

Errors

[start with page
about double-precision
numbers and machine precision!]

Mathematical errors / truncation errors

e.g. from stopping a series expansion

$$u''_i = \frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} + \mathcal{O}(h^2)$$

truncation
error

Round-off errors

Numbers only stored with accuracy \sim machine precision

For double : ~ 15 digits

So almost all numbers stored are approximations :

- True number : a

- Floating-point representation of a : $f(a)$

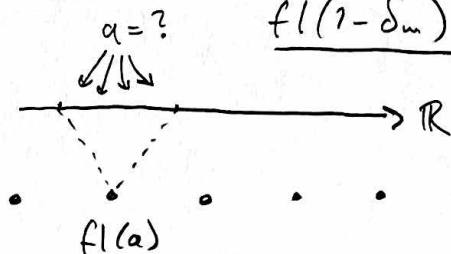
Given a , your $f(a)$ will be in range

$$a(1 - \delta_m) < f(a) < a(1 + \delta_m)$$

where δ_m is machine precision
(here : $\delta_m \sim 10^{-15}$)

Given $f(a)$, all you know is that
the true a is in the range

$$f(a)(1 - \delta_m) < a < f(a)(1 + \delta_m)$$



- Loss of numerical precision
- a.k.a loss of significance
- Typical case : Subtract similar numbers
→ loose significant digits, left with digits affected by round-off

◦ Example :

True : $a = 1.0054321$

Approx : $f1(a) = 1.005$

◦ 4 significant digits

◦ Rel. error in approx : $\left| \frac{a - f1(a)}{a} \right| \approx 4 \times 10^{-4}$

$b = 1.0040001$

$f1(b) = 1.004$

◦ 4 sig. digits

◦ Rel. err : $\left| \frac{b - f1(b)}{b} \right| \approx 1 \times 10^{-7}$

$f1(a)$ and $f1(b)$ are good approx.
to a and b .

Take difference :

True : $a - b = 0.0014320$

Approx : $f1(a) - f1(b) = \underline{0.001}$

◦ Only 1 sign. digit !

◦ Rel error : $\approx 3 \times 10^{-1}$ 30% !

⇒ Loss of precision !

- We are typically interested in relative errors

abs. err : $\Delta = |v_i - u_i|$

rel. err : $\epsilon = \left| \frac{v_i - u_i}{u_i} \right|$ will study this in proj. 7

[Often look at $\log_{10}(\epsilon)$ vs $\log_{10}(h)$]

- Typical case for us

- For "large" step sizes : truncation error dominates
- For tiny step sizes : round-off errors lead to loss of precision \rightarrow garbage

Some optimal stepsize gives smallest overall error

- I will put out a code example for this
- You will study this in proj. 7