

NUMERICAL SOLUTION OF NONLINEAR EQUATIONS BY A TWELTH-ORDER ITERATIVE METHOD WITH MEMORY

ABSTRACT. In this paper, we introduce three point derivative free scheme derived from King's family of methods for solving non-linear equations. This is an addition in three point with memory technique with twelfth-order of convergence introducing new parameters. These parameters not only minimized extra functional and derivative evaluation but also increase convergence order of method. Efficiency index remarkably enhanced from 1.86121 to 1.867551871. It is noted that new scheme has improved performance to solve non linear equation. Numerical analysis of recent methods with new method are given which shows the efficiency of presented method.

1. Introduction

Solving nonlinear equation $f(x) = 0$ is a classical problem in field of numerical computation [1]-[6] and references therein. There are enough publications to find solution of non-linear equation with different convergence order. Approximating simple roots of non-linear equations with a suitable level of accuracy has an interesting and vital importance in all fields of Science.

Jarratt and Ostrowski presented some well-known two-point techniques. King [11] presented one of the most famous optimal 4th order iterative method. But flaws of this scheme is that it requires first derivative in each step. Many authors modified King's method to get more accurate results such as Chun [9] introduced King's like methods of order four, but computing the first derivative within the iteration is also needed.

Chun and Lee presented a new 4th order optimal root-finding method to solve non-linear equations which describe the conjugacy classes and dynamics of the presented optimal method for complex polynomials of degree two and three. They obtained Jarratt's scheme of fourth order as a special case. Behl [2]-[4] introduced a fourth-order derivative-free scheme which is a modification in King's method by using weight functions.

To achieve more accuracy with less computation many authors introduced Optimal methods of eighth-order of convergence i.e Chun et al. [9] Cordero et al. [7]-[8], Behl et al. [2]-[4] and Geum et al. [10]. Geum and Neta [10] developed a 16th order simple root finding optimal method with general weight functions.

Our aim is to present a derivative-free iterative method with memory of twelfth order using King's iterative schemes and steffensen approaches. In the next section we establish our new method and gave convergence analysis. In section 3, solution

Key words and phrases. Nonlinear equation, iterative method with memory, twelfth-order, convergence order.

of some numerical examples with their comparison to other well-known iterative schemes are presented. Section 4 is a smart conclusion.

2. Establishment of new scheme of twelfth-order

We start with king's [11] technique that is one of important family for finding solution of nonlinear problems.

$$y_m = x_m - \frac{f(x_m)}{f'(x_m)},$$

$$x_{m+1} = y_m - \frac{f(y_m)}{f'(x_m)} \cdot \frac{f(x_m) + \gamma f(y_m)}{f(x_m) + (\gamma - 2)f(y_m)}, \quad (m = 0, 1, \dots), \quad \gamma \in R,$$

Here x_0 is a preliminary approximation of a simple zero α of f . First we derive an optimal derivative free scheme of 2 point method having convergence order 4. We consider Steffensen's scheme for the 1st & 2nd step approximation $f'(x_m)$

$$f'(x_n) \approx \frac{f[y_m, w_m]}{G(t_m)},$$

while $w_m = x_m - \beta f(x_m)$, $\beta \neq 0$, $f[y_m, w_m] = \frac{f(y_m) - f(w_m)}{y_m - w_m}$, $t_m = \frac{f(y_m)}{f(x_m)}$ Here G is actual function.

Hence, we obtain a fourth order scheme,

$$\begin{cases} y_m = x_m - \frac{\beta f(x_m)^2}{f(x_m) - f(w_m)}, \\ x_{m+1} = y_m - \frac{f(x_m) + \gamma f(y_m)}{f(x_m) + (\gamma - 2)f(y_m)} \cdot \frac{f(y_m)}{f[y_m, w_m]} G(t_m). \end{cases} \quad (1)$$

Now we derive eighth-order scheme by adding Newton step in scheme (1), we have

$$\begin{cases} y_m = x_m - \frac{\beta f(x_m)^2}{f(x_m) - f(w_m)}, \\ z_m = y_m - \frac{f(x_m) + \gamma f(y_m)}{f(x_m) + (\gamma - 2)f(y_m)} \cdot \frac{f(y_m)}{f[y_m, w_m]} G(t_m). \\ x_{m+1} = z_m - \frac{f(z_m)}{f'(z_m)}. \end{cases} \quad (2)$$

It is observed the function is evaluated many times to make it derivative-free and optimal method. We approximate $f'(z_m)$ with Newton's interpolation of degree 3 at the point x_m, y_m , and z_m .

$$N_3(t; z_m, y_m, x_m, w_m) = f(z_m) + f[z_m, y_m](t - z_m) + f[z_m, y_m, x_m](t - z_m)(t - y_m) + f[z_m, y_m, x_m, w_m](t - z_m)(t - y_m)(t - x_m).$$

It can be seen

$$N_3(z_m) = f(z_m), \text{ and } N_3(t) |_{t=z_m} = f'(z_m).$$

So,

$$\begin{aligned} N_3(z_m) &= \left[\frac{d}{dt} N_3(t) \right]_{t=z_m} \\ &= f[z_m, y_m] + f[z_m, y_m, x_m](z_m - y_m) + f[z_m, y_m, x_m, w_m] \\ &\quad (z_m - y_m)(z_m - x_m). \end{aligned}$$

and hence we get

$$\begin{cases} y_m = x_m - \frac{\beta f(x_m)^2}{f(x_m) - f(w_m)}, \\ z_m = y_m - \frac{f(x_m) + \gamma f(y_m)}{f(x_m) + (\gamma - 2)f(y_m)} \cdot \frac{f(y_m)}{f(y_m, w_m)} G(t_m). \\ x_{m+1} = z_m - \frac{f(z_m)}{f[z_m, y_m] + f[z_m, y_m, x_m](z_m - y_m) + f[z_m, y_m, x_m, w_m](z_m - y_m)(z_m - x_m)}. \end{cases} \quad (3)$$

which is iterative scheme with memory of order 8. Now we extend this method to achieve convergence order 12.

This is done by using the speed acceleration parameters in scheme (3). If $\beta \neq 1/f'(\alpha)$ rate of convergence of scheme (3) is 8. When $\beta = 1/f'(\alpha)$ rate of convergence of method (3) could be twelve. Because the value of $f'(\alpha)$ is unavailable, we apply an approximation $f'(\alpha) \approx f(\alpha)$. Our objective is to create a with memory method that includes parameter calculations $\beta = \beta_m$ as iteration progresses by $\beta_m = 1/f'(\alpha)$ for $m = 1, 2, 3, \dots$. Initial value β_0 should be selected before beginning of iteration process. Here we use some characters \rightarrow, O and \sim according to Traub's iterative scheme.

If $\lim_{m \rightarrow \infty} f(x_m) = C$, we write down $f(x_m) \rightarrow C$ or $f \rightarrow C$, where C is a non zero contants. If $\mathcal{L}_m \rightarrow C$, we write $f = O(g)$ or $f \sim C(g)$.

By approximating $f'(\alpha)$ with $N_4(x_m)$ we get

$$\beta_m = \frac{1}{N_4(x_m)},$$

Here $N_4(t_m) := N_4(t; x_m, z_{m-1}, y_{m-1}, w_{m-1}, x_{m-1})$ is Newton's interpolation polynomial of 4th degree, place with 5 approximation $(x_m, z_{m-1}, y_{m-1}, w_{m-1}, x_{m-1})$.

$$\begin{aligned} N_4(x_m) &= \left[\frac{d}{dt} N_4(t) \right]_{t=x_m} = f[x_m, z_{m-1}] + f[x_m, z_{m-1}, y_{m-1}](x_m - z_{m-1}) \\ &\quad + f[x_m, z_{m-1}, y_{m-1}, w_{m-1}](x_m - z_{m-1})(x_m - y_{m-1}) \\ &\quad + f[x_m, z_{m-1}, y_{m-1}, w_{m-1}, x_{m-1}](x_m, z_{m-1})(x_m - y_{m-1}) \\ &\quad (x_m - w_{m-1}). \end{aligned}$$

We approximate in (2) $f'(z_m)$ with Newton's interpolation of degree 6 at the point $z_m, y_m, w_m, x_m, z_{m-1}$ and y_{m-1} .

$$\begin{aligned} N_5(t; z_m, y_m, w_m, x_m, z_{m-1}, y_{m-1}) &= f(z_m) + f[z_m, y_m](t - z_m) + f[z_m, y_m, w_m] \\ &\quad (t - z_m)(t - y_m) + f[z_m, y_m, w_m, x_m](t - z_m) \\ &\quad (t - y_m)(t - w_m) + f[z_m, y_m, w_m, x_m, z_{m-1}] \\ &\quad (t - z_m)(t - y_m)(t - w_m)(t - x_m) + \\ &\quad f[z_m, y_m, w_m, x_m, z_{m-1}, y_{m-1}](t - z_m) \\ &\quad (t - y_m)(t - w_m)(t - x_m)(t - z_{m-1}). \end{aligned}$$

It is clear that

$$N_5(z_m) = f(z_m), \text{ and } N_5'(t) |_{t=z_m} = f'(z_m).$$

Then

$$\begin{aligned} N_5'(z_m) &= \left[\frac{d}{dt} N_5(t) \right]_{t=z_m} = f[z_m, y_m] + f[z_m, y_m, w_m](z_m - y_m) + f[z_m, y_m, w_m, x_m] \\ &\quad (z_m - y_m)(z_m - w_m) + f[z_m, y_m, w_m, x_m, z_{m-1}](z_m - y_m)(z_m - w_m)(z_m - x_m) \\ &\quad + f[z_m, y_m, w_m, x_m, z_{m-1}, y_{m-1}](z_m - y_m)(z_m - w_m)(z_m - x_m)(z_m - z_{m-1}). \end{aligned}$$

and hence we get

$$\begin{aligned} y_m &= x_m - \frac{\beta f(x_m)^2}{f(x_m) - f(w_m)}, \\ z_m &= y_m - \frac{f(x_m) + \gamma f(y_m)}{f(x_m) + (\gamma - 2)f(y_m)} \cdot \frac{f(y_m)}{f(y_m, w_m)} G(t_m), \\ x_{m+1} &= z_m - \frac{f(z_m)}{f[z_m, y_m] + f[z_m, y_m, w_m](z_m - y_m) + \dots + f[z_m, y_m, w_m, x_m, z_{m-1}, y_{m-1}](z_m - y_m) \dots (z_m - w_{m-1})}. \end{aligned} \tag{4}$$

We denote the above scheme with AM12 which is a with memory method of convergence order 12. Now, we will prove the convergence order by applying the matrix method of Herzberger.

Theorem 1. *Let x_0 be a starting value which is close enough to 0 of $f(x)$ and the iterative scheme AM12 has 2 parameters which are repeatedly computed by the outline, then the scheme AM12 has 12.164414 order of convergence.*

Proof. By using Herzberger's matrix method we will find R-order of convergence which states that the spectral radius of matrix $M^{(u)} = (t_{p,q})(1 \leq p; q \leq u)$ related to a with-memory 1 step r-point scheme $x_k = \Phi(x_{k-1}, x_{k-2}, \dots, x_{k-u})$ is the lower bound of its rate of convergence. The elements of this method are as:

$$\begin{aligned} t_{p,q} &= \text{no. of functional evaluations needed at point } x_{k-q} = 1, 2, \dots, u \\ t_{p,q-1} &= 1 \text{ for } p=2, 3, \dots, u \\ t_{p,q} &= 0, \text{ otherwise.} \end{aligned}$$

Moreover, the spectral radius of product of the matrices B_1, B_2, \dots, B_m is the lower bound of order of an m-step method $\Phi = \Phi_1, \Phi_2, \dots, \Phi_m$ where the matrices B_k correspond to the iteration step $\Phi_k, 1 \leq k \leq m$. From the above equations we develop the associated matrices as follow:

$$X_{k+1} = \Phi_1(z_m, y_m, w_m, x_m, z_{m-1}, y_{m-1}).$$

$$M_1 = \begin{bmatrix} 1 & 1 & 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix}$$

$$Z_m = \Phi_2(y_m, w_m, x_m, z_{m-1}, y_{m-1}, w_{m-1}).$$

$$M_2 = \begin{bmatrix} 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix}$$

$$Y_m = \Phi_3(w_m, x_m, z_{m-1}, y_{m-1}, w_{m-1}, x_{m-1}).$$

$$M_3 = \begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix}$$

$$W_m = \Phi_4(x_m, z_{m-1}, y_{m-1}, w_{m-1}, x_{m-1}, z_{m-2}).$$

$$M_4 = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix}$$

$$M_1 \cdot M_2 = \begin{bmatrix} 2 & 2 & 2 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix}$$

$$M_1 \cdot M_2 \cdot M_3 = \begin{bmatrix} 4 & 4 & 0 & 0 & 0 & 0 \\ 2 & 2 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \end{bmatrix}$$

Hence we obtained,

$$\begin{aligned} M^{(4)} &= M_1 \cdot M_2 \cdot M_3 \cdot M_4 \\ &= \begin{bmatrix} 8 & 4 & 4 & 4 & 4 & 4 \\ 4 & 2 & 2 & 2 & 2 & 2 \\ 2 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \end{bmatrix} \end{aligned}$$

The eigen value of $M^{(4)}$ are

$$\lambda_1 = 12.164414$$

$$\lambda_2 = -0.164414003$$

$$\lambda_3 = 2.0470017e^{-15} - 3.78075673e^{-81}$$

$$\lambda_4 = 2.75578744e^{-15} - 3.78075673e^{-81}$$

$$\lambda_5 = 1.22388505e^{-16} + 1.13323081e^{-16}$$

Hence spectral radius of $M^{(4)}$ matrix is 12.164414.

□

3. Numerical Examples and comparison

3.1. **Method 1.** Consider G as weight function

$$G(t_m) = 1 - t_m, \quad (5)$$

where $t_m = \frac{f(y_m)}{f(x_m)}$. G in (5) satisfies the assumption of 2^{nd} Theorem, Hence we get the following scheme denoted by AM.1.

$$\begin{aligned} y_m &= x_m - \frac{\beta_m f(x_m)^2}{f(x_m) - f(w_m)}, w_m = x_m - \beta_m f(x_m), \beta_m = \frac{1}{N_5(x_m)}, \\ z_m &= y_m - \frac{f(x_m) + \gamma f(y_m)}{f(x_m) + (\gamma - 2)f(y_m)} \cdot \frac{f(x_m) - f(y_m)}{f(x_m)} \cdot \frac{f(y_m)}{f[y_m, w_m]} \\ x_{m+1} &= z_m - \left[\frac{f(z_m)}{f[z_m, y_m] + f[z_m, y_m, w_m](z_m - y_m) + \dots} \right. \\ &\quad \left. + f[z_m, y_m, w_m, x_m, z_{m-1}, y_{m-1}](z_m - y_m) \dots (z_m - w_{m-1}) \right] \end{aligned}$$

3.2. **Method 2.** Consider G as a weight function

$$G(t_m) = 1 - \frac{t_m}{1 + t_m},$$

where $t_m = \frac{f(y_m)}{f(x_m)}$.

$$\begin{aligned} y_m &= x_m - \frac{\beta_m f(x_m)^2}{f(x_m) - f(w_m)}, w_m = x_m - \beta_m f(x_m), \beta_m = \frac{1}{N_5(x_m)} \\ z_m &= y_m - \frac{f(x_m) + \gamma f(y_m)}{f(x_m) + (\gamma - 2)f(y_m)} \cdot \frac{f(x_m)}{f(x_m) + f(y_m)} \cdot \frac{f(y_m)}{f[y_m, w_m]}, \\ x_{m+1} &= z_m - \frac{f(z_m)}{f[z_m, y_m] + f[z_m, y_m, w_m](z_m - y_m) + \dots + f[z_m, y_m, w_m, x_m, z_{m-1}, y_{m-1}](z_m - y_m) \dots (z_m - w_{m-1})}. \end{aligned}$$

we call this method AM.2.

3.3. **Method 3.** Choose weight function G

$$G(t_m) = \frac{1 - 2t_m}{1 - t_m},$$

where $t_m = \frac{f(y_m)}{f(x_m)}$

$$\begin{aligned} y_m &= x_m - \frac{\beta_m f(x_m)^2}{f(x_m) - f(w_m)}, w_m = x_m - \beta_m f(x_m), \beta_m = \frac{1}{N_5(x_m)} \\ z_m &= y_m - \frac{f(x_m) + \gamma f(y_m)}{f(x_m) + (\gamma - 2)f(y_m)} \cdot \frac{f(x_m) - 2f(y_m)}{f(x_m) - f(y_m)} \cdot \frac{f(y_m)}{f[y_m, w_m]}, \\ x_{m+1} &= z_m - \frac{f(z_m)}{f[z_m, y_m] + f[z_m, y_m, w_m](z_m - y_m) + \dots + f[z_m, y_m, w_m, x_m, z_{m-1}, y_{m-1}](z_m - y_m) \dots (z_m - w_{m-1})}. \end{aligned}$$

we call this scheme AM.3.

3.4. **Method 4.** Consider G as weight function

$$G(t_m) = (1 - t_m)^{\frac{2t_m + 1}{t_m + 1}},$$

$$\begin{aligned}
 y_m &= x_m - \frac{\beta_m f(x_m)^2}{f(x_m) - f(w_m)}, w_m = x_m - \beta_m f(x_m), \beta_m = \frac{1}{N_5(x_m)} \\
 z_m &= y_m - \frac{f(x_m) + \gamma f(y_m)}{f(x_m) + (\gamma - 2)f(y_m)} \cdot \left(\frac{f(x_m) - f(y_m)}{f(x_m)} \right)^{\frac{2f(y_m) + f(x_m)}{f(y_m) + f(x_m)}} \cdot \frac{f(y_m)}{f[y_m, w_m]}, \\
 x_{m+1} &= z_m - \frac{f(z_m)}{f[z_m, y_m] + f[z_m, y_m, w_m](z_m - y_m) + \dots + f[z_m, y_m, w_m, x_m, z_{m-1}, y_{m-1}](z_m - y_m) \dots (z_m - w_{m-1})}.
 \end{aligned}$$

this scheme is named as AM.4.

3.5. Kung and Traub(KT). The derivative free method by Kung and Traub[14],

$$\begin{aligned}
 y_m &= x_m - \frac{f(x_m)}{f[x_m, w_m]}, w_m = x_m + \beta_m f(x_m), \beta_m = \frac{1}{N(x_m)} \\
 z_m &= y_m - \frac{f(y_m)f(w_m)}{(f(w_m) - f(y_m))f[x_m, w_m]}, \\
 x_{m+1} &= z_m - \frac{f(y_m)f(w_m)(y_m - x_m \frac{f(x_m)}{f[x_m, z_m]})}{(f(y_m) - f(z_m))(f(w_m) - f(z_m))} + \frac{f(y_m)}{f[y_m, z_m]}.
 \end{aligned}$$

this method is named as KT.

3.6. Sharma et al. The method by Sharma et al. [22]

$$\begin{aligned}
 y_m &= x_m - \frac{f(x_m)}{\varphi(x_m)}, \varphi(x_m) = \frac{f(w_m) - f(x_m)}{\beta_m f(x_m)}, w_m = x_m + \beta_m f(x_m), \beta_m = \frac{-1}{N(x_m)} \\
 z_m &= y_m - H(\mu_m v_m) \frac{f(y_m)}{\varphi(x_m)}, H(\mu_m v_m) = \frac{1 + \mu_m}{1 - v_m}, v_m = \frac{f(y_m)}{f(w_m)}, \mu_m = \frac{f(y_m)}{f(x_m)} \\
 x_{m+1} &= z_m - \frac{f(z_m)}{f[z_m, y_m] + f[z_m, y_m, x_m](z_m - y_m) + f[z_m, y_m, x_m, w_m](z_m - y_m)(z_m - x_m)}.
 \end{aligned}$$

3.7. Zheng et al. The method by Zheng et al. [29]

$$\begin{aligned}
 y_m &= x_m - \frac{f(x_m)}{f[x_m, w_m]}, w_m = x_m + \beta_m f(x_m), \beta_m = \frac{-1}{N(x_m)} \\
 z_m &= y_m - \frac{f(y_m)}{f[y_m, x_m] + f[y_m, x_m, w_m](y_m - x_m)}, \\
 x_{m+1} &= z_m - \frac{f(z_m)}{f[z_m, y_m] + f[z_m, y_m, x_m](z_m - y_m) + f[z_m, y_m, x_m, w_m](z_m - y_m)(z_m - x_m)}.
 \end{aligned}$$

In order to test presented with-memory methods, we select the following nonlinear functions with initial approximation as x_0 and exact solution α . Our comparison is based on computational order of convergence and error computation.

Table 1. Test functions, exact root α and initial approximation x_0

Test Functions	α	x_0
$f_1(x) = \ln(x^2 - 2x + 2) + e^{x^2 - 5x + 4} \sin(x - 1)$	1	1.05
$f_2(x) = e^{x^2 + x \cos(x) - 1} \sin(\pi x) + x \ln(x \sin(x) + 1)$	0	0.2
$f_3(x) = \frac{(1 - \sin(x^2))(1 + x^2)}{1 + x^3} + x \ln(x^2 - \pi + 1) - \frac{1 + \pi}{1 + \pi^{3/2}}$	1.77	1.74
$f_4(x) = e^{x^2 - 3x} \sin(x) + \ln(x^2 + 1)$	0	0.35
$f_5(x) = (4 + 3 \sin(x) - 2x^2)^4$	1.854	1.86

Table 2. Errors and coc of AM12

f_n	$ x_1 - \alpha $	$ x_2 - \alpha $	$ x_3 - \alpha $	COC
f_1	5.49×10^{-8}	1.65×10^{-84}	9.27×10^{-1003}	12.9997
f_2	4.13×10^{-6}	9.44×10^{-66}	1.93×10^{-781}	12.9997
f_3	9.97×10^{-12}	5.28×10^{-134}	2.72×10^{-1601}	12.9997
f_4	3.42×10^{-6}	5.63×10^{-63}	2.24×10^{-744}	12.9997
f_5	2.52×10^{-3}	9.61×10^{-4}	3.61×10^{-4}	12.9997

$x_0 = 1.05$

Table 3. Error and coc of methods at f_1

	Scheme AM.1	Scheme AM.2	Scheme AM.3	Scheme AM.4
$ x_1 - \alpha $	0.314×10^{-6}	0.314×10^{-5}	0.543×10^{-6}	5.49×10^{-8}
$ x_2 - \alpha $	0.110×10^{-66}	0.153×10^{-61}	0.715×10^{-65}	1.65×10^{-84}
$ x_3 - \alpha $	0.178×10^{-800}	0.914×10^{-739}	0.100×10^{-788}	9.27×10^{-1003}
COC	12.1378	12.1056	12.1238	12.9997

$x_0 = 0.2$

Table 4. Error and coc of methods f_2

	Scheme AM.1	Scheme AM.2	Scheme AM.3	Scheme AM.4
$ x_1 - \alpha $	0.900×10^{-4}	0.585×10^{-3}	0.308×10^{-4}	4.13×10^{-6}
$ x_2 - \alpha $	0.111×10^{-44}	0.713×10^{-38}	0.650×10^{-49}	9.44×10^{-66}
$ x_3 - \alpha $	0.255×10^{-536}	0.126×10^{-454}	0.419×10^{-587}	1.93×10^{-781}
COC	12.0178	11.9363	12.0465	12.9997

$x_0 = 1.74$

Table 5. Error and coc of methods at f_3

	Scheme AM.1	Scheme AM.2	Scheme AM.3	Scheme AM.4
$ x_1 - \alpha $	0.836×10^{-8}	0.164×10^{-7}	0.605×10^{-8}	9.97×10^{-12}
$ x_2 - \alpha $	0.102×10^{-94}	0.101×10^{-96}	0.624×10^{-97}	5.28×10^{-134}
$ x_3 - \alpha $	0.134×10^{-1138}	0.118×10^{-1090}	0.341×10^{-1165}	2.72×10^{-1601}
COC	12.0110	12.0173	12.0048	12.9997

$x_0 = 0.35$

Table 6. Error and coc of methods at f_4

	Scheme AM.1	Scheme AM.2	Scheme AM.3	Scheme AM.4
$ x_1 - \alpha $	0.451×10^{-6}	0.290×10^{-7}	0.539×10^{-8}	3.42×10^{-12}
$ x_2 - \alpha $	0.612×10^{-63}	0.716×10^{-94}	0.335×10^{-96}	5.63×10^{-135}
$ x_3 - \alpha $	0.354×10^{-744}	0.612×10^{-1090}	0.148×10^{-1138}	2.24×10^{-1475}
COC	12.0110	12.0173	12.0048	12.9997

$x_0 = 1.86$

Table 7. Error and coc of methods at f_5

	Scheme AM.1	Scheme AM.2	Scheme AM.3	Scheme AM.4
$ x_1 - \alpha $	4.12×10^{-3}	3.21×10^{-5}	2.23×10^{-8}	2.52×10^{-14}
$ x_2 - \alpha $	0.542×10^{-94}	0.453×10^{-96}	0.342×10^{-45}	9.61×10^{-165}
$ x_3 - \alpha $	0.341×10^{-1138}	0.614×10^{-1090}	0.360×10^{-999}	3.61×10^{-1340}
<i>COC</i>	12.0110	12.0173	12.0048	12.9997

$f_1, x_0 = 1.35$

Table 8. Comparison of different methods

	KT	Sharma	Zheng
$ x_1 - \alpha $	0.845×10^{-4}	0.308×10^{-6}	0.148×10^{-5}
$ x_2 - \alpha $	0.393×10^{-45}	0.179×10^{-67}	0.157×10^{-61}
$ x_3 - \alpha $	0.100×10^{-540}	0.126×10^{-812}	0.481×10^{-738}
<i>COC</i>	11.9906	12.1688	12.0973

$f_2, x_0 = 0.6$

Table 9. Comparison of different methods

	KT	Sharma	Zheng
$ x_1 - \alpha $	0.798×10^{-3}	0.891×10^{-4}	0.214×10^{-4}
$ x_2 - \alpha $	0.194×10^{-40}	0.541×10^{-45}	0.168×10^{-53}
$ x_3 - \alpha $	0.976×10^{-486}	0.274×10^{-543}	0.386×10^{-642}
<i>COC</i>	11.8387	12.0827	11.9875

$f_3, x_0 = 1.7$

Table 10. Comparison of different methods

	KT	Sharma	Zheng
$ x_1 - \alpha $	0.241×10^{-8}	0.757×10^{-8}	0.221×10^{-7}
$ x_2 - \alpha $	0.137×10^{-99}	0.267×10^{-96}	0.140×10^{-90}
$ x_3 - \alpha $	0.283×10^{-1196}	0.561×10^{-1158}	0.546×10^{-1089}
<i>COC</i>	12.0190	12.0028	12.0001

$f_4, x_0 = 0.35$

Table 11. Comparison of different methods

	KT	Sharma	Zheng
$ x_1 - \alpha $	0.832×10^{-5}	0.412×10^{-7}	0.143×10^{-6}
$ x_2 - \alpha $	0.514×10^{-46}	0.342×10^{-68}	0.156×10^{-62}
$ x_3 - \alpha $	0.231×10^{-542}	0.135×10^{-815}	0.567×10^{-740}
<i>COC</i>	12.9997	12.9997	12.9997

$f_5, x_0 = 1.86$

Table 12. Comparison of different methods

	KT	Sharma	Zheng
$ x_1 - \alpha $	0.871×10^{-9}	0.898×10^{-9}	0.234×10^{-8}
$ x_2 - \alpha $	0.542×10^{-99}	0.675×10^{-98}	0.156×10^{-91}
$ x_3 - \alpha $	0.413×10^{-1198}	0.546×10^{-1160}	0.342×10^{-1090}
<i>COC</i>	12.9997	12.9997	12.9997

By the above results we can conclude that our methods computations agrees with the existing schemes.

4. Conclusion

We have introduced three point derivative free scheme of 12th order derived from King's family of methods for solving non-linear equations. This is an addition in three point with memory technique with twelfth order of convergence introducing new parameters. These parameters not only minimized extra functional and derivative evaluation but also increase convergence order of method. Efficiency index remarkably enhanced from 1.86121 to 1.867551871. It is noted that new scheme has improved performance to solve non linear equation. Numerical analyzing of recent methods with new method are given which shows the efficacy of presented method.

REFERENCES

- [1] Bradie, B., A friendly introduction to numerical analysis, Pearson Education, New Jersey, 2006.
- [2] Behl, R., González, D., Maroju, P., Motsa, S.S., An optimal and efficient general eighth-order derivative free scheme for simple roots. *Appl. Math. J. Comput.* 330, 666-675, 2018.
- [3] Behl, R., Motsa, S.S., Kansal, M., Kanwar, V. Fourth-order derivative-free optimal families of King's and Ostrowski's methods, Springer, 359-371, 2015.
- [4] Behl, R., Argyros, I.K., Motsa, S.S., A new highly efficient and optimal family of eighth-order methods for solving nonlinear equations, *Appl. Math. Comput.* 282, 175-186, 2015.
- [5] Changbum Chun, Some variants of King's fourth-order family of methods for nonlinear equations, *Appl. Math. Comput.* 190, 57-62, 2007.
- [6] Chun, C., Lee, M.Y., A new optimal eighth-order family of iterative methods for the solution of nonlinear equations, *Appl. Math. Comput.* 223, 506-519, 2013.
- [7] Cordero, A., Torregrosa, J.R., Vassileva, M.P., Three-step iterative methods with optimal eighth-order convergence, *J. Comput. Appl. Math.* 235, 3189-3194, 2011.
- [8] Cordero, A., Fardi, M., Ghasemi, M., Torregrosa, J.R., Accelerated iterative methods for finding solutions of nonlinear equations and their dynamical behavior, *Calcolo*, 1-14, 2012.
- [9] Chun, C., Lee, M.Y., Neta, B., Dzunic, J. On optimal fourth-order iterative methods free from second derivative and their dynamics, *Appl. Math. Comput.* 218, 6427-6438, 2012.
- [10] Geum, Kim, Neta, B. Developing an Optimal Class of Generic Sixteenth-Order Simple-Root Finders and Investigating Their Dynamics, 2019.
- [11] King, R.F., Family of four order methods for nonlinear equations, *SIAM J. Numer. Anal.* 10, 876-879, 1973.

1

*