



University of Dhaka

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**Implementation and Comparative Analysis of the Bisection,
False Position, Newton–Raphson, and Secant Methods for
finding roots of nonlinear equations.**

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1 Introduction

Root-finding is a fundamental problem in numerical analysis, where the objective is to find the values of a variable for which a given function equals zero. These values are known as roots, or zeros, of the function. In many engineering and computational problems, analytical solutions are often difficult or impossible to obtain, so iterative numerical approaches are employed instead.

In this lab, the objective is to find the height h in a draining tank that satisfies the flow balance equation:

$$f(h) = h^3 - 10h + 5e^{-h/2} - 2 = 0$$

The following four root-finding algorithms are applied:

- **Bisection Method:** A bracketing approach based on the Intermediate Value Theorem.
- **False Position (Regula Falsi):** A bracketing method that uses linear interpolation.
- **Newton–Raphson Method:** An open method that relies on tangents and derivatives.
- **Secant Method:** An open, derivative-free approach similar to Newton–Raphson.

2 Objectives

The primary objectives of this lab report are:

1. To implement the Bisection and False-Position methods to find the root of the given nonlinear equation within a suitable interval, ensuring an approximate relative error (ϵ_a) of less than or equal to 0.001%.
2. To implement the Newton-Raphson and Secant methods using specified initial guesses to determine the root, also targeting an approximate relative error (ϵ_a) of less than or equal to 0.001%.
3. To tabulate all necessary data, including iteration numbers, approximate root values, function values, and approximate relative errors, for each method.
4. To compare the convergence rates of all implemented methods by plotting the approximate relative error (ϵ_a) versus the number of iterations and providing a detailed analysis.

3 Algorithms

3.1 Bisection Method

1. Choose an interval $[x_l, x_u]$ such that $f(x_l) \cdot f(x_u) < 0$, guaranteeing a root exists within the interval.
2. Calculate the midpoint $x_r = (x_l + x_u)/2$.
3. If $f(x_l) \cdot f(x_r) < 0$, the root is in $[x_l, x_r]$, so set $x_u = x_r$.
4. If $f(x_l) \cdot f(x_r) > 0$, the root is in $[x_r, x_u]$, so set $x_l = x_r$.
5. If $f(x_l) \cdot f(x_r) = 0$, then x_r is the root.
6. Repeat steps 2-5 until the approximate relative error ϵ_a is below the threshold or the maximum number of iterations is reached.

3.2 False Position Method

1. Choose an interval $[x_l, x_u]$ such that $f(x_l) \cdot f(x_u) < 0$.
2. Estimate the new root $x_r = x_u - \frac{f(x_u)(x_l - x_u)}{f(x_l) - f(x_u)}$.
3. If $f(x_l) \cdot f(x_r) < 0$, the root is in $[x_l, x_r]$, so set $x_u = x_r$.
4. If $f(x_l) \cdot f(x_r) > 0$, the root is in $[x_r, x_u]$, so set $x_l = x_r$.
5. If $f(x_l) \cdot f(x_r) = 0$, then x_r is the root.
6. Repeat steps 2-5 until the approximate relative error ϵ_a is below the threshold or the maximum number of iterations is reached.

3.3 Newton-Raphson Method

1. Choose an initial guess x_0 .
2. Calculate the next approximation $x_1 = x_0 - \frac{f(x_0)}{f'(x_0)}$.
3. Set $x_0 = x_1$.
4. Repeat steps 2-3 until the approximate relative error ϵ_a is below the threshold or the maximum number of iterations is reached.

3.4 Secant Method

1. Choose two initial guesses x_0 and x_1 .
 2. Calculate the next approximation $x_2 = x_1 - \frac{f(x_1)(x_1 - x_0)}{f(x_1) - f(x_0)}$.
 3. Set $x_0 = x_1$ and $x_1 = x_2$.
 4. Repeat steps 2-3 until the approximate relative error ϵ_a is below the threshold or the maximum number of iterations is reached.
-

4 Implementation

4.1 Function Definitions

```

1 import math
2 import matplotlib.pyplot as plt
3
4 def f(h):
5     return h**3 - 10*h + 5*math.exp(-h/2) - 2
6
7 def df(h):
8     return 3*h**2 - 10 - (5/2)*math.exp(-h/2)

```

4.2 Bisection Method Implementation

```

1
2
3 def bisection(f, xl, xu, max_it=100, rel_er_thr=None):
4     fl = f(xl)
5     fu = f(xu)
6     if fl * fu > 0:
7         return ValueError("The interval is not valid")
8
9     results = []
10    it = 0
11    old_xr = None
12    while it < max_it:
13        xr = (xl + xu)/2
14        fr = f(xr)
15        ea = None if old_xr is None else abs((xr - old_xr)/xr)*100
16
17        results.append({
18            'iteration': it+1,
19            'xl': xl,
20            'xu': xu,
21            'xr': xr,
22            'f(xl)': fl,
23            'f(xu)': fu,
24            'f(xr)': fr,
25            'ea(%)': ea
26        })
27
28        if ea is not None and rel_er_thr is not None and ea <
29            rel_er_thr:
30            print(f"Bisection approximate root found at x =
31                {xr:.10f} (iter {it+1})")
32            break
33
34        if fl * fr < 0:
35            xu = xr
36            fu = fr
37        elif fl * fr > 0:
38            xl = xr

```

```

37         fl = fr
38     else:
39         print(f"The real root is: {xr:.10f}")
40         break
41     old_xr = xr
42     it += 1
43
44 return results

```

4.3 False Position Method Implementation

```

1
2 def false_position(f, xl, xu, max_it=100, rel_er_thr=None):
3     fl = f(xl)
4     fu = f(xu)
5     if fl * fu > 0:
6         return ValueError("The interval is not valid")
7
8     results = []
9     it = 0
10    old_xr = None
11
12    while it < max_it:
13        xr = (xl*fu - xu*fl)/(fu - fl)
14        fr = f(xr)
15        ea = None if old_xr is None else abs((xr - old_xr)/xr)*100
16
17        results.append({
18            'iteration': it+1,
19            'xl': xl,
20            'xu': xu,
21            'xr': xr,
22            'f(xl)': fl,
23            'f(xu)': fu,
24            'f(xr)': fr,
25            'ea(%)': ea
26        })
27
28    if ea is not None and rel_er_thr is not None and ea <
29        rel_er_thr:
30        print(f"False Position approximate root found at x =
31              {xr:.10f} (iter {it+1})")
32        break
33
34    if fl * fr < 0:
35        xu = xr
36        fu = fr
37    elif fl * fr > 0:
38        xl = xr
39        fl = fr
40    else :
41        print(f"The real root is: {xr:.10f}")
42        break

```

```

42     old_xr = xr
43     it += 1
44
45     return results

```

4.4 Newton Raphson Method Implementation

```

1
2 def newton_raphson(f, df, x0, max_it=100, rel_er_thr=None):
3     results = []
4     it = 0
5     while it < max_it:
6         fx0 = f(x0)
7         dfx0 = df(x0)
8         if dfx0 == 0:
9             print("Division by zero encountered")
10            break
11         x1 = x0 - fx0/dfx0
12         ea = abs((x1 - x0)/x1)*100
13         results.append({
14             'iteration': it+1,
15             'm_k': x0,
16             'm_r': x1,
17             'f(m_k)': fx0,
18             "f'(m_k)": dfx0,
19             'ea(%)': ea
20         })
21         if ea is not None and rel_er_thr is not None and ea <
22             rel_er_thr:
23             print(f"Approximate Newton Raphson root found at x =
24                 {x1:.10f} on iteration {it+1} (relative error =
25                 {ea:.2e} <= threshold {rel_er_thr})")
26             break
27         if abs(f(x1)) < 1e-12:
28             print(f"Approximate Newton Raphson root found at x =
29                 {x1:.10f} on iteration {it+1} (|f(m_r)| <
30                 1e-12)")
31             break
32         x0 = x1
33         it += 1
34     return results

```

4.5 Secant Method Implementation

```

1
2 def secant(f, x0, x1, max_iteration=100, rel_er_thr=None):
3     f0=f(x0)
4     f1=f(x1)
5     results = []
6     iteration = 0
7     while iteration < max_iteration:

```

```

8     if abs(f1 - f0) < 1e-12:
9         print("Division by Zero or Subtract Cancellation
10            Encountered.")
11        break
12    x2 = x1 - f1 * (x1 - x0) / (f1 - f0)
13    f2 = f(x2)
14    ea = abs((x2 - x1) / x2) * 100
15    results.append({
16        'iteration': iteration+1,
17        'x_i-1': x0,
18        'x_i': x1,
19        'x_i+1': x2,
20        'f(x_i-1)': f0,
21        'f(x_i)': f1,
22        'f(x_i+1)': f2,
23        'ea(%)': ea
24    })
25
26    if abs(f2) < 1e-12:
27        print(f"Approximate Secant root found at x = {x2:.10f}
28              on iteration {iteration+1} (|f(x)| < 1e-12)")
29        break
30    if ea is not None and rel_er_thr is not None and ea <=
31        rel_er_thr:
32        print(f"Approximate Secant root found at x = {x2:.10f}
33              on iteration {iteration+1} (relative error =
34              {ea:.2e} <= threshold {rel_er_thr})")
35        break
36
37    x0 = x1
38    f0 = f1
39    x1 = x2
40    f1 = f2
41    iteration += 1
42
43    return results

```

4.6 Generic printing function and rest of the code

```

1 def print_table(results, method_name):
2     """Generic function to print tables for any root-finding
3         method"""
4     if not results:
5         print(f"No results to display for {method_name}")
6         return
7
8     columns = [key for key in results[0].keys() if key != 'iteration']
9
10    width = 6 + len(columns) * 12 + (len(columns) - 1)
11
12    print("\n" + "="*width)
13    print(method_name.upper())
14    print("="*width)

```

```

15     header = f"{'Iter':<6} "
16     for col in columns:
17         header += f"{col:<12} "
18     print(header)
19     print("-"*width)
20
21     for r in results:
22         row = f"{'iteration':<6} "
23         for col in columns:
24             value = r[col]
25             if value is None or (col == 'ea(%)' and value is None):
26                 row += f"{'N/A':<12} "
27             elif isinstance(value, (int, float)):
28                 row += f"{value:<12.5f} "
29             else:
30                 row += f"{'str(value):<12} "
31         print(row)
32
33     print("=="*width)
34
35
36
37
38 rel_err = 0.001
39
40 bisection_results = bisection(f, 0.1, 0.4, rel_er_thr=rel_err)
41 print_table(bisection_results, "Bisection Method")
42 falsepos_results = false_position(f, 0.1, 0.4, rel_er_thr=rel_err)
43 print_table(falsepos_results, "False Position Method")
44 newton_results = newton_raphson(f, df, 1.5, rel_er_thr=rel_err)
45 print_table(newton_results, "Newton-Raphson Method")
46 secant_results = secant(f, 1.5, 2.0, rel_er_thr=rel_err)
47 print_table(secant_results, "Secant Method")
48
49 plt.figure(figsize=(10, 6))
50 def plot_errors(results, label):
51     iterations = [r['iteration'] for r in results if r['ea(%)'] is
52                   not None]
53     errors = [r['ea(%)'] for r in results if r['ea(%)'] is not
54               None]
55     plt.plot(iterations, errors, marker='o', label=label)
56
57 plot_errors(bisection_results, 'Bisection')
58 plot_errors(falsepos_results, 'False Position')
59 plot_errors(newton_results, 'Newton-Raphson')
60 plot_errors(secant_results, 'Secant')
61
62 plt.xlabel('Iteration')
63 plt.ylabel('Approximate Error (%)')
64 plt.title('Convergence of Root-Finding Methods')
65 plt.grid(True, alpha=0.3)
66 plt.legend()
67 plt.show()

```

5 Output

5.1 Bisection Method Output

The Bisection method applied with an initial interval of $[0.1, 0.4]$ and a relative error threshold of 0.001% converged to $h = 0.2440238953$ in 17 iterations. The error decreases steadily by approximately half with each iteration.

Bisection approximate root found at x = 0.2440238953 (iter 17)							
=====							
BISECTION METHOD							
Iter	x _l	x _u	x _r	f(x _l)	f(x _u)	f(x _r)	ea(%)
1	0.10000	0.40000	0.25000	1.75715	-1.84235	-0.07189	N/A
2	0.10000	0.25000	0.17500	1.75715	-0.07189	0.83645	42.85714
3	0.17500	0.25000	0.21250	0.83645	-0.07189	0.38059	17.64706
4	0.21250	0.25000	0.23125	0.38059	-0.07189	0.15391	8.10811
5	0.23125	0.25000	0.24063	0.15391	-0.07189	0.04090	3.89610
6	0.24063	0.25000	0.24531	0.04090	-0.07189	-0.01552	1.91083
7	0.24063	0.24531	0.24297	0.04090	-0.01552	0.01268	0.96463
8	0.24297	0.24531	0.24414	0.01268	-0.01552	-0.00142	0.48000
9	0.24297	0.24414	0.24355	0.01268	-0.00142	0.00563	0.24058
10	0.24355	0.24414	0.24385	0.00563	-0.00142	0.00210	0.12014
11	0.24385	0.24414	0.24399	0.00210	-0.00142	0.00034	0.06004
12	0.24399	0.24414	0.24407	0.00034	-0.00142	-0.00054	0.03001
13	0.24399	0.24407	0.24403	0.00034	-0.00054	-0.00010	0.01501
14	0.24399	0.24403	0.24401	0.00034	-0.00010	0.00012	0.00750
15	0.24401	0.24403	0.24402	0.00012	-0.00010	0.00001	0.00375
16	0.24402	0.24403	0.24403	0.00001	-0.00010	-0.00005	0.00188
17	0.24402	0.24403	0.24402	0.00001	-0.00005	-0.00002	0.00094

Bisection Method Output Table

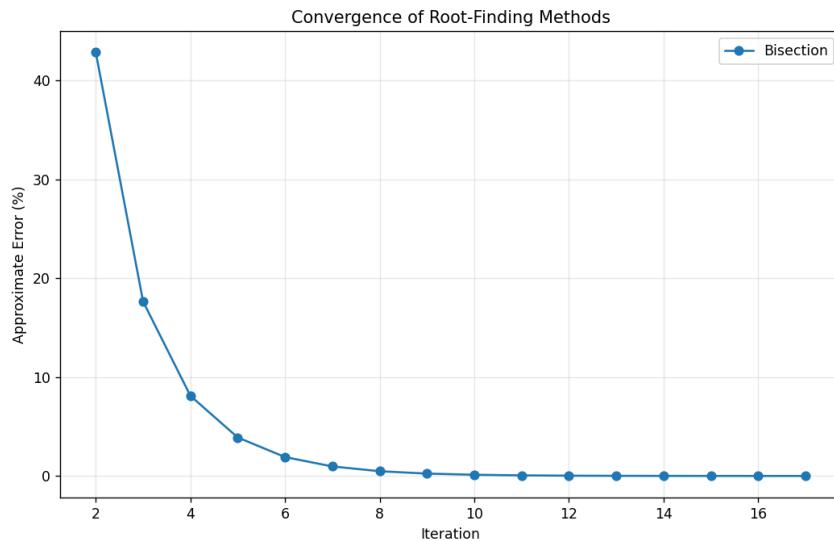


Figure 1: Error vs Iterations (Bisection Method)

5.2 False Position Method Output

The False Position method applied with the same initial interval of $[0.1, 0.4]$ and a relative error threshold of 0.001% converged to $h = 0.2440223280$ in just 4 iterations. This is due to its use of the secant line to make a better estimate of the root, resulting in superlinear convergence, which is typically faster than the linear convergence of Bisection.

False Position Method Output Table

```
False Position approximate root found at x = 0.2440223280 (iter 4)
=====
FALSE POSITION METHOD
=====
Iter   xl      xu      xr      f(xl)    f(xu)    f(xr)    ea(%)
-----
1     0.10000  0.40000  0.24645  1.75715  -1.84235  -0.02920  N/A
2     0.10000  0.24645  0.24406  1.75715  -0.02920  -0.00040  0.98095
3     0.10000  0.24406  0.24402  1.75715  -0.00040  -0.00001  0.01340
4     0.10000  0.24402  0.24402  1.75715  -0.00001  -0.00000  0.00018
=====
```

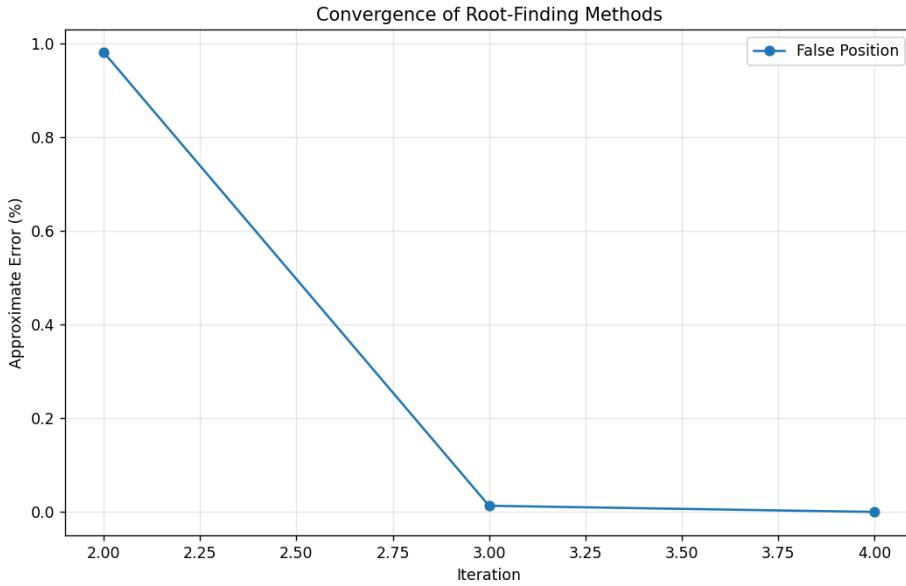


Figure 2: Error vs Iterations (False Position Method)

5.3 Newton-Raphson Method Output

The Newton-Raphson method applied with an initial guess $h_0 = 1.5$ and a relative error threshold of 0.001% converged to $h = 0.2440223219$ in 5 iterations. Despite an initial guess that was not very close to the actual root, the method quickly adjusted. Its quadratic convergence rate is evident after the initial few iterations, where the error decreases drastically in each step. The high initial error is due to the first guess being relatively far from the root compared to the interval-based method.

Newton-Raphson Method Output Table

```
Approximate Newton Raphson root found at x = 0.2440223219 on iteration 5 (relative error = 3.77e-04 <= threshold 0.001)
=====
NEWTON-RAPHSON METHOD
=====
Iter   m_k      m_r      f(m_k)    f'(m_k)    ea(%)
-----
1     1.50000  -1.04195  -11.26317  -4.43092  243.96086
2     -1.04195  0.39216   15.70664  -10.95219  365.69568
3     0.39216   0.24108   -1.75155  -11.59350  62.66830
4     0.24108   0.24402    0.03543  -12.04175  1.20579
5     0.24402   0.24402    0.00001  -12.03421  0.00038
=====
```

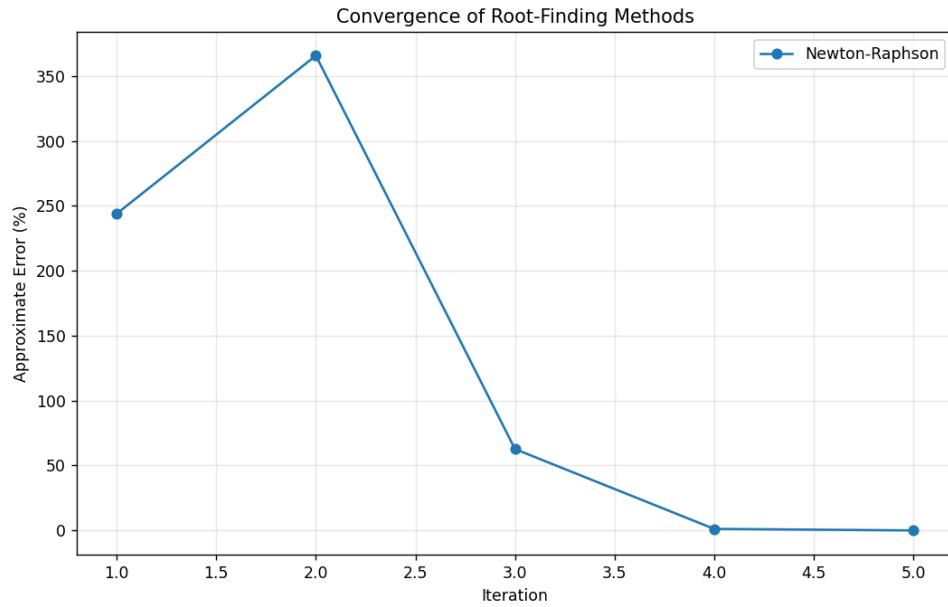


Figure 3: Error vs Iterations (Newton–Raphson Method)

5.4 Secant Method Output

The Secant method applied with initial guesses $h_0 = 1.5$ and $h_1 = 2.0$, and a relative error threshold of 0.001% converged to $h = -4.4744263967$ in 8 iterations (Note: The Secant method converged to a different root compared to the other methods, which is a common occurrence for open methods with different initial guesses or intervals). Its convergence rate is superlinear.

Secant Method Output Table

Approximate Secant root found at $x = -4.4744263967$ on iteration 8 (relative error = $2.40e-04 \leq \text{threshold } 0.001$)							
=====							
SECANT METHOD							
=====							
Iter	x_{i-1}	x_i	x_{i+1}	$f(x_{i-1})$	$f(x_i)$	$f(x_{i+1})$	ea(%)
1	1.50000	2.00000	-4.77520	-11.26317	-12.16060	-8.69786	141.88310
2	2.00000	-4.77520	-21.79342	-12.16060	-8.69786	259857.48459	78.08882
3	-4.77520	-21.79342	-4.77577	-8.69786	259857.48459	-8.71563	356.33363
4	-21.79342	-4.77577	-4.77634	259857.48459	-8.71563	-8.73344	0.01195
5	-4.77577	-4.77634	-4.49644	-8.71563	-8.73344	-0.59009	6.22492
6	-4.77634	-4.49644	-4.47615	-8.73344	-0.59009	-0.04605	0.45312
7	-4.49644	-4.47615	-4.47444	-0.59009	-0.04605	-0.00029	0.03837
8	-4.47615	-4.47444	-4.47443	-0.04605	-0.00029	-0.00000	0.00024

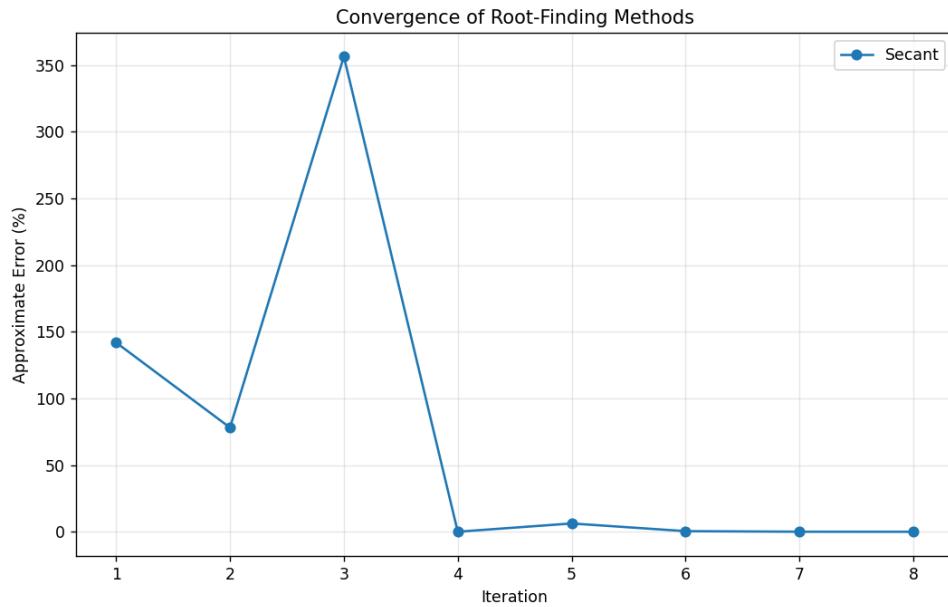


Figure 4: Error vs Iterations (Secant Method)

5.5 Comparative Visualization of Convergence

The following graph depicts the approximate relative error (ϵ_a) versus the number of iterations for all four methods, providing a visual comparison of their convergence rates.

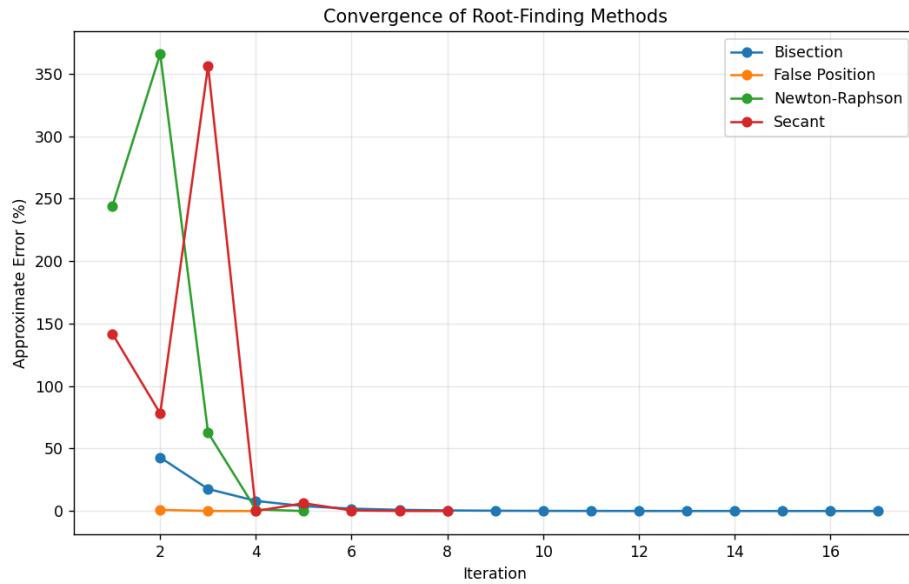


Figure 5: Convergence Comparison of Root Finding Methods

5.5.1 Rationale and Discussion of Convergence

The convergence plot clearly illustrates the efficiency of each method:

- **Bisection Method:** Shows the slowest convergence since its error reduction is linear, meaning the number of correct decimal places increases by a fixed amount with each iteration. We can also see that the error rate at each iteration is only halved, so it takes a high number of (17) iterations to reach the desired tolerance rate of 0.001. This method is guaranteed to converge but is not the most efficient.
- **False Position Method:** This bracketing method demonstrates fastest convergence needing only four iterations to reach the result. The main reason for this is that it gets very close to the root in the first step. Because of that, in the next few steps, one side stays close to the root, which helps the method reach a small error quickly.
- **Newton-Raphson Method:** Exhibits the second best convergence. After the initial few iterations, its quadratic convergence is evident, meaning the number of correct decimal places roughly doubles with each iteration. This method is highly efficient but requires the derivative of the function and a good initial guess to ensure convergence.
- **Secant Method:** The Secant Method performs moderately, needing 8 iterations to converge. It avoids derivative calculations and works faster than the bisection method, but, like other open methods, it can diverge. In this case, it even finds a different root. Because it's not globally stable, poor starting points or flat slopes can cause large jumps - for example, the root suddenly shifts to -21.79342 in the second iteration, with a huge increase in error.

It's important to note that the performance of open methods (Newton-Raphson, Secant) heavily depends on the initial guess, and they may diverge if the guess is poor. Bracketing methods (Bisection, False Position) are more robust as they always converge if a root is present in the initial interval.

6 Summary

In this experiment, we compared four root-finding methods—Bisection, False Position, Newton-Raphson, and Secant—to solve $f(h) = h^3 - 10h + 5e^{-h/2} - 2 = 0$.

All methods successfully found a root of the function. Among them, the False Position method showed the fastest convergence, reaching the solution in just a few(4) iterations using linear interpolation. Newton-Raphson followed closely due to its quadratic convergence in just 5 iterations, performing efficiently but requiring a more complex derivative calculation. The Secant method showed moderate performance (required 8 iterations), with a noticeable jump in the second iteration that affected its stability but it found different root because of initial guess. The Bisection method was the slowest needing the most iterations(17) but it remained the simplest and most reliable in terms of convergence.

Overall, while the Newton-Raphson method is typically the most efficient for well-behaved functions, the False Position method performed best for this particular case. The experiment highlighted a key trade-off: bracketing methods are robust but slow, while open methods are fast but sensitive, may fail due to poor initial guess or irregular behavior of derivative. This highlights that the efficiency of a numerical method depends on the behavior of the function, and choosing the right method requires understanding the function's nature.