

DELFT UNIVERSITY OF TECHNOLOGY  
FACULTY OF ELECTRICAL ENGINEERING, MATHEMATICS AND COMPUTER SCIENCE

**The final grade of the test:** (total number of points)/5

**Examiner responsible:** C. Vuik

**Examination reviewer:** V. Dwarka

**TEST SCIENTIFIC COMPUTING ( wi4201 )**  
**Wednesday January 22 2026, 13:30-16:30**

1. Below 5 statements are given. If the statement is true give a short proof. If the statement is wrong give a counter example or an explanation.

(a)  $I \in \mathbb{R}^{n \times n}$  is the identity matrix.  $\Rightarrow K_2(I) = 2$ . (2 pt.)

(b) For  $A \in \mathbb{R}^{n \times n}$  the matrix norm  $\|A\|_{max}$  is defined as  $\|A\|_{max} = \max_{1 \leq i \leq n, 1 \leq j \leq n} |a_{i,j}|$ .  
This norm has the multiplicative property. (2 pt.)

(c)  $A = \begin{pmatrix} 1 & -1 & 0 & 0 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ 0 & 0 & -2 & 2 \end{pmatrix}$  The spectral radius of  $A$  is larger than 6. (2 pt.)

(d)  $A \in \mathbb{R}^{n \times n}$ , and  $A = A^T \Rightarrow \|A\|_2 = \max_{1 \leq i \leq n} \lambda_i$  where  $\lambda_i$  is an eigenvalue of  $A$ .  
(2 pt.)

(e)  $A \in \mathbb{R}^{n \times n}$  is a lower triangular matrix with only zero elements on the main diagonal  $\Rightarrow A^n = 0$ . (2 pt.)

2. Consider the boundary value problem on the unit square

$$-\nabla \cdot (\alpha \nabla u)(x, y) + \sigma u(x, y) = f(x, y) \text{ for } (x, y) \in \Omega = (0, 1) \times (0, 1)$$

with homogeneous Dirichlet boundary conditions

$$u(x, y) = 0 \text{ for } (x, y) \in \partial\Omega.$$

Here,  $\alpha > 0$  and  $\sigma \in \mathbb{R}$  are constants. For the discretization we use a uniform grid with  $n$  intervals in each coordinate direction and elimination of the boundary conditions.

(a) Classify the PDE (elliptic, parabolic, hyperbolic). (1 pt.)

(b) Give a discretization such that the truncation error is  $O(h^2)$ . (2 pt.)

- (c) Take  $n = 3$ . Using a lexicographic ordering, give the full matrix  $A$  and vector  $f$  such that  $Au = f$ . Motivate why  $A$  is symmetric. (2 pt.)

We now continue by considering a general  $n$  and  $h$ .

- (d) Assume  $\alpha = 1$ . Take  $\sigma = 1$  and use Gershgorin's theorem to formulate a bound (as an interval in terms of  $h$ ) that contains all the eigenvalues of  $A$ . (2 pt.)
- (e) Now repeat the previous question for  $\sigma = -1$ . (1 pt.)
- (f) Using the bounds from the previous parts, characterize all  $\sigma$  for which you can guarantee  $A$  is nonsingular, including any nontrivial cases. (2 pt.)
3. Let  $A \in \mathbb{R}^{n \times n}$  be a symmetric positive definite matrix and consider solving

$$Ax = f$$

- (a) Consider a perturbation of the right hand side  $f \mapsto f + \Delta f$  with  $\|\Delta f\| \leq \delta \|f\|$ . Prove a bound of the form

$$\frac{\|\Delta x\|}{\|x\|} \leq \delta \kappa(A),$$

for any submultiplicative matrix norm. (2 pt.)

- (b) Prove that if  $A$  is symmetric positive definite, then

$$\kappa_2(A) = \|A\|_2 \|A^{-1}\|_2 = \frac{\lambda_{\max}(A)}{\lambda_{\min}(A)}.$$

Use the fact that we can diagonalize  $A$  with an orthogonal basis of eigenvectors. (3 pt.)

- (c) Assume the computed solution  $\hat{x}$  returned by a Cholesky based solver satisfies the backward error relation

$$(A + E)\hat{x} = f, \quad \|E\|_2 \leq \varepsilon \|A\|_2,$$

for some  $\varepsilon > 0$ . Using only norm inequalities, show that if

$$\varepsilon \kappa_2(A) < 1,$$

then

$$\frac{\|\hat{x} - x\|_2}{\|x\|_2} \leq \frac{\varepsilon \kappa_2(A)}{1 - \varepsilon \kappa_2(A)}.$$

You may use that if  $\|B\|_2 < 1$ , then  $(I + B)$  is invertible and

$$\|(I + B)^{-1}\|_2 \leq \frac{1}{1 - \|B\|_2}.$$

(3 pt.)

- (d) Now suppose  $A$  is the matrix obtained by the standard five point finite difference discretization of  $-\Delta u = f$  on  $\Omega = (0, 1) \times (0, 1)$  with Dirichlet boundary conditions, using a uniform grid with mesh size  $h$ . Assume you know the eigenvalue bounds

$$c_1 \leq \lambda_{\min}(A) \leq \lambda_{\max}(A) \leq \frac{c_2}{h^2},$$

with constants  $c_1, c_2 > 0$  independent of  $h$ . Prove that

$$\kappa_2(A) = O(h^{-2}),$$

and combine this with (a) to explain why the upper bound deteriorates as  $h \rightarrow 0$  if machine precision is fixed. (2 pt.)

4. (a) Given the linear system  $A\mathbf{u} = \mathbf{f}$  with an  $n$ -by- $n$  real-valued coefficient matrix  $A$ . Assume a splitting of this coefficient matrix of the form  $A = M - N$  where  $M$  is non-singular and assume that a basic iterative solution method for the linear system is derived from this splitting. Derive a recursion formula for the iterates  $\mathbf{u}^k$ . Derive a recursion formula for the residual vector  $\mathbf{r}^k$ . (2 pt.)

- (b) Give the relation between the error  $\mathbf{e}^k$  and the residual vector  $\mathbf{r}^k$ . Use this relation to derive the defect-correction scheme that use the approximation  $\hat{A}$  to  $A$ ; (2 pt.)

- (c) Consider a 2D Poisson equation with Dirichlet boundary conditions on a square domain, discretized by a 5-point stencil. Give the stencil notation of the Jacobi iteration matrix for an interior grid point. (2 pt.)

- (d) Derive the damped Jacobi method and give the damped Jacobi iteration matrix. (2 pt.)

- (e) Do 1 iteration with the backward Gauss-Seidel method to the following linear system, where we start with the zero vector.

$$\begin{pmatrix} 1 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \\ u_3 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

(2 pt.)

5. In this exercise we have to solve a linear system  $A\mathbf{u} = \mathbf{f}$ , where  $A$  is an  $n \times n$  non-singular matrix.

- (a) Take  $\mathbf{u}_1 = \alpha\mathbf{f}$ . Derive an expression for  $\alpha$  such that  $\|\mathbf{u} - \mathbf{u}_1\|_{A^T A}$  is minimal. (2 pt.)

- (b) The CGNR method is the CG method applied to  $A^T A \mathbf{u} = A^T \mathbf{f}$ . Show that the 2-norm of the residuals is monotone decreasing. (2 pt.)
- (c) Give a  $3 \times 3$  non-diagonal matrix such that CGNR converges in one iteration for every right-hand side vector  $\mathbf{f}$ . (2 pt.)
- (d) Given the algorithm

**Bi-CGSTAB method**

```

 $\mathbf{u}^0$  is an initial guess;  $\mathbf{r}^0 = \mathbf{f} - A\mathbf{u}^0$ ;
 $\bar{\mathbf{r}}^0$  is an arbitrary vector, such that  $(\bar{\mathbf{r}}^0)^T \mathbf{r}^0 \neq 0$ , e.g.,  $\bar{\mathbf{r}}^0 = \mathbf{r}^0$ ;
 $\rho_{-1} = \alpha_{-1} = \omega_{-1} = 1$ ;
 $\mathbf{v}^{-1} = \mathbf{p}^{-1} = \mathbf{0}$ ;
for  $i = 0, 1, 2, \dots$  do
     $\rho_i = (\bar{\mathbf{r}}^0)^T \mathbf{r}^i$ ;  $\beta_{i-1} = (\rho_i / \rho_{i-1})(\alpha_{i-1} / \omega_{i-1})$ ;
     $\mathbf{p}^i = \mathbf{r}^i + \beta_{i-1}(\mathbf{p}^{i-1} - \omega_{i-1}\mathbf{v}^{i-1})$ ;
     $\hat{\mathbf{p}} = M^{-1}\mathbf{p}^i$ ;
     $\mathbf{v}^i = A\hat{\mathbf{p}}$ ;
     $\alpha_i = \rho_i / (\bar{\mathbf{r}}^0)^T \mathbf{v}^i$ ;
     $\mathbf{s} = \mathbf{r}^i - \alpha_i \mathbf{v}^i$ ;
    if  $\|\mathbf{s}\|$  small enough then
         $\mathbf{u}^{i+1} = \mathbf{u}^i + \alpha_i \hat{\mathbf{p}}$ ; quit;
     $\mathbf{z} = M^{-1}\mathbf{s}$ ;
     $\mathbf{t} = A\mathbf{z}$ ;
     $\omega_i = \mathbf{t}^T \mathbf{s} / \mathbf{t}^T \mathbf{t}$ ;
     $\mathbf{u}^{i+1} = \mathbf{u}^i + \alpha_i \hat{\mathbf{p}} + \omega_i \mathbf{z}$ ;
    if  $\mathbf{u}^{i+1}$  is accurate enough then quit;
     $\mathbf{r}^{i+1} = \mathbf{s} - \omega_i \mathbf{t}$ ;
end for

```

The matrix  $M$  in this scheme represents the preconditioning matrix based on ILU(0). Determine the minimal amount of memory and flops per iteration, where we assume that  $A$  is based on a 5-point stencil. (2 pt.)

- (e) Give a comparison of the mathematical properties of the CGNR and Bi-CGSTAB method (both without preconditioning). (2 pt.)