# UNIVERSITY OF OSLO

# Faculty of mathematics and natural sciences

Exam in:	MAT3110 — Introduction to numerical analysis
Day of examination:	14 December 2018
Examination hours:	0900-1300
This problem set consists of 2 pages.	
Appendices:	None
Permitted aids:	None

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

All 8 part questions will be weighted equally.

### Problem 1 Matlab function

Write a Matlab (or Python) function [L,U] = mylu(A) that computes the LU factorization of an  $n \times n$  matrix A, assuming that no pivoting is required.

# Problem 2 Cholesky

Use the Cholesky algorithm to determine whether the matrix

$$A = \begin{bmatrix} 2 & 6 & -4 \\ 6 & 17 & 7 \\ -4 & 7 & 10 \end{bmatrix}$$

is positive-definite.

# Problem 3 Matrix norm and SVD

#### 3a

Recall that the 2-norm of a vector  $\mathbf{x} = (x_1, \ldots, x_n)$  in  $\mathbb{R}^n$  is

$$\|\mathbf{x}\|_{2} = \left(\sum_{i=1}^{n} x_{i}^{2}\right)^{1/2}$$

If U is an orthogonal  $n \times n$  matrix, what is the 2-norm of  $U\mathbf{x}$ ?

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#### 3b

What is the definition of the 2-norm of a real  $n \times n$  matrix A?

#### **3**c

Using the singular value decomposition of A, show that the 2-norm of A equals  $\sigma_1$ , the largest singular value of A.

### Problem 4 Iterative scheme

Consider the linear system  $A\mathbf{x} = \mathbf{b}$  where A is a non-singular  $n \times n$  matrix and **b** is a vector in  $\mathbb{R}^n$ . Suppose we try to find the solution **x** using the iterative scheme

$$(A - B)\mathbf{x}^{(k+1)} = -B\mathbf{x}^{(k)} + \mathbf{b}, \qquad k = 0, 1, 2,$$

with  $\mathbf{x}^{(0)}$  some initial guess. Assuming A - B is non-singular, what condition on A, B, and  $\mathbf{b}$  is both necessary and sufficient for the sequence  $\mathbf{x}^{(0)}, \mathbf{x}^{(1)}, \mathbf{x}^{(2)}, \ldots$  to converge to  $\mathbf{x}$ ? Explain your answer.

### Problem 5 Polynomial interpolation

The polynomial of degree  $\leq n$  that interpolates a function  $f : [-1, 1] \to \mathbb{R}$ at distinct points  $x_0, x_1, \ldots, x_n \in [-1, 1]$  can be expressed as

$$p_n(x) = \sum_{i=0}^n \prod_{\substack{j=0\\j \neq i}}^n \frac{x - x_j}{x_i - x_j} f(x_i).$$

For approximating f, what is a good choice of the points  $x_i$  when n is large?

### Problem 6 Non-linear least squares

Suppose we want to minimize a function  $f : \mathbb{R}^n \to \mathbb{R}$  of the form

$$f(\mathbf{x}) = \frac{1}{2} \sum_{i=1}^{m} r_i(\mathbf{x})^2, \quad \mathbf{x} \in \mathbb{R}^n,$$

where  $r_i : \mathbb{R}^n \to \mathbb{R}$ , i = 1, ..., n, are the so-called residuals. What are the Newton and Gauss-Newton methods for this problem, and what are their advantages and disadvantages?

Good luck!