

# Problem sheet 2

Please submit the question marked \* to your tutor.

## E2.1

Use elementary calculus to prove that

$$\max_{x \in [x_0, x_1]} (x - x_0)(x_1 - x) = \frac{1}{4}(x_0 - x_1)^2$$

and hence complete the proof of Corollary 2.2.

## E2.2 \*

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

(a) Write down the linear polynomial  $p_1$  that interpolates  $f$ .

(b) Use Corollary 2.2 to bound  $|e(x)|$  over  $x \in [0, 1]$  where  $e := f - p_1$ . Compare your bound with the actual error, which can be found analytically in this case.

(c) Find the quadratic polynomial  $p_2$  that interpolates  $f$  at  $x_0, x_1$ , and the additional point  $x_2 = 2$ .

(d) Next find the cubic polynomial  $p_3$  which interpolates  $f$  at  $x_0, x_1, x_2$  and the additional point  $x_3 = -1$ . Comment on the relation between  $f$  and  $p_3$ .

## E2.3

The following function interpolates a given function  $f$  between the two points  $x_0$  and  $x_1$  (input arguments  $x_0$  and  $x_1$ ).  $p_1$  is a vector containing all the values of the function  $p_1(x)$  at 100 equally spaced points in  $[x_0, x_1]$ .

```
In [1]: import numpy as np
def f(x):
    return np.sqrt(x)

def linear_interp(x0, x1, mesh):
    vecone = np.ones(100,)
    p1 = f(x0) + ((f(x1) - f(x0)) / (x1 - x0)) * (mesh - x0)
    return p1

mesh = np.linspace(0, 1, 100)
p1 = linear_interp(0, 1, mesh)
```

Write code to plot  $p_1(x)$  and  $f(x)$  (in different colours) on one graph, and then the error  $e(x) := f(x) - p_1(x)$  on another graph.

```
In [2]: import matplotlib.pyplot as plt
# fill in
```

## E2.4\*

In this exercise, numerically investigate

$$e_h := \max_{x \in [-h, h]} |f(x) - p_1(x)|$$

for  $h > 0$ , where the maximum is taken over 100 equally spaced points in  $[-h, h]$ .

Run your program for the case  $f(x) = \exp(x)$  and draw up a table of  $e_h$  against  $h$ , for  $h = 1/8, 1/16, 1/32, \dots, 1/256$ . Investigate the convergence rate as  $h \rightarrow 0$  experimentally, by making the conjecture  $e_h = Ch^\alpha$  where  $C$  and  $\alpha$  are constants and then finding approximations to  $\alpha$ .

Note that as in lectures, computing  $\log_2(e_{2h}/e_h)$  will give approximations to  $\alpha$ . This can be done either by editing the program or using a calculator.

Repeat the exercise for  $f(x) = \sin x$ . Explain your observations by appealing to the theory from lectures.

```
In [3]: def f(x):
    return np.exp(x)

#takes as input h and returns the maximum error eh as defined above
def max_err_linear_interp(h):
    #fill in
    return eh
```