2f'(x_n)} \quad \text{and} \quad \beta_n = -\frac{H_4'''(x_n)}{6f'(x_n)}, \quad (21)$$

where

$$H_4'''(x_n) = 6f[x_n, x_n, y_{n-1}, x_{n-1}] + 6f[x_n, x_n, y_{n-1}, x_{n-1}, x_{n-1}](2x_n - x_{n-1} - y_{n-1}).$$

**Method 2:**

$$\gamma_n = \frac{H_4''(x_n)}{2f'(x_n)} \quad \text{and} \quad \beta_n = -\frac{\bar{H}_3'''(y_n)}{6f'(x_n)}, \quad (22)$$

where  $\bar{H}_3'''(y_n) = 6f[y_n, x_n, x_n, y_{n-1}]$ .  $\bar{H}_3(x) = \bar{H}_3(x; y_n, x_n, x_n, y_{n-1})$  is a Hermite interpolatory polynomial of third degree.

**Method 3:**

$$\gamma_n = \frac{H_4''(x_n)}{2f'(x_n)} \quad \text{and} \quad \beta_n = -\frac{\bar{H}_4'''(y_n)}{6f'(x_n)}, \quad (23)$$

where

$$\bar{H}_4'''(y_n) = \bar{H}_3'''(y_n) + 6f[y_n, x_n, x_n, y_{n-1}, x_{n-1}](3y_n - y_{n-1} - 2x_n).$$

$\bar{H}_4(x) = \bar{H}_4(x; y_n, x_n, x_n, y_{n-1}, x_{n-1})$  is a Hermite interpolatory polynomial of fourth degree.

**Method 4:**

$$\gamma_n = \frac{H_4''(x_n)}{2f'(x_n)} \quad \text{and} \quad \beta_n = -\frac{\bar{H}_5'''(y_n)}{6f'(x_n)}, \quad (24)$$

where

$$\begin{aligned} \bar{H}_5'''(y_n) = & \bar{H}_4'''(y_n) + 6f[y_n, x_n, x_n, y_{n-1}, x_{n-1}, x_{n-1}] \\ & [(y_n - y_{n-1})(y_n - x_{n-1}) + (y_n - x_n)^2 + 2(y_n - x_n)(2y_n - y_{n-1} - x_{n-1})]. \end{aligned}$$

$\bar{H}_5(x) = \bar{H}_5(x; y_n, x_n, x_n, y_{n-1}, x_{n-1}, x_{n-1})$  is a Hermite interpolatory polynomial of fifth degree.

Here  $f[x_n, x_n] = f'(x_n)$  and  $f[x_n, t] = (f(t) - f(x_n))/(t - x_n)$  are two first-order divided differences. The higher order divided differences are defined recursively. The divided difference  $f[y_n, x_n, x_n, t_0, t_1, \dots, t_{m-3}]$  of order  $m$  ( $m \geq 3$ ) is defined as

$$f[y_n, x_n, x_n, t_0, t_1, \dots, t_{m-3}] = \frac{f[x_n, x_n, t_0, t_1, \dots, t_{m-3}] - f[y_n, x_n, x_n, t_0, t_1, \dots, t_{m-4}]}{t_{m-3} - y_n}.$$

The divided difference  $f[t_0, t_1, \dots, t_{m-3}]$  of order  $m$  ( $m \geq 3$ ) is defined as

$$f[t_0, t_1, \dots, t_{m-3}] = \frac{f[t_1, t_2, \dots, t_{m-3}] - f[t_0, t_1, \dots, t_{m-4}]}{t_{m-3} - t_0}.$$

**Remark 3.** Now, we can obtain the following two-parameter Newton-type iterative method with memory

$$\begin{cases} y_n = x_n - \frac{f(x_n)}{f'(x_n)} G(s_n), \\ x_{n+1} = y_n - \frac{f(y_n)}{2f[x_n, y_n] - f'(x_n)} H(s_n), \end{cases} \quad (25)$$

where  $s_n = f(x_n)/f'(x_n)$ ,  $G(s_n)$  satisfies the condition  $G(0) = 1$ ,  $G'(0) = \gamma_n$  and  $H(s_n)$  satisfies the condition  $H(0) = 1$ ,  $H'(0) = 0$ ,  $H''(0) = 2\beta_n$ . The parameters  $\gamma_n$  and  $\beta_n$  are calculated by using one of the formulas (21)-(24) and they depend on the data available from the current and the previous iterations. In this paper, the self-accelerating parameter  $\gamma_n$  is equal to the parameter  $-\lambda_n$  of Newton-type method (3).

**Remark 4.** Parameter (21) can be computed easily, because Hermite interpolation polynomial  $H_4(x)$  has been used in the parameter  $\gamma_n$ . Computation of self-correcting parameters  $\gamma_n$  and  $\beta_n$  is complicated in this paper. The main purpose of this study is to choose available nodes as good as possible to obtain the maximal convergence order of two-step Newton type methods with memory. Hence, some simple approximations of self-accelerating parameters  $\gamma_n$  and  $\beta_n$  are not considered in this paper.

The concept of the  $R$ -order of convergence [6] and the following assertion (see [1]) will be applied to estimate the convergence order of the iterative method with memory (25).

**Theorem 2.** If the errors of approximations  $e_j = x_j - a$  obtained in an iterative root-finding method IM satisfy

$$e_{k+1} \sim \prod_{i=0}^n (e_{k-i})^{m_i}, \quad k \geq k(\{e_k\}),$$

then the  $R$ -order of convergence of IM, denoted with  $O_R(IM, a)$ , satisfies the inequality  $O_R(IM, a) \geq s^*$ , where  $s^*$  is the unique positive solution of the equation  $s^{n+1} - \sum_{i=0}^n m_i s^{n-i} = 0$ .

**Lemma 1.** Let  $\bar{H}_m$  be the Hermite interpolating polynomial of the degree  $m$  that interpolates a function  $f$  at interpolation nodes  $y_n, x_n, t_0, \dots, t_{m-3}$  contained in an interval  $I$  and the derivative  $f^{(m+1)}$  is continuous in  $I$  and the Hermite interpolating polynomial  $\bar{H}_m(x)$  satisfied the condition  $\bar{H}_m(y_n) = f(y_n), \bar{H}_m(x_n) = f(x_n), \bar{H}'_m(x_n) = f'(x_n), \bar{H}_m(t_j) = f(t_j) (j = 0, \dots, m-3)$ . Define the errors  $e_{t,j} = t_j - a (j = 0, \dots, m-3)$  and assume that

1. all nodes  $y_n, x_n, t_0, \dots, t_{m-3}$  are sufficiently close to the zero  $a$ ;
2. the conditions  $e_n = x_n - a = O(e_{t,0} \dots e_{t,m-3})$  and  $e_{n,y} = y_n - a = O(e_n^2 e_{t,0} \dots e_{t,m-3})$  hold.

Then

$$\beta_n + c_3 \sim (-1)^{m-2} c_{m+1} \prod_{j=0}^{m-3} e_{t,j}. \tag{26}$$

**Proof.** The error of the Hermite interpolation can be expressed as follows

$$f(x) - \bar{H}_m(x) = \frac{f^{(m+1)}(\xi)}{(m+1)!} (x - y_n)(x - x_n)^2 \prod_{j=0}^{m-3} (x - t_j) \quad (\xi \in I). \tag{27}$$

Differentiating (27) at the point  $x = y_n$  we obtain

$$f'''(y_n) - \bar{H}_m'''(y_n) = 6 \frac{f^{(m+1)}(\xi)}{(m+1)!} \left\{ \prod_{j=0}^{m-3} (y_n - t_j) + 2(y_n - x_n) \sum_{k=0}^{m-3} \left( \prod_{\substack{j=0 \\ j \neq k}}^{m-3} (y_n - t_j) \right) \right\}$$

$$+ \frac{(y_n - x_n)^2}{2} \sum_{k=0}^{m-3} \sum_{\substack{i=0 \\ i \neq k}}^{m-3} \left( \prod_{\substack{j=0 \\ j \neq k \\ j \neq i}}^{m-3} (y_n - t_j) \right) \left. \vphantom{\sum} \right\} (\xi \in I), \quad (28)$$

$$\begin{aligned} \bar{H}_m'''(y_n) = f'''(y_n) - 6 \frac{f^{(m+1)}(\xi)}{(m+1)!} & \left\{ \prod_{j=0}^{m-3} (y_n - t_j) + 2(y_n - x_n) \sum_{k=0}^{m-3} \left( \prod_{\substack{j=0 \\ j \neq k}}^{m-3} (y_n - t_j) \right) \right. \\ & \left. + \frac{(y_n - x_n)^2}{2} \sum_{k=0}^{m-3} \sum_{\substack{i=0 \\ i \neq k}}^{m-3} \left( \prod_{\substack{j=0 \\ j \neq k \\ j \neq i}}^{m-3} (y_n - t_j) \right) \right\} (\xi \in I), \quad (29) \end{aligned}$$

Taylor's series of derivatives of  $f$  at the point  $x_n, y_n \in I$  and  $\xi \in I$  about the zero  $a$  of  $f$  give

$$f'(x_n) = f'(a)(1 + 2c_2e_n + 3c_3e_n^2 + 4c_4e_n^3 + O(e_n^4)), \quad (30)$$

$$f'(y_n) = f'(a)(1 + 2c_2e_{n,y} + 3c_3e_{n,y}^2 + 4c_4e_{n,y}^3 + O(e_{n,y}^4)), \quad (31)$$

$$f''(y_n) = f'(a)(2c_2 + 6c_3e_{n,y} + 12c_4e_{n,y}^2 + O(e_{n,y}^3)), \quad (32)$$

$$f'''(y_n) = f'(a)(6c_3 + 24c_4e_{n,y} + O(e_{n,y}^2)), \quad (33)$$

$$f^{(m+1)}(\xi) = f'(a)((m+1)!c_{m+1} + (m+2)!c_{m+2}e_\xi + O(e_\xi^2)), \quad (34)$$

where  $e_\xi = \xi - a$ . Substituting (33) and (34) into (29) we have

$$\begin{aligned} \bar{H}_m'''(y_n) \sim 6f'(a) & \left\{ c_3 + 4c_4e_{n,y} - c_{m+1} \left\{ \prod_{j=0}^{m-3} (y_n - t_j) + 2(y_n - x_n) \right. \right. \\ & \left. \left. \sum_{k=0}^{m-3} \left( \prod_{\substack{j=0 \\ j \neq k}}^{m-3} (y_n - t_j) \right) + \frac{(y_n - x_n)^2}{2} \sum_{k=0}^{m-3} \sum_{\substack{i=0 \\ i \neq k}}^{m-3} \left( \prod_{\substack{j=0 \\ j \neq k \\ j \neq i}}^{m-3} (y_n - t_j) \right) \right\} \right\}. \quad (35) \end{aligned}$$



Dividing (35) by (30) we obtain

$$\beta_n = -\frac{\bar{H}_m'''(y_n)}{6f'(x_n)} \sim -\left( c_3 - c_{m+1} \left\{ \prod_{j=0}^{m-3} (y_n - t_j) + 2(y_n - x_n) \right. \right. \\ \left. \left. \sum_{k=0}^{m-3} \left( \prod_{\substack{j=0 \\ j \neq k}}^{m-3} (y_n - t_j) \right) + \frac{(y_n - x_n)^2}{2} \sum_{k=0}^{m-3} \sum_{\substack{i=0 \\ i \neq k}}^{m-3} \left( \prod_{\substack{j=0 \\ j \neq k \\ j \neq i}}^{m-3} (y_n - t_j) \right) \right\} \right), \quad (36)$$

and

$$\beta_n + c_3 \sim c_{m+1} \left\{ \prod_{j=0}^{m-3} (y_n - t_j) + 2(y_n - x_n) \sum_{k=0}^{m-3} \left( \prod_{\substack{j=0 \\ j \neq k}}^{m-3} (y_n - t_j) \right) \right. \\ \left. + \frac{(y_n - x_n)^2}{2} \sum_{k=0}^{m-3} \sum_{\substack{i=0 \\ i \neq k}}^{m-3} \left( \prod_{\substack{j=0 \\ j \neq k \\ j \neq i}}^{m-3} (y_n - t_j) \right) \right\} \sim (-1)^{m-2} c_{m+1} \prod_{j=0}^{m-3} e_{t,j}. \quad (37)$$

The proof is completed.  $\square$

**Theorem 3.** *Let the varying parameters  $\gamma_n$  and  $\beta_n$  in the iterative method (25) be calculated by (21). If an initial approximation  $x_0$  is sufficiently close to a simple root  $a$  of  $f(x)$ , then the  $R$ -order of convergence of the iterative method (25) with memory is at least 5.3059.*

**Proof.** Let the sequence  $\{x_n\}$  be generated by an iterative method (IM) converging to the root  $a$  of  $f(x)$  with the  $R$ -order  $O_R(\text{IM}, a) \geq r$ , we write

$$e_{n+1} \sim D_{n,r} e_n^r, \quad e_n = x_n - a, \quad (38)$$

where  $D_{n,r}$  tends to the asymptotic error constant  $D_r$  of (IM) when  $n \rightarrow \infty$ . So,

$$e_{n+1} \sim D_{n,r} (D_{n-1,r} e_{n-1}^r)^r = D_{n,r} D_{n-1,r}^r e_{n-1}^{r^2}. \quad (39)$$

It is similar to the derivation of (39). We assume that the iterative sequence  $\{y_n\}$  has the  $R$ -order  $p$ ; then

$$e_{n,y} \sim D_{n,p} e_n^p \sim D_{n,p} (D_{n-1,r} e_{n-1}^r)^p = D_{n,p} D_{n-1,r}^p e_{n-1}^{rp}. \quad (40)$$

Using (11), (20),  $\gamma_n$  and  $\beta_n$ , we obtain the corresponding error relations for the methods with memory (25)

$$e_{n,y} = y_n - a \sim (c_2 - \gamma_n)e_n^2, \quad (41)$$

$$e_{n+1} = x_{n+1} - a \sim (c_2 - \gamma_n)[c_2(c_2 - \gamma_n) - (c_3 + \beta_n)]e_n^4. \quad (42)$$

Here, we omitted higher order terms in (41)-(42).

Hermite interpolating polynomial  $H_4(x)$  satisfied the conditions  $H_4(x_n) = f(x_n)$ ,  $H_4'(x_n) = f'(x_n)$ ,  $H_4(y_{n-1}) = f(y_{n-1})$ ,  $H_4(x_{n-1}) = f(x_{n-1})$  and  $H_4'(x_{n-1}) = f'(x_{n-1})$ . The error of the Hermite interpolation can be expressed as follows:

$$f(x) - H_4(x) = \frac{f^{(5)}(\xi)}{5!}(x - x_n)^2(x - x_{n-1})^2(x - y_{n-1}) \quad (\xi \in I). \quad (43)$$

Differentiating (43) at the point  $x = x_n$  we obtain

$$f''(x_n) - H_4''(x_n) = 2\frac{f^{(5)}(\xi)}{5!}(x_n - x_{n-1})^2(x_n - y_{n-1}) \quad (\xi \in I), \quad (44)$$

$$H_4''(x_n) = f''(x_n) - 2\frac{f^{(5)}(\xi)}{5!}(x_n - x_{n-1})^2(x_n - y_{n-1}) \quad (\xi \in I), \quad (45)$$

$$f'''(x_n) - H_4'''(x_n) = 6\frac{f^{(5)}(\xi)}{5!}[2(x_n - x_{n-1})(x_n - y_{n-1}) + (x_n - x_{n-1})^2] \quad (\xi \in I), \quad (46)$$

$$\begin{aligned} H_4'''(x_n) &= f'''(x_n) - 6\frac{f^{(5)}(\xi)}{5!}[2(x_n - x_{n-1})(x_n - y_{n-1}) \\ &\quad + (x_n - x_{n-1})^2] \quad (\xi \in I). \end{aligned} \quad (47)$$

Taylor's series of derivatives of  $f$  at the point  $x_n, \xi \in I$  about the zero  $a$  of  $f$  give

$$f''(x_n) = f'(a)(2c_2 + 6c_3e_n + 12c_4e_n^2 + O(e_n^3)), \quad (48)$$

$$f'''(x_n) = f'(a)(6c_3 + 24c_4e_n + O(e_n^2)), \quad (49)$$

$$f^{(5)}(\xi) = f'(a)(5!c_5 + 6!c_6e_\xi + O(e_\xi^2)), \quad (50)$$

where  $e_\xi = \xi - a$ .

Substituting (48) and (50) into (45) we have

$$H_4''(x_n) \sim 2f'(a)(c_2 + c_5e_{n-1,y}e_{n-1}^2 + 3c_3e_n). \quad (51)$$

Substituting (49) and (50) into (47) we have

$$H_4'''(x_n) \sim 6f'(a)(c_3 - c_5(2e_{n-1,y}e_{n-1} + e_{n-1}^2) + 4c_4e_n). \quad (52)$$

Dividing (51) by (30) we obtain

$$\gamma_n = \frac{H_4''(x_n)}{2f'(x_n)} \sim (c_2 + c_5e_{n-1,y}e_{n-1}^2 + (3c_3 - 2c_2^2)e_n), \quad (53)$$

$$c_2 - \gamma_n \sim -c_5e_{n-1,y}e_{n-1}^2. \quad (54)$$

Dividing (52) by (30) we obtain

$$\beta_n = -\frac{H_4'''(x_n)}{6f'(x_n)} \sim - (c_3 - c_5(2e_{n-1,y}e_{n-1} + e_{n-1}^2) + (4c_4 - c_2c_3)e_n), \quad (55)$$

$$c_3 + \beta_n \sim (c_5(2e_{n-1,y}e_{n-1} + e_{n-1}^2) - (4c_4 - c_2c_3)e_n) \sim c_5e_{n-1}^2. \quad (56)$$

According to (41), (42), (54) and (56), we get

$$\begin{aligned} e_{n,y} &\sim (c_2 - \gamma_n)e_n^2 \sim -c_5e_{n-1,y}e_{n-1}^2(D_{n-1,r}e_{n-1}^r)^2 \\ &\sim -c_5D_{n-1,p}D_{n-1,r}^2e_{n-1}^{2r+p+2}, \end{aligned} \quad (57)$$

$$\begin{aligned} e_{n+1} &\sim c_5e_{n-1,y}e_{n-1}^2[c_2c_5e_{n-1,y}e_{n-1}^2 + c_5e_{n-1}^2]e_n^4 \\ &\sim c_5^2e_{n-1}^4D_{n-1,p}e_{n-1}^p(D_{n-1,r}e_{n-1}^r)^4 \sim c_5^2D_{n-1,p}D_{n-1,r}^4e_{n-1}^{4r+p+4}. \end{aligned} \quad (58)$$

By comparing exponents of  $e_{n-1}$  appearing in two pairs of relations ((40),(57)) and ((39),(58)), we get the following system of equations

$$\begin{cases} 2r + p + 2 = rp, \\ 4r + p + 4 = r^2. \end{cases} \quad (59)$$

Positive solution of system (59) is given by  $r \approx 5.3059$  and  $p \approx 2.9209$ . Therefore, the  $R$ -order of the methods with memory (25), when  $\gamma_n$  is calculated by (21), is at least 5.3059.  $\square$

Using the results of Lemma 1 and Theorem 3, we get Theorem 4 as follows:

**Theorem 4.** *Let the varying parameters  $\gamma_n$  and  $\beta_n$  in the iterative method (25) be calculated by (22), (23) and (24), respectively. If an initial approximation  $x_0$  is sufficiently close to a simple root  $a$  of  $f(x)$ , then the  $R$ -order of convergence of iterative methods (25) with memory is at least 5.4356, 5.5708 and 5.7016, respectively.*

**Proof.** Method 2,  $\gamma_n$  and  $\beta_n$  are calculated by (22):

Using Lemma 1 for  $m = 3$  and  $t_0 = y_{n-1}$ , we obtain

$$c_3 + \beta_n \sim -c_4e_{n-1,y}. \quad (60)$$

According to (42), (54) and (60), we get

$$\begin{aligned} e_{n+1} &\sim c_5e_{n-1}^2e_{n-1,y}[c_2c_5e_{n-1}^2e_{n-1,y} - c_4e_{n-1,y}]e_n^4 \sim -c_4c_5e_{n-1}^2e_{n-1,y}^2e_n^4 \\ &\sim -c_4c_5e_{n-1}^2D_{n-1,p}^2e_{n-1}^{2p}(D_{n-1,r}e_{n-1}^r)^4 \sim -c_4c_5D_{n-1,p}^2D_{n-1,r}^4e_{n-1}^{4r+2p+2}. \end{aligned} \quad (61)$$

By comparing exponents of  $e_{n-1}$  appearing in two pairs of relations ((40),(57)) and ((39),(61)), we get the following system of equations

$$\begin{cases} 2r + p + 2 = rp, \\ 4r + 2p + 2 = r^2. \end{cases} \quad (62)$$

Positive solution of system (62) is given by  $r \approx 5.4356$  and  $p \approx 2.9018$ . Therefore, the  $R$ -order of the methods with memory (25), when  $\gamma_n$  is calculated by (22), is at least 5.4356.

Method 3,  $\gamma_n$  and  $\beta_n$  are calculated by (23):

Using Lemma 1 for  $m = 4$ ,  $t_0 = y_{n-1}$  and  $t_1 = x_{n-1}$ , we obtain

$$c_3 + \beta_n \sim c_5 e_{n-1,y} e_{n-1}, \quad (63)$$

According to (42), (54) and (63), we get

$$\begin{aligned} e_{n+1} &\sim c_5 e_{n-1}^2 e_{n-1,y} [c_2 c_5 e_{n-1}^2 e_{n-1,y} + c_5 e_{n-1,y} e_{n-1}] e_n^4 \sim c_5^2 e_{n-1}^3 e_{n-1,y}^2 e_n^4 \\ &\sim c_5^2 e_{n-1}^3 D_{n-1,p}^2 e_{n-1}^{2p} (D_{n-1,r} e_{n-1}^r)^4 \sim c_5^2 D_{n-1,p}^2 D_{n-1,r}^4 e_{n-1}^{4r+2p+3}. \end{aligned} \quad (64)$$

By comparing exponents of  $e_{n-1}$  appearing in two pairs of relations ((40), (57)) and ((39), (64)), we get the following system of equations

$$\begin{cases} 2r + p + 2 = rp, \\ 4r + 2p + 3 = r^2. \end{cases} \quad (65)$$

Positive solution of system (65) is given by  $r \approx 5.5708$  and  $p \approx 2.8751$ . Therefore, the  $R$ -order of the methods with memory (25), when  $\gamma_n$  is calculated by (23), is at least 5.5708.

Method 4,  $\gamma_n$  and  $\beta_n$  are calculated by (24):

Hermite interpolating polynomial  $\bar{H}_5(x)$  satisfied the condition  $\bar{H}_5(y_n) = f(y_n)$ ,  $\bar{H}_5(x_n) = f(x_n)$ ,  $\bar{H}'_5(x_n) = f'(x_n)$ ,  $\bar{H}_5(y_{n-1}) = f(y_{n-1})$ ,  $\bar{H}_5(x_{n-1}) = f(x_{n-1})$  and  $\bar{H}'_5(x_{n-1}) = f'(x_{n-1})$ . The error of the Hermite interpolation can be expressed as follows:

$$f(x) - \bar{H}_5(x) = \frac{f^{(5)}(\xi)}{6!} (x - y_n)(x - x_n)^2(x - x_{n-1})^2(x - y_{n-1}) \quad (\xi \in I). \quad (66)$$

Differentiating (66) at the point  $x = y_n$  we obtain

$$\begin{aligned} f'''(y_n) - \bar{H}_5'''(y_n) &= 6 \frac{f^{(5)}(\xi)}{6!} [(y_n - y_{n-1})(y_n - x_{n-1})^2 + 2(y_n - x_n)(y_n - x_{n-1})^2 \\ &\quad + 4(y_n - x_n)(y_n - x_{n-1})(y_n - y_{n-1}) + 2(y_n - x_n)^2(y_n - x_{n-1}) \\ &\quad + (y_n - x_n)^2(y_n - y_{n-1})] \quad (\xi \in I), \end{aligned} \quad (67)$$

$$\begin{aligned} \bar{H}_5'''(y_n) &= f'''(y_n) - 6 \frac{f^{(5)}(\xi)}{6!} [(y_n - y_{n-1})(y_n - x_{n-1})^2 \\ &\quad + 2(y_n - x_n)(y_n - x_{n-1})^2 + 4(y_n - x_n)(y_n - x_{n-1})(y_n - y_{n-1}) \\ &\quad + 2(y_n - x_n)^2(y_n - x_{n-1}) \\ &\quad + (y_n - x_n)^2(y_n - y_{n-1})] \quad (\xi \in I). \end{aligned} \quad (68)$$

Taylor's series of derivatives of  $f$  at the point  $\xi \in I$  about the zero  $a$  of  $f$  give

$$f^{(5)}(\xi) = f'(a)(6!c_6 + 7!c_7e_\xi + O(e_\xi^2)), \quad (69)$$

where  $e_\xi = \xi - a$ .

Substituting (33) and (69) into (68) we have

$$\bar{H}_5'''(y_n) \sim 6f'(a)(c_3 + c_6 e_{n-1,y} e_{n-1}^2). \quad (70)$$

Dividing (70) by (30) we obtain

$$\beta_n = -\frac{\bar{H}_5'''(y_n)}{6f'(x_n)} \sim -(c_3 + c_6 e_{n-1,y} e_{n-1}^2), \tag{71}$$

$$c_3 + \beta_n \sim -c_6 e_{n-1,y} e_{n-1}^2. \tag{72}$$

According to (42), (54) and (72), we get

$$\begin{aligned} e_{n+1} &\sim c_5 e_{n-1,y} e_{n-1}^2 [c_2 c_5 e_{n-1,y} e_{n-1}^2 - c_6 e_{n-1,y} e_{n-1}^2] e_n^4 \sim c_5 (c_2 c_5 - c_6) e_{n-1}^4 e_{n-1,y}^2 e_n^4 \\ &\sim c_5 (c_2 c_5 - c_6) D_{n-1,p}^2 D_{n-1,r}^4 e_{n-1}^{4r+2p+4}. \end{aligned} \tag{73}$$

By comparing exponents of  $e_{n-1}$  appearing in two pairs of relations ((40),(57)) and ((39),(73)), we get the following system of equations

$$\begin{cases} 2r + p + 2 = rp, \\ 4r + 2p + 4 = r^2. \end{cases} \tag{74}$$

Positive solution of system (74) is given by  $r \approx 5.7016$  and  $p \approx 2.8508$ . Therefore, the  $R$ -order of the methods with memory (25), when  $\gamma_n$  is calculated by (24), is at least 5.7016.

The proof is completed. □

**Remark 5.** *The efficiency index of iterative method (25) with the corresponding expressions (21)-(24) of  $\gamma_n$  and  $\beta_n$  is  $(5.3059)^{1/3} \approx 1.7442$ ,  $(5.4356)^{1/3} \approx 1.7583$ ,  $(5.5708)^{1/3} \approx 1.7727$  and  $(5.7016)^{1/3} \approx 1.7865$ , respectively. The efficiency indices of our methods with memory are higher than the efficiency index  $(5^{1/3} \approx 1.7100)$  of Newton-type method (3) with memory.*

### 4. Numerical results

The new methods (18) and (19) with and without memory are used to solve nonlinear functions  $f_i(x)(i = 1, 2)$  and the computation results are compared with other Newton-type iterative methods (1) and (3) with memory, see Tables 1-2. The absolute errors  $|x_k - a|$  in the first four iterations are given in Tables 1-2, where  $a$  is the exact root computed with 1200 significant digits. The computational order of convergence  $\rho$  is defined by [3]:

$$\rho \approx \frac{\ln(|x_{n+1} - x_n| / |x_n - x_{n-1}|)}{\ln(|x_n - x_{n-1}| / |x_{n-1} - x_{n-2}|)}. \tag{75}$$

The following test functions are used:

$$\begin{aligned} f_1(x) &= x e^{x^2} - \sin^2(x) + 3 \cos(x) + 5, \quad a \approx -1.2076478271309189, \quad x_0 = -1.3, \\ f_2(x) &= \arcsin(x^2 - 1) - 0.5x + 1, \quad a \approx 0.59481096839836918, \quad x_0 = 0.7. \end{aligned}$$

The numerical results shown in Tables 1-2 are in concordance with the theory developed in this paper. We can notice that the results obtained by our methods with memory are better than the other two-step Newton-type methods with memory. The highest order of convergence of our methods with memory is 5.7016, which is higher than methods (1) and (3).

Methods	$ x_1 - a $	$ x_2 - a $	$ x_3 - a $	$ x_4 - a $	$\rho$
(18), $\gamma = \beta = -1$	0.23e-4	0.37e-19	0.23e-78	0.40e-315	4.00
(19), $\gamma = \beta = -1$	0.23e-4	0.36e-19	0.20e-78	0.22e-315	4.00
(1)	0.53e-5	0.12e-23	0.25e-108	0.18e-494	4.56
(3), $\lambda_0 = -0.01, \beta = -2$	0.34e-4	0.66e-21	0.32e-105	0.90e-527	5.00
(18), (21), $\gamma_0 = \beta_0 = -1$	0.23e-4	0.85e-24	0.47e-127	0.62e-675	5.31
(18), (22), $\gamma_0 = \beta_0 = -1$	0.23e-4	0.45e-24	0.45e-132	0.89e-719	5.43
(18), (23), $\gamma_0 = \beta_0 = -1$	0.23e-4	0.34e-25	0.32e-141	0.13e-787	5.57
(18), (24), $\gamma_0 = \beta_0 = -1$	0.23e-4	0.11e-25	0.53e-148	0.24e-844	5.69
(19), (21), $\gamma_0 = \beta_0 = -1$	0.23e-4	0.82e-24	0.38e-127	0.20e-675	5.31
(19), (22), $\gamma_0 = \beta_0 = -1$	0.23e-4	0.43e-24	0.36e-132	0.28e-719	5.43
(19), (23), $\gamma_0 = \beta_0 = -1$	0.23e-4	0.32e-25	0.26e-141	0.41e-788	5.57
(19), (24), $\gamma_0 = \beta_0 = -1$	0.23e-4	0.10e-25	0.40e-148	0.47e-845	5.69

Table 1: Numerical results for  $f_1(x)$  by the methods with and without memory

Methods	$ x_1 - a $	$ x_2 - a $	$ x_3 - a $	$ x_4 - a $	$\rho$
(18), $\gamma = \beta = -0.6$	0.56e-4	0.48e-17	0.27e-69	0.28e-278	4.00
(19), $\gamma = \beta = -0.6$	0.56e-4	0.49e-17	0.30e-69	0.42e-278	4.00
(1)	0.37e-8	0.70e-41	0.13e-189	0.84e-868	4.56
(3), $\lambda_0 = -0.1, \beta = -1.75$	0.40e-5	0.25e-27	0.96e-139	0.88e-696	5.00
(18), (21), $\gamma_0 = \beta_0 = -0.6$	0.56e-4	0.65e-24	0.33e-128	0.77e-683	5.32
(18), (22), $\gamma_0 = \beta_0 = -0.6$	0.56e-4	0.41e-24	0.16e-133	0.13e-727	5.43
(18), (23), $\gamma_0 = \beta_0 = -0.6$	0.56e-4	0.55e-25	0.28e-141	0.37e-789	5.57
(18), (24), $\gamma_0 = \beta_0 = -0.6$	0.56e-4	0.62e-26	0.15e-149	0.26e-855	5.71
(19), (21), $\gamma_0 = \beta_0 = -0.6$	0.56e-4	0.66e-24	0.38e-128	0.17e-682	5.32
(19), (22), $\gamma_0 = \beta_0 = -0.6$	0.56e-4	0.42e-24	0.19e-133	0.31e-727	5.43
(19), (23), $\gamma_0 = \beta_0 = -0.6$	0.56e-4	0.57e-25	0.33e-141	0.92e-789	5.57
(19), (24), $\gamma_0 = \beta_0 = -0.6$	0.56e-4	0.64e-26	0.19e-149	0.85e-855	5.71

Table 2: Numerical results for  $f_2(x)$  by the methods with and without memory

## 5. Dynamical analysis

Dynamical properties of the rational function give us important information about numerical features of the iterative method as its stability and reliability. In what follows, we compare our methods with and without memory to methods with memory (1) and (3) by using the basins of attraction for three complex polynomials  $f(z) = z^k - 1$ ,  $k = 2, 3, 4$ . We use a similar like technique as in [15] to generate the basins of attraction. To generate the basins of attraction for the zeros of a polynomial and an iterative method we take a grid of  $400 \times 400$  points in a rectangle  $D = [-3.0, 3.0] \times [-3.0, 3.0] \subset \mathbb{C}$  and we use these points as  $z_0$ . If the sequence generated by the iterative method reaches a zero  $z^*$  of the polynomial with a tolerance  $|z_k - z^*| < 10^{-15}$  and a maximum of 25 iterations, we decide that  $z_0$  is in the basin of attraction of the zero and we paint this point blue for this root. In the same basin of attraction, the number of iterations needed to achieve the solution is shown in darker or

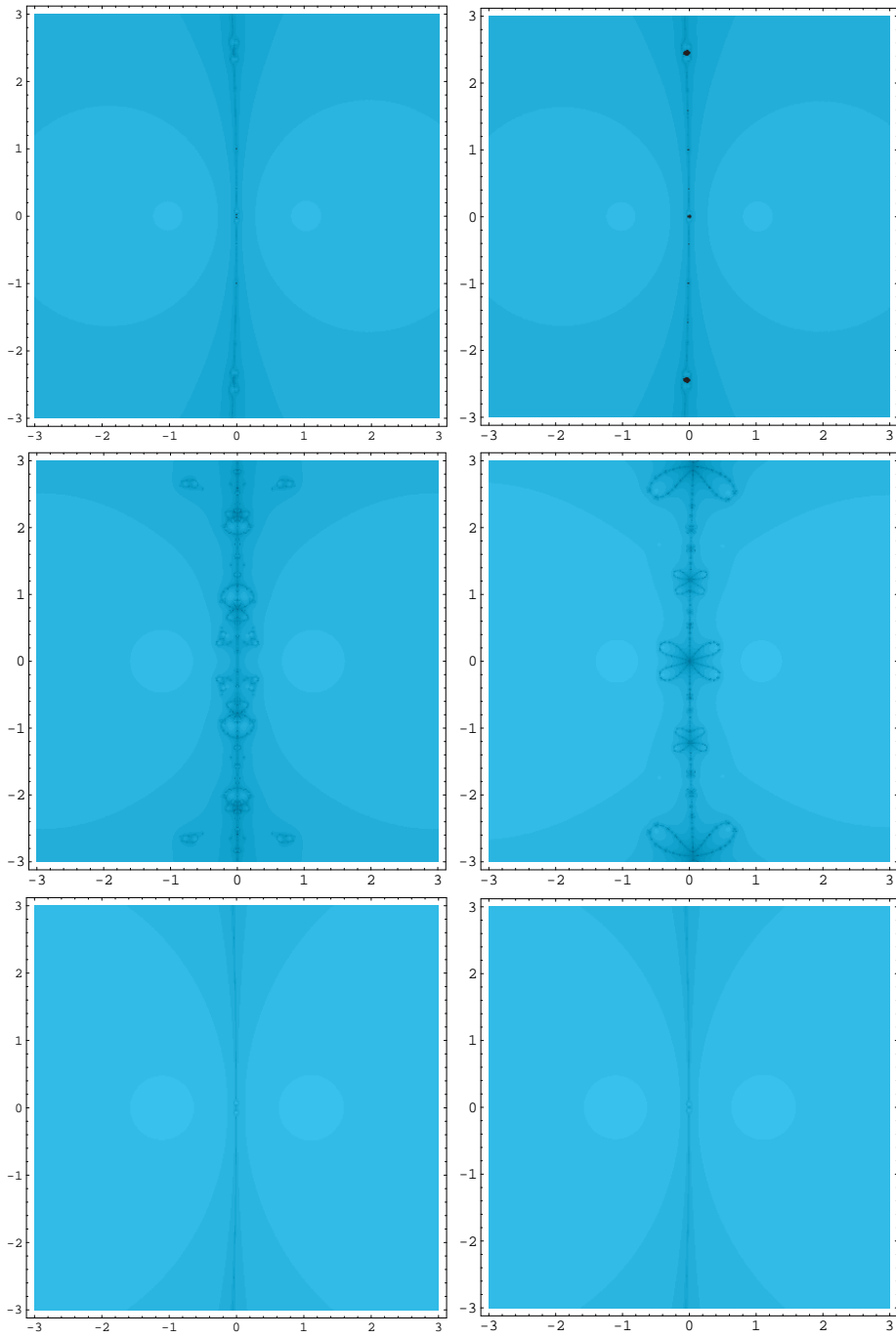


Figure 1: Top row: Methods without memory (18)(left) and (19)(right). Middle row: Methods (1)(left) and (3)(right). Bottom row: Methods (18) with (24)(left) and (19) with (24)(right). The results are for the polynomial  $z^2 - 1$ .

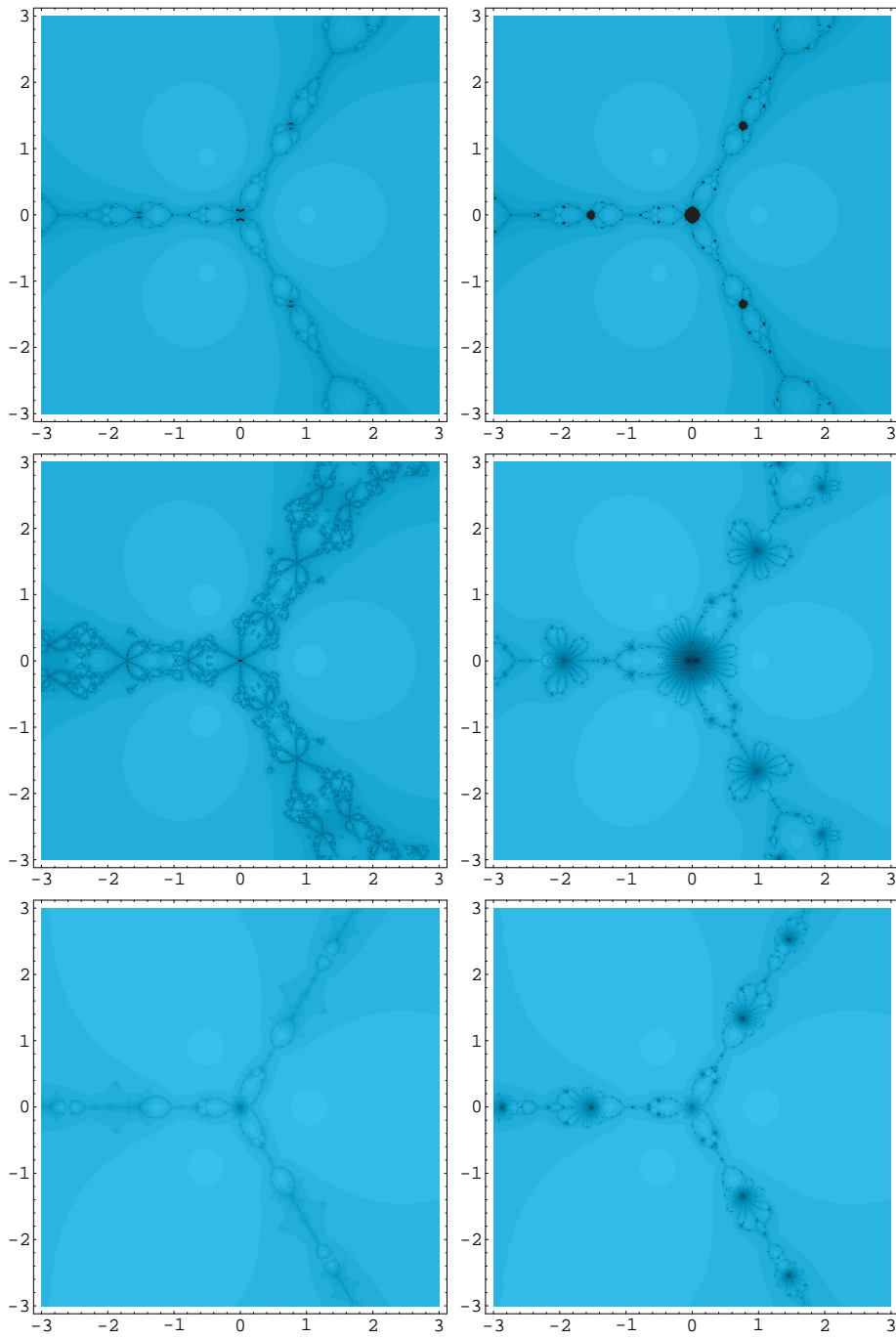


Figure 2: Top row: Methods without memory (18)(left) and (19)(right). Middle row: Methods (1)(left) and (3)(right). Bottom row: Methods (18) with (24)(left) and (19) with (24)(right). The results are for the polynomial  $z^3 - 1$ .



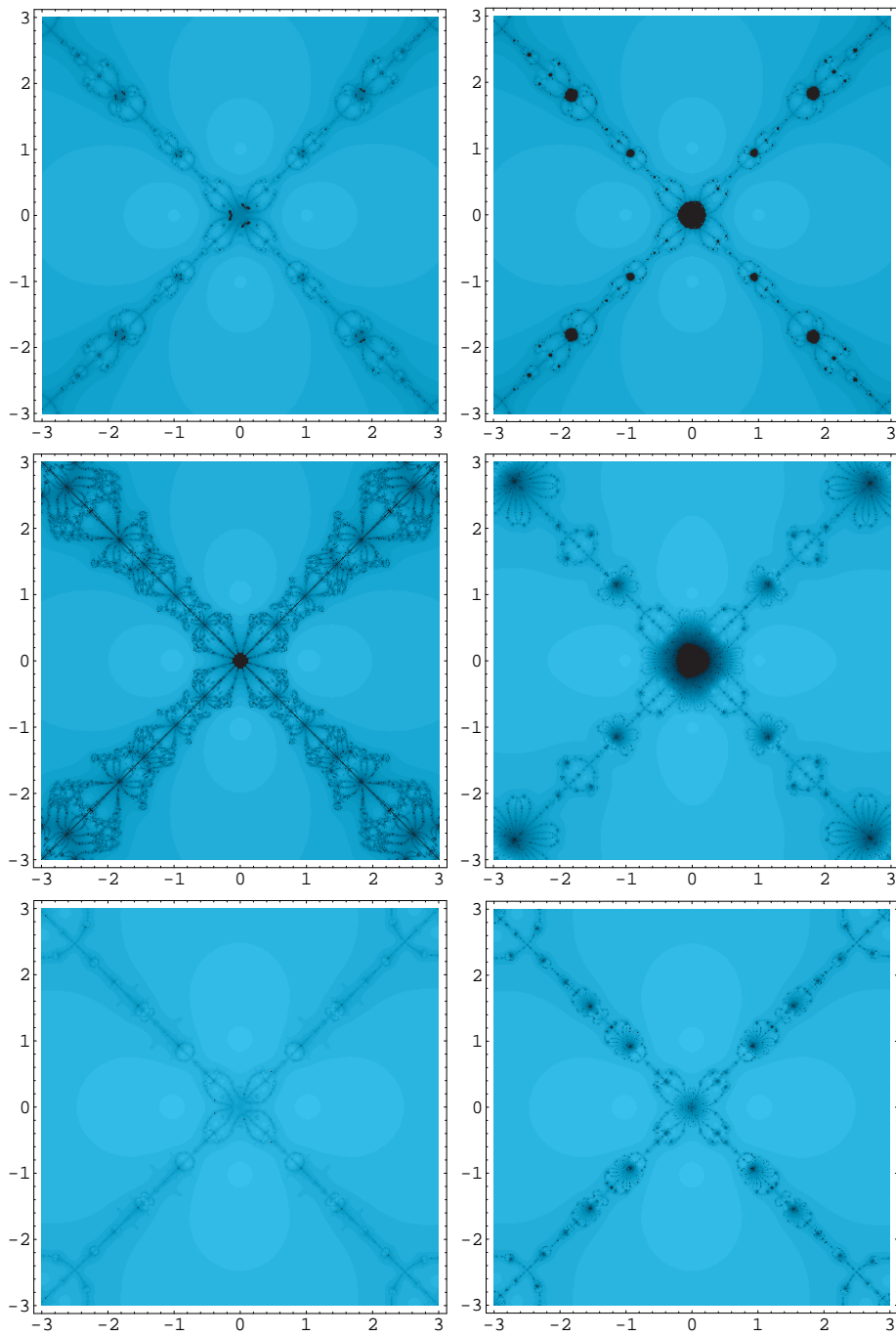


Figure 3: Top row: Methods without memory (18)(left) and (19)(right). Middle row: Methods (1)(left) and (3)(right). Bottom row: Methods (18) with (24)(left) and (19) with (24)(right). The results are for the polynomial  $z^4 - 1$ .

brighter colors (the less iterations, the brighter color). Black color denotes lack of convergence to any of the roots (with the maximum of iterations established) or convergence to the infinity. The fractals of our methods with memory are almost similar, we only give the basins of attraction of our methods with memory of the  $R$ -order 5.7016. The parameters used in iterative methods (18)–(19) without memory are  $\gamma = 0.01$  and  $\beta = -0.01$ . The parameters  $\lambda_0 = 0.01$  and  $\beta = -0.01$  are used in method (3) with memory. Our methods with memory use the parameters  $\gamma_0 = 0.01$  and  $\beta_0 = -0.01$  in the first iteration.

Figures 1-3 show that the new methods with memory have very little diverging points compared to other methods. Figures 2-3 show that the basins of attraction for our methods with memory are larger than the other methods and the convergence speed of our methods with memory are faster than our methods (18)–(19) without memory. On the whole, we can see that the new methods with memory are better than the other methods in this paper.

## 6. Conclusions

In this paper, we have proposed a new family of two-step Newton-type iterative methods with and without memory for solving nonlinear equations. Using two self-correcting parameters calculated by Hermite interpolatory polynomials the  $R$ -order of convergence of the new Newton-type method with memory is increased from 4 to 5.7016 without any additional calculations. The main contribution of this paper is that we obtain the maximal order of Newton-type methods without any additional function evaluations. The new methods are compared in performance and computational efficiency with some existing methods by numerical examples. We observed that the computational efficiency indices of the new methods with memory are better than those of other existing two-step methods.

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