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# Hanoi Fall-School 2013 on Numerical Analysis

# Exercise sheet 1

## Numerical Linear Algebra - "LU factorization"

You will find all the material for the first exercise in:

https://github.com/ukandler/hanoi2013 numerical analyses

**Exercise 1.1** Solve the system 
$$\begin{bmatrix} 4 & 1 & 4 \\ 8 & 4 & 6 \\ 8 & 5 & 6 \end{bmatrix} x = \begin{bmatrix} 1 \\ 4 \\ 0 \end{bmatrix}$$
 using the LU decomposition

- (a) without pivoting.
- (b) with partial pivoting.

**Definition 1** Let  $A \in \mathbb{R}^{n \times n}$  and let  $A^{(k)} \in \mathbb{R}^{n \times n}$  define the converted system after the k-th step of the Gaussian elimination. Then the growth factor of A is defined by

$$g_n = \frac{\max_{i,j,k} |a_{ij}^{(k)}|}{\max_{i,j} |a_{ij}|}.$$

#### Programming 1

- (a) Write in Octave function [LU,g] = LUfac(A) that
  - given a matrix  $A \in \mathbb{R}^{n \times n}$
  - computes the LU factorization of A and the growth factor  $g_n$ . (Hint: L and U are stored in place, i.e. in the same array as A)
- (b) Write a second program function x = forback(LU,b) that
  - using the LU factorization of A and the right hand side  $b \in \mathbb{R}^n$
  - computes the solution of the linear system Ax = b.

Test your program on the following examples:

i) 
$$\begin{bmatrix} 5 & 3 & -1 & 0 \\ 2 & 5 & 0 & 1 \\ -1 & 0 & 5 & -2 \\ 0 & 1 & -2 & 5 \end{bmatrix} \mathbf{x} = \begin{bmatrix} 27 \\ 24 \\ 4 \\ 4 \end{bmatrix}$$
 ii) 
$$\begin{bmatrix} \varepsilon & 2 \\ 1 & 1 \end{bmatrix} \mathbf{x} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
 with  $\varepsilon = 1e-16$ .

Compare your results with the one you get from  $x = A \setminus b$ . Explain numerically what happens in the case ii).

- (c) Write an OCTAVE program function [LU,p,g] = LUfac(A,pivot) that
  - given a matrix  $A \in \mathbb{R}^{n \times n}$
  - computes the LU factorization with or without partial pivoting of A and the growth factor  $g_n$ . (Hint: Include an option in your first function such that the user can decide whether partial pivoting is used (pivot=1) or not (pivot=0).)

Start the runme.m file. If your program is working correctly a little car is driving over a bridge. (Hint: It will take some seconds to see the result! )

Run your program on 10 random matrices of dimension 10, 30, 100, 300 and 1000 and store the different growth factors in a 10 by 5 matrix.

- What is the maximum growth factor with and without partial pivoting?
- In the case of partial pivoting plot the growth factors as well as  $f(n) = \sqrt{n}$ . Explain the result. (Hint: Use a loglog scale plot.)

**Exercise 1.2** How many operations (divisions and multiplications) are necessary to perform an LU decomposition without pivoting?

**Exercise 1.3** Show that for LU factorization with partial pivoting applied to any matrix  $A \in \mathbb{R}^{n \times n}$  the growth factor  $g_n$  satisfies  $g_n \leq 2^{n-1}$ .

**Definition 2** A matrix  $A \in \mathbb{R}^{n \times n}$  is called diagonally dominant, if

$$|a_{ii}| \ge \sum_{\substack{j=1\\j\neq i}}^{n} |a_{ij}|, \quad i = 1, \dots, n.$$

**Exercise 1.4** Let  $A \in \mathbb{R}^{n \times n}$  be a nonsingular and diagonally dominant. Show that partial pivoting is not needed to perform the LU factorization.

**Lemma 1** (Perturbation Lemma) Let  $P \in \mathbb{R}^{n \times n}$  and ||P|| < 1. Then I-P is nonsingular and fulfills the estimation

$$||(I-P)^{-1}|| \le \frac{1}{(1-||P||)}.$$

**Exercise 1.5** Proof the following Theorem.

**Theorem 1** Let Ax = b be a linear system with  $A \in \mathbb{R}^{n \times n}$  and the solution  $x = A^{-1}b$ . Let  $(A + \Delta A)(x + \Delta x) = b + \Delta b$  be the corresponding perturbed system with  $\Delta A \in \mathbb{R}^{n \times n}$ . If  $\|\Delta A\| \leq \|A\|/\kappa(A)$ , then  $A + \Delta A$  is invertible and it holds

$$\frac{\|\Delta x\|}{\|x\|} \le \frac{\kappa(A)}{1 - \kappa(A) \frac{\|\Delta A\|}{\|A\|}} \left( \frac{\|\Delta A\|}{\|A\|} + \frac{\|\Delta b\|}{\|b\|} \right).$$

**Definition 3** The matrix  $A \in \mathbb{R}^{n \times n}$  is positive definite if and only if for all  $x \in \mathbb{R}$ ,  $x \neq 0$  it holds that  $x^T A x > 0$ . The matrix A is symmetric if  $A = A^T$ .

**Exercise 1.6** Let  $A \in \mathbb{R}^{n \times n}$  be symmetric and positive definite. Show that

- (a)  $a_{jj} > 0$  for  $j = 1, 2, \dots n$ .
- (b)  $a_{jk}^2 < a_{jj}a_{kk}$  for  $j, k = 1, 2, \dots n, \quad j \neq k$ .
- (c) the largest entry of A in magnitude is on the main diagonal.

**Exercise 1.7** Estimate the condition number of the matrix

$$\mathbf{A} = \begin{bmatrix} 1 & -2 & 0 & \dots & 0 \\ 0 & 1 & -2 & 0 & \dots & 0 \\ & \ddots & \ddots & & & \\ & & \ddots & \ddots & & \\ 0 & & \dots & 0 & 1 & -2 \\ 0 & & & \dots & 0 & 1 \end{bmatrix} \in \mathbb{R}^{n \times n}.$$

(Hint: Consider  $\mathbf{A}\mathbf{x} = \mathbf{b}$  with  $\mathbf{x} = \begin{bmatrix} 2^{n-1} & \dots & 8 & 4 & 2 & 1 \end{bmatrix}^T$ .)

#### Additional exercises

**Definition 4** For  $A \in \mathbb{R}^{m \times n}$ 

$$||A||_1 := \max_{x \neq 0} \frac{||Ax||_1}{||x||_1} \quad ||A||_{\infty} := \max_{x \neq 0} \frac{||Ax||_{\infty}}{||x||_{\infty}}$$
$$||A||_2 := \max_{x \neq 0} \frac{||Ax||_2}{||x||_2} \quad ||A||_F := \sqrt{\sum_{i=1}^m \sum_{k=1}^n |a_{ik}|^2}$$

## Exercise 1.8\* (norm properties)

Let  $\|.\|$  be a vector norm in  $\mathbb{R}^n$  and

$$||A|| := \max_{x \neq 0} \frac{||Ax||}{||x||} = \max_{||x|| = 1} ||Ax||$$

the induced matrix norm. Let  $A, B \in \mathbb{R}^{n \times n}$  and  $b \in \mathbb{R}^n$ . Show that:

- (a)  $\kappa(AB) \le \kappa(A)\kappa(B)$
- (b)  $1 \le \kappa(A)$
- (c)  $\frac{1}{\|A^{-1}b\|} \le \frac{\|A\|}{\|b\|}$

(d) 
$$||A||_1 = \max_{k=1,\dots,n} \sum_{i=1}^m |a_{ik}|$$

(e) 
$$||A||_{\infty} = \max_{i=1,\dots,m} \sum_{k=1}^{n} |a_{ik}|$$

(f) 
$$||A||_2 = \sqrt{\lambda_{\max}(A^T A)}$$
 (where  $\lambda_{\max}(A)$  denotes the largest eigenvalue of  $A$ )

**Definition 5** A floating point numbers  $\mathcal{F}(\beta, t, e_{\min}, e_{\max})$  are characterized by the base  $\beta$ , the precision t and the exponent range  $[e_{\min}, e_{\max}]$ .  $\mathcal{F}$  consists of all numbers f of the form

$$f = \pm d_1 d_2 \dots d_t \times \beta^e$$
,  $0 \le d_i < \beta$ ,  $d_1 \ne 0$   $e_{\min} \le e \le e_{\max}$ .

## Exercise 1.9\* (rounding)

Let  $x_1, x_2 \in \mathbb{R}$  and  $\bar{x} = \frac{x_1 + x_2}{2}$ . In exact arithmetic it holds the inequality

$$\min\{x_1, x_2\} \le \bar{x} \le \max\{x_1, x_2\} \tag{1}$$

In general this is not true in floating point arithmetic with  $\bar{x} = fl(fl(x_1 + x_2)/2)$ . Find  $x_1, x_2 \in \mathcal{F}(10, 2, -3, 3)$  such that (1) is violated.

How should  $\bar{x}$  be modified in floating point arithmetic such that (1) holds?