

## Sheet 5

Discussion of the sheet: Th., 02.05.2019

1. Let  $A \in \mathbb{R}^{n \times n}$  be SPD, set  $\phi(x) := \frac{1}{2}(Ax, x) - (b, x)$ , and let  $x_*$  be the exact solution of  $Ax_* = b$ .

a) Show that

$$\phi(x) - \phi(x_*) = \frac{1}{2}\|x - x_*\|_A^2.$$

- b) Fix  $x_0 \in \mathbb{R}^n$  and let  $(x_0, x_1, \dots, x_k)$  be the sequence of CG approximations. Let  $\hat{x}_k \in \mathbb{R}^n$  be the result of a steepest-descent iteration step with search direction  $0 \neq d = r_{k-1} = b - Ax_{k-1}$ , starting from the CG-vector  $x_{k-1}$ . Show

$$\|x_k - x_*\|_A \leq \|\hat{x}_k - x_*\|_A,$$

i.e. the CG-step is at least as good as the SD-step.

Conclude the local contraction property

$$\|x_k - x_*\|_A \leq q\|x_{k-1} - x_*\|_A \quad \text{with} \quad q := \frac{\kappa_2(A) - 1}{\kappa_2(A) + 1} < 1.$$

2. (Exercise 9.5 from the Lecture notes)

Assume that for  $m = 1, 2, \dots$ , the tridiagonal matrices

$$T_m = \begin{pmatrix} \alpha_1 & \beta_2 & & & \\ \beta_2 & \alpha_2 & \beta_3 & & \\ & \ddots & \ddots & \ddots & \\ & & \beta_{m-1} & \alpha_{m-1} & \beta_m \\ & & & \beta_m & \alpha_m \end{pmatrix} \in \mathbb{R}^{m \times m}$$

admit LU-decompositions  $T_m = L_m U_m$ . Of course,  $L_m$  and  $U_m$  are bidiagonal, and we adopt the notation

$$T_m = L_m U_m = \begin{pmatrix} 1 & & & & \\ \lambda_2 & 1 & & & \\ & \ddots & \ddots & & \\ & & \lambda_{m-1} & 1 & \\ & & & \lambda_m & 1 \end{pmatrix} \begin{pmatrix} \eta_1 & \omega_2 & & & \\ & \eta_2 & \omega_3 & & \\ & & \ddots & \ddots & \\ & & & \eta_{m-1} & \omega_m \\ & & & & \eta_m \end{pmatrix} \quad (1)$$

- a) Verify the following recursive formulas for the values  $\lambda_m, \omega_m, \eta_m$  in (1):

$$\omega_m = \beta_m, \quad \lambda_m = \frac{\beta_m}{\eta_{m-1}}, \quad \eta_m = \begin{cases} \alpha_1 & m = 1, \\ \alpha_m - \lambda_m \omega_m, & m > 1. \end{cases} \quad (2)$$

Conclude that the matrices  $L_m$  and  $U_m$  are recursively obtained from  $L_{m-1}, U_{m-1}$  by adding one row and column, i.e.,

$$L_m = \left( \begin{array}{c|c} L_{m-1} & 0 \\ \hline 0^T & \lambda_m \end{array} \right), \quad U_m = \left( \begin{array}{c|c} U_{m-1} & 0 \\ \hline 0^T & \eta_m \end{array} \right) \quad (3)$$

- b) Given the factors  $L, U \in \mathbb{R}^{m \times m}$  of the LU-decomposition of a matrix  $T \in \mathbb{R}^{m \times m}$  in the form (3),

$$L = \left( \begin{array}{c|c} L' & 0 \\ \hline \ell^T & 1 \end{array} \right), \quad U = \left( \begin{array}{c|c} U' & u \\ \hline 0^T & \eta \end{array} \right)$$

with  $L', U' \in \mathbb{R}^{(m-1) \times (m-1)}$  and  $\ell, u \in \mathbb{R}^{m-1}$ , conclude that  $L^{-1}$  and  $U^{-1}$  can be written as

$$L^{-1} = \left( \begin{array}{c|c} L'^{-1} & 0 \\ \hline -\ell^T L'^{-1} & 1 \end{array} \right), \quad U^{-1} = \left( \begin{array}{c|c} U'^{-1} & -\frac{1}{\eta} U'^{-1} u \\ \hline 0^T & \frac{1}{\eta} \end{array} \right)$$

This means that after adding one row and column to the tridiagonal matrix  $T$  to obtain the new tridiagonal matrix  $T'$ , the triangular inverses  $L'^{-1}$ ,  $U'^{-1}$  can be also obtained from  $L^{-1}$  and  $U^{-1}$  by simply adding one row and column.

- 3.** We consider the *minimal residual iteration* for solving  $Ax = b$ . In this method, in order to determine the iteration  $x_{k+1}$  in each step, the residual  $\Phi(x) := \|b - Ax\|_2^2$  is minimized over the one dimensional affine space  $x_k + \text{span}\{r_k\}$ .

- a) Formulate the algorithm for computing the iterations  $x_1, x_2, \dots$
- b) Assume, that  $A$  is positive definite (but not necessarily symmetric). Define

$$\mu := \frac{1}{2} \lambda_{\min}(A + A^T), \quad \sigma := \|A\|_2.$$

Show that  $\mu \leq \sigma$ .

- c) Show that under the assumptions of b), the *minimal residual iteration* converges and the following estimate holds

$$\|r_{k+1}\|_2^2 \leq \left(1 - \frac{\mu^2}{\sigma^2}\right) \|r_k\|_2^2.$$