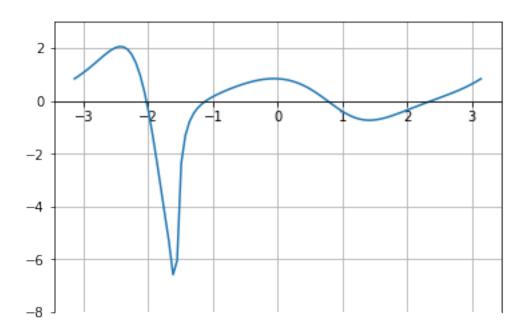
# MACS\_PS\_2

## January 21, 2019

# 0.1 Name: Hyunwoo Roh

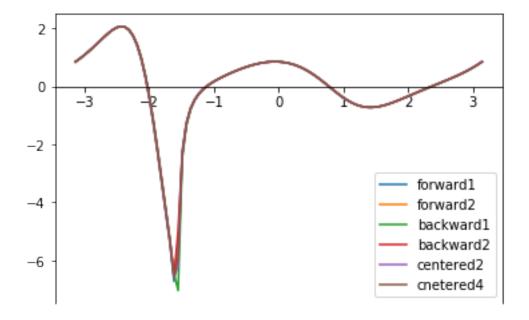
```
In [ ]: from matplotlib import pyplot as plt
      from sympy import *
      import sympy as sy
      import numpy as np
      import math
Problem 1 #
        x = sy.symbols('x')
        y = (\sin(x)+1)**(\sin(\cos(x)))
Out [504]: (\sin(x) + 1)**\sin(\cos(x))
In [520]: def sympy_diff(x_vals):
           x = sy.symbols('x')
           f = (sy.sin(x) + 1) ** (sy.sin(sy.cos(x)))
           fp = f.diff(x)
           fp_np = sy.lambdify(x, fp, 'numpy')
           return fp_np(x_vals)
In [421]: dy=sy.diff(y,x)
In [422]: dy_lamb=sy.lambdify(x,dy,'numpy')
        y_lamb=sy.lambdify(x,y,'numpy')
        xvals = np.linspace(-np.pi, np.pi, 100)
        dy_lamb(5)
Out[422]: -0.4117441338415354
In [423]: ax = plt.gca()
        ax.spines["bottom"].set_position("zero")
```

```
ax.plot(xvals,dy_lamb(xvals))
plt.ylim(-8,3)
plt.grid()
plt.show()
```



```
return (f(x+h)-f(x-h))/(2*h)
          def ct4(f, x, h):
              return (f(x-2*h)-8.*f(x-h)+8.*f(x+h)-f(x+2*h))/(12*h)
          def centered_o4(f, x, h):
             num = f(x - 2 * h) - 8 * f(x - h) + 8 * f(x + h) - f(x + 2 * h)
              return num / (12 * h)
In [425]: # Plot the results
          xvals=x_vals(100)
          ax = plt.gca()
          ax.spines["bottom"].set_position("zero")
          ax.plot(xvals, fw1(y_lamb,xvals,0.01),label='forward1')
          ax.plot(xvals, fw2(y_lamb,xvals,0.01),label='forward2')
          ax.plot(xvals, bw1(y_lamb,xvals,0.01),label='backward1')
          ax.plot(xvals, bw2(y_lamb,xvals,0.01),label='backward2')
          ax.plot(xvals, ct2(y_lamb,xvals,0.01),label='centered2')
          ax.plot(xvals, ct4(y_lamb,xvals,0.01),label='cnetered4')
          plt.legend(loc='lower right')
          plt.show
```

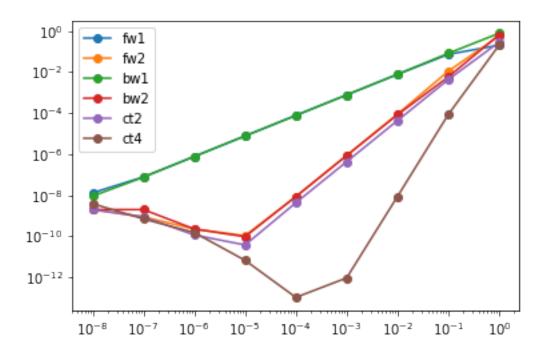
Out[425]: <function matplotlib.pyplot.show(\*args, \*\*kw)>



In [426]: # compare them to the results of Problem 1
# We can think of calculating difference from true value.

```
def test(f,n):
                               \#return sum(abs(dy_lamb(x_vals(n))-f))
                               return np.sqrt(sum((dy_lamb(x_vals(n))-f)**2))
In [427]: # calculate the error for each method when n=100
                      n=100
                      print ((test(fw1(y_lamb,x_vals(n),0.01),n),
                      test(fw2(y_lamb,x_vals(n),0.01),n),
                      test(bw1(y_lamb,x_vals(n),0.01),n),
                      test(bw2(y_lamb,x_vals(n),0.01),n),
                      test(ct2(y_lamb,x_vals(n),0.01),n),
                      test(ct4(y_lamb,x_vals(n),0.01),n)))
In [428]: # calculate the error for each method when n=300
                      n=300
                      print (test(fw1(y_lamb,x_vals(n),0.01),n),
                      test(fw2(y_lamb,x_vals(n),0.01),n),
                      test(bw1(y_lamb,x_vals(n),0.01),n),
                      test(bw2(y_lamb,x_vals(n),0.01),n),
                      test(ct2(y_lamb,x_vals(n),0.01),n),
                      test(ct4(y_lamb,x_vals(n),0.01),n))
1.7343772059566547 \ \ 0.9469287658772867 \ \ 2.735736878923265 \ \ 3.8208682613690983 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.583737575762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.5837375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \ 0.58375762903856 \ \
Problem 3
                      ############################
                       # In problem 1, we defined dy_lamb that compute the derivative.X_{-}0 = 1
                      dy_lamb(1), y_lamb(1)
Out [429]: (-0.3965403874194623, 1.3689877347067858)
In [430]: test_func = [fw1, fw2, bw1, bw2, ct2, ct4]
                      def error_f(x):
                               hvals=np.logspace(-8,0,9)
                               abs_error = np.zeros((9,6))
                               for i in range(9):
                                        for j in range(6):
                                                  abs_error[i,j]=abs(test_func[j](y_lamb,x,hvals[i])-dy_lamb(x))
                               plt.figure()
                               labels = ["fw1", "fw2", "bw1", "bw2", "ct2", "ct4"]
                               for j in range(6):
                                        plt.loglog(hvals,abs_error[:,j], label=labels[j], marker='o')
                               plt.legend(loc='upper left')
                               plt.show()
```

#### In [431]: error\_f(1)



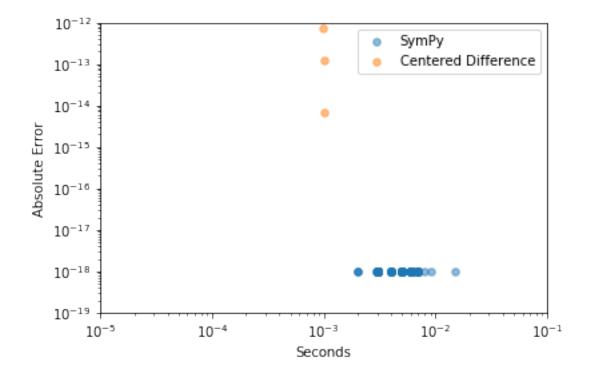
```
Problem 4
          #############################
          data = np.load(r"C:\Users\ericr\Desktop\persp-model-econ_W19\ProblemSets\PS2\plane.n
          t=data[:0]
          alpha=np.deg2rad(data[:,1])
          beta=np.deg2rad(data[:,2])
In [433]: x_pos = 500*np.tan(beta)/(np.tan(beta)-np.tan(alpha))
          y_pos = 500*np.tan(alpha)*np.tan(beta)/(np.tan(beta)-np.tan(alpha))
          x_veloc = np.zeros(len(x_pos))
          y_veloc = np.zeros(len(y_pos))
          t = 0
          while t \le 7:
              if t == 0:
                  x_{veloc[t]} = (x_{pos[t+1]} - x_{pos[t]})
                  y_{veloc}[t] = (y_{pos}[t+1] - y_{pos}[t])
              elif t < 7:
                  x_{\text{veloc}}[t] = (x_{\text{pos}}[t+1] - x_{\text{pos}}[t-1])/2
                  y_{veloc[t]} = (y_{pos[t+1]} - y_{pos[t-1]})/2
```

```
else:
                 x_{veloc[t]} = (x_{pos[t]} - x_{pos[t-1]})
                 y_{veloc}[t] = (y_{pos}[t] - y_{pos}[t-1])
             t += 1
         veloc = np.sqrt(x_veloc**2 + y_veloc**2)
         print(veloc)
[46.42420062 47.00103938 48.99880514 50.09944163 48.29035084 51.56455905
 53.92303355 51.51480057]
Problem 5
          ############################
          # Approximate the Jacobian matrix of f at x using the second order centered differen
         def jacobian(f,x_0):
             h=0.0001
             f_{len} = len(f(x_0))
             Iden = np.identity(f_len)
              jacob = np.zeros((f_len,len(x_0)))
             for i in range(f_len):
                 for j in range(len(x_0)):
                      e = Iden[i,:]
                     f_h_p = f(x_0+h*e)[j]
                     f_h_n = f(x_0-h*e)[j]
                      jacob[j,i]=((f_h_p-f_h_n)/(2*h))
             return jacob
          # Test
         def func(x):
             return np.array([x[0]**2, x[0]**3 - x[1]])
         values = np.array([1.,1.])
         jacobian(func, values)
Out[434]: array([[ 2.
                                          ],
                         , 0.
                 [3.0000001, -1.
                                         11)
In [435]: x,y=sy.symbols('x,y', real=True)
         J = Function('J')(x,y)
         f1=x**2
         f2=x**3-y
         f1x=diff(f1,x)
         f1y=diff(f1,y)
         f2x=diff(f2,x)
         f2y=diff(f2,y)
         J = sy.Matrix([[f1x,f1y],[f2x,f2y]])
         J.subs([(x,1), (y,1)])
```

```
# We can use jacobian function directly
         F = sy.Matrix([f1,f2])
         F.jacobian([x,y])
         F.jacobian([x,y]).subs([(x,1), (y,1)])
Out[435]: Matrix([
         [2, 0],
         [3, -1]]
  We can see that element(2,1) is not exactly same.
Problem 7
         from autograd import grad
         import time
         def problem7(n):
             func = y_lamb
             time_sympy = np.zeros(n)
             time_autograd = np.zeros(n)
             time_centered = np.zeros(n)
             err_centered = np.zeros(n)
             err_autograd = np.zeros(n)
             for i in range(n):
                 x_0=np.random.rand()
                 start_time_sympy=time.time()
                 exact = sympy_diff(x_0)
                 time_sympy[i]=time.time()-start_time_sympy
                 start_time_ct4=time.time()
                 err_centered[i] = np.abs(ct4(func, x_0, 1e-4) - exact)
                 time_centered[i] = time.time() - start_time_ct4
                 #start_time_auto=time.time()
                 #dfqrad = qrad(func)(x_0)
                 \#err_autograd[i]=np.abs(dfgrad)-dy(x_0)
                 #time_autograd[i]=time.time()-start_time_auto
             plt.scatter(time_sympy, 1e-18 * np.ones(n), alpha=0.5, label='SymPy', s=30)
             plt.scatter(time_centered, err_centered, alpha=0.5, label='Centered Difference',
             #plt.scatter(time_autograd, err_autograd, alpha=0.5, label='Autograd', s=30)
             plt.loglog()
             plt.xlim(1e-5, 1e-1)
             plt.ylim(1e-19, 1e-12)
```

```
plt.xlabel('Seconds')
plt.ylabel('Absolute Error')
plt.legend()
plt.show()
```

#### In [540]: problem7(200)



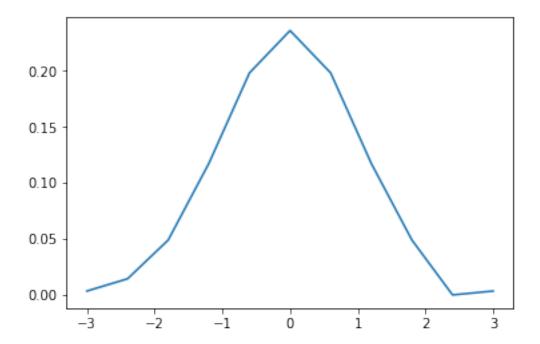
# 0.2 Integration Part

#### **Question 14.1**

```
from scipy import integrate
          from scipy import stats as sts
In [440]: def num integral(func, a, b, N, method='midpoint'):
              bin_cuts = np.linspace(a, b, num=N)
              i = np.arange(len(bin_cuts))
              if method == 'midpoint':
                  xi vals = a + (((2*i+1)*(b - a))/(2*N))
                  g_vals = func(xi_vals)
                  w = (b - a)/N
                  result_mid = w * np.sum(g_vals[:-1])
                  return result_mid
              elif method == 'trapezoid':
                  xi_vals = a + (i*(b - a))/N
                  g_vals = np.sum(func(xi_vals)) + func(xi_vals[-1])
                  w = (b - a)/(2*N)
                  result_trap = w * (func(xi_vals[0]) + 2 * g_vals)
                  return result_trap
              elif method == 'simpsons':
                     h = (b - a) / N
                      s = func(a) + func(b)
                      for i in range(1, N, 2):
                          s += 4 * func(a + i * h)
                      for i in range(2, N-1, 2):
                          s += 2 * func(a + i * h)
                      result_simp = s * h / 3
                      return result_simp
In [441]: midpoint = num_integral(g, -10, 10, 10000, 'midpoint')
          print('Integral computed via midpoint method: ', midpoint)
          trap = num_integral(g, -10, 10, 10000, 'trapezoid')
          print('Integral computed via trapezoid method: ', trap)
          simp = num_integral(g, -10, 10, 10000, 'simpsons')
          print('Integral computed via Simpson method: ', simp)
Integral computed via midpoint method: 4374.1851216356135
Integral computed via trapezoid method: 4377.9757567908655
Integral computed via Simpson method: 4373.3333333333338
```

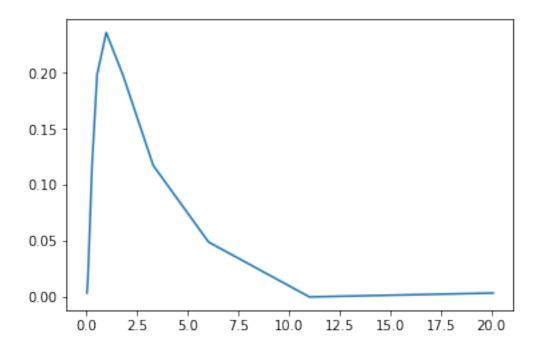
**Question 14.2** Write a python function that makes a Newton-Cotes discrete approximation of the distribution of the normally distributed variable Z~N(mu, sigma)

```
In [442]: from scipy.stats import norm
          import matplotlib.pyplot as plt
In [443]: def approx_normal(mu, sigma, N, k):
              Z = np.linspace(mu - k*sigma, mu + k*sigma, N)
              w = np.zeros_like(Z)
              w[0] = norm.cdf((Z[0] + Z[1]) / 2, loc=mu, scale=sigma)
              w[N-1] = 1 - norm.cdf((Z[N-2] + Z[N-1]) / 2, loc=mu, scale=sigma)
              for i in range(1, N - 2):
                  z_{min} = (Z[i - 1] + Z[i]) / 2
                  z_{max} = (Z[i] + Z[i + 1]) / 2
                  w[i] = norm.cdf(z_max, loc=mu, scale=sigma) - norm.cdf(z_min, loc=mu, scale=sigma)
              return Z, w
          Z, w = approx_normal(0,1,11,3)
          print('Weights:')
          print(w)
          print('Nodes:')
          print(Z)
          plt.plot(bin_cuts, w)
          plt.show()
Weights:
 \hbox{\tt [0.00346697\ 0.01439745\ 0.04894278\ 0.11725292\ 0.19802845\ 0.23582284] }
0.19802845 0.11725292 0.04894278 0.
                                              0.00346697]
Nodes:
[-3. -2.4 -1.8 -1.2 -0.6 0. 0.6 1.2 1.8 2.4 3.]
```



## **Question 14.3**

```
In [444]: def approx_lognormal(mu, sigma, N, k):
              Z, w_log = approx_normal(mu, sigma, N, k)
              Z_{log} = np.exp(Z)
              return Z_log, w_log
          Z_log, w_log = approx_lognormal(0,1,11,3)
          print('Weights:')
          print(w_log)
          print('Nodes:')
          print(Z_log)
          plt.plot(Z_log,w_log)
          plt.show()
Weights:
 \hbox{\tt [0.00346697\ 0.01439745\ 0.04894278\ 0.11725292\ 0.19802845\ 0.23582284] }
0.19802845 0.11725292 0.04894278 0.
                                               0.00346697]
Nodes:
[ 0.04978707  0.09071795  0.16529889  0.30119421  0.54881164  1.
  1.8221188
              3.32011692 6.04964746 11.02317638 20.08553692]
```



Weights are same in 14.2 and 14.3

#### Question 14.4

```
In [445]: mu = 10.5
          sigma = 0.8
          N = 100
          k = 3
          Y, w = approx_lognormal(mu,sigma,N,k)
In [446]: exact = np.exp(mu+(sigma**2)/2)
          print ('Exact expected value :
                                               ', exact)
          EY = Y@w
          print ('Approximated expected value :
                                                  ', EY)
Exact expected value :
                              50011.087008521754
Approximated expected value :
                                     49739.23910845465
In [447]: EY = sum(w[i]*Y[i] for i in range(len(Y)))
          ΕY
Out [447]: 49739.23910845463
```

**Question 14.5** Approximate the integral of the function in Exercise 2.1

```
In [448]: # write down all equations as in 14.11 as N=3
         import scipy as sc
          import scipy.optimize as opt
         g = lambda x: .1 * x ** 4 - 1.5 * x ** 3 + .53 * x ** 2 + 2 * x + 1
         def equations(parameters):
              w1, w2, w3, x1, x2, x3 = parameters
              eq0 = (b - a) - (w1 + w2 + w3)
              eq1 = ((1/2) * b ** 2 - (1/2) * a ** 2) - (w1 * x1 + w2 * x2 + w3 * x3)
              eq2 = ((1/3) * b ** 3 - (1/3) * a ** 3) - (w1 * x1**2 + w2 * x2**2 + w3 * x3**2)
              eq3 = ((1/4) * b ** 4 - (1/4) * a ** 4) - (w1 * x1**3 + w2 * x2**3 + w3 * x3**3)
              eq4 = ((1/5) * b ** 5 - (1/5) * a ** 5) - (w1 * x1**4 + w2 * x2**4 + w3 * x3**4)
              eq5 = ((1/6) * b ** 6 - (1/6) * a ** 6) - (w1 * x1**5 + w2 * x2**5 + w3 * x3**5)
             return (eq0, eq1, eq2, eq3, eq4, eq5)
         a = -10
         b = 10
         parameters = opt.root(equations,np.ones((6,1)), tol = 1e-8).x
         omega, Z = parameters[:3], parameters[-3:]
         integ_val = sum(omega[i]*g(Z[i]) for i in range(len(Z)))
          # Calculate the true known value of the integral
         import sympy as sy
         z = sy.symbols('z')
         exact=sy.integrate(0.1*z**4 - 1.5*z**3 + 0.53*z**2 + 2*z + 1, (z, a, b))
         print ('Gaussian Quadrature: ', integ_val)
         print ('True known value: ', exact)
Gaussian Quadrature:
                           4373.333333340382
```

**Question 14.6** Use the python Gaussian quadrature command to numerically approximate integral

4373.33333333333

True known value:

```
In [449]: sc.integrate.quad(g, -10, 10)[0]
Out[449]: 4373.333333333333
Question 14.7 Use Monte Carlo integration to aaproximate \pi
In [582]: import math
          np.random.seed(seed=25)
          def monte(g, domain, N):
              draws = np.zeros((N,2))
              monte_sum = 0
              for i in range(N):
                  for j in range(2):
                      draws[i,j] = np.random.uniform(domain[j,0],domain[j,1])
                  monte_sum+=g(draws[i,:])
              return (4/N)*monte_sum
          g = lambda x: 1 if x[0]**2+x[1]**2 <= 1 else 0
          domain = np.array([[-1,1],[-1,1]])
          monte(g,domain,20000)
Out [582]: 3.1338000000000004
In [584]: np.random.seed(seed=25)
          toler = 1e-5
          diff = 1e3
          n = 10
          max_N = 100000
          while diff > toler and n < max_N:</pre>
              pi = monte(g, domain, n)
              diff = np.abs(pi - 3.1415)
```

smallest number of random draws 4432 3.1415030467163167

print("smallest number of random draws", n, pi)

**Question 14.8** Define a function in that returns the -th element of a -dimensional equidistributed sequence. It should have support for the four sequences in the Table in Section 4.2.

n += 1

```
break
                  else:
                      primes.append(num)
                  num += 2
              return primes
In [497]: def equidistr(n, d, seq):
              if seq == 'weyl':
                  prime = n_p(d)
                  x = n * np.power(prime, .5)
                  xn = x - np.floor(x)
              elif seq == 'haber':
                  prime = n_p(d)
                  x = (n * (n + 1.0)) / 2 * np.power(prime, .5)
                  xn = x - np.floor(x)
              elif seq == 'niederreiter':
                  ar = np.arange(1, d + 1)
                  x = n * (2 ** (ar / (d + 1)))
                  x_floor = np.floor(x)
                  xn = x - x_floor
              elif seq == 'baker':
                  prime = n_p(d)
                  x = n * np.exp(prime)
                  xn = x - np.floor(x)
              else:
                  raise ValueError('Sequence does not exist.')
              return xn
In [498]: print("weyl:", equidistr(10, 2, 'weyl'))
          print("haber:", equidistr(10, 2, 'haber'))
          print("niederreiter", equidistr(10, 2, 'niederreiter'))
          print("baker", equidistr(5, 2, 'baker'))
weyl: [0.14213562 0.32050808]
haber: [0.78174593 0.26279442]
niederreiter [0.5992105 0.87401052]
baker [0.94528049 0.42768462]
```

**Question 14.9** approximate the value of , this time using quasi-Monte Carlo integration.

```
pi = 0
              iterate = 0
              s = np.zeros((N, 2), dtype=np.float64)
              for i in range(1, N+1):
                  s[i-1, :] = equidistr(i, 2, seq)
              for j in range(N):
                  if s[j,0] ** 2 + s[j,1] ** 2 <= 1:
                      iterate += 1
              pi = 4.0 * iterate / N
              return pi
In [ ]: np.random.seed(seed=25)
        toler = 1e-5
        diff = 1e3
        n = 10
        max_N = 100000
        while diff > toler and n < max_N:</pre>
            pi = quasi_monte(n,'weyl')
            diff = np.abs(pi - 3.1415)
        print("weyl:", quasi_monte(100000, 'weyl'), "convergence")
In [546]: print("weyl:", quasi_monte(100000, 'weyl'))
          print("haber:", quasi_monte(100000, 'haber'))
          print("niederreiter", quasi_monte(100000, 'niederreiter'))
          print("baker", quasi_monte(100000, 'baker'))
weyl: 3.14036
haber: 3.14396
niederreiter 3.1414
baker 3.1416
In [570]: np.random.seed(seed=25)
          toler = 1e-5
          diff = 1e3
          n = 10
          max_N = 100000
          while diff > toler and n < max_N:</pre>
              pi_wey = quasi_monte(n,'weyl')
              diff = np.abs(pi_wey - 3.1415)
          print("weyl:", quasi_monte(100000, 'weyl'), "smallest number of random draws", n)
weyl: 3.14036 smallest number of random draws 1683
```

```
In [571]: np.random.seed(seed=25)
          toler = 1e-5
          diff = 1e3
          n = 10
          \max N = 100000
          while diff > toler and n < max_N:</pre>
              pi_haber = quasi_monte(n, 'haber')
              diff = np.abs(pi_haber - 3.1415)
          print("haber:", quasi_monte(100000, 'haber'), "smallest number of random draws", n)
haber: 3.14396 smallest number of random draws 1054
In [573]: np.random.seed(seed=25)
          toler = 1e-5
          diff = 1e3
          n = 10
          max_N = 100000
          while diff > toler and n < max_N:</pre>
              pi_nieder = quasi_monte(n, 'niederreiter')
              diff = np.abs(pi_nieder - 3.1415)
          print("niederreiter:", quasi_monte(100000, 'niederreiter'), "smallest number of rando
niederreiter: 3.1414 smallest number of random draws 1888
In [586]: np.random.seed(seed=25)
          toler = 1e-5
          diff = 1e3
          n = 10
          max_N = 100000
          while diff > toler and n < max_N:</pre>
              pi_baker = quasi_monte(n, 'baker')
              diff = np.abs(pi_baker - 3.1415)
              n += 1
          print("baker:", quasi_monte(100000, 'baker'), "smallest number of random draws", n)
        KeyboardInterrupt
                                                    Traceback (most recent call last)
        <ipython-input-586-28fe198addc1> in <module>
          5 \text{ max}_N = 100000
          6 while diff > toler and n < max_N:
    ----> 7
                pi_baker = quasi_monte(n,'baker')
                diff = np.abs(pi_baker - 3.141)
```

```
<ipython-input-553-a8bed455b5bf> in quasi_monte(N, seq)
            s = np.zeros((N, 2), dtype=np.float64)
           for i in range(1, N+1):
               s[i-1, :] = equidistr(i, 2, seq)
----> 8
     10
           for j in range(N):
    <ipython-input-497-9ef1e4db3589> in equidistr(n, d, seq)
           elif seq == 'baker':
               prime = n_p(d)
     20
---> 21
               x = n * np.exp(prime)
               xn = x - np.floor(x)
     22
     23
            else:
```

KeyboardInterrupt:

9

n += 1

In terms of number of draws, haber < weyl < niederreiter