

Error Analysis

If A is an approximation of an exact number E then

$$\begin{aligned}\Delta &= E - A && \text{true error} \\ |\Delta| &= |E - A| && \text{absolute error} \\ \epsilon &= \frac{|\Delta|}{|E|} && \text{relative error.}\end{aligned}$$

→ Sources of errors.

- 1) Input errors: Errors in data given to software
- 2) Procedural errors: Errors arising from approximation scheme of algorithm.
- 3) Propagation error: Results from amplification of input error.
- 4) Computational error: Results from finite representation of real numbers.

Ill-conditioned: An algorithm with significant propagation error.

Unstable algorithm: Significant computational error.

→ Estimating propagation error

Consider the computation

$$y = f(x_1, x_2, x_3, \dots, x_n)$$

and let e_1, e_2, \dots, e_n be the input errors relative errors of x_1, x_2, \dots, x_n . The error e of y is estimated by:

$$e = \frac{\Delta y}{y} = \sum_{a=1}^n \frac{x_a}{f(\vec{x})} \frac{\partial f(\vec{x})}{\partial x_a} e_a$$

- $f(x_1, x_2) = x_1 x_2 \Rightarrow e = e_1 + e_2$

- $f(x_1, x_2) = x_1/x_2 \Rightarrow e = e_1 - e_2$

- $f(x_1, x_2) = x_1 + x_2 \Rightarrow e = \frac{x_1 e_1 + x_2 e_2}{x_1 + x_2}$

For $x_1 \geq 0$ and $x_2 \geq 0 \Rightarrow e \leq e_1 + e_2$

- $f(x_1, x_2) = x_1 - x_2 \Rightarrow e = \frac{x_1 e_1 - x_2 e_2}{x_1 - x_2}$

For $x_1 \approx x_2 \rightarrow e$ is amplified by denominator!

} Well conditioned

Thus it is not safe to subtract numbers that are nearly equal to each other.

↳ Eliminate/reduce subtractions.

• $f(x_i) = x_i^p \Rightarrow e = p e_1$

Thus, roots ($0 < p \leq 1$) are well conditioned but powers with large $p \gg 1$ are ill-conditioned

example 1: Quadratic formula $ax^2 + bx + c = 0$
Assume $b > 0$.

For numerical accuracy rewrite

$$x_{1,2} = \frac{-b \pm \sqrt{\Delta}}{2a} \quad \text{as: } \begin{cases} x_1 = \frac{-b - \sqrt{\Delta}}{2a} \\ x_2 = \frac{2ac}{-b - \sqrt{\Delta}} \end{cases}$$

example 2: Evaluation of $p(x) = a_n x^n + \dots + a_1 x + a_0$.

We use the Horner scheme.

$$p(x) = (\dots ((a_n x + a_{n-1})x + a_{n-2})x + \dots)x + a_0$$

which can be written recursively as:

$$\begin{cases} b_n = a_n \\ b_k = b_{k+1}x + a_k, \text{ for } k = n-1, \dots, 0 \end{cases}$$

↓

$$p(x) = b_0.$$

The Horner scheme alternates addition with multiplication.

→ Understanding numerical stability.

Consider the iteration

$$\begin{cases} x_0 = 1, \quad x_1 = 1/3 \\ x_{n+1} = \frac{13}{3}x_n - \frac{4}{3}x_{n-1} \end{cases} \quad (1)$$

By induction, we can easily prove that

$$x_n = (1/3)^n.$$

A numerical evaluation of this iteration diverges. The origin of this problem is the "unstable" mode in the general solution of (1)

$$x_n = A(1/3)^n + B \cdot (4)^n.$$