

Homework 7

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MATH 4650

1)

```
In [6]: from scipy.special import roots_laguerre
import numpy as np
```

a)

```
In [242]: import numpy as np
import math
import matplotlib.pyplot as plt

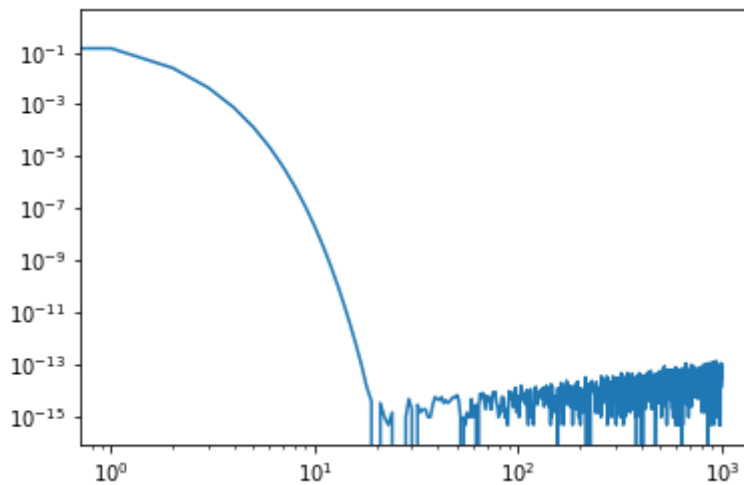
def f(x):
    return 1/(1+x**2)

def integrate(deg):
    sum = 0
    x, w = np.polynomial.legendre.leggauss(deg)
    for i in range(deg):
        sum += w[i]*f(x[i])
    return sum;

N = 1000
I = []
err = []
for i in range(1,N):
    q = 2*integrate(i)
    er = abs(q-np.pi)
    I.append(q)
    err.append(er)
```

```
In [243]: plt.loglog(err)
```

```
Out[243]: [matplotlib.lines.Line2D at 0x114eaa240]
```



```
In [248]: print('The min error is ', min(err), 'and occurs when d = ', err.index(min(err)))
```

```
The min error is 0.0 and occurs when d = 20
```

It looks like as the number of degrees increases we quickly become more accurate and we hover around an error of 10^{-15} . That being said Because π is transcendental when we have an error of 0 im sure we are just at machine precision.

b)

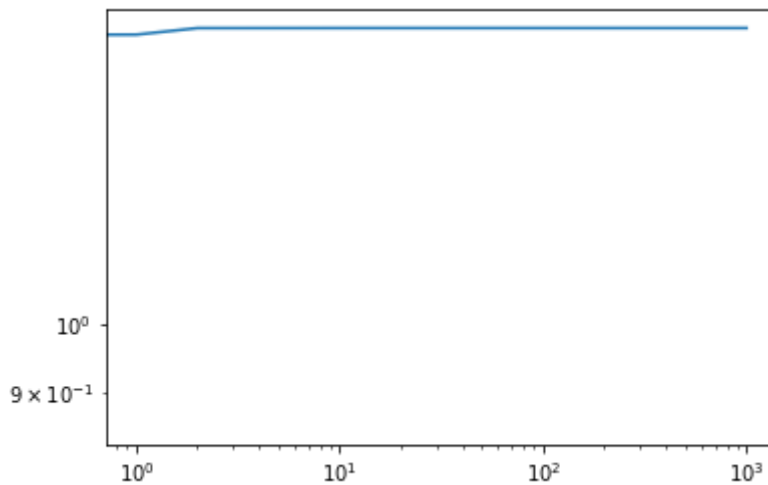
Note, with $x = e^t - 1$, our function is now e^{-t}

```
In [283]: f = lambda x: math.exp(-x)
def integrate(deg):
    sum = 0
    x, w = np.polynomial.legendre.leggauss(deg)
    for i in range(deg):
        sum += w[i]*f(x[i])
    return sum;

N = 1000
I = []
err = []
for i in range(1,N):
    q = 2*integrate(i)
    er = abs(q-np.pi)
    I.append(q)
    err.append(er)
```

```
In [284]: plt.loglog(err)
```

```
Out[284]: [<matplotlib.lines.Line2D at 0x1515b82b70>]
```



Well this looks very odd. It appears that our error is fixed at a rather large value no matter what d is. I will talk more about what I think is happening in part c.

c)

So for part b I think we are seeing some weird things happening with the weights from Legendre polynomial and our function that is now just e^{-x} . I think this has to do with the fact that the weights are orthogonal to the weight function, which just so happens to be e^{-x} so I think we are getting some weird cancellation. So for this problem I don't think the change of variables would be the best and I would just use part a instead.

2

a)

```
In [202]: from scipy.integrate import quadrature, quad, dblquad
import time
```

```

In [183]: def baseQuadratureRule(f,a,b):
    Q, errEstimate = quad(f,a,b,epsabs=1.5e-8,epsrel=1.5e-8,limit=50)
    return Q

def my2D_quadrature(f,a,b,c,d, baseQuadRule ):
    @np.vectorize
    def g(x):

        def f_as_fcn_of_y(y):
            return f(x,y)

        Q = baseQuadRule( f_as_fcn_of_y, c, d )
        return Q

    Q = baseQuadRule( g, a, b )
    return Q

def my3D_quadrature(f,a1,b1,a2,b2,a3,b3, my2D_quadrature ):
    @np.vectorize
    def g(x):

        def f_as_fcn_of_yz(y,z):
            return f(x,y,z)

        Q = my2D_quadrature( f_as_fcn_of_yz, a2, b2, a3, b3, baseQuadratureRule )
        return Q

    Q = baseQuadratureRule( g, a1, b1 )
    return Q

def my4D_quadrature(f,a1,b1,a2,b2,a3,b3,a4,b4, my3D_quadrature ):
    @np.vectorize
    def g(x):

        def f_as_fcn_of_yzw(y,z,w):
            return f(x,y,z,w)

        Q = my3D_quadrature( f_as_fcn_of_yzw, a2, b2, a3, b3, a4, b4, my2D_quadrature )
        return Q

    Q = baseQuadratureRule( g, a1, b1 )
    return Q

```

b)

```
In [147]: from scipy.linalg import hilbert
d=2
A = hilbert(d)
def f(x,y):
    args = np.array([x,y])
    return args.T.dot(args.dot(A))
def p(x,y):
    return((1/(2*np.pi)**(2/2))*np.exp(-(x**2+y**2)/2))
def g(x,y):
    return(f(x,y)*p(x,y))
#my2D_quadrature(g,-5,5,-5,5, baseQuadratureRule(g, -5,5))
```

```
In [148]: my2D_quadrature(g,-5,5,-5,5, baseQuadratureRule)
```

```
Out[148]: 1.3333119816098895
```

```
In [174]: d=3
A = hilbert(d)
def f(x,y,z):
    args = np.array([x,y,z])
    return args.T.dot(args.dot(A))
def p(x,y,z):
    return((1/(2*np.pi)**(3/2))*np.exp(-(x**2+y**2+z**2)/2))
def g(x,y,z):
    return(f(x,y,z)*p(x,y,z))
```

```
In [179]: my3D_quadrature(g,-5,5,-5,5,-5,5, my2D_quadrature )
```

```
Out[179]: 1.533307899800628
```

```
In [184]: d=4
A = hilbert(d)
def f(x,y,z,w):
    args = np.array([x,y,z,w])
    return args.T.dot(args.dot(A))
def p(x,y,z,w):
    return((1/(2*np.pi)**(4/2))*np.exp(-(x**2+y**2+z**2+w**2)/2))
def g4(x,y,z,w):
    return(f(x,y,z,w)*p(x,y,z,w))
```

```
In [185]: my4D_quadrature(g4,-5,5,-5,5,-5,5,-5,5,my3D_quadrature)
```

```
Out[185]: 1.6761617121244665
```

i)

As we can see from the line of code above, $E[f(x)] \approx 1.6761676$, when $d = 4$.

ii)

My error tolerance (both absolute and relative) for 1D is very low so it doesn't surprise that 4D took a very long time to run (it was approx. 5 min). For both 2D and 3D the functions were evaluated much quicker as would be expected in a problem like this.

c)

```
In [219]: def MonteCarlo(d, N, Verbose = True):
            start_time = time.time()
            A = hilbert(d)
            I = 0.0
            for i in range(N):
                x = np.random.randn(d)
                f = x.T.dot(x.dot(A))
                I += f
            I = (1/N)*I
            end_time = time.time()
            if Verbose == True:
                print('Execution time is:', end_time-start_time, ' seconds')
            return I
```

Running our MonteCarlo for d = 2,3,4:

```
In [218]: I2 = MonteCarlo(2,10**6, Verbose = False)
            I3 = MonteCarlo(3,10**6, Verbose = False)
            I4 = MonteCarlo(4,10**6, Verbose = False)

            print("For d=2: ", I2)
            print("For d=3: ", I3)
            print("For d=4: ", I4)

            For d=2:  1.3342270837817078
            For d=3:  1.53219163268725
            For d=4:  1.678155905224539
```

Now finding the runtime and error for d = 4,12,100

```
In [221]: #runtime and error for d=4  
I4 = MonteCarlo(4,10**6, Verbose = True)  
print('Error for d=4 is:', abs(I4 - 1.6761617121244665))
```

```
Excution time is: 2.024691104888916  seconds  
Error for d=4 is: 0.00175656939114055
```

```
In [222]: #runtime and error for d=12  
I12 = MonteCarlo(12,10**6, Verbose = True)  
print('Error for d=12 is:', abs(I12 - 2.224352838648))
```

```
Excution time is: 2.243312120437622  seconds  
Error for d=12 is: 0.0021304304003479935
```

```
In [224]: #runtime and error for d=100  
I100 = MonteCarlo(100,10**6, Verbose = True)  
print('Error for d=100 is:', abs(I100 - 3.284342189302))
```

```
Excution time is: 5.517652988433838  seconds  
Error for d=100 is: 0.002000626990464216
```

d)

It seems like quadrature is probably more accurate than Monte Carlo is expically in low levels of d. However quadrature would probably be almost impossible to run on my compute for d =100 ao I think when we have igh dimensions Monte Carlo is the way to go!