

# hw2 Problem2

January 20, 2019

## 0.0.1 Import Packages

```
In [1]: import numpy as np
import scipy as sp
from scipy.stats import norm
from scipy.integrate import quad
```

## 0.0.2 2.1

```
In [2]: def integrate(g,a,b,N,method='midpoint'):
    if method not in ['midpoint','trapezoid', 'Simpsons']:
        raise ValueError
    else:
        if method == 'midpoint':
            counter = 0
            for i in range(N):
                counter += g(a+(2*i+1)*(b-a)/(2*N))
            return (b-a)*counter/N
        if method == 'trapezoid':
            counter = g(a)+g(b)
            for i in range(1,N):
                counter += 2*g(a+i*(b-a)/N)
            return (b-a)*counter/(2*N)
        if method == 'Simpsons':
            counter = g(a)+g(b)+4*g(a+(2*N-1)*(b-a)/(2*N))
            for i in range(1,N):
                counter += 4*g(a+(2*i-1)*(b-a)/(2*N))
                counter += 2*g(a+2*i*(b-a)/(2*N))
            return (b-a)*counter/(6*N)
test_func = lambda x: 0.1*x**4-1.5*x**3+0.53*x**2+2*x+1
exact = 0.02*(10**5-(-10)**5)+0.53/3*(10**3-(-10)**3)+20
for method in ['midpoint','trapezoid', 'Simpsons']:
    integr = integrate(test_func, -10, 10, 100000, method)
    print(method,'method value:',integr)
    print("Difference between",method,'method and true value',abs(integr-exact))
```

midpoint method value: 4373.333331964723

Difference between midpoint method and true value 1.3686103557120077e-06

trapezoid method value: 4373.333336070682  
 Difference between trapezoid method and true value 2.7373489501769654e-06  
 Simpsons method value: 4373.333333333185  
 Difference between Simpsons method and true value 1.482476363889873e-10

### 0.0.3 2.2

```
In [3]: def N_C(N, mu=0, sigma=1, k=3):
        Z = np.linspace(mu-k*sigma, mu+k*sigma, N)
        weight = np.zeros(N)
        weight[0] = norm.cdf((Z[0]+Z[1])/2, loc=mu, scale=sigma)
        for i in range(1,N-1):
            func = lambda x: norm.pdf(x, loc=mu, scale=sigma)
            weight[i] = quad(func, (Z[i-1]+Z[i])/2, (Z[i+1]+Z[i])/2)[0]
        weight[N-1] = 1-norm.cdf((Z[N-2]+Z[N-1])/2, loc=mu, scale=sigma)
        return Z, weight
Z, weight = N_C(11)
print('Z =',Z)
print('weight = ',weight)
```

Z = [-3. -2.4 -1.8 -1.2 -0.6 0. 0.6 1.2 1.8 2.4 3. ]  
 weight = [0.00346697 0.01439745 0.04894278 0.11725292 0.19802845 0.23582284  
 0.19802845 0.11725292 0.04894278 0.01439745 0.00346697]

### 0.0.4 2.3

```
In [4]: def new_N_C(N, mu=0, sigma=1, k=3):
        Z = np.linspace(mu-k*sigma, mu+k*sigma, N)
        A = np.e**Z
        weight = np.zeros(N)
        weight[0] = norm.cdf((Z[0]+Z[1])/2, loc=mu, scale=sigma)
        for i in range(1,N-1):
            func = lambda x: norm.pdf(x, loc=mu, scale=sigma)
            weight[i] = quad(func, (Z[i-1]+Z[i])/2, (Z[i+1]+Z[i])/2)[0]
        weight[N-1] = 1-norm.cdf((Z[N-2]+Z[N-1])/2, loc=mu, scale=sigma)
        return A, weight
A, weight = new_N_C(11)
print('A =',A)
print('weight = ',weight)
```

A = [ 0.04978707 0.09071795 0.16529889 0.30119421 0.54881164 1.  
 1.8221188 3.32011692 6.04964746 11.02317638 20.08553692]  
 weight = [0.00346697 0.01439745 0.04894278 0.11725292 0.19802845 0.23582284  
 0.19802845 0.11725292 0.04894278 0.01439745 0.00346697]

## 0.0.5 2.4

```
In [5]: A, weight = new_N_C(1001, mu=10.5, sigma=0.8, k=10)
Myresult = A@weight.reshape(-1,1)
Exactresult = np.e**(10.5+0.5*(0.8**2))
print('The difference between my result and the exact expectation is:')
print(Myresult[0]-Exactresult)
```

The difference between my result and the exact expectation is:  
0.5334533017885406

## 0.0.6 3.1

```
In [6]: def Gaussian(g,a,b,N=3):
    init_weight = [1/N for i in range(N)]
    init_x = [a+i*(b-a)/(N-1) for i in range(N)]
    init = init_weight+init_x
    def func(x):
        result = []
        for i in range(2*N):
            weight = x[:N]
            node = x[N:]
            Sum = sum(weight[k]*(node[k]**i) for k in range(N))
            result.append((b**(i+1)-a**(i+1))/(i+1)-Sum)
        return tuple(k for k in result)
    Vector = [k for k in sp.optimize.root(func, init)['x']]
    weight = Vector[:N]
    node = Vector[N:]
    counter = 0
    for i in range(N):
        counter += weight[i]*g(node[i])
    return counter
test_func = lambda x: 0.1*x**4-1.5*x**3+0.53*x**2+2*x+1
Gauss = Gaussian(test_func, -10, 10)
Newton = integrate(test_func, -10, 10, 10000, "Simpsons")
Exact = 0.02*(10**5-(-10)**5)+0.53/3*(10**3-(-10)**3)+20
print("The result of Gaussian approximate is", Gauss)
print('The absolute error of Gaussian approximate is', abs(Gauss-Exact))
print("The result of Newton-Cotes approximate is", Newton)
print('The absolute error of Newton-Cotes approximate is', abs(Newton-Exact))
```

The result of Gaussian approximate is 4373.333333189601  
The absolute error of Gaussian approximate is 1.4373199519468471e-07  
The result of Newton-Cotes approximate is 4373.333333333337  
The absolute error of Newton-Cotes approximate is 3.728928277269006e-11

### 0.0.7 3.2

```
In [7]: Quad = quad(lambda x: 0.1*x**4-1.5*x**3+0.53*x**2+2*x+1, -10, 10)[0]
        print("The result of Python Gaussian approximate is", Quad)
        print('The absolute error of Python Gaussian approximate is', abs(Quad-Exact))
```

The result of Python Gaussian approximate is 4373.333333333334

The absolute error of Python Gaussian approximate is 9.094947017729282e-13

### 0.1 4

```
In [8]: def isPrime(n):
        '''
        -----
        This function returns a boolean indicating whether an integer n is a
        prime number
        -----
        INPUTS:
        n = scalar, any scalar value

        OTHER FUNCTIONS AND FILES CALLED BY THIS FUNCTION: None

        OBJECTS CREATED WITHIN FUNCTION:
        i = integer in [2, sqrt(n)]

        FILES CREATED BY THIS FUNCTION: None

        RETURN: boolean
        -----
        '''
        for i in range(2, int(np.sqrt(n) + 1)):
            if n % i == 0:
                return False

        return True
def primes_ascend(N, min_val=2):
    '''
    -----
    This function generates an ordered sequence of N consecutive prime
    numbers, the smallest of which is greater than or equal to 1 using
    the Sieve of Eratosthenes algorithm.
    (https://en.wikipedia.org/wiki/Sieve\_of\_Eratosthenes)
    -----
    INPUTS:
    N          = integer, number of elements in sequence of consecutive
                prime numbers
    min_val    = scalar >= 2, the smallest prime number in the consecutive
                sequence must be greater-than-or-equal-to this value
```

*OTHER FUNCTIONS AND FILES CALLED BY THIS FUNCTION:*

*isPrime()*

*OBJECTS CREATED WITHIN FUNCTION:*

*primes\_vec* = (N,) vector, consecutive prime numbers greater than *min\_val*

*MinIsEven* = boolean, =True if *min\_val* is even, =False otherwise

*MinIsGrtrThn2* = boolean, =True if *min\_val* is greater-than-or-equal-to 2, =False otherwise

*curr\_prime\_ind* = integer  $\geq 0$ , running count of prime numbers found

*FILES CREATED BY THIS FUNCTION: None*

*RETURN: primes\_vec*

```
-----  
'''  
primes_vec = np.zeros(N, dtype=int)  
MinIsEven = 1 - min_val % 2  
MinIsGrtrThn2 = min_val > 2  
curr_prime_ind = 0  
if not MinIsGrtrThn2:  
    i = 2  
    curr_prime_ind += 1  
    primes_vec[0] = i  
i = min(3, min_val + (MinIsEven * 1))  
while curr_prime_ind < N:  
    if isPrime(i):  
        curr_prime_ind += 1  
        primes_vec[curr_prime_ind - 1] = i  
    i += 2  
  
return primes_vec
```

#### 0.1.1 4.1

```
In [9]: def M_C(N, func=None, omega=[-1,1,-1,1]):  
    counter = 0  
    x_1 = np.random.uniform(omega[0],omega[1],size=N)  
    x_2 = np.random.uniform(omega[2],omega[3],size=N)  
    def g(x,y):  
        if x**2+y**2<=1:  
            return 1  
        else:  
            return 0  
    for i in range(N):  
        x,y = x_1[i],x_2[i]  
        if func is None:
```

```

        counter += g(x,y)
    else:
        counter += func(x,y)
    return 4*counter/N
np.random.seed(25)
judge = False
min_N = 0
while judge is False:
    min_N += 1
    judge = (round(M_C(min_N),4)==3.1415)
print("The smallest number of random draws N is", min_N)

```

The smallest number of random draws N is 615

#### 0.1.2 4.2

```

In [10]: def equidistribution(n,d,Type='weyl'):
    prime_vector = primes_ascend(d)
    def rational_list(d):
        return [1/(i+1) for i in range(d)]
    def cut(x):
        return x-x//1
    if Type == 'weyl':
        return tuple(cut(n*np.sqrt(prime_vector[i])) for i in range(d))
    elif Type == 'haber':
        return tuple(cut(n*(n+1)*0.5*np.sqrt(prime_vector[i])) for i in range(d))
    elif Type == 'nie':
        return tuple(cut(n*(2**(i/(n+1)))) for i in range(d))
    elif Type == 'baker':
        return tuple(cut(n*(np.e**(rational_list(d)[i]))) for i in range(d))

```

#### 0.1.3 4.3

```

In [11]: def quasi_M_C(N,Type, func=None,omega=[-1,1,-1,1]):
    counter = 0
    x = []
    for k in range(N):
        x.append((2*equidistribution(k,2,Type)[0]-1,2*equidistribution(k,2,Type)[1]-1))
    def g(X):
        x,y = X[0], X[1]
        if x**2+y**2<=1:
            return 1
        else:
            return 0
    for i in range(N):
        X = x[i]
        if func is None:

```

```

        counter += g(X)
    else:
        counter += func(X)
    return 4*counter/N
Text = "The smallest number of random draws N"
print(Text, "for M-C method is", min_N)
for method in ['weyl', 'haber', 'baker']:
    judge = False
    min_N2 = 0
    while judge is False:
        min_N2 += 1
        judge = (round(quasi_M_C(min_N2, Type = method), 4) == 3.1415)
    print(Text, "for quasi-M-C method with type", method, "is", min_N2)

```

The smallest number of random draws N for M-C method is 615

The smallest number of random draws N for quasi-M-C method with type weyl is 1230

The smallest number of random draws N for quasi-M-C method with type haber is 2064

The smallest number of random draws N for quasi-M-C method with type baker is 205

Since the smallest number of random draws N for quasi-Monte Carlo integration with Baker sequence that matches the true value of  $\pi$  to the 4th decimal 3.1415 is the smallest, the rates of convergence of quasi-Monte Carlo integration with Baker sequence is the fastest. We can show the convergence rate of quasi-Monte Carlo integration with Niederreiter sequence is the slowest or with the following code:

```
In [12]: round(quasi_M_C(100000, Type = method), 4) - 3.1415
```

```
Out[12]: -0.00030000000000000189
```