

**Exercise 1.** Determine the LU decomposition with row swaps, i.e.  $L$ ,  $U$  and the corresponding permutation vector  $p$  for the following matrices. Then solve the system  $Ax = e_1$ . Here,  $e_1 = (1, 0, \dots, 0)^T$ .

$$\text{a) } A = \begin{pmatrix} 0 & 0 & 1 \\ 2 & 0 & 1 \\ 2 & 2 & 4 \end{pmatrix} \qquad \text{b) } A = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 2 & 4 \\ 1 & 1 & 2 & 4 \end{pmatrix}.$$

**Suggested Solution.**

(a) Step 1: Swap rows 1 and 2, i.e.  $p = (2)$  and eliminate the 1st column.

$$\Rightarrow P_1 A = \begin{pmatrix} 2 & 0 & 1 \\ 0 & 0 & 1 \\ 2 & 2 & 4 \end{pmatrix} = \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix}}_{L_1^{-1}} \underbrace{\begin{pmatrix} 2 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & 2 & 3 \end{pmatrix}}_{A^{(1)}}$$

Step 2: Swap rows 2 and 3, i.e.  $p = (2, 3)$ .

$$\Rightarrow P_2 P_1 A = \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}}_{P_2 L_1^{-1} P_2 = L} \underbrace{\begin{pmatrix} 2 & 0 & 1 \\ 0 & 2 & 3 \\ 0 & 0 & 1 \end{pmatrix}}_{P_2 A^{(1)} = U} = LU$$

(b) Step 1: Swap rows 1 and 2, i.e.  $p = (2)$  and eliminate the 1st column.

$$P_1 A = \begin{pmatrix} 2 & 0 & 2 & 1 \\ 0 & 0 & 0 & 1 \\ 2 & 2 & 2 & 4 \\ 1 & 1 & 2 & 4 \end{pmatrix} = \underbrace{\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ \frac{1}{2} & 0 & 0 & 1 \end{pmatrix}}_{L_1^{-1}} \underbrace{\begin{pmatrix} 2 & 0 & 2 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 2 & 0 & 3 \\ 0 & 1 & 1 & \frac{7}{2} \end{pmatrix}}_{A^{(1)}}$$

Step 2: Swap rows 2 and 4, i.e.  $p = (2, 4)$ .

$$\Rightarrow P_2 P_1 A = \underbrace{\begin{pmatrix} 1 & 0 & 0 & 0 \\ \frac{1}{2} & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}}_{P_2 L_1^{-1} P_2} \underbrace{\begin{pmatrix} 2 & 0 & 2 & 1 \\ 0 & 1 & 1 & \frac{7}{2} \\ 0 & 2 & 0 & 3 \\ 0 & 0 & 0 & 1 \end{pmatrix}}_{P_2 A^{(1)}}$$

Eliminate the 2nd column:

$$\Rightarrow P_2 P_1 A = \underbrace{\begin{pmatrix} 1 & 0 & 0 & 0 \\ \frac{1}{2} & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}}_{P_2 L_1^{-1} P_2 L_2^{-1} = L} \underbrace{\begin{pmatrix} 2 & 0 & 2 & 1 \\ 0 & 1 & 1 & \frac{7}{2} \\ 0 & 2 & 0 & 3 \\ 0 & 0 & 0 & 1 \end{pmatrix}}_{L_2 P_2 A^{(1)} = A^{(2)} = U} = LU$$

Now we respectively solve  $Ax = e_1$ : Note that  $Ax = b \Leftrightarrow PAx = LUx = Pb$ .

(a)  $p = (2, 3)$ , i.e.

$$b = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \xrightarrow{I \leftrightarrow II} \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \xrightarrow{II \leftrightarrow III} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = Pb$$

(1) Solve  $Ly = Pb$

$$\Leftrightarrow \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} y = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \Rightarrow \begin{cases} y_1 = 0 \\ y_2 = -y_1 = 0 \\ y_3 = 1 \end{cases}$$

(2) Solve  $Rx = y$ :

$$\Leftrightarrow \begin{pmatrix} 2 & 0 & 1 \\ 0 & 2 & 3 \\ 0 & 0 & 1 \end{pmatrix} x = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \Rightarrow \begin{cases} x_1 = -\frac{1}{2}x_3 = -\frac{1}{2} \\ x_2 = -\frac{3}{2}x_3 = -\frac{3}{2} \\ x_3 = 1 \end{cases}$$

(b)  $p = (2, 4)$ , i.e.

$$b = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} \xrightarrow{I \leftrightarrow II} \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \end{pmatrix} \xrightarrow{II \leftrightarrow IV} \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} = Pb$$

(1) Solve  $Ly = Pb$

$$\Leftrightarrow \begin{pmatrix} 1 & 0 & 0 & 0 \\ \frac{1}{2} & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} y = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \Rightarrow \begin{cases} y_1 = 0 \\ y_2 = 0 \\ y_3 = 0 \\ y_4 = 1 \end{cases}$$

(2) Solve  $Rx = y$ :

$$\Leftrightarrow \begin{pmatrix} 2 & 0 & 2 & 1 \\ 0 & 1 & 1 & \frac{7}{2} \\ 0 & 0 & -2 & -4 \\ 0 & 0 & 0 & 1 \end{pmatrix} x = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \Rightarrow \begin{cases} x_1 = -x_3 - \frac{1}{2}x_4 = \frac{3}{2} \\ x_2 = -\frac{7}{2}x_4 - x_3 = -\frac{3}{2} \\ x_3 = -2x_4 = -2 \\ x_4 = 1 \end{cases}$$

**Exercise 2.** Let  $A = (a_{ij}) \in \mathbb{R}^{n \times n}$  be arbitrary. The Frobenius norm is defined by

$$\|A\|_F = \left( \sum_{i=1}^n \sum_{j=1}^n a_{ij}^2 \right)^{\frac{1}{2}}.$$

Proof the following:

(a) For  $n \geq 2$  the Frobenius norm is not induced by any norm.

**Hint:** Exercise Sheet 2, Exercise 2(d).

(b)  $\|A\|_F^2 = \text{trace}(A^T A)$

(c) The Frobenius norm is compatible with the Euclidean norm  $\|\cdot\|_2$ , i.e. for all  $x \in \mathbb{R}^n$  it is

$$\|Ax\|_2 \leq \|A\|_F \|x\|_2.$$

**Hint:** Consider the squared inequality and use the (squared) Cauchy-Schwarz inequality to estimate  $\|Ax\|_2^2$ .

(d)  $\|A\|_2 \leq \|A\|_F$ .

**Hint:** Use subtask (c)

**Suggested Solution.** (a) Assume the contrary. From Sheet 2, Ex. 2(d) we conclude that  $\|E\|_F = 1$ . But this contradicts

$$\|E\|_F^2 = \sum_{ij} \delta_{ij}^2 = \sum_{i=1}^n 1 = n$$

at least for  $n \geq 2$ .

(b) With  $A = [a_1, \dots, a_n]$  we obtain

$$\text{trace}(A^T A) = \sum_j \langle e_j, A^T A e_j \rangle = \sum_j \langle A e_j, A e_j \rangle = \sum_j \langle a_j, a_j \rangle = \sum_{ij} a_{ij}^2.$$

(c) Let  $a_i^T$  be the  $i$ th row of  $A$  and  $x \in \mathbb{R}^n$  be a unit vector. Then

$$\begin{aligned} \|Ax\|_2^2 &= \sum_i \langle a_i, x \rangle^2 \\ &\leq \sum_i \|a_i\|_2^2 \|x\|_2^2 \\ &= \sum_{ij} a_{ij}^2. \end{aligned}$$

(d) Take the supremum over all  $x \neq 0$  in

$$\frac{\|Ax\|_2}{\|x\|_2} \leq \|A\|_F$$

yields the result.

**Exercise 3.** Let  $A \in \mathbb{R}^{n \times n}$  be *nilpotent*, i.e. for some  $N \in \mathbb{N}$  it is

$$A^N = 0.$$

Are the following statements true or false (give a reason for each!):

(a)  $\det(A) = 0$

(b)  $\sigma(A) = \{0\}$

(c)  $\text{trace}(A) = 0$

(d)  $A = 0$  or  $A$  is not diagonalizable

(e)  $E - A$  is invertible

**Suggested Solution.** (a) true, since

$$0 = \det(A^N) = (\det A)^N.$$

(b)  $\sigma(A) = \{0\}$  is true: Let  $\lambda \in \sigma(A)$  be any eigenvalue and  $v$  be a corresponding eigenvector. Then by induction we get  $A^N v = \lambda^N v$ . Since  $v \neq 0$ ,  $0 = A^N v = \lambda^N v$  implies that  $\lambda^N = 0$ , hence  $\lambda = 0$ .

(c)  $\text{trace}(A) = 0$  is true because  $\sigma(A) = \{0\}$  implies

$$\text{trace}(A) = \sum_{\lambda \in \sigma(A)} \mu(\chi_A, \lambda) \lambda = 0.$$

(d) true: Assume the contrary, i.e.  $A \neq 0$  is diagonalizable. This means that

$$0 \neq A = SDS^{-1}$$

for some  $D = \text{diag}(d_1, \dots, d_n) \neq 0$  (otherwise  $SDS^{-1} = 0$ ) and  $S \in \text{Gl}(n, \mathbb{R})$ . By induction we can deduce

$$0 = A^N = (SDS^{-1})^N = SD^N S^{-1} = S \begin{pmatrix} d_1^N & & 0 \\ & \ddots & \\ 0 & & d_n^N \end{pmatrix} S^{-1}.$$

This implies that  $D^N = 0 \Rightarrow D = 0$ , contradicting  $D \neq 0$ .

(e)  $E - A$  is invertible since  $E - A$  is injective:

$$(E - A)v = 0 \Rightarrow Av = 1 \cdot v \xrightarrow{\sigma(A)=\{0\}} v = 0.$$

Another possibility:

$$(E - A) \sum_{j=0}^{N-1} A^j = E - A^N = E,$$

i.e.  $(E - A)$  is invertible with  $(E - A)^{-1} = \sum_{j=0}^{N-1} A^j$ .

**Exercise 4.** Write a program that will implement the LU decomposition (without permutations) preferably with Octave/Matlab. The program should receive an arbitrary  $(n \times n)$ -matrix  $A$  as input and the matrices  $L$  (unipotent lower triangular matrix) and  $U$  (upper triangular matrix) as output, such that

$$A = LU.$$

Test your program with the matrices from Sheet 3, Ex. 3.

**Suggested Solution.**

```
function [L,U] = LU_wo_permutations(A)
% Determine the matrix dimension
[n,n] = size(A);
% initialize L with a unit matrix
L = eye(n);
% Outer for loop: Iterates through all columns to be eliminated
for j=1:(n-1)
% Inner for loop: Eliminates the jth column below the diagonal
for i=(j+1):n
L(i,j) = A(i,j)/A(j,j);
end
```

```
% Carry out row transformations in A
for i=(j+1):n
    A(i,:) = A(i,:) - L(i,j)*A(j,:);
end
end
U=A;
end
```