

# Numerical Mathematics for Engineers II

## Homework 1

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### Exercise 1

**Claim 1.** Let  $\mathbf{A} \in \mathbb{R}^{n \times n}$  be symmetric.  $\mathbf{A}$  is positive definite iff all eigenvalues of  $\mathbf{A}$  are positive.

*Proof.*  $\Rightarrow$  Let  $x$  be an eigenvector of  $\mathbf{A}$  with  $\lambda x = \mathbf{A}x$ .  $0 < x^T \mathbf{A}x = \lambda x^T x = \lambda |x|^2$ . Since  $|x|^2 > 0 \Rightarrow \lambda > 0$ .

$\Leftarrow$  Since  $\mathbf{A} \in \mathbb{R}^{n \times n}$  and  $\mathbf{A} = \mathbf{A}^T$  there exists an orthonormal basis  $(b_1, \dots, b_n)$  consisting of the eigenvectors of  $\mathbf{A}$ . So for any vector exists  $c_i$  such that  $x = \sum_{i=1}^n c_i b_i$ .  $x^T \mathbf{A}x = \langle x, \mathbf{A}x \rangle = \langle \sum_{i=1}^n c_i b_i, \sum_{i=1}^n \lambda_i c_i b_i \rangle = \sum_{i,j=1}^n \langle c_j b_j, \lambda_i c_i b_i \rangle = \sum_{i=1}^n \langle c_i b_i, \lambda_i c_i b_i \rangle = \sum_{i=1}^n \lambda_i c_i^2 > 0$ .

In the next to last step, we have used that the basis vectors are orthogonal. In the last step we have used that all  $\lambda_i$  are positive and the sum of them multiplied by a positive factor  $c_i^2$ .

□

### Exercise 2

Let  $\mathbf{A} \in \mathbb{R}^{n \times n}$  be a symmetric and positive definite matrix.

(a)

**Claim 2.**  $\mathbf{A}$  is invertible and  $\mathbf{A}^{-1}$  is SPD

*Proof.* Let's assume that  $\mathbf{A}$  is not invertible. That means  $\exists x$  such that  $\mathbf{A}x = 0$ . This further implies  $x^T \mathbf{A}x = x^T 0 = 0$ , which contradicts the fact that  $\mathbf{A}$  is positive definite.

Then it holds  $\mathbf{A} \cdot \mathbf{A}^{-1} = \mathbf{I}$  and  $\mathbf{A}^{-1} \mathbf{A} = \mathbf{I}$ . With  $(\mathbf{A}^{-1} \mathbf{A})^T = \mathbf{A}^T (\mathbf{A}^{-1})^T = \mathbf{A} (\mathbf{A}^{-1})^T = \mathbf{I}$ . And thus it holds  $\mathbf{A}^{-1} = (\mathbf{A}^{-1})^T$ .

The positive definiteness follows from:  $0 < x^T \mathbf{A}x = x^T \mathbf{A} \mathbf{A}^{-1} \mathbf{A}x = (\mathbf{A}x)^T \mathbf{A}^{-1} (\mathbf{A}x)$ . Since  $\mathbf{A}$  is regular  $\forall \tilde{x} : \exists x : \tilde{x} = \mathbf{A}x$  and thus with  $\tilde{x} = \mathbf{A}x$  it holds that  $\tilde{x}^T \mathbf{A}^{-1} \tilde{x} > 0$ .

□

(b)

**Claim 3.** The diagonal elements  $\mathbf{a}_{ii}$  of  $\mathbf{A}$  are positive.

*Proof.* Choose  $x = e_i$  with  $i \in [1, n]$ .  $x^T \mathbf{A} x = x^T \mathbf{a}_i$  with  $\mathbf{a}_i$  being the  $i$ th column. Then  $x^T \mathbf{a}_i = \mathbf{a}_{ii} > 0$  because  $\mathbf{A}$  is positive definite.  $\square$

### Exercise 3

Let  $\mathbf{Q} \in \mathbb{R}^{n \times n}$  be an orthogonal matrix.

(a)

**Claim 4.**  $\|\mathbf{Q}x\|_2 = \|x\|_2 \ \forall x \in \mathbb{R}^n$

*Proof.* Since  $\mathbf{Q}$  is an orthogonal matrix, its row vectors form a basis  $(r_1, \dots, r_n)$  of  $\mathbb{R}^n$  with  $\|r_i\|_2 = 1$ . Thus for every  $x \in \mathbb{R}^n$  exist  $c_i$  such that  $x = \sum_i^n c_i r_i$ .  
 $\|\mathbf{Q}x\|_2 = \|\sum_i^n c_i \mathbf{Q}r_i\|_2 = \|\sum_i^n c_i e_i\|_2 = \sqrt{\sum_i^n c_i^2} = \sqrt{\langle \sum_i^n c_i r_i, \sum_i^n c_i r_i \rangle} = \|x\|_2$   $\square$

(b)

**Claim 5.**  $\forall \lambda$  of  $\mathbf{Q}$ :  $|\lambda| = 1$ .

*Proof.* Let  $x$  be an eigenvector of  $\mathbf{Q}$ .  $\|\mathbf{Q}x\|_2 = \|\lambda x\|_2 = |\lambda| \cdot \|x\|_2 = \|x\|_2$ . Thus  $|\lambda| = 1$ .  $\square$

(c)

**Claim 6.**  $\kappa_2(\mathbf{Q}) = 1$

*Proof.*  $\|\mathbf{Q}\|_2 := \sup\{\|\mathbf{Q}y\|_2 : y \in \mathbb{R}^n, \|y\|_2 = 1\}$ . According to (a):  $\|\mathbf{Q}\|_2 = \sup\{\|y\|_2 = 1\} = 1$ . Since  $\mathbf{Q}^{-1} = \mathbf{Q}^T$  and  $\mathbf{Q}^T$  is an orthogonal matrix, it follows that  $\|\mathbf{Q}^{-1}\|_2 = 1$  and thus  $\kappa_2(\mathbf{Q}) = 1$ .  $\square$

(d)

**Claim 7.** Let  $\mathbf{A} \in \mathbb{R}^{n \times n}$ .  $\|\mathbf{A}\|_2 = \|\mathbf{Q}\mathbf{A}\|_2 = \|\mathbf{A}\mathbf{Q}\|_2$ .

*Proof.*  $\|\mathbf{A}\|_2 = \sup\{\|\mathbf{A}y\|_2 : y \in \mathbb{R}^n, \|y\|_2 = 1\} = \sup\{\|x\|_2 : x = \mathbf{A}y, y \in \mathbb{R}^n, \|y\|_2 = 1\} = \sup\{\|\mathbf{Q}x\|_2 : x = \mathbf{A}y, y \in \mathbb{R}^n, \|y\|_2 = 1\} = \sup\{\|\mathbf{Q}\mathbf{A}y\|_2 : y \in \mathbb{R}^n, \|y\|_2 = 1\} = \|\mathbf{Q}\mathbf{A}\|_2$   
 $\|\mathbf{A}\|_2 = \sup\{\|\mathbf{A}y\|_2 : y \in \mathbb{R}^n, \|y\|_2 = 1\} = \sup\{\|\mathbf{A}\mathbf{Q}x\|_2 : \mathbf{Q}x \in \mathbb{R}^n, \|\mathbf{Q}x\|_2 = 1\} = \sup\{\|\mathbf{A}\mathbf{Q}x\|_2 : x \in \mathbb{R}^n, \|\mathbf{Q}x\|_2 = 1\} = \sup\{\|\mathbf{A}\mathbf{Q}x\|_2 : x \in \mathbb{R}^n, \|x\|_2 = 1\} = \|\mathbf{A}\mathbf{Q}\|_2$   $\square$