# UNIVERSITY OF OSLO

## Faculty of mathematics and natural sciences

Exam in:	MAT3110/MAT4110 — Introduction to numerical analysis
Day of examination:	19 January 2021
Examination hours:	09:00-13:00
This problem set con	sists of 7 pages.
Appendices:	None
Permitted aids:	All written aids

### Please make sure that your copy of the problem set is complete before you attempt to answer anything.

### Note:

- There are in total 11 subproblems (1, 2a, 2b, ...), and you can get 5–10 points for each sub-problem, for a total of 100 points.
- All answers must be justified.

### Problem 1 Root finding

Let  $f(x) = \cos(x) - x$ . This function has a single root  $x_0$  somewhere in [0, 1], and we wish to compute it.

### 1a (10 points)

Perform two steps with both the bisection method and Newton's method. Justify your choice of starting values.

### 1b (10 points)

Which of the two methods can we expect to be the most accurate after several iterations? Justify your answer.

#### Solution:

#### 1a

For the bisection method we choose  $x_0 = 0, x_1 = 1$ . Then  $f(x_0) = 1 > 0$ and  $f(x_1) = \cos(1) - 1 < 0$ . Since f is continuous, it has a zero in  $(x_0, x_1)$ , and the bisection method will be able to find it. We compute

$$x_2 = \frac{x_0 + x_1}{2} = \frac{1}{2}$$

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and note that  $f(x_2) = \cos(1/2) - 1/2 > 0$ , and therefore the new interval will be  $(x_2, x_1)$ . We finally get

$$x_3 = \frac{x_1 + x_2}{2} = \frac{3}{4}.$$

For Newton's method, we note that  $f'(x) = -\sin(x) - 1$ , which is nonzero in [0, 1]. Hence, as long as the iteration stays in [0, 1], the method will converge. We set e.g.  $x_0 = 0$  and get

$$x_1 = x_0 - \frac{f(x_0)}{f'(x_0)} = -\frac{1}{-1} = 1,$$
  
$$x_2 = x_1 - \frac{f(x_1)}{f'(x_1)} = 1 - \frac{\cos(1) - 1}{-\sin(1) - 1} \approx 0.7504.$$

1b

The bisection method converges linearly, while Newton's method converges quadratically, so we can expect Newton's method to be the most accurate.

### Problem 2 Polynomial interpolation (10 points)

Let  $f: [0,2] \to \mathbb{R}$  be a given function and let  $n \in \mathbb{N}$ . We wish to interpolate f using an *n*-th order polynomial p.

- Explain how we should do this in order to minimize the maximal error  $||f p||_{C([0,2])} = \sup_{x \in [0,2]} |f(x) p(x)|.$
- Give an estimate of  $||f p||_{C([0,2])}$ .

**Solution:** Assume first that we are on the interval [-1, 1], and let  $x_0, \ldots, x_n \in [-1, 1]$  be distinct interpolation points. We let p interpolate f over these points:

$$p(x) = \sum_{k=0}^{n} f(x_k) \prod_{\substack{l=0,...,n\\l \neq k}} \frac{x - x_l}{x_k - x_l}.$$

The basic error estimate is

$$||f - p||_{C([-1,1])} \leq \frac{||f^{(n+1)}||_{C([-1,1])}}{(n+1)!} ||w_n||_{C([-1,1])}$$

where  $w_n(x) = \prod_{k=0}^n (x - x_k)$ . The term  $||w_n||_{C([-1,1])}$  is the least possible when  $x_0, \ldots, x_n$  are chosen as the Chebysheff points, yielding  $||w_n||_{C([-1,1])} = 2^{-n}$ , and therefore

$$||f - p||_{C([-1,1])} \leq \frac{||f^{(n+1)}||_{C([-1,1])}}{2^n(n+1)!}$$

(Continued on page 3.)

Any other choice of interpolation points will yield a larger right-hand side.

To transform this analysis to the interval [0, 2], it is enough to note that the two intervals are of the same length, and that translating a function does not change its norm, so the same results apply:

$$||f - p||_{C([0,2])} \leq \frac{||f^{(n+1)}||_{C([0,2])}}{2^n(n+1)!}$$

### Problem 3 Polynomial interpolation

Let  $f: [0,1] \to \mathbb{R}$  be the function  $f(x) = \cos(2x) - e^x$ . For some  $n \in \mathbb{N}$ , let p be the *n*-th order polynomial which interpolates f over the uniform grid  $0, 1/n, \ldots, 1$ .

#### 3a (10 points)

Prove that  $||f - p||_{C([0,1])} \to 0$  as  $n \to \infty$ . (Here,  $||f - p||_{C([0,1])} = \sup_{x \in [0,1]} |f(x) - p(x)|$ .)

#### 3b (10 points)

How large must n be in order to guarantee that  $||f - p||_{C([0,1])} \leq 10^{-10}$ ?

*Hint:* In this problem you might (or might not) need Stirling's approximation:

 $m! \geqslant m^m e^{-m}.$ 

**Solution:** The basic error estimate is: For every  $x \in [0, 1]$  there is some  $\xi \in [0, 1]$  such that

$$|f(x) - p(x)| \leq \frac{|f^{(n+1)}(\xi)|}{(n+1)!} \prod_{k=0}^{n} |x - x_k|.$$

We have

$$|f^{(m)}(\xi)| \leqslant |\frac{d^m}{d\xi^m}\cos(2\xi)| + |\frac{d^m}{d\xi^m}e^{\xi}| \leqslant 2^m + e.$$

Moreover,  $|x - x_k| \leq 1$ , so we get

$$|f(x) - p(x)| \leq \frac{2^{n+1} + 1}{(n+1)!} \quad \forall x \in [0,1].$$

Using Stirling's formula we get

$$||f - p||_{C([0,1])} \leq \frac{2^{n+1} + 1}{(n+1)!} \leq \frac{e^{n+1}(2^{n+1} + 1)}{(n+1)^{n+1}}.$$

3a

It is clear that the expression above converges to zero as  $n \to \infty$ .

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### 3b

Testing different values of n shows that n = 29 gives an upper bound of  $\approx 4.3 \times 10^{-11}$ .

Alternative solution: If, say,  $x \in [x_m, x_{m+1}]$  then

$$\prod_{k=0}^{n} |x - x_k| = \frac{1}{n^{n+1}} \prod_{k=0}^{n} |xn - k| \leq \frac{(m+1)!(n-m)!}{n^{n+1}} \leq \frac{(n+1)!}{n^{n+1}},$$

where the last inequality follows from  $1 \leq \binom{n+1}{m+1} = \frac{(n+1)!}{(m+1)!(n-m)!}$ . We get

$$|f(x) - p(x)| \leq \frac{\|f^{(n+1)}\|}{(n+1)!} \frac{(n+1)!}{n^{n+1}} \leq \frac{2^{n+1}+1}{n^{n+1}}.$$

3a

It is clear that  $\frac{2^{n+1}+1}{n^{n+1}} \to 0$  as  $n \to \infty$ .

### 3b

With this improved estimate we find that n = 12 gives an upper bound  $\sim 7.66 \times 10^{-11}$ .

### Problem 4 QR factorization

Let A, Q and R be the matrices

$$A = \begin{pmatrix} 0 & 1\\ \sqrt{2} & 3\sqrt{2}\\ 0 & 1 \end{pmatrix}, \quad R = \sqrt{2} \begin{pmatrix} 1 & 3\\ 0 & 1\\ 0 & 0 \end{pmatrix}, \quad Q = \frac{1}{\sqrt{2}} \begin{pmatrix} 0 & 1 & 1\\ \sqrt{2} & 0 & 0\\ 0 & 1 & -1 \end{pmatrix}.$$

Note that A = QR (you don't have to show this).

### 4a (5 points)

Explain what it means that QR is the QR factorization of A. Justify your answer.

### 4b (10 points)

Find the least squares solution of the equation

$$Ax = b$$
, where  $b = \begin{pmatrix} 1\\ 2\\ 3 \end{pmatrix}$ .

Solution:

### 4a

A QR factorization consists of an orthogonal matrix Q and an upper triangular matrix R (with 1's as its first nonzero element in each row, if the factorization is in normal form). It is straightforward to see that  $Q^{\mathsf{T}}Q = I$ , so Q is orthogonal, and that R is upper triangular.

#### 4b

We wish to minimize  $||Ax - b|| = ||Rx - Q^{\mathsf{T}}b||$ . Write

$$R = \begin{pmatrix} R_1 \\ 0 \end{pmatrix}, \quad Q^{\mathsf{T}}b = \begin{pmatrix} c_1 \\ c_2 \end{pmatrix}, \quad R_1 = \sqrt{2} \begin{pmatrix} 1 & 3 \\ 0 & 1 \end{pmatrix}, \quad c_1 \in \mathbb{R}^2, c_2 \in \mathbb{R}.$$

Then  $||Ax - b||^2 = ||Rx - Q^{\mathsf{T}}b||^2 = ||R_1x - c_1||^2 + ||c_2||^2$ , so we need to minimize the first term; to this end, we solve  $R_1x = c_1$ . We compute

$$c_1 = \begin{pmatrix} 2\\ 2\sqrt{2} \end{pmatrix} \Rightarrow x = R_1^{-1}c_1 = \begin{pmatrix} \sqrt{2} - 6\\ 2 \end{pmatrix}.$$

### Problem 5 SVD (10 points)

Compute the singular value decomposition (SVD) of

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 6 \end{pmatrix}.$$

*Hint:* You may use the fact that one of the eigenpairs of the normal matrix  $A^{\mathsf{T}}A$  is  $\lambda_1 = 50$ ,  $\mathbf{v}_1 = \frac{1}{\sqrt{5}} \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ .

**Solution:** We see that A is non-invertible, so  $A^{\mathsf{T}}A$  must also be non-invertible, whence the second eigenvalue is  $\lambda_2 = 0$ . The second eigenvector is chosen such that  $V = (\mathbf{v}_1 \ \mathbf{v}_2)$  is orthogonal; this is achieved by letting  $\mathbf{v}_2 = \frac{1}{\sqrt{5}} \begin{pmatrix} 2 \\ -1 \end{pmatrix}$ . We get the two singular values

$$\sigma_1 = \sqrt{50} = 5\sqrt{2}, \qquad \sigma_2 = 0.$$

Setting  $S = \text{diag}(\sigma_1, \sigma_2)$  we want to find an orthogonal matrix U such that  $A = USV^{\mathsf{T}}$ , or US = AV. Writing  $U = (\mathbf{u}_1 \quad \mathbf{u}_2)$ , we have

$$US = \begin{pmatrix} \sigma_1 \mathbf{u}_1 & 0 \end{pmatrix} \Rightarrow \mathbf{u}_1 = \frac{1}{\sigma_1} A \mathbf{v}_1 = \frac{1}{\sqrt{10}} \begin{pmatrix} 1 \\ 3 \end{pmatrix}.$$

Finally, we let  $\mathbf{u}_2$  be such that U is orthogonal:  $\mathbf{u}_2 = \frac{1}{\sqrt{10}} \begin{pmatrix} 3 \\ -1 \end{pmatrix}$ . Thus,  $A = USV^{\mathsf{T}}$  with  $U = \frac{1}{\sqrt{10}} \begin{pmatrix} 1 & 3 \\ 3 & -1 \end{pmatrix}$ ,  $S = \begin{pmatrix} 5\sqrt{2} \\ 0 \end{pmatrix}$ ,  $V = \frac{1}{\sqrt{5}} \begin{pmatrix} 1 & 2 \\ 2 & -1 \end{pmatrix}$ .

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### Problem 6

We wish to approximate the integral  $I(f) = \int_0^{20} f(x) dx$  of a function f.

#### 6a (5 points)

If we wish to approximate I(f) using an 5-point quadrature rule, which quadrature rule should we choose to make the error as small as possible? Justify your answer.

#### 6b (10 points)

Recall that the Gauss quadrature of order 3 on the interval [-1, 1] is

$$\int_{-1}^{1} g(x) \, dx \approx f\left(-\sqrt{1/3}\right) + f\left(\sqrt{1/3}\right). \tag{1}$$

Write down the composite integration rule over N = 2 subintervals which approximates I(f). Use the quadrature rule (1) in the composite method.

### Solution:

#### **6**a

The *n*-point Gauss quadrature rule has order 2n-1, which is the largest possible. We should therefore use the 3-point Gauss quadrature rule.

#### **6**b

Translating the interval [-1, 1] to [0, 10] gives quad. points and weights

$$x_0 = 5 - 5/\sqrt{3}, \quad x_1 = 5 + 5/\sqrt{3}, \quad w_0 = w_1 = 5$$

and on the interval [10, 20]

$$x_2 = 15 - 5/\sqrt{3}, \quad x_3 = 15 + 5/\sqrt{3}, \quad w_2 = w_3 = 5.$$

Thus, the composite quadrature rule is

$$I(f) \approx 5 \Big( f(5-5/\sqrt{3}) + f(5+5/\sqrt{3}) + f(15+5/\sqrt{3}) + f(15+5/\sqrt{3}) \Big).$$

### Problem 7 Runge–Kutta method (10 points)

Consider the ODE

$$\begin{cases} x'(t) = f(x(t), t) \\ x(0) = x_0 \end{cases}$$

where f is a given smooth function, and the Runge–Kutta method

$$k = f(y_n + hk/2, t_n + h/2)$$
$$y_{n+1} = y_n + hk.$$

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Set  $f(x,t) = \lambda x$  for some  $\lambda \in \mathbb{C}$  with  $\operatorname{Re}(\lambda) < 0$ . Find the stability function of this method, and determine whether the method is unconditionally stable or not.

Hint: If you are unable to determine stability, it's enough to insert  $h\lambda = 1, 10, 100$  in the stability function and conclude based on that.

**Solution:** We insert  $f(x,t) = \lambda x$  and get  $k = \lambda(y_n + hk/2) \quad \Leftrightarrow \quad k = \frac{h\lambda}{1 - h\lambda/2}y_n$  $y_{n+1} = y_n + hk = y_n \left(1 + \frac{h\lambda}{1 - h\lambda/2}\right) = y_n R(h\lambda)$ 

where  $R(z) = 1 + \frac{z}{1-z/2} = \frac{1+z/2}{1-z/2} = \frac{2+z}{2-z}$ . If  $\operatorname{Re}(z) < 0$  then |2+z| < |2-z|, and therefore |R(z)| < 1 for all such z. We conclude that the method is unconditionally stable.