



## Exercise Sheet 7.

Due: Monday, 10.11.2025, 12:00.

**Exercise 1** (Computation of the singular value decomposition). Let  $\mathbf{A} \in \mathbb{R}^{n \times n}$ . Consider a sequence of matrices  $\mathbf{A} = \mathbf{A}_0, \mathbf{A}_1, \mathbf{A}_2, \dots$ , where  $\mathbf{A}_{k+1}$  is derived from  $\mathbf{A}_k$  by applying an orthogonal transformation  $\mathbf{P}_k$ ,  $\mathbf{A}_{k+1} = \mathbf{P}_k \mathbf{A}_k$ , and, for an even integer  $k$ , the matrices are of the form

$$\mathbf{A}_k = \begin{bmatrix} a_{k,1} & b_{k,1} & & 0 \\ & a_{k,2} & \ddots & \\ & & \ddots & b_{k,n-1} \\ 0 & & & a_{k,n} \end{bmatrix} \quad \text{and} \quad \mathbf{A}_{k+1} = \begin{bmatrix} a_{k+1,1} & & & 0 \\ b_{k+1,1} & \ddots & & \\ & \ddots & a_{k+1,n-1} & \\ 0 & & b_{k+1,n-1} & a_{k+1,n} \end{bmatrix}.$$

Show that:

- (a) There exists orthogonal matrices  $\mathbf{Q}_1, \mathbf{Q}_2 \in \mathbb{R}^{n \times n}$  such that  $\mathbf{Q}_1 \mathbf{A} \mathbf{Q}_2 = \tilde{\mathbf{A}}$ , where  $\tilde{\mathbf{A}}$  is a bidiagonal matrix.
- (b) The limit  $\lim_{k \rightarrow \infty} b_{k,j} = 0$  holds for all  $j = 1, \dots, n-1$ .
- (c) There exists an integer  $K \in \mathbb{N}$ , such that  $a_{k,1} \geq a_{k,2} \geq \dots \geq a_{k,n}$  holds for all  $k > K$ .

**Exercise 2** (Estimates of singular values). Let  $\mathbf{A} \in \mathbb{R}^{m \times n}$ . Prove the following inequalities, where  $\sigma_{\max}$  and  $\sigma_{\min}$  are the biggest and smallest singular values, respectively.

- (a) For a matrix  $\mathbf{E} \in \mathbb{R}^{m \times n}$ ,

$$\sigma_{\max}(\mathbf{A} + \mathbf{E}) \leq \sigma_{\max}(\mathbf{A}) + \|\mathbf{E}\|_2 \quad \text{and} \quad \sigma_{\min}(\mathbf{A} + \mathbf{E}) \geq \sigma_{\min}(\mathbf{A}) - \|\mathbf{E}\|_2.$$

- (b) For  $\mathbf{z} \in \mathbb{R}^m$  let  $[\mathbf{A} \mid \mathbf{z}] \in \mathbb{R}^{m \times (n+1)}$  be the matrix defined by adding the column  $\mathbf{z}$  to the matrix  $\mathbf{A}$ . Show that

$$\sigma_{\max}([\mathbf{A} \mid \mathbf{z}]) \geq \sigma_{\max}(\mathbf{A}) \quad \text{and} \quad \sigma_{\min}([\mathbf{A} \mid \mathbf{z}]) \leq \sigma_{\min}(\mathbf{A}).$$

**Exercise 3** (Eckart-Young-Mirsky theorem). Let the matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$  be defined by its singular value decomposition  $(\{\sigma_i\}_{i=1}^r, \{\mathbf{u}_i\}_{i=1}^m, \{\mathbf{v}_i\}_{i=1}^n)$ . Consider the following low rank approximations of  $\mathbf{A}$

$$\mathbf{A}_k = \sum_{i=1}^k \sigma_i \mathbf{u}_i \mathbf{v}_i^T,$$

where  $1 \leq k < r$ . Show the following statements:

- (a) It holds  $\|\mathbf{A} - \mathbf{A}_k\|_2 = \sigma_{k+1}$ .
- (b) For every matrix  $\mathbf{B} \in \mathbb{R}^{m \times n}$  with  $\text{rank}(\mathbf{B}) = k$ , there exists a vector  $\mathbf{z} \in \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_{k+1}\}$ ,  $\mathbf{z} \neq \mathbf{0}$ , such that  $\mathbf{B}\mathbf{z} = \mathbf{0}$ .

- (c) Show that, among all matrices  $\mathbf{B} \in \mathbb{R}^{m \times n}$  with rank at most equal to  $k$ ,  $\mathbf{A}_k$  is the best approximation of  $\mathbf{A}$  with respect to the  $\|\cdot\|_2$ -Norm:

$$\min_{\substack{\mathbf{B} \in \mathbb{R}^{m \times n}, \\ \text{rank}(\mathbf{B}) \leq k}} \|\mathbf{A} - \mathbf{B}\|_2 = \|\mathbf{A} - \mathbf{A}_k\|_2.$$

Hint. Estimate  $\|\mathbf{A} - \mathbf{B}\|_2$  using question (b) and compare it with question (a).

**Exercise 4** (Lower semi-continuity of the rank function). A function  $f : \mathbb{R}^{m \times n} \rightarrow \mathbb{R}$  is called *lower semi-continuous* at  $\mathbf{A}_0 \in \mathbb{R}^{m \times n}$  if for every  $\varepsilon > 0$ , there exists a  $\delta > 0$  such that

$$f(\mathbf{A}) \geq f(\mathbf{A}_0) - \varepsilon$$

holds for all matrices  $\mathbf{A} \in \mathbb{R}^{m \times n}$  with  $d(\mathbf{A}, \mathbf{A}_0) \leq \delta$ , where  $d(\cdot, \cdot)$  is a distance function. We call  $f$  *upper semi-continuous* at  $\mathbf{A}_0 \in \mathbb{R}^{m \times n}$  if  $-f$  is lower semi-continuous at  $\mathbf{A}_0$ . In this exercise, use the distance function induced by the matrix 2-norm, i.e.  $d(\mathbf{A}, \mathbf{B}) = \|\mathbf{A} - \mathbf{B}\|_2$ .

- (a) Prove that the rank function  $\mathbf{A} \mapsto \text{rank}(\mathbf{A})$  is lower semi-continuous at every matrix  $\mathbf{A}_0 \in \mathbb{R}^{m \times n}$ .

Hint. Use the Eckart-Young-Mirsky theorem to prove a contradiction.

- (b) Construct an example to show that in general, the rank of a matrix is not upper semi-continuous.