Hermite-Gauss Quadrature Lab Report for Assignment No. 5

SHASHVAT JAIN (2020PHY1114) HARSH SAXENA (2020PHY1162)

S.G.T.B. Khalsa College, University of Delhi, Delhi-110007, India.

March 10, 2022

Submitted to Dr. Mamta "32221401 - MATHEMATICAL PHYSICS III"

Contents

1	Theory		
	1.1	Hermite-Gauss Quadrature	1
	1.2	Hermite polynomials	1
	1.3	Construction of 2-point Hermite gauss quadrature	2
2 Programming		3	
3	Res	ults and Analysis	6

1 Theory

1.1 Hermite-Gauss Quadrature

One such family of Polynomials is Hermite Polynomials which are known to be the solution of Hermite differential equation which is a second order linear differential equation.

$$y'' + -2xy' + \lambda y = 0$$
 where $\lambda \in \mathbb{R}$ (1)

Whenever $\lambda \in \mathbb{N}$ the solution of this differential equation gives the Hermite polynomial.

The weighting function and integration interval of these polynomials is -

$$w(x) = e^{-x^2}$$
 on $[-\infty, \infty)$ (2)

It is evident by seeing the limit of integration that this gaussian quadrature rule is used in calculating the improper of integral of type.

$$\int_{-\infty}^{\infty} f(x)dx \tag{3}$$

Specifically when we include the weighting function e^{-x^2} then.

$$\int_{-\infty}^{\infty} f(x)e^{-x^2}dx = \sum_{1}^{n} w_i f(x_i)$$

where n is the n point quadrature used. We can write

$$\int_{-\infty}^{\infty} f(x)dx = \int_{-\infty}^{\infty} f(x)e^{x^2}e^{-x^2}dx = \int_{-\infty}^{\infty} g(x)e^{-x^2}dx$$
$$g(x) := f(x)e^{x^2}$$

1.2 Hermite polynomials

The Solution of above differential equation results in the following recurrence relation.

$$2xH_n(x) = 2nH_{n-1}(x) - H_{n+1}(x)$$
(4)

If we know the polynomials H_0 and H_1 we can determine all the other polynomials

$$H_n(x) = \sum_{k=0}^n \frac{(-1)^k (n!)}{(n-2k)!k!} (2x)^{n-2k}$$
(5)

From these relations we can derive first five Hermite polynomials

$$H_0 = 1 H_1 = 2x (6)$$

$$H_2 = 4x^2 - 2 (7)$$

$$L_3 = 8x^3 - 12x \tag{8}$$

$$L_4 = 16x^4 + 48x^2 + 12\tag{9}$$

$$L_5 = 32x^5 - 160x^3 + 120x \tag{10}$$

Orthogonality relation

$$\int_{-\infty}^{\infty} H_m(x)H_n(x)e^{-x^2}dx = 2^n n! \sqrt{\pi}\delta_{mn}$$
(11)

Where δ_{mn} is the kronecker delta function

1.3 Construction of 2-point Hermite gauss quadrature

Since this formula is to have degree of precision equal to 3,Then the weights and abcissas must satisfy-that for constant,linear,quadratic and cubic functions the integral must be exatly equal to the weights multiplied by the function evaluation at these decided abcissa.

1. f(x) = 1

$$\int_{-\infty}^{\infty} 1e^{-x} dx = w_1 f(x_1) + w_2 f(x_2)$$

$$w_1 + w_2 = \sqrt{\pi}$$
(12)

2. f(x) = x

$$\int_{-\infty}^{\infty} x e^{-x^2} dx = w_1 f(x_1) + w_2 f(x_2)$$
(13)

$$w_1 x_1 + w_2 x_2 = \left[\frac{-e^{-x^2}}{2}\right]_{-\infty}^{\infty} \tag{14}$$

$$w_1 x_1 + w_2 x_2 = 0 (15)$$

3. $f(x) = x^2$

$$\int_{-\infty}^{\infty} x^2 e^{-x^2} dx = w_1 f(x_1) + w_2 f(x_2)$$
(16)

(17)

$$w_1 x_1^2 + w_2 x_2^2 = \frac{\sqrt{\pi}}{2} \tag{18}$$

4. $f(x) = x^3$

$$\int_{-\infty}^{\infty} x^3 e^{-x^2} dx = w_1 f(x_1) + w_2 f(x_2) \tag{19}$$

$$w_1 x_1^3 + w_2 x_2^3 = \left[-(x^2 + 1)e^{-x^2} \right]_{-\infty}^{\infty}$$
(20)

$$w_1 x_1^3 + w_2 x_2^3 = 0 (21)$$

After solving the following equations we get the following weights-

x_i	w_i
-0.7071067811865475244008	0.8862269254527580136491
0.7071067811865475244008	0.886226925452758013649

Table 1: Roots and weights for 2-point formula

2 Programming

```
1 # -*- coding: utf-8 -*-
3 Created on Sun Mar 6 15:41:06 2022
5 @author: harsh
8 from Myintegration import *
9 from scipy import stats
10 import matplotlib.pyplot as plt
11
plt.style.use('bmh')
13
  def MyHermiteQuad(func,n):
      x,w = roots_hermite(n)
      return np.dot(w,func(x)*np.exp(x))
16
17
18 MyHermiteQuad = np.vectorize(MyHermiteQuad)
19
20 #
_{21} # Validation
23 listp = ['one','two','three']
24 \text{ func} = []
_{25} P = []
_{26} K = []
for i in range(len(listp)):
      k = input(f"function {listp[i]}: ")
28
      K.append(k)
      func.append(lambda x,i = i: eval(K[i],{'x':x,'np':np}))
30
31
32
  for j in range(len(func)):
      p1 = lambda x, j=j : func[j](x)*np.exp(-(x**2))
34
      P.append(p1)
35
36
mat = np.zeros((2,4))
n_r = [2,4]
39 mat[:,0] = n_r
40 for i in range (1,4):
     mat[:,i] = MyHermiteQuad(P[i-1],n_r)
_{43} np.savetxt("validate-herm-1162.out", mat, fmt="%.7g", delimiter=",", header="n, $I_1$,
     $I_2$")
44 print (mat)
45
46 ######
I1 = lambda x : np.exp(-x**2)/(1+x**2)
48 I2 = lambda x : 1/(1+x**2)
49
dat = np.zeros((7,3))
dat[:,0] = 2**np.arange(1,8)
```

```
53 n_arr = dat[:,0]
54 dat[:,1] = MyHermiteQuad(I1,n_arr)
55 dat[:,2] = MyHermiteQuad(I2,n_arr)
_{56} np.savetxt("quad-herm-1114.out", dat, fmt="%.7g", delimiter=",", header="n,$I_1$,$I_2$")
  def hermite_tol(func,n_max,rtol):
       n_{arr} = np.arange(5, n_{max}, 5)
59
       I = np.zeros(n_arr.shape)
       r_{err} = np.zeros(len(n_{arr})-1)
61
       for i in range(len(n_arr)):
63
           I[i] = MyHermiteQuad(func, n_arr[i])
64
           if i==0:
65
              continue
66
           r_{err[i-1]} = abs((I[i] - I[i-1])/(I[i]))
67
           if r_err[i-1] <= rtol:</pre>
68
              return r_err[:i], I[:i], n_arr[:i]
69
70
71
_{72} def simpsonHermite(f, max_b= int(2e16), rtol = 0.5*10**(-4)):
       b = 2**np.arange(1,np.floor(np.log2(max_b)))
       print(b)
74
       I = np.zeros(b.shape)
75
       r_{err} = np.zeros(len(b)-1)
76
       for i, bi in enumerate(b):
           I[i] = MySimp(f,0,bi,m = int(1e7),d = 5)[0]
78
           if i == 0:
79
               continue
           r_{err}[i-1] = abs((I[i] - I[i-1])/(I[i]))
82
           if r_err[i-1] <= rtol:</pre>
83
               return r_err[:i], I[:i], b[:i]
84
       return I,b
87 n_arrt = np.arange(5,100,5)
88 inty = MyHermiteQuad(I2,n_arrt)
90 rl_l, inte_l,n_ar = hermite_tol(I2,500,0.5*10**(-12))
91 rl_s, inte_s,b_ar = simpsonHermite(I2,int(2e16),0.5*10**(-4))
93 slope, intercept, r_value, p_value, std_err = stats.linregress(rl_s,inte_s)
94 Y = slope*rl_s + intercept
96 print('Slope of relative tol vs Integral line for function 1/(1+x^2)',slope)
97 print(n_ar)
98 #Plotting and comparison
99
100 fig,ax = plt.subplots()
101 ax.plot(rl_l,inte_l,'--o',label ='improper integral of 1/(1+x^2) by hermite in range
       -inf to inf')
102 ax.plot(rl_s,inte_s,'* ',label = 'improper integral by simpson')
ax.plot(rl_s,Y,label = 'Regression line')
ax.set_xlabel('relative tolerence')
ax.set_ylabel('Integral')
106 ax.set_title('Comparsion between integraal calculated by simpson and Hermite')
107 ax.legend()
109 fig,(ax1,ax2) = plt.subplots(1,2)
ax1.plot(n_arrt,inty,'--v')
ax1.set_xlabel('n point formula used')
ax1.set_ylabel('Integral')
ax1.set_title('A')
```

```
ax2.plot(b_ar,inte_s,'r--x')
ax2.set_xlabel('Upper limit on the integral')
ax2.set_ylabel('Integral')
ax2.set_title('B')
```

3 Results and Analysis

From the data obtained, shown below, we notice that both the integrals I_1 and I_2 converge as the Gaussian Quadrature is calculated at greater number of absiccas. (n in the n-point formula, which is also what we expected to happen as the Gaussian quadrature is able to calculate functions differing from the usual polynomial functions more accurately with an increase in n-points.

n	I_1	I_2
2	0.9034642	1.489561
4	1.101324	1.558053
8	1.153702	1.575974
16	1.159536	1.578104
32	1.159775	1.578192
64	1.159777	1.578193
128	1.159777	1.578193

Table 2: Data from quad-herm-1114.out

However we also expect this change in integral to die out which is explained by the second figure. The simpson method seems to converge much more sharply, but this is really not the case, do not forget that in simpson's case we are increasing the upper-bound thus introducing many more sub-intervals to compute the formula over whereas in the case of n-points, we are varying n-points not so much in comparison to Simpson.

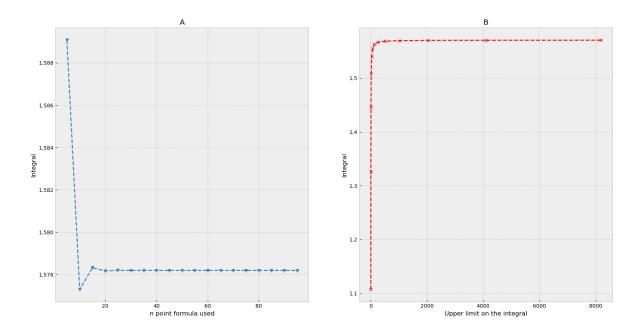


Figure 1: (A) Hermite-Gauss Quadrature (B) Simpson's composite rule to calculate $f(x) = \frac{1}{1+x^2}$

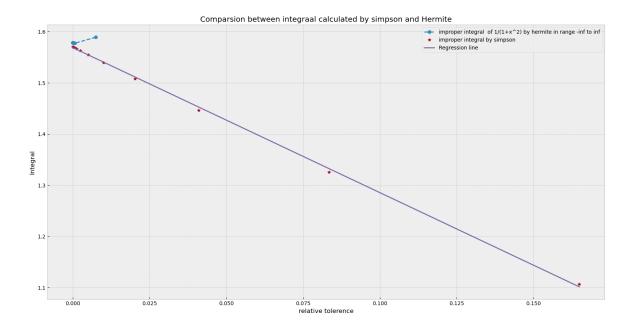


Figure 2: Integral vs relative toler