

## Linear algebra

## **Exercise sheet 8 / Model solutions**

1. For any  $A \in \mathbb{C}^{n \times n}$  and  $\lambda \in \mathbb{C}$  it holds that

$$(\lambda I + A)^k = \sum_{i=0}^k {k \choose i} \lambda^{k-i} A^i, \qquad k \in \mathbb{N}_0 = \{0, 1, 2, 3, \dots\}.$$

(a) Let  $T \in \mathbb{C}^{n \times n}$  be a *nilpotent matrix*. That is, there exists some  $p \in \mathbb{N}$  such that  $T^p = \mathbf{0} \in \mathbb{C}^{n \times n}$ . Show that

$$(\lambda I + T)^k = \sum_{i=0}^{\min\{k, p-1\}} {k \choose i} \lambda^{k-i} T^i.$$

(b) Let

$$B = \begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} + \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}.$$

Write an explicit formula for  $B^k$ , for general  $k \in \mathbb{N}$ .

Solution.

- (a) This is immediate by the given formula and the assumption on p.
- (b) Observe that

$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}^2 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}.$$

Hence, using item (a) with  $\lambda = 2$ , for any  $k \ge 1$  it holds

$$B^{k} = \sum_{i=0}^{\min\{k,1\}} \binom{k}{i} 2^{k-i} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}^{i} = 2^{k} I + \binom{k}{1} 2^{k-1} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 2^{k} & k 2^{k-1} \\ 0 & 2^{k} \end{bmatrix}.$$

2. Let

$$A = \begin{bmatrix} 3 & -1 \\ 1 & 1 \end{bmatrix}.$$

- (a) Is the matrix A diagonalizable?
- (b) Compute  $e^{tA}$  using the Jordan decomposition

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^{-1}.$$

*Hints*: (a) Use Example 1.1 in section 2. (b) Use Problem 1(b). Note that  $\sum_{k=1}^{\infty} \frac{1}{k!} kt^k 2^{k-1} = te^{2t}$ .

Solution.

- (a) No. Indeed,  $\det(A \lambda I) = (\lambda 2)^2$  but  $N(A 2I) = \operatorname{span}([1\ 1]^T)$ . This means that 2 is an eigenvalue of A with algebraic multiplicity 2 but geometric multiplicity 1.
- (b) Let

$$X = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}, \qquad B = \begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix}.$$

Then, by item (b) of Problem 1,

$$e^{tA} = \sum_{k=0}^{\infty} \frac{t^k}{k!} A^k = X \sum_{k=0}^{\infty} \frac{t^k}{k!} B^k X^{-1} = X \left( I + \sum_{k=1}^{\infty} \begin{bmatrix} \frac{(2t)^k}{k!} & t \frac{(2t)^{k-1}}{(k-1)!} \\ 0 & \frac{(2t)^k}{k!} \end{bmatrix} \right) X^{-1}.$$

Summing entry by entry, this is in turn equal to

$$X \begin{bmatrix} e^{2t} & te^{2t} \\ 0 & e^{2t} \end{bmatrix} X^{-1} = \begin{bmatrix} (t+1)e^{2t} & -te^{2t} \\ te^{2t} & (1-t)e^{2t} \end{bmatrix}.$$

3. Consider the system of differential equations

$$\mathbf{x}'(t) = A\mathbf{x}(t) \quad \text{for } t > 0, \qquad \mathbf{x}(0) = \begin{bmatrix} 1 \\ 0 \end{bmatrix},$$
 (1)

where  $A \in \mathbb{C}^{2 \times 2}$  and  $x : \mathbb{R}_+ \to \mathbb{C}^2$ . Solve (1) by using the matrix exponential, when

(a)

$$A = \begin{bmatrix} 3 & -1 \\ 1 & 1 \end{bmatrix},$$

(b)

$$A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}.$$

*Hints*: (a) Use Problem 2(b). (b) Observe that A is symmetric.

Solution. From the lectures we know that the solution of (1) is

$$\boldsymbol{x}(t) = e^{tA} \boldsymbol{x}(0) = e^{tA} \begin{bmatrix} 1 \\ 0 \end{bmatrix}.$$

(a) By Problem 2(b) we know that

$$e^{tA} = \begin{bmatrix} (1+t)e^{2t} & -te^{2t} \\ te^{2t} & (1-t)e^{2t} \end{bmatrix},$$

hence

$$\boldsymbol{x}(t) = e^{tA} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} (1+t)e^{2t} \\ te^{2t} \end{bmatrix} = \begin{bmatrix} 1+t \\ t \end{bmatrix} e^{2t}.$$

(b) Let us diagonalize A. We know that this happens unitarily and with real eigenvalues, because  $A^* = A$ . We have the characteristic polynomial

$$p_A(A) = \det(A - \lambda I) = \begin{vmatrix} 1 - \lambda & 3 \\ 3 & 1 - \lambda \end{vmatrix}$$
$$= \lambda^2 - 2\lambda - 8.$$

Therefore the eigenvalues are  $\lambda_1 = -2$  and  $\lambda_2 = 4$ . Let us now compute the eigenvectors. We have  $v \in E_{-2} = N(A + 2I)$  if and only if

$$\begin{bmatrix} 1+2 & 3 \\ 3 & 1+2 \end{bmatrix} \boldsymbol{v} = \boldsymbol{0},$$

which means that  $\mathbf{v} = \alpha \begin{bmatrix} 1 & -1 \end{bmatrix}^T$ , for any  $\alpha \in \mathbb{R}$ . Similarly, we have  $\mathbf{v} \in E_4 = N(A-4I)$  if and only if  $\mathbf{v} = \alpha \begin{bmatrix} 1 & 1 \end{bmatrix}^T$ , for any  $\alpha \in \mathbb{R}$ . Let us pick unitary eigenvectors:

$$\boldsymbol{v}_1 = rac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \in E_{-2}, \qquad \boldsymbol{v}_2 = rac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \in E_4.$$

Since A is symmetric,  $v_1$  and  $v_2$  are orthogonal with respect to the euclidean inner product (eays). So, if we set

$$V = egin{bmatrix} oldsymbol{v}_1 & oldsymbol{v}_2 \end{bmatrix} = egin{bmatrix} rac{1}{\sqrt{2}} & rac{1}{\sqrt{2}} \ -rac{1}{\sqrt{2}} & rac{1}{\sqrt{2}} \end{bmatrix},$$

then

$$V^{-1} = V^{T} = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}.$$

We can diagonalize the matrix A in the form

$$A = V \begin{bmatrix} -2 & 0 \\ 0 & 4 \end{bmatrix} V^T = V \Lambda V^T.$$

And now we can easily compute

$$\begin{split} e^{tA} &= V e^{t\Lambda} V^T = V \begin{bmatrix} e^{-2t} & 0 \\ 0 & e^{4t} \end{bmatrix} V^T \\ &= \left(\frac{1}{\sqrt{2}}\right)^2 \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} e^{-2t} & 0 \\ 0 & e^{4t} \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \\ &= \frac{1}{2} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} e^{-2t} & -e^{-2t} \\ e^{4t} & e^{4t} \end{bmatrix} \\ &= \frac{1}{2} \begin{bmatrix} e^{-2t} + e^{4t} & -e^{-2t} + e^{4t} \\ -e^{-2t} + e^{4t} & e^{-2t} + e^{4t} \end{bmatrix}. \end{split}$$

And then the solution to (1) in this case is

$$x(t) = e^{tA} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} e^{-2t} + e^{4t} \\ -e^{-2t} + e^{4t} \end{bmatrix}.$$

4. Consider the matrix

$$A = \begin{bmatrix} 1 & 1 \\ 0 & 1 + \epsilon \end{bmatrix},$$

where  $\epsilon \in \mathbb{R}$ .

- (a) For which values of  $\epsilon$  is the matrix A diagonalizable?
- (b) Let  $\epsilon$  be such that A is diagonalizable. Find an invertible  $V \in \mathbb{C}^{2\times 2}$  and a diagonal matrix  $\Lambda \in \mathbb{C}^{2\times 2}$  so that  $A = V\Lambda V^{-1}$ . Scale the columns of V so that the first row of V is  $\begin{bmatrix} 1 & 1 \end{bmatrix}$ .
- (c) Compute the condition number  $\kappa_2(V)$  using the Matlab function cond. Plot the condition number as a function of  $\epsilon$  on the intervall  $\epsilon \in [10^{-4}, 1]$ . Use semilogarithmic scale, see help semilogy. What happens when A is very close to a non-diagonalizable matrix?
- (d) Set  $\epsilon = 0$  and try to compute V and  $\Lambda$  using the Matlab function eig. What is the condition number  $\kappa_2(V)$ ? Is the diagonalization given by Matlab plausible? (Compare the result to (a).)

*Hints*: (a) If a  $(2 \times 2)$ -matrix has two distinct eigenvalues, it is diagonalizable (see Section 2, Theorem 1.1 of the lecture notes); if this is not the case, one has to check that the geometric and algebraic multiplicities of each eigenvalue meet. (b) Note that  $\Lambda$  and V depend on the parameter  $\epsilon$ .

Solution.

(a,b) The eigenvalues of an upper triangular matrix can be seen directly on the diagonal: they are  $\lambda_1=1$  and  $\lambda_2=1+\epsilon$ . The matrix is surely diagonalizable when the two eigenvalues are distinct, that is, when  $\epsilon\neq 0$  (in which case the algebraic multiplicity of both eigenvalues is 1). Let's assume now that  $\epsilon\neq 0$ . Let's find the eigenvectors for  $\lambda_1=1$ :

$$(A-1I)\boldsymbol{x} = \begin{bmatrix} 0 & 1 \\ 0 & \epsilon \end{bmatrix} \boldsymbol{x} = 0 \qquad \Rightarrow \qquad \boldsymbol{x} = t \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad t \in \mathbb{R}.$$

And eigenvectors for  $\lambda_2 = 1 + \epsilon$ :

$$(A - (1 + \epsilon)I)\boldsymbol{x} = \begin{bmatrix} -\epsilon & 1 \\ 0 & 0 \end{bmatrix} \boldsymbol{x} = 0 \qquad \Rightarrow \qquad \boldsymbol{x} = t \begin{bmatrix} 1 \\ \epsilon \end{bmatrix}, \quad t \in \mathbb{R}.$$

So a diagonalization is

$$A = X\Lambda X^{-1} = \begin{bmatrix} 1 & 1 \\ 0 & \epsilon \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 + \epsilon \end{bmatrix} \begin{bmatrix} 1 & -\frac{1}{\epsilon} \\ 0 & \frac{1}{\epsilon} \end{bmatrix}.$$

In case  $\epsilon=0$ ,we have  $\lambda_1=\lambda_2=1$ , and we get:

$$(A-1I)\boldsymbol{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \boldsymbol{x} = 0 \qquad \Rightarrow \qquad \boldsymbol{x} = t \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad t \in \mathbb{R}.$$

The geometric multiplicity is then 1, hence the matrix is not diagonalizable.

(c) The condition number of the matrix X is

$$\kappa_2(X) = ||X||_2 ||X^{-1}||_2 = \sigma_{\max} \sigma_{\min}^{-1}.$$

Let us compute the squares of the singular values: since we have

$$X^T X = \begin{bmatrix} 1 & 0 \\ 1 & \epsilon \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & \epsilon \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 1 + \epsilon^2 \end{bmatrix},$$

we get  $\det(X^TX - \lambda I) = (1 - \lambda)(1 + \epsilon^2 - \lambda) - 1$ , and by setting this qual to zero we get  $\lambda^2 - (2 + \epsilon^2)\lambda + \epsilon^2 = 0$ , whose solutions are

$$\lambda = \frac{2 + \epsilon^2 \pm \sqrt{\epsilon^4 + 4}}{2}.$$

And hence we get

$$\kappa_2(X) = \sigma_{\max} \sigma_{\min}^{-1} = \sqrt{\frac{2 + \epsilon^2 + \sqrt{\epsilon^4 + 4}}{2 + \epsilon^2 - \sqrt{\epsilon^4 + 4}}} = \sqrt{\frac{8 + 4\epsilon^2 + 2\epsilon^4 + (4 + 2\epsilon^2)\sqrt{\epsilon^4 + 4}}{4\epsilon^2}}.$$

And then we get  $\kappa_2(X) \approx \sqrt{\frac{4}{\epsilon^2}} = \frac{2}{|\epsilon|}$  for small values of  $\epsilon$ . Hence the condition number grows to  $+\infty$  when the matrix gets close to being non-diagonalizable.

(d) The following is Matlab code which includes the output.

V =

$$\begin{array}{ccc} 1.0000 & -1.0000 \\ 0 & 0.0000 \end{array}$$

D =

>> cond(V)