Sheet 8

Discussion of the sheet: Th., 06.06.2019

- **1.** (Exercise 12.1) Let $A, M \in \mathbb{R}^{n \times n}$ be SPD. Define the M-inner product $(\cdot, \cdot)_M$ by $(x, y)_M = (Mx, y) = x^{\mathrm{T}} M y$. Show:
 - a) $M^{-1}A$ is selfadjoint with respect to the inner product $(\cdot,\cdot)_M$, i.e., $(M^{-1}Ax,y)_M = (x,M^{-1}Ay)_M$ for all $x,y\in\mathbb{R}^n$.
 - **b)** $M^{-1}A$ is positive definite with respect to $(\cdot, \cdot)_M$, i.e., $(M^{-1}Ax, x)_M > 0$ for all $0 \neq x \in \mathbb{R}^n$.
 - c) The matrix AM^{-1} is symmetric positive definite with respect to the M^{-1} -inner product $(\cdot,\cdot)_{M^{-1}}$.
- **2.** (Exercise 12.3) Assume that the Cholesky decomposition $M = L L^T$ of an SPD preconditioner M is available. Consider the split-preconditioned system with $M_L = L$ and $M_R = L^T$, i.e., applying the CG algorithm to the SPD system $L^{-1}AL^{-T}u = L^{-1}b$.

Show: The iterates $x_m = L^{-T}u_m$ coincide with those of the left-preconditioned CG method, i.e., Alg. 12.2 with preconditioner M.

(Note that the same argument holds for any SPD preconditioner M implicitly specified by a regular matrix C with $M = C C^T$.)

- **3.** (Exercise 12.4)
 - a) Show: The speed of convergence of the PCG iteration can be estimated by

$$\|e_m\|_A \le 2\left(\frac{\sqrt{\kappa}-1}{\sqrt{\kappa}+1}\right)^m \|e_0\|_A, \quad \kappa = \kappa_\sigma(M^{-1}A) = \frac{\lambda_{max}}{\lambda_{min}}$$

where λ_{max} and λ_{min} are the largest and smallest eigenvalues of $M^{-1}A$, and $\kappa_{\sigma}(M^{-1}A) = \lambda_{max}/\lambda_{min}$ is the spectral condition number (in general, this is not identical with $\kappa_2(M^{-1}A)$).

b) Show that a characterization of λ_{min} and λ_{max} which may be easier to check is the following: λ_{min} is the largest and λ_{max} is the smallest number such that

$$\lambda_{min} M \leq A \leq \lambda_{max} M$$

is valid.

4. (Exercise 12.5) Let $X \in \mathbb{R}^{n \times n}$ and $\varphi(X) := \|AX - I\|_F^2$. Expand $\varphi(X+H)$ to conclude $\varphi(X+H) = \varphi(X) - 2(A^{\mathrm{T}}R, H)_F + \|AH\|_F^2 \quad \text{with the residual matrix } R = I - AX.$

Note that this expansion is exact because $\varphi(X)$ is a quadratic functional.