

Example For $f(x) = \tan^{-1} x$, estimate $f'(\sqrt{2})$ using a sequence of $h \rightarrow 0$. ($f'(x) = \frac{1}{1+x^2}$ so $f'(\sqrt{2}) = \frac{1}{3}$)

Using $h = 2^{-k}$, set $r = \frac{f(\sqrt{2}+h) - f(\sqrt{2})}{h}$

k	h	r
4	$0.62 \cdot 10^{-1}$	0.32374
12	$0.24 \cdot 10^{-3}$	0.33325 $\approx 1/3$
20	$0.95 \cdot 10^{-6}$	0.31250
24	$0.60 \cdot 10^{-7}$	1.00000
26	$0.15 \cdot 10^{-7}$	0.00000 (Why do we see $r=0$?)

Error = $\mathcal{O}(h)$!

Error = $\mathcal{O}(h^2)$ would work better:

Note $f(x+h) = f(x) + hf'(x) + \frac{h^2}{2} f''(x) + \frac{h^3}{3!} f'''(\xi_1)$ Error

$f(x-h) = f(x) - hf'(x) + \frac{h^2}{2} f''(x) - \frac{h^3}{3!} f'''(\xi_2)$ Error

$\Rightarrow f(x+h) - f(x-h) = 2hf'(x) + \frac{h^3}{3!} (f'''(\xi_1) + f'''(\xi_2))$

$\Rightarrow f'(x) = \underbrace{\frac{1}{2h} (f(x+h) - f(x-h))}_{\text{Estimate}} - \underbrace{\frac{h^2}{12} (f'''(\xi_1) + f'''(\xi_2))}_{\text{Error}}$

Error = $\mathcal{O}(h^2)$ but applicable only if f''' exists.

In fact, if $f \in C^3[x-h, x+h]$ then

$\exists x-h < \xi < x+h$ s.t. $\frac{f'''(\xi_1) + f'''(\xi_2)}{2} = f'''(\xi)$ (Why?)

Then

$f'(x) = \frac{1}{2h} (f(x+h) - f(x-h)) - \frac{h^2}{6} f'''(\xi)$ Error

(Similarly, we can obtain $f''(x) = \frac{1}{h^2} (f(x+h) - 2f(x) + f(x-h)) - \frac{h^2}{12} f^{(4)}(\xi)$)

Example For $f(x) = \tan^{-1} x$, estimate $f'(\sqrt{2})$ using a sequence of $h \rightarrow 0$. (Recall $f'(\sqrt{2}) = \frac{1}{3}$)

Using $h = 2^{-k}$, set $r = \frac{f(\sqrt{2}+h) - f(\sqrt{2}-h)}{2h}$

k	h	r
2	0.25	0.33719
10	$0.98 \cdot 10^{-3}$	0.33334 $\approx 1/3$
18	$0.38 \cdot 10^{-5}$	0.32813
26	$0.15 \cdot 10^{-7}$	0.00000

Note that even if we compute $x \pm h$ very accurately, the error in $f(x \pm h)$ (or $f(x+h) - f(x-h)$) gets multiplied by $\frac{1}{2h}$.

This magnifies the error since h is very small.

Thus, we should be very careful with numerical differentiation of empirical data. In fact, we should avoid it if possible.

Differentiation via Polynomial Interpolation

Recall: Lagrange form of interpolation polynomial

$p(x) = \sum_{i=0}^n f(x_i) l_i(x)$ where $l_i(x) = \prod_{\substack{j=0 \\ j \neq i}}^n \frac{x - x_j}{x_i - x_j}$

Recall: Thm 2 of Section 6.1:

$f(x) = p(x) + \frac{1}{(n+1)!} f^{(n+1)}(\xi_x) w(x)$ where $w(x) = \prod_{i=0}^n (x - x_i)$ ξ_x depends on x ! $\frac{dy}{dx}$ not necessarily 0.

$\Rightarrow f'(x) = \sum_{i=0}^n f(x_i) l_i'(x) + \frac{1}{(n+1)!} \frac{d}{dx} (f^{(n+1)}(\xi_x) w(x)) + \frac{1}{(n+1)!} f^{(n+1)}(\xi_x) w'(x)$

Since $w(x) = \prod_{i=0}^n (x - x_i)$, ① $w(x_k) = 0$ $(f_0(x) f_2(x) f_3(x) \dots f_n(x))'$

② $w'(x) = \sum_{i=0}^n \prod_{\substack{j=0 \\ j \neq i}}^n (x - x_j)$ $= f_1(x) f_2(x) \dots f_n(x)$
 $+ f_1(x) f_2(x) \dots f_{n-1}(x)$
 $+ \dots + f_1(x) f_2(x) \dots f_n(x)$

② \Rightarrow ③ $w'(x_k) = \prod_{\substack{j=0 \\ j \neq k}}^n (x_k - x_j)$

① & ③ $\Rightarrow f'(x_k) = \sum_{i=0}^n f(x_i) l_i'(x_k) + \frac{f^{(n+1)}(\xi_{x_k})}{(n+1)!} \prod_{\substack{j=0 \\ j \neq k}}^n (x_k - x_j)$

Example: ($n=2, k=1$) x_0, x_1, x_2

$l_0(x) = \frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)}$ $l_0'(x) = \frac{2x - x_1 - x_2}{(x_0-x_1)(x_0-x_2)}$

$l_1(x) = \frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)}$ \Rightarrow $l_1'(x) = \frac{2x - x_0 - x_2}{(x_1-x_0)(x_1-x_2)}$

$l_2(x) = \frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_2-x_1)}$ $l_2'(x) = \frac{2x - x_0 - x_1}{(x_2-x_0)(x_2-x_1)}$

$l_0'(x_1) = \frac{x_1 - x_2}{(x_0-x_1)(x_0-x_2)}$

\Rightarrow $l_1'(x_1) = \frac{2x_1 - x_0 - x_2}{(x_1-x_0)(x_1-x_2)}$

$l_2'(x_1) = \frac{x_1 - x_0}{(x_2-x_0)(x_2-x_1)}$

Thus, $f'(x_1) = f(x_0) \frac{x_1 - x_2}{(x_0-x_1)(x_0-x_2)} + f(x_1) \frac{2x_1 - x_0 - x_2}{(x_1-x_0)(x_1-x_2)} + f(x_2) \frac{x_1 - x_0}{(x_2-x_0)(x_2-x_1)} + \text{Error}$ $(x_1-x_0) + (x_1-x_2)$

where $\text{Error} = \frac{1}{6} f'''(y_x) (x_1 - x_0)(x_1 - x_2)$ $w^1(x) = (x_1 - x_0)(x_1 - x_2)$

Example Say x_0, x_1, x_2 are equally spaced, that is $x_0 = x_1 - h$
 $x_2 = x_1 + h$ for some $h > 0$. Then

$$f'(x_1) = f(x_1 - h) \frac{(-h)}{(-h)(-2h)} + f(x_1) \frac{h-h}{-h \cdot h} + f(x_1 + h) \frac{h}{2h(h)}$$

$$+ \frac{1}{6} f'''(y_x) h h$$

or $= \frac{f(x_1 + h) - f(x_1 - h)}{2h} + \frac{1}{6} f'''(y_x) h^2$ as before!

Richardson Extrapolation

Taylor series: $f(x+h) = \sum_{k=0}^{\infty} \frac{1}{k!} h^k f^{(k)}(x) = f(x) + f'(x)h + \frac{f''(x)}{2}h^2 + \frac{f'''(x)}{3!}h^3 + \dots$

$$f(x-h) = \sum_{k=0}^{\infty} \frac{1}{k!} (-h)^k f^{(k)}(x) = f(x) - f'(x)h + \frac{f''(x)}{2}h^2 - \frac{f'''(x)}{3!}h^3 + \dots$$

$$\Rightarrow f(x+h) - f(x-h) = 2f'(x)h + \frac{f'''(x)}{3}h^3 + O(h^5)$$

$$\Rightarrow f'(x) = \frac{f(x+h) - f(x-h)}{2h} - \frac{f'''(x)h^2}{3} + O(h^4)$$

Quantity of interest Estimate with param. h Error term in powers of $h = O(h^2)$

$$\Rightarrow L = \underbrace{\mathcal{Q}(h)}_{\text{Quantity of interest}} + \underbrace{a_2 h^2 + a_4 h^4 + a_6 h^6 + \dots}_{\text{Estimate with param. } h} \quad \forall h$$

$$\Rightarrow L = \mathcal{Q}\left(\frac{h}{2}\right) + a_2 \frac{h^2}{4} + a_4 \frac{h^4}{16} + a_6 \frac{h^6}{64} + \dots$$

$$\Rightarrow 4L - L = 4\mathcal{Q}\left(\frac{h}{2}\right) - \mathcal{Q}(h) + 4a_2 \frac{h^2}{4} - a_2 h^2 + 4a_4 \frac{h^4}{16} - a_4 h^4 + 4a_6 \frac{h^6}{64} - a_6 h^6 + \dots$$

$$\Rightarrow L = \frac{4}{3} \mathcal{Q}\left(\frac{h}{2}\right) - \frac{1}{3} \mathcal{Q}(h) - \frac{1}{4} a_4 h^4 - \frac{5}{16} a_6 h^6 - \dots$$

formula for L Error = $O(h^4)$

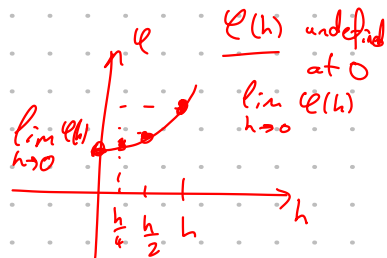
$$L = \Psi(h) + b_4 h^4 + b_6 h^6 + b_8 h^8 + \dots \quad \text{where } \Psi(h) = \frac{4}{3} \mathcal{Q}\left(\frac{h}{2}\right) - \frac{1}{3} \mathcal{Q}(h)$$

$$\Rightarrow L = \Psi\left(\frac{h}{2}\right) + b_4 \frac{h^4}{4^2} + b_6 \frac{h^6}{4^3} + \dots$$

$$\Rightarrow 4^2 L - L = 4^2 \Psi\left(\frac{h}{2}\right) - \Psi(h) + O(h^6)$$

$$\Rightarrow L = \frac{4^2 \Psi\left(\frac{h}{2}\right) - \Psi(h)}{4^2 - 1} + O(h^6)$$

formula for L Error term



$$L = \Theta(h) + O(h^6) \quad \text{where } \Theta(h) = \frac{16 \Psi\left(\frac{h}{2}\right) - \Psi(h)}{15}$$

The same process can be repeated over and over again to produce error terms with higher powers of h .

This is called the Richardson extrapolation.

The algorithm:

M : number of steps, \mathcal{Q} : the initial "formula"

1. Select a value for h (e.g. $h=1$)

for $0 \leq n \leq M$, compute $D(n, 0) = \mathcal{Q}(h/2^n)$.

2. for $k=1$ to M

for $n=k$ to M

$$D(n, k) = \frac{4^k}{4^k - 1} D(n, k-1) - \frac{1}{4^k - 1} D(n-1, k-1)$$

(the second param represents which "formula" we are using $\mathcal{Q}, \Psi, \Theta, \dots$)

end for

$$D(1, 1) = \frac{4}{3} \mathcal{Q}\left(\frac{h}{2}\right) - \frac{1}{3} \mathcal{Q}(h) = \Psi(h)$$

e.g. $D(0, 0) = \mathcal{Q}(h)$

$$D(1, 0) = \mathcal{Q}\left(\frac{h}{2}\right) \rightarrow D(1, 1) = \Psi(h)$$

$$D(2, 0) = \mathcal{Q}\left(\frac{h}{4}\right) \rightarrow D(2, 1) = \Psi\left(\frac{h}{2}\right) \rightarrow D(2, 2) = \Theta(h)$$

Thm (on Richardson extrapolation)

$$D(n, k-1) = L + O(h^{2k}) \quad \text{as } h \rightarrow 0$$

more precisely,

$$D(n, k-1) = L + \sum_{j=k}^{\infty} A_{jk} \left(\frac{h}{2^n}\right)^{2j}$$

Example Approximate $\frac{d}{dx} \tan^{-1} x \Big|_{x=\sqrt{2}}$

n	$D(n, 0)$	$D(n, 1)$	$D(n, 2)$	$D(n, 3)$
0	0.39269 91			
1	0.34877 10	0.33412 83		
2	0.33719 38	0.33333 48	0.33328 19	
3	0.33429 81	0.33333 29	0.33333 28	0.33333 36

Warning: Richardson extrapolation is still vulnerable to subtractive cancellation.

7.2 Numerical Integration Based on Interpolation

FIC: If h is an antiderivative of f , $\int_a^b f(x) dx = h(b) - h(a)$

However, many elementary functions do not have simple antiderivatives.

e.g. $\int e^{x^2} dx = \sum_{k=0}^{\infty} \frac{x^{2k+1}}{(2k+1)k!} + C$ ($e^{x^2} = \sum \frac{x^{2k}}{k!}$)

Computing the Taylor series is also not always possible.

e.g. if we only know $(x, f(x))$ at finitely many points.

One robust strategy is replace f with g s.t. $g \approx f$ and g is easy to integrate. Of course, we will use polynomial interpolation.

Given $[a, b]$, we select nodes x_0, x_1, \dots, x_n . Then

$$f(x) \approx p(x) = \sum_{i=0}^n f(x_i) l_i(x)$$

$$\Rightarrow \int_a^b f(x) dx \approx \sum_{i=0}^n f(x_i) \int_a^b l_i(x) dx$$

Set $A_i = \int_a^b l_i(x) dx$ Then $\int_a^b f(x) dx \approx \sum_{i=0}^n A_i f(x_i)$

Note that A_i is independent of f ! so the formula

$$\textcircled{*} \int_a^b f(x) dx \approx \sum_{i=0}^n A_i f(x_i) \text{ can be used on any function } f!$$

If x_0, x_1, \dots, x_n are equally spaced, $\textcircled{*}$ is called a Newton-Cotes formula.

Trapezoid Rule

If $n=1$, $x_0=a$, $x_1=b$, then

$$l_0(x) = \frac{b-x}{b-a} \quad l_1(x) = \frac{x-a}{b-a}$$

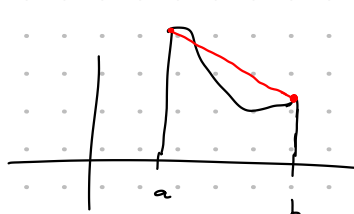
$$\Rightarrow A_0 = \int_a^b l_0(x) dx = \left. -\frac{1}{2} \frac{(b-x)^2}{(b-a)} \right|_a^b = \frac{1}{2}(b-a)$$

$$A_1 = \int_a^b l_1(x) dx = \left. \frac{1}{2} \frac{(x-a)^2}{(b-a)} \right|_a^b = \frac{1}{2}(b-a)$$

Thus, the quadrature formula is

a formula that gives the area.

$$\int_a^b f(x) dx \approx \frac{b-a}{2} [f(a) + f(b)]$$



$$\left(\text{Error} = -\frac{1}{12} (b-a)^3 f''(\eta) \quad \eta \in (a, b) \right)$$