

# Computational Homework9 Report

**Student:** 纪浩正, [jihz2023@mail.sustech.edu.cn](mailto:jihz2023@mail.sustech.edu.cn)

## 1 Bisection Method

### 1.1 Introduction

The Bisection Method is a numerical technique for finding a root of a continuous function  $f(x) = 0$  in a given interval  $[a, b]$ , provided that  $f(a)f(b) < 0$ . This guarantees that there is at least one root between  $a$  and  $b$  by the Intermediate Value Theorem.

In the implemented Python class `BisectionMethod`, the method works as follows:

1. Compute the midpoint of the interval:

$$m = \frac{a + b}{2}$$

and store it in `self.points` for tracking iterations.

2. Check if the interval is sufficiently small:

$$|b - a| < 2 \cdot \text{tol}$$

3. Determine in which subinterval the root lies:

if  $f(a)f(m) < 0$ , then the root is in  $[a, m]$  else if  $f(m)f(b) < 0$ , then the root is in  $[m, b]$

4. Repeat the above steps recursively until the interval satisfies the tolerance condition.

```
class BisectionMethod:
    def __init__(self, f, tol):
        self.f=f
        self.tol=tol
        self.points=[]
        self.roots=[]
```

```

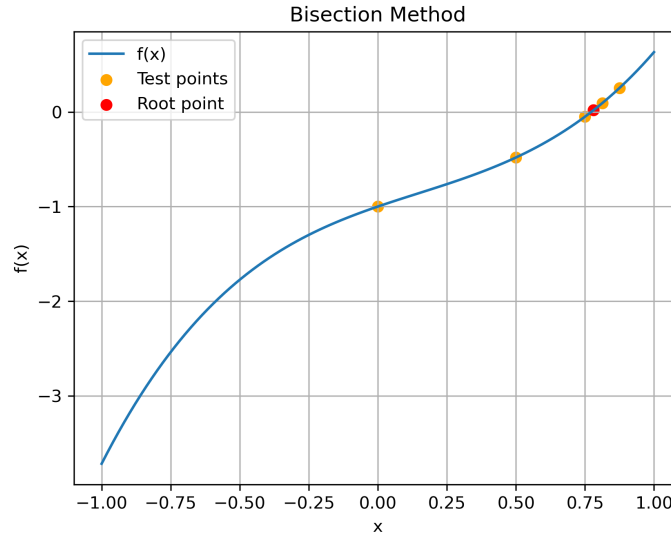
def _bi(self,a,b):
    m=(a+b)/2
    self.points.append(m)
    if (b-a)<2*self.tol:
        if self.f(a)*self.f(b)<0:
            self.roots.append(m)
            return m
        else:
            raise ValueError("please change the interval")
    else:
        if abs(self.f(m))<self.tol:
            self.roots.append(m)
            return m
        elif self.f(a)*self.f(m)<0:
            return self._bi(a,m)
        elif self.f(m)*self.f(b)<0:
            return self._bi(m,b)
        else:
            raise ValueError("please change the interval")

def bisection(self,a,b):
    self.points=[]
    self.roots=[]
    if self.f(a)==0:
        self.roots.append(a)
    if self.f(b)==0:
        self.roots.append(b)
    if self.f(a)*self.f(b)<0:
        self._bi(a,b)
    return np.array(sorted(self.roots)), np.array(self.points)
else:
    raise ValueError("please change the interval")

```

## 1.2 Result

The following table summarizes the iterations of the Bisection Method applied to  $f(x) = x^3 - e^{-x}$ :



**Figure 1** Visualization of the Bisection Method iterations, showing the convergence of test points to the final root.

Iteration	$x_n$	$f(x_n)$
0	0.000000	-1.000000e+00
1	0.500000	-4.815307e-01
2	0.750000	-5.049155e-02
3	0.875000	2.530599e-01
4	0.812500	9.262964e-02
5	0.781250	1.900380e-02
<b>Final</b>	<b>0.765625</b>	<b>-1.624787e-02</b>

**Table 1** Iteration results for the Bisection Method. The final approximation of the root is  $x \approx 0.765625$  with  $f(x) \approx -1.624787 \times 10^{-2}$ .

```
def bisection_method():
    def f(x):
        return x**3-np.exp(-x)
    S=BisectionMethod(f,0.02)
    r,n =S.bisection(-1,1)
    print("##### Bisection Method #####")
    for count in range(len(n)):
```

```
print(f"Iteration {count}: x = {n[count]:.6f}, f(x)={f(n[count]):.6e}")
print(f"Final fixed point:{r[0]:.6f} with value {f(r[0]):.6e}")
```

## 2 Fixed-point Method

### 2.1 Introduction

The Fixed-point Method is an iterative technique used to find a solution  $x^*$  to an equation  $f(x) = 0$ . In this homework, the original equation is:

$$f(x) = x^3 - e^{-x} = 0.$$

To apply the Fixed-point Method, we rewrite it in the form:

$$x = f(x) + x = g(x),$$

so that the iteration can be written as:

$$x_{n+1} = \varphi(x_n) = x_n + f(x_n).$$

For better convergence, we can transform the original equation into a different fixed-point form. For example:

$$x^3 - e^{-x} = 0 \quad \text{can be rewritten as} \quad e^{-x/3} - x = 0.$$

In the iteration, we can define:

$$f(x) = e^{-x/3} - x, \quad f'(x) = -\frac{1}{3}e^{-x/3} - 1.$$

A critical requirement for convergence is the Lipschitz (Lei) condition:

$$|f'(x) + 1| < 1,$$

which ensures that the iterative process will converge. If this condition is not satisfied, the iteration may diverge. By choosing a suitable transformation  $\varphi(x)$ , we can reduce  $|\varphi'(x)|$  near the root and satisfy the Lipschitz condition.

```
class FixedPointMethod():
    def __init__(self, f, f1, x0, tol):
```

```

self.f=f
self.f1=f1
self.x0=x0
self.tol=tol
self.points=[]

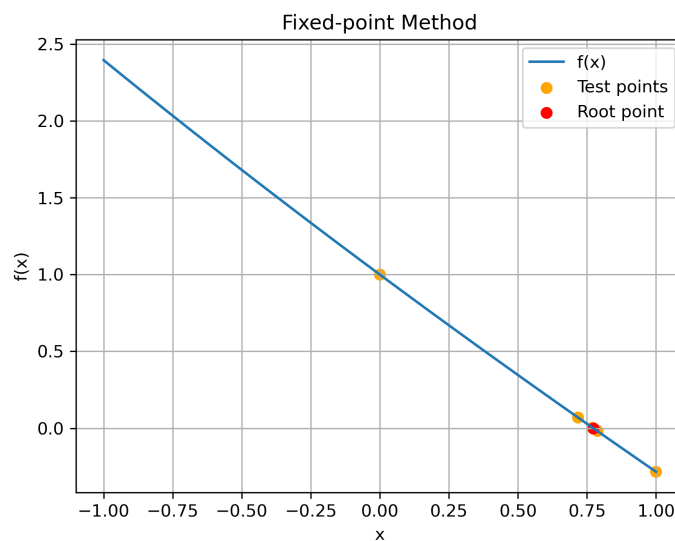
def _fixed_point(self,x0,count=0):
    print(f"Iteration {count}:x = {x0:.6f}, f(x0)={self.f(x0):.6e}")
    if abs((self.f1(x0)+1))>1:
        raise ValueError("enter the correct condition to satisfy the lipschitz condition")
    x_new=x0+self.f(x0)
    self.points.append(x_new)
    if abs(x_new-x0)<self.tol:
        return x_new
    else:
        return self._fixed_point(x_new,count+1)

def fixed_point(self):
    x0=self.x0
    self.points=[]
    return self._fixed_point(x0,np.array(sorted(self.points)))

```

## 2.2 Result

The iteration results of the Fixed-point Method are summarized in Table 2. The final fixed point is  $x^* = 0.773854$  with function value  $f(x^*) = -1.221010 \times 10^{-3}$ .



**Figure 2** Visualization of the Fixed-point Method iterations, showing the convergence of the iteration points towards the final root.

**Table 2 Fixed-point Method Iteration Results**

Iteration	$x_n$	$f(x_n)$
0	100.000000	-1.000000e+02
1	0.000000	1.000000e+00
2	1.000000	-2.834687e-01
3	0.716531	7.100660e-02
4	0.787538	-1.842126e-02
5	0.769117	4.737230e-03
<b>Final</b>	0.773854	-1.221010e-03

```
def fixed_point_method():
    def f(x):
        return np.exp(-x/3)-x
    def f1(x):
        return -1/3*np.exp(-x/3)-1
    x0,tol=100,1e-2
    print("\n##### Fixed-point Method #####")
    S=FixedPointMethod(f,f1,x0,tol)
    r,n=S.fixed_point()
    print(f"Final fixed point:{r:.6f} with value {f(r):.6e}")
```

### 3 Newton Method

#### 3.1 Introduction

The Newton Method is an iterative technique to find a root  $x^*$  of a function  $f(x) = 0$ . The method uses both the function and its derivative  $f'(x)$  to iteratively improve the approximation of the root.

Starting from an initial guess  $x_0$ , the iteration formula is:

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

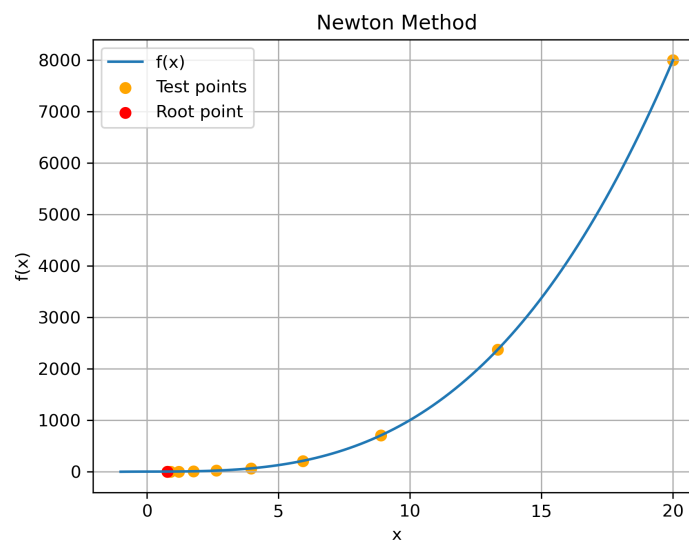
In the provided Python implementation (Newton\_method), all intermediate points are stored in the points list for visualization and analysis. The convergence check is based on the change of successive iterates:

$$|x_{n+1} - x_n| < \text{tol}.$$

```
def Newton_method(f,f1,x0,tol,points=None,count=0):
    if points is None:
        points = []
    points.append(x0)
    print(f"Iteration {count}:x = {x0:.6f}, f(x)={f(x0):.6e}")

    x_new=x0-f(x0)/f1(x0)
    if abs(x_new-x0)<tol:
        points.append(x_new)
        print(f"Final fixed point: {x_new:.6f}, f(x)={f(x_new):.6e}")
        return x_new,np.array(points)
    else:
        return Newton_method(f,f1,x_new,tol,points,count+1)
```

## 3.2 Result



**Figure 3** Convergence of the Newton Method. The orange points indicate the iteration steps, and the red point is the final root.

**Table 3** Iteration results for the Newton Method. The final approximation of the root is  $x \approx 0.772883$  with  $f(x) \approx 3.759 \times 10^{-8}$ .

Iteration	$x_n$	$f(x_n)$
0	20.000000	$8.000000 \times 10^3$
1	13.333333	$2.370370 \times 10^3$
2	8.888889	$7.023318 \times 10^2$
3	5.925928	$2.080959 \times 10^2$
4	3.950694	$6.164313 \times 10^1$
5	2.634748	$1.821841 \times 10^1$
6	1.762946	5.307662
7	1.203979	1.445248
8	0.893086	$3.029361 \times 10^{-1}$
9	0.784979	$2.756830 \times 10^{-2}$
10	0.773017	$3.024780 \times 10^{-4}$
<b>Final</b>	<b>0.772883</b>	$3.759421 \times 10^{-8}$

```
def newton_method():
    def f(x):
        return x**3-np.exp(-x)
    def f1(x):
        return 3*x**2+np.exp(-x)

    x0,tol=20,1e-2
    print("\n##### Newton Method #####")
    r,n=Newton_method(f,f1,x0,tol)
```

## 4 Secant Method

### 4.1 Introduction

The Secant Method is an iterative root-finding algorithm that uses two initial guesses  $x_0$  and  $x_1$  to approximate a root of  $f(x) = 0$ . Unlike the Newton Method, it does not require the derivative of  $f(x)$ , and instead approximates it using the slope of the line connecting  $(x_0, f(x_0))$  and  $(x_1, f(x_1))$ :

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

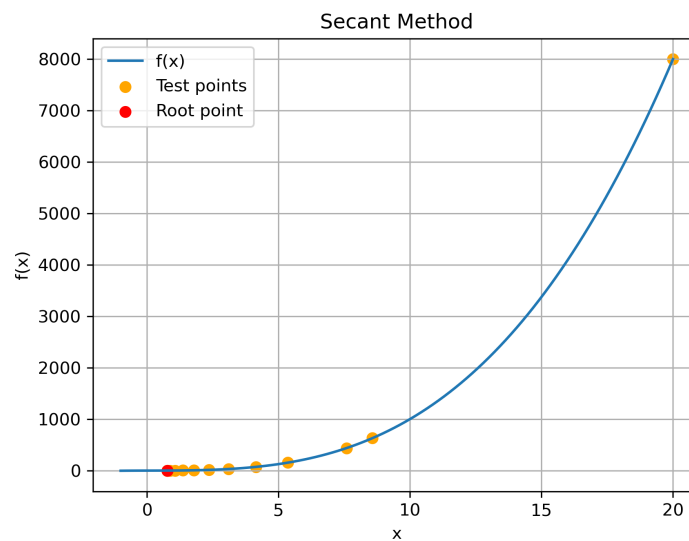


The iteration continues until convergence is reached, i.e.,  $|x_{n+1} - x_n| < \text{tol}$ .

```
def Secant_method(f,x0,x1,tol,points=None,count=0):
    if points is None:
        points=[]
    points.append(x1)
    print(f"Interaction {count}:x = {x0:.6f}, f(x)={f(x0):.6e}")

    x_new=x1-(x1-x0)*f(x1)/(f(x1)-f(x0))
    if abs(x_new-x1)<tol:
        points.append(x_new)
        print(f"Final fixed point: {x_new:.6f}, f(x)={f(x_new):.6e}")
        return x_new,np.array(points)
    else:
        return Secant_method(f,x1,x_new,tol,points,count+1)
```

## 4.2 Result



**Figure 4** Convergence of the Secant Method. Orange points show iteration steps and the red point indicates the final root.

**Table 4** Iteration results for the Secant Method. The final approximation of the root is  $x \approx 0.772937$  with  $f(x) \approx 1.215 \times 10^{-4}$ .

Iteration	$x_n$	$f(x_n)$
0	10.000000	$1.000000 \times 10^3$
1	20.000000	$8.000000 \times 10^3$
2	8.571429	$6.297374 \times 10^2$
3	7.594937	$4.380988 \times 10^2$
4	5.362612	$1.542112 \times 10^2$
5	4.149986	$7.145688 \times 10^1$
6	3.102905	$2.982991 \times 10^1$
7	2.352566	$1.292531 \times 10^1$
8	1.778855	5.460040
9	1.359246	2.254423
10	1.064148	$8.600296 \times 10^{-1}$
11	0.882137	$2.725524 \times 10^{-1}$
12	0.797696	$5.722401 \times 10^{-2}$
<b>Final</b>	<b>0.772937</b>	$1.215491 \times 10^{-4}$