



MALMÖ UNIVERSITY

Computational Physics: Introductory Course

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Assignment 9: Monte-Carlo methods

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Exercise 1

Compute the definite integral

$$\int_0^{10} x e^{-x} dx$$

using a Monte-Carlo method with $10^2, 10^3, \dots, 10^7$ random numbers. Compare with the analytical result. Present your results in a table

```

1 import numpy as np
2
3 # Define the function to integrate
4 def f(x):
5     return x * np.exp(-x)
6
7 # Define the limits of integration
8 a, b = 0, 10
9
10 # Define the analytical solution
11 analytical_solution = 1 - np.exp(-b) * (b + 1)
12
13 # Define the number of random samples to use
14 num_samples = [10**2, 10**3, 10**4, 10**5, 10**6, 10**7]
15
16 # Define a list to store the results
17 results = []
18
19 # Loop over the number of samples
20 for n in num_samples:
21     # Generate n random samples in the interval [a, b]
22     x = np.random.uniform(a, b, n)
23
24     # Evaluate the function at the random samples
25     y = f(x)
26
27     # Compute the integral approximation using the Monte Carlo method
28     integral_approximation = (b - a) * np.mean(y)
29
30     # Compute the error with respect to the analytical solution
31     error = np.abs(integral_approximation - analytical_solution)
32
33     # Append the results to the list
34     results.append([n, integral_approximation, error])
35
36 # Print the results in a table
37 print("Number of samples\tIntegral approximation\tAnalytical Solution\tError")
38 for n, integral, error in results:
39     print(f"{n}\t\t\t\t\t{integral:.6f}\t\t\t\t\t{analytical_solution:.6f}\t\t\t\t\t{error:.6f}")

```

40				
41	Number of samples	Integral	Analytical Solution	Error
42	100	1.046059	0.999501	0.046559
43	1000	0.945703	0.999501	0.053798
44	10000	0.998178	0.999501	0.001323
45	100000	1.002517	0.999501	0.003016
46	1000000	0.999850	0.999501	0.000350
47	10000000	0.998962	0.999501	0.000538

As it is clear the number of samples makes a difference in the error. We can see that higher number of samples gives a more precise approximation when using the Monte Carlo simulation.

Exercise 2

The density in g/cm³ in the interior of the sun can be approximated by the following expression:

$$\rho(x) = (2139x + 155)e^{13.8x},$$

where x is a radial coordinate, stretching from the center ($x = 0$) to the surface of the sun ($x = 1$). Using a Monte-Carlo method (see section 22.6 in the lecture notes), compute the mass and moment of inertia of the sun and compare your result with other sources. For the radius of the sun, use 6.96×10^8 m.

```

1 import numpy as np
2
3 # Constants
4 R_sun = 6.96e8 # Radius of the sun in meters
5 N = 100000 # number of points in D0
6
7 # Define density function
8 def rho(x):
9     return (2139*x + 155)*np.exp(-13.8*x)
10
11 # Initialize variables
12 fsum = 0
13
14 # Generate random points and compute sums
15 for i in range(N):
16     # Generate random radial coordinate
17     x = -R_sun + 2*R_sun*np.random.rand()
18     y = -R_sun + 2*R_sun*np.random.rand()
19     z = -R_sun + 2*R_sun*np.random.rand()
20     r2 = x**2 + y**2 + z**2
21     r = np.sqrt(r2/R_sun**2)
22     # Compute density at current radial coordinate
23     if r2 < R_sun**2:
24         fsum = fsum + rho(r)*(x**2+ y**2)

```

```

25     mass = fsum + rho(r)
26
27
28 # Compute moment of inertia contribution and update moment of inertia sum
29 I = 8 * R_sun ** 3/N*fsum
30
31
32
33 # Print results
34 print("Moment of inertia: %6.4e g*cm^3" %I)
35 print("Mass: ", mass)
36
37 # Moment of inertia: 6.0182e+43 g*cm^3
38 # Mass: 2.2312626819951719e+21

```

Exercise 3

In assignment 6 and exercise 2, we studied Rutherford scattering in two dimensions by solving a set of coupled first order differential equations. In particular, for a given impact parameter b you could find the corresponding numerical value of the scattering angle θ . In this exercise, you will extend the code and simulate thousands of scattered alpha particles and investigate how many alpha particles that are scattered at different angles. Finally, you will compare your results with predictions from the differential cross section for Rutherford scattering. To solve the exercise, you need to consult the Physics background section below. Assume the same values for the parameters as in assignment 6 and exercise 2. That is, $K = 3.47710^{13} \text{ fm}^3 \text{ fs}^{-2} e^{-2}$, $Q = 79e$, $q = 2e$ and $v = 1.5310^7 \text{ fm/fs (m/s)}$. The scattering angles that will be investigated are in the interval $50 - 180^\circ$. Find the impact parameter b_{\max} corresponding to 50° . Now follow the steps below:

- For each alpha particle, construct coordinates b_y and b_z of the impact parameter b by assigning them uniformly distributed random numbers between 0 and b_{\max} . The impact parameter is then given as $b = \sqrt{b_y^2 + b_z^2}$
- If $b < b_{\max}$, use the code obtained in assignment 6, exercise 2 to determine the numerical scattering angle. Store the scattering angle in an array.
- Repeat steps (a) and (b) for at least 1 000 alpha particles and plot a histogram in the range $50^\circ - 180^\circ$ based on the array with the scattering angles. Choose a suitable bin number.
- Compare the resulting histogram with the predicted number of scattered alpha particles in each bin. See figure 1 for expected results.

```
1 import numpy as np
2 from scipy.integrate import solve_ivp
3 import matplotlib.pyplot as plt
4
5 # Define the constants
6 K = 3.477e13 # fm^3 fs^-2 e^-2
7 q = 2 # elementary charge
8 Q = 79 # gold nucleus charge
9 v = 1.53e7 # fm/fs
10 theta = np.deg2rad(50)
11 m = 6.64e-27
12 b_max = (K * Q * q / v**2) * (1 / np.tan(theta/2))
13 # Generate random impact parameters for 1000 alpha particles
14 np.random.seed(42)
15 by = np.random.uniform(0, b_max, size=1000)
16 bz = np.random.uniform(0, b_max, size=1000)
17 b_values = np.sqrt(by**2 + bz**2)
18
19
```

```

20 print("B is equal to: ", b_values)
21 print("B_max is equal to: ",b_max)
22
23
24
25 def f(t, y):
26     x, y, vx, vy = y
27     r = np.sqrt(x**2 + y**2)
28     fx = (K * q * Q * x) / r**3
29     fy = (K * q * Q * y) / r**3
30     return [vx, vy, fx, fy]
31
32 # Define the time range
33 t = np.linspace(0, 0.1, 1000)
34
35 # Define the initial conditions
36 x0 = -1e6
37 y0 = 25
38 vx0 = v
39 vy0 = 0
40
41 # Solve the differential equations
42 sol = solve_ivp(f, [0, 0.1] , [x0,y0,vx0,vy0] , t_eval= t)
43
44 # extract the solution
45 x, y, vx, vy = sol.y
46 theta_numerical = []
47
48
49 # Solve the differential equations and calculate the scattering angles
50 for i,b in enumerate(b_values):
51     if(b<b_max):
52         y0 = [x0, b, vx0, vy0]
53         sol = solve_ivp(f, [0, 0.1], y0, t_eval=t, rtol=1e-10, atol=1e-10)
54         x, y, vx, vy = sol.y
55         dy = y[1000-1]-y[1000-11]
56         dx = x[1000-1]-x[1000-11]
57         thetanum = np.rad2deg(np.arctan(dy/dx))
58         if dy < 0 and dx < 0:
59             thetanum = 180 - thetanum
60         elif dy > 0 and dx < 0:
61             thetanum = 180 + thetanum
62         elif dy < 0 and dx > 0:
63             thetanum = - thetanum
64
65         if thetanum >= 50 and thetanum <= 180:
66             theta_numerical.append(thetanum)
67
68
69 # Specify the range and number of bins for the histogram
70 binedge = np.arange(50,181,10)

```

```

71
72 # Plot the histogram and the analytical solution
73 plt.hist(theta_numerical, bins=binedge)
74
75 # Add labels and title
76 plt.xlabel('Scattering angle (degrees)')
77 plt.ylabel('Frequency')
78 plt.title('Histogram of Scattering Angles')
79
80 # Add legend
81 plt.legend()
82
83 # Show the plot
84 plt.show()
85
86 # I tried to add
87 # theta_analytical = np.linspace(50, 180, 100)
88 # def total_cross_section(theta):
89     #return (4 * np.pi * K**2 * q**2 * Q**2) / (m * v**2 * np.sin(theta)
90         /2)**4
91 # cross_sections_analytical = total_cross_section(theta_analytical)
92 # Plot the histogram and the analytical solution
93 # plt.hist(theta_numerical, bins=binedge, density=True, label='Numerical')
94 # plt.plot(theta_analytical, cross_sections_analytical, label='Analytical
    ')

```

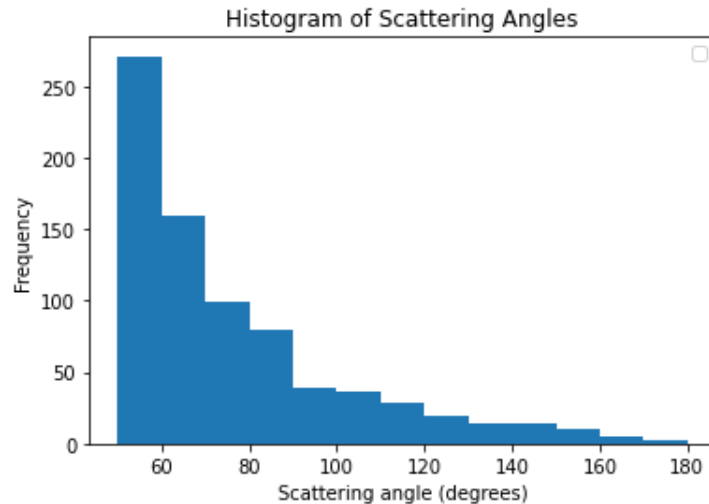


Figure 1: Histogram when simulating Rutherford with 1000 alpha particles

Now it is working properly I think for the histogram. However, I tried to add the analytical solution but I did not manage to make it work. I think I have a huge problem understanding this exercise to be honest. See the comments please on the code and let me know if there is anything to be fixed. Thank you in advance.