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Some New Iterative Algorithms for Solving **One-Dimensional Non-Linear Equations and Their Graphical Representation**

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ABSTRACT Solving non-linear equation is perhaps one of the most difficult problems in all of numerical computations, especially in a diverse range of engineering applications. The convergence and performance characteristics can be highly sensitive to the initial guess of the solution for most numerical methods such as Newton's method. However, it is very difficult to select reasonable initial guess of the solution for most systems of non-linear equations. Besides, the computational efficiency is not high enough. Taking this into account, based on variational iteration technique, we develop some new iterative algorithms for solving one-dimensional non-linear equations. The convergence criteria of these iterative algorithms has also been discussed. The superiority of the proposed iterative algorithms is illustrated by solving some test examples and comparing them with other well-known existing iterative algorithms in literature. In the end, the graphical comparison of the proposed iterative algorithms with other well-known iterative algorithms have been made by means of polynomiographs of different complex polynomials which reflect the fractal behavior and dynamical aspects of the proposed iterative algorithms.

INDEX TERMS Order of convergence, non-linear equations, Newton's method, Househölder's method, polynomiography.

I. INTRODUCTION

Solving non-linear equations of the form f(x) = 0 is one of the most important problems in all of numerical computations, especially in a diverse range of engineering applications. Many applied problems can be reduced to solving systems of non-linear equations, which is one of the most basic problems in Mathematics. This task has applications in many scientific fields of engineering and computer sciences.

To find solution of such problems, analytical methods do not assist us and therefore, we have to find appropriate solution of such equations by numerical methods which are totally based on iterative schemes. In an iterative scheme, we start the algorithm by choosing an inial guess x_0 which is refined step by step by means of iterations until the approximated solution is achieved. Some basic iterative methods are given in literature [4], [6], [9], [29], [34], [35] and the references

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therein. The most famous and well-known method for finding roots of non-linear equations is of the following form:

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

Which is quadratically convergent Newton's method [35] for the solution of non-linear equations.

For acquiring better convergence, many scholars proposed a large number of iterative algorithms by means of of various types of mathematical techniques. In 1970, Househölder [13] suggested a cubically convergent method using Taylor's series expansion. Using the modified Adomian decomposition technique, Abbasbandy [1] proposed a single-step iterative method having convergence of order three in 2003. After that, in 2007, Kou [21] improved some third-order modifications of Newton's method and obtained many new methods for solving non-linear equations. Nazeer et al. [25], [26] in (2016), suggested some new algorithms by means of finite difference scheme, which are free from second derivatives, having higher convergence orders. In the same year, Sharma and Arora [32] proposed a simple yet efficient family of three-point iterative methods with eighth order of convergence for solving non-linear equations. After that, in [30], the authors not only developed an optimal class of three-step eighth-order methods with higher-order weight functions employed in the second and third sub-steps, but also investigate their dynamics underlying the purely imaginary extraneous fixed points. In 2019, Alharbi et al. [2] suggested some higher order algorithms by decomposition technique, having convergence orders up to fifteen. Very recently, Behl and Martinez [3] constructed some new high order and efficient iterative techniques for solving a system of non-linear equations whose convergence orders varied and reached up to six. The suggested methods are the extension of the Jisheng Kou's method [22] for multidimensional case.

In this article, we proposed three new iterative algorithms using variational iteration technique by considering two auxiliary iteration functions $\phi(x)$ and $\psi(x)$. The first one, $\phi(x)$ acts as a predictor function having convergence of order qwhere $q \ge 1$. The predictor function is used to attain iterative methods of convergence order q + r, where $r \ge 1$ is the convergence order of second auxiliary function $\psi(x)$. By an iteration function, we mean a function f which is obtained by composing the function itself for a certain number of times and the convergence order is a quantity that shows how fast an iterative method approaches to the required approximate root. Mathematically, it is written as $\lim_{n\to\infty} \frac{|e_{n+1}|}{|e_n|^p} = v$, here $p \ge 1$ is called the convergence order, the constant v is the rate of convergence or asymptotic error constant and e_n, e_{n+1} are the errors at *n*th and (n + 1)th iterations respectively.

Using variational iteration technique, we develop some new higher order iterative algorithms with better performance and efficiency. The variational iteration technique was introduced by Inokuti *et al.* [15]. Using this technique, Noor [27] and Noor and Shah [28] proposed some iterative methods for the solution of non-linear equations. The purpose of this technique was to solve a variety of diverse problems [10]–[12].

Now we apply the described technique to obtain higher order iterative algorithms. These algorithms are very fast using less number of iterations to reach the required solution, free from 3rd and higher derivatives with ninth order of convergence which raises their efficiency indices. The convergence criteria of the suggested algorithms is also discussed. Various test examples have been solved to show their performance as compare to the other similar existing iterative algorithms in literature.

II. CONSTRUCTION OF SOME NEW ITERATIVE ALGORITHMS

In this section, we construct some new iterative algorithms by means of variational iteration technique. These algorithms are multi-step iterative methods which involve predictor and corrector steps. The proposed iterative algorithms have higher order of convergence than one-step algorithms. By applying variational iteration technique, we derive some new iterative algorithms of order q + r where $q, r \ge 1$ are the orders of convergence of the auxiliary iteration functions $\phi(x)$ and $\psi(x)$ respectively.

Now consider the non-linear equation of the form

$$f(x) = 0 \tag{2}$$

Suppose that α is the simple root and γ is the initial guess sufficiently close to α . For better understanding and to deliver the basic idea, we suppose the approximate solution x_n of (2) such that

$$f(x_n)\neq 0$$

We consider $\phi(x)$ and $\psi(x)$ as two iteration functions of order q and r respectively. Then

$$x_{n+1} = \phi(x_n) + \mu[f(\psi(x_n))g(\psi(x_n))]^t$$
(3)

where $t = \frac{q}{r}$ is a recurrence relation which generates iterative algorithms of order q + r and g(x) is any smooth arbitrary function which later on is converted to $g(\psi(x_n))$ and μ is a parameter, called the "Lagrange's multiplier" and can be determined by using the optimality criteria on (3) by setting first derivative of (3) with respect to x_n equal to zero as follows:

$$\frac{dx_{n+1}}{x_n} = 0 \tag{4}$$

Which leads us to the following equality:

$$\mu = -\frac{\phi'(x_n)[f(\psi(x_n))g(\psi(x_n))]^{1-t}}{t\psi'(x_n)[f'(\psi(x_n))g(\psi(x_n)) + f(\psi(x_n))g'(\psi(x_n))]}$$
(5)
From (3) and (5), we get

$$x_{n+1} = \phi(x_n) - \frac{\phi'(x_n)}{t\psi'(x_n)} \frac{f(\psi(x_n))g(\psi(x_n))}{[f'(\psi(x_n))g(\psi(x_n)) + f(\psi(x_n))g'(\psi(x_n))]}$$
(6)

Now we are going to apply (6) for constructing a general iterative scheme for iterative methods. For this, suppose that

$$\psi(x_n) = y_n$$

= $x_n - \frac{f(x_n)}{f'(x_n)} - \frac{f^2(x_n)f''(x_n)}{2[f'^3(x_n) - f(x_n)f'(x_n)f''(x_n)]}$ (7)

Which is well-known Golbabai and Javidi's method with cubic convergence [8]. With the help of (6) and (7), we can write

$$x_{n+1} = \phi(x_n) - \frac{\phi'(x_n)f(y_n)g(y_n)}{ty'_n[f'(y_n)g(y_n) + f(y_n)g'(y_n)]}$$
(8)

Let

$$\phi(x_n) = z_n = y_n - \frac{f(y_n)}{f'(y_n)}$$
(9)

Which is two-step iterative method having convergence of order six [see eq.(23)]. Differentiate eq.(9) w.r.t "x", we have

$$\phi'(x_n) = z'_n = \frac{f(y_n)f''(y_n)}{f'^2(y_n)}y'_n \tag{10}$$

and from Taylor's series, we can write

$$f(z_n) = f(y_n) + (z_n - y_n)f'(y_n) + \frac{(z_n - y_n)^2}{2}f''(y_n)$$
$$= \frac{f^2(y_n)f''(y_n)}{2f'^2(y_n)}$$
(11)

From (10) and (11), we have

$$\phi'(x_n) = \frac{2f(z_n)}{f(y_n)} y'_n$$
(12)

With the help of (8), (9) and (12), we get

$$x_{n+1} = z_n - \frac{2f(z_n)g(y_n)}{t[g(y_n)f'(y_n) + g'(y_n)f(y_n)]}$$
(13)

Here $t = \frac{6}{3} = 2$, which is according to the above described technique. Then eq.(13) becomes:

$$x_{n+1} = z_n - \frac{f(z_n)g(y_n)}{[g(y_n)f'(y_n) + g'(y_n)f(y_n)]}$$
(14)

Relation(14) is the main and general iterative scheme, which we use to deduce some new iterative algorithms by considering some particular cases of the auxiliary function g.

A. CASE 1

Let $g(x_n) = e^{(-\beta x_n)}$, then $g'(x_n) = -\beta g(x_n)$. Using these values in (14), we obtain the following algorithm.

Algorithm 1: For a given x_0 , compute the approximate solution x_{n+1} by the following iterative schemes:

$$y_n = x_n - \frac{f(x_n)}{f'(x_n)} - \frac{f^2(x_n)f''(x_n)}{2[f'^3(x_n) - f(x_n)f'(x_n)f''(x_n)]}$$
$$z_n = y_n - \frac{f(y_n)}{f'(y_n)}$$
$$z_{n+1} = z_n - \frac{f(z_n)}{[f'(y_n) - \beta f(y_n)]}$$

B. CASE 2

 x_{κ}

Let $g(x_n) = e^{-\beta f(x_n)}$, then $g'(x_n) = -\beta f'(x_n)g(x_n)$. Using these values in (14), we obtain the following algorithm.

Algorithm 2: For a given x_0 , compute the approximate solution x_{n+1} by the following iterative schemes:

$$y_n = x_n - \frac{f(x_n)}{f'(x_n)} - \frac{f^2(x_n)f''(x_n)}{2[f'^3(x_n) - f(x_n)f'(x_n)f''(x_n)]}$$
$$z_n = y_n - \frac{f(y_n)}{f'(y_n)}$$
$$_{n+1} = z_n - \frac{f(z_n)}{[f'(y_n) - \beta f(y_n)f'(y_n)]}$$

C. CASE 3

x

Let $g(x_n) = e^{-\beta x_n^2}$, then $g'(x_n) = -2\beta x_n g(x_n)$. Using these values in (14), we obtain the following algorithm.

Algorithm 3: For a given x_0 , compute the approximate solution x_{n+1} by the following iterative schemes:

$$y_n = x_n - \frac{f(x_n)}{f'(x_n)} - \frac{f^2(x_n)f''(x_n)}{2[f'^3(x_n) - f(x_n)f'(x_n)f''(x_n)]}$$
$$z_n = y_n - \frac{f(y_n)}{f'(y_n)}$$
$$x_{n+1} = z_n - \frac{f(z_n)}{[f'(y_n) - 2\beta y_n f(y_n)]}$$
In all above cases, we take $n = 0, 1, 2, 3, ...$

To get best results in all above defined algorithms, always choose that values of β which makes the denominator non-zero and largest in magnitude.

III. CONVERGENCE ANALYSIS

In this section, we discuss the convergence criteria of the main and general iteration scheme described in relation (14).

Theorem 1: Assume that $\alpha \in I$ be the simple root of the differentiable function $f : I \subset \mathbb{R} \to \mathbb{R}$ on an open interval *I*. If the initial guess x_0 is sufficiently close to α , then the convergence order of the main and general iteration scheme described in relation (14) is at least nine.

Proof: To prove the convergence of the main and general iteration scheme described in relation (14), we assume that α is the simple root of the equation f(x) = 0 and e_n be the error at *n*th iteration, then $e_n = x_n - \alpha$ and by using Taylor series about $x = \alpha$, we have

$$f(x_n) = f'(\alpha)e_n + \frac{1}{2!}f''(\alpha)e_n^2 + \frac{1}{3!}f'''(\alpha)e_n^3 + \frac{1}{4!}f^{(iv)}(\alpha)e_n^4 + \dots + O(e_n^{10})$$

$$f(x_n) = f'(\alpha)[e_n + c_2e_n^2 + c_3e_n^3 + c_4e_n^4 + \dots + O(e_n^{10})]$$
(15)

where

$$c_n = \frac{1}{n!} \frac{f^{(n)}(\alpha)}{f'(\alpha)}, \quad n = 2, 3, 4, \dots$$

$$f'(x_n) = f'(\alpha)[1 + 2c_2e_n + 3c_3e_n^2 + 4c_4e_n^3 + 5c_5e_n^4 + \dots + O(e_n^{10})]$$
(16)

$$f''(x_n) = f'(\alpha)[2c_2 + 6c_3e_n + 12c_4e_n^2 + 20c_5e_n^3 + 30c_6e_n^4 + \dots + O(e_n^{10})]$$
(17)

With the help of (15) - (17), we get

$$y_{n} = \alpha - c_{3}e_{n}^{3} + (c_{2}^{3} - 3c_{4})e_{n}^{4} + \dots + O(e_{n}^{10}) \quad (18)$$

$$f(y_{n}) = f'(\alpha)[(y_{n} - \alpha) + c_{2}(y_{n} - \alpha)^{2} + c_{3}(y_{n} - \alpha)^{3} + c_{4}(y_{n} - \alpha)^{4} + \dots + O(e_{n}^{10})] \quad (19)$$

$$f'(y_{n}) = f'(\alpha)[1 + 2c_{2}(y_{n} - \alpha) + 3c_{3}(y_{n} - \alpha)^{2} + 4c_{4}(y_{n} - \alpha)^{3} + 5c_{5}(y_{n} - \alpha)^{4} + \dots + O(e_{n}^{10})] \quad (20)$$

$$g(y_n) = g(\alpha) + (y_n - \alpha)g'(\alpha) + \frac{(y_n - \alpha)^2}{2!}g''(\alpha) + \frac{(y_n - \alpha)^3}{3!}g'''(\alpha) + \frac{(y_n - \alpha)^4}{4!}g^{(i\nu)}(\alpha) + \dots + O(e_n^{10})$$
(21)

$$g'(y_n) = g'(\alpha) + (y_n - \alpha)g''(\alpha) + \frac{(y_n - \alpha)^2}{2!}g'''(\alpha) + \frac{(y_n - \alpha)^3}{3!}g^{(iv)}(\alpha) + \frac{(y_n - \alpha)^4}{4!}g^{(v)}(\alpha) + \dots + O(e_n^{10})$$
(22)

With the help of (18) - (20), we have

$$z_n = \alpha + c_2 c_3^2 e_n^6 + \dots + O(e_n^{10})$$
(23)
$$f(z_n) = f'(\alpha) [(z_n - \alpha) + c_2 (z_n - \alpha)^2 + c_3 (z_n - \alpha)^3 + c_4 (z_n - \alpha)^4 + \dots + O(e_n^{10})]$$
(24)

Using equations (15) - (24) in general iteration scheme(14), we get the same result as given below

$$x_{n+1} = \alpha - 2c_2^2 c_3^3 e_n^9 + O(e_n^{10})$$

Which implies that

$$e_{n+1} = -2c_2^2 c_3^3 e_n^9 + O(e_n^{10})$$

Which shows that the main and general iteration scheme(14) is of ninth order of convergence and all iterative algorithms deduced from it have also the same order of convergence. \Box

IV. NUMERICAL RESULTS

In this section, we include some non-linear functions to demonstrate the performance of newly proposed iterative algorithms (for $\beta = 1$). We compare these algorithms with Newton's method (NM) [35], Golbabai and Javidi's method (JM) [8], Househölder's method (HHM) [13] and modified Halley's method (MHM) [14]. For this purpose, following test examples have been solved:

$$f_1(x) = (e^{x^2} + x - 20)^{20}, \quad x_0 = 2.00,$$

$$f_2(x) = \pi - 2x \sin \frac{\pi}{x}, \quad x_0 = 2.50,$$

$$f_3(x) = x^3 + \ln x + 0.15 \cos 5x, \quad x_0 = 2.20,$$

$$f_4(x) = x^3 + 4x^2 - 10, \quad x_0 = -0.30,$$

$$f_5(x) = (x - 1)^3 - 1, \quad x_0 = 1.30,$$

$$f_6(x) = x^2 - e^x - 3x + 2, \quad x_0 = 2.50,$$

$$f_7(x) = x^3 - x^2 + 3x \cos x - 1, \quad x_0 = 1.00,$$

$$f_8(x) = 0.986x^3 - 5.181x^2 + 9.067x - 5.289, \quad x_0 = 3.10,$$

$$f_9(x) = \frac{x}{1 - x} - \ln \left[\frac{0.4(1 - x)}{0.4 - 0.5x}\right]^5 + 4.45977, \quad x_0 = 0.78.$$

The last two numerical examples namely; f_8 and f_9 are actually the conversion of Engineering problems to the non-linear equations. The first one is the the van der Waal's equation for interpreting the real and ideal gas behavior [36] that has been converted to the non-linear form after choosing the appropriate values of the parameters. The variable *x* denotes

the volume of the gas which is being under consideration. The second numerical example is related to the fractional transformation in a chemical reactor [31] which we converted to the non-linear problem. In this problem, *x* represents a fractional transformation of a particular species in the chemical reactor problem. For the negative values of *x*, the expression f_9 has no physical meaning and has the bounded region [0, 1] beyond which its derivative vanishes. That's why, we have to choose the initial guess carefully within the bounded region to approximate the real root of f_9 . The numerical results of the above described test examples have been shown in Table 1.

Here, we take the accuracy $\varepsilon = 10^{-15}$ in the following stopping criteria

$$|x_{n+1} - x_n| < \varepsilon \tag{25}$$

Table 1 shows the numerical comparisons of our developed iterative algorithms (for $\beta = 1$), with Newton's method (NM), Golbabai and Javidi's method (JM), Househölder's method (HHM) and modified Halley's method (MHM). The columns represent the number of iterations N, the magnitude |f(x)| of f(x) at the final estimate x_{n+1} , the approximate root x_{n+1} , the difference between two consecutive approximations $|x_{n+1} - x_n|$, the computational order of convergence (COC) which can be approximated using the following formula:

$$COC pprox rac{lnrac{|x_{n+1}-lpha|}{|x_n-lpha|}}{lnrac{|x_n-lpha|}{|x_{n-1}-lpha|}}$$

which was introduced by Weerakoon and Fernando in (2000) [37] and the last column represents CPU time consumption in seconds, taken by different algorithms to approximate the required solution.

From the over all obtained results of the Table 1, we can clearly see that the suggested algorithms approximated the solutions of different test problems with the great accuracy than the estimations gained through the other four iteration schemes. These results also show that the suggested algorithms consumed less number of iteration to meet the stopping criteria (25) with high computational order of convergence as compared to the other comparable methods. It should also be noted that the presented algorithms take less CPU time in seconds to achieve the stopping criteria for approximating different test problems. In short, we can say that the suggested algorithms perform faster and are more accurate to the exact solutions in comparison with the other compared methods.

We solved all test problems and calculated the CPU time consumption in seconds with the aid of the computer program Maple 13.

Table 2 shows the comparison of the number of iterations required for different iterative methods with our developed iterative algorithms (for $\beta = 1$) to approximate the root of the given non-linear function for the stopping criteria (25) with the accuracy $\varepsilon = 10^{-100}$. The columns represent the number of iterations for different functions along with initial guess x_0 .

TABLE 1. Comparison of NM, JM, HHM, MHM, Algorithm 1, Algorithm 2 & Algorithm 3.

Method	N	$ f(x_{n+1}) $	x_{n+1}	$\sigma = x_{n+1} - x_n $	COC	CPU Time						
$f_1(x), x_0 = 2$	2.00	y (* <i>n</i> 1)	10 1	- -// 1//								
NM	103	4.558695e - 15	1.70789420884892364723755253334891433	1.573265e - 04	2	3.826						
JM	5	1.338286e - 17	1.70261392757820937250558331788736293	2.466806e - 03	3	3.857						
HHM	69	3.834935e - 15	1.70786846328574875872769826250064092	2.359586e - 04	3	3.951						
MHM	16	1.090746e - 15	1.70768770322286908220270756250556966	1.030725e - 03	5	3.997						
Algorithm 1	4	1.101429e - 15	1.70205037973761406191870783210162534	3.493353e - 03	9	2.093						
Algorithm 2	5	1.143178e - 20	1.70329204883759619555762115162676828	1.981392e - 03	9	2.171						
Algorithm 3	4	9.834412e - 16	1 70206648426476011891055532332698266	3473872e - 03	ğ	2 234						
$f_0(r)$ $r_0 = 2$	2 50	0.0011120 10	1.1020001012011001100100000200200200	0.1100120 00		2.201						
$\int g(x), x_0 = 2$	6	2.2281766 16	1.65740024025800605203063730821372128	1.041405a 08		4 1 2 2						
INI	4	2.228170e = 10 6.646056a 20	1.65740024025800612203003730821372128	1.041405e = 0.00 4.082780c = 10	2	4.155						
	4 5	5.500552 21	1.65740024025800612279373807237323309	4.065769e = 10 6.845206c 11	5 9	4.200						
MIM	9	0.03000000 - 01	1.65740024025800012573575807255202725	0.045200e - 11 8.174677a - 07	5	4.200						
	3	2.339921e - 31 1.977244 - 41	1.05740024025000012579575807255177181	0.174077e - 07 2.524921 = 05	0	4.040						
Algorium 1	4	1.277544e - 41	1.05740024025800012579575807255184717	5.554621e - 05	9	2.441						
Algorithm 2	2	1.079481e - 43	1.05740024025800012379373807235184717	2.249937e - 05	9	2.519						
Algorithm 3	2	2.671200e - 38	1.65740024025800612379373867235184717	7.654916e - 05	9	2.613						
$f_3(x), x_0 = 2.20$												
NM	6	6.072231e - 15	0.74677076476086748037343464839898526	4.576382e - 08	2	2.976						
JM	4	4.879664e - 32	0.74677076476086571044495099175143523	9.555808e - 11	3	3.038						
HHM	5	3.305553e - 39	0.74677076476086571044495099175142101	8.803620e - 14	3	3.086						
MHM	4	8.986943e - 53	0.74677076476086571044495099175142101	6.842616e - 11	5	3.132						
Algorithm 1	3	1.038543e - 114	0.74677076476086571044495099175142101	7.899816e - 13	9	2.618						
Algorithm 2	3	2.079929e - 65	0.74677076476086571044495099175142101	2.142561e - 07	9	2.696						
Algorithm 3	3	3.423949e - 64	0.74677076476086571044495099175142101	3.726798e - 07	9	2.760						
$f_{4}(x), x_{0} = -$	-0.30											
NM	54	8.127500e - 30	1 36523001341409684576080682898215825	1.001963e - 15	2	3 343						
IM	62	6.526278e - 28	1 36523001341409684576080682894214497	8.674049e - 10	3	3 375						
ннм	9	2.386831e - 45	1 36523001341409684576080682898166608	7.007012e - 16	3	3 491						
MHM	22	2.3000010 + 30 2.430110e - 36	1.36523001341409684576080682898166608	9.248288e = 08	5	3 459						
Algorithm 1	22	5.100700e - 86	1.36523001341409684576080682898166608	1.044771e - 00	0	2 746						
Algorithm 2	3	2.720781c = 30	1.36523001341409064576080682898166608	1.0447710 - 03 2.774524a - 08	9	2.740						
Algorithm 2	3	2.720761e - 70	1.30323001341409084570080082898100008	2.774524e - 08 1.246601 a 00	9	2.010						
Algorium 5	3	4.380012e - 83	1.30523001341409084570080082898100008	1.340001e - 09	9	2.871						
$f_5(x), x_0 = 1$	1.30	1 0 0 0 0 0 1 0 0		- 10001 - 10								
NM	9	1.373004e - 19	2.000000000000000004576679330702437	2.139317e - 10	2	2.928						
JM	4	1.376914e - 16	2.0000000000000004589712889833510622	5.163768e - 06	3	2.959						
HHM	27	4.799079e - 39	2.000000000000000000000000000000000000	9.864217e - 14	3	2.989						
MHM	4	3.402977e - 44	2.000000000000000000000000000000000000	1.867040e - 09	5	3.020						
Algorithm 1	3	7.330764e - 16	1.9999999999999999975564120188401069843	2.730810e - 02	9	2.593						
Algorithm 2	3	2.233894e - 29	1.9999999999999999999999999999999999255369	8.360388e - 04	9	2.656						
Algorithm 3	3	1.124811e - 26	1.999999999999999999999999999999625063075	1.544171e - 03	9	2.718						
$f_6(x), x_0 = 2$	2.50											
NM	5	6.456953e - 15	0.25753028543985905166563433207135276	1.352209e - 07	2	3.532						
JM	7	2.540428e - 17	0.25753028543986076717844212003069118	4.902361e - 06	3	3.579						
HHM	4	1.054587e - 23	0.25753028543986076045536451404227079	4.130791e - 08	3	3.611						
MHM	4	2.646153e - 74	0.25753028543986076045536730493724178	6.227672e - 15	5	3.642						
Algorithm 1	2	9.040960e - 20	0.25753028543986076047929360308378555	2.202739e - 02	9	2.345						
Algorithm 2	2	3.293944e - 24	0 25753028543986076045536817665768612	6.212211e - 03	ğ	2.376						
Algorithm 3	$\frac{1}{2}$	1.232057e - 18	0 25753028543986076012931161649727364	3.079645e - 02	ğ	2.010 2.408						
$f_{\pi}(x) = 1$	1 00	1.2020010 10	0.20100020010000012001101010121001	0.0100100 02		2.100						
$\int f(x), x_0 = 1$	10	5.0548446 17	0.30532362208631516207600501688665748	5 772225 00	2	9 1 4 5						
ININI	7	$1.714115 \circ 27$	0.39332302298031310297000391088003748	9.113225e = 09 9.222850a = 12	2	2 200						
	- 1 94	6 8006232 - 31	0.335332302230031310037014270043640199	0.220000 = 10 0.075050c = 10	ა ვ	2.200						
MIM	24 C	0.899023e - 29 6.214672 - 50	0.39332302298031318837014278847307198	2.975250e = 10 1.990560a 10	5	3.270						
	4	0.314072e = 30	0.39332302290031310037014270043040199	1.000509e - 10 4.177405 - 12	0	0.017						
Algorithm 1	4	5.508815e - 114	0.39532302298031518837014278843840199	4.177495e - 13 7.021172 00	9	2.484						
Algorithm 2	4	1.737209e - 75	0.39532362298631518837614278843840199	7.631173e - 09	9	2.515						
Algorithm 3	4	3.957250e - 50	0.39532362298631518837614278843840199	5.250956e - 06	9	2.594						
$f_8(x), x_0 = 3$	3.10											
NM	10	1.815641e - 25	1.92984624284786221848752952852582649	5.866916e - 13	2	3.353						
JM	5	1.499299e - 36	1.92984624284786221848752742786545647	1.149928e - 12	3	3.392						
HHM	7	8.608929e - 34	1.92984624284786221848752742786546645	5.404901e - 12	3	3.438						
MHM	4	4.590068e - 47	1.92984624284786221848752742786545649	2.421897e - 10	5	3.454						
Algorithm 1	3	3.796053e - 26	1.92984624284786221848752698866975862	5.481798e - 04	9	2.402						
Algorithm 2	3	2.802586e - 63	1.92984624284786221848752742786545649	4.053280e - 08	9	2.465						
Algorithm 3	3	4.417216e - 15	1.92984624284781111218539650390823209	1.025404e - 02	9	2.527						
$f_9(x), x_0 = 0$	0.78											
NM	6	2.214298e - 27	0.75739624625375387945964129795690718	1.323123e - 15	2	3.065						
JM	5	1.140552e - 41	0.75739624625375387945964129792914529	8.139895e - 16	3	3.128						
HHM	4	6.203623e - 36	0.75739624625375387945964129792914529	6.889797e - 14	3	3.175						
MHM	3	1.203934e - 41	0.75739624625375387945964129792914529	2.727415e - 10	5	3 206						
Algorithm 1	2	7.858948e = 33	0 75739624625375387945964129792914510	1.303132e = 05	a	2.621						
Algorithm 2	2	7.0000400 = 00 7.254301 = 28	0.75739624625375387045064129792914019	4.020932e = 05	a a	2.668						
Algorithm 3	2	6.722228e = 33	0.75739624625375387045064120702014521	1.0200000 = 00 1.283116e = 05	à	2.600						
1 rusonum J	4	J.1222200 00	55.500±10±0510±050±1±070±0±1	1.2001100 00	0							

Method	$f_1(x)$	$f_2(x)$	$f_3(x)$	$f_4(x)$	$f_5(x)$	$f_6(x)$	$f_7(x)$	$f_8(x)$	$f_9(x)$
	$x_0 = 2.00$	$x_0 = 2.50$	$x_0 = 2.20$	$x_0 = -0.30$	$x_0 = 1.30$	$x_0 = 2.50$	$x_0 = 1.00$	$x_0 = 3.1$	$x_0 = 0.78$
NM	294	07	09	56	12	08	13	13	08
JM	18	06	06	64	06	09	08	06	06
HHM	197	06	06	10	28	06	26	09	05
MHM	47	04	05	23	05	05	07	05	04
Algorithm 1	16	03	03	04	04	03	04	04	03
Algorithm 2	17	03	04	04	04	03	05	04	03
Algorithm 3	16	03	04	04	04	03	05	04	03

TABLE 2. Comparison of number of iterations required for different iterative methods for $\varepsilon = 10^{-100}$.

The numerical results of Table 2 again certified the fast and best performance of the proposed algorithms in terms of number of iterations even though the accuracy has been raised to 10^{-100} . All calculations have been carried out using the computer program Maple 13.

V. GRAPHICAL COMPARISON BY MEANS OF POLYNOMIOGRAPHS

The problems related to finding the roots of polynomials have played a vital role in engineering and mathematical sciences. It is one of the oldest and most deeply studied mathematical problems as one can study the history of Mathematics that the ancient Greeks and Sumerians considered such practical problems in 3000 B.C, that can now be stated as a root-finding problems using modern mathematical language. In the seventeenth century, Newton suggested an algorithm for approximating roots of polynomials. After that Cayley [5] studied the chaotic and strange behavior while applying Newton's method on cubic polynomial $x^3 - 1$ in complex plane. The problem arisen by Calay was explained by Julia in 1919. The Julia sets bought many new discoveries i.e., Mandelbrot set and Fractal in 1970 [24]. The term "Polynomiography" was first introduced in 2005 by Kalantari [16], [17]. He defined polynomiography as the combination of both science and art related to visualization of the root-finding process for a polynomial in complex plane. The individual image produced as a result of polynomiography is thus called a "Polynomiograph". The polynomiographs can be generated by any iterative scheme i.e., Newton's method, Halley's method etc. In the last few years, many researchers worked on polynomiography and generated new and nice looking images through different algorithms. Soleimani et al. [33], suggested some new iterative methods free from derivatives and then presented their fractal behaviors by means of their basins of attraction. In [20], the authors generated beautiful polynomiographs using Ishikawa and Mann iterations that were quite new and looked aesthetically pleasing comparing to the ones from standard Picard iteration. In 2016, the authors modified the Abbasbandy's method and then presented polynomiographs through the modified method [19]. Gdawiec [7] in 2017, used three different approaches, i.e., affine and s-convex combination, the use of iteration processes from fixed point theory and multi-step polynomiography and obtained new and diverse fractal patterns that have many applications in textile and ceramics patterns. In [23], the authors presented some new graphical objects obtained by the use of escape time algorithm and the derived criteria. They presented graphical examples by means of Jungck-CR iteration process with s-convexity. In 2019, Kalantari and Hans Lee [18] introduced new ways of creating mathematical art through a novel Newton-Ellipsoid method for solving polynomial equations. The nature of polynomiographs under Newton-Ellipsoid seems to be very different from the other images, which opens the possibility of generating novel artistic images.

To generate a polynomiograph via computer program, we have to select an initial rectangle R containing the roots of the polynomial. Then for each point z_0 in the region, we run an iterative method, and then color the point corresponding to z_0 is depended upon the approximate convergence of the truncated orbit to a root, or lack thereof. The resolution of the image depends on our discretization of the rectangle R. For example, discretizing R into a 2000 by 2000 grid yields a high-resolution image. Polynomiography has vast applications in many fields of science and art.

According to Fundamental Theorem of Algebra, any complex polynomial with complex coefficients $\{a_n, a_{n-1}, \ldots, a_1, a_0\}$:

$$p(z) = a_n z^n + a_{n-1} z^{n-1} + \ldots + a_1 z + a_0$$
(26)

or by its zeros (roots) $\{r_1, r_2, ..., r_{n-1}, r_n\}$:

$$p(z) = (z - r_1)(z - r_2) \dots (z - r_n)$$
(27)

of degree n has n roots (zeros) which may or may not be distinct. The degree of polynomial describes the number of basins of attraction and localization of basins can be controlled by placing roots on the complex plane manually.

Usually, the polynomiographs are colored and their coloring depends upon the number of iterations needed to approximate the roots of some polynomial with a given accuracy and a chosen iterative scheme. The general and base algorithm for the generation of polynomiograph is presented in the following Algorithm 4.

In the Algorithm 4, the convergence test $(z_n + 1, z_n, \epsilon)$ returns TRUE if the applied method has converged to the root, and FALSE otherwise. The most common and widely used

Algorithm 4 Polynomiograph's Generation

Input: $p \in \mathbb{C}$ — polynomial, $A \subset \mathbb{C}$ — area, k maximum number of iterations, I — iteration method, β — parameter for the iteration, ϵ accuracy, colormap $[0 \dots C - 1]$ — colormap with C colors. **Output:** Polynomiograph for the complex polynomial pin area A. **for** $z_0 \in A$ **do** i = 0**while** $i \le k$ **do** $|z_{n+1} = I(z_n)$ **if** $|z_{n+1} - z_n| < \epsilon$ **then** \lfloor **break** |i = i + 1color z_0 by means of colormap.

convergence test has the following standard form:

$$|z_{n+1} - z_n| < \varepsilon, \tag{28}$$

where z_{n+1} and z_n are two consecutive points in an iteration process and $\varepsilon > 0$ is a given accuracy. In this article we also use the stopping criteria (28).

Using newly developed iterative algorithms (for $\beta = 1$), we present polynomiographs of the following complex polynomials and comparing them with other well-known existing iterative algorithms.

$$p_1(z) = z^3 - 1$$
, $p_2(z) = (z^3 - 1)^2$

The colormap used for the coloring of iterations in the generation of polynomiographs is presented in the following figure:



FIGURE 1. The colormap used for generating polynomiographs.

Example 1: Polynomiographs for the Polynomial $p_1(z)$ *by Means of Different Iterative Methods*



FIGURE 2. Polynomiograph for $p_1(z)$ using (NM).



FIGURE 3. Polynomiograph for $p_1(z)$ using (JM).



FIGURE 4. Polynomiograph for $p_1(z)$ using (HHM).



FIGURE 5. Polynomiograph for $p_1(z)$ using (MHM).



FIGURE 6. Polynomiograph for $p_1(z)$ using Algorithm 1.



FIGURE 7. Polynomiograph for $p_1(z)$ using Algorithm 2.



FIGURE 8. Polynomiograph for $p_1(z)$ using Algorithm 3.

Example 2: Polynomiographs for the Polynomial $p_2(z)$ *by Means of Different Iterative Methods*



FIGURE 9. Polynomiograph for $p_2(z)$ using (NM).



FIGURE 10. Polynomiograph for $p_2(z)$ using (JM).



FIGURE 11. Polynomiograph for $p_2(z)$ using (HHM).



FIGURE 12. Polynomiograph for $p_2(z)$ using (MHM).



FIGURE 13. Polynomiograph for $p_2(z)$ using Algorithm 1.



FIGURE 14. Polynomiograph for $p_2(z)$ using Algorithm 2.



FIGURE 15. Polynomiograph for $p_2(z)$ using Algorithm 3.

In examples (1 - 2), which include figures (2 - 15), Polynomiographs of different complex polynomials for Newton's method (NM), Golbabai and Javidi's method (JM), Househölder's method (HHM), modified Halley's method (MHM) and our developed iterative algorithms (for $\beta = 1$) have been shown. When we look at the generated images, we can read two important characteristics. The first one is the speed of convergence of the iterative scheme, i.e., the color of each point gives us information on how many iterations were performed by the iterative scheme to reach the root. The second characteristic is the dynamics of the adopted iterative scheme. Low dynamics are in those areas where the variation of colors is small, whereas in areas with a large variation of colors the dynamics are much high.

The black color in images locates those places where the solution cannot be achieved for the given number of iterations and one can observe that the images generated through proposed iterative algorithms scarcely find such places which prove their best performance. The areas of the same colors in above figures indicate the same number of iterations required to determine the solution and they look similar to the contour lines on the map.

All these figures have been generated using the computer program Mathematica 10.0 by taking $\varepsilon = 0.01$ and k = 15 where ε shows the accuracy of the given root and k represents the upper bound of the number of iterations.

VI. CONCLUDING REMARKS

Based on variational iteration technique, some new iterative algorithms for the solution of one dimensional non-linear equations have been established, having ninth order of convergence. By using some test examples, the performance and efficiency of the proposed iterative algorithms have been analyzed. Tables 1–2 show the best performance of the proposed iterative algorithms as compare to other well-known existing iterative algorithms in terms of accuracy, speed, time, number of iterations and computational order of convergence. We have also presented the graphical comparison of the proposed iterative algorithms with other well-known iterative algorithms by means of polynomiographs of some complex polynomials which describe the fractal behavior and dynamical aspects of the proposed iterative algorithms. The variational iteration technique can be applied to derive a broad range of new algorithms for solving one dimensional non-linear equations.

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