

# Worksheet 10: Integration

June 6, 2014

## General Remarks

- The deadline for the worksheets is **Wednesday, 25 June 2014, 10:00** for the tutorial on Friday and **Friday, 27 June 2014, 10:00** for the tutorials on Tuesday and Wednesday.
- On this worksheet, you can achieve a maximum of 10 points.
- To hand in your solutions, send an email to
  - Johannes (zeman@icp.uni-stuttgart.de; Tuesday, 9:45–11:15)
  - Tobias (richter@icp.uni-stuttgart.de; Wednesday, 15:45–17:15)
  - Shervin (shervin@icp.uni-stuttgart.de; Friday, 11:30–13:00)

## Task 10.1: One-dimensional Integration (7 points)

The error function is a sigmoidal function that is used in statistics and some other areas. It is defined as

$$\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-\tau^2} d\tau \quad (1)$$

The job in this task is to implement the error function.

- **10.1.1** (1 point) Implement a Python function `integrate1d_midpoint(f,a,b,N)` that approximates  $\int_a^b f(x) dx$  using the midpoint rule with  $N$  support points.
- **10.1.2** (1 point) Implement a Python function `integrate1d_romberg(f,a,b,N)` that approximates  $\int_a^b f(x) dx$  using Romberg's method with  $N$  support points on the finest level, i.e. use  $h = \frac{b-a}{N-1}$  as smallest step size.

**Hint** Note that the implementation in the lecture script uses  $N = 2^{kmax}$  support points, so choose  $kmax$  accordingly.

- **10.1.3** (1 point) Implement a Python function `integrate1d_mc(f,a,b,N)` that approximates  $\int_a^b f(x) dx$  using  $N$  steps of Monte-Carlo Integration.
- **10.1.4** (1 point) Implement the Python functions `erf_midpoint(x,N)`, `erf_romberg(x,N)` and `erf_mc(x,N)` that compute the error function for  $N$  points with the corresponding integration rules.
- **10.1.5** (1 point) Evaluate the error function  $\operatorname{erf}(x)$  on the interval  $[0, 2.0]$  at 100 equidistant points  $x$  using `erf_romberg` with  $N = 64$ , and plot the resulting approximation of the error function.
- **10.1.6** (1 point) Compute a reference value  $e_{\text{ref}} = \operatorname{erf}(1)$  using Romberg's method with  $N = 512$ . Create a loglog plot of the accuracy of all implemented methods when computing  $\operatorname{erf}(1)$  for  $N \in \{4, 8, 16, 32, 64, 128, 256, 512\}$ .

- **10.1.7** (1 point) Use the `integrate1d_*` functions to approximate  $\pi$  via

$$\frac{\pi}{4} = \int_0^1 \sqrt{1-x^2} dx \quad (2)$$

Create a loglog plot of the accuracy of the different integration schemes versus the number of points  $N \in \{4, 8, 16, \dots, 512\}$ . Use `numpy.pi` as reference value. Why does the accuracy behave differently this time?

### Task 10.2: Multidimensional Monte-Carlo Integration (3 points)

In Gaussian units ( $\epsilon = 1$ ), the interaction energy of two homogenously charged bodies  $V_1$  and  $V_2$  is given by

$$E = \int_{V_1} \int_{V_2} \frac{1}{|\vec{x}_1 - \vec{x}_2|} d\vec{x}_1 d\vec{x}_2$$

In this task, we consider the interaction of two homogenously charged cubes (in 3D;  $Q = 1$ )  $V_1$  and  $V_2$  with side length 1 at a distance  $d$  (see the figure below).

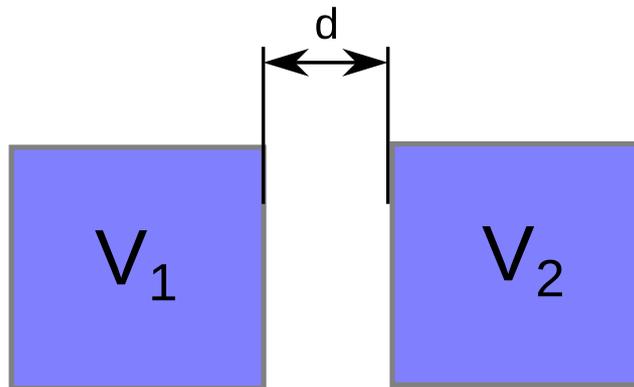


Figure 1: Two homogenously charged cubes  $V_1$  and  $V_2$  at distance  $d$ .

- **10.2.1** (2 points) Write a Python function that computes the interaction energy of the cubes at distance  $d$  using Monte-Carlo integration with  $N = 1000$  steps.
- **10.2.2** (1 point) Compute the interaction energy for 10 different values of  $d$  between  $10^{-2}$  and  $10^2$ . Plot the interaction energy  $E$  versus the distance  $d$  in log-log-scale. In the same figure, add a plot of the interaction energy of two unit point charges at distance  $d$ .