

lecture 5: error estimation

learning goals

- estimate error in a numerical integral
- search for and use standard integration methods

to estimate integration error, start with a function expansion around 2 points

$$f(x) = \begin{cases} f(x_{i-1}) + (x-x_{i-1})f'(x_{i-1}) + (x-x_{i-1})^2 f''(x_{i-1}) + \dots \\ f(x_i) + (x-x_i)f'(x_i) + (x-x_i)^2 f''(x_i) + \dots \end{cases}$$

then we can integrate

$$\int_{x_{i-1}}^{x_i} dx f(x) = \begin{cases} \int_0^{\Delta x} du [f(x_{i-1}) + u f'(x_{i-1}) + u^2 f''(x_{i-1}) + \dots] \\ \int_{-\Delta x}^0 dv [f(x_i) + v f'(x_i) + v^2 f''(x_i) + \dots] \end{cases}$$

averaging these gives

$$\int_{x_{i-1}}^{x_i} dx f(x) = \frac{1}{2} \left\{ \Delta x [f(x_{i-1}) + f(x_i)] + \frac{1}{2} \Delta x^2 [f'(x_{i-1}) - f'(x_i)] + \frac{1}{6} \Delta x^3 [f''(x_{i-1}) + f''(x_i)] + \dots \right\}$$

note the minus sign

we can then write the whole integral as a sum over discrete intervals

$$\begin{aligned} \int_a^b dx f(x) &= \sum_i \int_{x_{i-1}}^{x_i} dx f(x) \\ &= \sum_i \Delta x \frac{f(x_{i-1}) + f(x_i)}{2} \leftarrow \text{trapezoid} \\ &\quad + \sum_i \frac{\Delta x^2}{4} [f'(x_{i-1}) - f'(x_i)] \rightarrow = \frac{\Delta x^2}{4} [f'(a) - f'(b)] \\ &\quad + \sum_i \frac{\Delta x^3}{6} \frac{f''(x_{i-1}) + f''(x_i)}{2} \leftarrow \propto \text{trapezoidal integral of } f'' \\ &\quad + \mathcal{O}(\Delta x^4) \end{aligned}$$

integrating the trapezoidal sum for f' and canceling terms, we get error on the trapezoidal integral

$$\epsilon = \frac{\Delta x^2}{12} [f'(a) - f'(b)] + \mathcal{O}(\Delta x^4)$$

note: for other integration rules, need a different expansion, see Newman

$$\text{for Simpson's rule, } \epsilon = \frac{\Delta x^4}{90} [f'''(a) - f'''(b)] + \mathcal{O}(\Delta x^6)$$

theoretical error is great, but what if we have no analytical expression for f ?

start by performing integral w/ N_1 steps, record answer I_1

integrate again with $N_2 = 2N_1$ steps, so $\Delta x_1 = 2\Delta x_2$

then write

$$I = I_1 + c\Delta x_1^2 = I_1 + 4c\Delta x_2^2 = I_2 + \overset{\text{some constant}}{c}\Delta x_2^2$$

$$I_2 - I_1 = 3c\Delta x_2^2, \text{ so } e_2 = c\Delta x_2^2 = \frac{1}{3}(I_2 - I_1)$$

aside

how important is rounding error?

each numerical computation includes a rounding error that is well-approximated by a normal (Gaussian) random variable w/ mean 0 and standard deviation

$$\epsilon = Cx, \text{ w/ } C \sim 10^{-16}$$

w/out going into detail, sum of N normal variables w/ std ϵ has typical size $\propto \sqrt{N}\epsilon$

but trapezoidal rule multiplies by $\Delta x \propto 1/N$, so cumulative error drops like $1/\sqrt{N}$!

*notebook example w/ scipy package