

1. Show that

$$\frac{x_{n+1} - x_n}{x_n - x_{n-1}} \approx \lambda$$

Suppose that $x_{n+1} - \bar{x}$, $x_n - \bar{x}$ and $x_{n-1} - \bar{x}$ take

$$e_n = x_n - \bar{x} \quad \text{--- (1)}$$

$$e_{n+1} = x_{n+1} - \bar{x} \quad \text{--- (2)}$$

e_{n+1} can be expressed in terms of e_n using:

$$e_{n+1} = \lambda e_n \quad \text{--- (3)}$$

Substituting (1) and (2) into (3) we obtain:

$$x_{n+1} - \bar{x} = \lambda (x_n - \bar{x}) \quad \text{--- (4)}$$

Similarly:

$$e_n = \lambda e_{n-1} \quad \text{--- (5)}$$

Obtaining

$$(x_n - \bar{x}) = \lambda (x_{n-1} - \bar{x}) \quad \text{--- (5)}$$

Subtracting (4) from (5)

$$x_n - x_{n+1} = \lambda x_{n-1} - \lambda x_n$$

~~to~~ Simplifying to

$$\frac{x_{n+1} - x_n}{x_n - x_{n-1}} \approx \lambda$$

2. For Superlinear Convergence $\alpha=1$, $\lambda=0$, hence

$$\lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|} = 0$$

Consider $\lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|^\alpha} = \lambda$, for $\lambda \neq 0$ and $\alpha > 1$

So

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|} &= \lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|^{(\alpha - \alpha + \alpha)}} \\ &= \lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|^\alpha |x_n - \bar{x}|^{1-\alpha}} \\ &= \lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|^\alpha} |x_n - \bar{x}|^{\alpha-1} \end{aligned}$$

Since $\lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|^\alpha} = \lambda$, then,

$$\lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|} = \lambda \lim_{n \rightarrow \infty} |x_n - \bar{x}|^{\alpha-1}$$

Since $\alpha > 1$, it means fast convergence therefore x_n tends quickly to \bar{x} , hence $\lim_{n \rightarrow \infty} |x_n - \bar{x}|^{\alpha-1} = 0$ as α increases. Thus,

$$\lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|} = \lambda \cdot 0 = 0$$

which is Superlinear Convergence since $\alpha=1$, and $\lambda=0$

3. Suppose that $\{x_n\}$ Converges Superlinearly to \bar{x} .
Show that.

$$\lim_{n \rightarrow \infty} \frac{|x_{n+1} - x_n|}{|x_n - \bar{x}|} = 1$$

$$\lim_{n \rightarrow \infty} \frac{|x_{n+1} - x_n|}{|x_n - \bar{x}|} = \lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x} + \bar{x} - x_n|}{|x_n - \bar{x}|}$$

$$\leq \lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|} + \lim_{n \rightarrow \infty} \frac{|x_n - \bar{x}|}{|x_n - \bar{x}|}$$

$$\lim_{n \rightarrow \infty} \frac{|x_{n+1} - x_n|}{|x_n - \bar{x}|} \leq 1 + \lim_{n \rightarrow \infty} \frac{|x_n - \bar{x}|}{|x_n - \bar{x}|}$$

$$\text{For Super linear Convergence, } \lim_{n \rightarrow \infty} \frac{|x_{n+1} - \bar{x}|}{|x_n - \bar{x}|} = 0$$

$$\lim_{n \rightarrow \infty} \frac{|x_{n+1} - x_n|}{|x_n - \bar{x}|} \leq 1 + 0$$

hence

$$\lim_{n \rightarrow \infty} \frac{|x_{n+1} - x_n|}{|x_n - \bar{x}|} = 1$$

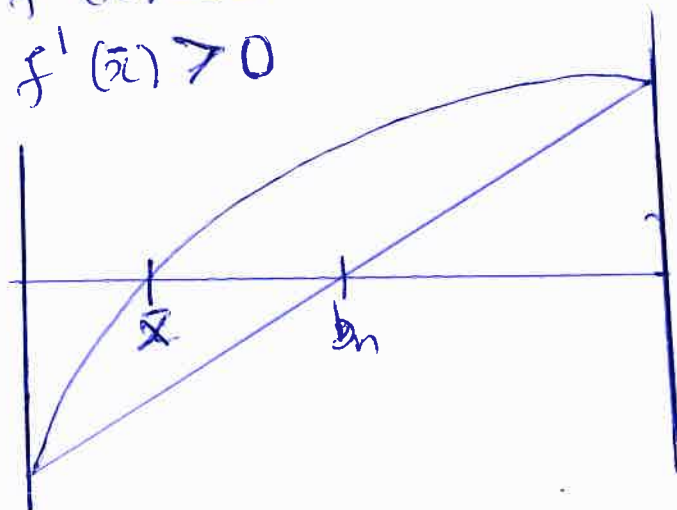
While developing root-finding schemes we are interested in how fast the solution converges to the root, so this makes us to be interested in the error between the solution and the root. So if the error be at all possible solutions near the root are the same, giving 1 after division, this means that the scheme approximates well the root. therefore $|x_n - \bar{x}| = |x_{n+1} - x_n|$ gives more information on how fast the scheme will converge to the ~~solution~~ root.

1. Show that $\lambda \approx \frac{L f''(\bar{x})}{2f'(\bar{x}) + L f''(\bar{x})}$ satisfies $|\lambda| < 1$.

Using the four possible cases.

$$\lambda \approx \frac{L f''(\bar{x})}{2f'(\bar{x}) + L f''(\bar{x})}, \text{ where } L = \begin{cases} b_n - \bar{x}, & b_n \text{ fixed} \\ a_n - \bar{x}, & a_n \text{ fixed} \end{cases}$$

Case 1: $f''(\bar{x}) < 0$
 $f'(\bar{x}) > 0$



If b_n is fixed, $L = b_n - \bar{x}$.

from the diagram, $b_n - \bar{x} > 0$ and $f''(\bar{x}) < 0$

therefore $(b_n - \bar{x}) f''(\bar{x}) < 0$.

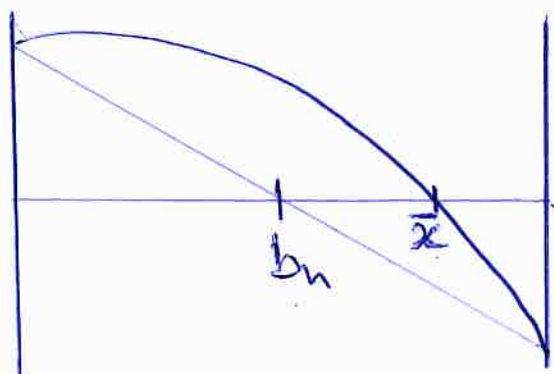
Since $f'(\bar{x}) > 0$, it follows that

$$2f'(\bar{x}) + (b_n - \bar{x}) f''(\bar{x}) > (b_n - \bar{x}) f''(\bar{x})$$

$$1 > \frac{(b_n - \bar{x}) f''(\bar{x})}{2f'(\bar{x}) + (b_n - \bar{x}) f''(\bar{x})}$$

$$1 > \left| \frac{(b_n - \bar{x}) f''(\bar{x})}{2f'(\bar{x}) + (b_n - \bar{x}) f''(\bar{x})} \right| = |\lambda|, \text{ hence } \underline{\underline{|\lambda| < 1}}$$

Case 2: $f''(\bar{x}) < 0$
 $f'(\bar{x}) < 0$



If b_n is fixed, $L = b_n - \bar{x}$
 from the diagram, $b_n - \bar{x} < 0$ and $f''(\bar{x}) < 0$
 $|(b_n - \bar{x}) f''(\bar{x})| > 0$

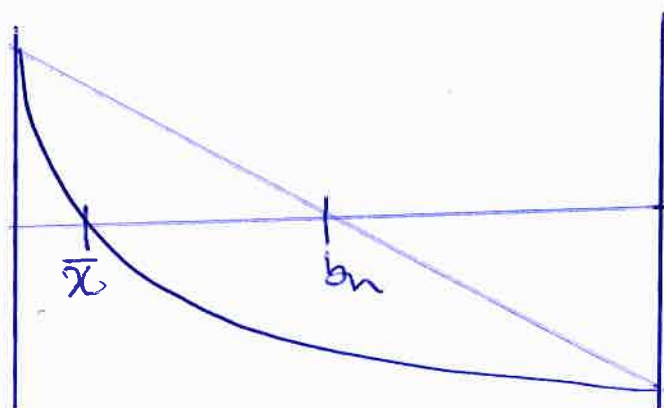
Since $f'(\bar{x}) < 0$, it follows that

$$|2f'(\bar{x}) + (b_n - \bar{x})f''(\bar{x})| > |(b_n - \bar{x})f''(\bar{x})|$$

$$1 > \left| \frac{(b_n - \bar{x})f''(\bar{x})}{2f'(\bar{x}) + (b_n - \bar{x})f''(\bar{x})} \right| = |\lambda|$$

$$\text{hence } |\lambda| < 1$$

Case 3: $f''(\bar{x}) > 0$
 $f'(\bar{x}) < 0$



fixing b_n , $(b_n - \bar{x}) > 0$

Since $f''(\bar{x}) > 0$

then $|(b_n - \bar{x}) f''(\bar{x})| > 0$

Since $f'(\bar{x}) < 0$, then it follows that

$$|2f'(\bar{x}) + (b_n - \bar{x})f''(\bar{x})| > |(b_n - \bar{x})f''(\bar{x})|$$

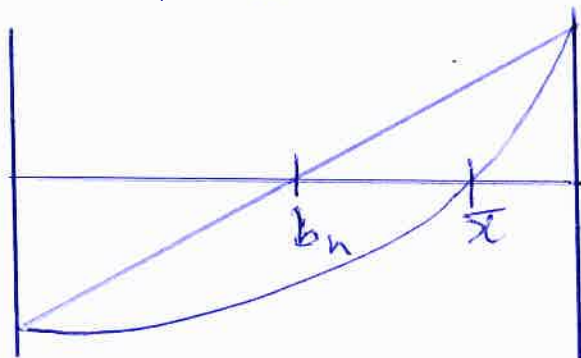
$$1 > \left| \frac{(b_n - \bar{x})f''(\bar{x})}{2f'(\bar{x}) + (b_n - \bar{x})f''(\bar{x})} \right| = |\lambda|$$

hence

$$|\lambda| < 1$$

Case 4: $f''(\bar{x}) > 0$

$$f'(\bar{x}) > 0$$



fixing b_n , $(b_n - \bar{x}) < 0$ and $f''(\bar{x}) > 0$

$$(b_n - \bar{x}) f''(\bar{x}) < 0$$

Since $f'(\bar{x}) > 0$, it follows that

$$2f'(\bar{x}) + (b_n - \bar{x})f''(\bar{x}) > (b_n - \bar{x})f''(\bar{x})$$

$$1 > \left| \frac{(b_n - \bar{x})f''(\bar{x})}{2f'(\bar{x}) + (b_n - \bar{x})f''(\bar{x})} \right| = |\lambda|$$

Hence

$$1 > |\lambda|$$


```
In [17]: %matplotlib notebook
%matplotlib inline
Using matplotlib backend: nbAgg
Populating the interactive namespace from numpy and matplotlib
```

No.4

```
In [18]: f=lambdax:(x**2)-(1e51)

def bisection(f,a,b):
    tol=1e-10
    fa=f(a)
    kmax=int(log2((b-a)/tol)+1)
    print("Estimated Number of iterations =",kmax,"iterations")
    for k in range(kmax):
        c=(a+b)/2
        fc=f(c)
        if sign(f(a))!=sign(f(c)):
            b=c
        else:
            a=c
            fa=fc
        k+=1
    if abs(b-a)<tol:
        print("Converged in {0} iterations".format(k))
        break
    return c

xroot=bisection(f,2e25,4e25)
#print(xroot)
#print(f(xroot))
print("x=:{24.10f}".format(xroot))
print("f(x)=:{24.4e}".format(f(xroot)))

Estimated Number of iterations = 118 iterations
x = 31622776601683794681921536.000000000000000000
f(x) = 1.6615e+35

Can we actually achieve this tolerance?

No

Why?

Because we know that the iterations in bisection method is always converging to the xroot, and f(xroot) must be zero which is not the case in our calculations, then we cant achieve this tolerance.

How would you choose a more appropriate tolerance?

Since the appropriate tolerance depends on the number of iterations N, and N depends only on the initial interval [a_0,b_0] bracketing the root. Therefore the interval length after N iterations is  $\frac{b_0-a_0}{2^N}$  and this must be less or equal to  $\tau$  to obtain an accuracy of  $\tau$  i.e.
```

No. 5

```
In [19]: def fixed_point(f,x0,beta):
x=x0 #initial guess
kmax=100
tol=1e-10
for k in range(kmax):
    x1=f(x)
    if abs(x1-x)<tol:
        print("Tolerance achieved\n")
        xroot=x1
        break
    x=x1
print('The root = ',x1)
print('Number of iterations = ',k)

In [20]: f=lambdax:(1/3)*(x**3)-(x**2)+(4/3)*beta
x0=0.1
fixed_point(f,x0,beta)

Tolerance achieved

The root = 0.1195995366904333
Number of iterations = 13

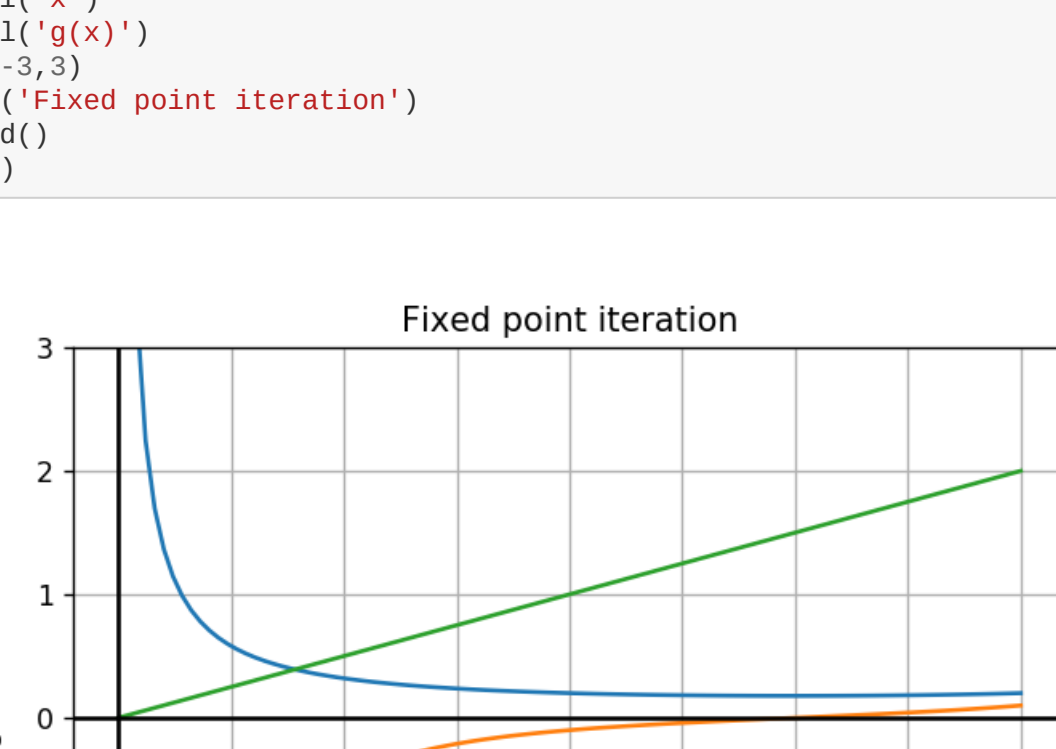
In [21]: f=lambdax:(1/3)*(x**3)-(x**2)+(4/3)*beta
x0=0.1
fixed_point(f,x0,beta)

The root = 0.7942962233406211
Number of iterations = 99
```

$\beta = 0.1$

```
In [22]: beta=0.1
#let f(x)=0 be written in the form x=g(x)
#first function g(x)
x=linspace(0,2,100)
g=lambdax:((1/3)*(x**3)-((4/3)*beta))**(1/3)
gprime=lambdax:((3*(x**2))-((4*beta)))/(2*sqrt((4*beta)-(x**3)))
y=lambdax:x

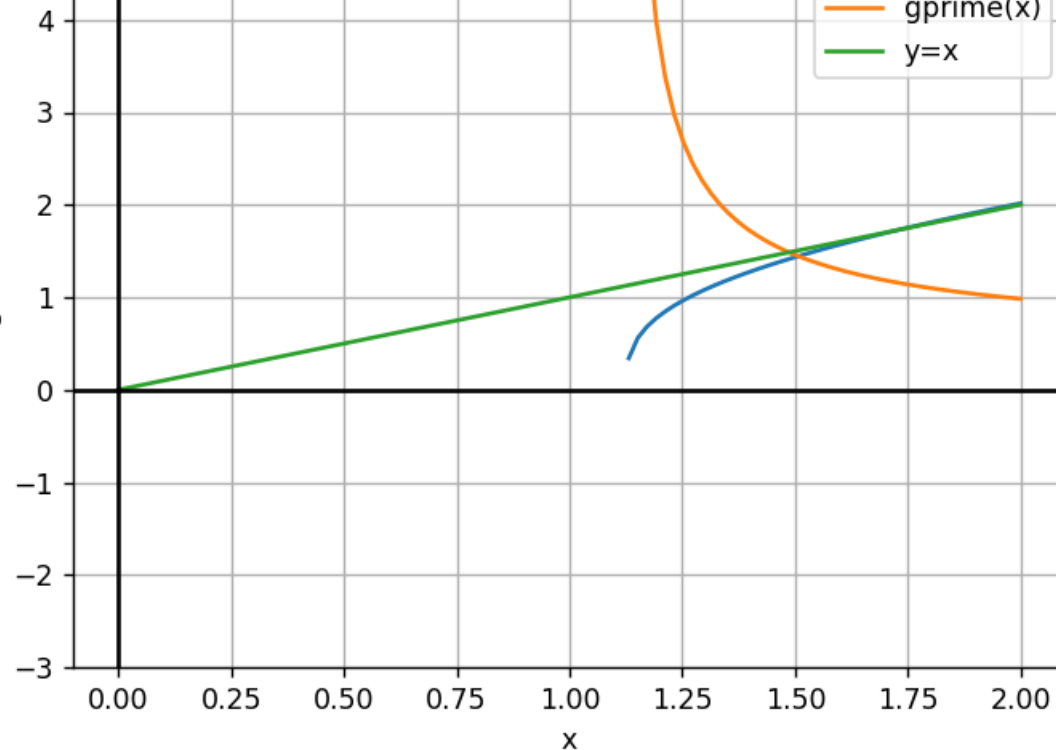
figure(1)
#plot(x,g(x),label='g(x)')
plot(x,gprime(x),label='gprime(x)')
plot(x,y(x),label='y=x')
xlabel('x')
ylabel('g(x)')
title('Fixed point iteration')
axhline(y=0, color='k')
axvline(x=0, color='k')
grid()
legend()
show()
```



From the graph above, it is easily seen that for all $x \in [0, 2]$, $g(x)$ does not take every value in the interval $[0, 2]$, even though it intersects with $y=x$. And also $g(x)$ exists but doesn't satisfy $|g(x)| \leq r$ for r between $(0, 1)$, therefore $x=g(x)$ doesn't have a unique solution, hence doesn't converge for $g(x) \in [0, 2]$.

```
In [23]: beta=0.1
#let f(x)=0 be written in the form x=g(x)
#second function g(x)
x=linspace(0,2,100)
g=lambdax:((1/3)*(x**3)-((4/3)*beta))**(1/2)
gprime=lambdax:(((3*(x**2))-((4*beta)))/(2*sqrt((4*beta)-(x**3))))
y=lambdax:x

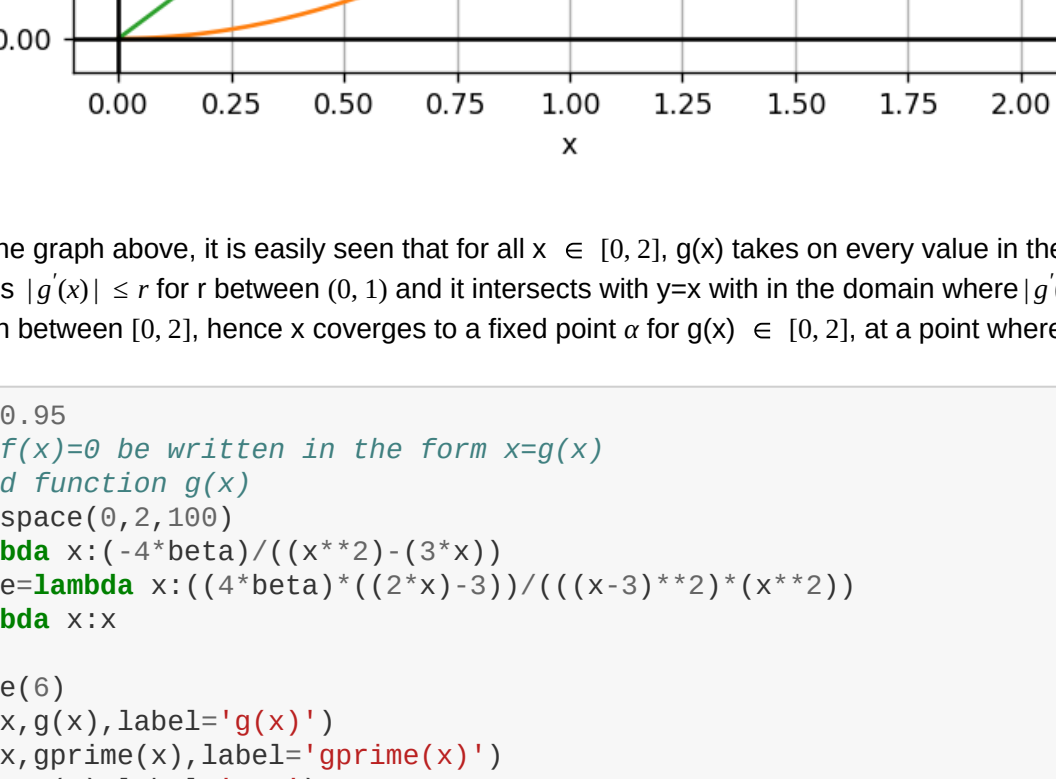
figure(2)
plot(x,g(x),label='g(x)')
plot(x,gprime(x),label='gprime(x)')
plot(x,y(x),label='y=x')
xlabel('x')
ylabel('g(x)')
title('Fixed point iteration')
axhline(y=0, color='k')
axvline(x=0, color='k')
grid()
legend()
show()
```



From the graph above, it is easily seen that for all $x \in [0, 2]$, $g(x)$ takes on every value in the interval $[0, 2]$. Since also $g(x)$ exists and satisfies $|g(x)| \leq r$ for r for r between $(0, 1)$ and it intersects with $y=x$ with in the domain where $|g(x)| \leq r$. Therefore $x=g(x)$ has a unique solution between $[0, 2]$, hence x converges to a fixed point α for $g(x) \in [0, 2]$, at a point where $g(x)$ intersects $y=x$.

```
In [24]: beta=0.1
#let f(x)=0 be written in the form x=g(x)
#third function g(x)
x=linspace(0,2,100)
g=lambdax:((4*beta)-((2*x)-3))/(((x-3)**2)-(x**2))
y=lambdax:x

figure(3)
plot(x,g(x),label='g(x)')
plot(x,gprime(x),label='gprime(x)')
plot(x,y(x),label='y=x')
xlabel('x')
ylabel('g(x)')
title('Fixed point iteration')
axhline(y=0, color='k')
axvline(x=0, color='k')
grid()
legend()
show()
```

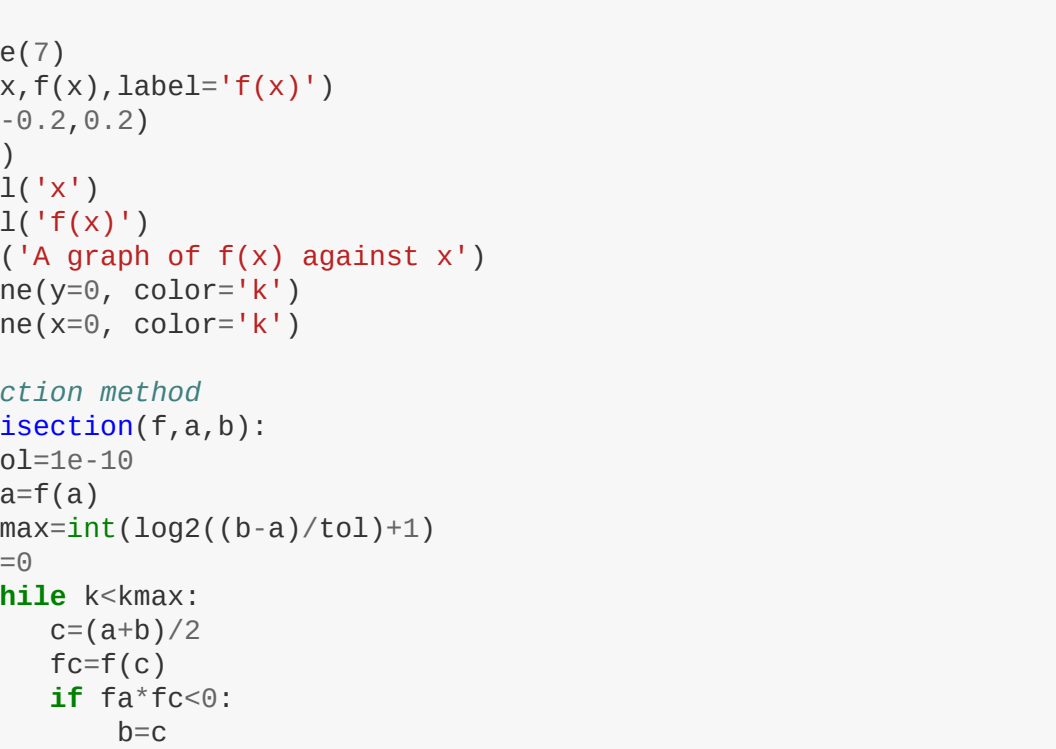


From the graph above, it is easily seen that for all $x \in [0, 2]$, $g(x)$ does not take every value in the interval $[0, 2]$, since the last point is $-\infty$ hence diverges, even though it intersects with $y=x$, and that $g(x)$ exists and satisfies $|g(x)| \leq r$ for r between $(0, 1)$. Therefore $x=g(x)$ doesnot have a unique solution, hence doesn't converge for $g(x) \in [0, 2]$.

$\beta = 0.95$

```
In [25]: beta=0.95
#let f(x)=0 be written in the form x=g(x)
#third function g(x)
x=linspace(0,2,100)
g=lambdax:((4*beta)-((2*x)-3))/(((x-3)**2)-(x**2))
y=lambdax:x

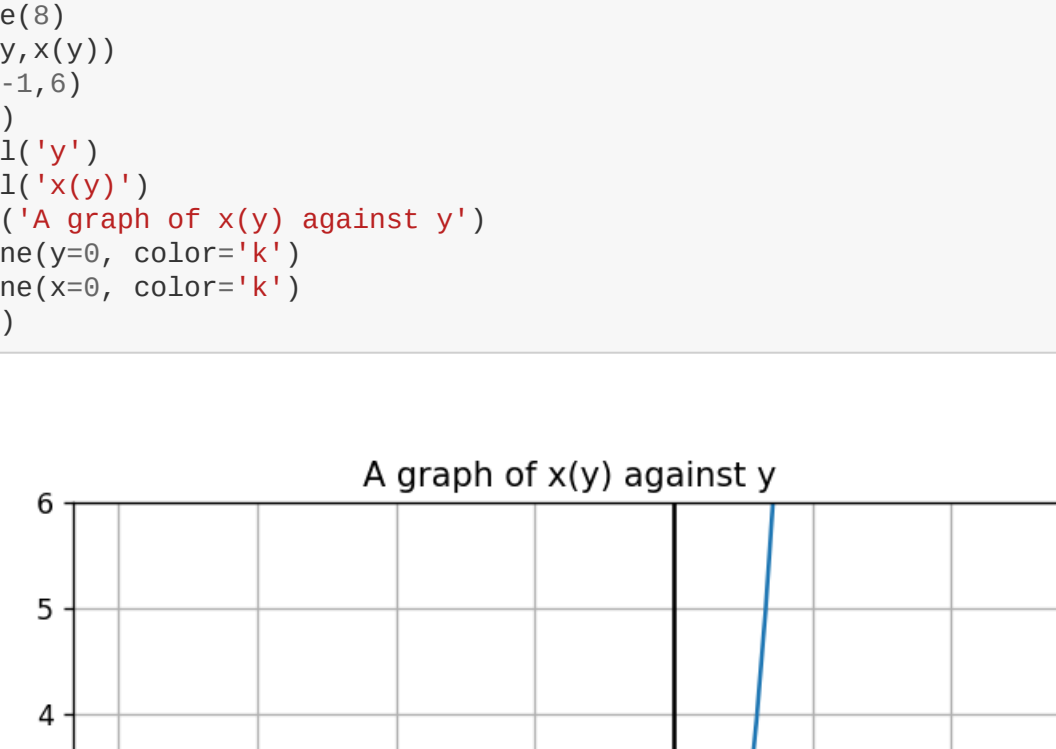
figure(4)
plot(x,g(x),label='g(x)')
plot(x,gprime(x),label='gprime(x)')
plot(x,y(x),label='y=x')
xlabel('x')
ylabel('g(x)')
title('Fixed point iteration')
axhline(y=0, color='k')
axvline(x=0, color='k')
grid()
legend()
show()
```



From the graph above, it is easily seen that for all $x \in [0, 2]$, $g(x)$ does not take every value in the interval $[0, 2]$, since the last point is $-\infty$ hence diverges, even though it intersects with $y=x$, and that $g(x)$ exists and satisfies $|g(x)| \leq r$ for r between $(0, 1)$. Therefore $x=g(x)$ doesnot have a unique solution, hence doesn't converge for $g(x) \in [0, 2]$.

```
In [26]: beta=0.95
#let f(x)=0 be written in the form x=g(x)
#second function g(x)
x=linspace(0,2,100)
g=lambdax:((1/3)*(x**3)-((4/3)*beta))**(1/2)
gprime=lambdax:(((3*(x**2))-((4*beta)))/(2*sqrt((4*beta)-(x**3))))
y=lambdax:x

figure(5)
plot(x,g(x),label='g(x)')
plot(x,gprime(x),label='gprime(x)')
plot(x,y(x),label='y=x')
xlabel('x')
ylabel('g(x)')
title('Fixed point iteration')
axhline(y=0, color='k')
axvline(x=0, color='k')
grid()
legend()
show()
```



From the graph above, it is easily seen that for all $x \in [0, 2]$, $g(x)$ takes on every value in the interval $[0, 2]$. Since also $g(x)$ exists and satisfies $|g(x)| \leq r$ for r for r between $(0, 1)$ and it intersects with $y=x$ with in the domain where $|g(x)| \leq r$. Therefore $x=g(x)$ has a unique solution between $[0, 2]$, hence x converges to a fixed point α for $g(x) \in [0, 2]$, at a point where $g(x)$ intersects $y=x$.

```
In [27]: beta=0.95
#let f(x)=0 be written in the form x=g(x)
#third function g(x)
x=linspace(0,2,100)
g=lambdax:((4*beta)-((2*x)-3))/(((x-3)**2)-(x**2))
y=lambdax:x

figure(6)
plot(x,g(x),label='g(x)')
plot(x,gprime(x),label='gprime(x)')
plot(x,y(x),label='y=x')
xlabel('x')
ylabel('g(x)')
title('Fixed point iteration')
axhline(y=0, color='k')
axvline(x=0, color='k')
grid()
legend()
show()
```



From the graph above, it is easily seen that for all $x \in [0, 2]$, $g(x)$ does not take every value in the interval $[0, 2]$, since the last point is $-\infty$ hence diverges, even though it intersects with $y=x$, and that $g(x)$ exists and satisfies $|g(x)| \leq r$ for r between $(0, 1)$. Therefore $x=g(x)$ doesnot have a unique solution, hence doesn't converge for $g(x) \in [0, 2]$.

No.6

```
In [28]: r=2 #radius
dw=0.04 #density of marble
dm=0.998 #density of water

beta=dm/dw

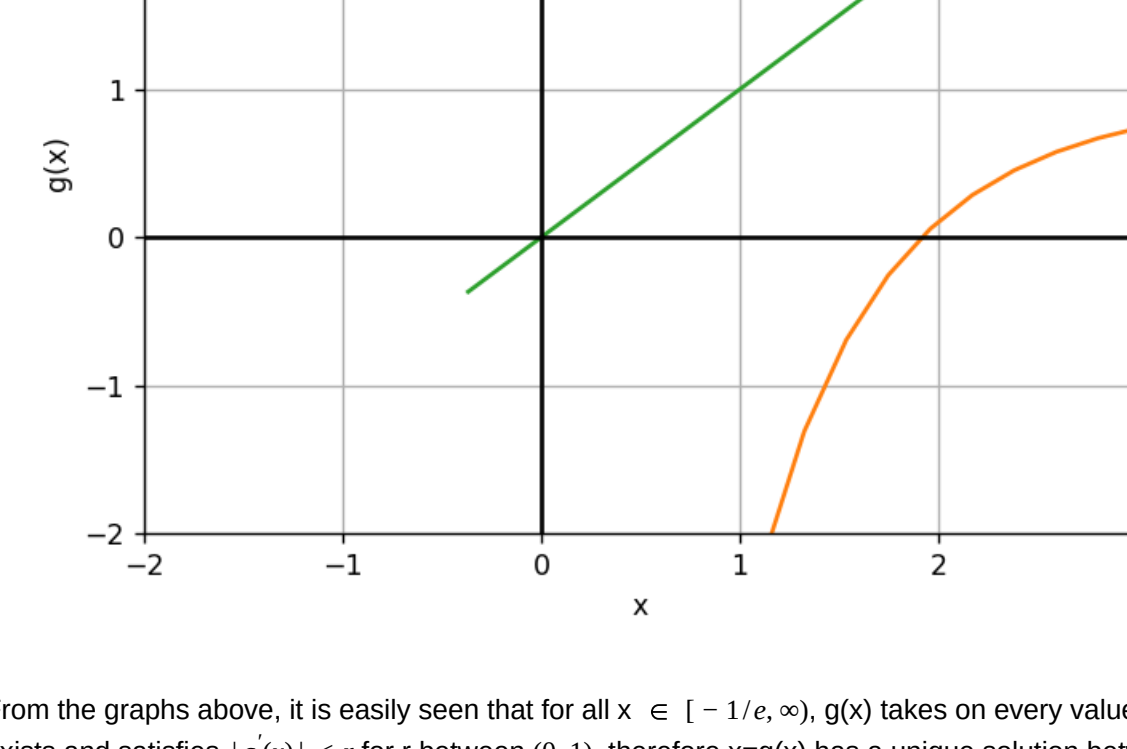
f=lambdax:(1/3)*(x**3)-(x**2)+(4/3)*beta
n=lambdax:r*x
ao,bo=-0.5,2
n=100
x=linspace(ao,bo,n)

#location of t_g
root=zeros(len(x))
mark=zeros(len(x))
mark[0]=0

figure(7)
plot(x,f(x),label='f(x)')
ylim(-0.2,0.2)
grid()
xlabel('x')
ylabel('f(x)')
title('A graph of f(x) against x')
axhline(y=0, color='k')
axvline(x=0, color='k')

#bisection method
def bisection(f,a,b):
    tol=1e-10
    fa=f(a)
    kmax=int(log2((b-a)/tol)+1)
    k=0
    while k<kmax:
        c=(a+b)/2
        fc=f(c)
        if fa*fc<0:
            b=c
        else:
            a=c
            fa=fc
        k+=1
    if abs(b-a)<tol:
        break
    return c

xroot=bisection(f,ao,bo)
root[0]=xroot
plot(root,mark,marker='xroot', ls='', marker='o', label='xroot')
legend()
show()
print('The depth is: ',h(xroot))
```



The depth is: 0.48211818968411535

No.8

a)

```
In [29]: x=lambdax:y.*exp(y)
#prime=lambdax:(y+1)*exp(y)
x=linspace(-8,5,100)
figure(8)
plot(y,x(y),label='x(y)')
ylim(-1,6)
grid()
xlabel('y')
ylabel('x(y)')
title('A graph of x(y) against y')
axhline(y=0, color='k')
axvline(x=0, color='k')
show()
```



According to the graph above we have a turning point between 0 and 2, therefore the curve $x = y \exp(y)$ fails the horizontal test. So since $L(x) = \frac{dx}{dy} = (y+1)\exp(y)$, at turning point:

$$\frac{dx}{dy} = 0 \implies (y+1)\exp(y) = 0$$

Thus,

either

$$\exp(y) = 0 \implies y = \infty$$

or

$$y+1=0 \implies y=-1$$

For the turning points,using $x = y \exp(y)$:

for $y = -1 \implies x = -1/e$ and for $y = \infty$ respectively.

Therefore concluding that the range of $x = y \exp(y)$ is $[-1/e, \infty)$, since $L(x)$ is the inverse of $y \exp(y)$, and its known that range of $y \exp(y)$ is the domain of its inverse, $L(x)$.

b)

```
In [30]: def L(x):
f=lambdax:y.*exp(y)
fprime=lambdax:(y+1)*exp(y)
tol=1e-10
n=100
#initial guess
if x<0:
    yo=-0.01
else:
    yo=log(x)
i=0
while i<n:
    fo=f(yo)
    fpo=fprime(yo)
    yn=yo-(fo-x)/fpo
    if abs(yn-yo)<tol:
        break
    yo=yn
    i+=1
return yn
```

c)

```
In [60]: def g(x,y):
return (x/(x+1))^(x+y/(x*exp(x)))

def gprime(x,y):
return ((x+2)/((x+1)**2))*(x-y/(exp(x)))

x=linspace(-1/e,1e),10)

for y in range(11,14): #sample values of y
figure(9)
plot(x,g(x,y),label='g(x)')
plot(x,gprime(x,y),label='g'(x)')
plot(x,y,y=x)
grid()
axhline(y=0, color='k')
axvline(x=0, color='k')
xlabel('g(x)')
ylabel('g'(x)')
ylim(-2,3)
xlim(-2,3)
title('A graph of g(x) against x')
legend()
show()
```


From the graphs above, it is easily seen that for all $x \in [-1/e, \infty)$, $g(x)$ takes on every value in the interval $[-1/e, \infty)$. Since also $g(x)$ exists and satisfies $|g(x)| \leq r$ for r between $(0, 1)$, therefore $x=g(x)$ has a unique solution between $[-1/e, \infty)$, and it intersects with $y=x$ with in the domain where $|g(x)| \leq r$. Hence x converges to a fixed point α for $g(x) \in [-1/e, \infty)$, at a point where $g(x)$ intersects $y=x$ for $x \in [-1/e, \infty)$. Hence since $x=g(x)$ converges then also Newton's method will also converge in all graphs, therefore we can suppose that it will converge for other values of y for any x in the domain $[-1/e, \infty)$.

d)

```
In [33]: #check that L(xe^x)=x
x=linspace(0,10,1)
for i in x:
    print('x=',i,', L(1*exp(1))')
```

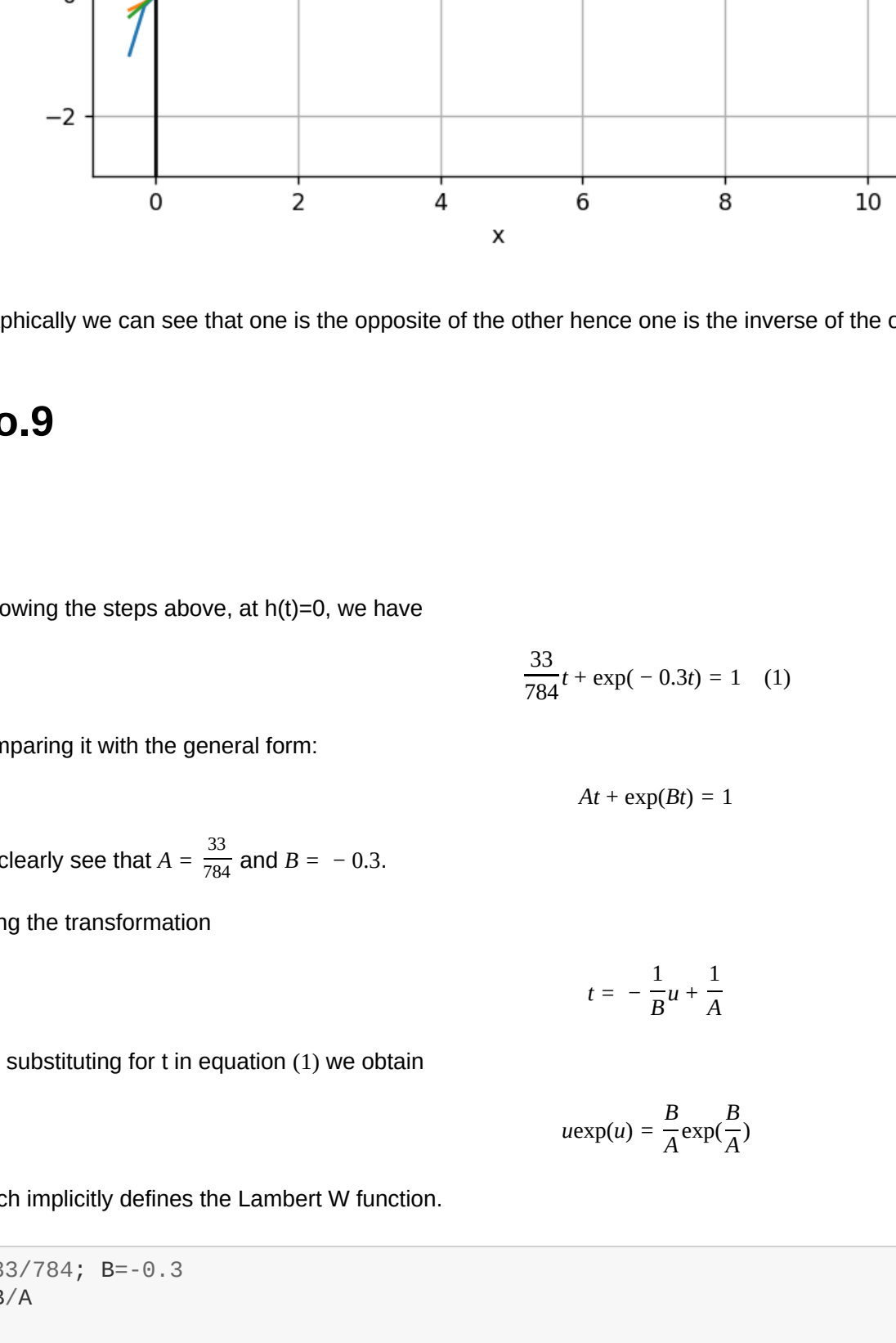
Hence we are accurately computing $L(\exp(x)) = x$

e)

In [34]: `x=linspace(-1/6,10) #for aprinciple branch`
`L1=vectorize(L)`

```
y=Lambda x:x*exp(x)

figure(8)
plot(x,L1(x),label='L(x)')
plot(x,y(x),label='y=x*exp(x)')
plot(x,x,label='y=x')
axhline(y=0,color='k')
axvline(x=0,color='k')
xlabel('L(x) and y')
ylabel('x')
title('A graph of L(x) against x')
legend()
ylim(-3,7)
grid()
show()
```



Graphically we can see that one is the opposite of the other hence one is the inverse of the other

No.9

a)

Following the steps above, at $h(t)=0$, we have

$$\frac{33}{784}t^4 + \exp(-0.3t) = 1 \quad (1)$$

Comparing it with the general form:

$$At + \exp(Bt) = 1$$

we clearly see that $A = \frac{33}{784}$ and $B = -0.3$.

Using the transformation

$$t = -\frac{1}{B}u + \frac{1}{A}$$

and substituting for t in equation (1) we obtain

$$u\exp(u) = \frac{B}{A} - \frac{B}{A^2}$$

which implicitly defines the Lambert W function.

In [35]: `A=33/784; B=-0.3`

```
C=B/A

#calling the Lambert W function L(x) to use it to calculate u
u=L(C*exp(C))

#time t_g for the rocket to hit the ground
t_g=Lambda u:-:(1/B)*u+(1/A)
print("t_g = :8f)".format(t_g(u)),"Seconds")

t_g = 23.738391 Seconds
```

b)

In [36]: `h=Lambda t:-33*t + 784*(1-exp(-0.3*t))`

```
t=linspace(8,36)
root=zeros(len(t))
root[0]=t_g(u)
mark=zeros(len(t))
mark[0]=0

figure(10)
plot(t,h(t),label='h(t)')
plot(root,mark,markevery=t_g(u), ls="", marker="o", label="tg")
axhline(y=0,color='k')
axvline(x=0,color='k')
xlabel('h(t)')
ylabel('t')
title('A graph of h(t) against t')
legend()
grid()
show()
```



The value of t_g is correct, since we considered that at $h = 0$ is the ground, and looking on the graph, this is the exact time at which $h(t)$ is zero.

c)

Since in our equation $h(t)$, on the right hand side t appears both inside and outside in the exponential function. So this means solving this equation directly will give us only on solution, which is not right, since $h(t)$ is a multivalued function, thus generally it has morethan on solution. And also since we are dealing with rocket height, we expect large positive values of both t and h, with these large values t, $\exp(t)$, and $\exp(t)$ grow similarly implying that also their inverse functions will have similar asymptotes. So $\exp(t)$ will have two real branches. Giving better results than solving it directly.

In []: