

## EXERCISE 4 – Low-rank approximation techniques

**Problem 1.** Prove that the volume of a square matrix is equal to the product of its singular values.

**Problem 2.** For  $A \in \mathbb{R}^{n \times m}$  and  $B \in \mathbb{R}^{m \times n}$ , with  $m > n$ , let  $A_i$  be a  $n \times (m-1)$  submatrix of  $A$  without  $i$ -th column, and  $B_i$  be  $(m-1) \times n$  submatrix of  $B$  without  $i$ -th row. Prove that it holds that

$$\det(AB) = \frac{1}{m-n} \sum_{i=1}^m \det(A_i B_i).$$

**Problem 3.** This exercise is concerned with the lemma from slide 4 of Lecture 4.

1. Show that the inequality of the lemma cannot be improved in general by constructing a matrix  $U = \begin{pmatrix} I_r \\ B \end{pmatrix}$  with  $\max |b_{ij}| \leq 1$  and  $\|B\|_2 = \sqrt{(n-r)r}$ .
2. Develop an extension of the result of the lemma. Prove that for an arbitrary  $n \times r$  matrix  $U$  of rank  $r$ , there is an index set  $I$  with  $\#I = r$  such that

$$\frac{1}{\sigma_{\min}(U(I, :))} \leq \sqrt{r(n-r)+1} \frac{1}{\sigma_{\min}(U)}.$$

**Problem 4.** For the matrix  $Z_n$  defined on Slide 10, prove  $(Z_n^{-1})_{ij} = 2^{j-i-1}$  for  $j > i$ .

**Problem 5.** Prove that any projector  $\Pi \notin \{0, I\}$  satisfies  $\|I - \Pi\|_2 = \|\Pi\|_2$ .

**Problem 6.** *Quite likely, this problem is very hard.* Let  $A \in \mathbb{R}^{n \times n}$  be symmetric positive definite and let  $k \in [1, n]$ . Prove or give a counterexample for the conjecture

$$\begin{aligned} & \max\{|\det(A(I, J))| : I, J \subset \{1, \dots, n\}, \#I = \#J = k\} \\ &= \max\{|\det(A(I, I))| : I \subset \{1, \dots, n\}, \#I = k\}. \end{aligned}$$

In particular, this would imply that there is a symmetric positive definite  $k \times k$  submatrix of maximal volume.

**Problem 7.** MATLAB exercise.

1. Implement the greedy method from Slide 9.
2. Given an  $m \times n$  matrix  $A$  and an integer  $r \leq \min\{m, n\}$ , implement the following method for obtaining a rank- $r$  approximation:
  - Apply  $r$  steps of the greedy method to left singular vectors  $U_r$  of  $A$  to determine  $I$ .
  - Apply  $r$  steps of the greedy method to right singular vectors  $V_r$  of  $A$  to determine  $J$ .
  - Return rank- $r$  approximation  $A(:, J)(A(I, J))^{-1}A(I, :)$ .

Apply your implementation to the two matrices from Slide 34 for  $r = 1, \dots, 30$  (for Hilbert matrix) and  $r = 1, \dots, 100$  (for exponential). Plot the singular values and the obtained approximation error in the spectral norm.

Prof. Dr. D. Kressner  
Dr. L. Periša