

Assignment 2

Q1.

From solution to Griffith 2.7, we have

$$E_z = \frac{1}{4\pi\epsilon_0} (2\pi R^2 \sigma) \int_0^\pi \frac{(z - R\cos\theta)\sin\theta}{(R^2 + z^2 - 2Rz\cos\theta)^{\frac{3}{2}}} d\theta$$

Let $u = \cos\theta$

$$= \frac{1}{4\pi\epsilon_0} (2\pi R^2 \sigma) \int_{-1}^1 \frac{z - Ru}{(R^2 + z^2 - 2Rzu)^{\frac{3}{2}}} du$$

From this we can set up our integrator.

Assume all coefficients are 1

```
import numpy as np
from matplotlib import pyplot as plt
from scipy import integrate
#define function to integrate according to the solution to Griffith 2.7
def fun_int(x,z):
    return (z-x)/((1+(z**2)-(2*z*x))**(3/2))

#define integrator that uses Simpsons rule
def integrator(z):
    x = np.linspace(-1,1,1001) #integrate from -1 to 1
    y = fun_int(x,z)
    dx=np.median(np.diff(x))
    #integrate according to Simpson rules
    ints_simp=dx/3.0*(y[0]+y[-1]+4*np.sum(y[1::2])+2*np.sum(y[2:-1:2]))
    return ints_simp

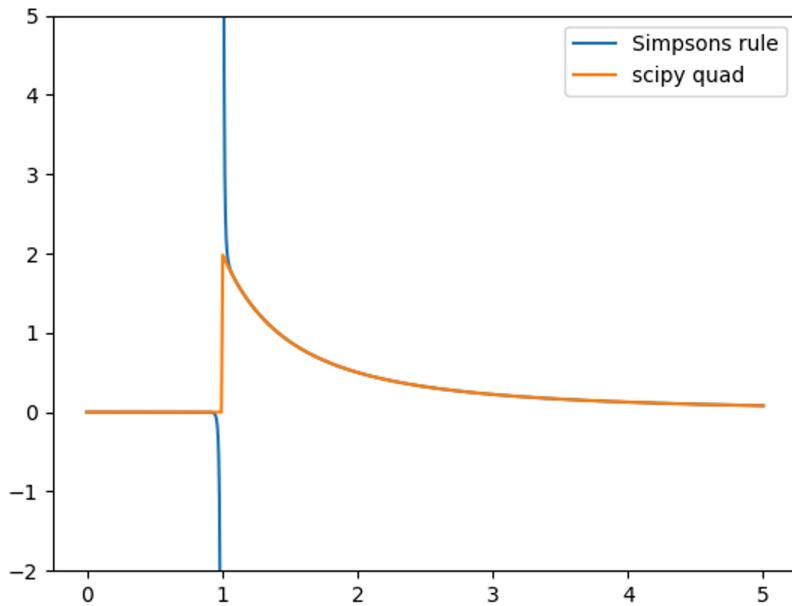
simp_val = []
quad_val = []
z_arr = np.linspace(0,5,1001)
for z in z_arr: #note z = 1 is included in this linspace
    simp_temp = integrator(z)
    quad_temp,err = integrate.quad(fun_int,-1,1,args=(z,))
    simp_val = np.append(simp_val,simp_temp)
```

```

quad_val = np.append(quad_val,quad_temp)

plt.clf()
plt.plot(z_arr,simp_val)
plt.plot(z_arr,quad_val)
plt.ylim(-2, 5) #set y-axis limits so that the singularity does not make the plot
look too bad
plt.legend(['Simpsons rule','scipy quad'])
plt.savefig('question_1.png')
plt.show()

```



As we can see, there is a singularity
at $z = R$.

Our integrator using Simpsons rule
breaks up near the singularity,
while `scipy.integrate.quad` does not
care.

Our integrator breaking up is to be

expected, since near the singularity
we would have very large value
for function value due to a
small value dividing a even smaller
value, which is pretty bad for
computer calculation.

Q2

```
import numpy as np
function_call = 0 #global variable to see how many function calls we did
def exp_int(x):
    global function_call
    #update the total number of function calls
    function_call = function_call+np.size(x)
    return np.exp(x)
def integrate(fun,a,b,tol,extra=None):
    if extra is None:
        x=np.linspace(a,b,5)
        dx=x[1]-x[0]
        y=fun(x)
        #do the 3-point integral
        i1=(y[0]+4*y[2]+y[4])/3*(2*dx)
        i2=(y[0]+4*y[1]+2*y[2]+4*y[3]+y[4])/3*dx
        myerr=np.abs(i1-i2)
        if myerr<tol:
            return i2
        else:
            mid=(a+b)/2
            #pass the already calculated function value into recursive call
            int1=integrate(fun,a,mid,tol/2,extra=np.array([y[0], y[1], y[2]]))
            int2=integrate(fun,mid,b,tol/2,extra=np.array([y[2], y[3], y[4]]))
            return int1+int2
    else:
        x=np.linspace(a,b,5)
        #we already have the 1st, middle and last function value from extra
        #need only to calculate 2nd and 4th value
        y=np.array([extra[0],0,extra[1],0,extra[2]])
```

```

y[1] = fun(x[1])
y[3] = fun(x[3])
dx=x[1]-x[0]
i1=(y[0]+4*y[2]+y[4])/3*(2*dx)
i2=(y[0]+4*y[1]+2*y[2]+4*y[3]+y[4])/3*dx
myerr=np.abs(i1-i2)
if myerr<tol:
    return i2
else:
    mid=(a+b)/2
    int1=integrate(fun,a,mid,tol/2,extra=np.array([y[0], y[1], y[2]]))
    int2=integrate(fun,mid,b,tol/2,extra=np.array([y[2], y[3], y[4]]))
    return int1+int2

def integrate_old(fun,a,b,tol):
    x=np.linspace(a,b,5)
    dx=x[1]-x[0]
    y=fun(x)
    #do the 3-point integral
    i1=(y[0]+4*y[2]+y[4])/3*(2*dx)
    i2=(y[0]+4*y[1]+2*y[2]+4*y[3]+y[4])/3*dx
    myerr=np.abs(i1-i2)
    if myerr<tol:
        return i2
    else:
        mid=(a+b)/2
        int1=integrate_old(fun,a,mid,tol/2)
        int2=integrate_old(fun,mid,b,tol/2)
        return int1+int2
ans_new = integrate(exp_int,1,2,1e-6)
print(ans_new)
# ans_old = integrate_old(exp_int,1,2,1e-6)
# print(ans_old)
print(function_call)

```

we can compare our number of function calls by commenting out different part of the last section of our code.

For our new code :

The integrated value is: 4.670774295215381

The number of function calls is: 33

For the old code in class :

The integrated value is: 4.670774295215381

The number of function calls is: 75

We can see that the value is unchanged, but the number of function calls is drastically reduced.

Q3 :

For cheb fit of \log_2 from 0.5 to 1, we need to renormalize our x value to be

$4x - 3$ to have x range from -1 to 1.

```
import numpy as np
from matplotlib import pyplot as plt
x_true = np.linspace(0.5, 1, 10001) # x array from 0.5 to 1
# renormalize x array to be from -1 to 1 for cheb fit
x_renorm = x_true * 4 - 3
y = np.log2(x_true)
cheb_coeff = np.polynomial.chebyshev.chebfit(x_renorm, y, 30)
print(cheb_coeff)
```

The cheb coefficient of degree 30 is

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```
[ -4.56893394e-01 4.95054673e-01 -4.24689768e-02 4.85768297e-03  
-6.25084976e-04 8.57981013e-05 -1.22671891e-05 1.80404307e-06  
-2.70834256e-07 4.13047281e-08 -6.37810023e-09 9.94832637e-10  
-1.56469112e-10 2.47866900e-11 -3.95524365e-12 6.38836021e-13  
-1.08832208e-13 2.28575622e-14 -9.50961038e-15 6.55421517e-15  
-6.32919459e-15 5.86956403e-15 -5.59628545e-15 5.11556488e-15  
-5.39073805e-15 4.30425692e-15 -4.66669035e-15 3.68445846e-15  
-4.49188576e-15 2.67766742e-15 -2.39329224e-15]
```

From this we see that to have an error less than 10^{-6} , it is enough to only pick the first 8 coefficients.

```
nuse = 8  
cheb_coeff_cut = cheb_coeff[:nuse]
```

For natural log, we could use `numpy.frexp(x)`. It split a float x into mantissa and exponent such that

$$x = \text{mantissa} \times 2^{\text{exponent}}$$

Note that for positive number, `frexp` always gives mantissa to be from 0.5 to 1.

Now

$$\log_2(x) = \log_2(\text{mantissa} \times 2^{\text{exponent}})$$

$$= \log_2(\text{mantissa}) + \log_2(2^{\text{exponent}})$$

$$= \log_2(\text{mantissa}) + \text{exponent}.$$

Then we could use

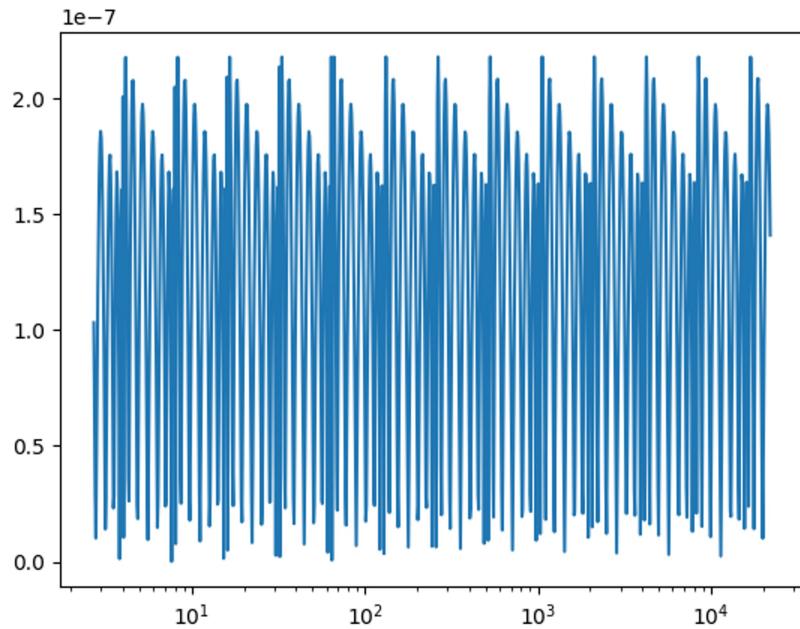
$$\log_e(x) = \frac{\log_2(x)}{\log_2(e)}$$

to get natural log.

```
def mylog2(x):
    #stop calculation if any of the x is smaller or equal to zero
    if np.any(x<=0):
        print('natural log only accept positive number')
    else:
        #use frexp to get mantissa and exponent
        mantissa, exponent = np.frexp(x)
        #use the previously calculated cheb coefficients
        global cheb_coeff_cut
        #renormalize the mantissa for chebval
        mantissa_renorm = mantissa*4-3
        mantissa_log =
    np.polynomial.chebyshev.chebval(mantissa_renorm, cheb_coeff_cut)
        #from our previous calculation
        log2_val = mantissa_log + exponent
        #convert to natural log
        natural_log = log2_val/np.log2(np.exp(1))
        return natural_log
#test the error of mylog2
y_test_val_true = np.linspace(1,10,1001)
x_test_val = np.exp(y_test_val_true)
y_test_val_cheb = mylog2(x_test_val)
err = np.abs(y_test_val_true-y_test_val_cheb)
print('the std of error is: ', np.std(err))
plt.clf()
plt.xscale('log')
plt.plot(x_test_val,err)
plt.savefig('question_3.png')
plt.show()
```

The output is :

the std of error is: 6.024236269321544e-08



we can see that indeed we have
a natural log calculator with
error less than 10^{-6}