

Q1

```
import numpy as np
def bisection(f,a,b,tol,L,H,C_ocean,C_land):
    if f(a,L,H,C_ocean,C_land)*f(b,L,H,C_ocean,C_land) >= 0:
        print("Bisection method fails.")
        return None
    a_n = a
    b_n = b
    error_metric = f(a,L,H,C_ocean,C_land) - f(b,L,H,C_ocean,C_land)
    while error_metric<=tol:
        m_n = (a_n + b_n)/2
        f_m_n = f(m_n,L,H,C_ocean,C_land)
        if f(a_n,L,H,C_ocean,C_land)*f_m_n < 0:
            a_n = a_n
            b_n = m_n
        elif f(b_n,L,H,C_ocean,C_land)*f_m_n < 0:
            a_n = m_n
            b_n = b_n
        elif f_m_n == 0:
            print("Found exact solution.")
            return m_n
        else:
            print("Bisection method fails.")
            return None
        error_metric = f(a_n,L,H,C_ocean,C_land) -
f(b_n,L,H,C_ocean,C_land)
    return (a_n + b_n)/2
def func(x,L,H,C_ocean,C_land):
    p1 = np.sqrt(H**2+x**2)
    p2 = L-x
    total_cost = p1*C_ocean + p2*C_land

    return total_cost
def my_pipe_builder(C_ocean, C_land, L, H):
    a = 0
    b= L
    tol = 1e-6
    print("Soltuion ", bisection(func,a,b,tol,L,H,C_ocean,C_land))
# Test case
my_pipe_builder(2,-1.5,1,0.5)
```

Found exact solution.
Soltuion 0.2529042306098583

Q2

```
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap
```

```
TOL = 1.e-8
```

```
def newton(z0, f, fprime, MAX_IT=1000):
    """The Newton-Raphson method applied to f(z).

    Returns the root found, starting with an initial guess, z0, or
    False
    if no convergence to tolerance TOL was reached within MAX_IT
    iterations.
```

```
    """
```

```
    z = z0
```

```
    for i in range(MAX_IT):
```

```
        dz = f(z)/fprime(z)
```

```
        if abs(dz) < TOL:
```

```
            return z
```

```
        z -= dz
```

```
    return False
```

```
def get_root_index(roots, r):
```

```
    """Get the index of r in the list roots.
```

```
    If r is not in roots, append it to the list.
```

```
    """
```

```
    try:
```

```
        return np.where(np.isclose(roots, r, atol=TOL))[0][0]
```

```
    except IndexError:
```

```
        roots.append(r)
```

```
        return len(roots) - 1
```

```
def newton_fractal(f, fprime, n=200, domain=(-1, 1, -1, 1)):
```

```
    """Plot a Newton Fractal by finding the roots of f(z).
```

```
    The domain used for the fractal image is the region of the complex
    plane
```

```
    (xmin, xmax, ymin, ymax) where  $z = x + iy$ , discretized into n
    values along
    each axis.
```

```
    """
```

```
    roots = []
```

```
    m = np.zeros((n, n))
```

```

xmin, xmax, ymin, ymax = domain
for ix, x in enumerate(np.linspace(xmin, xmax, n)):
    for iy, y in enumerate(np.linspace(ymin, ymax, n)):
        z0 = x + y*1j
        r = newton(z0, f, fprime)
        if r is not False:
            ir = get_root_index(roots, r)
            m[iy, ix] = ir
nroots = len(roots)
for i in range(len(roots)):
    print("Root_"+str(i+1)+" : ", roots[i])
# return forth root
return roots[3]

def f(z):
    return z**4-1
def fprime(z):
    return 4*z**3
iterations = 250
newton_fractal(f, fprime, n=iterations)

Root_1 : (-3.2665559226941115e-13+1.00000000000003602j)
Root_2 : (1.0000000000000022+1.955133178273462e-15j)
Root_3 : (3.256421917228081e-09-0.999999997807702j)
Root_4 : (-0.9999999999999997-6.198137121753741e-16j)

(-0.9999999999999997-6.198137121753741e-16j)

```

Q3

```

import matplotlib.pyplot as plt
import numpy as np
from scipy.interpolate import lagrange
def langarange_(x, y, x_p):
    yp = 0
    # Implementing Lagrange Interpolation
    for i in range(len(x)):
        p = 1
        for j in range(len(y)):
            if i != j:
                p = p * (x_p - x[j])/(x[i] - x[j])
        yp = yp + p * y[i]
    return yp

```

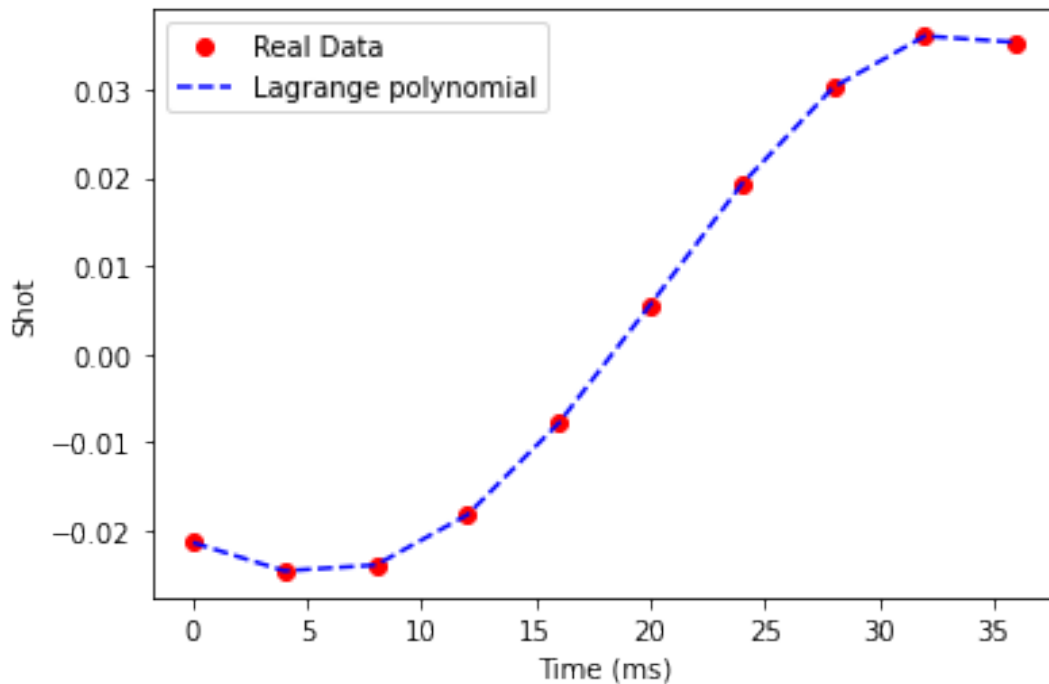
```

# def lagrange_main():
x= list(range(0,40,4))
y=[-0.021373, -0.024578,-0.023914, -0.018227,-
0.00781,0.005602,0.019264, 0.030235, 0.036059, 0.035334]
predict = []
for xp in x:
    predict.append(langrange_(x, y, xp))

plt.plot(x,y, 'ro', label='Real Data')
plt.plot(x,predict, '--b', label='Lagrange polynomial')
plt.xlabel("Time (ms)")
plt.ylabel("Shot")
plt.legend()

<matplotlib.legend.Legend at 0x7f18231d6210>

```



Q4

```

import numpy as np
from scipy import linalg
import matplotlib.pyplot as plt
#pip install PyPolynomial
from pypoly import Polynomial
N=100
X=np.linspace(np.pi/2, np.pi, N)
y = X**2*np.cos(X)

y_out = []

```

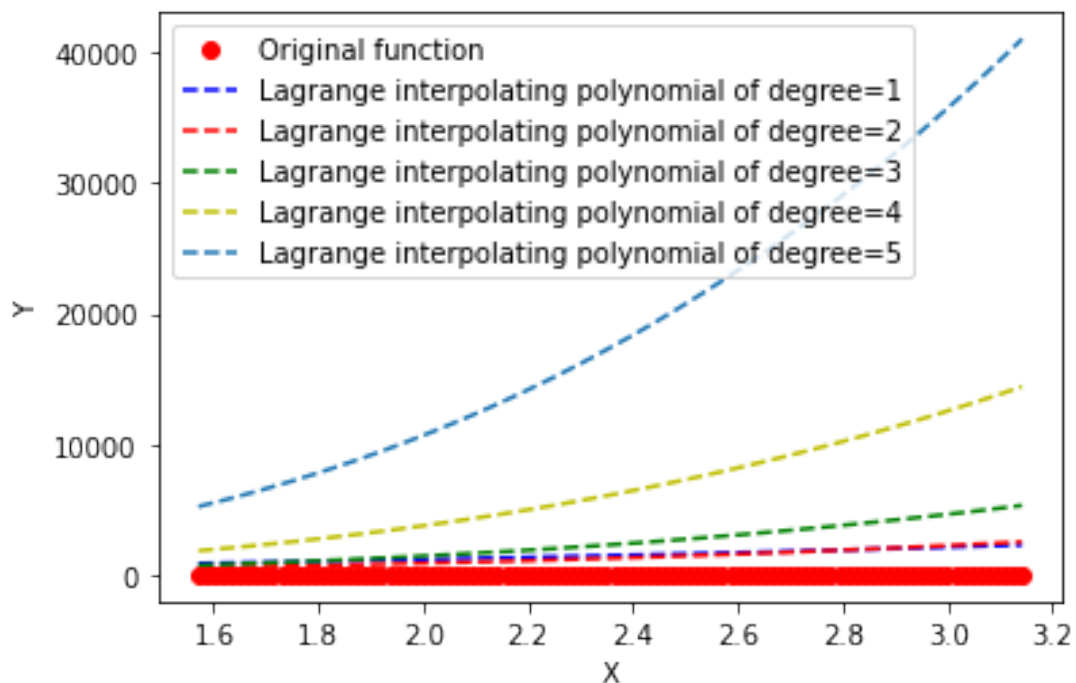
```

for order in [1, 2,3,4,5]: # reuired order of equatioin
    equations = np.array([[po ** order] for po in X])
    coefficients =linalg.lstsq(equations, y)
    p = Polynomial(*coefficients)
    out = []
    for x in X:
        out.append(p(x))
    y_out.append(out)
len(y_out)
plt.plot(X,y, 'ro', label='Original function')
plt.plot(X,y_out[0], '--b', label='Lagrange interpolating polynomial
of degree=1')
plt.plot(X,y_out[1], '--r', label='Lagrange interpolating polynomial
of degree=2')
plt.plot(X,y_out[2], '--g', label='Lagrange interpolating polynomial
of degree=3')
plt.plot(X,y_out[3], '--y', label='Lagrange interpolating polynomial
of degree=4')
plt.plot(X,y_out[4], '--', label='Lagrange interpolating polynomial of
degree=5')

plt.xlabel("X")
plt.ylabel("Y")
plt.legend()

```

<matplotlib.legend.Legend at 0x7f1817a6a6d0>

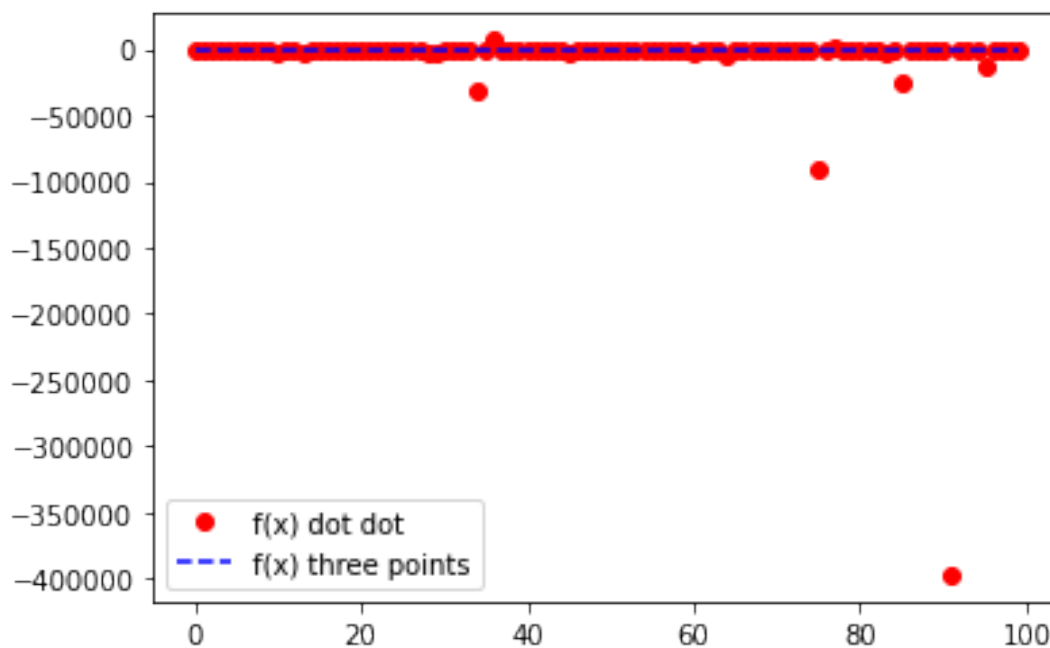


Q5

```
import numpy as np
import matplotlib.pyplot as plt
def Gaussian_Func_dot_dot(x,std, mean):
    fx = 1/(np.sqrt(2*np.pi)*std)
    fx = fx * (x-mean)**2/(std**2) - 1/(std**2)
    fx = fx * np.exp(-(x-mean)**2/x*std)
    return fx
def Gaussian_Func(x,std, mean):
    fx = 1/(np.sqrt(2*np.pi)*std)
    fx = fx * np.exp(-(x-mean)**2/x*std)
    return fx
def Gaussian_three_points(x,std, mean,dx):
    #  $f(x+dx) - 2f(x) + f(x-dx) / dx^2$ 
    return (Gaussian_Func(x+dx,std, mean)-2*Gaussian_Func(x,std, mean)
+Gaussian_Func(x-dx,std, mean))/ dx**2

# initialization
mu, sigma = 0, 0.1 # mean and standard deviation
x = np.random.normal(mu, sigma, 100)
dx=1e-6
y_dot = Gaussian_Func_dot_dot(x,s, mu)
y_3P = Gaussain_three_points(x,s, mu,dx)
plt.plot(list(range(len(x))),y_dot, 'ro', label='f(x) dot dot ')
plt.plot(list(range(len(x))),y_3P, '--b', label='f(x) three points')
plt.legend()
```

<matplotlib.legend.Legend at 0x7f4fcb9e9cd0>



#Q6

```
def f(x):
    return 1/(1+(x-np.pi)**2)
def midpoint(f, a, b, n):
    h = float(b-a)/n
    result = 0
    for i in range(n):
        result += f((a + h/2.0) + i*h)
    result *= h
    return result

def trapz(f,a,b,N=50):

    x = np.linspace(a,b,N+1) # N+1 points make N subintervals
    y = f(x)
    y_right = y[1:] # right endpoints
    y_left = y[:-1] # left endpoints
    dx = (b - a)/N
    T = (dx/2) * np.sum(y_right + y_left)
    return T
def simpson(f,a,b,N=50):

    if N % 2 == 1:
        raise ValueError("N must be an even integer.")
    dx = (b-a)/N
    x = np.linspace(a,b,N+1)
    y = f(x)
    S = dx/3 * np.sum(y[0:-1:2] + 4*y[1::2] + y[2::2])
    return S
def Composite_simpson(f,a,b,N, tol):
    exactValue = np.arctan(5-np.pi)+np.arctan(np.pi)
    absError = 0
    for n in range(2, N, 2):
        h = (b-a)/n
        xI0 = f(a) + f(b)
        xI1 = 0
        xI2 = 0
        # Use simpsons composite rule to eventually approximate the
integral
        for i in range(1,n):
            x = a + i*h
            if i%2 == 0: #even
                xI2 = xI2 + f(x)
            else:
                xI1 = xI1 + f(x)
        xI = h*(xI0 + 2*xI2 + 4*xI1)/3 # our approximation
        absError = abs(xI - exactValue)
        if absError < tol:
            break
```

```

    return absError
n=50
a = 0
b = 5
exactValue = np.arctan(5-np.pi)+np.arctan(np.pi)
f_midpoint= midpoint(f, a, b, n)
f_trapz = trapz(f,a,b,n)
f_simps = simps(f,a,b,n)
tol = 1e-4
f_c_simp_error =Composite_simps(f,a,b,n, tol)
print("Mid Point Rule error      : ", abs(exactValue-f_midpoint) )
print("Trapzoid rule error      : ", abs(exactValue-f_trapz))
print("Simpson rule error       : ", abs(exactValue-f_simps))
print("Composite simpson rule error : ",f_c_simp_error )

Mid Point Rule error      :    0.0001001957355359906
Trapzoid rule error      :    0.00020042542611919956
Simpson rule error       :    1.809228722393641e-07
Composite simpson rule error :    3.7766006237838212e-06

```

Q7

```

import math
from matplotlib import pyplot as plt

```

```

def step_one(t, p):
    """
    M = mean anomaly
    M = 2pi * t
    -----
    P
    """
    return (2 * math.pi * t) / p

def step_two(m, e):
    """
    M = mean anomaly
    E = eccentric anomaly
    e = eccentricity

    M = E - esinE
    """
    def M(E): return E - (e * math.sin(E))
    E = 0
    while m > M(E):
        E += 1
    while M(E) > m:
        E -= 0.00001
    return E

```



```

def step_three(e, E):
    """
     $(1 - e)\tan^2(\theta/2) = (1 + e)\tan^2(E/2)$ 
    e = eccentricity
    theta = true anomaly
    E = eccentric anomaly
    """
    def l(theta): return (1-e)*(math.tan(theta/2))**2
    r = (1+e)*(math.tan(E/2))**2
    theta = 0
    while l(theta) < r:
        theta += 0.1
    while r < l(theta):
        theta -= 0.00001
    return [theta, 2*(math.pi - theta) + theta]

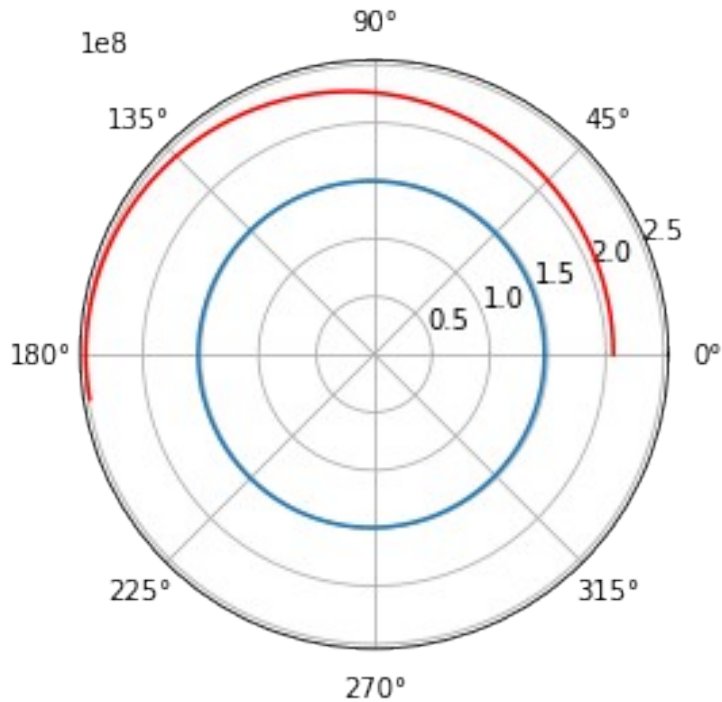
def step_four(a, e, E):
    """
    a = semi-major axis
    e = eccentricity
    E = eccentric anomaly

     $r = a(1 - e\cos E)$ 
    """
    return a * (1 - (e * math.cos(E)))

def calculate(e, t, p, a):
    M = step_one(t, p)
    E = step_two(M, e)
    if list(math.modf(float(t) / p))[0] > 0.5:
        theta = step_three(e, E)[1]
    if list(math.modf(float(t) / p))[0] < 0.5:
        theta = step_three(e, E)[0]
    r = step_four(a, e, E)
    return [theta, r]

e_theta, e_r = [], []
m_theta, m_r = [], []
for x in range(0, 365):
    e_coords = calculate(0.0167, x, 365, 1.496E8)
    e_theta.append(e_coords[0])
    e_r.append(e_coords[1])
    m_coords = calculate(0.0935, x, 687, 2.2792E8)
    m_theta.append(m_coords[0])
    m_r.append(m_coords[1])
plt.polar(e_theta, e_r)
plt.polar(m_theta, m_r, 'r')
plt.show()

```



Q8:

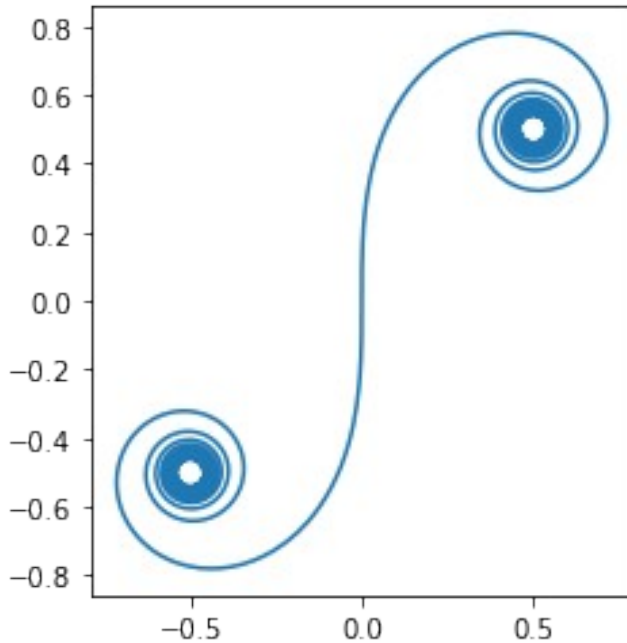
```
from scipy.special import fresnel
from scipy import linspace
import matplotlib.pyplot as plt
```

```
t = linspace(-8, 8, 1000)
s, c = fresnel(t)
```

```
plt.plot(s, c)
plt.axes().set_aspect("equal")
plt.show()
# S is smoother
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:5:
DeprecationWarning: scipy.linspace is deprecated and will be removed
in SciPy 2.0.0, use numpy.linspace instead
"""
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:9:
MatplotlibDeprecationWarning: Adding an axes using the same arguments
as a previous axes currently reuses the earlier instance. In a future
version, a new instance will always be created and returned.
Meanwhile, this warning can be suppressed, and the future behavior
ensured, by passing a unique label to each axes instance.
if __name__ == '__main__':
```



Q9

```

from scipy.optimize import curve_fit
from matplotlib import pyplot
from numpy import arange
def objective(strain, a,b):
    # integral of given form: stress = (a*strain)/(1-b*strain)
    return a*strain/(1-b*strain)
x = [0,
0.05 ,
0.10 ,
0.15 ,
0.20 ,
0.25 ,
0.30 ,
0.35 ,
0.40 ,
0.45 ,
0.50 ]
y= [0, 0.252,0.531,0.840,1.184,1.558,1.975,2.444,2.943, 3.500,4.115]

popt, _ = curve_fit(objective, x, y)
# summarize the parameter values

a, b= popt
print('y = %.5f * x + %.5f' % (a, b))
# plot input vs output
pyplot.scatter(x, y)
# define a sequence of inputs between the smallest and largest known

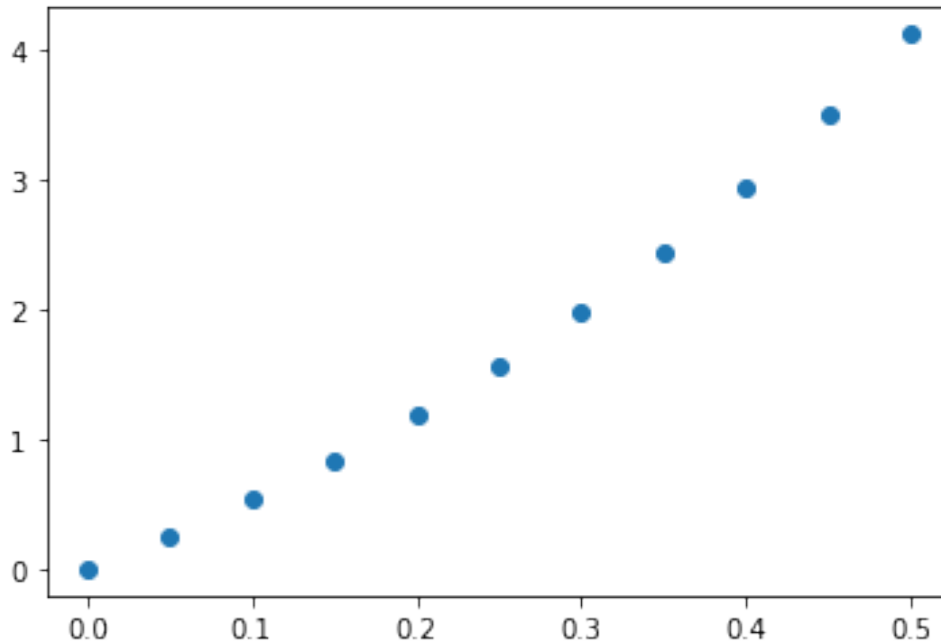
```

```

inputs
x_line = arange(min(x), max(x), 1)
# calculate the output for the range
y_line = objective(x_line, a, b)
# create a line plot for the mapping function
pyplot.plot(x_line, y_line, '--', color='red')
pyplot.show()

y = 5.03745 * x + 0.78037

```



Q10

Euler forward method

```

import numpy as np
import matplotlib.pyplot as plt

plt.style.use('seaborn-poster')
%matplotlib inline

# Define parameters
f = lambda y, t: 0.2*y - 0.01*y**2 + np.sin(2*np.pi*t) # ODE
h = 0.2 # Step size
t = np.arange(2000, 2020, h) # Numerical grid
s0 = 6 # Initial Condition

# Explicit Euler Method
s = np.zeros(len(t))
s[0] = s0

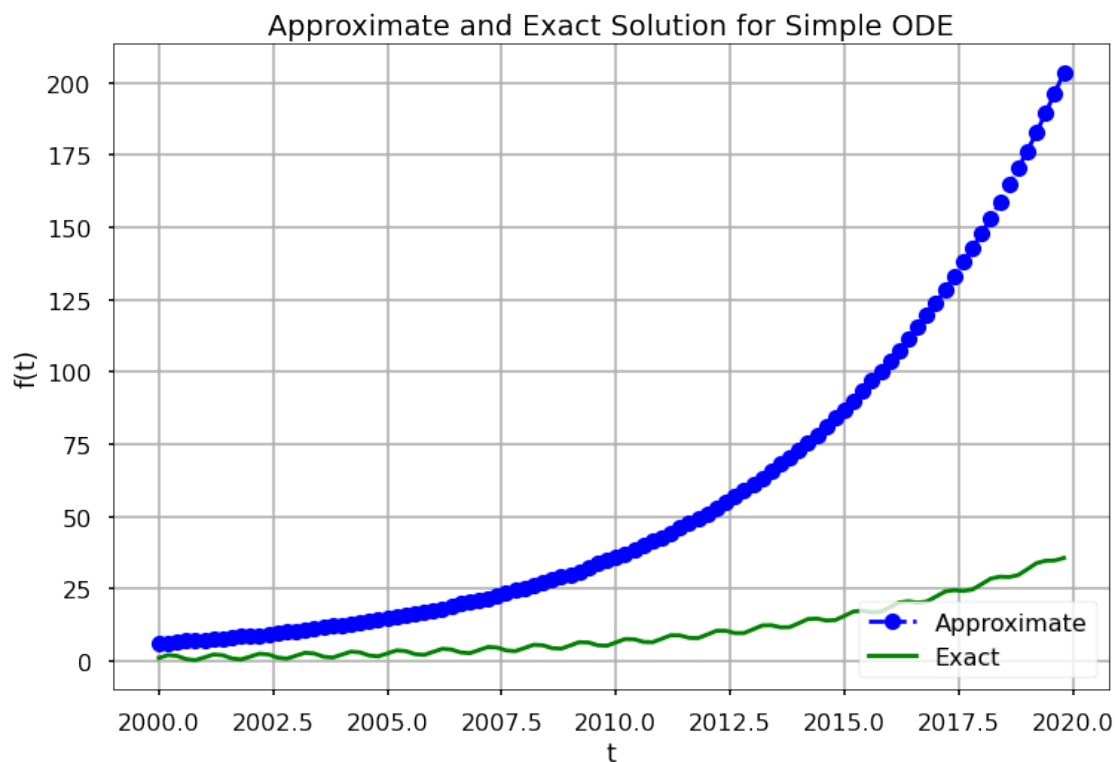
```

```

for i in range(0, len(t) - 1):
    s[i + 1] = s[i] + h*f(s[i], t[i])

plt.figure(figsize = (12, 8))
plt.plot(t, s, 'bo--', label='Approximate')
plt.plot(t, f(s,t), 'g', label='Exact')
plt.title('Approximate and Exact Solution \
for Simple ODE')
plt.xlabel('t')
plt.ylabel('f(t)')
plt.grid()
plt.legend(loc='lower right')
plt.show()

```



Euler Backward

```

def backward_euler(f_prime, y_0, endpoints, h):
    n = int((endpoints[-1] - endpoints[0])/h)
    x = endpoints[0]
    y = y_0
    x_out, y_out = np.array([x]), np.array([y])
    for i in range(n):
        y_prime = f_prime(x, y)/(1 + h * 8)
        y += h * y_prime
        x += h
        x_out = np.append(x_out, x)

```

```

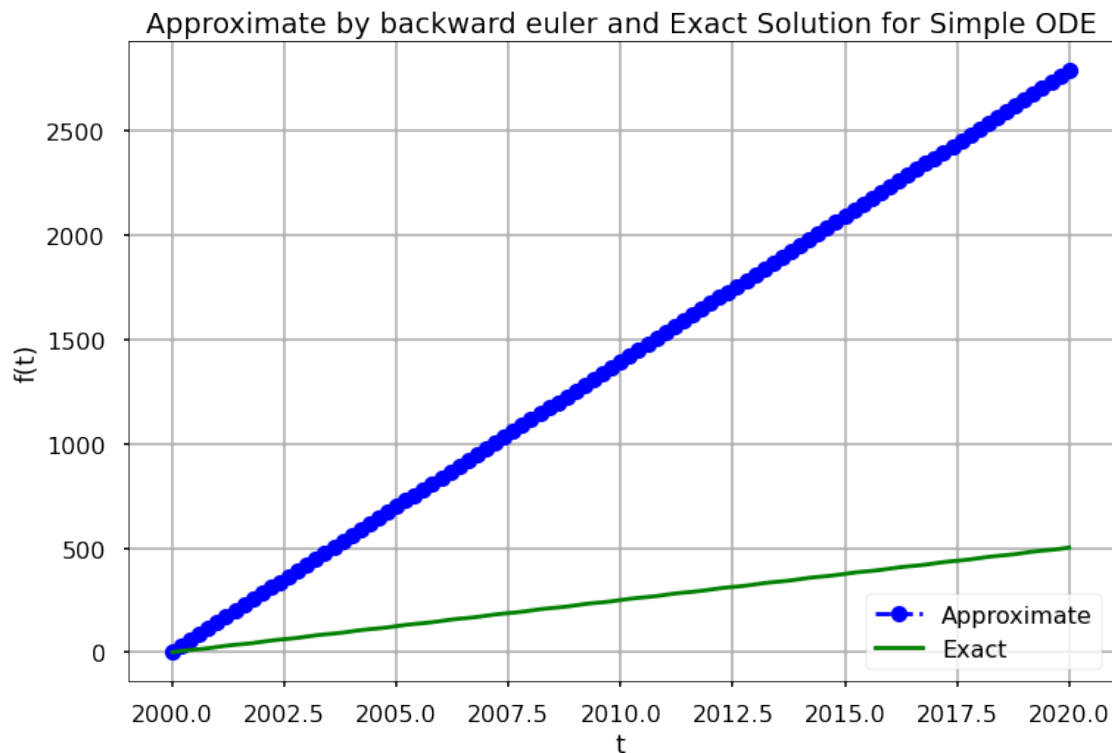
y_out = np.append(y_out, y)

    return x_out, y_out
f = lambda y, t: 0.2*y - 0.01*y**2 + np.sin(2*np.pi*t) # ODE
h = 0.2 # Step size
t = np.arange(2000, 2020, h) # Numerical grid
s0 = 6 # Initial Condition

t, s = backward_euler(f, s0, [2000,2020], h)

plt.figure(figsize = (12, 8))
plt.plot(t, s, 'bo--', label='Approximate')
plt.plot(t, f(s,t), 'g', label='Exact')
plt.title('Approximate by backward euler and Exact Solution \
for Simple ODE')
plt.xlabel('t')
plt.ylabel('f(t)')
plt.grid()
plt.legend(loc='lower right')
plt.show()

```



4th Order Runge Kutta Method

```
def rk4(f,x0,y0,xn,n):
    n = int((xn-x0)/h)
    x = []
    y= []
    print('\n-----SOLUTION-----')
    print('-----')
    print('x0\ty0\tyn')
    print('-----')
    for i in range(n):
        x.append(x0)
        y.append(y0)
        k1 = h * (f(x0, y0))
        k2 = h * (f((x0+h/2), (y0+k1/2)))
        k3 = h * (f((x0+h/2), (y0+k2/2)))
        k4 = h * (f((x0+h), (y0+k3)))
        k = (k1+2*k2+2*k3+k4)/6
        yn = y0 + k
        print('%.4f\t%.4f\t%.4f'% (x0,y0,yn) )
        print('-----')
        y0 = yn
        x0 = x0+h

    return x,y

f = lambda y, t: 0.2*y- 0.01*y*2 + np.sin(2*np.pi*t) # ODE
h = 0.2 # Step size
t = np.arange(2000, 2020, h) # Numerical grid
s0 = 6 # Initial Condition

t, s = rk4(f,2000,6,2020,h)

plt.figure(figsize = (12, 8))
plt.plot(t, s, 'bo--', label='Approximate')
plt.title('Approximate by 4th Order Runge Kutta Method and Exact
Solution \
for Simple ODE')
plt.xlabel('t')
plt.ylabel('f(t)')
plt.grid()
plt.legend(loc='lower right')
plt.show()
```

```
-----SOLUTION-----
-----
x0      y0      yn
-----
2000.0000  6.0000  78.0056
```

```
-----
2000.2000  78.0056   150.0391
-----
2000.4000  150.0391   222.1586
-----
2000.6000  222.1586   294.3670
-----
2000.8000  294.3670   366.4768
-----
2001.0000  366.4768   438.5154
-----
2001.2000  438.5154   510.5312
-----
2001.4000  510.5312   582.5402
-----
2001.6000  582.5402   654.5471
-----
2001.8000  654.5471   726.5534
-----
2002.0000  726.5534   798.5595
-----
2002.2000  798.5595   870.5657
-----
2002.4000  870.5657   942.5719
-----
2002.6000  942.5719  1014.5782
-----
2002.8000  1014.5782  1086.5847
-----
2003.0000  1086.5847  1158.5913
-----
2003.2000  1158.5913  1230.5980
-----
2003.4000  1230.5980  1302.6050
-----
2003.6000  1302.6050  1374.6121
-----
2003.8000  1374.6121  1446.6195
-----
2004.0000  1446.6195  1518.6272
-----
2004.2000  1518.6272  1590.6352
-----
2004.4000  1590.6352  1662.6436
-----
2004.6000  1662.6436  1734.6525
-----
2004.8000  1734.6525  1806.6620
-----
2005.0000  1806.6620  1878.6722
```

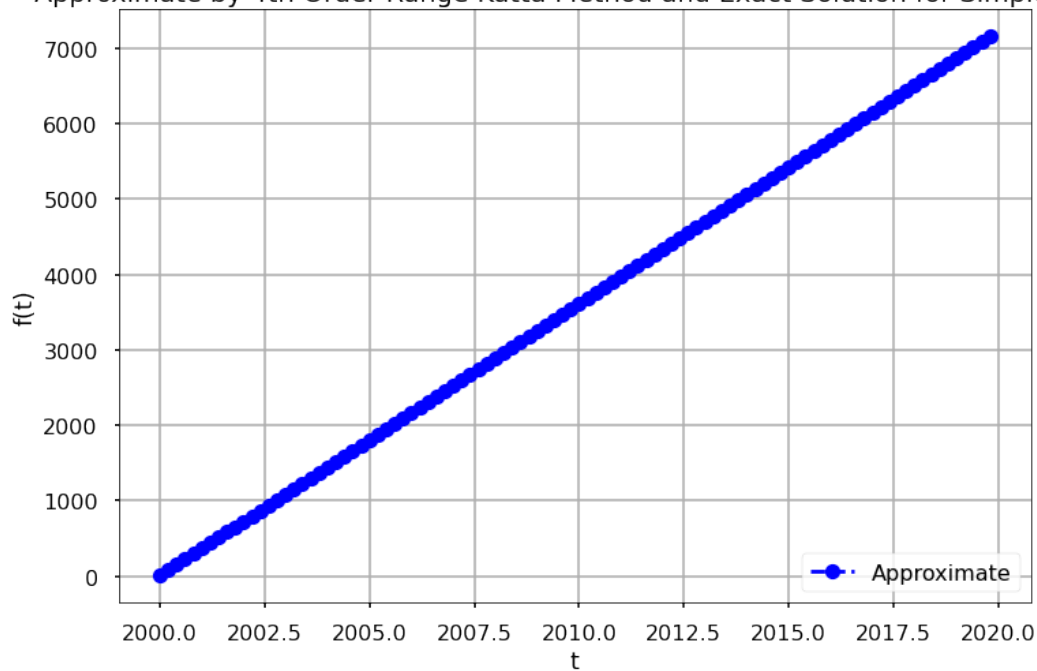


```
-----
2005.2000  1878.6722  1950.6833
-----
2005.4000  1950.6833  2022.6957
-----
2005.6000  2022.6957  2094.7098
-----
2005.8000  2094.7098  2166.7263
-----
2006.0000  2166.7263  2238.7468
-----
2006.2000  2238.7468  2310.7743
-----
2006.4000  2310.7743  2382.8162
-----
2006.6000  2382.8162  2454.8985
-----
2006.8000  2454.8985  2527.1370
-----
2007.0000  2527.1370  2599.5172
-----
2007.2000  2599.5172  2671.6655
-----
2007.4000  2671.6655  2743.7448
-----
2007.6000  2743.7448  2815.8288
-----
2007.8000  2815.8288  2888.0001
-----
2008.0000  2888.0001  2960.4123
-----
2008.2000  2960.4123  3032.6541
-----
2008.4000  3032.6541  3104.7694
-----
2008.6000  3104.7694  3176.9155
-----
2008.8000  3176.9155  3249.2923
-----
2009.0000  3249.2923  3321.6292
-----
2009.2000  3321.6292  3393.7781
-----
2009.4000  3393.7781  3465.9780
-----
2009.6000  3465.9780  3538.4340
-----
2009.8000  3538.4340  3610.7035
-----
2010.0000  3610.7035  3682.8817
```

```
-----
2010.2000  3682.8817  3755.2845
-----
2010.4000  3755.2845  3827.6615
-----
2010.6000  3827.6615  3899.8581
-----
2010.8000  3899.8581  3972.2618
-----
2011.0000  3972.2618  4044.6672
-----
2011.2000  4044.6672  4116.8903
-----
2011.4000  4116.8903  4189.3538
-----
2011.6000  4189.3538  4261.7230
-----
2011.8000  4261.7230  4334.0062
-----
2012.0000  4334.0062  4406.5458
-----
2012.2000  4406.5458  4478.8242
-----
2012.4000  4478.8242  4551.2681
-----
2012.6000  4551.2681  4623.7136
-----
2012.8000  4623.7136  4696.0460
-----
2013.0000  4696.0460  4768.6084
-----
2013.2000  4768.6084  4840.9126
-----
2013.4000  4840.9126  4913.4675
-----
2013.6000  4913.4675  4985.8278
-----
2013.8000  4985.8278  5058.3395
-----
2014.0000  5058.3395  5130.7862
-----
2014.2000  5130.7862  5203.2803
-----
2014.4000  5203.2803  5275.7713
-----
2014.6000  5275.7713  5348.2713
-----
2014.8000  5348.2713  5420.7787
-----
2015.0000  5420.7787  5493.3046
```

```
-----
2015.2000  5493.3046  5565.8065
-----
2015.4000  5565.8065  5638.3692
-----
2015.6000  5638.3692  5710.8467
-----
2015.8000  5710.8467  5783.4511
-----
2016.0000  5783.4511  5855.8993
-----
2016.2000  5855.8993  5928.5512
-----
2016.4000  5928.5512  6000.9956
-----
2016.6000  6000.9956  6073.6690
-----
2016.8000  6073.6690  6146.1959
-----
2017.0000  6146.1959  6218.8036
-----
2017.2000  6218.8036  6291.4365
-----
2017.4000  6291.4365  6363.9404
-----
2017.6000  6363.9404  6436.6498
-----
2017.8000  6436.6498  6509.2187
-----
2018.0000  6509.2187  6581.8475
-----
2018.2000  6581.8475  6654.5501
-----
2018.4000  6654.5501  6727.0944
-----
2018.6000  6727.0944  6799.7888
-----
2018.8000  6799.7888  6872.4787
-----
2019.0000  6872.4787  6945.0399
-----
2019.2000  6945.0399  7017.7716
-----
2019.4000  7017.7716  7090.4765
-----
2019.6000  7090.4765  7163.0656
-----
2019.8000  7163.0656  7235.8054
-----
```

Approximate by 4th Order Runge Kutta Method and Exact Solution for Simple ODE



Trapezoidal

```
def trapz(f,a,b,t,N=50):
```

```
    x = np.linspace(a,b,N) # N+1 points make N subintervals
    y = f(x,t)
    y_right = y[1:] # right endpoints
    y_left = y[:-1] # left endpoints
    dx = (b - a)/N
    T = (dx/2) * np.sum(y_right + y_left)
    return x,y
f = lambda y, t: 0.2*y- 0.01*y*2 + np.sin(2*np.pi*t) # ODE
h = 0.2 # Step size
t = np.arange(2000, 2020, h) # Numerical grid
s0 = 6 # Initial Condition
```

```
t, s = trapz(f, 2000, 2020, t, int((2020-2000)/h))
```

```
plt.figure(figsize = (12, 8))
plt.plot(t, s, 'bo--', label='Approximate')
plt.plot(t, f(s,t), 'g', label='Exact')
plt.title('Approximate by trapezoid and Exact Solution \
for Simple ODE')
plt.xlabel('t')
plt.ylabel('f(t)')
plt.grid()
```

```
plt.legend(loc='lower right')  
plt.show()
```

