# Validation for scientific computations Error analysis

#### **Cours de recherche master informatique**

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### **References for today's lecture**

#### **Condition number:**

- Ph. Langlois: HDR (LIP, 2001) and chapter in A. Barraud et al.: Outils d'analyse numérique pour l'automatique, Hermes, 2002
- N.H. Higham: Accuracy and stability of numerical algorithms, SIAM, 1996 (1st edition)
- P. Lascaux and R. Théodor: Analyse numérique matricielle appliquée à l'art de l'ingénieur, Masson, 1993 (2nd edition)

#### Forward and backward analyses:

- Ph. Langlois: HDR (LIP, 2001) and chapter in A. Barraud et al.: Outils d'analyse numérique pour l'automatique, Hermes, 2002
- N.H. Higham: Accuracy and stability of numerical algorithms, SIAM, 1996 (1st edition)
- A. Neumaier: Introduction to numerical analysis, Cambridge University Press, 2001

### Definition of well-posed problem and ill-posed problem

**Problem**  $\mathcal{P}$ :

compute

$$y = f(x)$$

where x: input, y: output or result. It can also be given implicitely:

$$F(x, y) = 0 \text{ or } x = F(y).$$

# Definition of well-posed problem and ill-posed problem

#### Well-posed problem $\mathcal{P}$ :

y = f(x) exists, is unique and depends continuously of x. If the problem is given as x = F(y), this means that  $F^{-1}(x)$  exists, is unique and  $F^{-1}$  is continuous.

If the problem is not well-posed, then it is **ill-posed**.

In what follows, we assume that every problem is well-posed.

#### **Relative error, absolute error**

If  $\in \mathbb{R}$ ,  $\hat{x}$  is an approximation of x and  $\Delta x = \hat{x} - x$ absolute error on x:  $|\Delta x| = |\hat{x} - x|$ relative error on x, if  $x \neq 0$ :  $|\frac{\Delta x}{x}| = |\frac{\hat{x} - x}{x}|$ 

If 
$$x \in \mathbb{R}^n$$
,  
absolute error on  $x$ :  $\|\Delta x\|_a = \|\hat{x} - x\|$   
relative error on  $x$ , if  $x \neq 0$ :

• normwise: 
$$\|\Delta x\|_r = \frac{\|\Delta x\|}{\|x\|} = \frac{\|\hat{x} - x\|}{\|x\|}$$

• componentwise: 
$$|\Delta x||_r = \max_{1 \le i \le n} \left| \frac{\Delta x_i}{x_i} \right|$$

#### **Definition of condition number**

In french: conditionnement ou nombre de condition. Problem  $\mathcal{P}$ : compute y = f(x) with

If x and y do not vanish,

$$\kappa(\mathcal{P}, x) = \lim_{\delta \to 0} \sup_{\substack{\|\Delta x\|_D \\ \|x\|_D} \le \delta} \frac{\|\Delta y\|_R / \|y\|_R}{\|\Delta x\|_D / \|x\|_D}$$

#### **Definition of condition number**

Problem  $\mathcal{P}$ : compute y = f(x) with

If x or y vanish, one uses absolute errors instead of relative errors and one gets an absolute condition number:

$$\kappa(\mathcal{P}, x) = \lim_{\delta \to 0} \sup_{\|\Delta x\|_D \le \delta} \frac{\|\Delta y\|_R}{\|\Delta x\|_D}$$

### Interpretation of condition number

The condition number is the factor of amplification (or reduction) of the error:



# Well-conditioned, ill-conditioned problem

A problem is well-conditioned if its condition number is small, i.e., close to 1,

it is ill-conditioned if its condition number is large, i.e.,  $\gg 1$ .

It is a qualitative notion. . .

#### **Condition number:** subtraction $y = x_1 - x_2$

Let us consider  $\hat{x_1} = x_1(1 + \delta_1) \ \hat{x_2} = x_2(1 + \delta_2)$ , and  $\hat{y} = y + \Delta y = x_1(1 + \delta_1) - x_2(1 + \delta_2)$ , thus

$$\|\Delta y\|_{r} = \frac{|\Delta y|}{|y|} = \frac{|x_{1}\delta_{1} - x_{2}\delta_{2}|}{|x_{1} - x_{2}|} \le \frac{|x_{1}| + |x_{2}|}{|x_{1} - x_{2}|} \max(|\delta_{1}|, |\delta_{2}|).$$

If we use the componentwise relative error,

$$\kappa(-, (x_1, x_2)) = \lim_{\delta \to 0} \sup_{x: \|\Delta x\|_r \le \delta} \frac{|\Delta y|/|y|}{\|\Delta x\|_r} \le \frac{|x_1| + |x_2|}{|x_1 - x_2|}$$

and there is an equality, when  $x_1 \cdot x_2 \ge 0$  and when  $\delta_1 = -\delta_2$ .

**Condition number: subtraction**  $y = x_1 - x_2$ 

$$\kappa(-, (x_1, x_2)) = \frac{|x_1| + |x_2|}{|x_1 - x_2|}$$

This condition number can be arbitrarily large when  $x_1$  and  $x_2$  are close, i.e., when  $|x_1 - x_2| \ll |x_1| + |x_2|$ .

This means that, if  $x_1$  and  $x_2$  are computed results with some uncertainty, this uncertainty is magnified by this subtraction: catastrophic cancellation.

### **Condition number:** subtraction $y = x_1 - x_2$

Cancellation is not always a bad thing:

- no problem when  $x_1$  and  $x_2$  are error-free, cf. divided difference schemes;
- cancellation may be a symptom of intrinsic ill-conditioning and may therefore be unavoidable;
- the result may be harmless, for instance in  $(x_1 x_2) + x_3$  when  $x_3 \gg x_1 x_2$ .

# **Condition number:** division $y = x_1/x_2$

Let us consider  $\hat{x_1} = x_1(1 + \delta_1) \ \hat{x_2} = x_2(1 + \delta_2)$ , and  $\hat{y} = y + \Delta y = \frac{x_1(1+\delta_1)}{x_2(1+\delta_2)}$ , i.e.,  $\Delta y = \frac{x_1}{x_2} \cdot \frac{\delta_1 - \delta_2}{1+\delta_2}$ , thus

$$\|\Delta y\|_r = \frac{|\Delta y|}{|y|} = \frac{|\delta_1 - \delta_2|}{|1 + \delta_2|} \le \frac{2\max(|\delta_1|, |\delta_2|)}{1 - |\delta_2|}.$$

If we use the componentwise relative error,

$$\kappa(-, (x_1, x_2)) = \lim_{\delta \to 0} \sup_{x: \|\Delta x\|_r \le \delta} \frac{|\Delta y|/|y|}{\|\Delta x\|_r} \le 2$$

i.e., the division is always well-conditioned.

### **Condition number:** evaluation of p(x)

Let  $p(x) = \sum_{i=0}^{n} a_i x^i$  a polynomial.

Let us evaluate  $\hat{p}$  in x, where  $\hat{p}(x) = \sum_{i=0}^{n} (a_i + \Delta a_i) x^i$ :

$$\Delta y = \hat{p}(x) - p(x) = \sum_{i=0}^{n} \Delta a_i x^i$$

and thus

$$\frac{|\Delta y|}{|y|} = \frac{|\sum_{i=0}^{n} \Delta a_i x^i|}{|\sum_{i=0}^{n} a_i x^i|} \le \frac{\sum_{i=0}^{n} |\Delta a_i| \cdot |x|^i}{|\sum_{i=0}^{n} a_i x^i|} \le \|\Delta a\|_r \frac{\sum_{i=0}^{n} |a_i| \cdot |x|^i}{|\sum_{i=0}^{n} a_i x^i|}$$

and there is an equality if  $\Delta a_i = sign(a_i x^i) . \|\Delta a\|_r$ .

# **Condition number: evaluation of** p(x)

This implies

$$\kappa(p(x), p) = \frac{\sum_{i=0}^{n} |a_i| \cdot |x|^i}{|p(x)|}.$$

In other words, the evaluation of a polynomial p at x can be arbitrarily ill-conditioned when x is close to a root of p.

### **Condition number: solving the linear system** Ax = b

$$A = \begin{pmatrix} 10 & 7 & 8 & 7 \\ 7 & 5 & 6 & 5 \\ 8 & 6 & 10 & 9 \\ 7 & 5 & 9 & 10 \end{pmatrix} \quad b = \begin{pmatrix} 32 \\ 23 \\ 33 \\ 31 \end{pmatrix}$$
solution of  $Ax = b$ :  $x = (1 \ 1 \ 1 \ 1)^t$ .  
If the system is perturbed in

$$A + \Delta A = \begin{pmatrix} 10 & 7 & 8.1 & 7.2 \\ 7.08 & 5.04 & 6 & 5 \\ 8 & 5.98 & 9.89 & 9 \\ 6.99 & 4.99 & 9 & 9.98 \end{pmatrix},$$

the solution of  $(A + \Delta A)x = b$  is  $(-81 \ 137 \ -34 \ 22)^t$ .

### **Condition number: solving the linear system** Ax = b

If x is the solution of Ax = b and  $\hat{x}$  is the solution of  $Ax = b + \Delta b$ ,

$$A\Delta x = \Delta b$$
  

$$\Rightarrow \Delta x = A^{-1}\Delta b$$
  

$$\Rightarrow \|\Delta x\| = \|A^{-1}\Delta b\| \leq \|A^{-1}\| \|\Delta b\|$$

We also have 
$$Ax = b$$
  
 $\Rightarrow ||Ax|| = ||b||$   
 $\Rightarrow ||A|| \cdot ||x|| \ge ||b||$   
 $\Rightarrow \frac{||A||}{||b||} \ge \frac{1}{||x||}$   
and thus  $\frac{||\Delta x||}{||x||} \le ||A|| \cdot ||A^{-1}|| \frac{||\Delta b||}{||b||}$ 

#### **Condition number: solving the linear system** Ax = b

 $\kappa(A) = \|A\| . \|A^{-1}\|$ 

It also holds that, if  $x + \Delta x$  is the solution of  $(A + \Delta A)x = b$ ,

$$\frac{\|\Delta x\|}{\|x\|} \le \kappa(A) \cdot \frac{\|\Delta A\|}{\|A\|}$$

and if  $x + \Delta x$  is the solution of  $(A + \Delta A)x = b + \Delta b$ ,

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

#### **Condition number: link with a first-order analysis**

Problem  $\mathcal{P}$ : compute y = f(x).

and thus  

$$\begin{aligned} y + \Delta y &= f(x + \Delta x) \simeq f(x) + f'(x) . \Delta x + \mathcal{O}(\Delta x^2) \\ \frac{|\Delta y|}{|y|} \simeq \frac{|f'(x)| . |\Delta x|}{|f(x)|} = \frac{|f'(x)| . |x|}{|f(x)|} . \frac{|\Delta x|}{|x|} \end{aligned}$$
i.e.,  

$$\kappa(\mathcal{P}, x) \simeq \left| \frac{f'(x) . x}{f(x)} \right|.$$

### **Condition number: link with a first-order analysis**

When x and y are not scalars, but vectors: the absolute condition number is

$$\kappa_a(\mathcal{P}, x) = \|f'(x)\|$$

it can also be interpreted as the Lipschitz constant for f', and the relative condition number is

$$\kappa(\mathcal{P}, x) = \frac{\|f'(x)\| \cdot \|x\|}{\|f(x)\|}.$$

# **Condition number: remarks**

- 1. the choice of the norms  $\|.\|_D$  and  $\|.\|_R$  modifies the condition number; the choice of the authorized perturbations  $\Delta x$  as well: one may preserve the structure of the problem and restrict  $\Delta x$  (sparsity of a matrix...)
- 2. quite often, the condition number measures the inverse of the distance to the nearest singular problem (or ill-posed problem)
- 3. the condition number depends only on the problem  $\ensuremath{\mathcal{P}}$  and not on the algorithm that solves it
- 4. only small perturbations are considered here, for large perturbations consider interval arithmetic.