

Samir Banjara 09/18/23

Question 1: Program Newton's Method and apply it to the function  $f(x)=x^3+4x^2-10$ . This function has a root in  $[1,2]$

In [ ...

```
import numpy as np
import pandas as pd

def fixed_point (p0,e, max_it,g):
    p = np.zeros(max_it)
    p[0] = p0
    i = 1
    while i < max_it:
        try:
            p[i] = g(p[i-1])
        except:
            print('Arithmetic error')
            return(p)
        if abs(p[i] - p[i-1]) <= e:
            return(p)

        i += 1

    print('max number of iteration exceeded')
    return(p)

def newton_bisection(a, b, e, max_it, f):
    FA = f(a)
    i = 0
    p = np.zeros(max_it)
    while i< max_it:
        p[i] = (a + b) / 2
        FP = f(p[i])
        if (abs(b - a) <= e/2):
            return p
        if FA * FP < 0:
            b = p[i]
        else:
            a = p[i]
            FA = FP
        i += 1
    return(p)
    print('Warning. Max Iter Reached!')

def f(x):
    y = x**3 + 4 * x**2 - 10
    return y

def g1(x):
    y = x - (x**3 + 4 * x**2 - 10)
    return y
```

```

def g2(x):
    y = np.sqrt(10/x - 4 * x)
    return y

def g3(x):
    y = (1/2) * np.sqrt(10 - x**3)
    return y

def g4(x):
    y = np.sqrt(10/(4+x))
    return y

def g5(x):
    y = x - (x**3 + 4 * x**2 - 10)/(3 * x**2 + 8 * x)
    return y

```

```

-----
ModuleNotFoundError                                Traceback (most recent call last)
/Users/samirbanjara/Downloads/math625_assignment_1_Samir_Banjara.ipynb Cell
3 line 1
----> <a href='vscode-notebook-cell:/Users/samirbanjara/Downloads/math625_as
signment_1_Samir_Banjara.ipynb#W2sZmlsZQ%3D%3D?line=0'>1</a> import numpy as
np
      <a href='vscode-notebook-cell:/Users/samirbanjara/Downloads/math625_as
signment_1_Samir_Banjara.ipynb#W2sZmlsZQ%3D%3D?line=1'>2</a> import pandas a
s pd
      <a href='vscode-notebook-cell:/Users/samirbanjara/Downloads/math625_as
signment_1_Samir_Banjara.ipynb#W2sZmlsZQ%3D%3D?line=3'>4</a> def fixed_point
(p0,e, max_it,g):
ModuleNotFoundError: No module named 'numpy'

```

```

In [ ... p0 = 1.35
max_it = 30
e = 1e-8
a = 1
b = 2

p1 = fixed_point(p0, e, max_it, g1)
p2 = fixed_point(p0, e, max_it, g2)
p3 = fixed_point(p0, e, max_it, g3)
p4 = fixed_point(p0, e, max_it, g4)
p5 = fixed_point(p0, e, max_it, g5)
pn = newton_bisection(a, b, e, max_it, f)

```

```

max number of iteration exceeded
max number of iteration exceeded
max number of iteration exceeded
max number of iteration exceeded
max number of iteration exceeded

```

```

In [ ... np.array([p1, p2, p3, p4, p5, pn]).reshape(6,max_it).transpose()

```

```
Out[ ... array([[1.35      , 1.35      , 1.35      , 1.35      , 1.35      ,
1.5       ],
[1.599625 , 1.41683006, 1.37291888, 1.36717185, 1.36534501,
1.25      ],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.375     ],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.3125    ],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.34375   ],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.359375  ],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.3671875 ],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36328125],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36523438],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36425781],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36474609],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36499023],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.3651123  ],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36517334],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36520386],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36521912],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36522675],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36523056],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36522865],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36522961],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36523008],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36522985],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36522996],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36523002],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36522999],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36523001],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36523002],
[0.       , 0.       , 0.       , 0.       , 0.       ,
1.36523001],
```

```

[0.          , 0.          , 0.          , 0.          , 0.          ,
 1.36523001],
[0.          , 0.          , 0.          , 0.          , 0.          ,
 0.          ]]

```

In [ ... `pd.DataFrame(np.array([p1,p2,p3,p4,p5,pn]).reshape(6,max_it).transpose(), co`

Out[ ...

		g1	g2	g3
0	1.3500000000000000	1.3500000000000000	1.3500000000000000	1.3500000000000000
1	1.5996249999999998	1.416830055937340	1.372918879613796	1.3670000000000000
2	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
3	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
4	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
5	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
6	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
7	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
8	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
9	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
10	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
11	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
12	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
13	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
14	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
15	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
16	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
17	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
18	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
19	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
20	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
21	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
22	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
23	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
24	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
25	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000

	g1	g2	g3
26	0.0000000000000000	0.0000000000000000	0.0000000000000000
27	0.0000000000000000	0.0000000000000000	0.0000000000000000
28	0.0000000000000000	0.0000000000000000	0.0000000000000000

```
In [ ... ## Using a Pandas data frame, we can look at the convergence.
all_ps = {'p1':p1, 'p2':p2, 'p3':p3, 'p4':p4, 'p5':p5, 'pn':pn}
#print(all_ps)

pd.DataFrame (dict([(k, pd.Series(v)) for k, v in all_ps.items()]])
```

Out[ ...

		p1	p2	p3
0	1.3500000000000000	1.3500000000000000	1.3500000000000000	1.3500000000000000
1	1.5996249999999998	1.416830055937340	1.372918879613796	1.3670000000000000
2	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
3	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
4	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
5	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
6	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
7	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
8	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
9	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
10	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
11	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
12	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
13	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
14	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
15	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
16	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
17	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
18	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
19	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
20	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
21	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
22	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
23	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
24	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000
25	0.0000000000000000	0.0000000000000000	0.0000000000000000	0.0000000000000000

	p1	p2	p3
26	0.0000000000000000	0.0000000000000000	0.0000000000000000 0.000
27	0.0000000000000000	0.0000000000000000	0.0000000000000000 0.000
28	0.0000000000000000	0.0000000000000000	0.0000000000000000 0.000

In [ ...

```

## Testing Newton's Method
def g1(x):
    y = np.cos(x)
    return y

def g2(x):
    y = x + (np.cos(x) - x)/(np.sin(x) + 1)
    return y

p0 = np.pi / 4

max_it = 10
e = 1e-8

p1 = fixed_point(p0, e, max_it, g1)
p2 = fixed_point(p0, e, max_it, g1)

all_ps = {'p1':p1, 'p2':p2}
D = pd.DataFrame(dict([(k,pd.Series(v)) for k, v in all_ps.items()]))
pd.options.display.float_format = '{:,.15f}'.format
print(D)

```

```

max number of iteration exceeded
max number of iteration exceeded

           p1           p2
0 0.785398163397448 0.785398163397448
1 0.707106781186548 0.707106781186548
2 0.000000000000000 0.000000000000000
3 0.000000000000000 0.000000000000000
4 0.000000000000000 0.000000000000000
5 0.000000000000000 0.000000000000000
6 0.000000000000000 0.000000000000000
7 0.000000000000000 0.000000000000000
8 0.000000000000000 0.000000000000000
9 0.000000000000000 0.000000000000000

```

In [ ...

```

## Multiple Roots
def fixed_point(p0, e, max_it, g):
    p = []
    p.append(p0)
    i = 1
    while i <= max_it:
        try:
            p.append(g(p0))
        except:
            print('Arithmetic error')
            return(p)

```



```

    if abs(p[i] - p0) <= e:
        return(p)
    p0 = p[i]
    i += 1

print('max number of iteration exceeded')
return(p)

```

Question 2: Use the `matplotlib` library to plot the function. Place a red dot on the figure where the estimated root is. Make sure that the xxx and yyy axes are visible.

```

In [ ... ] import matplotlib.pyplot as plt

# Get the estimated root using the newton_bisection method
pn = newton_bisection(a, b, e, max_it, f)
estimated_root = pn[0]

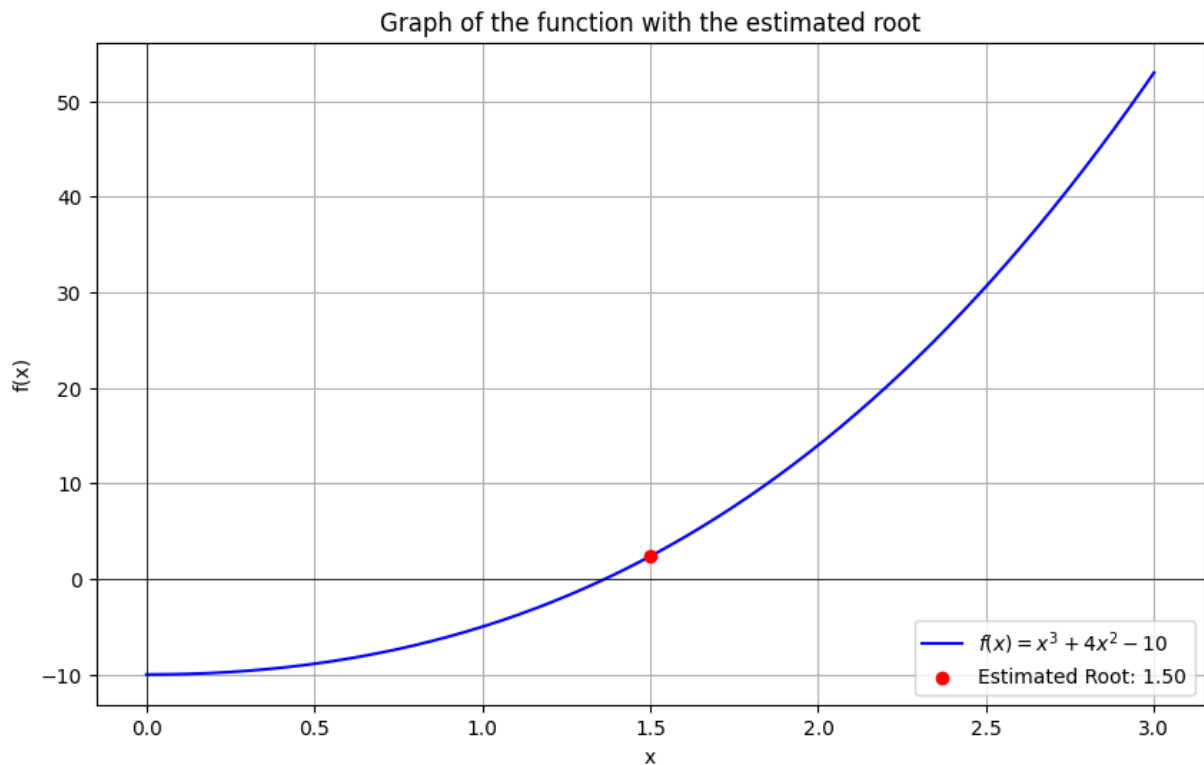
```

```

In [ ... ] x_values = np.linspace(0, 3, 400)
           y_values = f(x_values)

# Plotting the function and the estimated root on the same graph
plt.figure(figsize=(10, 6))
plt.plot(x_values, y_values, label=r'$f(x) = x^3 + 4x^2 - 10$', color='blue')
plt.scatter(estimated_root, f(estimated_root), color='red', zorder=5, label=)
plt.axhline(0, color='black', linewidth=0.5)
plt.axvline(0, color='black', linewidth=0.5)
plt.title("Graph of the function with the estimated root")
plt.xlabel("x")
plt.ylabel("f(x)")
plt.grid(True)
plt.legend()
plt.show()

```



Question 3: Show that for a fixed point iteration procedure  $p_{n+1} = g(p_n)$  with  $|g'(x)| \leq K < 1$ ,  $|p_n - p| \leq K^n |p_1 - p_0|$

**Proof:**

Given a fixed point iteration procedure defined by  $p_{n+1} = g(p_n)$  we aim to demonstrate that if  $|g'(x)| \leq K < 1$  for all  $x$ , then  $|p_n - p| \leq K^n |p_1 - p_0|$  for all  $n$ .

We will employ Mathematical Induction.

Consider the error iteration formula for fixed point iteration:  $e_{n+1} = |p_{n+1} - p|$ ,  $e_n = |p_n - p|$

By invoking the Mean Value Theorem, which states:

If a function  $(f)$  is continuous on the closed interval  $([a, b])$  and differentiable on the open interval  $((a, b))$ , then there exists at least one number  $(c)$  in the open interval  $((a, b))$  such that:  $f'(c) = \frac{f(b)-f(a)}{b-a}$

there exists a number  $(c)$  between  $(p)$  and  $(p_n)$  such that:  $g'(c) = \frac{g(p_n)-g(p)}{p_n-p}$

Rearranging, we obtain:  $g'(c)(p_n-p) = g(p_n) - g(p)$

Given  $(p_{n+1} = g(p_n))$  and  $(p = g(p))$  (since  $(p)$  is a fixed point), substituting into the above equation yields:  $g'(c)(p_{n+1}-p) = g(p_{n+1}) - g(p)$

Taking the absolute value, we get:  $|p_{n+1}-p| = |g'(c)| \times |p_n-p|$  or equivalently,  $e_{n+1} = |g'(c)| \times e_n$

Given the condition  $(|g'(x)| \leq K)$ , it follows that:  $e_{n+1} \leq K \times e_n$

**Base Case:** For  $(n = 1)$ :  $e_2 \leq K \times e_1$

This is validated by the previous equation.

### Inductive Step:

Assuming the inequality holds for  $(n = k)$ :  $e_{k+1} \leq K \times e_k$

From our derived inequality, we infer:  $e_{k+2} \leq K \times e_{k+1}$

Substituting our inductive assumption into this gives:

$$e_{k+2} \leq K \times K^{k+1-K} \times e_1 e_{k+2} \leq K \times K^{k+1-K} \times e_1 e_{k+2} \leq K \times \frac{K^k}{1-K} \times e_1 \text{ or} \\ e_{k+2} \leq K^{k+1-K} \times e_1 e_{k+2} \leq K^{k+1-K} \times e_1 e_{k+2} \leq \frac{K^{k+1}}{1-K} \times e_1$$

By induction, this inequality stands for all ( n ).

Lastly, given (  $e_1 = |p_1 - p_0|$  ), we deduce:

$$e_{n+1} \leq K^{n+1-K} \times |p_1 - p_0| e_{n+1} \leq K^{n+1-K} \times |p_1 - p_0| e_{n+1} \leq \frac{K^n}{1-K} \times |p_1 - p_0|$$

This completes the proof of the given inequality.