

## A Family of Predictor-Corrector Methods Based on Weight Combination of Quadratures for Solving Nonlinear Equations

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**Abstract:** In this paper, we propose and analyze some new predictor-corrector methods for solving nonlinear equations using the weight combination of mid-point, Trapezoidal and Simpson quadrature formulas. We prove that these new methods are better than the *newton* method. Several examples are given to illustrate the efficiency of the proposed methods.

**Keywords:** predictor-corrector methods; convergence; quadrature formulas; nonlinear equations; numerical examples

### 1 Introduction

Finding the roots of non-linear equations are common yet important problem in science and engineering. Analytical methods for solving such equations are difficult or almost non-existent. Therefore it is only possible to obtain approximate solutions by numerical techniques based on iteration procedures [1,5,6]. It is well known that the quadrature formulas[1], play an important and significant rule in the evaluation of the integrals. It has been shown [2] that these quadrature formulas can be used to develop some iterative methods for solving nonlinear equations. we suggest and analyze some new-iterative methods by using the weight combination of the midpoint, Trapezoidal and Simpson quadrature formulas. This method is an implicit-type method. To implement this, we can use the *Newton* and the *Halley* methods and some newly developed method by Noor[2,3,4], as predictor method and then use this new method as a corrector method .A comparison between these new methods with that of *Newton* method is given. Several examples are given to illustrate the efficiency and advantage of these two-step methods.

### 2 Iterative methods

Suppose that  $r$  be the simple zero of a sufficiently differentiable function and consider the numerical solution of equation  $f(x) = 0$ , then

$$f(x) = f(x_n) + \int_{x_n}^x f'(t)dt. \tag{1}$$

If we approximate  $\int_{x_n}^x f'(t)dt$  with average of midpoint and Simpson quadrature formulas then we have

$$\int_{x_n}^x f'(t)dt = \frac{x - x_n}{2} f'(\frac{x_n + x}{2}) + \frac{x - x_n}{12} [f'(x_n) + 4f'(\frac{x_n + x}{2}) + f'(x)]. \tag{2}$$

From (2.1) and (2.2), we have

$$f(x) = f(x_n) + \frac{x - x_n}{12} [f'(x_n) + 10f'(\frac{x_n + x}{2}) + f'(x)].$$

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Since  $f(x) = 0$  then

$$x = x_n - \frac{12f(x_n)}{f'(x_n) + 10f'(\frac{x_n+x}{2}) + f'(x)}.$$

With this fixed point formulation and any iterative method as predictor we will have following implicit iterative method.

**Algorithm 1** For a given  $x_0$ , compute the approximate solution  $x_{n+1}$  by iterative scheme.

$$x_{n+1} = x_n - \frac{12f(x_n)}{f'(x_n) + 10f'(\frac{x_n+x}{2}) + f'(x)}.$$

Using the predictor type technique, we suggest the following two-step method which is obtained by combining the *Halley* method.

**Algorithm 2** For a given  $x_0$ , compute the approximate solution  $x_{n+1}$  by the iterative scheme.

$$y_n = x_n - \frac{2f(x_n)f'(x_n)}{2(f'(x_n))^2 - f(x_n)f''(x_n)},$$

$$x_{n+1} = x_n - \frac{12f(x_n)}{f'(x_n) + 10f'(\frac{x_n+y_n}{2}) + f'(y_n)}.$$

For approximating  $\int_{x_n}^x f'(t)dt$ , if we combine Trapezoidal and Simpson quadrature formulas with weight factor  $\frac{1}{2}$ , then we have

$$\int_{x_n}^x f'(t)dt = \frac{x-x_n}{4}[f'(x_n) + f'(x)] + \frac{x-x_n}{12}[f'(x_n) + 4f'(\frac{x_n+x}{2}) + f'(x)]. \quad (3)$$

So from (2.1) and (2.3) and  $f(x) = 0$ , we can obtain

$$x = x_n - \frac{3f(x_n)}{f'(x_n) + f'(\frac{x_n+x}{2}) + f'(x)}.$$

In similar way we will have following algorithm which our predictor is the well-known *Newton* method.

**Algorithm 3** For a given  $x_0$ , compute the approximate solution  $x_{n+1}$  by the iterative scheme.

$$y_n = x_n - \frac{f(x_n)}{f'(x_n)},$$

$$x_{n+1} = x_n - \frac{3f(x_n)}{f'(x_n) + f'(\frac{x_n+y_n}{2}) + f'(y_n)}.$$

For approximating  $\int_{x_n}^x f'(t)dt$ , if we combine mid-point, Trapezoidal and Simpson quadrature formulas with  $\frac{1}{4}$ ,  $\frac{1}{4}$  and  $\frac{1}{2}$  weight factors respectively, then we will have

$$\int_{x_n}^x f'(t)dt = \frac{x-x_n}{4}f'(\frac{x_n+x}{2}) + \frac{x-x_n}{8}[f'(x_n) + f'(x)] + \frac{x-x_n}{12}[f'(x_n) + 4f'(\frac{x_n+x}{2}) + f'(x)]. \quad (4)$$

Since  $f(x) = 0$ , from (2.1) and (2.4) we obtain following fixed point formulation.

$$x = x_n - \frac{24f(x_n)}{5f'(x_n) + 14f'(\frac{x_n+x}{2}) + 5f'(x)}.$$

Same as algorithm 2. with selecting the *Halley* method as a predictor we will have following algorithm

**Algorithm 4** For a given  $x_0$ , compute the approximate solution  $x_{n+1}$  by the iterative scheme.

$$y_n = x_n - \frac{2f(x_n)f'(x_n)}{2(f'(x_n))^2 - f(x_n)f''(x_n)},$$

$$x_{n+1} = x_n - \frac{24f(x_n)}{5f'(x_n) + 14f'(\frac{x_n+y_n}{2}) + 5f'(y_n)}.$$

### 3 Convergence analysis

In this section, we consider the convergence of Algorithm 3. In similar way, one can prove the convergence of other two step algorithms.

**Theorem 5** Let  $r \in I$  be a sample zero of sufficiently differentiable function  $f : I \subseteq R \rightarrow R$  for an open interval  $I$ . If  $x_0$  is sufficiently close to  $r$ , then the two step method defined by Algorithm 3 has quadratic convergence and it's asymptotic convergence is  $\frac{f^{(2)}(r)}{6f'(r)}$ .

**Proof.** Consider to

$$y_n = x_n - \frac{f(x_n)}{f'(x_n)}, \tag{5}$$

$$x_{n+1} = x_n - \frac{3f(x_n)}{f'(x_n) + f'(\frac{x_n+y_n}{2}) + f'(y_n)}. \tag{6}$$

Let  $r$  be a simple zero of  $f$ . Since  $f$  is sufficiently differentiable, by expanding  $f(x_n)$  and  $f'(x_n)$  about  $r$ , we get

$$f(x_n) = f(r) + (x_n - r)f'(r) + \frac{(x_n - r)^2}{2!}f^{(2)}(r) + \frac{(x_n - r)^3}{3!}f^{(3)}(r) + \frac{(x_n - r)^4}{4!}f^{(4)}(r) + \dots,$$

then

$$f(x_n) = f'(r)[e_n + c_2e_n^2 + c_3e_n^3 + c_4e_n^4 + \dots], \tag{7}$$

and

$$f'(x_n) = f'(r)[1 + 2c_2e_n + 3c_3e_n^2 + 4c_4e_n^3 + 5c_5e_n^4 + \dots], \tag{8}$$

where  $c_k = \frac{1}{k!} \frac{f^{(k)}(r)}{f'(r)}$ ,  $k = 1, 2, 3, \dots$  and  $e_n = x_n - r$ .

Now, from (3.7) and (3.8), we have

$$\frac{f(x_n)}{f'(x_n)} = e_n - c_2e_n^2 + 2(c_2^2 - c_3)e_n^3 + (-7c_2c_3 + 4c_2^3 + 3c_4)e_n^4 + \dots, \tag{9}$$

From (3.5) and (3.9), we get

$$y_n = r + c_2e_n^2 + 2(c_3 - c_2^2)e_n^3 + (-7c_2c_3 + 4c_2^3 + 3c_4)e_n^4 + \dots \tag{10}$$

From (3.10), we get

$$f(y_n) = f'(r)[(y_n - r) + c_2(y_n - r)^2 + c_3(y_n - r)^3 + c_4(y_n - r)^4 + \dots]$$

and

$$\begin{aligned} f'(y_n) &= f'(r)[1 + 2c_2(y_n - r) + 3c_3(y_n - r)^2 + 4c_4(y_n - r)^3 + 5c_5(y_n - r)^4 + \dots] \\ &= f'(r)[1 + 2c_2^2e_n^2 + 4(c_2c_3 - c_2^3)e_n^3 + (-11c_2^2c_3 + 8c_2^4 + 6c_2c_4)e_n^4 + \dots]. \end{aligned}$$

Expanding  $f'(\frac{x_n+y_n}{2})$  about  $r$ , we get

$$\begin{aligned} f'(\frac{x_n+y_n}{2}) &= f'(r)[1 + 2c_2(\frac{x_n+y_n}{2} - r) + 3c_3(\frac{x_n+y_n}{2} - r)^2 + 4c_4(\frac{x_n+y_n}{2} - r)^3 + \dots] \\ &= f'(r)[1 + 2c_2e_n + (2c_2^2 + \frac{3}{4}c_3 + \frac{1}{2}c_4)e_n^2 + (4c_2c_3 - 4c_2^3)e_n^3 \\ &\quad + (\frac{-61}{4}c_2^2c_3 + 8c_2^4 + 6c_2c_4)e_n^4 + \dots]. \end{aligned}$$

Then

$$f'(x_n) + f'(y_n) + f'\left(\frac{x_n + y_n}{2}\right) = 3f'(r)\left[1 + \frac{4}{3}c_2e_n + \frac{1}{3}(4c_2^2 + \frac{15}{4}c_3 + \frac{1}{2}c_4)e_n^2 + \frac{1}{3}(4c_4 + 8c_2c_3 - 8c_2^3)e_n^3 + \frac{1}{3}\left(\frac{-97}{4}c_2c_3 + 5c_5 + 16c_2^4 + 12c_2c_4\right)e_n^4 + \dots\right].$$

From (3.6),  $e_{n+1} = x_{n+1} - r$  and  $e_n = x_n - r$

$$e_{n+1} = e_n - \frac{3f(x_n)}{f'(x_n) + f'\left(\frac{x_n + y_n}{2}\right) + f'(y_n)}.$$

Then we will have

$$e_{n+1} = e_n - [e_n + c_2e_n^2 + c_3e_n^3 + c_4e_n^4 + \dots]\left[1 - \left(\frac{4}{3}c_2e_n + \frac{1}{3}(4c_2^2 + \frac{15}{4}c_3 + \frac{1}{2}c_4)e_n^2 + \frac{1}{3}(4c_4 + 8c_2c_3 - 8c_2^3)e_n^3 + \dots\right) + \left(\frac{4}{3}c_2e_n + \frac{1}{3}(4c_2^2 + \frac{15}{4}c_3 + \frac{1}{2}c_4)e_n^2 + \dots\right)^2 + \dots\right].$$

Finally

$$e_{n+1} = e_n - \left(e_n + \left(c_2 - \frac{4}{3}c_2\right)e_n^2 + \left(-\frac{4}{3}c_2^2 - \frac{5}{4}c_3 - \frac{1}{6}c_4 - \frac{4}{3}c_2^2 + c_3\right)e_n^3 + \dots\right),$$

$$e_{n+1} = \frac{c_2}{3}e_n^2 + \left(\frac{8c_2^2}{3} + \frac{c_3}{4} + \frac{c_4}{6}\right)e_n^3 + \dots$$

$$\lim_{n \rightarrow \infty} \frac{e_{n+1}}{e_n^2} = \frac{c_2}{3} = \frac{f^{(2)}(r)}{6f'(r)}.$$

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Since asymptotic convergence of *Newton* method is  $c_2$  and from Theorem 5, we result that the convergence rate of Algorithm 3 is better than the *Newton* method.

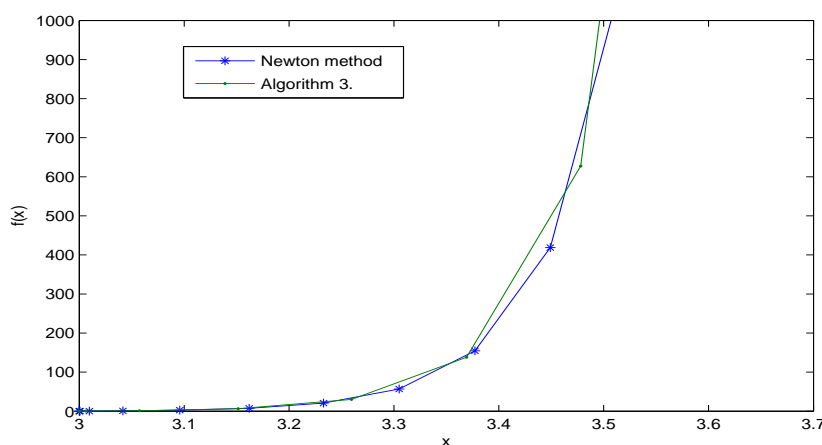


Figure 1: The number of iteration between the *Newton* method and Algorithm 3. with common starting value  $x_0 = 4$

## 4 Numerical experiments

In all of our examples, the maximum number of iteration is  $n = 200$  and our examples are tested with precision  $\varepsilon = 1 \times 10^{-15}$ . The following stopping criteria is used for computer programs:

(i)  $|f(x_{n+1})| < \varepsilon$ . (ii)  $|x_{n+1} - x_n| < \varepsilon$ .

Table 1 presents iteration number comparison of algorithms 2, 3 and 4 with the *Newton* method in given precision. In Table 2, the CPU time ( per second ) of our algorithms and *Newton* method are compared . All numerical results show here, are obtained on a pentium IV processor at 3.00 GHz.

Fig. 1 presents convergence comparison between the *Newton* method and Algorithm 3 for  $f(x) = e^{x^2+7x-30} - 1$  from starting value  $x_0 = 4$ .

Table 1: Examples and comparison between algorithms.

Equation	$x_0$	Newton	Algorithm 2	Algorithm 3	Algorithm 4	$x_n$
$e^{x^2+7x-30} - 1 = 0$	4	20	9	13	10	3.0000000000000000
$x^3 - 10 = 0$	1.5	7	4	5	4	2.154434690031884
$x^2 - e^x - 3x + 2 = 0$	2	6	4	4	4	-1.207647827130919
$\sin^2(x) - x^2 + 1 = 0$	-1	7	4	5	4	-1.404491648215341
$x^{10} - 1 = 0$	1.5	10	5	7	6	1.0000000000000000
$11x^{11} - 1 = 0$	0.7	8	4	6	4	0.804133097503664
$\sin(\frac{1}{x}) - x = 0$	2	6	4	4	4	0.897539461280487

## 5 Conclusion

In Theorem 1. we proved that asymptotic convergence of algorithm 3. is less than the *Newton* method. Then this two step method is better than the *Newton* method . One can prove that other two-step algorithms proposed here are better than the *Newton* method too. In Table 1. we can see accuracy and efficiency of our two-step methods when compared with the *Newton* method.

Table 2: The CPU time ( per second ) of algorithms.

Equation	Newton	Algorithm 2	Algorithm 3	Algorithm 4
$e^{x^2+7x-30} - 1 = 0$	0.171875	0.078125	0.109375	0.093750
$x^3 - 10 = 0$	0.078125	0.031250	0.031250	0.046875
$x^2 - e^x - 3x + 2 = 0$	0.046875	0.031250	0.031250	0.031250
$\sin^2(x) - x^2 + 1 = 0$	0.046875	0.031250	0.046875	0.031250
$x^{10} - 1 = 0$	0.078125	0.031250	0.046875	0.031250
$11x^{11} - 1 = 0$	0.062500	0.031250	0.031250	0.031250
$\sin(\frac{1}{x}) - x = 0$	0.046875	0.031250	0.031250	0.031250

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