

Problem Sheet 1

discussion: week of Monday, 17.10.2022

1.1. Consider the approximation of the integral $\int_0^1 f(x)$ by the trapezoidal rule

$$\int_0^1 f(x) dx \approx T(h) := \sum_{i=0}^{N-1} \frac{h}{2} (f(x_i) + f(x_{i+1})), \quad x_i = ih, \quad h = \frac{1}{N}$$

as well as the box rule

$$\int_0^1 f(x) dx \approx R(h) := \sum_{i=0}^{N-1} h f(x_i), \quad x_i = ih, \quad h = \frac{1}{N}.$$

- a) Write one program that, given N and a function f , returns $T(h)$ and one that returns $R(h)$ with $h = 1/N$.
- b) For $f(x) = e^x$ and $N(i) = 2^i$, $i = 1, \dots, 10$, set $h(i) = 1/N(i)$. For these 10 values of h , compute the actual errors $\mathbf{err}(i) = \int_0^1 e^x dx - T(h(i)) = e - 1 - T(h(i))$. Plot the (absolute value) of the true error in a “log-log” plot. That is, in **matlab** notation with the vectors h and \mathbf{err} (both of length 10)

`loglog(h, abs(err))`

Does the true error behave like Ch^2 ?

Do the same thing for the box rule as well. What error behaviour do you see?

Remark: In **python** `loglog` plotting is realized with `matplotlib.pyplot.loglog` and the array limits have to be adapted since arrays start at 0 in **python**.

1.2. Again, consider the approximation of the integral $\int_0^1 f(x)$ by the trapezoidal rule $T(h)$ from exercise 1.1. The aim is to design an error estimator as it was done for the “box scheme” in the lecture.

- a) For the error estimator, make the ansatz

$$\int_0^1 f(x) dx - T(h) \stackrel{!}{=} Ch^2$$

and compute C from (calculated) $T(h)$ and $T(h/2)$. The error estimator is then $E(h) = Ch^2$.

- b) Modify your program from exercise 1.1(a) such that it also computes the error estimator $E(h)$. For the test-case in exercise 1.1(b) do also compute the ratios $\mathbf{r}(i) := \mathbf{err}(i)/E(h(i))$ of the true error to estimated error. Plot the ratio \mathbf{r} in a semilogarithmic plot versus h . That is, in **matlab** notation with the vectors h and \mathbf{r} (both of length 10)

`semilogx(h, r)`

1.3. Let the data $(0, 0)$, $(1, 2)$, $(4, 8)$ be given.

- a) Write down the (quadratic) interpolating polynomial p in the Lagrange form, i.e., $p(x) = \sum_{i=0}^2 f_i \ell_i(x)$.
- b) The quadratic interpolating polynomial $p(x) = a_0 + a_1x + a_2x^2$ can also be determined by solving a linear system of equations for the coefficients a_0 , a_1 , a_2 . Formulate the system and solve for the a_i .
- c) Check your approximation $p(2)$ using the **matlab** or **numpy** routines `polyfit(x, f, n)` and `polyval`. Note that `polyfit` also returns the coefficients a_i of part c). Compare.

Note: The use of `polyfit` and `polyval` in the context of polynomial interpolation should rather be viewed as a “quick and dirty” method since `polyfit` is based on polynomial approximation in the monomial basis $\{1, x, x^2, \dots, x^n\}$ and therefore numerically unsuitable for large/largish n .

- 1.4.** a) Write a routine that has as input the vectors \mathbf{x} and \mathbf{f} of knots and data (both vectors have the length $n + 1$) and returns the value of the interpolating polynomial at the point 0, i.e., $p(0)$ and the polynomial p of degree n satisfies $p(\mathbf{x}_i) = \mathbf{f}_i$ for all i . To simplify matters, you may use `polyfit` and `polyval`.
- b) Define the values $h_i := 2^{-i}$, $i = 0, 1, \dots$. Based on a), write a function whose input is a function handle f and integers $m, n \in \mathbb{N}_0$. The output is a matrix $N \in \mathbb{R}^{(m+1) \times (n+1)}$ with entries $N_{j,k} := p_{j,k}(0)$, where $j = 0, \dots, m$, $k = 0, \dots, n$ and $p_{j,k}$ is the polynomial of degree k that interpolates the data $(h_{j+\ell}, f(h_{j+\ell}))$, $\ell = 0, \dots, k$.
- Remark:* This matrix N is the so-called *Neville-scheme* for interpolation although it is usually set up differently. (This will be done later in class.)
- c) Apply your routine of b) to the function

$$f(h) = \frac{e^h - 1}{h}$$

and select $m = 10$ and $n = 2$. You expect the columns of N to converge to $\lim_{h \rightarrow 0} f(h) = 1$. To check that, plot (in `matlab` notation)

`loglog(h(1:m+1), abs(N(:,1)-1), h(1:m+1), abs(N(:,2)-1), h(1:m+1), abs(N(:,3)-1))`

What convergence behavior do you observe for these 3 graphs?

- 1.5.** A common way to approximate a function by a polynomial is to use Taylor polynomials around a point x_0 given as

$$f(x) \simeq \sum_{j=0}^n \frac{f^{(j)}(x_0)}{j!} (x - x_0)^j = f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2} (x - x_0)^2 + \dots$$

Compute the Taylor polynomial of degree $n = 2$ for the function $f = \sin(x)$ at $x_0 = 0$ and at $x_0 = \frac{\pi}{2}$. Compare it with the interpolating polynomial of order 2 from the lecture, what do you observe?