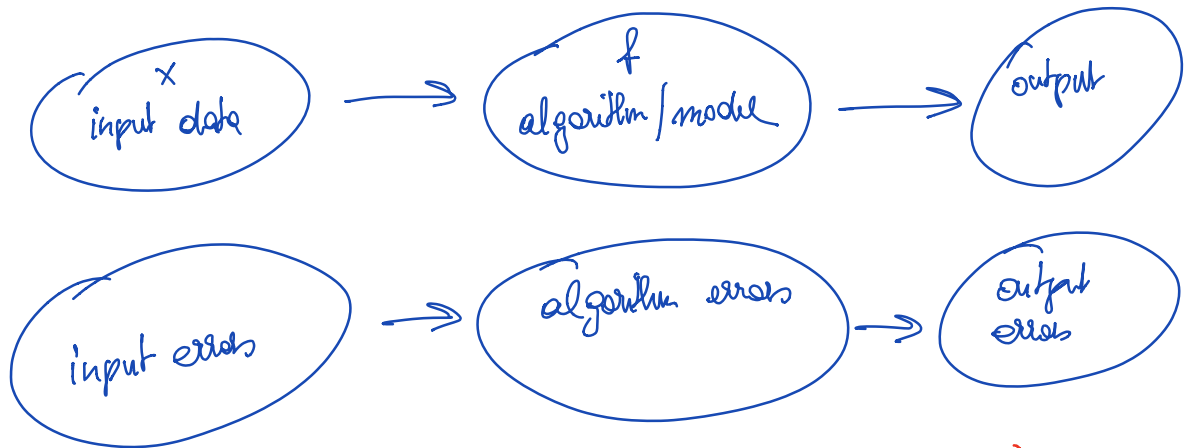


Error analysis (§ 2 in Deuffhard/Hohmann)



Sources of errors:

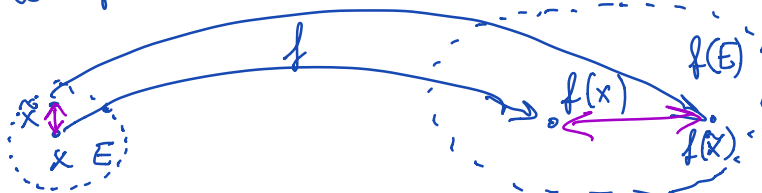
- model errors (improve model)
- data/input errors (improve measurement devices)

Comput. errors {

- truncation/discretization errors (i.e. approx. by finite steps)
- rounding errors / errors in the algorithm due to finite precision

Conditioning of problems

How do perturbations/errors in input affect the output.
This concept is independent from the algorithm.



Examples for f : 1) $f: d \in \mathbb{R} \rightarrow$ roots of $y^2 = d$
(input has two outputs)

2.) $x, b \in \mathbb{R}^n, A \in \mathbb{R}^{n \times n}$

• $f: x \mapsto Ax$ multiplication with a matrix

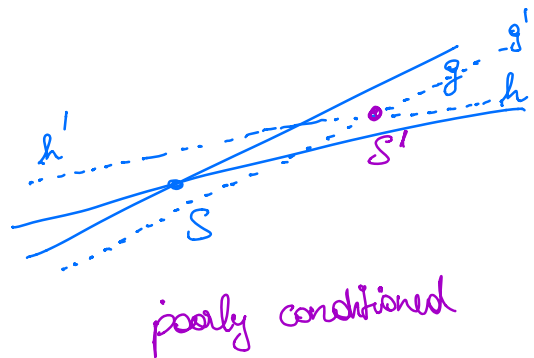
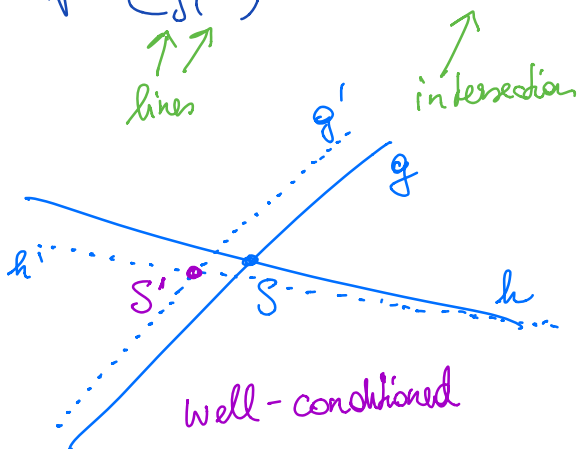
• $f: b \mapsto A^{-1}b = x$ rhs to solution of $Ax = b$

• $f: A \mapsto Ax$ perturbations w.r. to A .

3.) x, g functions $x, g: [0, 1] \rightarrow \mathbb{R}$

$f: g$ to the solution of $-x'' = g, x(0) = x(1) = 0$

4.) $f: (g, h) \rightarrow S$



Definition: $f: U \subset \mathbb{R}^n \rightarrow \mathbb{R}^m, x \in U, U \text{ open}, m, n \geq 1$

$\delta > 0$

$\|\tilde{x} - x\| \leq \delta$

absolute neighborhood around x

$\|\tilde{x} - x\| \leq \delta \|x\|$

relative neighborhood around x

Absolute condition number of f is the smallest

$K_{abs} \geq 0$ s.t.

$\|f(\tilde{x}) - f(x)\| \leq K_{abs} \|\tilde{x} - x\|$ for $\tilde{x} \rightarrow x$

Relative condition number of f at x is the smallest $K_{rel} \geq 0$ s.t.

$$\frac{\|f(\tilde{x}) - f(x)\|}{\|f(x)\|} \leq K_{rel} \frac{\|\tilde{x} - x\|}{\|x\|} \text{ for } \tilde{x} \rightarrow x$$

well-conditioned if K_{abs} / K_{rel} small

poorly-conditioned if K_{abs} / K_{rel} is large or infinity

Relation between condition numbers and derivatives:

f is differentiable: $f(x) = y$ original problem

$f(x + \delta x) = y + \delta y$ perturbed problem

$$\delta y = f(x + \delta x) - f(x) \stackrel{\text{Taylor}}{=} f'(x) \delta x + o(\|\delta x\|)$$

use
norms \Rightarrow

$$\frac{\|\delta y\|}{\|\delta x\|} \leq \|f'(x)\|$$

Jacobian, in $\mathbb{R}^{n \times m}$ if $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$

$$\rightarrow K_{abs} = \|f'(x)\|$$

$$K_{rel} = \|f'(x)\| \frac{\|x\|}{\|f(x)\|}$$

Examples:

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}$$

$$\begin{pmatrix} a \\ b \end{pmatrix} \mapsto a + b$$

$$f' \begin{pmatrix} a \\ b \end{pmatrix} = (f_{,a} \quad f_{,b}) \in \mathbb{R}^{1 \times 2}$$

$$= (1 \quad 1)$$

We use the 1-norm in \mathbb{R}^2 , i.e. $\| \begin{pmatrix} a \\ b \end{pmatrix} \| = |a| + |b|$

$$\|f'(a/b)\| = \|(1, 1)\| \quad \text{matrix-norm induced by 1-norm}$$

$$\Rightarrow \kappa_{\text{abs}} = 1$$

$$\kappa_{\text{rel}} = 1 \cdot \frac{|a|+|b|}{|a+b|}$$

Addition of numbers with same sign is well-conditioned
 ——— " ——— with different sign can be poorly conditioned if the numbers are close in absolute value.

As a consequence, roundoff can have bad effects, e.g.:

$$a = 0.12345 * \dots \quad * \text{ means errors}$$

$$b = 0.12356 * \dots$$

$$a-b = 0.00011 * \dots \quad \text{So } a, b \text{ were accurate for 5 digits, but } a-b \text{ is only accurate to 2 digits.}$$

Warning: roundoff errors can have significant negative effect.

Example 2: $f: x \mapsto Ax$

$$f'(x) = A, \quad \kappa_{\text{abs}} = \|A\|$$

$$\kappa_{\text{rel}} = \|f'(x)\| \cdot \frac{\|x\|}{\|f(x)\|} = \|A\| \cdot \frac{\|x\|}{\|Ax\|}$$

Example 3: $f: b \mapsto A^{-1}b = x$ Ax = b

$$K_{\text{abs}} = \|A^{-1}\|$$

$$K_{\text{rel}} = \|A^{-1}\| \cdot \frac{\|b\|}{\|A^{-1}b\|} = \|A^{-1}\| \frac{\|Ax\|}{\|x\|} \leq$$

$$\leq \|A^{-1}\| \frac{\|A\| \|x\|}{\|x\|} = \|A\| \|A^{-1}\|$$

K_A condition number of A

Example:

$$f: A \rightarrow A^{-1}b = x$$

$\mathbb{R}^{n \times n}$, invertible

Derivative of f ,

$$F: A \rightarrow A^{-1}$$

$\mathbb{R}^{n \times n} \rightarrow \mathbb{R}^{n \times n}$

$C \in \mathbb{R}^{n \times n}$

$$F'(A)C = \underline{\underline{-A^{-1}CA^{-1}}}$$