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*Article*

# A New Wave of Hybrid Algorithms for Roots of Transcendental and Non-Linear Equations

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**Abstract**— Finding the roots of an equation is a fundamental problem in diverse fields. Optimization problem leads to non-linear equations for calculating the roots of the equation efficiently. Numerical iterative techniques are frequently applied when analytic solution is not available. Hybrid techniques to find roots of an equation is new to this fundamental problem. Recently, some attempts emerged to design hybrid algorithms for efficient solutions. A new heuristic algorithm is designed which is an intuitive approach to hybridization. In this paper, we implement two new algorithms: (1) a new BTsection algorithm blending the Bisection and Trisection algorithm, (2) a hybrid algorithm, that promises to be more efficient than the existing hybrid algorithms. The new hybrid algorithm is a blend of BTsection algorithm and Regula Falsi algorithm. The implementation results validate that the new algorithm, Hybrid4, surpasses the efficiency of existing hybrid algorithms Hybrid1, Hybrid2, and Hybrid3 algorithms. This paper contributes an essential hybrid algorithm to the repertoire of hybrid root finding algorithms.

**Keywords:** regula false; newton-raphson; bisection; trisection; BTsection; hybrid algorithm

## 1. Introduction

Numerical iterative techniques are applied to find roots of an equation when analytic solution is not available [1–4]. Hybrid algorithm is a new concept to iterative solutions. Envisioning hybrid techniques to find roots of an equation is an inspiring technique to solve the fundamental problem in diverse fields. Recently, there have been some hybrid algorithms designing efficient solutions. Here we present a new hybrid algorithm which is a more intuitive and heuristic approach to hybridization.

Mostly optimization problems lead to solving non-linear equations for optimizing calculation of the value of a parameter, that is the root of the equation. We design and implement a new algorithm Hybrid4 that is more efficient hybrid of BTsection and False Position methods. The implementation results validate that the new algorithm surpasses the existing hybrid algorithms rooted on standalone Bisection and Trisection algorithms in conjunction with False Position and Newton Raphson algorithms. We designed and implemented an essential hybrid algorithm to the root finding algorithms.

First, why do we need another root finding algorithm? Even though the classical methods have been developed and used for decades, enhancements are progressively made to improve the performance of these methods [5]. There are several factors in determining the efficiency of an algorithm: accuracy, CPU time efficiency, and generality that it works most of the time on all functions and domain intervals as expected. Accuracy may be measured in terms of number of decimal digits. For iterative methods, efficiency could mean the number of iterations to arrive at a solution coupled with CPU times, and computational complexity of iteration etc.

Finding the roots of an equation is a fundamental problem in diverse fields in physical and social sciences including Computer Science, Engineering (Biological, Civil, Electrical, Mechanical), and Social Sciences (Psychology, Economics, Businesses, Stock Analysis) etc. They look for the optimal solution to the recurring non-linear problems. The problems such as minimization, Target Shooting, Orbital Motion, Plenary Motion, Social Sciences, Financial Market Stock prediction analysis etc, lend themselves to finding roots of non-linear functional equations [6,7]. There is thorough study by Sapna and Mohan in the financial sector away from mathematics [8].

Thus, a root-finding algorithm implies solving an equation defined by a function that may be linear, non-linear or transcendental. Some root-finding algorithms do not guarantee that they will find a root. If such an algorithm does not find a root, that does not mean that a root does not exist. No single classical method outperforms another method on all functions [9]. A method may outperform other methods on one input and may produce inferior results on another input. Different methods have their own strengths/weaknesses.

There are classical root-finding algorithms: Bisection, False Position, Newton-Raphson, Secant, methods for finding the roots of an equation  $f(x) = 0$ . Every textbook on Numerical Techniques has details of these methods [1–3]. In the presence of classical methods, enhancements are progressively made to improve the performance of these methods [10–12].

Recently papers are making a headway on seeking better performing methods. In response, the better algorithms that are hybrid of classical methods Bisection, Trisection, with False Position, Newton Raphson methods been developed, namely Hybrid1 [10], Hybrid2 [11], and Hybrid3 [12]. Inspired by these three algorithms, we have crafted a new proficient algorithm, Hybrid4, that is a blend of BTsection and Regula Falsi algorithms. This blended algorithm is comparatively better than Hybrid1, Hybrid2, and Hybrid3 with respect to computational efficiency, solution accuracy (less error) and iteration count required to terminate within the specified error tolerance. This algorithm, Hybrid4, further optimizes these algorithms, first by blending Bisection and Trisection algorithms, then the resulting algorithm is blended with False Position method to evolve into Hybrid 4 **algorithm hybridizing** one step further by eliminating some of the computing time and increasing the efficiency of the algorithm.

The paper is organized as follows. Section II is brief description of classical methods Bisection, Regula Falsi, Newton-Raphson, Secant; their strengths and pitfalls. In addition, Trisection and new BTsection algorithms are included. Section III describes hybrid algorithms and a new algorithm, Hybrid4, that blends BTsection and False Position algorithms. Section IV presents experimental results simulating the performance of new algorithm Hybrid4 and validating it by comparing its performance with the previous hybrid algorithms. Section V is conclusion.

## 2. Related Background

The classical algorithms Bisection, False Position, Newton-Raphson, Secant methods are readily found in any text book in detail and iterated in most articles [1–3]. The classical algorithms, that are ubiquitous are available everywhere, These are iteration based methods. BIS-Bisection method compute a new approximation at the mid point of the interval, FP-FalsPosition obtains a new approximation by the secant line joining the endpoints of the function. NR- Newton Raphson derives new approximating by the tangent line at the iterated point on the function. All these algorithms are iterative for calculations of better approximations.

We will discuss the recent algorithms Trisection and BTsection. As Bisection method divides an interval into two equal parts, the Trisection algorithm subdivides the interval in three equal parts to get better estimate at the next iteration. BTsection is new algorithm that subdivides the interval adaptively into three parts: first Bisection into two, then the promising one half is adaptively trisected into parts to get better approximation, see algorithm in sectio3 [14].

Bisection algorithm states that “If  $f$  is continuous on a closed interval  $[a, b]$  and  $f$  is of opposite signs at the end points, i.e.  $f(a)f(b) < 0$ , then there is a root in the open interval  $(a, b)$ .” Standard algorithm implementation expects that  $f(a)f(b) < 0$  to begin with. Because,  $f(a)f(b)=0$  in the following, it will fail to proceed on this root finding problem: “ $x^2-x-2=0$  on interval  $[2,4]$ , or  $[0, 2]$ , or  $[-2,-1]$  or  $[-3,-1]$ . A robust bisection algorithm first confirms that  $f(a)f(b) < 0$  before proceeding to iterate. For example, for  $f(x) = (x-1)*(x-2)*(x-3)$  on the interval  $[1,3]$ , the original version of algorithm will not start, but will succeed on  $[a,b]$  only when  $a, b$  are not 1, 2, 3.

In order to be successful on any interval including special cases, we reconsider the wording in this theorem and reformulate this theorem to include closed interval all the way. “If  $f$  is continuous on a closed interval  $[a, b]$  and  $f$  is not of same sign at the end points, i.e.  $f(a)f(b) \leq 0$ , then there is a root in the closed interval  $[a, b]$ .” The Bisection algorithm is generalized from “ $f(a)f(b) < 0$ ” to “ $f(a)f(b)$

$\leq 0$ " to include the interval boundary also in definition. For example, the equation  $x^2-x-2=0$  may show up in some applications with an interval like  $[2,4]$ ,  $[0, 2]$ . Though, it may look simple but if  $f(a)f(b) = 0$ , then a or b must be a root, but we still want to know which one of a and b?

For approximate solutions using iterative methods, we have some idea about where the root may be. We should have initial guess as close to the location of root as far as possible. That is, we provide a start point or a guess or initial bracketing interval to the algorithm to iterate in search for the true approximate value of actual root, within some acceptable tolerance.

Figure 1. Modified version of Bisection on  $[1,3]$ , succeeds in determining a root  $x=2$  of  $(x-1)(x-2)(x-3)=0$ , in one iteration.

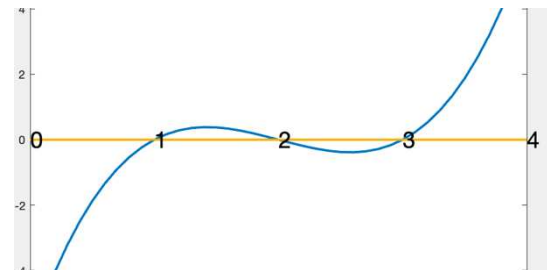


Table 1      Function $f(x) = (x-1)(x-2)(x-3)$					
Max Iterations = 40			Error tolerance = 0.000000000001		
Method	Interval	Root	Error	Iterations	CPU Time
Bisection	[1,3]	2.000000000000	0.000000000000	1	0.003404917000

Since the Bisection, Regula Falsi, and Newton Raphson methods are readily available in the literature, their standalone derivations are skipped in this background section.

To enhance the performance of Bisection method, Bader et.al designed a Trisection method that supersedes Bisection method in finding an approximate root. Trisection method reduces the number of iterations performed, computation time, and error of approximation at a small cost on number of function computations. It inspired us to combine Bisection and Trisection algorithms to a BTsection algorithm at no additional compute cost and is at par with Trisection algorithm in computation CPU time, and error in approximate root, but BTsection requires fewer number of iterations, see examples in this section. These algorithms are blended with False Position and Newton-Raphson methods to construct hybrid algorithms. The effectiveness and efficacy of root approximation is measured by number of iterations in root calculation and the approximation accuracy of the root at the termination of algorithms. The heuristic metrics for measuring error, the number iterations and stopping criteria are elaborated here first.

## Heuristics for comparing algorithms

### 2.1. Metric for Approximation Error Measurement

There are various ways to measure the error in the approximate root of an equations,  $f(x) = 0$ , at successive iterations to continue to a relatively more accurate approximation to the actual root. At iteration  $n$ , to determine  $r_n$  for which  $f(r_n) \cong 0$ , we proceed to analyze as follows.

The iterated root approximation error can be

$$\text{RelativeRootError} = \left| \frac{r_n - r_{n-1}}{r_n} \right|$$

or

$$\text{AbsoluteRootError} = |r_n - r_{n-1}|$$

Since a root can be zero, in order to avoid division by small numbers, it is preferable to use absolute error  $|r_n - r_{n-1}|$  for convergence test. Another reason for this is that if  $r_n = 2^{-n}$ , then  $\left| \frac{r_n - r_{n-1}}{r_n} \right|$  is always 1, it can never be less than 1, so the root-error tolerance test cannot be satisfied effectively, this test does not work. Since function value is expected to be zero at the root, an alternate cognitively

more appealing error test is to use  $f(r_n)$  for error consideration instead of  $r_n$ . There are three versions for this concept, for comparison criteria, they are

$$\text{RelativeValueError} = \left| \frac{f(r_n) - f(r_{n-1})}{f(r_n)} \right|$$

or

$$\text{AbsoluteVlaueError} = |f(r_n) - f(r_{n-1})|$$

or

$$\text{TrueValueError} = |f(r_n)|$$

Since  $f(r_n)$  is to be close to zero near the root, in order to avoid divide by small numbers, we discard using  $\left| \frac{f(r_n) - f(r_{n-1})}{f(r_n)} \right|$ . Further, since  $|f(r_n) - f(r_{n-1})|$  can be close to zero without  $|f(r_n)|$  being close to zero, we discard using  $|f(r_n) - f(r_{n-1})|$  also in favor of using only  $|f(r_n)|$ , trueValue error. For example,  $f(r_n) = (n-1)/n$  is such an example. We avoid using the first two criteria for this reason and exploit the last one,  $|f(r_n)|$ .

Now we are left with two options  $|r_n - r_{n-1}|$  and  $|f(r_n)|$  to consider for error analysis. Again,  $r_n$  and  $r_{n-1}$  can be closer to each other without  $f(r_n)$  being closer to zero. For example  $r_n = 1 + 1/n$ ,  $f(r_n) = r_n$ . Between the options  $|r_n - r_{n-1}|$  and  $|f(r_n)|$ , we find that  $|f(r_n)|$  is the only reliable metric for analyzing the approximation error. Hence, we use,  $|f(r_n)|$ , as the criteria for for comparing with tolerance error analysis for all the methods uniformly.

## 2.2. Metric for Iterations Stopping Criteria, Halting Condition

Stopping criteria plays a major role in simulations. The iteration termination (stopping) criteria for False Position method is different from Bisection method. Tradeoff between accuracy and efficiency is accuracy of the outcome. In order to obtain  $n$  significant digit accuracy, let  $\epsilon_s$  be stopping error and let  $\epsilon_a$  be the approximation error at any iteration. If  $\epsilon_a < \epsilon_s$ , the algorithm stops iterations. With  $\epsilon_s = 5/10^{n-1}$ , we have  $n$  significant digit accuracy in the outcome [1]. The bisection algorithm is trivial [2], Trisection and BTsection algorithms are described in Section III.

Here we describe two enhancements to bisection algorithm. Trisection algorithm and BTsection [section III] algorithm, each algorithm has five comparison tests and reference to four function values. The BTsection algorithm uses the same amount of computation resources as Trisection algorithm.

But the number of iterations  $bn$  (Bisection),  $tn$  (Trisection),  $btn$  (BTsection) required by the Bisection, Trisection, and BTsection algorithms on  $[a, b]$  with stopping tolerance  $tol$  are

$$\frac{b-a}{2^{bn}} < tol, \quad \frac{b-a}{3^{tn}} < tol, \quad \frac{b-a}{6^{bn}} < tol,$$

$$bn = \log\{(b-a)/tol\} \quad tn = (0.63) \log\{(b-a)/tol\} \quad btn = (0.39) \log\{(b-a)/tol\}.$$

Trisection algorithm takes 37% fewer iterations than Bisection algorithm to converge within the desired tolerance.

BTsection algorithm takes 24% fewer iterations than Trisection algorithm to converge to the root within the desired tolerance. In addition, as observed below in both Trisection and BTsection algorithms, in each iteration, there is no change in the computation time: five comparison tests and references to four function values.

## 2.3. Criteria for Performance

This is an experimental science. The methods are numerical and iterative, not analytic solutions. A method may perform well in one case, and fail miserably in another case, see examples below. No one method outperforms all other methods all the time on all the intervals of definition [9], [15]. We determine that the proposed algorithm performs better than the existing related algorithms.

There is no easy way to declare that an algorithm superior to other exiting algorithms. There are several factors that must be taken into consideration for distinguishing between the competing algorithms for a root finding problem.

Smart way is to represent new algorithm as to how the proposed algorithm differs from the other algorithms. To emphasize the new idea, it is also desirable that the new solution be applicable to a



larger spectrum, devoid of incomplete domain. A simple customized example does not give a quantitative or qualitative view. We need to check the performance and applicability of the proposed approach in a broader scope. To comprehend this further, let us take the following example.

For example, the following problem shows that Trisection algorithm outperforms the other algorithms, see Table2 [Bader], [12], it gets the correct root in one iteration whereas other cited algorithms take more iterative steps, and more CPU time as well. One can infer from this example that the Trisection algorithm 2 and 3 work well and are superior to other algorithms. Of course, it is when input test function is  $x^2 - x - 2$  and interval is [1,4].

**Table 2.** In this example, algorithm termination upper bound is 40 for iterations and upto 12 decimal digits for approximation in the acceptable root.

Function $f(x) = x^2 - x - 2$ ,					
Method	Interval	Root	Error	Iterations	CPU Time
Hybrid1: Bisection-FalsePos	[1,4]	2.000000000000	0.000000000000	2	0.020951834000
Hybrid2: Trisection-FalsePos	[1,4]	2.000000000000	0.000000000000	1	0.006400042000
Hybrid3: Trisection-NewtonR	[1,4]	2.000000000000	0.000000000000	1	0.010491333000
Hybrid4: BTsection-FalsePos	[1,4]	2.000000000000	0.000000000000	7	0.007028250000

Now, consider the same function, and same four test algorithms. The initial input interval [1, 4] is changed to [1,3] or [1,5], the Table3 indicates that the inference from Table 2 does not hold good, the tables are turned, see Tables 3, 4. Here *algorithm4* solves the same problem in one iteration and uses the lowest amount of CPU time. Again, from this example, it is unfair to declare that the algorithm 4 outperforms Algorithms 1, 2, 3. Considering Table 4, it gives a clue to inference. Even though algorithm4 doesn't **terminate in one iteration**, it does terminate before the other algorithms.

Table 3      Function $f(x) = x^2 - x - 2$ ,					
Method	Interval	Root	Error	Iterations	CPU Time
Hybrid1: Bisection-FalsePos	[1,3]	2.000000000000	0.000000000000	1	0.003802500000
Hybrid2: Trisection-FalsePos	[1,3]	2.000000000000	0.000000000000	7	0.006985792000
Hybrid3: Trisection-NewtonR	[1,3]	2.000000000000	0.000000000000	13	0.026495625000
Hybrid4: BTsection-FalsePos	[1,3]	2.000000000000	0.000000000000	1	0.003495625000

Table 4      Function $f(x) = x^2 - x - 2$ ,					
Method	Interval	Root	Error	Iterations	CPU Time
Hybrid1: Bisection-FalsePos	[1,5]	2.000000000000	0.000000000000	6	0.028778667000
Hybrid2: Trisection-FalsePos	[1,5]	2.000000000000	0.000000000000	7	0.014464750000
Hybrid3: Trisection-NewtonR	[1,5]	2.000000000000	0.000000000000	14	0.031316708000
Hybrid4: BTsection-FalsePos	[1,5]	2.000000000000	0.000000000000	1	0.010769959000

We have seen that in one case algorithms 2, 3 and in the other two cases algorithms 1, 4 come out ahead because they solve the problem in one iteration and use the lowest CPU time. The dilemma is to determine which algorithm will work satisfactorily on the majority of cases, if not, all the cases. We have to consider not only this function but other functions as well; not one interval, but other related intervals as well. Let us first see the next examples in Tables 5,6,7,8,9 where none of these four algorithms succeed in finding the approximation in one iteration. We used different functions and different intervals for these examples in Tables 5,6,7,8,9. In all these examples, the algorithm 4 has the lowest number of iterations and lowest amount of CPU time. We have tested on number other examples as well to validate the same phenomena. The following examples use different functions and different initial intervals [1, 6], [1,7], [1, 8], even [0,1] as well to make a intuitive heuristic that

algorithm 4 may be more efficient than the others. In all the ensuing examples in this paper, algorithm halting upperbound is 40 for iterations and upto 12 decimal digits for approximation in the acceptable root.

Table 5. Function $f(x) = 0.986x^3 - 5.181x^2 + 9.067x - 5.289$					
Method	Interval	Root	Error	Iterations	CPU Time
Hybrid1: Bisection-FalsePos	[1,6]	1.929846242840	0.000000000001	10	0.005306959000
Hybrid2: Trisection-FalsePos	[1,6]	1.929846242844	0.000000000000	9	0.004571250000
Hybrid3: Trisection-NewtonR	[1,6]	1.929846242850	0.000000000000	11	0.017164083000
Hybrid4: BTsection-FalsePos	[1,6]	1.929846242848	0.000000000000	7	0.005307125000

Table 6. Function $f(x) = x^2 - x - 2$					
Method	Interval	Root	Error	Iterations	CPU Time
Hybrid1: Bisection-FalsePos	[1,7]	2.000000000000	0.000000000000	9	0.003208750000
Hybrid2: Trisection-FalsePos	[1,7]	2.000000000000	0.000000000000	7	0.003186292000
Hybrid3: Trisection-NewtonR	[1,7]	2.000000000000	0.000000000001	13	0.018284292000
Hybrid4: BTsection-FalsePos	[1,7]	2.000000000000	0.000000000000	6	0.002163208000

Table 7. Function $f(x) = x^2 - 2$					
Method	Interval	Root	Error	Iterations	CPU Time
Hybrid1: Bisection-FalsePos	[1,3]	1.414213562373	0.000000000000	8	0.003360958000
Hybrid2: Trisection-FalsePos	[1,3]	1.414213562373	0.000000000000	7	0.002914916000
Hybrid3: Trisection-NewtonR	[1,3]	1.414213562373	0.000000000000	13	0.005550834000
Hybrid4: BTsection-FalsePos	[1,3]	1.414213562373	0.000000000000	6	0.007887542000

Table 8. Function $f(x) = x - \exp(-x)$					
Method	Interval	Root	Error	Iterations	CPU Time
Hybrid1: Bisection-FalsePos	[0,2]	0.567143290410	0.000000000000	8	0.046667333000
Hybrid2: Trisection-FalsePos	[0,2]	0.567143290410	0.000000000000	7	0.017032042000
Hybrid3: Trisection-NewtonR	[0,2]	0.567143290409	0.000000000001	11	0.015795000000
Hybrid4: BTsection-FalsePos	[0,2]	0.567143290410	0.000000000000	6	0.014784125000

Table 9. Function $f(x) = x - \cos(x)$					
Method	Interval	Root	Error	Iterations	CPU Time
Hybrid1: Bisection-FalsePos	[0,1]	0.739085133215	0.000000000000	7	0.008797250000
Hybrid2: Trisection-FalsePos	[0,1]	0.739085133215	0.000000000000	6	0.004603500000
Hybrid3: Trisection-NewtonR	[0,1]	0.739085133215	0.000000000001	11	0.011723000000
Hybrid4: BTsection-FalsePos	[0,1]	0.739085133215	0.000000000001	5	0.002825583000

## 2.4. Algorithms

### 2.4.1. Bisection, Regula Falsi, Newton-Raphson algorithm

These algorithms are ubiquitous [1], [2], [3]]. We use Regula Falsi method for analysis of hybrid algorithms. There are multiple reasons for neglecting Newton Raphson from this analysis: (a) it requires that function be differentiable, (2) it depends heavily on the start point, (3) iterated approximations are not bracketed, (4) it fails if start point is not close to the root. Although False Position method is preferably used in hybrid algorithms, it does not fit well in the category of Bisection, Trisection and BTsection, we will first compare these algorithms.

#### 2.4.2. Trisection Algorithm [11].

Bader extended bisection algorithm to trisection in order to create a better algorithm to hybridize it with false position algorithm. The algorithm is as follows.

Input: Function  $f(x)$ , Initial approximations  $[a, b]$  and absolute error  $eps$ .

Output: Approximate root  $r$ , *bracketing interval*, and number of iterations  $k$

```

for k=1 to n
    p := (2*a + b)/3; q := (a + 2*b)/3;
    if |f(p)| < |f(q)|
        r := p
    else
        r := q
    endif;
    if |f(r)| < eps
        return r, a, b, k;
    else if f(a)*f(p) < 0
        b:=p;
    else if f(p)*f(q) < 0
        a:=p;
        b:=q;
    else
        a:=q;
    end if;
end for

```

---

#### 2.4.3. BTsection Algorithm

In this paper, we adapted both Bisection and Trisection algorithms to craft a BTsection algorithm to hybridize, more heuristically, with False Position method [14]. BTsection algorithm incurs no more computational cost than the Trisection algorithm, but requires fewer iterations, in general, to reach an approximate solution. The algorithm is as follow.

Input: function  $f(x)$ , Initial interval  $[a, b]$  and absolute error  $eps$ .

Output: Approximate root  $r$ , *bracketing interval*, and number of iterations  $k$

```

for i=1:imax
    bisection step
        r=(a+b)/2;
        if (f(a)*f(r)) < 0
            b=r; r=(a+2*b)/3;
        elseif (f(r)*f(b)) < 0
            a=r; r=(2*a+b)/3;
        end if;
    end for

```



```

else
return i, r, a, b
endif
trisection step
if (f(a)*f(r)) < 0
b=r;
elseif (f(r)*f(b)) < 0
a=r;
else
return i, r, a, b
endif
if |f(r)| < eps
return i,r,a,b ;
endif
endfor

```

The Trisection algorithm and a new BTsection algorithm are conceptually equivalent in each iteration. From the following tables, it is clear that BTsection algorithm requires fewer iterations to converge. It shows that BTsection algorithm competes successfully with the other algorithms with respect to number of loop iterations, CPU time, and accuracy in approximation of the root. These benchmark functions appear in recent papers in the literature. See Tables 10-15 for comparing the performance of the sectioning algorithms. The functions and the interval of definition are generic of any algorithm. In all the ensuing examples in this paper, algorithm halting upperbound is 40 for iterations and upto 12 decimal digits for approximation in the acceptable root.

Table 10. Function  $x^2 - x - 2$ 

Method	Interval	Root	Error.	Iterations	CPU Time
Bisection	[1,6]	1.000000000000	0.000000000003	40	0.006641375000
Trisection	[1,6]	2.000000000000	0.000000000001	27	0.007940666000
BTsection	[1,6]	2.000000000000	0.000000000001	17	0.006606125000_

Table 11. Function  $0.986x^3 - 5.181x^2 + 9.067x - 5.289$ 

Method	Interval	Root	Error.	Iterations	CPU Time
Bisection	[1,5]	1.929846242856	0.000000000001	37	0.002006000000
Trisection	[1,5]	1.929846242855	0.000000000001	24	0.002049208000
BTsection	[1,5]	1.929846242843	0.000000000000	20	0.003908708000_

Table 12. Function  $\exp(x).(x-1)$ 

Method	Interval	Root	Error.	Iterations	CPU Time
Bisection	[0,4]	1.000000000000	0.000000000000	2	0.012940750000
Trisection	[0,4]	1.000000000000	0.000000000001	29	0.016367250000
BTsection	[0,4]	1.000000000000	0.000000000001	2	0.017235792000

Table 13. Function  $(x-1).(x-2).(x-3)$

Method	Interval	Root	Error.	Iterations	CPU Time
Bisection	[1,3]	2.000000000000	0.000000000000	1	0.003182459000
Trisection	[1,3]	2.000000000000	0.000000000000	26	0.003927291000
BTsection	[1,3]	2.000000000000	0.000000000000	1	0.003374833000

Table 14. Function  $x-\cos(x)$

Method	Interval	Root	Error.	Iterations	CPU Time
Bisection	[0,1]	0.739085133215	0.000000000001	39	0.026511958000
Trisection	[0,1]	0.739085133215	0.000000000000	24	0.017229750000
BTsection	[0,1]	0.739085133215	0.000000000000	21	0.021565208000

Table 15. Function  $x-\exp(-x)$

Method	Interval	Root	Error.	Iterations	CPU Time
Bisection	[0,1]	0.567143290409	0.000000000001	38	0.009934417000
Trisection	[0,1]	0.567143290410	0.000000000000	25	0.008384250000
BTsection	[0,1]	0.567143290409	0.000000000001	19	0.013742375000

### 3. Hybrid Algorithms

Now we present recent hybrid algorithms and a new proposed hybrid algorithm. The existing hybrid algorithms have one thing in common. At each iteration, they compute the bracketing interval for each of the two hybridizing algorithms, and compute the interval common to the two algorithms to use at the next iteration. It incurs a computation step. Here in the new algorithm, there is no need to perform such computation because we can have the common interval readily available without performing this computation.

First we describe the original hybrid algorithm, namely, Hybrid1 based on classical Bisection and False Position algorithms. Since the classical algorithms can be found in any text book, those algorithms are not described here.

In Hybrid1 algorithm, at each iteration, more promising root between the Bisection and False Position approximate roots is selected, and common interval is computed for the next iteration. This curtails the unnecessary iterations in either method. It was succeeded by more efficient algorithms using Trisection method in place of Bisection algorithm: Hybrid2 and Hybrid3 [12]. These algorithms lead the way for us to discover more heuristics to design a new blended algorithm Hybrid4 which is more efficient than the previous three hybrid algorithms. All the four algorithms are described here for reference. For ease in readability, the prefixed variables use symbols: b for Bisection, p for False Position, t for Trisection, and bt for BTsection. The variables without prefix are algorithms local working variable.

#### Hybrid1: Bisection and False Position Algorithm [10]

Input:  $f$ ,  $[a, b]$ ,  $\epsilon_s$ ,  $\maxIterations$

Output: root  $r$ ,  $k$ -number of iterations, error of approximation  $\epsilon_a$ , bracketing interval  $[a_{k+1}, b_{k+1}]$

//initialize

$k = 0$ ;  $a_1 = a$ ,  $b_1 = b$

Initialize bounded interval for bisection and false position

$pa_{k+1}=ba_{k+1}=a_1$ ;  $pb_{k+1}=bb_{k+1}=b_1$

repeat

$pa_{k+1}=ba_{k+1}=a_k$ ;  $pb_{k+1}=bb_{k+1}=b_k$

compute the mid point the error

```

    m =  $\frac{a_k + b_k}{2}$ , and  $\epsilon_m = |f(m)|$ 
    compute the False Position point and error,
    s =  $a_k - \frac{f(a_k)(b_k - a_k)}{f(b_k) - f(a_k)}$  and  $\epsilon_p = |f(s)|$ 
    if  $|f(m)| < |f(s)|$ ,
    f(m) is closer to zero, Bisection method determines bracketing interval  $[b_{k+1}, bb_{k+1}]$ 
    r = m
     $\epsilon_a = \epsilon_m$ 
    if  $f(a_k) \cdot f(r) > 0$ ,
         $b_{k+1} = r$ ;  $bb_{k+1} = b_k$ ;
    else
         $b_{k+1} = a_k$ ;  $bb_{k+1} = r$ ;
    endif
else
    f(s) is closer to zero, False Position method determines bracketing interval  $[p_{k+1}, pb_{k+1}]$ 
    r = s
     $\epsilon_a = \epsilon_p$ 
    if  $f(a_k) \cdot f(r) > 0$ ,
         $p_{k+1} = r$ ;  $pb_{k+1} = b_k$ ;
    else
         $p_{k+1} = a_k$ ;  $pb_{k+1} = r$ ;
    endif
endif
Since the root is bracketed by both  $[b_{k+1}, bb_{k+1}]$  and  $[p_{k+1}, pb_{k+1}]$ , set
 $[a_{k+1}, b_{k+1}] = [b_{k+1}, bb_{k+1}] \cap [p_{k+1}, pb_{k+1}]$  or
 $a_{k+1} = \max(b_{k+1}, p_{k+1})$ ;
 $b_{k+1} = \min(bb_{k+1}, pb_{k+1})$ ;
outcome: iteration complexity, root, and error of approximation
iterationCount = k
r = rk
error =  $\epsilon_a = |f(r)|$ 
k = k + 1
until  $|f(r)| < \epsilon_s$  or k > maxIterations

```

---

### Hybrid2: Trisection and False Position Algorithm[11]

This function implements a blend of trisection and false position methods.

Input: The function f; the interval [a, b] where  $f(a)f(b) < 0$  and the root lies in [a, b],

The absolute error (eps).

Output: The root (x), The value of f(x), Numbers of iterations (n), the interval [a, b] where the root lies in

$n = 0$ ;  $a1 := a$ ;  $a2 := a$ ;  $b1 := b$ ,  $b2 := b$

while true do

n := n + 1

$xT1 := (b + 2*a)/3$

$xT2 := (2*b + a)/3$

```

    xF := a - (f(a)*(b - a))/(f(b) - f(a))
    x := xT1
    fx := fxT1
    if |f(xT2)| < |f(x)|
    x := xT2
    if |f(xF)| < |f(x)|
        x := xF
    if |f(x)| <= eps
        return x, f(x), n, a, b
    if fa * f(xT1) < 0
        b1 := xT1
    else if f(xT1) * f(xT2) < 0
        a1 := xT1
        b1 := xT2
    else
        a1 := xT2
    if fa*f(xF) < 0
        b2 := xF;
    else
        a2 := xF;
    a := max(a1, a2) ; b := min(b1, b2)
end (while)

```

---

### Hybrid 3: Trisection and Newton-Raphson Algorithm

This algorithm is along the same lines as Hybrid 2, but with (1) instead of false position method, it uses Newton-Raphson algorithm which requires differentiability of the function, (2) improved iteration count and accuracy in the hybridization step: namely, the common interval in each iteration is computed by analyzing the five function values and then mapped to parameter values for the optimal interval.

The Algorithm is as follows.

Input: Function  $f(x)$ , an Initial approximations  $x_0$  and absolute error  $eps$ .

Output: Root  $x$  and number of iterations  $n$

```

df(x):=f'(x); k:=0;
for k=1:n
    p := (2*a + b)/3;
    q := (a + 2*b)/3;
    if |f(p)| < |f(q)|
    then r := p - f(p)/df(p);
    else r := q - f(q)/df(q);
    end if;
    if |f(r)| < eps
    then
    return r, k;
    else

```

```

find fv:={f(a),f(b),f(r),f(p),f(q)};
a := xa where fv max -ve;
b := xb where fv min +ve
end if;
end.

```

---

Recall, in these three algorithms, there are two common steps to coordinate the two algorithms to hybridize. At each iteration, they determine (1) the promising approximation root out of the two roots (2) the common interval bracketing the approximate root. In Hybrid1 and Hybrid 2 algorithms, this simply reduces to intersection of two intervals so that common interval contains the approximate root. No function evaluation is involved in the search for common interval to contain the predicted approximate root. In the Hybrid3 algorithm, it searches among five function values used to determine two function values pertaining the common interval. From these two function values, the function parameters are determined to create the common bracketing interval.

#### ***New Hybrid Algorithm***

Hybrid4 algorithm provides a more proficient approach to optimization: (1) BTsection algorithm is used instead of bisection or trisection, (2) it eliminates the computation of common interval required by the foregoing algorithms used to hybridize, There is no work needed to determine the better of the two roots. This leads to more efficiency for optimal root approximation and readily available common interval. It is based on common sense Occam's razor principle [13], Figure 2. The Occam's razor principle is a heuristic, not a proof. That is, when faced with competing choices, we use the simplest from what we have. It will be shown that Occam's Razor Principle works quite well in this case.



Figure2 <https://conceptually.org/concepts/occams-razor>

#### **Hybrid4.**

This algorithm is a blend of BTsection and False Position methods to find iteration approximations with minimal effort.

Input  $a_0, b_0, r_0, \text{eps}, \text{imax}, f$

Output  $k, a_k, b_k, r_k$

for  $k=1:\text{imax}$

*BTsection iteration step determines*

$btak, btb_k, btr_k$  from  $btak-1, btb_{k-1}, btr_{k-1}$

reliable

$btak, btb_k, btr_k$  to  $a_k, b_k, r_k$

*False-position iteration step*

input is  $a_k, b_k, r_k$  instead of old  $pa_{k-1}, pb_{k-1}, pr_{k-1}$

false position iteration step determines

$pa_k, pb_k, pr_k$  from  $a_k, b_k, r_k$  instead of from old  $pa_{k-1}, pb_{k-1}, pr_{k-1}$   
 This makes  $[pa_k, pb_k]$  as the common interval  $[a_k, b_k]$  without any computation.  
 The use of  $a_k, b_k, r_k$  instead of old  $pa_{k-1}, pb_{k-1}, pr_{k-1}$ , makes this step more optimal for next approximation

```

    if  $f(pr_k) < \epsilon$ 
        return  $k, pa_k, pb_k, pr_k$ 
    end
    relabel  $pa_k, pb_k, pr_k$  to  $qa_k, qb_k, qr_k$ 
endFor

```

---

Summarizing the foregoing algorithms, succinctly the *iteration step* in the algorithms are:

#### Hybrid1

```

 $[ba_k, bb_k, br_k] = \text{Bisection}(a_{k-1}, b_{k-1}, r_{k-1}, f)$ 
 $[pa_k, pb_k, pr_k] = \text{FalsePosition}(a_{k-1}, b_{k-1}, r_{k-1}, f)$ 

```

The results of hybridization step are:  
 $r_k$  is better of  $br_k$   $pr_k$ ,  
 $[a_k, b_k]$  is common to  $[ba_k, bb_k]$ ,  $[pa_k, pb_k]$ ,  
 $r_k$  belongs to  $[a_k, b_k]$

#### Hybrid 2

```

 $[ta_k, tb_k, tr_k] = \text{Trisection}(a_{k-1}, b_{k-1}, r_{k-1}, f)$ 
 $[pa_k, pb_k, pr_k] = \text{FalsePosition}(a_{k-1}, b_{k-1}, r_{k-1}, f)$ 

```

The outcomes of hybridization are:  
 $r_k$  is better of  $tr_k$   $pr_k$ ,  
 $[a_k, b_k]$  is common to  $[ta_k, tb_k]$ ,  $[pa_k, pb_k]$ ,  
 $r_k$  belongs to  $[a_k, b_k]$

#### Hybrid3

```

 $[ta_k, tb_k, tr_k] = \text{Trisection}(a_{k-1}, b_{k-1}, r_{k-1}, f)$ 
 $[na_k, nb_k, nr_k] = \text{NewtonRaphson}(a_{k-1}, b_{k-1}, r_{k-1}, f)$ 

```

The upshot of hybridization is:  
 $r_k$  as better of  $tr_k$   $nr_k$ ,  
 $[a_k, b_k]$  is common from  $[ba_k, bb_k]$ ,  $[pa_k, pb_k]$  and  $nr_k$  by analyzing  $\{f(ba_k), f(bb_k), f(na_k), f(nb_k), f(nr_k)\}$  then finding from  $\{\{ba_k, bb_k, pa_k, pb_k, nr_k\}\}$ .  
 $r_k$  belongs to  $[a_k, b_k]$

#### Hybrid4

```

 $[a, b, r] = \text{BTsection}(a_{k-1}, b_{k-1}, r_{k-1}, f)$ 
 $[pa_k, pb_k, pr_k] = \text{FalsePosition}(a, b, r, f)$ 

```

The conclusion of hybridization is: there is no need to do any work calculation:  $[a_k, b_k]$  is the same interval  $[pa_k, pb_k]$  containing  $r_k$  is the same as  $pr_k$ , the desired root. This algorithm is optimal in the number of iterations and the accuracy in approximate root.



#### 4. Discussion

Many researchers focused their attention toward using such methods to solve their problems. The roots are calculated iteratively, along with the number of iterations within a specified tolerance. In this section, all the existing hybrid methods are compared. Error analysis is performed. It is validated here that Hybrid4 algorithm preferable than the existing algorithms Hybrid1, Hybrid2, and Hybrid3.

##### 4.1. Empirical Evidence Testing.

We have tested our new algorithm Hybrid4 with other hybrid algorithms on diverse examples of functions found in articles in the literature to validate that new algorithm outperforms the other three hybrid algorithms.

##### 4.2. Experiments in Matlab

Moazzam used Microsoft visual C++ to find roots, we used Matlab R2022A 64 bit (maci64) on MacBook Pro MacOS Sonoma 14.1.1 16 GB Apple M1Pro . Along with numerous different functions and varying intervals, we use the same frequently cited functions. Overall we have validated that the new algorithm performs better than all the hybrid algorithms. These tests indicate that hybrid with false position algorithm is still preferable to hybridization with Newton-Raphson possibly because the NR method requires that the function be differentiable.

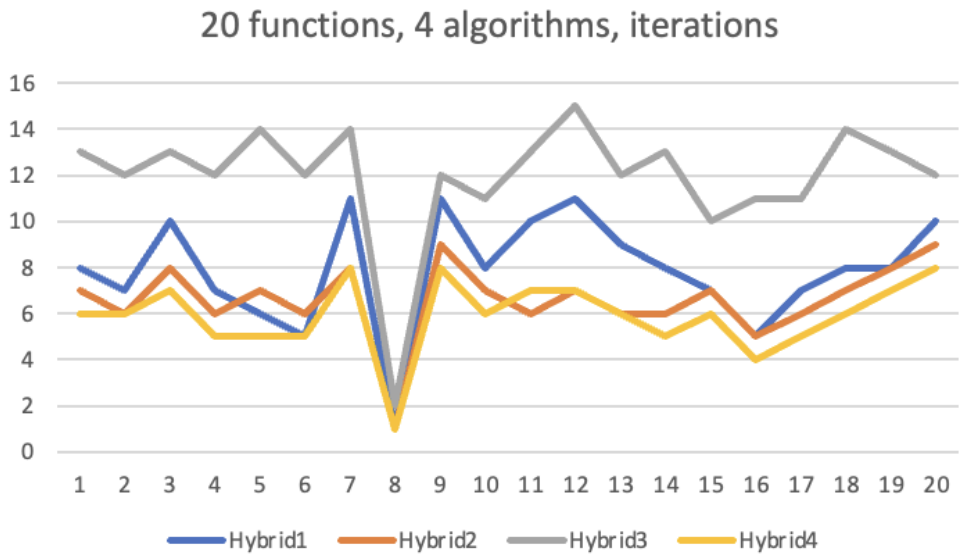
For all these Tables 16-34, the functions are frequently occurring in the literature. For all the functions the upperbound on the number of iterations is 40 and acceptable approximation error bound is  $10^{-12}$  . These tables are for comparison of four algorithms number of iterations, CPU compute time, accuracy of the solution.

Frequently used twenty functions are compared for iterations efficiency. These functions are listed in the table. This table lists functions used in the simulations. Columns represent the functions, the interval of function's domains, the number of iterations required by each of four algorithm to reach an approximate root within tolerance. Figure3 is a visual representation of the table.

**Table 15.** Listing of functions, intervals and iterations in four hybrid algorithms.

Function	interval	iterations			
		Hybrid1	Hybrid2	Hybrid3	Hybrid4
$x^2 - 2$	[1,3]	8	7	13	6
$x^2 - 3$	[1,2]	7	6	12	6
$x^2 - 5$	[2,7]	10	8	13	7
$x^2 - 10$	[3,4]	7	6	12	5
$x^2 - x - 2$	[1,5]	6	7	14	5
$x^2 + 2x - 7$	[1,3]	5	6	12	5
$x^3 - 2$	[1,8]	11	8	14	8
$(x-1).(x-2).(x-3)$	[1,3]	1	1	2	1
$x^{10} - 1$	[0,1.4]	11	9	12	8
$x - \exp(-x)$	[0,2]	8	7	11	6
$x \cdot \exp(x) - 7$	[0,2]	10	6	13	7
$x \cdot \exp(x) - 7$	[0,3]	11	7	15	7
$\exp(x) - 3x - 2$	[2,3]	9	6	12	6
$x - \exp(-x)$	[1,2]	8	6	13	5
$\sin(x) - x^2$	[0.6,1]	7	7	10	6
$x \cos(x) - 1$	[0,2]	5	5	11	4
$x - \cos(x)$	[0,1]	7	6	11	5
$\sin(x) \sinh(x) + 1$	[3,4]	8	7	14	6
$\exp(x).(x-1)$	[0,4]	8	8	13	7
$0.986x^3 - 5.181x^2 + 9.067x - 5.289$	[1,5]	10	9	12	8

In the chart below, functions are along x-axis, number of iterations along y-axis, legend color indicates algorithms 1,2,3,4: Hybrid1 is series1 (blue) Hybrid2 is series2(magenta), Hybrid3 is serie3(gray), Hybrid4 is series 4 (yellow). Four algorithms are in the body of the chart.



**Figure 3.** Comparing iterations in twenty functions by four algorithms.

The tables 16–34 depict the Matlab numerical outputs of the simulation on 20 functions.

Table 16. Function  $x^2 - 2$

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[1,3]	1.414213562373	0.000000000000	8	0.020685333000
Hybrid2: Trisection-FalsePos	[1,3]	1.414213562373	0.000000000000	7	0.013117792000
Hybrid3: Trisection-NewtonR	[1,3]	1.414213562373	0.000000000000	13	0.026385250000
Hybrid4: BTsection-FalsePos	[1,3]	1.414213562373	0.000000000000	6	0.013243375000

Table 17. Function  $x^2 - 3$

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[1,2]	1.732050807569	0.000000000000	7	0.025633708000
Hybrid2: Trisection-FalsePos	[1,2]	1.732050807569	0.000000000000	6	0.010531792000
Hybrid3: Trisection-NewtonR	[1,2]	1.732050807569	0.000000000000	12	0.021213458000
Hybrid4: BTsection-FalsePos	[1,2]	1.732050807569	0.000000000000	6	0.011047208000

Table 18. Function  $x^2 - 5$

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[2,7]	2.236067977500	0.000000000000	10	0.006093541000
Hybrid2: Trisection-FalsePos	[2,7]	2.236067977500	0.000000000000	8	0.002769917000
Hybrid3: Trisection-NewtonR	[2,7]	2.236067977500	0.000000000000	13	0.018985375000
Hybrid4: BTsection-FalsePos	[2,7]	2.236067977500	0.000000000000	7	0.002470208000

Table 19. Function  $x^2 - 10$

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[3,4]	3.162277660168	0.000000000000	7	0.002982000000
Hybrid2: Trisection-FalsePos	[3,4]	3.162277660168	0.000000000000	6	0.001419541000
Hybrid3: Trisection-NewtonR	[3,4]	3.162277660168	0.000000000001	12	0.005096542000
Hybrid4: BTsection-FalsePos	[3,4]	3.162277660168	0.000000000000	5	0.001676208000

Table 20. Function  $x^2 - x - 2$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[1,5]	2.000000000000	0.000000000000	6	0.007671917000
Hybrid2: Trisection-FalsePos	[1,5]	2.000000000000	0.000000000000	7	0.006940042000
Hybrid3: Trisection-NewtonR	[1,5]	2.000000000000	0.000000000000	14	0.008945583000
Hybrid4: BTsection-FalsePos	[1,5]	2.000000000000	0.000000000000	5	0.005738792000

Table 21. Function  $x^2 + 2x - 7$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[1,3]	1.828427124746	0.000000000000	5	0.004073625000
Hybrid2: Trisection-FalsePos	[1,3]	1.828427124746	0.000000000000	6	0.003551708000
Hybrid3: Trisection-NewtonR	[1,3]	1.828427124746	0.000000000001	12	0.005001833000
Hybrid4: BTsection-FalsePos	[1,3]	1.828427124746	0.000000000000	5	0.002713375000

Table 22 . Function  $x^3 - 2$ 

Method	Interval	Root	Error	Iterations.	CPU Time
Hybrid1: Bisection-FalsePos	[1,8]	1.259921049895	0.000000000000	11	0.002263292000
Hybrid2: Trisection-FalsePos	[1,8]	1.259921049895	0.000000000000	8	0.002984666000
Hybrid3: Trisection-NewtonR	[1,8]	1.259921049895	0.000000000000	14	0.006186750000
Hybrid4: BTsection-FalsePos	[1,8]	1.259921049895	0.000000000000	8	0.002300708000

Table 23. Function  $(x-1).(x-2).(x-3)$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[1,3]	1.000000000000	0.000000000000	1	0.011055917000
Hybrid2: Trisection-FalsePos	[1,3]	1.000000000000	0.000000000000	1	0.009512750000
Hybrid3: Trisection-NewtonR	[1,3]	1.000000000000	0.000000000000	2	0.015483084000
Hybrid4: BTsection-FalsePos	[1,3]	1.000000000000	0.000000000000	1	0.009636834000

Table .24 Function  $x^{10} - 1$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[0,1.4]	1.000000000000	0.000000000000	11	0.003317792000
Hybrid2: Trisection-FalsePos	[0,1.4]	1.000000000000	0.000000000000	9	0.004301584000
Hybrid3: Trisection-NewtonR	[0,1.4]	1.000000000000	0.000000000000	12	0.005507042000
Hybrid4: BTsection-FalsePos	[0,1.4]	1.000000000000	0.000000000000	8	0.003067708000

Table 25. Function  $x\text{-exp}(-x)$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[0,2]	0.567143290410	0.000000000000	8	0.040281833000
Hybrid2: Trisection-FalsePos	[0,2]	0.567143290410	0.000000000000	7	0.021879167000
Hybrid3: Trisection-NewtonR	[0,2]	0.567143290409	0.000000000001	11	0.033102500000
Hybrid4: BTsection-FalsePos	[0,2]	0.567143290410	0.000000000000	6	0.018546208000

Table 26. Function  $x*\exp(x) - 7$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[0,3]	1.524345204984	0.000000000000	11	0.002657292000
Hybrid2: Trisection-FalsePos	[0,3]	1.524345204984	0.000000000000	7	0.001618333000
Hybrid3: Trisection-NewtonR	[0,3]	1.524345204984	0.000000000000	15	0.005757834000
Hybrid4: BTsection-FalsePos	[0,3]	1.524345204984	0.000000000000	7	0.002397334000

Table 27. Function  $\exp(x)-3x-2$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[2,3]	2.125391198811	0.000000000000	9	0.003195209000
Hybrid2: Trisection-FalsePos	[2,3]	2.125391198811	0.000000000000	6	0.001899459000
Hybrid3: Trisection-NewtonR	[2,3]	2.125391198811	0.000000000001	12	0.015803167000
Hybrid4: BTsection-FalsePos	[2,3]	2.125391198811	0.000000000000	6	0.002095750000

Table 28 Function  $x\text{-exp}(-x)$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[1,2]	1.593550763288	0.000000000000	8	0.003090166000
Hybrid2: Trisection-FalsePos	[1,2]	1.593550763288	0.000000000000	6	0.002133250000
Hybrid3: Trisection-NewtonR	[1,2]	1.593550763288	0.000000000000	13	0.006343166000
Hybrid4: BTsection-FalsePos	[1,2]	1.593550763288	0.000000000000	5	0.002137834000

Table 29. Function  $\sin(x)-x^2$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[0.6,1]	0.876726215395	0.000000000001	7	0.006993375000
Hybrid2: Trisection-FalsePos	[0.6,1]	0.876726215395	0.000000000000	7	0.001796917000
Hybrid3: Trisection-NewtonR	[0.6,1]	0.876726215395	0.000000000000	10	0.009451666000
Hybrid4: BTsection-FalsePos	[0.6,1]	0.876726215395	0.000000000000	6	0.002477750000

Table 30. Function  $x\text{-cos}(x)$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[-2,4]	2.073932809091	0.000000000000	9	0.003396375000
Hybrid2: Trisection-FalsePos	[-2,4]	2.073932809091	0.000000000000	7	0.003089125000
Hybrid3: Trisection-NewtonR	[-2,4]	2.073932809091	0.000000000001	11	0.028336042000
Hybrid4: BTsection-FalsePos	[-2,4]	2.073932809091	0.000000000000	5	0.005645708000

Table 31 Function  $\mathbf{x\cos(x)-1}$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[0,2]	1.114157140872	0.000000000000	5	0.009676417000
Hybrid2: Trisection-FalsePos	[0,2]	1.114157140872	0.000000000000	5	0.002748458000
Hybrid3: Trisection-NewtonR	[0,2]	1.114157140871	0.000000000001	11	0.018048625000
Hybrid4: BTsection-FalsePos	[0,2]	1.114157140872	0.000000000000	4	0.004841666000

Table 32. Function  $\mathbf{\sin(x)\sinh(x) + 1}$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[3,4]	3.221588399094	0.000000000000	8	0.023326292000
Hybrid2: Trisection-FalsePos	[3,4]	3.221588399094	0.000000000000	7	0.034500000000
Hybrid3: Trisection-NewtonR	[3,4]	3.221588399094	0.000000000000	14	0.062597125000
Hybrid4: BTsection-FalsePos	[3,4]	3.221588399094	0.000000000000	6	0.024583292000

Table 33. Function  $\mathbf{0.986*x.^3 -5.181* x.^2 +9.067*x-5.289}$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[1,5]	1.929846242844	0.000000000000	10	0.003417250000
Hybrid2: Trisection-FalsePos	[1,5]	1.929846242848	0.000000000000	9	0.003345834000
Hybrid3: Trisection-NewtonR	[1,5]	1.929846242850	0.000000000000	12	0.005083042000
Hybrid4: BTsection-FalsePos	[1,5]	1.929846242848	0.000000000000	7	0.002059334000

Table 34 . Function  $\mathbf{\exp(x).*(x-1)}$ 

Method	Interval	Root	Error	Iterations	CPUTime
Hybrid1: Bisection-FalsePos	[0,4]	1.000000000000	0.000000000000	8	0.002467334000
Hybrid2: Trisection-FalsePos	[0,4]	1.000000000000	0.000000000000	8	0.001703875000
Hybrid3: Trisection-NewtonR	[0,4]	1.000000000000	0.000000000001	13	0.005289167000
Hybrid4: BTsection-FalsePos	[0,4]	1.000000000000	0.000000000000	7	0.001907000000

## V. Conclusion

We have designed and implemented a new algorithm Hybrid4, an efficient algorithm hybrid of BTsection and Regula Falsi methods. The algorithm was implemented in Matlab R2022A 64 bit (maci64) on MacBook Pro MacOS Sonoma 14.1.1 16 GB Apple M1Pro. The implementation tests Tables15-34 and numerical and visual iteration count graph indicate that Hybrid4 outperforms the Hybrid1, Hybrid2, and Hybrid3 algorithms. The experiments on numerous datasets used in the literature validate that the new algorithm Hybrid4 is effective both conceptually and computationally. This algorithm is applicable to both differentiable and non-differentiable functions. This paper provides an essential algorithm for the practitioners and contribution to the repertoire of hybrid algorithms.

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