

# UNIVERSITY OF OSLO

Faculty of mathematics and natural sciences

Exam in: MAT3110/MAT4110 — Introduction to numerical analysis

Day of examination: 26 November 2020

Examination hours: 15:00–19:00

This problem set consists of 8 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 14 subproblems (1a, 1b, 2a, etc.), and you can get between 5 and 10 points for each sub-problem.
- All answers must be justified.

## Problem 1 QR factorization

Let  $\delta > 0$  and let

$$A = \begin{pmatrix} 3 & 3 & 0 \\ 0 & \delta & 1 \\ 4 & 4 & 0 \end{pmatrix}.$$

**1a**

Compute the QR factorization of  $A$  using the Gram–Schmidt algorithm.

**1b**

If  $\delta$  is very small, what can go wrong if we run this algorithm on a computer? Be as specific as you can.

**Solution:**

**1a**

We run through the algorithm:

$$n = 1: w \leftarrow \begin{pmatrix} 3 \\ 0 \\ 4 \end{pmatrix}, r_1 \leftarrow \begin{pmatrix} 5 \\ 0 \\ 0 \end{pmatrix}, q_1 \leftarrow \begin{pmatrix} 3/5 \\ 0 \\ 4/5 \end{pmatrix}, k \leftarrow 1.$$

(Continued on page 2.)

$$n = 2: w \leftarrow \begin{pmatrix} 0 \\ \delta \\ 0 \end{pmatrix}, r_2 \leftarrow \begin{pmatrix} 5 \\ \delta \\ 0 \end{pmatrix}, q_2 \leftarrow \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, k \leftarrow 2.$$

$$n = 3: w \leftarrow \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, r_3 \leftarrow \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}.$$

Since  $k < n$  at the end of the algorithm, we add a unit vector  $q_3$  which is orthogonal to  $q_1, q_2$ , for instance  $q_3 = \begin{pmatrix} -4/5 \\ 0 \\ 3/5 \end{pmatrix}$ . (Only  $\pm q_3$  are correct answers.) We get

$$A = QR, \quad Q = \begin{pmatrix} 3/5 & 0 & -4/5 \\ 0 & 1 & 0 \\ 4/5 & 0 & 3/5 \end{pmatrix}, \quad R = \begin{pmatrix} 5 & 5 & 0 \\ 0 & \delta & 1 \\ 0 & 0 & 0 \end{pmatrix}.$$

### 1b

If  $\delta \approx 0$  then we might run into roundoff errors when we compute  $w$  in the second step:

$$w = \begin{pmatrix} 3 \\ \delta \\ 4 \end{pmatrix} - \left(\frac{9}{5} + \frac{16}{5}\right) \begin{pmatrix} 3/5 \\ 0 \\ 4/5 \end{pmatrix} = \begin{pmatrix} \tilde{0} \\ \delta \\ \hat{0} \end{pmatrix}$$

where  $\tilde{0}, \hat{0} \approx 0$ . If the round off error is of the same order as  $\delta$ , then the relative error is large, and  $q_2 = w/\|w\|$  will be incorrect; in particular,  $q_2$  will not be orthogonal to  $q_1$ . As a consequence, the computation of  $w$  in the third step will be incorrect, leading to incorrect  $r_3$  and  $q_3$ .

## Problem 2

Let  $A \in \mathbb{R}^{n \times n}$  be a given matrix and define

$$f(x) = \frac{\|Ax\|}{\|x\|} \quad \text{for } x \in \mathbb{R}^n, x \neq 0$$

(where  $\|\cdot\|$  is the Euclidean norm,  $\|x\| = \sqrt{\sum_{i=1}^n (x_i)^2}$ .)

### 2a

In what way does  $f(x)$  tell us how sensitive  $A$  is to  $x$ ?

### 2b

Assume that  $n = 2$  and that  $A$  can be decomposed as

$$A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 8 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 3/5 & 4/5 \\ -4/5 & 3/5 \end{pmatrix}.$$

What type of decomposition is this?

(Continued on page 3.)

**2c**

Let  $A$  be as in problem **2b**. Find nonzero vectors  $y, z \in \mathbb{R}^2$  such that

$$f(y) = \max_{\substack{x \in \mathbb{R}^2 \\ x \neq 0}} f(x), \quad f(z) = \min_{\substack{x \in \mathbb{R}^2 \\ x \neq 0}} f(x).$$

**Solution:****2a**

Let  $e \in \mathbb{R}^n$ . Then  $f(x)$  is the relative error when we evaluate  $A(e+x)$  instead of  $Ae$ :

$$\frac{\|A(e+x) - Ae\|}{\|(e+x) - e\|} = \frac{\|Ax\|}{\|x\|} = f(x).$$

Put another way: It tells us how sensitive the map  $e \mapsto Ae$  is to changes in the input.

**2b**

This is an SVD decomposition,  $A = USV^T$ , with

$$U = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}, \quad S = \begin{pmatrix} 8 & 0 \\ 0 & 1 \end{pmatrix}, \quad V = \begin{pmatrix} 3/5 & -4/5 \\ 4/5 & 3/5 \end{pmatrix}.$$

Indeed,  $S$  is clearly a diagonal matrix with nonnegative entries, and both  $U$  and  $V$  are orthogonal matrices:  $UU^T = VV^T = I$ .

**2c**

Write  $U = (u_1 \ u_2)$ ,  $V = (v_1 \ v_2)$ ,  $S = \text{diag}(\sigma_1, \sigma_2)$ . Then  $Ax = u_1\sigma_1\langle v_1, x \rangle + u_2\sigma_2\langle v_2, x \rangle$ . Since  $u_1, u_2$  are orthonormal, the norm of  $Ax$  is

$$\|Ax\| = \sqrt{\sigma_1^2\langle v_1, x \rangle^2 + \sigma_2^2\langle v_2, x \rangle^2}.$$

If, say,  $\|y\| = 1$  and  $f(y)$  is as large as possible, then necessarily we need  $\langle v_1, y \rangle = 1$  and  $\langle v_2, y \rangle = 0$  — that is,  $y = v_1$ . Likewise, we need  $z = v_2$ .

Thus:

$$y = \begin{pmatrix} 3/5 \\ 4/5 \end{pmatrix}, \quad z = \begin{pmatrix} -4/5 \\ 3/5 \end{pmatrix}.$$

**Problem 3 Newton's method**

Consider the system of equations

$$\begin{aligned} x^3 - y^3 &= 3 \\ x^2 + y^2 &= 4. \end{aligned} \tag{1}$$

**3a**

Write down Newton's method for solving (1), and perform one iteration when starting at  $(x_0, y_0) = (1, 1)$ .

(Continued on page 4.)

**3b**

What can go wrong if we start at, or close to, the  $x$ - or  $y$ -axes?

**Solution:****3a**

Let

$$f(x, y) = \left( \frac{x^3 - y^3 - 3}{x^2 + y^2 - 4} \right).$$

Newton's method is

$$\begin{pmatrix} x_{n+1} \\ y_{n+1} \end{pmatrix} = \begin{pmatrix} x_n \\ y_n \end{pmatrix} - \nabla f(x_n, y_n)^{-1} f(x_n, y_n).$$

We compute

$$\nabla f(x, y) = \begin{pmatrix} 3x^2 & -3y^2 \\ 2x & 2y \end{pmatrix} \Rightarrow \nabla f(x, y)^{-1} = \frac{1}{6x^2y + 6xy^2} \begin{pmatrix} 2y & 3y^2 \\ -2x & 3x^2 \end{pmatrix}.$$

This gives

$$\begin{aligned} \begin{pmatrix} x_1 \\ y_1 \end{pmatrix} &= \begin{pmatrix} 1 \\ 1 \end{pmatrix} - \frac{1}{12} \begin{pmatrix} 2 & 3 \\ -2 & 3 \end{pmatrix} \begin{pmatrix} -3 \\ -2 \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ 1 \end{pmatrix} - \frac{1}{12} \begin{pmatrix} -12 \\ 0 \end{pmatrix} \\ &= \begin{pmatrix} 2 \\ 1 \end{pmatrix}. \end{aligned}$$

**3b**

The Jacobian  $\nabla f(x, y)$  is not invertible when  $x = 0$  or  $y = 0$  (the first or second column becomes zero, respectively). As a consequence the Newton method breaks down.

This happens because near, say, the  $x$ -axis (i.e., when  $y \approx 0$ ), the function  $f$  changes little or nothing at all when moving in the  $y$ -direction. As a consequence, Newton's method dictates moving very far (or infinitely far) in the  $y$ -direction.

**Problem 4 Polynomial interpolation**

Let  $f : \mathbb{R} \rightarrow \mathbb{R}$  be a function satisfying

$$f(0) = 2, \quad f(1) = 1, \quad f(3) = 5.$$

**4a**

Find the polynomial  $p$  of the lowest possible order which interpolates  $f$  through these points.

(Continued on page 5.)

**4b**

Assume that  $f \in C^3([0, 4])$  satisfies

$$\|f^{(3)}\|_\infty \leq M, \quad \text{where } \|f^{(3)}\|_\infty = \sup_{x \in [0, 4]} |f^{(3)}(x)|$$

for some  $M > 0$ . Estimate the error  $\|f - p\|_\infty$  in terms of  $M$ .

*Hint: You don't need the solution from problem 4a in order to solve problem 4b.*

**Solution:****4a**

We have 3 interpolation points, hence there exists a unique interpolating polynomial  $p \in \mathbb{P}_n$  when  $n = 2$ . We use the Lagrange form for  $p$ :

$$\begin{aligned} p(x) &= \sum_{i=0}^2 f(x_i) L_i(x) = 2 \frac{x-1}{0-1} \frac{x-3}{0-3} + 1 \frac{x-0}{1-0} \frac{x-3}{1-3} + 5 \frac{x-0}{3-0} \frac{x-1}{3-1} \\ &= \frac{2}{3}(x-1)(x-3) - \frac{1}{2}x(x-3) + \frac{5}{6}x(x-1) \\ &= x^2 - 2x + 2. \end{aligned}$$

**4b**

The standard error estimate is

$$|f(x) - p(x)| = \frac{|f^{(n+1)}(\xi)|}{(n+1)!} \prod_{i=0}^n |x - x_i| \leq \frac{M}{6} |x(x-1)(x-3)|$$

where  $n = 2$ , for some  $\xi \in (0, 4)$ . The last term can be estimated in various ways, for instance by upper bounding each of the terms  $x$ ,  $x-1$  and  $x-3$  by 4, 3 and 3, resp. (since  $x \in [0, 4]$ ). This gives:

$$\|f - p\|_\infty \leq 6M.$$

(The function  $x \mapsto x(x-1)(x-3)$  attains its largest absolute value at  $x = 4$ , with the value 12; thus, the sharpest upper bound is  $2M$ .)

**Problem 5 Numerical quadrature**

Consider the quadrature rule  $I(f) \approx J(f)$ , where

$$I(f) = \int_0^4 f(x) dx \quad \text{and} \quad J(f) = f(0)w_0 + f(x_1)w_1 + f(4)w_2$$

where  $w_0, w_1, w_2 \in \mathbb{R}$  and  $x_1 \in (0, 4)$ .

**5a**

Let  $x_1 \in (0, 4)$  be fixed. Let  $m$  denote the largest integer such that the quadrature rule is exact for all  $f \in \mathbb{P}_m$ . Show that  $w_0, w_1, w_2$  can be chosen such that  $m \geq 2$ .

(Continued on page 6.)

*Hint: You need to find expressions for  $w_0, w_1, w_2$ , and to show that  $I(f) = J(f)$  for all  $f \in \mathbb{P}_2$ .*

**5b**

Is there a choice of  $x_1 \in (0, 4)$  which makes the order  $m$  of the quadrature rule greater than 2? Explain why, or why not.

**Solution:****5a**

Let  $f \in \mathbb{P}_2$ . Then  $f = f(0)L_0 + f(x_1)L_1 + f(4)L_2$ , where  $L_i$  is the  $i$ -th Lagrange function. Hence,

$$\begin{aligned} \int_0^4 f(x) dx &= f(0) \underbrace{\int_0^4 L_0(x) dx}_{=w_0} + f(x_1) \underbrace{\int_0^4 L_1(x) dx}_{=w_1} + f(4) \underbrace{\int_0^4 L_2(x) dx}_{=w_2} \\ &= w_0 f(0) + w_1 f(x_1) + w_2 f(4). \end{aligned}$$

From the definition it is clear that the method is exact for all  $f \in \mathbb{P}_2$ .

**5b**

Yes, the midpoint  $x_1 = 2$  would yield Simpson's rule,

$$J(f) = \frac{4}{6}f(0) + \frac{16}{6}f(2) + \frac{4}{6}f(4).$$

We only need to check that  $J$  integrates  $f(x) := x^3$  exactly:

$$\begin{aligned} I(f) &= 64, \\ J(f) &= \frac{16}{6}2^3 + \frac{4}{6}4^3 = \frac{16 \cdot 8 + 4 \cdot 4^3}{6} = 64. \end{aligned}$$

An alternative solution is to recall the formula  $f(x) - p(x) = \frac{f^{(3)}(\xi)}{6}(x-0)(x-x_1)(x-4)$ , for some  $\xi \in (0, 4)$ , where  $p \in \mathbb{P}_2$  interpolates  $f$  through 0,  $x_1$ , 4. If  $f \in \mathbb{P}_3$  then  $f^{(3)} \equiv c$  for some  $c \in \mathbb{R}$ , so

$$I(f) - J(f) = \int_0^4 f(x) - p(x) dx = \frac{c}{6} \int_0^4 (x-0)(x-x_1)(x-4) dx.$$

We see that the choice  $x_1 = 2$  would make the integral vanish, and hence  $I(f) = J(f)$ .

**Problem 6 Numerical methods for ODEs**

Consider the ordinary differential equation (ODE)

$$\begin{cases} \dot{x}(t) = f(x(t), t) & \text{for } t > 0 \\ x(0) = x_0 \end{cases} \quad (2)$$

(Continued on page 7.)

We partition the time domain  $t \in [0, T]$  into subintervals with endpoints  $t_n = hn$ , where  $h = \frac{T}{N}$ , for some  $N \in \mathbb{N}$ . The solution is approximated as  $y_n \approx x(t_n)$ .

**6a**

Consider the Runge–Kutta method with the Butcher tableau

$$\begin{array}{c|cc} \alpha & 0 & 0 \\ 2/3 & 1/3 & \beta \\ \hline & 1/4 & \gamma \end{array}$$

for numbers  $\alpha, \beta, \gamma \in \mathbb{R}$ . Determine  $\alpha, \beta, \gamma$  such that the method is consistent. Is the method explicit or implicit?

**6b**

For the Runge–Kutta method from problem **6a**, write down a formula for  $y_{n+1}$  in terms of  $y_n$ .

**6c**

We consider the linear problem (2) with  $f(x, t) = \lambda x$  for some  $\lambda \in \mathbb{C}$  with  $\operatorname{Re} \lambda < 0$ . We solve the problem using the numerical method

$$\begin{cases} y_{n+1} = y_n + hf(y_n + \frac{h}{2}f(y_n, t_n), t_n + \frac{h}{2}) & \text{for } n = 0, 1, \dots, \\ y_0 = x_0. \end{cases} \quad (3)$$

Find the stability function for (3). Is the method unconditionally stable?

**Solution:****6a**

We need  $\alpha = 0 + 0 = 0$ ,  $2/3 = 1/3 + \beta$ , and  $1/4 + \gamma = 1$ , so

$$\alpha = 0, \quad \beta = 1/3, \quad \gamma = 3/4.$$

The method is implicit since the Runge–Kutta matrix has nonzero entries along the diagonal.

**6b**

We get

$$\begin{aligned} k_1 &= f(y_n, t_n) \\ k_2 &= f(y_n + h(k_1 + k_2)/3, t_n + 2/3) \\ y_{n+1} &= y_n + h(k_1 + 3k_2)/4. \end{aligned}$$

(Continued on page 8.)

**6c**

If we insert  $f(x, \lambda) = \lambda x$  then we get

$$y_{n+1} = y_n + h\lambda\left(y_n + \frac{h}{2}\lambda y_n\right) = y_n\left(1 + h\lambda + \frac{1}{2}(h\lambda)^2\right).$$

Hence,  $y_{n+1} = R(h\lambda)y_n$  with

$$R(z) = 1 + z + \frac{z^2}{2}.$$

If, say,  $\lambda \in \mathbb{R}$  and  $\lambda \ll 0$  then  $|R(\lambda h)|$  is a large number (larger than 1) unless  $h$  is proportionally small. In other words, the method is *not* unconditionally stable.