

Scientific Computing

(1)

Designing + Analyzing algorithms for solving
computational problems involving continuous quantities

⇒ making approximation

⇒ errors.

Example : surface area of Earth

Radius : 6371 km.

$$6371 \times 6371$$

π : 3.14

$$\text{Surface area of sphere} = 4\pi r^2$$

$$= 4 \times 3.14 \times 6371 \times 6371$$

$$= 509.8 \times 10^6 \text{ km}^2.$$

$$\text{Actual} - 510.1 \times 10^6 \text{ km}^2$$

What are the sources of errors?

Earth not a sphere

— modelling
Error

Radius measured empirically

— Empirical
measure
Truncation.

Values of π truncated

} Truncation
Error

∴ radius truncated.

} Before
Computation

} During
Computation

Example of a computation problem.

use mathematical model to understand the universe.

(2)

- ① Developed a mathematical model. — $4\pi r^2$.
- ② Developed an algorithm to solve the equation numerically
- ③ Implemented the algorithm
- ④ Run the algo. and interpret the result.

The focus of this course

We're going to focus on well-posed problem

- ① a solution exists.
- ② solution is unique
- ③ result depends continuously on the problem data

Opposite: ill-posed problems

~~map~~ midterm avg for the class → midterm grade for each student.

Image denoising: noisy \rightarrow clean img

finding the root of $x^2 + 1$

Typical Computation Problem:

Compute $f: \mathbb{R} \rightarrow \mathbb{R}$ (or $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$)

Quantify error

$$\begin{aligned}\text{Absolute Error} &= \text{Approx Value} - \text{True Value} \\ & (509.8 \times 10^6 \text{ km}^2 - 510.1 \times 10^6 \text{ km}^2) \\ & = -0.3 \times 10^6 \text{ km}^2\end{aligned}$$

$$\begin{aligned}\text{Relative Errr} &= \frac{\text{Absolute Errr}}{\text{True Value}} \\ & \left(\frac{-0.3 \times 10^6 \text{ km}^2}{510.1 \times 10^6 \text{ km}^2} \right) \\ & = -0.06\%\end{aligned}$$

Data Errr vs. Computation Errr

x — true input

\hat{x} — approx input

\hat{f} — approx function

want: $f(x)$

①

Total Errr = $\hat{f}(\hat{x}) - f(x)$

$$= (\hat{f}(\hat{x}) - f(\hat{x})) + (f(\hat{x}) - f(x))$$

① computational
errr

② propagated data
errr

④ Computational error:

- Truncation Error (true result) - (result of using exact arithmetic)
- Rounding Error (result using exact arithmetic) - (result from finite precision arithmetic)

eg // approximate $\cos(x) = 1 - \frac{x^2}{2}$

Truncation Error $\sum_{k=2}^{\infty} \frac{(-1)^k x^{2k}}{(2k)!}$

Forward and Backward Error

Want to compute $y = f(x)$ $f: \mathbb{R} \rightarrow \mathbb{R}$.

But. \hat{f} approximate $\Rightarrow \hat{y} = \hat{f}(x)$ (abs. error. $\hat{y} - y$, relative error $\frac{\hat{y} - y}{y}$)

Absolute Forward Error : $\Delta y = \hat{y} - y$.

Problem : difficult to estimate

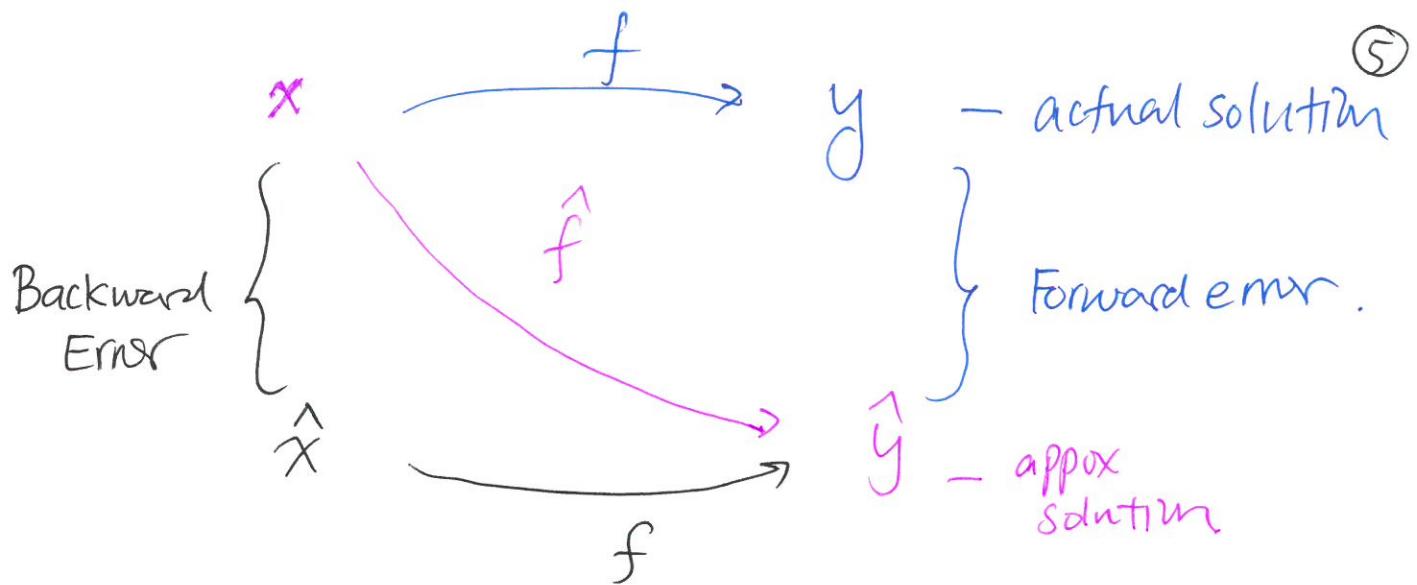
worse-case assumptions give pessimistic bounds.

Backward Error : $\Delta x = \hat{x} - x$. different \hat{x} than last page

$$\hat{y} = f(\hat{x})$$

$$\hat{x} := f^{-1}(\hat{y})$$

Approx. Solutions are "good" if it is the exact Solution to a nearby problem.



Example $f(x) = \cos(x)$

$$\hat{f}(x) = 1 - \frac{x^2}{2}$$

$$\Delta y = \left(1 - \frac{x^2}{2}\right) - \cos(x)$$

$$\Delta x = \arccos(\hat{y}) - x$$

where
 $\hat{y} = \hat{f}(x)$.

Sensitivity — describe a problem.

a problem is:

- insensitive or well-conditioned if

 a relative change in input cause similar relative change in solution.

- sensitive or ill-conditioned

 if relative change in solution is much larger than the ^{change in} input.

$$\begin{aligned}
 \text{Condition number} &= \frac{|\Delta y / y|}{|\Delta x / x|} && \begin{array}{l} \text{relative change in output} \\ \text{relative change in input.} \end{array} \\
 (\text{of a problem}) & & &
 \end{aligned}$$

If $\underline{\text{C.N.}} \gg 1$, the problem is ill-conditioned.
 \hookrightarrow represents an "amplification factor"

$$\left| \frac{\text{Relative Forward Error}}{\text{Error}} \right| = \text{C.N.} \times \left| \frac{\text{Relative Backward Error}}{\text{Error}} \right|.$$

$$\left| \frac{\Delta y}{y} \right| = \text{C.N.} \times \left| \frac{\Delta x}{x} \right|$$

\uparrow
 \approx when C.N. is estimated
 by an upper bound.

For $f: \mathbb{R} \rightarrow \mathbb{R}$. — continuous, differentiable.

$\hat{x} = x + \Delta x$ approximate value,

$$\begin{aligned}
 \Rightarrow \text{Forward Error} &= \frac{f(x + \Delta x) - f(x)}{f(x)} \frac{\Delta x}{\Delta x} \\
 &= \left(\frac{f(x + \Delta x) - f(x)}{\Delta x} \right) \frac{\Delta x}{f(x)} \\
 &\approx \boxed{f'(x)} \frac{\Delta x}{f(x)}
 \end{aligned}$$

$\Delta x \rightarrow 0$

$$C.N = \frac{|\text{Rel. Forward Error}|}{|\text{Rel. Backward Error}|}$$

$$= \left| \frac{f'(x) \frac{\Delta x}{f(x)}}{\frac{\Delta x}{x}} \right|$$

$$= \left| \frac{x f'(x)}{f(x)} \right| =: K_f(x).$$

Stability of Algorithms

An algo. is stable if the result is relatively insensitive to perturbation during computation.

backward error: an algo is stable if result produced is the exact solution of nearby problem.

Accuracy: closeness of computed solution to actual solution:

- Require
 1. problem is well-conditioned
 2. algorithm is stable