



Aalto University

# Linear algebra

## Exercise sheet 5 / Model solutions

1. (a) Let  $A \in \mathbb{R}^{n \times n}$  be an invertible matrix. Show that  $\lambda$  is an eigenvalue of  $A$  if and only if  $\lambda^{-1}$  is an eigenvalue of  $A^{-1} \in \mathbb{R}^{n \times n}$ . *Hint:* Show that each eigenvector of  $A$  is also an eigenvector of  $A^{-1}$  and vice versa.
- (b) Let  $A \in \mathbb{R}^{n \times n}$  be an invertible matrix. Show that the matrices  $AA^T$  and  $A^T A$  have the same eigenvalues. *Hint:* Show that  $\det(AA^T - \lambda I) = \det(A^T A - \lambda I)$ .
- (c) Let  $A \in \mathbb{R}^{n \times n}$  be an invertible matrix. Show that

$$\|A^{-1}\|_2^2 = \frac{1}{\lambda_{\min}(A^T A)},$$

where  $\lambda_{\min}(A^T A)$  is the smallest eigenvalue of matrix  $A^T A$ . *Hint:* Use (a) and (b).

*Solution.*

- (a) Note first that  $\lambda \neq 0$  as otherwise  $A$  would not be invertible. Suppose  $Av = \lambda v$  for a nonzero  $v$ . Multiplying this equation on the left by  $\lambda^{-1}A^{-1}$  we get

$$\lambda^{-1}v = A^{-1}v.$$

Conversely, suppose  $A^{-1}w = \lambda^{-1}w$ , and multiply on the left by  $\lambda A$  to conclude that  $\lambda w = Aw$ .

- (b) *Method 1:* Follow the hint which was provided. By direct calculation, we have

$$\begin{aligned} \det(AA^T - \lambda I) &= \det(AA^T - \lambda I) \det(A^{-1}A) \\ &= \det(A^{-1}) \det(AA^T - \lambda I) \det(A) \\ &= \det(A^{-1}(AA^T - \lambda I)A) \\ &= \det(A^T A - \lambda I). \end{aligned}$$

The eigenvalues are the roots of the characteristic polynomial, and this completes the proof. Notice that this proof can be generalized: we can consider any other square matrix  $B$  in place of  $A^T$ . We only needed  $A$  to be invertible for this argument to work.

*Method 2:* Suppose that for a nonzero  $v$  we have  $A^T Av = \lambda v$ ; let also  $w = Av$  and note that  $w$  is nonzero (else,  $v \in N(A)$ , which contradicts the assumption that  $A$  is invertible). Then, multiplying by  $A$  on the left, we have that  $\lambda w = AA^T w$ . Therefore, if  $\lambda$  is an eigenvalue of  $A^T A$  with eigenvector  $v$ , then it is an eigenvalue of  $AA^T$  with eigenvector  $Av$ . Similarly, one can show that if  $\lambda$  is an eigenvalue of  $AA^T$  with eigenvector  $w$  then it is eigenvalue of  $A^T A$  with eigenvector  $A^T w$ .

IMPORTANT: This proof is not valid when either  $A$  or  $A^T$  has a nontrivial null space (for example if  $A$  is square but not invertible, or rectangular), because a crucial step fails!

(c) From results in the lecture notes,

$$\|A^{-1}\|_2^2 = \lambda_{\max}((A^{-1})^T A^{-1}).$$

Using item (a) the right hand side is equal to  $1/\lambda_{\min}(AA^T)$ , which using item (b) is in turn equal to  $1/\lambda_{\min}(A^T A)$ .

2. (a) Let  $\lambda$  be an eigenvalue of  $A \in \mathbb{R}^{n \times n}$ . Show that  $\lambda^2$  is an eigenvalue of the matrix  $A^2$ .  
 (b) Let  $\lambda$  be an eigenvalue of the matrix  $A^2$ , with  $A \in \mathbb{R}^{n \times n}$ . Show that  $\sqrt{\lambda}$  or  $-\sqrt{\lambda}$  is an eigenvalue of  $A$ . (For a complex number  $\lambda$  the notation  $\sqrt{\lambda}$  stands for the main branch of the square root.) *Hint:*  $\det(A^2 - \lambda I) = \det((A - \sqrt{\lambda}I)(A + \sqrt{\lambda}I)) = \det(A - \sqrt{\lambda}I) \det(A + \sqrt{\lambda}I)$ .  
 (c) Let  $A \in \mathbb{R}^{n \times n}$  be symmetric. Show that

$$\|A\|_2 = |\lambda_{\max}(A)|,$$

where  $\lambda_{\max}(A) \in \mathbb{R}$  is an eigenvalue of  $A$  with largest absolute value. *Hint:* Use Lemma 2.5 from the lecture notes and parts (a), (b).

(d) Let  $A \in \mathbb{R}^{n \times n}$  be symmetric and invertible. Show that

$$\|A^{-1}\|_2 = \frac{1}{|\lambda_{\min}(A)|},$$

where  $\lambda_{\min}(A) \in \mathbb{R}$  is an eigenvalue of  $A$  with smallest absolute value. *Hint:* Use Problem 1(a) and (c).

*Solution.*

- (a) Assume that  $A\mathbf{v} = \lambda\mathbf{v}$  for some nonzero  $\mathbf{v}$ . Then,  $A^2\mathbf{v} = A(A\mathbf{v}) = \lambda A\mathbf{v} = \lambda^2\mathbf{v}$ .  
 (b) By the hint, if  $\lambda$  is an eigenvalue of  $A^2$  then we have

$$\det(A - \sqrt{\lambda}I) = 0 \quad \text{or} \quad \det(A + \sqrt{\lambda}I) = 0$$

(or both).

(c) If  $A$  is symmetric then  $A^T A = A^2$ . Hence, by item (a),

$$\|A\|_2^2 = \lambda_{\max}(A^T A) = \lambda_{\max}(A)^2.$$

Taking square roots the statement follows.

(d) This follows immediately by the hint.

3. Let  $A, B \in \mathbb{R}^{n \times n}$  be symmetric. In addition, assume that

$$|\lambda_{\min}(A)| \geq 2 \quad \text{and} \quad |\lambda_{\max}(B)| \leq 1,$$

where  $\lambda_{\min}(A)$  is an eigenvalue of  $A$  with smallest absolute value and  $\lambda_{\max}(B)$  is an eigenvalue of  $B$  with largest absolute value. Show that:

- (a)  $\|A^{-1}\|_2 \leq \frac{1}{2}$ ,  
 (b) The matrix  $A + B$  is invertible.  
 (c) Taking for granted the formula

$$\|(A + B)^{-1}\|_2 \leq \frac{\|A^{-1}\|_2}{1 - \|B\|_2\|A^{-1}\|_2},$$

a solution  $\mathbf{x}$  to the equation  $(A + B)\mathbf{x} = \mathbf{b}$  satisfies  $\|\mathbf{x}\|_2 \leq \|\mathbf{b}\|_2$ .

*Hints:* For (a), use Problem 2(d). For (b), use Theorem 3.1 and Problem 2. For (c), use Problem 2.

*Solution.*

- (a) By Problem 2(b) and by assumption,

$$\|A^{-1}\|_2 = \frac{1}{|\lambda_{\min}(A)|} \leq \frac{1}{2}.$$

- (b) Since  $A$  is invertible,  $A + B$  is also invertible if

$$\|B\|_2 < \underbrace{1/\|A^{-1}\|_2}_{\geq 2},$$

(see the proof of Theorem 3.1). On the other hand, by Problem 2(a),  $\|B\|_2 \leq |\lambda_{\max}(B)| \leq 1$ , so that  $A + B$  is invertible.

- (c) By the given formula, by Problem 2 and by the assumptions, we get

$$\begin{aligned} \|(A + B)^{-1}\|_2 &\leq \frac{\|A^{-1}\|_2}{1 - \|B\|_2\|A^{-1}\|_2} \\ &\leq \frac{1/2}{1 - \|B\|_2\|A^{-1}\|_2} \leq \frac{1/2}{1/2} = 1, \end{aligned}$$

since  $\|B\|_2\|A^{-1}\|_2 \leq 1/2$ . If we have  $(A + B)\mathbf{x} = \mathbf{b}$ , which is equivalent to  $\mathbf{x} = (A + B)^{-1}\mathbf{b}$ , then we get

$$\|\mathbf{x}\|_2 = \|(A + B)^{-1}\mathbf{b}\|_2 \leq \underbrace{\|(A + B)^{-1}\|_2}_{\leq 1} \|\mathbf{b}\|_2 \leq \|\mathbf{b}\|_2.$$

4. Let

$$A = \begin{bmatrix} 1 + \epsilon^2 & 1 \\ 1 & 1 + \epsilon^2 \end{bmatrix} \quad \text{and} \quad \mathbf{b} = \begin{bmatrix} 2 \\ 2 - \delta \end{bmatrix},$$

where  $\epsilon, \delta \in \mathbb{R}$  are free parameters.

- (a) Compute  $\kappa_2(A) = \|A\|_2\|A^{-1}\|_2$ . What happens to the condition number  $\kappa_2(A)$  when  $\epsilon \rightarrow 0$ ?

- (b) Compute  $R(A)$  and  $N(A)$ , when  $\epsilon \neq 0$ . What about  $\epsilon = 0$ ?  
 (c) For  $\epsilon \neq 0$ , solve  $A\mathbf{x} = \mathbf{b}$  using the formula ([Cramer's rule](#))

$$A^{-1} = \frac{1}{\det A} \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix}.$$

How does  $\mathbf{x}$  behave, when  $\delta = 0$  and  $\epsilon \rightarrow 0$ ? What about  $\delta \neq 0$  and  $\epsilon \rightarrow 0$ ?

*Hint:* For (a), use Problem 2.

*Solution.*

- (a) By definition,  $\kappa_2(A) = \|A\|_2 \|A^{-1}\|_2$ . Problem 2 gives easy expressions for  $\|A\|_2$  and  $\|A^{-1}\|_2$ , so our goal is to use those instead of the definition. The eigenvalues of  $A$  are quick to compute: we set equal to zero the determinant

$$\begin{vmatrix} 1 + \epsilon^2 - \lambda & 1 \\ 1 & 1 + \epsilon^2 - \lambda \end{vmatrix} = (1 + \epsilon^2 - \lambda)^2 - 1,$$

so that we get  $(1 + \epsilon^2 - \lambda)^2 = 1$ , meaning that  $1 + \epsilon^2 - \lambda = \pm 1$ . So the two eigenvalues are  $\lambda = \epsilon^2$  and  $\lambda = \epsilon^2 + 2$ . Using the same notation as Problem 2, we then get

$$\kappa_2(A) = \frac{|\lambda_{\max}(A)|}{|\lambda_{\min}(A)|} = \frac{\epsilon^2 + 2}{\epsilon^2}.$$

This goes to  $+\infty$  for  $\epsilon \rightarrow 0$ .

- (b) When  $\epsilon \neq 0$ , we have  $\det(A) = 2\epsilon^2 + \epsilon^4 > 0$ , hence  $A$  is invertible (it follows also from part (a)). So then  $R(A) = \mathbb{R}^2$  and  $N(A) = \{\mathbf{0}\}$ . When  $\epsilon = 0$ , then we easily see that  $R(A) = \text{span}\{[1 \ 1]^T\}$  and also  $N(A) = \text{span}\{[1 \ -1]^T\}$ .  
 (c) By Cramer's rule, we get

$$A^{-1} = \frac{1}{2\epsilon^2 + \epsilon^4} \begin{bmatrix} 1 + \epsilon^2 & -1 \\ -1 & 1 + \epsilon^2 \end{bmatrix}.$$

Multiplying  $A\mathbf{x} = \mathbf{b}$  by  $A^{-1}$  on the left, we get

$$\mathbf{x} = \frac{1}{2\epsilon^2 + \epsilon^4} \begin{bmatrix} 2\epsilon^2 + \delta \\ \epsilon^2(2 - \delta) - \delta \end{bmatrix}.$$

For  $\delta = 0$  and  $\epsilon \rightarrow 0$ , this is

$$\mathbf{x} = \begin{bmatrix} \frac{2}{2+\epsilon^2} \\ \frac{2}{2+\epsilon^2} \end{bmatrix} \rightarrow \begin{bmatrix} 1 \\ 1 \end{bmatrix},$$

whereas for  $\delta \neq 0$  we get

$$\mathbf{x} = \begin{bmatrix} \frac{2\epsilon^2 + \delta}{2\epsilon^2 + \epsilon^4} \\ \frac{\epsilon^2(2 - \delta) - \delta}{2\epsilon^2 + \epsilon^4} \end{bmatrix} \rightarrow \begin{bmatrix} \pm\infty \\ \mp\infty \end{bmatrix},$$

where the sign depends on the sign of  $\delta$ .