

Numerical Linear Algebra - Sheet 8

to be handed in until December 13, 2023, 11am.

Problem 1. Let $\mathbf{Ax} = \mathbf{b}$ be a linear system with a symmetric, positive definite matrix $\mathbf{A} \in \mathbb{R}^{n \times n}$ that has extremal eigenvalues λ_{\min} and λ_{\max} and spectral condition number $\text{cond}(\mathbf{A})$. Consider the sequence $\{\mathbf{x}^{(k)}\}$ of one-dimensional projection processes with $K = L = \text{span}\{\mathbf{e}_i\}$, where \mathbf{e}_i denotes the i -th unit vector in \mathbb{R}^n . The index i is selected at each step k to be the index of a component of largest absolute value in the current residual vector $\mathbf{r}^{(k)} = \mathbf{b} - \mathbf{Ax}^{(k)}$.

(a) For $\mathbf{d}^{(k)} = \mathbf{A}^{-1}\mathbf{b} - \mathbf{x}^{(k)}$ show that

$$\|\mathbf{d}^{(k+1)}\|_{\mathbf{A}} \leq \left(1 - \frac{1}{n \text{cond}(\mathbf{A})}\right)^{\frac{1}{2}} \|\mathbf{d}^{(k)}\|_{\mathbf{A}}.$$

Hint: You can use the expression

$$\langle \mathbf{Ad}^{(k+1)}, \mathbf{d}^{(k+1)} \rangle = \langle \mathbf{Ad}^{(k)}, \mathbf{d}^{(k)} \rangle - \frac{\langle \mathbf{r}^{(k)}, \mathbf{e}_i \rangle^2}{a_{ii}},$$

as well as the inequality $|\mathbf{e}_i^T \mathbf{r}^{(k)}| \geq n^{-1/2} \|\mathbf{r}^{(k)}\|_2$. a_{ii} denotes the i -th diagonal element of \mathbf{A} .

(b) Does (a) prove that the algorithm converges?

Problem 2. Consider the linear system $\mathbf{Ax} = \mathbf{b}$, where \mathbf{A} is a symmetric positive definite matrix. We define a projection method which uses a two-dimensional space at each step. At a given step, take $L = K = \text{span}\{\mathbf{r}, \mathbf{Ar}\}$, where $\mathbf{r} = \mathbf{b} - \mathbf{Ax}$ is the current residual.

- For a basis of K use the vector \mathbf{r} and the vector \mathbf{p} obtained by orthogonalizing \mathbf{Ar} against \mathbf{r} with respect to the \mathbf{A} -inner product. Give the formula for computing \mathbf{p} (no need to normalize the resulting vector).
- Write the algorithm for performing the projection method described above.
- Can you explain, not compute, why the algorithm converges for any initial guess \mathbf{x}_0 ? *Hint: Exploit the convergence results for one-dimensional projection techniques.*

Problem 3 (Programming).

- Implement the steepest decent method (Algorithm 3.3.13 in the lecture notes).
- Use your implementation of the steepest decent method to solve the 2D Laplace problem

$$\mathbf{L}_2 \mathbf{x} = \mathbf{b}$$

as in Problem 2 on Sheet 7 with right hand side vector $\mathbf{b} = (1, \dots, 1)^T$ and initial guess $\mathbf{x}^{(0)} = (0, \dots, 0)^T$ with $n = 50$ and $n = 100$.

- (c) Compute and discuss the observed convergence rate (see 3.2.21/Definition 3.2.22 in the Lecture Notes).

Problem 4 (Programming).

- (a) Implement the minimal residual method (Algorithm 3.3.20 in the lecture notes).
- (b) Use your implementation of the steepest decent method to solve the 2D Laplace problem

$$\mathbf{L}_2 \mathbf{x} = \mathbf{b}$$

as in Problem 2 on Sheet 7 with right hand side vector $\mathbf{b} = (1, \dots, 1)^T$ and initial guess $\mathbf{x}^{(0)} = (0, \dots, 0)^T$ with $n = 50$ and $n = 100$.

- (c) Compute and discuss the observed convergence rate (see 3.2.21/Definition 3.2.22 in the Lecture Notes). Compare it with the convergence rate observed in Problem 3.