

NUR A lecture 4

In this lecture:

Discretization

(Extended) trapezoid rule

(Extended) Simpson's rule

Higher orders

Open (extended) formulas/midpoint

Romberg integration and errors

Richardson extrapolation

Handling improper integrals

Integrating in log vs linear space

Basics of Gaussian quadrature

Multidimensional integration

Numerical Recipes for Astrophysics A

Lecture 4

Questions about last week?

Integrating functions

NR Ch 4

Integrating functions

- Discretize and evaluate

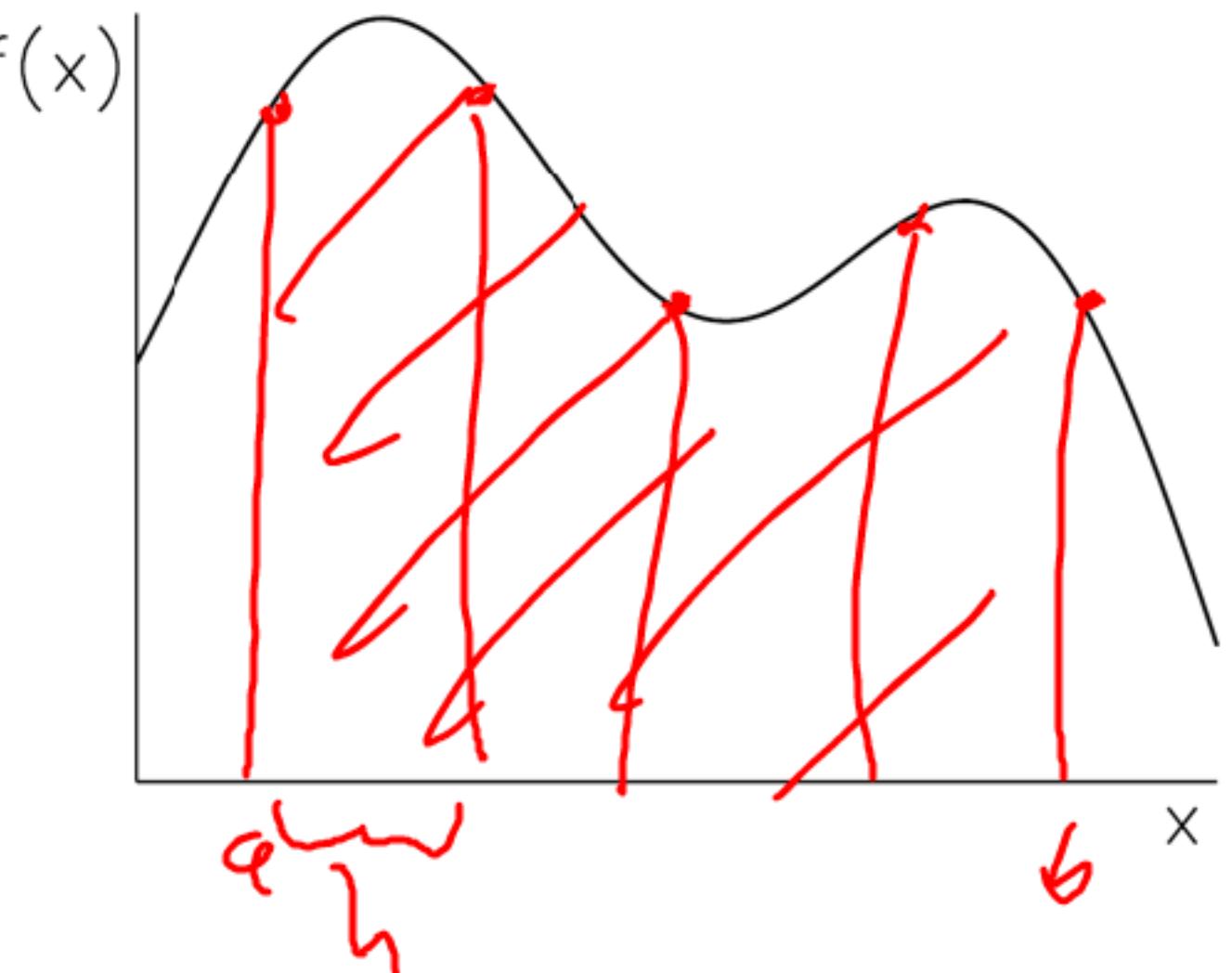
N intervals $\Rightarrow N + 1$ points (x_i, f_i)

If equally spaced: step $h = \frac{b - a}{N}$

$f(x)$ may be expensive: keep N low

Higher N : more accurate, unless N is so high/ h so small that round-off dominates

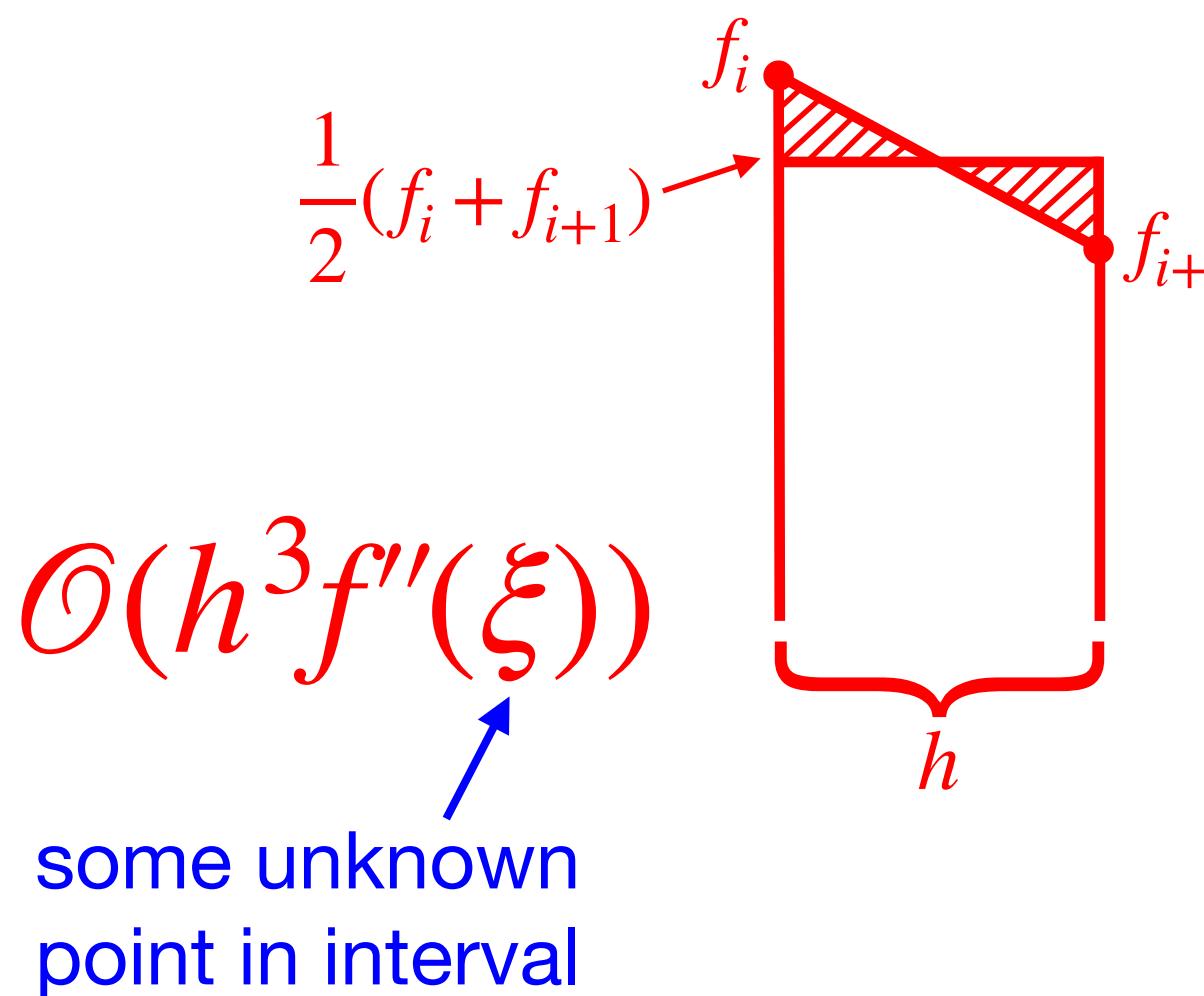
- Undefined $f(x)$ /divergence
- Able to not evaluate $f(x)$ at certain x



Equally spaced abscissae

- Trapezoid rule and extension

$$\int_{x_i}^{x_{i+1}} f(x) dx = \frac{h}{2}(f_i + f_{i+1}) + \mathcal{O}(h^3 f''(\xi))$$

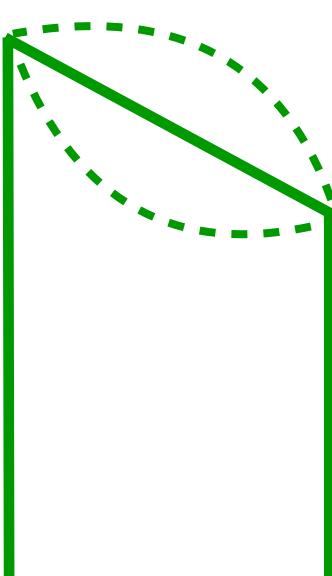


Extended trapezoid:

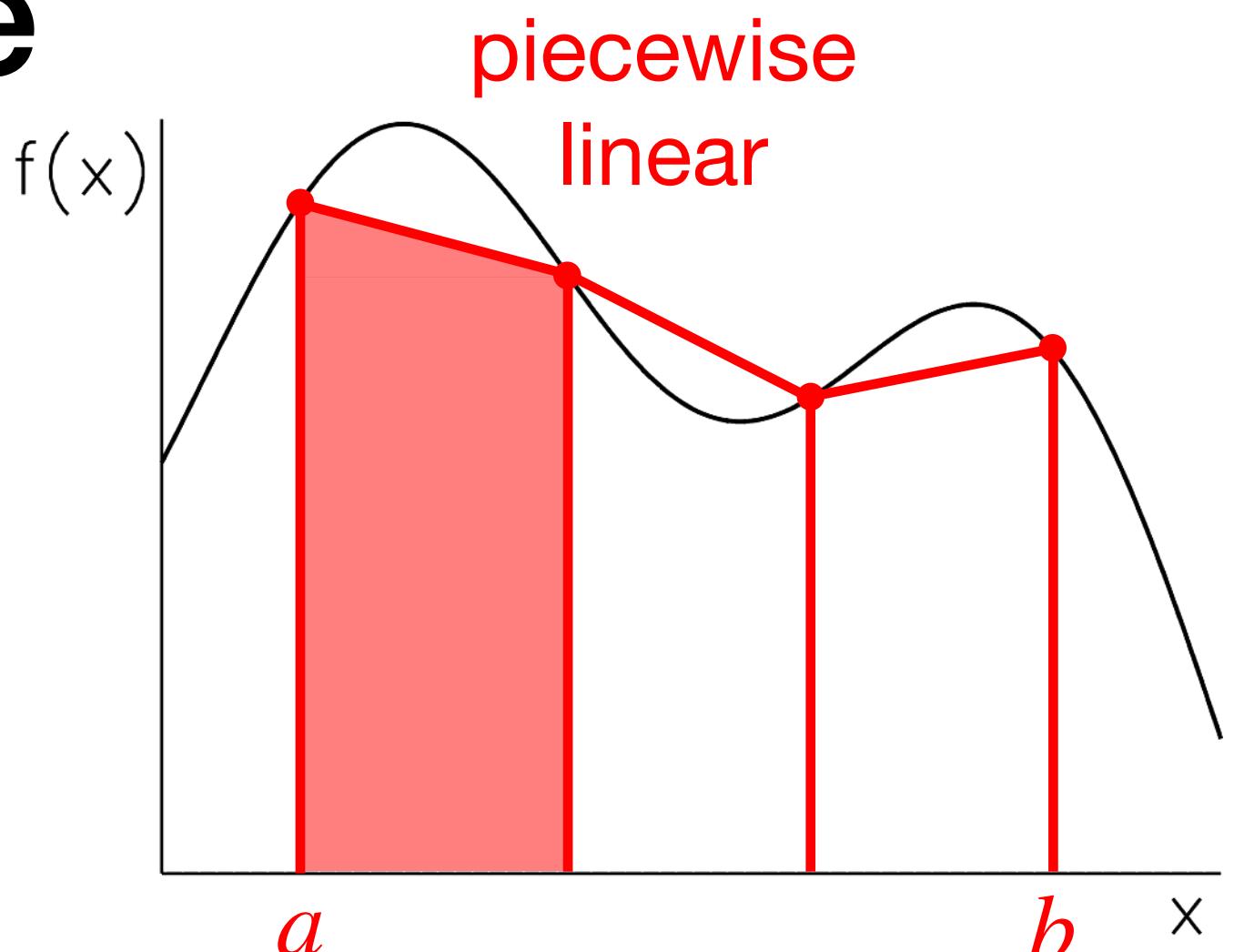
$$\int_a^b f(x) dx = \sum_{i=0}^{N-1} \int_{x_i}^{x_{i+1}} f(x) dx = h \left(\frac{f_0}{2} + \sum_{i=1}^{N-1} f_i + \frac{f_N}{2} \right) + \mathcal{O}\left(\frac{(b-a)^3}{N^2} f''\right)$$

$\left(\frac{f_0}{2} + \frac{f_1}{2} \right) + \left(\frac{f_1}{2} + \frac{f_2}{2} \right) + \dots$

$\propto N h^3$



(Note that trapezoid always underestimates concave-down functions, and always overestimates concave-up ones)



all error terms are even powers of N

$$\left(\frac{1}{N^2}, \frac{1}{N^4}, \frac{1}{N^6}, \dots \right)$$

Equally spaced abscissae

- Higher order: Simpson's rule

$$g(x) = ax^2 + bx + c \Rightarrow G(x) = \frac{1}{3}ax^3 + \frac{1}{2}bx^2 + cx$$

$$\int_{x_0}^{x_2} f(x) dx \approx G(x_2) - G(x_0) = \frac{2}{3}ah^3 + 2ch$$

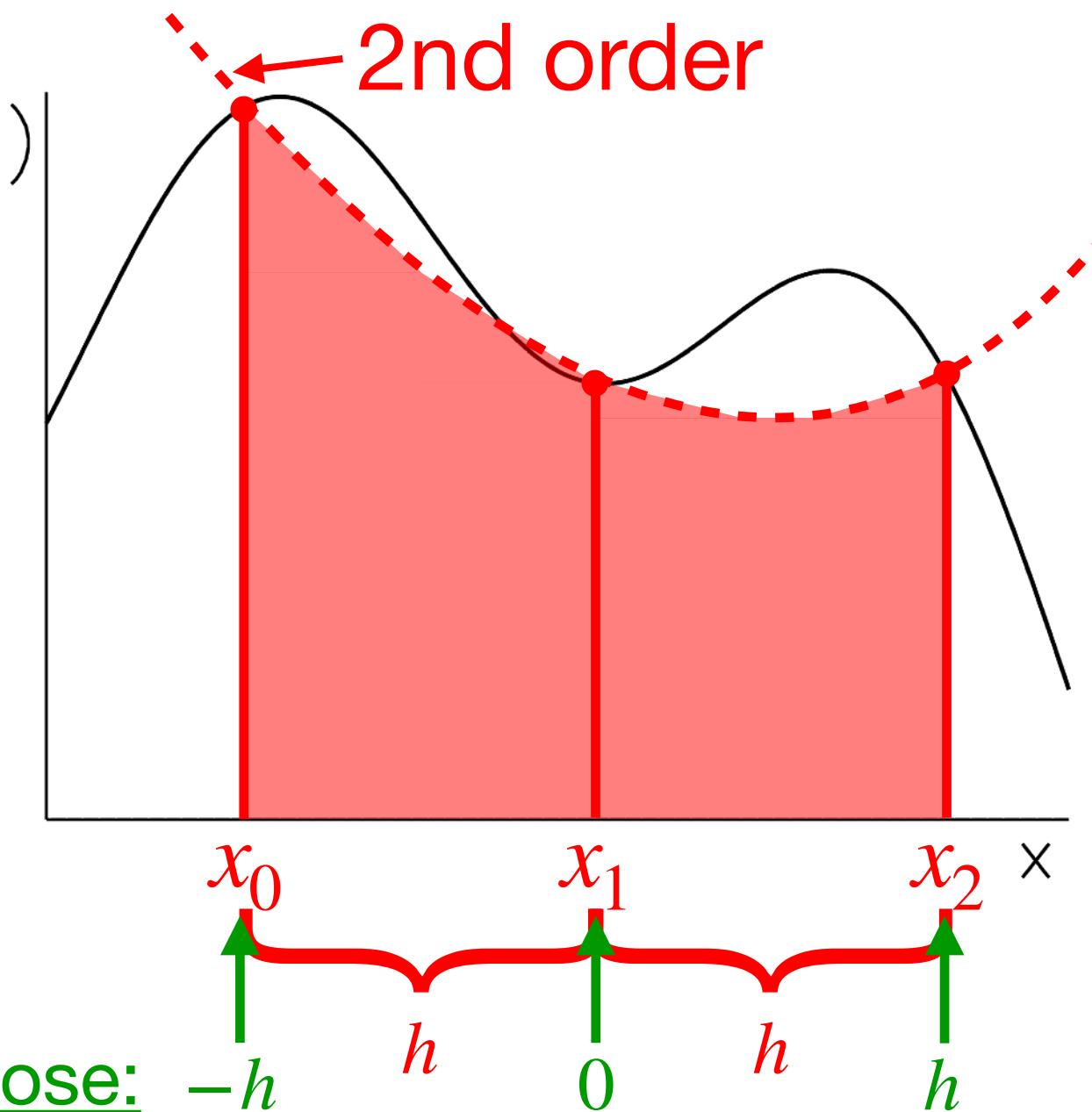
$$(1) \quad g(x_0) = f_0 = ah^2 - bh + c \quad (1) + (3) \Rightarrow a = \frac{f_0 - 2f_1 + f_2}{2h^2}$$

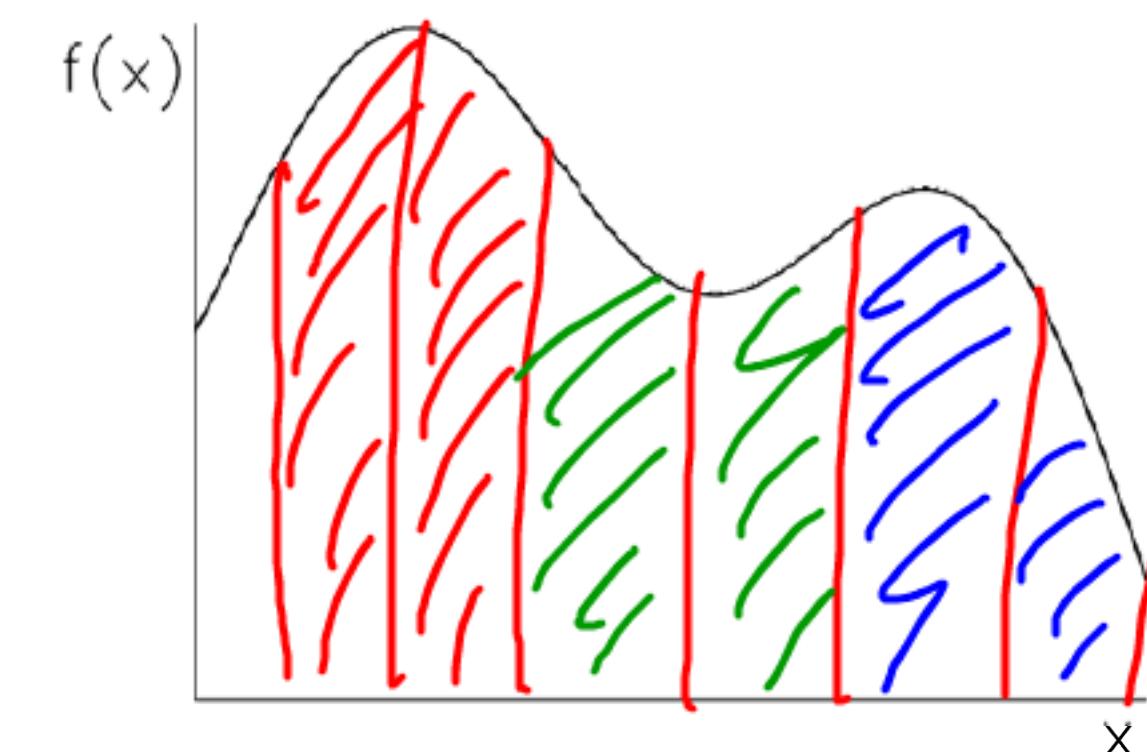
$$(2) \quad g(x_1) = f_1 = c$$

$$(3) \quad g(x_2) = f_2 = ah^2 + bh + c \quad (3) - (1) \Rightarrow b = \frac{f_2 - f_0}{2h} \quad (h^4, h^6, \dots \text{cancel})$$

$$\int_{x_0}^{x_2} f(x) dx = \frac{h}{3}(f_0 + 4f_1 + f_2) + \mathcal{O}(h^5 f^{(4)}) \Rightarrow$$

accurate to
3rd order!
(symmetry)





Equally spaced abscissae

- Extended Simpson's rule

$$\int_{x_0}^{x_N} f(x) dx = \frac{h}{3} (f_0 + 4f_1 + 2f_2 + 4f_3 + \dots + 2f_{N-2} + 4f_{N-1} + f_N) + \mathcal{O}\left(\frac{1}{N^4}\right)$$

$$1 + 4 + \cancel{1} \\ \cancel{1} + 4 + 1$$

$$\left(+ \sim \frac{1}{N^6} + \sim \frac{1}{N^8} + \dots \right)$$

- Higher order still

Simpson's $\frac{3}{8}$ rule:

(still) 3rd order

$$\int_{x_0}^{x_3} f(x) dx = \frac{3h}{8}(f_0 + 3f_1 + 3f_2 + f_3) + \mathcal{O}(h^5 f^{(4)})$$

Boole's rule:

(not “Bode’s”!)

$$\int_{x_0}^{x_4} f(x) dx = \frac{2h}{45}(7f_0 + 32f_1 + 12f_2 + 32f_3 + 7f_4) + \mathcal{O}(h^7 f^{(6)})$$

no cancellation

cancellation
again

Open formulas

- Deriving open formulas

$$\int_{-2h}^{2h} f(x) dx \approx \frac{4}{3}h(2f_1 - f_2 + 2f_3) : \text{Milne's rule}$$

Only extrapolate to the left:

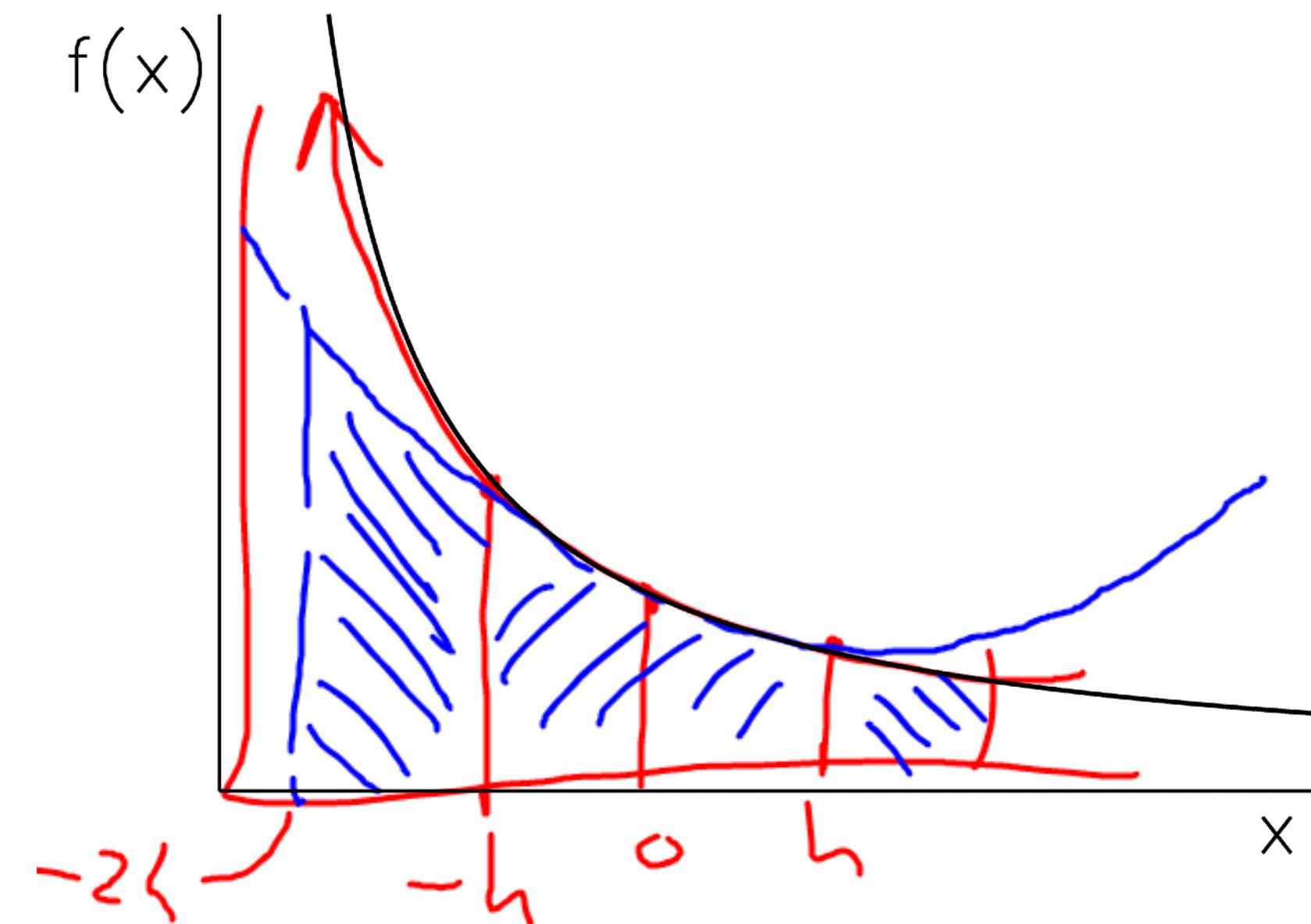
$$\int_{-2h}^h f(x) dx \approx \frac{h}{12}(23f_1 - 16f_2 + 5f_3)$$

- Open extended formulas

only open at endpoints, closed in between

- Semi-open

only open at a or b



Open formulas

- Second-order extended (Simpson's)

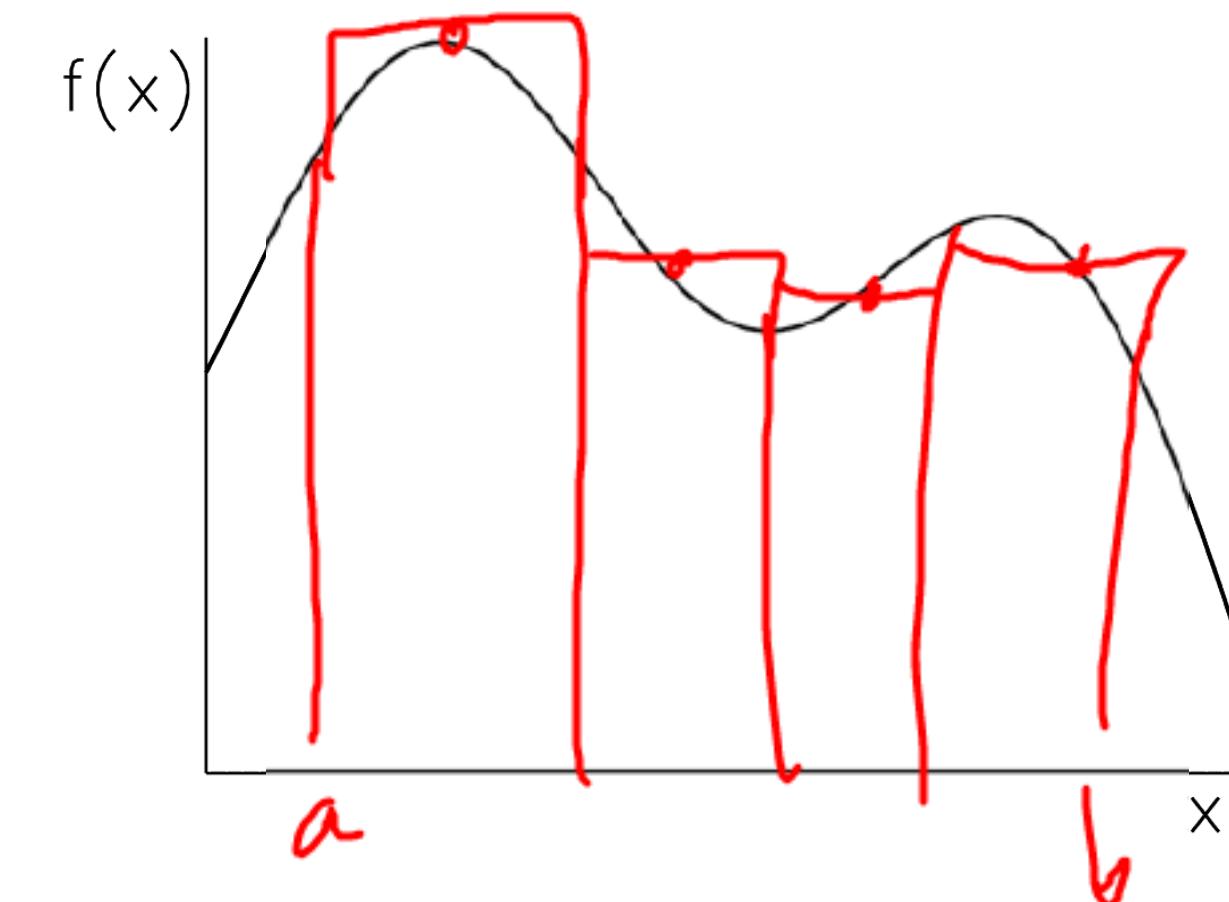
$$\int_{x_0}^{x_N} f(x) dx = \frac{h}{3} \left(\frac{27}{4}f_1 + \frac{13}{4}f_3 + 4f_4 + 2f_5 + 4f_6 + \cdots + 2f_{N-5} + 4f_{N-4} + \frac{13}{4}f_{N-3} + \frac{27}{4}f_{N-1} \right) + \mathcal{O}\left(\frac{1}{N^4}\right)$$

“Simpson’s pattern”

- Extended midpoint rule = middle Riemann sum

$$\int_{x_0}^{x_N} f(x) dx = h \left(f_{1/2} + f_{3/2} + f_{5/2} + \cdots + f_{N-3/2} + f_{N-1/2} \right) + \mathcal{O}\left(\frac{1}{N^2}\right)$$

open equivalent of the trapezoid rule
 (also: only even terms $\sim \frac{1}{N^{2k}}$)



Romberg integration

- $= N + 1$ evaluations
 N intervals: $\int_a^b f(x) dx \approx S_0$ (using $h_0 = h = (b - a)/N$)

Now with $N_1 = 2N$ intervals: $\int_a^b f(x) dx \approx S_1$ (using $h_1 = h_0/2$; h_i uses $h_0/2^i$)

$$\text{Error for } S_0: \mathcal{O}\left(\frac{1}{N^2}\right); \text{ error for } S_1: \mathcal{O}\left(\frac{1}{N_1^2}\right) = \frac{1}{4} \mathcal{O}\left(\frac{1}{N^2}\right)$$

$$S = \frac{4}{3}S_1 - \frac{1}{3}S_0$$

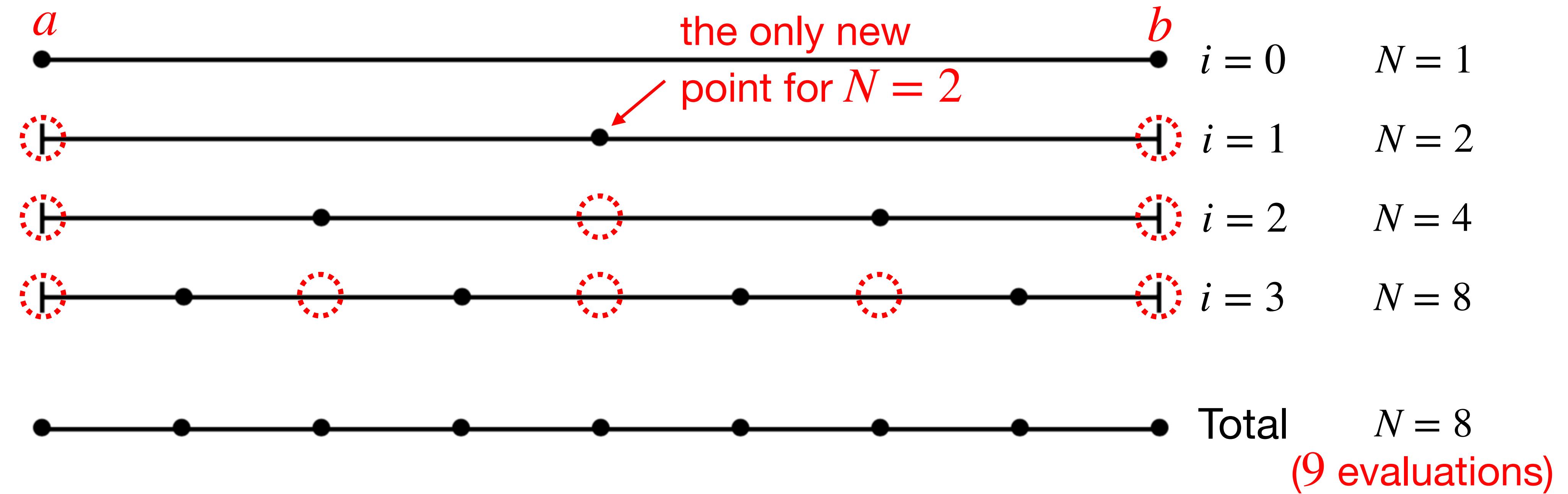
$$= \frac{4}{3} \frac{h}{2} \left(\frac{f_0}{2} + \sum_{i=1}^{2N-1} f_i + \frac{f_{2N}}{2} \right) + \frac{4}{3} \mathcal{O}\left(\frac{1}{4N^2}\right) - \frac{1}{3} h \left(\frac{f_0}{2} + \sum_{i=1}^{N-1} f_{2i} + \frac{f_{2N}}{2} \right) - \frac{1}{3} \mathcal{O}\left(\frac{1}{N^2}\right)$$

$$= \frac{h_1}{3} \left(f_0 + 4f_1 + 2f_2 + 4f_3 + \dots + 2f_{N_1-2} + 4f_{N_1-1} + f_{N_1} \right) + \mathcal{O}\left(\frac{1}{N^4}\right) \leftarrow \text{exactly Simpson's rule!}$$

But why stop there?

(except more efficient
and with less code)

Romberg integration



$$S_{i+1} = \frac{1}{2} S_i + \frac{b-a}{N_{i+1}} \sum_j f_j = \frac{1}{2} \left(S_i + \Delta_{i+1} \sum_j f_j \right)$$

uses $h_{i+1} = \frac{h_i}{2}$

$h_i \rightarrow \frac{h_i}{2}$

contributions of new points

space between new points ($= h_i$)

Analogue to Neville's algorithm

$N + 1$ evals: $h :$

$$S_0 \quad \mathcal{O}\left(\frac{1}{N^2}\right)$$

$$\frac{4S_1 - S_0}{3}$$

$$S_1 \quad \frac{1}{4}\mathcal{O}\left(\frac{1}{N^4}\right)$$

$$\frac{16S_{...} - S_{...}}{15}$$

$$S_2 \quad \frac{1}{64}\mathcal{O}\left(\frac{1}{N^6}\right)$$

$$\frac{64S_{...} - S_{...}}{63}$$

$$S_3 \quad \frac{1}{4096}\mathcal{O}\left(\frac{1}{N^8}\right)$$

$2N + 1$ evals: $h/2 :$

$$\frac{1}{4}\mathcal{O}\left(\frac{1}{N^2}\right)$$

$$S_2 \quad \frac{1}{64}\mathcal{O}\left(\frac{1}{N^4}\right)$$

$$S_3 \quad \frac{1}{4096}\mathcal{O}\left(\frac{1}{N^6}\right)$$

$4N + 1$ evals: $h/4 :$

$$\frac{1}{16}\mathcal{O}\left(\frac{1}{N^2}\right)$$

$$S_3 \quad \frac{1}{1024}\mathcal{O}\left(\frac{1}{N^4}\right)$$

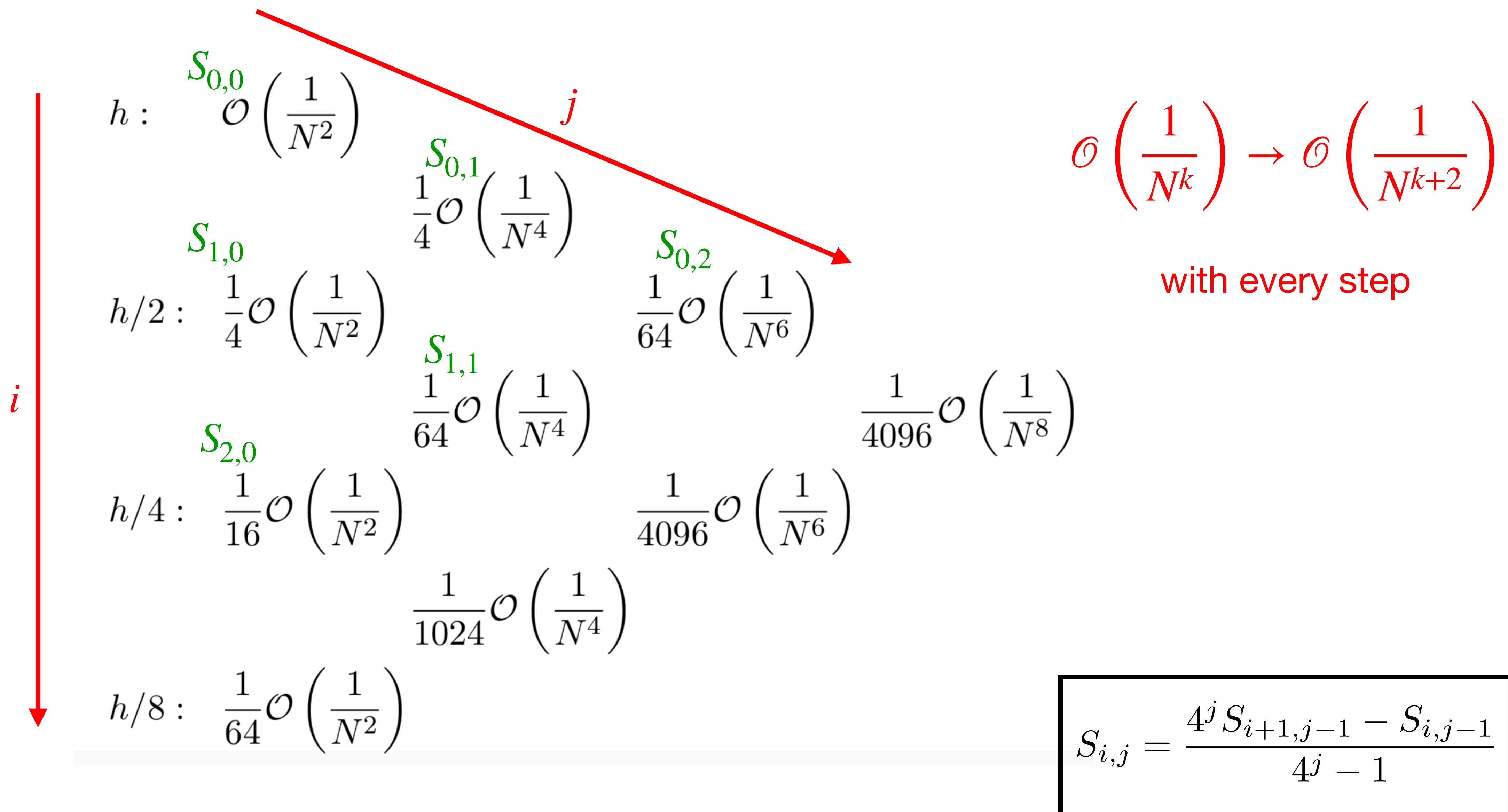
$8N + 1$ evals: $h/8 :$

$$\frac{1}{64}\mathcal{O}\left(\frac{1}{N^2}\right)$$

Romberg integration
 (= Richardson extrapolation
 applied to integration/
 trapezoid)

still uses only the same
 $8N + 1$ evaluations!

Analogue to Neville's algorithm



Romberg integration

1. Decide on an initial step size h and an “order” m , for example $h = b - a$ and $m = 5$.
2. Make an array r of size m for your estimates, up to step size $h/2^{m-1}$; set these to 0.
3. Calculate the initial estimate r_0 : for $h = b - a$, it's $r_0 = \frac{1}{2}h[f(a) + f(b)]$. **(trapezoid for $N = 1$)**
4. Set $N_p = 1$, loop over i from 1 through $m - 1$: **(fill out first column)**
5. Add in new function evaluations: $r_i = 0$, $\Delta = h$, $h = \frac{1}{2}h$, $x = a + h$, then N_p times do: $r_i = r_i + f(x)$, $x = x + \Delta$.
6. Now combine the new points with the already calculated ones for an estimate for $h/2^i$: $r_i = \frac{1}{2}[r_{i-1} + \Delta r_i]$, then double N_p .
7. Finish the loop over i . We now have the initial “column” of estimations.
now do weighted combinations
8. Reset N_p to 1, **start a new loop over i from 1 through $m - 1$:** **columns**
9. Upgrade the previous results by combining them: multiply N_p by 4, then for j from 0 up to $m - i$: $r_j = [N_p r_{j+1} - r_j]/(N_p - 1)$.
rows
10. Finish the loops over j and i ; r_0 now contains the best estimate for the integral.
error $\approx |r_0 - r_1|$

The magic in practice: error function

$$\operatorname{erf}(\underline{1}) = \frac{2}{\sqrt{\pi}} \int_0^{\underline{1}} e^{-x^2} dx \approx 0.8427007929 \quad (\text{correct answer to 10 decimals})$$

$N = 1, h = 1 :$

$N = 2, h = 1/2 :$

$N = 4, h = 1/4 :$

$N = 8, h = 1/8 :$

$N = 16, h = 1/16 :$

The magic in practice: error function

$$\operatorname{erf}(1) = \frac{2}{\sqrt{\pi}} \int_0^1 e^{-x^2} dx \approx 0.8427007929$$

all from trapezoid

$$N = 1, h = 1 : \quad 0.7717433323$$

$$N = 2, h = 1/2 : \quad 0.8252629556$$

$$N = 4, h = 1/4 : \quad 0.8383677774$$

$$N = 8, h = 1/8 : \quad 0.8416192212$$

$$N = 16, h = 1/16 : \quad 0.8424305055$$

16+1 unique evaluations

Now we combine with worse estimates that
use a subset of the same information (f_i)
and get out something much better!

The magic in practice: error function

$$\operatorname{erf}(1) = \frac{2}{\sqrt{\pi}} \int_0^1 e^{-x^2} dx \approx 0.8427007929$$

$N = 1, h = 1 :$	0.7717433323
	0.8431028300
$N = 2, h = 1/2 :$	0.8252629556
	0.8427360514
$N = 4, h = 1/4 :$	0.8383677774
	0.8427030358
$N = 8, h = 1/8 :$	0.8416192212
	0.8427009336
$N = 16, h = 1/16 :$	0.8424305055

Take these numbers and try it yourself!

Note: machine precision/round-off still limits us

The magic in practice: error function

$$\text{erf}(1) = \frac{2}{\sqrt{\pi}} \int_0^1 e^{-x^2} dx \approx 0.8427007929$$

$N = 1, h = 1 :$ 0.7717433323

0.8431028300

$N = 2, h = 1/2 :$ 0.8252629556

0.8427360514

$N = 4, h = 1/4 :$ 0.8383677774

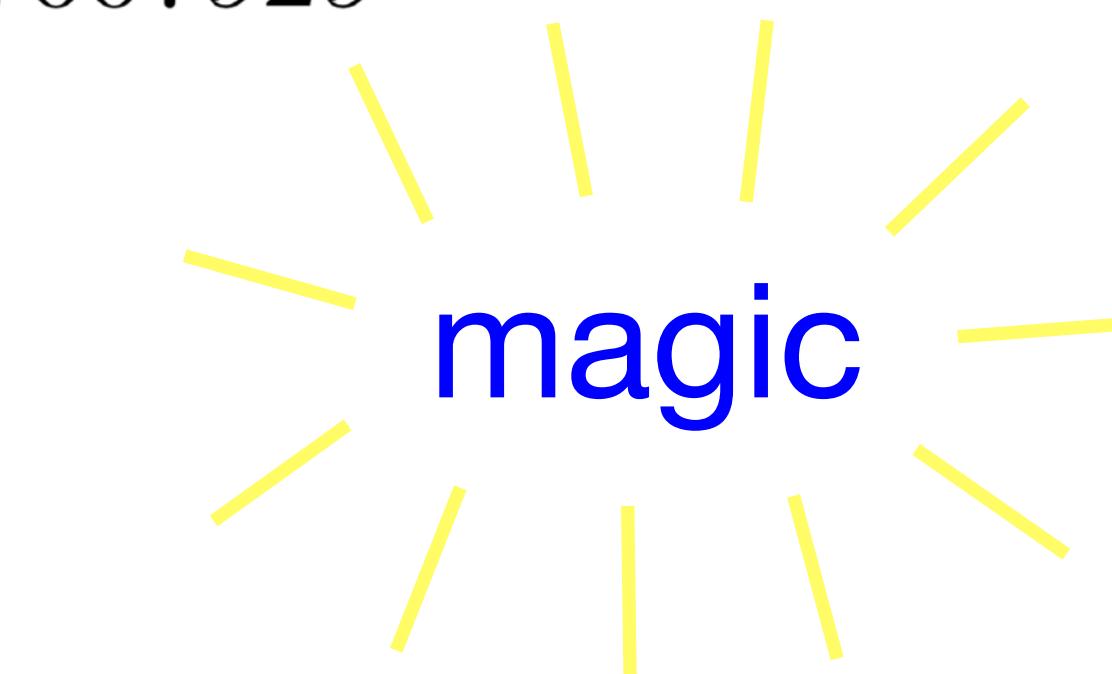
0.8427030358

$N = 8, h = 1/8 :$ 0.8416192212

0.8427009336

$N = 16, h = 1/16 :$ 0.8424305055

$\epsilon \sim 3 \times 10^{-4}$



(disclaimer: not actually magic,
we're exploiting that $f(x)$ is
smooth and has predictable errors)

$\epsilon \sim 4 \times 10^{-10}$
(6 orders of magnitude better
for ZERO extra f_i !)

error equivalent to trapezoid

with $h \approx 1/16384$

$\Rightarrow \sim 1000 \times$ more evaluations!

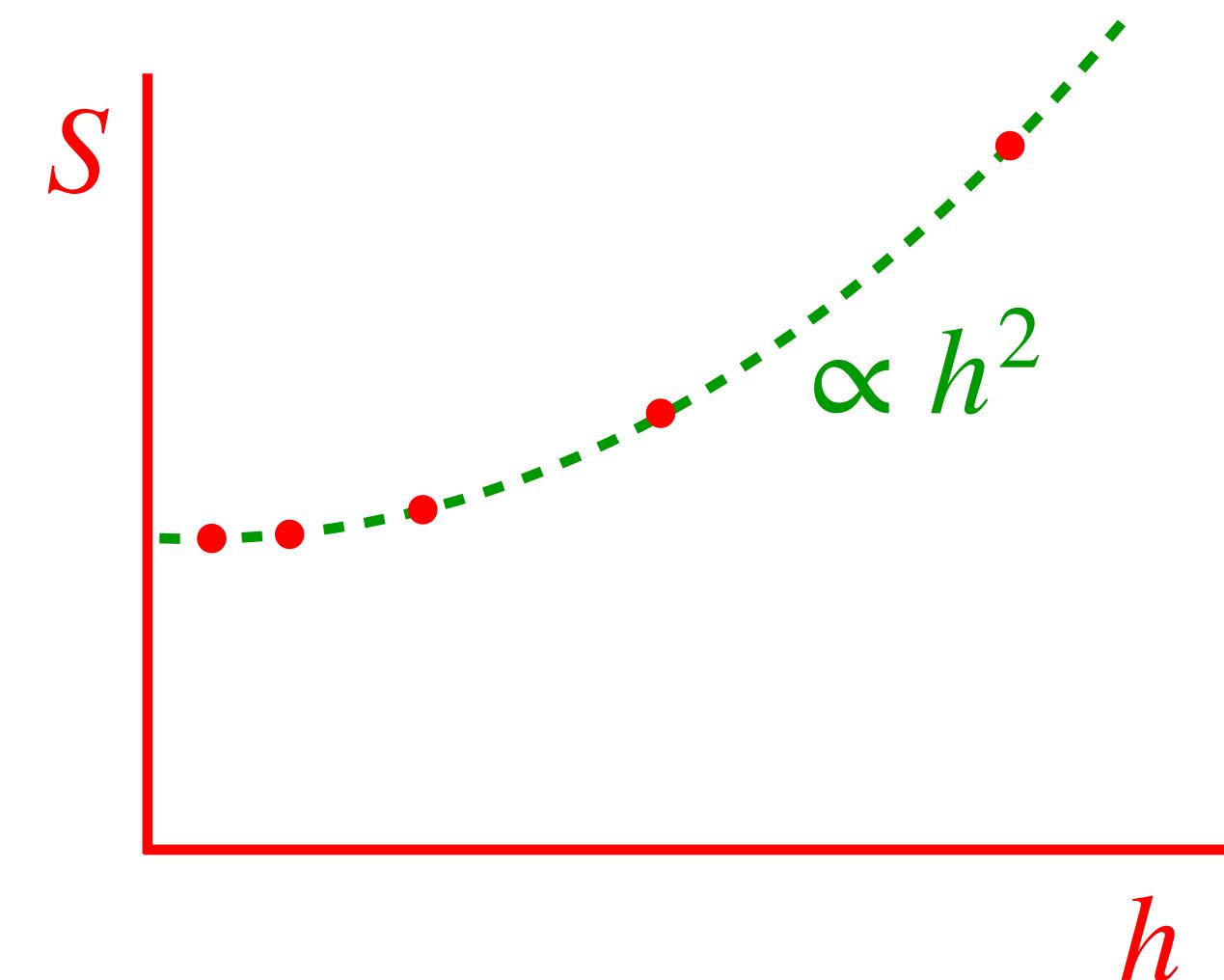
(Richardson) Extrapolation

↑
except now explicitly instead of cancelling error terms

- Explicitly extrapolating to $h = 0$

$$(h_i^2, S_i) \rightarrow (0, S) \quad \sim \text{Linear in } h^2$$

↑
extrapolate results to $h = 0$ limit based on
approximations/estimates for different h_i



- Where does this fail?

Same as Romberg: when f is not smooth/“predictable” \Rightarrow (scaling of) the error is unknown

Improper integrals

- Improper but not infinite or ill-defined:

- ✓ • The integrand is limited and has finite limits but cannot be evaluated at one of the limits (i.e. $\sin x/x$ at $x = 0$).
- ? • The upper or lower limit is $\pm\infty$.
- ✓ • It has an integrable singularity at either limit (e.g. $x^{-1/2}$ at $x = 0$). open method
- ? • It has an integrable singularity at some known place between the limits.
- ✗ • It has an integrable singularity at some unknown place between the limits.

- Extended midpoint Romberg

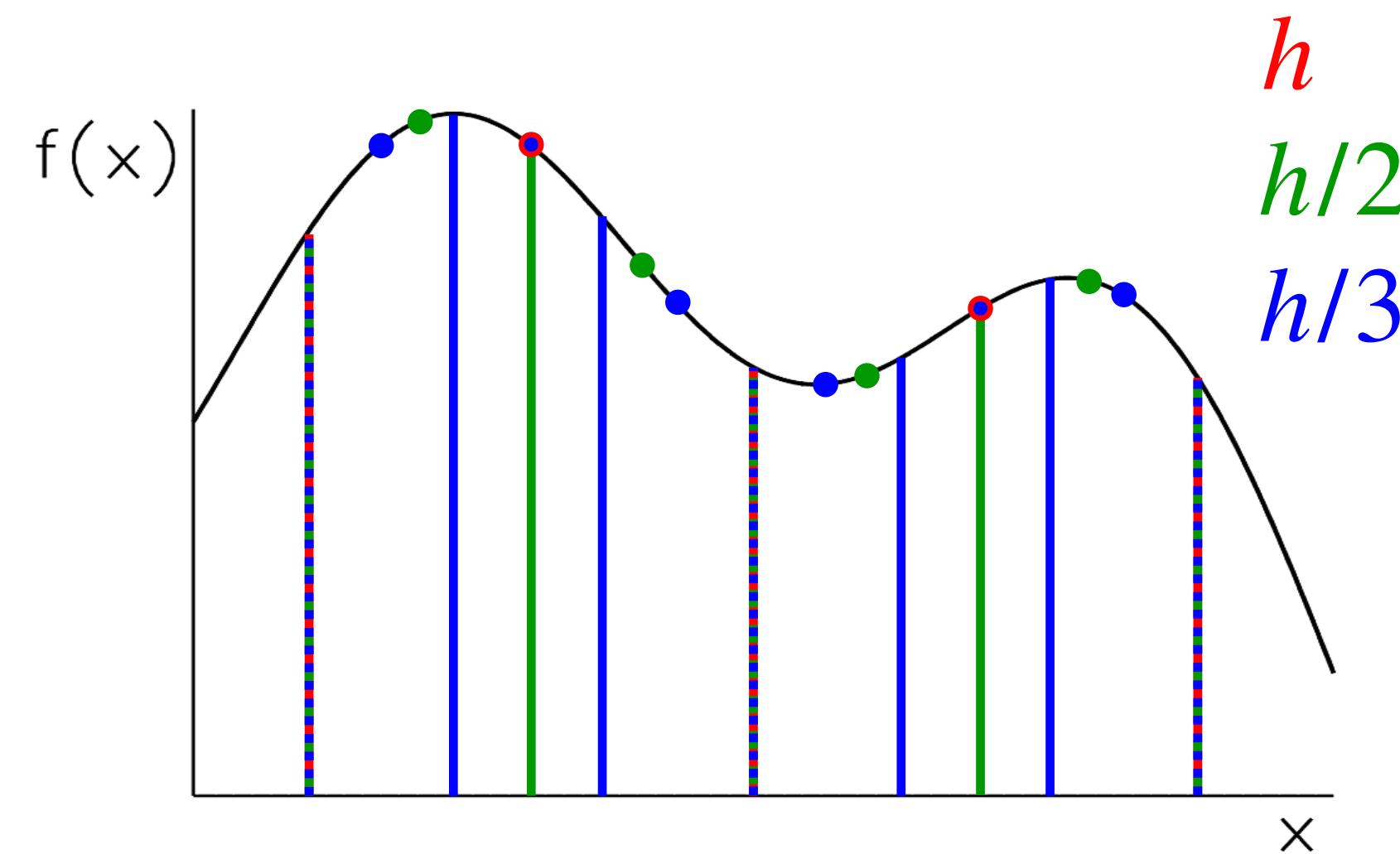
Just replace trapezoid by midpoint \Rightarrow open!

But: instead of $N \rightarrow 2N \rightarrow 4N \rightarrow \dots$

we have to do $N \rightarrow 3N \rightarrow 9N \rightarrow \dots$

in order to re-use previous evaluations

(so also $h/3$ instead of $h/2$, 9^j instead of 4^j , etc)



Change of variables

= rewrite integral

- Limits to infinity

Example: substitute $x = \frac{1}{t}$ (when $ab > 0$ and $f(x)$ decreases faster than x^{-2} as $x \rightarrow \pm\infty$)

$$\Rightarrow \int_a^b f(x) dx = \int_{1/b}^{1/a} \frac{1}{t^2} f\left(\frac{1}{t}\right) dt$$

- Slow power-law divergence

Example: substitute $x = a + t^{1/(1-\gamma)}$ (when $f(x)$ diverges as $(x - a)^{-\gamma}$ with $0 \leq \gamma < 1$)

$$\Rightarrow \int_a^b f(x) dx = \int_0^{(b-a)^{1-\gamma}} t^{\gamma/(1-\gamma)} f\left(a + t^{1/(1-\gamma)}\right) dt$$

- Many implementations exist! E.g. numpy/scipy

Summary: make substitutions not to find the primitive, but to alleviate infinite ranges and/or infinities

Breaking up the integral

- Splitting the limits

\downarrow

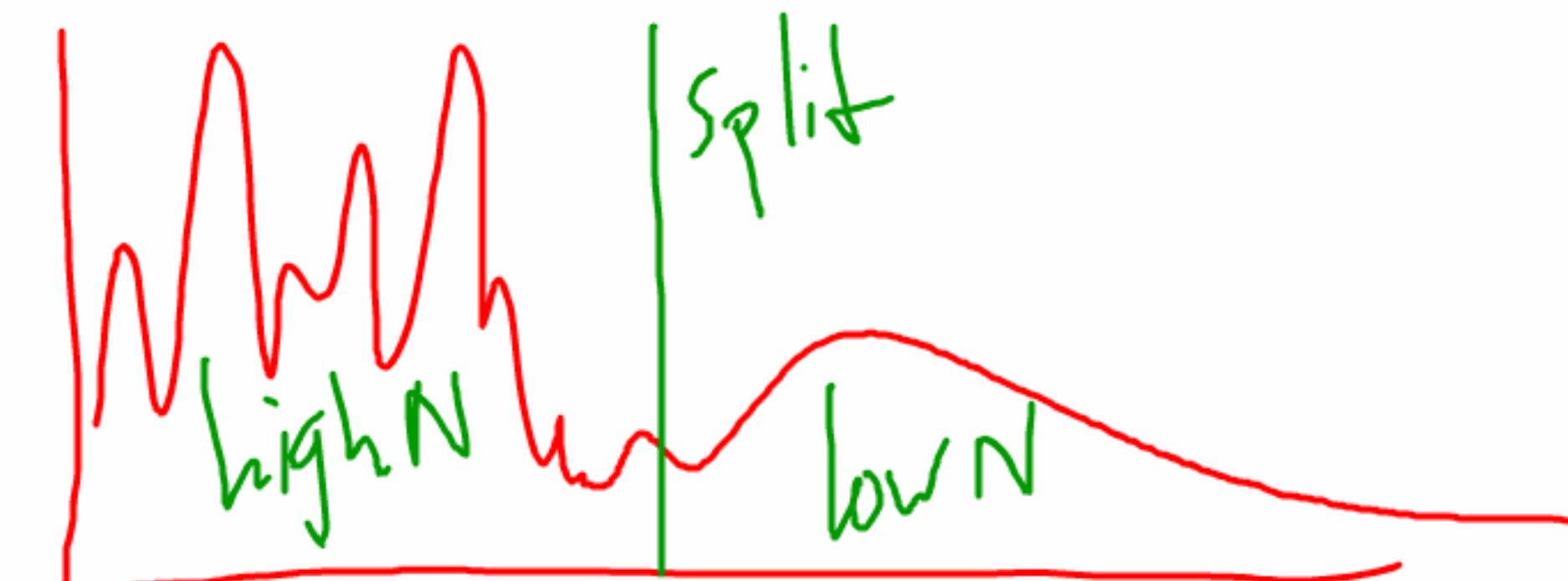
+ use
(semi-)open
method

$$\int_a^b f(x) dx = \int_a^m f(x) dx + \int_m^b f(x) dx$$
$$\left(= \int_a^{m_0} f(x) dx + \int_{m_0}^{m_1} f(x) dx + \cdots + \int_{m_N}^b f(x) dx \right)$$

E.g. $[a, b] = [-5, \infty) \rightarrow [-5, 2], [2, \infty)$

Note: we can use a different N and/or method for each sub-integral!

- Basics of adaptive step sizes



Linear vs logarithmic

- Range, space and measure

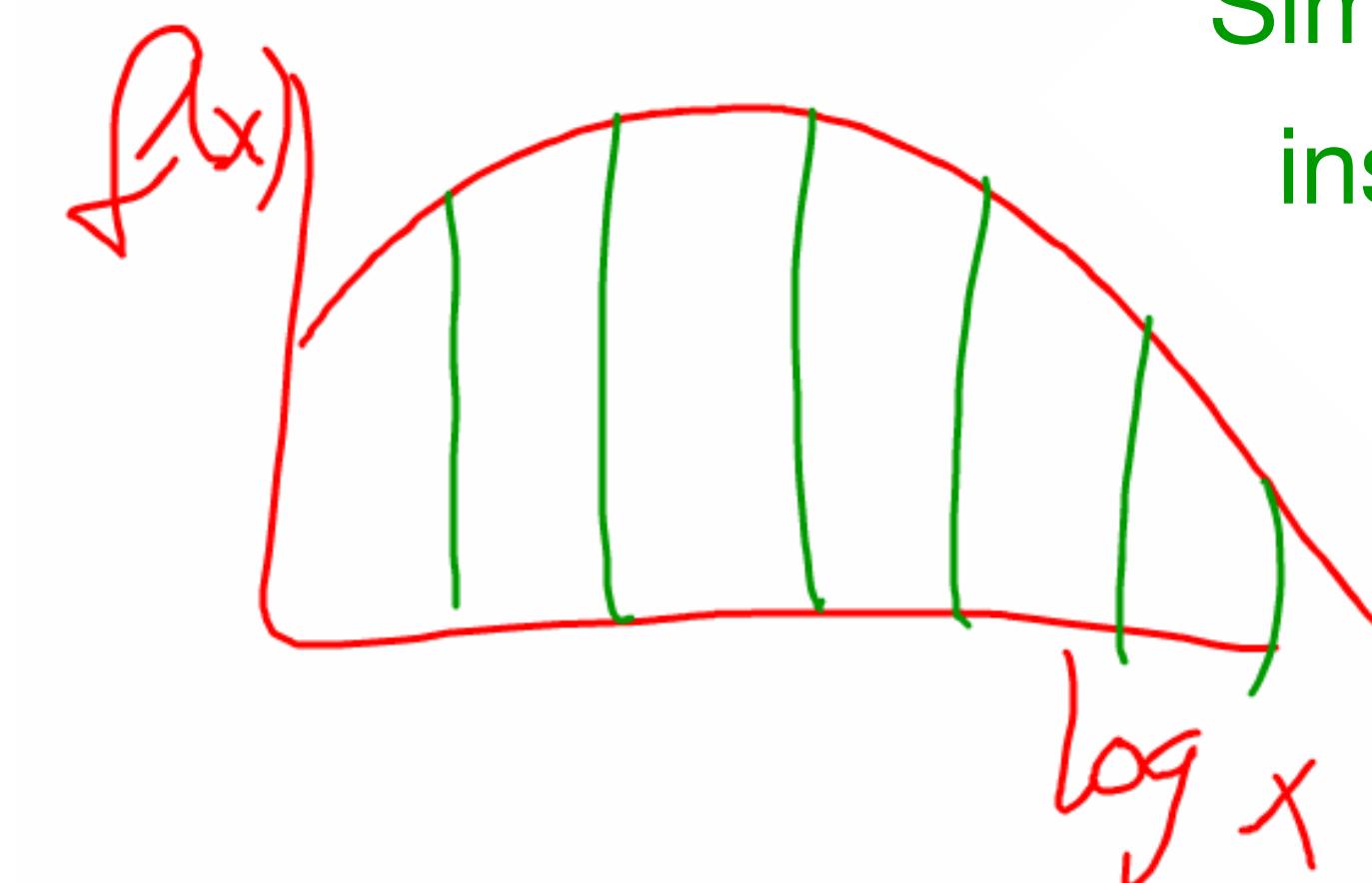
$$dx \rightarrow d \ln x$$

$$x \rightarrow e^t$$

$$x \rightarrow \ln t$$

- Appropriate transformations

Fewer evaluations for a given accuracy by choosing spaces/measures that match $f(x)$



Similarly, we could return $\ln f_i$ instead of f_i (as long as we transform back later)

Examples: $t = \ln x \Rightarrow \frac{d \ln x}{dx} = \frac{1}{x} \Rightarrow dx = x d \ln x = x dt \Rightarrow \int_a^b f(x) dx = \int_{\ln a}^{\ln b} f(e^t) e^t dt$

or: $t = e^{-x} \Rightarrow \int_a^\infty f(x) dx = \int_0^{e^{-a}} f(-\ln t) \frac{1}{t} dt$

Gaussian quadrature

- Additional freedom: unequal spacing

Say $f(x) = g(x)W(x)$ with $g(x)$ well-behaved, $W(x)$ troublesome

E.g. $f(x) = \frac{g(x)}{\sqrt{1 - x^2}}$, $f(x) = g(x)\sin x$

⇒ Choose x_j and weights w_j such that:

$$\int f(x) dx \approx \sum w_j g(x_j)$$

$N + N = 2N$ degrees

of freedom (x_j & w_j)

⇒ Exact if g is a polynomial with degree $< 2N$

- Convergence

Exponential convergence if x_j and w_j match $W(x)$

(but deriving these
is very difficult)

Gaussian quadrature

- Default rule: Gauss-Legendre $W(x) = 1$
- Gauss-Chebyshev: $W(x) = (1 - x^2)^{-1/2}$
- Gauss-Laguerre: $W(x) = \exp(-x)$
- Gauss-Hermite: $W(x) = \exp(-x^2)$
- w/x -generation in NumPy
 - Uses derived rules (hard!) or hard-coded coefficients (relatively recent alternatives: root-finding algorithms and asymptotic formulas)
 - See e.g. `scipy.integrate.quadrature`

Multidimensional integration

e.g. $\int \int f(x, y) dx dy$

- Reducing the dimensionality Rewrite integral, exploit symmetry, etc
 - E.g. surface of sphere is 3D in x, y, z , but 2D in ϕ, θ
Otherwise a lot of f_i are needed (exponential with dimension!)
- Function evaluation = integral
 - E.g. $\int f(x) dx$ with $f_i = \int g(x_i, y) dy$ (etc)
Note: we're again free to use different methods or N for each (sub)integral
- Pre-calculation and interpolation
 - If “inner” integral is expensive, consider pre-calculating it (e.g. on grid (x_i, y_i)) and interpolate its value to get evaluations for the “outer” integral(s) (less accurate, but much faster)

Next week:
Differentiation and random numbers