

Technical University of Moldova

Homework nr.4

on Numerical Analysis

executed in Python programming language

by the student from FAF – 213 academic group

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Problem 4.1:

Solution and output:

```
Fresnel integrals.py x Own integral code.py x Pi approximation.py x Gamma integration.py x
1  # this task requires studying Gamma function
2  # and the corresponding integral for computing it
3  from scipy import integrate, special
4      from numpy import linspace
5      import math
6  import warnings
7
8
9  # it is more convenient to ignore the accuracy warnings occurring
10 # while using the built-in integration methods
11 warnings.filterwarnings('ignore')
12
13
14 def Gamma_function(t, parameter):
15     return pow(t, parameter - 1) * pow(math.e, -t)
16
17
18 # first up let us apply the built in integration methods
19 # we will use Gaussian quadrature method
20 print('Gaussian quadrature deviation:')
21 x = 1
22 while x <= 4:
23     value = integrate.quadrature(lambda t: pow(t, x - 1) * pow(math.e, -t), 0, x)
24     print(f'x = {x}; relative_error = {round(abs(value[0] - special.gamma(x)) / special.gamma(x), 5)}')
25     x += 0.25
26 print()
27 print()
28
29
30 # secondly we apply composite Simpson's rule
31 print('Composite Simpson method deviation:')
32 x = 1
33 while x <= 5:
34     x_args = linspace(1, 10, 1000)
35     y_args = list()
36     for i in range(0, len(x_args)):
37         y_args.append(Gamma_function(x_args, x))
38
39     value = integrate.simps(y_args, x_args)
40     print(f'x = {x}; relative_error = {round(abs(value[0] - special.gamma(x)) / special.gamma(x), 5)}')
41     x += 0.25
42 print()
43 print()
44
45
46 # also we can try to use general purpose quadrature method
47 # and see the error in the integration itself
48 print('General purpose quadrature method deviation and calculation error:')
49 x = 1
50 while x <= 4:
51     value = integrate.quad(lambda t: pow(t, x - 1) * pow(math.e, -t), 0, x)
52     print(f'x = {x}; integral_error = {value[1]}; relative_deviation = {round(abs(value[0] - special.gamma(x)) / special.gamma(x), 5)}')
53     x += 0.25
54 print()
55 print()
```

```
Gamma integration ×
C:\Users\User\PycharmProjects\Laboratories\
Gaussian quadrature deviation:
x = 1; relative_error = 0.36788
x = 1.25; relative_error = 0.38146
x = 1.5; relative_error = 0.39162
x = 1.75; relative_error = 0.39957
x = 2.0; relative_error = 0.40601
x = 2.25; relative_error = 0.41135
x = 2.5; relative_error = 0.41588
x = 2.75; relative_error = 0.41978
x = 3.0; relative_error = 0.42319
x = 3.25; relative_error = 0.4262
x = 3.5; relative_error = 0.42888
x = 3.75; relative_error = 0.43129
x = 4.0; relative_error = 0.43347
```

```
Composite Simpson method deviation:
x = 1; relative_error = 0.63217
x = 1.25; relative_error = 0.5263
x = 1.5; relative_error = 0.42776
x = 1.75; relative_error = 0.34
x = 2.0; relative_error = 0.26474
x = 2.25; relative_error = 0.20232
x = 2.5; relative_error = 0.1521
x = 2.75; relative_error = 0.11286
x = 3.0; relative_error = 0.08307
x = 3.25; relative_error = 0.06118
x = 3.5; relative_error = 0.04573
x = 3.75; relative_error = 0.03546
x = 4.0; relative_error = 0.02932
x = 4.25; relative_error = 0.02652
x = 4.5; relative_error = 0.02644
x = 4.75; relative_error = 0.02867
x = 5.0; relative_error = 0.03291
```

General purpose quadrature method deviation and calculation error:

```
x = 1; integral_error = 7.017947987503856e-15; relative_deviation = 0.36788
x = 1.25; integral_error = 3.438695106439127e-09; relative_deviation = 0.38147
x = 1.5; integral_error = 6.827106657780746e-10; relative_deviation = 0.39163
x = 1.75; integral_error = 5.831357619001665e-11; relative_deviation = 0.39957
x = 2.0; integral_error = 6.594659821447934e-15; relative_deviation = 0.40601
x = 2.25; integral_error = 4.199973702156967e-12; relative_deviation = 0.41135
x = 2.5; integral_error = 5.199964141632447e-09; relative_deviation = 0.41588
x = 2.75; integral_error = 3.019555108556365e-09; relative_deviation = 0.41978
x = 3.0; integral_error = 1.2807753055302941e-14; relative_deviation = 0.42319
x = 3.25; integral_error = 1.3611053254449236e-08; relative_deviation = 0.4262
x = 3.5; integral_error = 7.2569228759849354e-09; relative_deviation = 0.42888
x = 3.75; integral_error = 3.1896394752304286e-08; relative_deviation = 0.43129
x = 4.0; integral_error = 3.7738470990419905e-14; relative_deviation = 0.43347
```

```
58 # as we can see from our results the smallest errors are provided by
59 # the composite Simpson's method; the rule is better to use for x values
60 # especially from 2 to 4 because at that interval the deviation is bounded
61 # and is decreasing; at the same time for the general purpose quadrature method
62 # it is permanently increasing and the error in computing the integral is
63 # not stable; it is also important to mention that the Gaussian quadrature
64 # method provides us with the bounded deviation on the interval from 1 to 2;
65 # also, both first and the third methods show smaller ratio of the errors
66 # when x is increasing;
```

Problem 4.2:

Solution and output:

```
Fresnel integrals.py × Own integral code.py × Pi approximation.py ×
1 # in this task we will compute the integral which approximates
2 # the number pi using different numerical integration methods
3 from math import pi
4 from scipy import integrate
5 import timeit
6
7
8 def pi_function(x):
9     return 4 / (1 + x ** 2)
```

```

12  # let us start from the midpoint rule
13  # we are working on the interval [a, b]
14  # h is the step size, n - number of sub-intervals
15  def midpoint_rule(a, b, n):
16      h = (b - a) / n
17      integral_value = 0
18      for i in range(n):
19          integral_value += pi_function((a + h / 2.0) + i * h)
20
21      return integral_value * h
22
23
24  # composite trapezoidal rule
25  def trapezoidal_rule(a, b, n):
26      h = (b - a) / n
27      integral_value = 0.5 * (pi_function(a) + pi_function(b))
28
29      for i in range(1, n):
30          k = a + i * h
31          integral_value += pi_function(k)
32
33      integral_value *= h
34
35      return integral_value

```

```

38  # composite Simpson rule
39  def Simpson_rule(a, b, n):
40      h = (b - a) / n
41
42      x_args = list()
43      y_args = list()
44
45      # calculating values of x and f(x)
46      i = 0
47      while i <= n:
48          x_args.append(a + i * h)
49          y_args.append(pi_function(x_args[i]))
50          i += 1

```

```

52         # calculating the value of the integral
53         integral_value = 0
54         i = 0
55         while i <= n:
56             if (i == 0) or (i == n):
57                 integral_value += y_args[i]
58             elif i % 2 != 0:
59                 integral_value += 4 * y_args[i]
60             else:
61                 integral_value += 2 * y_args[i]
62             i += 1
63
64         integral_value *= (h / 3)
65
66         return integral_value

```

```

69 # and now let's start to study the behaviour of each method
70 # by measuring the error and the time needed for computations
71 # we will vary the value of n for each method
72
73 start = timeit.default_timer()
74
75 print('Midpoint rule:')
76 N = 2
77 prev_error = 1
78 while N <= 2**21:
79     experimental_value = midpoint_rule(0, 1, N)
80     error = abs(pi - experimental_value)
81     if N < 128:
82         ratio = round(prev_error / error, 4)
83         prev_error = error
84         print(f'N = {N}; error = {error}; ratio = {ratio}')
85     else:
86         print(f'N = {N}; error = {error}')
87     N *= 2
88
89 end = timeit.default_timer()
90 print(f'\nTotal time required: {round(end - start, 5)} seconds')
91 print('\n\n')

```



```
94     start = timeit.default_timer()
95
96     print('Composite trapezoidal rule:')
97     N = 2
98     prev_error = 1
99     while N <= 2**21:
100         experimental_value = trapezoidal_rule(0, 1, N)
101         error = abs(pi - experimental_value)
102         if N < 128:
103             ratio = round(prev_error / error, 4)
104             prev_error = error
105             print(f'N = {N}; error = {error}; ratio = {ratio}')
106         else:
107             print(f'N = {N}; error = {error}')
108         N *= 2
109
110     end = timeit.default_timer()
111     print(f'\nTotal time required: {round(end - start, 5)} seconds')
112     print('\n\n\n')
```









```
115     start = timeit.default_timer()
116
117     print('Composite Simpson rule:')
118     N = 2
119     while N <= 2**21:
120         experimental_value = Simpson_rule(0, 1, N)
121         error = abs(pi - experimental_value)
122         prev_error = error
123         print(f'N = {N}; error = {error}')
124         N *= 2
125
126     end = timeit.default_timer()
127     print(f'\nTotal time required: {round(end - start, 5)} seconds')
128     print('\n\n\n')
```

```

131 # we can do the same procedure with the in-built
132 # Gaussian quadrature routine
133
134 start = timeit.default_timer()
135
136 print('Gaussian quadrature:')
137 for j in range(1, 10):
138     tolerance = 10 ** (-j)
139     experimental_value = integrate.quadrature(lambda x: pi_function(x), 0, 1, tol=tolerance)
140     error = abs(pi - experimental_value[0])
141     print(f'tolerance = {tolerance}; error = {error}')
142
143 end = timeit.default_timer()
144 print(f'\nTotal time required: {round(end - start, 5)} seconds')

```

Run:  Pi approximation 


  C:\Users\User\PycharmProjects\Laboratories\venv\Script
 Midpoint rule:
 N = 2; error = 0.02076028758667725; ratio = 48.1689
 N = 4; error = 0.005207864804149587; ratio = 3.9863
 N = 8; error = 0.0013020760018958022; ratio = 3.9997
 N = 16; error = 0.0003255207187669029; ratio = 4.0
 N = 32; error = 8.138020654424594e-05; ratio = 4.0
N = 64; error = 2.0345052054171475e-05; ratio = 4.0
N = 128; error = 5.086263021425452e-06
N = 256; error = 1.2715657535800062e-06
N = 512; error = 3.1789143761784544e-07
N = 1024; error = 7.947285940446136e-08
N = 2048; error = 1.9868213740892315e-08
N = 4096; error = 4.9670516588662394e-09
N = 8192; error = 1.2417640249395845e-09
N = 16384; error = 3.104312362722794e-10
N = 32768; error = 7.761213893786589e-11
N = 65536; error = 1.9383605831535533e-11
N = 131072; error = 4.8401282981558325e-12
N = 262144; error = 1.2803091919977305e-12
N = 524288; error = 3.1663560662309465e-13
N = 1048576; error = 3.9968028886505635e-15
N = 2097152; error = 1.2878587085651816e-14

Total time required: 2.20122 seconds

Composite trapezoidal rule:

```
N = 2; error = 0.04159265358979303; ratio = 24.0427
N = 4; error = 0.010416183001557666; ratio = 3.9931
N = 8; error = 0.0026041590987042618; ratio = 3.9998
N = 16; error = 0.0006510415484042298; ratio = 4.0
N = 32; error = 0.00016276041481821935; ratio = 4.0
N = 64; error = 4.0690104138985106e-05; ratio = 4.0
N = 128; error = 1.0172526041074548e-05
N = 256; error = 2.5431315116009046e-06
N = 512; error = 6.357828716829772e-07
N = 1024; error = 1.5894572102936877e-07
N = 2048; error = 3.973643369903357e-08
N = 4096; error = 9.934109534981417e-09
N = 8192; error = 2.4835209444518114e-09
N = 16384; error = 6.208913383431991e-10
N = 32768; error = 1.5522472196494164e-10
N = 65536; error = 3.8819614189833374e-11
N = 131072; error = 9.673151168954064e-12
N = 262144; error = 2.3936408410918375e-12
N = 524288; error = 6.235012506294879e-13
N = 1048576; error = 1.270095140171179e-13
N = 2097152; error = 8.570921750106208e-14
```

Total time required: 2.13879 seconds

```
Run:  Pi approximation ×

Composite Simpson rule:
N = 2; error = 0.008259320256459812
N = 4; error = 2.4026138812693887e-05
N = 8; error = 1.5113108631226169e-07
N = 16; error = 2.3649704417039175e-09
N = 32; error = 3.695710404372221e-11
N = 64; error = 5.782041512247815e-13
N = 128; error = 9.769962616701378e-15
N = 256; error = 4.440892098500626e-16
N = 512; error = 4.440892098500626e-16
N = 1024; error = 1.3322676295501878e-15
N = 2048; error = 1.7763568394002505e-15
N = 4096; error = 0.0
N = 8192; error = 4.440892098500626e-15
N = 16384; error = 7.105427357601002e-15
N = 32768; error = 1.0658141036401503e-14
N = 65536; error = 2.6645352591003757e-15
N = 131072; error = 5.240252676230739e-14
N = 262144; error = 7.638334409421077e-14
N = 524288; error = 3.863576125695545e-14
N = 1048576; error = 2.353672812205332e-14
N = 2097152; error = 3.9968028886505635e-14

Total time required: 4.044 seconds
```

```
Gaussian quadrature:
tolerance = 0.1; error = 0.005948330016764203
tolerance = 0.01; error = 0.0005245136266260886
tolerance = 0.001; error = 1.925165601246448e-05
tolerance = 0.0001; error = 1.3705040213807251e-08
tolerance = 1e-05; error = 4.240220574658338e-08
tolerance = 1e-06; error = 4.240220574658338e-08
tolerance = 1e-07; error = 4.240220574658338e-08
tolerance = 1e-08; error = 4.240220574658338e-08
tolerance = 1e-09; error = 4.240220574658338e-08

Total time required: 0.00234 seconds
```

```

147 # in conclusion I would say that analyzing the elapsed time for each method
148 # we can state that the Gaussian quadrature routine is the most optimal
149 # we may also notice that the Simpson's method works pretty slow in comparison
150 # with the others but it is important that at the same time its error converges
151 # much quicker than the error in the midpoint and trapezoidal rules
152
153 # the converges I am talking about means that at a certain point of decreasing h
154 # the error will stop decreasing and will become approximately constant or even larger
155
156 # that's because for midpoint and trapezoidal rules we get the error formulas equal to:
157 #  $C/n^p$ , where  $2^p = \text{ratio} = 4$  and therefore  $p = 2$ ,  $h = (b-a)/n$  and the obtained result is
158 #  $C \cdot h^2$  (because  $h = 1/n$ );
159 # for the Simpson's rule however things are a little bit different because the ratio
160 # does not converge to a certain value and therefore it may vary from 150 to 60
161 # therefore I will write the expression  $C \cdot h^4$  as mentioned in the lecture presentations
162
163 # it is obvious now that if we increase the value of h to the infinity the error will
164 # tend to zero and therefore further increasings will not be necessary

```

Problem 4.3:

Solution and output:

```

Fresnel integrals.py x Own integral code.py x
1 # in this problem we will work with Fresnel integrals
2 # which are applied in the light diffraction
3 # then we will plot the obtained functions
4 from scipy import special, integrate
5 from numpy import linspace
6 import matplotlib.pyplot as plot
7 import math
8
9
10 # first of all we will compute their real values
11 # on the interval from zero to five
12 arguments = linspace(0, 5, 1000)
13 S, C = special.fresnel(arguments)
14
15 print("The numerical results for Fresnel integrals are:")
16 print(f"scipy.special.fresnel method: C(0:5) = {C[len(C) - 1]}; S(0:5) = {S[len(S) - 1]}")
17
18
19 # plotting the original accurate Fresnel functions
20 plot.subplot(1, 3, 1)
21 plot.plot(arguments, C, color='blue', label='C(x)')
22 plot.plot(arguments, S, color='red', label='S(x)')
23 plot.title('Fresnel integrals real values', color='violet', fontsize=14)
24 plot.xlabel('x - axis', color='green', fontsize=12)
25 plot.ylabel('y - axis', color='green', fontsize=12)
26 plot.legend()
27 plot.grid()

```

```

30     # then we use the general purpose quadrature method
31     def C(t):
32         return math.cos(math.pi * pow(t, 2) / 2)
33
34
35     def S(t):
36         return math.sin(math.pi * pow(t, 2) / 2)
37
38
39     C_quad = list()
40     S_quad = list()
41     for i in range(0, len(arguments)):
42         C1 = integrate.quad(lambda x: C(x), 0, arguments[i])
43         S1 = integrate.quad(lambda x: S(x), 0, arguments[i])
44         C_quad.append(C1[0])
45         S_quad.append(S1[0])
46
47
48     # plotting the obtained functions
49     plot.subplot(1, 3, 2)
50     plot.plot(arguments, C_quad, color='blue', label='C(x)')
51     plot.plot(arguments, S_quad, color='red', label='S(x)')
52     plot.title('General purpose quad method', color='violet', fontsize=14)
53     plot.xlabel('x - axis', color='green', fontsize=12)
54     plot.ylabel('y - axis', color='green', fontsize=12)
55     plot.legend()
56     plot.grid()

```

```

58     print(f"general purpose quadrature method: C(0:5) = {C_quad[len(C_quad) - 1]}; S(0:5) = {S_quad[len(S_quad) - 1]}")
59
60
61     # finally, we apply composite Simpson's rule
62     arguments = linspace(0, 5, 1000)
63     simps_C = list()
64     simps_S = list()
65     for i in range(0, len(arguments)):
66         x_args = linspace(0, arguments[i], 1000)
67         y_args_1 = list()
68         y_args_2 = list()
69
70         for j in range(0, len(x_args)):
71             y_args_1.append(C(x_args[j]))
72             y_args_2.append(S(x_args[j]))
73
74         simps_C.append(integrate.simps(y_args_1, x_args))
75         simps_S.append(integrate.simps(y_args_2, x_args))

```

```

78 # plotting the results
79 plot.subplot(1, 3, 3)
80 plot.plot(arguments, simps_C, color='blue', label='C(x)')
81 plot.plot(arguments, simps_S, color='red', label='S(x)')
82 plot.title('Composite Simpson method', color='violet', fontsize=14)
83 plot.xlabel('x - axis', color='green', fontsize=12)
84 plot.ylabel('y - axis', color='green', fontsize=12)
85 plot.legend()
86 plot.grid()
87 plot.show()
88
89
90 print(f"composite Simpson's method: C(0:5) = {simps_C[len(simps_C) - 1]}; ", end='')
91 print(f"S(0:5) = {simps_S[len(simps_S) - 1]};")

```

```

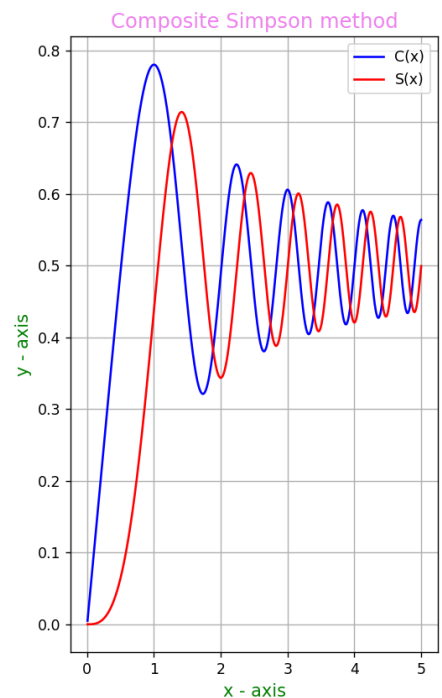
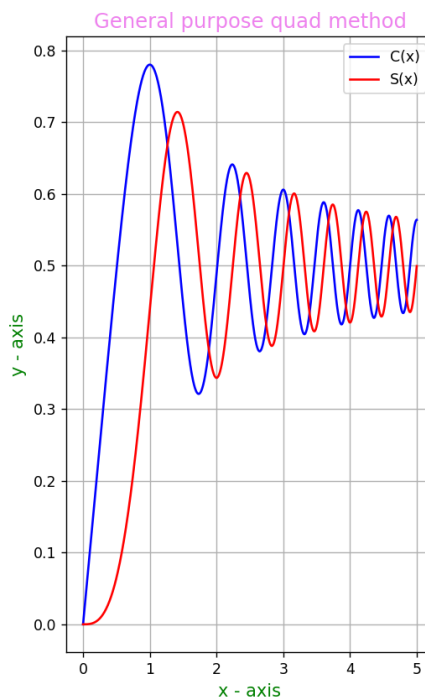
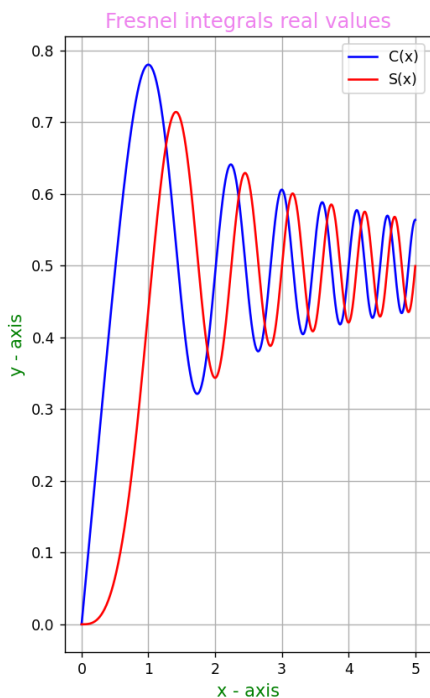
C:\Users\User\PycharmProjects\Laboratories\venv\Scripts\python.exe "C:/Users/User/Pycha
The numerical results for Fresnel integrals are:
scipy.special.fresnel method: C(0:5) = 0.5636311887040122; S(0:5) = 0.4991913819171169

```

```

general purpose quadrature method: C(0:5) = 0.5636311887040117; S(0:5) = 0.49919138191711704
composite Simpson's method: C(0:5) = 0.563631135195887; S(0:5) = 0.4991901114424573;

```



Problem 4.4:

Solution and output:

```
Own integral code.py ×
1  # in this exercise we will not use special python methods
2  # in order to apply numerical integration of different types
3  # but instead will use our own computer routine and develop
4  # some code for Simpson's rule and Gaussian quadrature
5  from scipy.special import legendre
6  import numpy as np
7
8
9  # the first rule finds the corresponding quadratic polynomial
10 # in order to bound the respective region
11 # a and b are the ends of the interval
12 # n is the number of its divisions
13
14 # it is also important to mention that we are dealing with improper
15 # integral of type 2 with x = 1 being the vertical asymptote
16 # as the function tends to infinity
17 # therefore we will have to integrate not exactly from 1
18 def function(x):
19     return pow(x - 1, -5 / 2)
20
21
22 def Simpson_method(f, a, b, n):
23     h = (b - a) / n
24     x = np.linspace(a, b, n + 1)
25     y = f(x)
26
27     riemann_sum = (h / 3) * np.sum(y[0:-1:2] + 4 * y[1::2] + y[2::2])
28     return riemann_sum
29
30
31 # showing the results
32 print('Results for the Simpson method:')
33 N = 4
34 for i in range(0, 5):
35     print(f'n = {N}; integral value = {Simpson_method(lambda x: function(x), 1 + 1e-8, 4, N)}')
36     N *= 2
```

```

39     # then we apply the Gaussian quadrature which initially
40     # works for the interval [-1; 1]
41     # for that we will need to generate nodes and weights
42     # of the quadrature using Legendre polynomials of degree n
43     # their solutions will become nodes and the weights will be
44     # determined using the special formula
45     def Legendre_derivative(coefficients, degree, x):
46         value = 0
47         index = degree
48         j = 1
49         while index != 0:
50             value += coefficients[j - 1] * index * pow(x, degree - j)
51             j += 1
52             index -= 1
53
54         return value

```

```

57     # a and b are the limits of integration
58     def Legendre_function(a, b, t):
59         return pow((b + a + t * (b - a)) / 2 - 1, -5/2)
60
61
62     def Gaussian_quadrature(nod, weight, a, b):
63         gaussian_sum = 0
64         for index in range(0, len(nod)):
65             gaussian_sum += weight[index] * Legendre_function(a, b, nod[index])
66
67         return gaussian_sum * (b - a) / 2

```


```

70     # the integral will be calculated by multiplying
71     # the corresponding nodes and weights consecutively
72     print()
73     print()
74     print()
75     print('Results for the Gaussian quadrature method:')
76     N = 4
77     for i in range(0, 5):
78         coefs = list(legendre(N))
79         nodes = np.roots(coefs)
80         weights = list()
81         for k in range(0, len(nodes)):
82             weights.append(2 / ((1 - pow(nodes[k], 2)) * (Legendre_derivative(coefs, N, nodes[k])) ** 2))
83
84         print(f'n = {N}; integral value = {Gaussian_quadrature(nodes, weights, 1, 4)}')
85     N *= 2

```



```
88 # in conclusion I would say that the Gaussian quadrature method
89 # converges much faster than the Simpson's method in approaching
90 # to the real value of the integral, but we I also observed that
91 # the quadrature at first tends to infinity and later starts to
92 # tend to the real value; however we can state that the Gaussian
93 # quadrature converges as n goes to infinity
```

Run:  Own integral code ×

C:\Users\User\PycharmProjects\Laboratories\venv\Scripts

Results for the Simpson method:

n = 4; integral value = 2.500000029650861e+19
n = 8; integral value = 1.2500000148254304e+19
n = 16; integral value = 6.250000074127152e+18
n = 32; integral value = 3.125000037063576e+18
n = 64; integral value = 1.5625000185317883e+18

Results for the Gaussian quadrature method:

n = 4; integral value = 27.564429010282502
n = 8; integral value = 183.50164887430475
n = 16; integral value = 1336.8142693912134
n = 32; integral value = 9271.89927886087
n = 64; integral value = (0.4305645559152461+0j)