

# Homework 4

Ramaseshan Parthasarathy, Saurabh Prasad

## Problem 1

1. The method does converge to  $f(x) = 0$ :

Assume the method converges, knowing  $\lim_{k \rightarrow \infty} x_k = a$ , and  $f'(a) \neq 0$ :

$$a = \frac{a + a - \frac{f(a)}{f'(a)}}{2} \Rightarrow 2a = 2a - \frac{f(a)}{f'(a)} \Rightarrow \frac{f(a)}{f'(a)} = 0$$

$\therefore$  The method converges to  $f(x) = 0$

2. This method does converge under the same conditions as Newton's method:

$$g(x) = x - \frac{f(x)}{2f'(x)} \Rightarrow g'(x) = 1 - \frac{f'^2(x) - f(x)f''(x)}{2f'^2(x)}$$

Three key assumptions were made in class notes with regards to Newton's Method:

(1)  $f(a) = 0$  (i.e. it is a solution to  $f(x) = 0$ )

(2)  $f'(a) \neq 0$ , and

(3)  $f''(a)$  is bounded near  $a$  (i.e.  $f''$  is continuous)

Using the above assumptions, we arrive at the following:  $\lim_{x \rightarrow a} g'(x) = \frac{1}{2}$

$\therefore g(x)$  must converge given it's a contraction on  $(a - \delta, a + \delta)$

3. The order of convergence is 1:

$$\text{From the class notes, } g(x_k) = g(a) + g'(a)(x_k - a) + \frac{g''(c)}{2}(x_k - a)^2$$

$$\text{This can be rewritten as } |e_{k+1}| = \frac{1}{2}|e_k| + \frac{g''(c)}{2}|e_k|^2, |e_{k+1}| \leq L|e_k|$$

Moreover, since  $|e_{k+1}| \leq L|e_k|^d$ , where  $d = 2$ , we have a first order convergence!

## Problem 2

1. Analysis of convergence properties:

$$|g'_1(x)| = \frac{|2x|}{3} = \frac{4}{3} > 1 \text{ (divergence)}$$

$$|g'_2(x)| = \left| \frac{3}{2\sqrt{3x-2}} \right| = \frac{3}{4} < 1 \text{ (linear convergence with constant 0.75)}$$

$$|g'_3(x)| = \left| \frac{2}{x^2} \right| = \frac{1}{2} < 1 \text{ (linear convergence with constant 0.50)}$$

$$|g'_4(x)| = \left| \frac{-2(x^2 - 2)}{(2x - 3)^2} + \frac{2x}{2x + 3} \right| = 0 \text{ (quadratic convergence)}$$

Showing linear convergence is equivalent to saying  $g_i$  is a contraction.

2. Verifying my analysis:

```
def fixedp(f, x0, eps=10e-6, n=100):
    """ Fixed point algorithm """
    e = 1
    itr = 0
    xp = []
    x = 2.5
    while(e > eps):
        x = f(x0)
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        #print(x)
        e = np.linalg.norm(np.fabs(x-x0))
        x0 = x
        xp.append(x0)
        #print(xp[itr])
        itr = itr + 1
        print(itr)
    return x,xp

def main():
    g1 = lambda x: (pow(x, 2) + 2)/3
    g2 = lambda x: math.sqrt(3*x - 2)
    g3 = lambda x: 3 - 2/x
    g4 = lambda x: (pow(x,2)-2)/(2*x - 3)

    x,xp = fpiters(g1, 2.1, 100)

    print(x)

if __name__ == "__main__":
    main()

```

The results came out as expected.

## Problem 3

The code for Bisection, Secant, and Newton methods for solving 1-D nonlinear equations is given below:

```

import numpy as np
import math
maxIterations=1000000
threshold=0.00001
def f1(x):
    return x**3-2*x-5.
def f1prime(x):
    return 3.*x**2-2
def f2(x):
    return math.exp(-x)-x
def f2prime(x):
    return -math.exp(-x)-1
def f3(x):
    return x*math.sin(x)-1.
def f3prime(x):
    return x*math.cos(x)+math.sin(x)
def f4(x):
    return x**3-3.*x**2+3.*x-1.
def f4prime(x):
    return 3*x**2-6*x+3

def Newton(f,fprime,x0):
    x=x0
    sol=np.zeros(100) #Create array of computed solutions
    sol[0]=x
    xnew=x-f(x)/fprime(x) # Compute first iteration
    iterations=1
    sol[iterations]=xnew
    while(math.fabs(x-xnew)>threshold and iterations<maxIterations):
        x=xnew
        xnew=x-f(x)/fprime(x)
        iterations+=1
        sol[iterations]=xnew

    ek=np.zeros(iterations);
    for i in range(iterations):

```

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        ek[i]=math.log10(math.fabs(xnew-sol[i])) # Compute error at each iteration
    z=np.polyfit(ek[0:ek.shape[0]-1],ek[1:ek.shape[0]],1) # Fit line through (x,y)->(log|e(k)|,log|e(k+
    print('Order of convergence',z[0])
    print('Iterations=',iterations)
    return xnew

def Secant(f,x1,x0):
    sol=np.zeros(100)#Create array of computed solutions
    xold=x1
    xoldest=x0
    sol[0]=xoldest
    sol[1]=xold
    iterations=2
    xnew=xold-(f(xold)/(f(xold)-f(xoldest)))/(xold-xoldest)# Compute first iteration
    sol[iterations]=xnew
    while(math.fabs(xnew-xold)>threshold and iterations<maxIterations):
        xoldest=xold
        xold=xnew
        xnew=xold-f(xold)/((f(xold)-f(xoldest))/(xold-xoldest))
        iterations+=1
        sol[iterations]=xnew

    ek=np.zeros(iterations);
    for i in range(iterations):
        ek[i]=math.log10(math.fabs(xnew-sol[i]))# Compute error at each iteration
    z=np.polyfit(ek[0:ek.shape[0]-1],ek[1:ek.shape[0]],1)# Fit line thru (x,y)->(log|e(k)|,log|e(k+1)|)
    print('Order of convergence',z[0])
    print('Iterations=',iterations)
    return xnew

def bisection(f,low,high):
    iterations=0;
    sol=np.zeros(100)#Create array of computed solutions
    while(low<=high and iterations<maxIterations):
        mid=(low*(1.)+high*(1.))/2. # Compute midpoint of interval
        sol[iterations]=mid
        if math.fabs(f(mid))<threshold:
            low=high+1
        elif f(mid)*f(low)<0:
            high=mid
        else:
            low=mid
        iterations+=1

    ek=np.zeros(iterations);
    for i in range(iterations):
        if not math.fabs(sol[i])==0:
            ek[i]=math.log10(math.fabs(sol[i]))# Compute error at each iteration
        else:
            ek[i]=threshold
    z=np.polyfit(ek[0:ek.shape[0]-1],ek[1:ek.shape[0]],1)# Fit line through (x,y)->(log|e(k)|,log|e(k+1
    print('Order of convergence',z[0])
    print('Iterations=',iterations)
    return mid;

print("Newton's method on problem 1")
sol=Newton(f1,f1prime,1)
print('x=', sol)

print("Secant method on problem 1")
sol=Secant(f1,1.1,1.)
print('x=', sol)

print("Bisection method on problem 1")
sol=bisection(f1,0,5)
print('x=', sol)

```

The termination criterion that was used here was checking the absolute difference of  $x$  is greater than epsilon (threshold or not).

## Problem 4

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Recall the secant update to be  $x_{k+1} = x_k - \frac{f(x_k)(x_k - x_{k-1})}{f(x_k) - f(x_{k-1})}$

$$x_k = x^* \Rightarrow x_{k+1} = x^* - \frac{f(x^*)(x^* - x_{k-1})}{f(x^*) - f(x_{k-1})} = x^* - \frac{0(x^* - x_{k-1})}{0 - f(x_{k-1})} = x^*$$

$$x_{k-1} = x^* \Rightarrow x_{k+1} = x_k - \frac{f(x_k)(x_k - x^*)}{f(x_k) - f(x^*)} = x_k - \frac{f(x_k)(x_k - x^*)}{f(x_k)} = x^*$$