

A new optimal iterative algorithm for solving nonlinear equations

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Abstract: The aim of this paper is to propose a new iterative algorithm (scheme or method) for solving algebraic and transcendental equations, considering a fixed point and an initial guess value on the x -axis. The concepts of the slope of a line and the Taylor series are used in the derivation. The algorithm has second-order convergence and requires two function evaluations in each step, which shows that it is optimal with a computational efficiency index of 1.414 and an informational efficiency of 1. The validity of the algorithm is examined by solving some examples and their comparisons with Newton's method.

Keywords: iterative algorithm; second-order convergence; nonlinear equations; Newton's method; optimal method

1. Introduction

Solving an equation $f(x) = 0$ always receives attraction for investigators due to its various applications in science, technology, and engineering. Many investigators [1–8] developed new iterative methods for better and faster convergence. Among various numerical methods, optimal methods draw significant attention due to their computational and informational efficiency. Any n -step iterative method is optimal when it reaches the order of convergence 2^n and requires $(n + 1)$ function evaluations. As mentioned above, the most important parameters of computational efficiency and informational efficiency are usually expressed [9,10] by

$$E = r^{\frac{1}{c}} \text{ and } I = \frac{r}{c},$$

respectively, where r is the order of convergence and c is the computational cost (number of function evaluations in each step).

In this paper, a new optimal iterative algorithm (scheme or method) for solving an equation $f(x) = 0$ is proposed, wherein a fixed point x_p and an initial guess value x_0 on the x -axis are considered. The concepts of the slope of a line and Taylor's series are used in the derivation. The algorithm is a single-step and has second-order convergence. The validity of the algorithm is examined by solving some examples and their comparisons with Newton's method. The algorithm works for both when $f'(x) \neq 0$ and $f'(x) = 0$.

2. The proposed optimal iterative algorithm

Figure 1 shows a geometric view of the proposed algorithm for solving $f(x) = 0$. Let ξ be the exact (simple) root in $I \subset R$, where I is some open interval and R is a set of real numbers. Let $f(x)$ and its first and second-order derivatives are

continuous nearer to ξ . In order to develop a proposed algorithm, consider a fixed point $A(x_p, 0)$ and the initial guess value (if possible, sufficiently close to ξ) $B(x_0, 0)$ on the x -axis. Draw a perpendicular from B to meet the curve $y = f(x)$ at $C(x_0, f(x_0))$ as shown in figure. Then for $0 < k < 1$, let $D(x_0, kf(x_0))$ be a point on inside the line segment \overline{BC} . Draw a line segment \overline{AD} . By considering the same slope α of \overline{AD} , draw another line segment \overline{BE} giving the first approximation x_1 (say $F(x_1, 0)$) by the following procedure, where $x_1 = x_0 + h$.

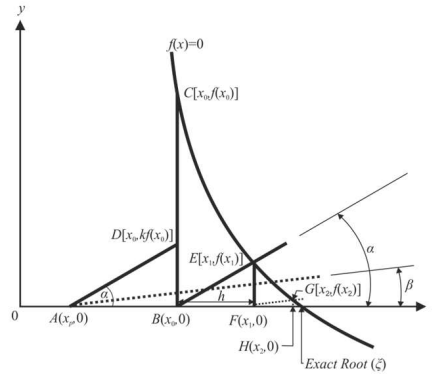


Figure 1. The proposed algorithm.

As discussed above, Slope of $\overline{AD} = \text{Slope of } \overline{BE}$ implies

$$\frac{kf(x_0)}{x_0 - x_p} = \frac{f(x_1)}{x_1 - x_0} \tag{1}$$

Using $x_1 = x_0 + h$ and Taylor series $f(x_1) = f(x_0 + h) \approx f(x_0) + hf'(x_0)$, Equation (1) becomes

$$x_1 = \frac{(x_0 - x_p)[f(x_0) - x_0f'(x_0)] + kx_0f(x_0)}{kf(x_0) - (x_0 - x_p)f'(x_0)}.$$

The repetition of the above procedure gives general algorithm as

$$x_{n+1} = \frac{(x_n - x_p)[f(x_n) - x_nf'(x_n)] + kx_nf(x_n)}{kf(x_n) - (x_n - x_p)f'(x_n)}; 0 < k < 1, n = 0, 1, 2, \dots \tag{2}$$

This algorithm converges to the root ξ , provided $kf(x_n) - (x_n - x_p)f'(x_n) \neq 0$. Here, the value of k is restricted to $0 < k < 1$ because of the following reasons.

- (a) At $k = 0, \alpha = 0$ and the Algorithm (2) leads to Newton's formula.
- (b) At $k = 1, \alpha$ takes the maximum possible angle, which may result in more iterations and delay in convergence.

It should be noted here that after the first iteration, due to $kf(x_1)$, there is a decrease in slope (say $\beta, \beta < \alpha$), which continues for further iterations as well.

3. Convergence analysis

In this article, the generalized convergence rate is considered similar to that of Newton's formula by Taylor series expansion.

Let ε_n and ε_{n+1} be the errors in the n^{th} and $(n + 1)^{\text{th}}$ iterations, respectively, then

$$\varepsilon_n = \xi - x_n, \varepsilon_{n+1} = \xi - x_{n+1}. \tag{3}$$

Using Equation (3), right-hand side of Algorithm (2) with Taylor's expansion

up to $O(\varepsilon_n^2)$ and the fact that $f(\xi) = 0$, gives

$$\begin{aligned} \frac{(x_n - x_p)[f(x_n) - x_n f'(x_n)] + kx_n f(x_n)}{kf(x_n) - (x_n - x_p)f'(x_n)} &= \frac{a + b\varepsilon_n + c\varepsilon_n^2}{d + e\varepsilon_n + g\varepsilon_n^2} = \frac{a + b\varepsilon_n + c\varepsilon_n^2}{d\left(1 + \frac{e}{d}\varepsilon_n + \frac{g}{d}\varepsilon_n^2\right)} = \frac{\frac{a}{d} + \frac{b}{d}\varepsilon_n + \frac{c}{d}\varepsilon_n^2}{1 + \left(\frac{e}{d}\varepsilon_n + \frac{g}{d}\varepsilon_n^2\right)} \\ &= \left(\frac{a}{d} + \frac{b}{d}\varepsilon_n + \frac{c}{d}\varepsilon_n^2\right) \left[1 - \left(\frac{e}{d}\varepsilon_n + \frac{g}{d}\varepsilon_n^2\right) + \left(\frac{e}{d}\varepsilon_n + \frac{g}{d}\varepsilon_n^2\right)^2\right] \\ &= \frac{a}{d} + \left(\frac{bd - ae}{d^2}\right)\varepsilon_n + \left(\frac{c}{d} - \frac{be}{d^2} - \frac{ag}{d^2} + \frac{ae^2}{d^3}\right)\varepsilon_n^2, \end{aligned}$$

where

$$\begin{aligned} a &= -\xi^2 f'(\xi) + \xi x_p f'(\xi), \\ b &= \xi^2 f''(\xi) + \xi f'(\xi) - \xi x_p f''(\xi) - k\xi f'(\xi), \\ c &= -\frac{3}{2}\xi f''(\xi) - \frac{1}{2}\xi^2 f'''(\xi) + \frac{x_p}{2} f''(\xi) + \frac{1}{2}\xi x_p f'''(\xi) + \frac{k\xi}{2} f''(\xi) + kf'(\xi), \\ d &= -\xi f'(\xi) + x_p f'(\xi), \\ e &= -kf'(\xi) + \xi f''(\xi) + f'(\xi) - x_p f''(\xi), \\ g &= \frac{k}{2} f''(\xi) - \frac{\xi}{2} f'''(\xi) - f''(\xi) + \frac{x_p}{2} f'''(\xi). \end{aligned}$$

Since

$$\frac{a}{d} = \xi \text{ and } \frac{bd - ae}{d^2} = 0,$$

Algorithm (2) becomes

$$\varepsilon_{n+1} \approx \varepsilon_n^2 \text{ (Finite quantity).}$$

Thus, the algorithm converges quadratically.

4. Results and discussion

From the above sections, it is noticed that the proposed algorithm is single-step, has second-order convergence, and uses two function evaluations in each step with the computational efficiency index $E = 2^{1/2} = 1.414$ and informational efficiency $I = \frac{2}{2} = 1$, which is same as Newton’s method. Thus, the proposed algorithm for solving an equation $f(x) = 0$ is optimal, as per the definition given in the introduction section.

The comparative study between the proposed Algorithm (2) and Newton’s method is given in **Tables 1–6**, where $|x_{n+1} - x_n| < 0.00001$.

Table 1. Comparative study of the proposed Algorithm (2) with Newton’s method for $x^3 - x^2 - x - 1 = 0$ considering different values of x_p , and same values of x_0 and k .

(1) Initial guess value x_0	(2) Value of x_p	(3) Value of k	(4) Number of iterations required by the proposed Algorithm (2)	(5) Number of iterations required by Newton’s method	(6) Solution of the given equation by the proposed Algorithm (2)	(7) Solution of the given equation by Newton’s method
	-3.0					
1.5	-1.5	0.00001	5	5	1.83928675521416	1.83928675521416
	3.0					
	6.0					

Table 2. Comparative study of the proposed Algorithm (2) with Newton’s method for different equations considering the same values of x_0 , x_p and k .

(1) Equations	(2) Initial guess value x_0	(3) Value of x_p	(4) Value of k	(5) Number of iterations required by the proposed Algorithm (2)	(6) Number of iterations required by Newton’s method	(7) Solution of the given equations by the proposed Algorithm (2)	(8) Solution of the given equations by Newton’s method
$x^3 - x^2 - x + 1 = 0$	1.5	0.5		16	16	1.000009413559562	1.000009413617196
$4x^4 - 4x^2 = 0$	$\frac{\sqrt{21}}{7}$	1.65		19	26	0.000007441381281	0.000007484300647
$x^3 - e^{-x} = 0$	0	-1		5	5	0.772882959152200	0.772882959152202
$\sin x = 0$	1.5	0.5	0.00001	4	4	-12.566370614359171	-12.566370614359172
$x^{10} - 1 = 0$	0.5	-0.5		43	43	1.000000000000000	1.000000000000000
$e^{x^2+7x-30} - 1 = 0$	3.5	2		11	11	3.000000000000253	3.000000000000253
$x^3 + x^2 - 2 = 0$	2.2	0		6	6	1.000000000000000	1.000000000000000
$(x - 2)^{23} - 1 = 0$	3.5	-4		13	13	3.000000000022981	3.000000000022981

Table 3. Comparative study of the proposed Algorithm (2) with Newton’s method for $xe^{x^2} - \sin^2 x + 3 \cos x + 5 = 0$ considering different values of x_0 and the same values of x_p and k .

(1) Equation	(2) Initial guess value x_0	(3) Value of x_p	(4) Value of k	(5) Number of iterations required by the proposed Algorithm (2)	(6) Number of iterations required by Newton’s method	(7) Solution of the given equation by the proposed Algorithm (2)	(8) Solution of the given equation by Newton’s method
$xe^{x^2} - \sin^2 x + 3 \cos x + 5 = 0$	-1			5	5	-1.207647827130919	-1.207647827130919
	-2	0	0.00001	7	7	-1.207647827173526	-1.207647827173531

Table 4. Comparative study of the proposed Algorithm (2) with Newton’s method for $\sin x = 0$ considering different values of k and the same values of x_0 and x_p .

(1) Initial guess value x_0	(2) Value of x_p	(3) Value of k	(4) Number of iterations required by the proposed Algorithm (2)	(5) Number of iterations required by Newton’s method	(6) Solution of the given equation by the proposed Algorithm (2)	(7) Solution of the given equation by Newton’s method
1.5	0.5	0.5	6	4	3.141592653589793	-12.566370614359172
		0.25	5		6.283185307179575	
		0.125	7		18.849555921538759	
		0.0625	6		-116.238928182822363	
		0.03125	7		-31.415926535897924	
		0.015625	5		-15.707963267948967	
		0.0078125	5		-18.849555921538759	
		0.00390625	5		-12.566370614359171	
		0.001953125	4		-12.566370614359164	
		0.0009765	4		-12.566370614359172	
0.00001	4	-12.566370614359171				

Table 5. Comparative study of the proposed Algorithm (2) with Newton’s method for $x^{10} - 1 = 0$ considering different values of k and same values of x_0 and x_p .

(1) Initial guess value x_0	(2) Value of x_p	(3) Value of k	(4) Number of iterations required by the proposed Algorithm (2)	(5) Number of iterations required by Newton’s method	(6) Solution of the given equation by the proposed Algorithm (2)	(7) Solution of the given equation by Newton’s method
0.5	-0.5	0.5	13		1.00000000005467	
		0.25	19		1.000000000000000	
		0.125	24		1.000000000000095	
		0.0625	29		1.00000000002376	
		0.03125	34		1.000000000000000	
		0.015625	37	43	1.00000000000537	1.000000000000000
		0.0078125	39		1.00000000226450	
		0.00390625	41		1.00000000000017	
		0.001953125	42		1.000000000000000	
		0.0009765	42		1.00000000002626	
	0.00001	43		1.000000000000000		

Table 6. Comparative study of the proposed Algorithm (2) with Newton’s method for $x^3 + x^2 - 2 = 0$ with the same values of x_0 , x_p and k .

(1) Initial guess value x_0	(2) Value of x_p	(3) Value of k	(4) Number of iterations required by the proposed Algorithm (2)	(5) Number of iterations required by Newton’s method	(6) Solution of the given equation by the proposed Algorithm (2)	(7) Solution of the given equation by Newton’s method
0	2	0.00001	36	Fails	1.00000000020332	Fails

Table 1 shows the result for $x^3 - x^2 - x - 1 = 0$ considering different values of x_p and the same values of x_0 and k , where the value $x_0 = 1.5$ is chosen by using the intermediate value theorem (IVT). It should be noted here that the use of IVT is not necessary. The table shows that the proposed algorithm gives the same result as Newton’s method correct up to at least fourteen decimal places with the same number of iterations when $k = 0.00001$.

Table 2 shows results for different equations considering the same values of x_0 , x_p and k ($k = 0.00001$).

Tables 1 and **2** show that for this value of k , the proposed algorithm has almost good agreement of results with Newton’s method.

Table 3 shows results for $xe^{x^2} - \sin^2 x + 3 \cos x + 5 = 0$ considering different values of x_0 and the same values of x_p and k . The table shows that a change in x_0 leads to a change in the number of iterations for getting the same result by both proposed and Newton’s methods.

Table 4 shows results for $\sin x = 0$ considering different values of k and the same values of x_0 and x_p . This table shows the case of multiple roots of an equation. It is observed that the proposed algorithm gives different roots for different values of k with the variation in the number of iterations.

Table 5 shows results for $x^{10} - 1 = 0$ considering different values of k , and the same values of x_0 and x_p . Here, it is observed that the proposed algorithm gives the same root as that of Newton’s method with the variation in the number of iterations.

Table 6 shows results for $x^3 + x^2 - 2 = 0$ with the same values of x_0, x_p and k . It is observed that the proposed algorithm works even if $f'(x) = 0$, which is the limitation of the methods suggested by authors [1–7]. Of course, Newton’s method is also unable to give a solution, but according to Traub [10], in such cases, it converges linearly.

5. Conclusions

This paper proposes a new single-step, second-order optimal iterative algorithm for finding the root of an equation $f(x) = 0$, wherein the concepts of the slope of a line and Taylor’s series are used in the derivation. The algorithm has a computational efficiency index of 1.414 and an informational efficiency of 1, and it works for both $f'(x) \neq 0$ and $f'(x) = 0$. Usually, the condition $f'(x) = 0$ is the limitation of the methods suggested by authors [1–7]. The algorithm may be considered as generalization of Newton’s method, involving two parameters k and x_p . The proposed algorithm is valid if

$$kf(x_n) - (x_n - x_p)f'(x_n) \neq 0.$$

If $kf(x_n) - (x_n - x_p)f'(x_n) = 0$, then

$$kf(x_n) = (x_n - x_p)f'(x_n). \tag{4}$$

As $0 < k < 1$, and x_n and x_p are two different points, so $k \neq 0$ and $(x_n - x_p) \neq 0$. Hence, $kf(x_n) = (x_n - x_p)f'(x_n)$ is satisfied when $f(x_n) = 0$ and $f'(x_n) = 0$ simultaneously.

Again, rearranging Equation (4) implies

$$\frac{k}{x_n - x_p} = \frac{f'(x_n)}{f(x_n)} \tag{5}$$

The satisfaction of the above condition is dependent on k and x_p . If, at any stage of iteration, Equation (5) is satisfied, then the proposed algorithm fails, but it happens rarely.

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