

# HW3

September 25, 2020

## Numerical Computation

**Andrew Pickner** I worked alone on this assignment.

### 1 Compute the derivative of $f(x)$

$$f(x) = \frac{1}{1+e^x} - \frac{1}{2}$$

My differentiation skills are not ideal when it comes to anything that isn't a polynomial, but luckily I know some great online resources.

Initially, I wanted to see what  $f(x)$  looked like so I checked out Desmos. This function seemed relatively easy to visualize  $f'(x)$ .

Wolfram Alpha and Symbolab both gave me  $f'(x) = -\frac{e^x}{(1+e^x)^2}$ , which I then plugged back into Desmos and all looked correct! Plus, I confirmed the real root  $r$  is 0.

### 2 Implement Newton's method

So obviously class and Sauer's textbook were helpful in developing pseudocode, I found this [website](#) to have a nice and concise write-up and I liked the code style they used so I modeled mine after theirs. When thinking about the 'interface', I chose to add a parameter for  $f'(x)$  as opposed to using scipy to differentiate  $f(x)$  because this seemed a little hacky

```
[22]: # it might be helpful to have tolerance as a parameter
def newtons_method(f,Df,x0,epsilon,max_iter):
    xn = x0
    for n in range(0,max_iter):
        fxn = f(xn)
        if abs(fxn) < epsilon:
            # print('Found solution after',n,'iterations.')
            return xn
        Dfxn = Df(xn)
        if Dfxn == 0:
            # print('Zero derivative. No solution found.')
            return None
```

```

        xn = xn - fxn/Dfxn
#     print('Exceeded maximum iterations. No solution found.')
    return None

import numpy as np

f      = lambda x: (((1) / (1 + np.exp(x))) - (1/2))
f_prime = lambda x: ((- np.exp(x)) / ((1 + np.exp(x)) ** 2))

print(newtons_method(f, f_prime, 0.5, 0.000000001, 20))

```

1.5516754209104165e-16

### 3 Choose an interval around $r=0$ and check whether the points converge or not

```

[36]: import numpy as np

# function we are testing and its derivative
f      = lambda x: (((1) / (1 + np.exp(x))) - (1/2))
f_prime = lambda x: ((- np.exp(x)) / ((1 + np.exp(x)) ** 2))

# change this number to play with the interval...
boundry = 0.177

grid_num = 2000
for i in range(0, 5):
    xs = np.linspace(-boundry - i, boundry + i, num=grid_num)
    print("[{}, {}".format(-boundry - i, boundry + i))
    count = 0
    for each in xs:
        ret = newtons_method(f, f_prime, each, 0.00001, 25)
        if ret != None:
            count += 1
    print("{}%".format((count / grid_num) * 100))

```

[-0.177, 0.177]

100.0%

[-1.177, 1.177]

100.0%

[-2.177, 2.177]

100.0%

[-3.177, 3.177]

/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:13: RuntimeWarning:  
overflow encountered in double\_scalars

```

del sys.path[0]
/usr/local/lib/python3.7/site-packages/ipykernel_launcher.py:5: RuntimeWarning:
invalid value encountered in double_scalars
"""
/usr/local/lib/python3.7/site-packages/ipykernel_launcher.py:4: RuntimeWarning:
overflow encountered in exp
after removing the cwd from sys.path.
/usr/local/lib/python3.7/site-packages/ipykernel_launcher.py:5: RuntimeWarning:
overflow encountered in exp
"""
/usr/local/lib/python3.7/site-packages/ipykernel_launcher.py:5: RuntimeWarning:
overflow encountered in double_scalars
"""

68.5%
[-4.177, 4.177]
52.1%

```

I didn't do a very elegant experiment to find the interval on which everything converges. However, by screwing around with the variable `boundry`, I was able to see that 100% of points on the interval `[-2.177, 2.177]` converge and only 99.9% converge when we make the interval: `[-2.178, 2.178]`.

## 4 Report the interval for which all initial guesses converge

```
[-2.177, 2.177]
```

This is the interval I found to be completely convergent in my little experiment above.

## 5 Plot the function and the interval on which all guesses converge

```

[53]: %matplotlib inline
import matplotlib.pyplot as plt
plt.style.use('seaborn-whitegrid')

n_left = -5
n_right = 5
xs = np.linspace(n_left, n_right, num=8000)

fig = plt.figure()
ax1 = plt.axes()

ax1.set_xlabel("x", fontsize=12)
ax1.set_ylabel("$f(x)$", fontsize=12)
ax1.set_title("$\\bf{f(x)}$ and the interval of convergence$", fontsize=12)

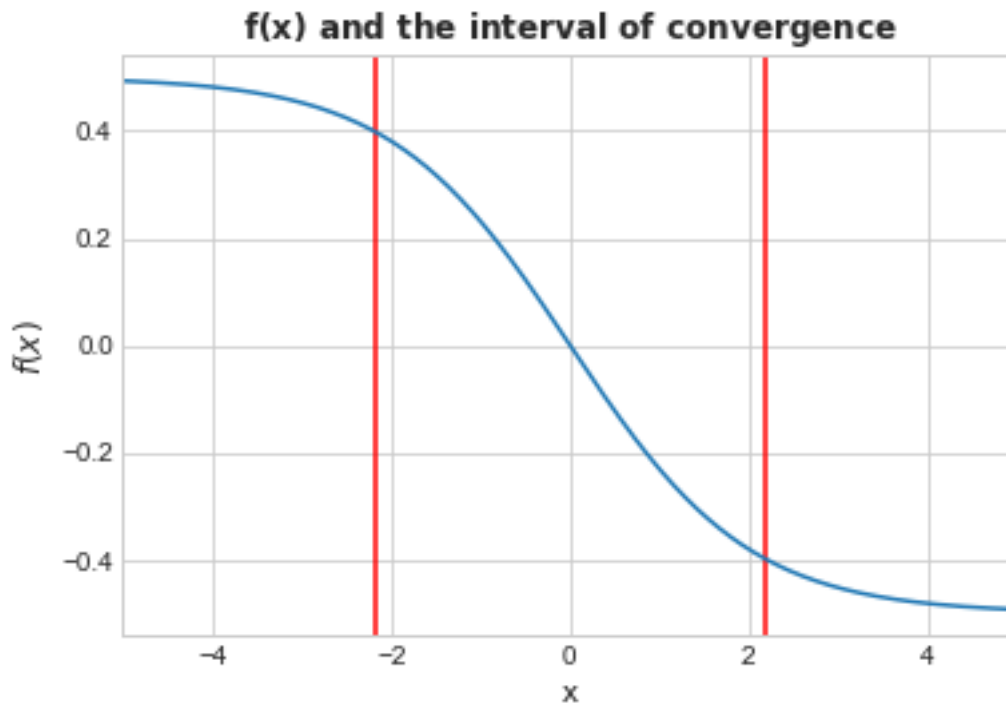
plt.xlim([n_left, n_right])

```

```
plt.axvline(x=-2.177, color="red")
plt.axvline(x=2.177, color="red")

plt.plot(xs, f(xs))
```

[53]: [<matplotlib.lines.Line2D at 0x1116b53d0>]



So I'm not entirely sure why the function converges on the interval it does, it seems odd that it only takes a change of 0.001 to make to function diverge.

[ ]: