

# Solving linear systems

The problem: given  $\underline{b} \in \mathbb{R}^n$ , and  $A \in \mathbb{R}^n \times \mathbb{R}^n$ , we look for  $\underline{x} \in \mathbb{R}^n$  solution of

$$A\underline{x} = \underline{b} \quad (1)$$

Problem (1) has a unique solution *if and only if* the matrix  $A$  is non-singular (or invertible), i.e.,  $\exists A^{-1}$  such that  $A^{-1}A = AA^{-1} = I$ ; necessary and sufficient condition for  $A$  being invertible is that  $\det(A) \neq 0$ . Then the solution  $\underline{x}$  is formally given by  $\underline{x} = A^{-1}\underline{b}$ .

**Beware:** never invert a matrix unless really necessary, due to the costs, as we shall see later on. (Solving a system with a general full matrix is also expensive, but not nearly as expensive as matrix inversion).

## Some example of linear systems

The simplest systems to deal with are diagonal systems:

$$D\underline{x} = \underline{b}$$

$$D = \begin{bmatrix} d_{11} & 0 & \cdots & 0 \\ 0 & d_{22} & \cdots & 0 \\ 0 & \cdots & \ddots & 0 \\ 0 & \cdots & & d_{nn} \end{bmatrix} \longrightarrow x_i = \frac{b_i}{d_{ii}} \quad i = 1, 2, \dots, n$$

The cost in terms of number of operations is negligible, given by  $n$  division.

## Lower Triangular matrices

Triangular matrices are also easy to handle. If  $A = L$  is lower triangular, the system can be solved “forward”:

$$L = \begin{bmatrix} l_{11} & 0 & \cdots & 0 \\ l_{21} & l_{22} & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ l_{n1} & l_{n2} & \cdots & l_{nn} \end{bmatrix} \longrightarrow \begin{cases} x_1 = \frac{b_1}{l_{11}} \\ x_2 = \frac{b_2 - l_{21}x_1}{l_{22}} \\ \vdots \\ x_n = \frac{b_n - \sum_{j=1}^{n-1} l_{nj}x_j}{l_{nn}} \end{cases}$$

Counting the operations: for  $x_1$  we have 1 product and 0 sums; for  $x_2$  2 products and 1 sum, ..., and for  $x_n$   $n$  products and  $n - 1$  sums, that is,  $1 + 2 + \cdots + n = \frac{n(n+1)}{2}$  products, plus  $1 + 2 + \cdots + n - 1 = \frac{(n-1)n}{2}$  sums, for a **total number of operations**  $= n^2$ .

# Forward Substitution

What follows is an algorithm, for solving  $L\underline{x} = \underline{b}$ , with  $L$  lower triangular.

## forward substitution

Input:  $L \in \mathbb{R}^{n \times n}$ , lower t., and  $b \in \mathbb{R}^n$

**for**  $i = 1, \dots, n$

**for**  $j = 1, \dots, i - 1$

$$b_i = b_i - l_{i,j}b_j$$

**end**

$$b_i = \frac{b_i}{l_{ii}}$$

**end**

the solution is in  $\underline{b}$ , i.e.,  $\underline{x} \leftarrow \underline{b}$

## Homework

Write the above algorithm in MATLAB

# A special case of lower triangular matrix (useful later...)

Assume  $L$  has only 1 on the diagonal, e.g.,  $l_{ii} = 1$ :

$$L = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ l_{21} & 1 & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ l_{n1} & l_{n2} & \cdots & 1 \end{bmatrix} \rightarrow \begin{cases} x_1 = b_1 \\ x_2 = b_2 - l_{21}x_1 \\ \vdots \\ x_n = b_n - \sum_{j=1}^{n-1} l_{nj}x_j \end{cases}$$

Then the following two algorithms, for solving  $L\underline{x} = \underline{b}$  are equivalent:

forward substitution as above

Input:  $L \in \mathbb{R}^{n \times n}$ , lower t., and  $b \in \mathbb{R}^n$

**for**  $i = 1, \dots, n$

**for**  $j = 1, \dots, i - 1$

$b_i = b_i - l_{i,j}b_j$

**end**

**end**

the solution is in  $\underline{b}$ , i.e.,  $\underline{x} \leftarrow \underline{b}$

equivalent to forward substitution

Input:  $L \in \mathbb{R}^{n \times n}$ , lower t., and  $b \in \mathbb{R}^n$

**for**  $k = 1, \dots, n - 1$

**for**  $i = k + 1, \dots, n$

$b_i = b_i - l_{i,k}b_k$

**end**

**end**

the solution is in  $\underline{b}$ , i.e.,  $\underline{x} \leftarrow \underline{b}$

# Upper Triangular matrices

Upper triangular systems (if  $A = U$  is upper triangular) are also easy to deal with, and can be solved “backward” with the same costs as lower triangular systems:

$$U = \begin{bmatrix} u_{11} & u_{12} & \cdots & u_{1n} \\ 0 & u_{22} & \cdots & u_{2n} \\ \vdots & \vdots & & \vdots \\ 0 & & \cdots & u_{nn} \end{bmatrix} \longrightarrow \begin{cases} x_n = \frac{b_n}{u_{nn}} \\ x_{n-1} = \frac{b_{n-1} - u_{n-1n}x_n}{u_{n-1n-1}} \\ \vdots \\ x_1 = \frac{b_1 - \sum_{j=2}^n u_{1j}x_j}{u_{11}} \end{cases}$$

# Backward Substitution

What follows is an algorithm, for solving  $U\underline{x} = \underline{b}$ , with  $U$  upper triangular.

## backward substitution

Input:  $L \in \mathbb{R}^{n \times n}$ , upper t., and  
 $b \in \mathbb{R}^n$

**for**  $i = n, \dots, 1$

**for**  $j = i + 1, \dots, n$

$$b_i = b_i - u_{i,j}b_j$$

**end**

$$b_i = \frac{b_i}{u_{ii}}$$

**end**

the solution is in  $\underline{b}$ , i.e.,  $\underline{x} \leftarrow \underline{b}$

## Homework

Write the above algorithm in MATLAB

## Costs for solving triangular systems

We saw that for solving the system we perform  $n^2$  operations.

To have an idea of the time necessary to solve a triangular system, suppose that the number of equations is  $n = 100.000$ , and the computer performance is a TERAFL0P =  $10^{12}$  FLOPS (floating-point operations per second); the time in seconds is given by

$$t = \frac{\#operations}{\#flops} = \frac{10^{10}ops}{10^{12}flops} = \frac{1}{100}sec.$$

(pretty quick)

The next target of High Performance Computing is EXAFLOP  
=  $10^{18}$  FLOPS