## Lecture 5

Exercises

**Computational Mathematics** 

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### Exercises

- Exercise 1: Consider  $K_n = [b \mid Ab \mid ... \mid A^{n-1}b]$  (Krylov matriz) for a given matriz  $A \in \mathbb{R}^{n \times n}$  and a vector  $b \in \mathbb{R}^n$ . In Matlab: Kn = gallery('krylov', A, b, n)
  - 1. Varying n between 2 and 20, choose a random matrix A and b and determine the condition number

$$cond(K_n) = \|K_n\|_2 \|K_n^{-1}\|_2$$

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of matrix  $K_n$  and plot the data in logarithmic scale.

2. Let us suppose that we want to solve the system  $K_n x = c$ , where *c* is determined such that  $x = [1, 1, ..., 1]^T$  is the exact solution of the system. Plot the evolution of the relative error with respect the calculated approximation  $\bar{x}$ ,

$$r(\bar{x}) = \|x - \bar{x}\|_2 / \|x\|_2,$$

as a function of n (make the range from 2 to 20).

3. Prove that, if  $K_n \bar{x} = \bar{c}$ ,

$$r(\bar{x}) \leq \operatorname{cond}(K_n)r(\bar{c})$$

and plot the evolution of this upper bond for the relative error as a function of n (make the range from 2 to 20).

# Exercises

#### • Exercise 2: Consider the SOR method for Ax = b.

- 1. Consider the tridiagonal matrix A with 2 on the diagonal and -1 above and below the diagonal. Construct the right-hand side vector so that  $x = [1, 1, ..., 1]^T$  is the true solution.
- For each value of ω = 1, 1.01, 1.02, ..., 1.99, 2.0, apply 100 iterations of SOR starting with x<sup>(0)</sup> = 0. Do this for A of order 10, 20, and 50. Measure the error at the end of 100 iterations, call it e, and set p = <sup>100</sup>√e. The value of p is the "average" rate of convergence of the iteration; the error was reduced by this much on each iteration. Note that the error of x<sup>(0)</sup> is 1.
- 3. For each of the three cases make a performance plot of p versus  $\omega$  and estimate the optimum value of  $\omega$ . If the plot is too coarse, make additional runs to fill in the gaps.
- 4. Discuss the behavior of the performance profiles and their implications for the difficulty of finding optimum SOR factors.

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## Exercises

- Exercise 3: In the last lectures we considered the Conjugate Gradient (CG) method, the Krylov subspace methods and Richardson iteration method.
  - 1. Explain the relationship between these three methods.
  - How does the CG method fit into the broader framework of Krylov subspace methods? Furthermore, discuss how Richardson iteration can be viewed as a special case of both the CG method and Krylov methods.
  - 3. Provide insights into the similarities and differences between these iterative techniques, and discuss the implications of understanding this relationship for solving linear systems efficiently.
  - 4. Prove that, for SPD matrices, solving Ax = b is equivalent to finding the minimizer  $x \in \mathbb{R}^n$  of the quadratic form

$$\phi(\mathbf{y}) = \frac{1}{2} \mathbf{y}^{\mathsf{T}} \mathbf{A} \mathbf{y} - \mathbf{y}^{\mathsf{T}} \mathbf{b}$$

and explain how this can be used to develop an interactive method for solving a system of linear equations.