

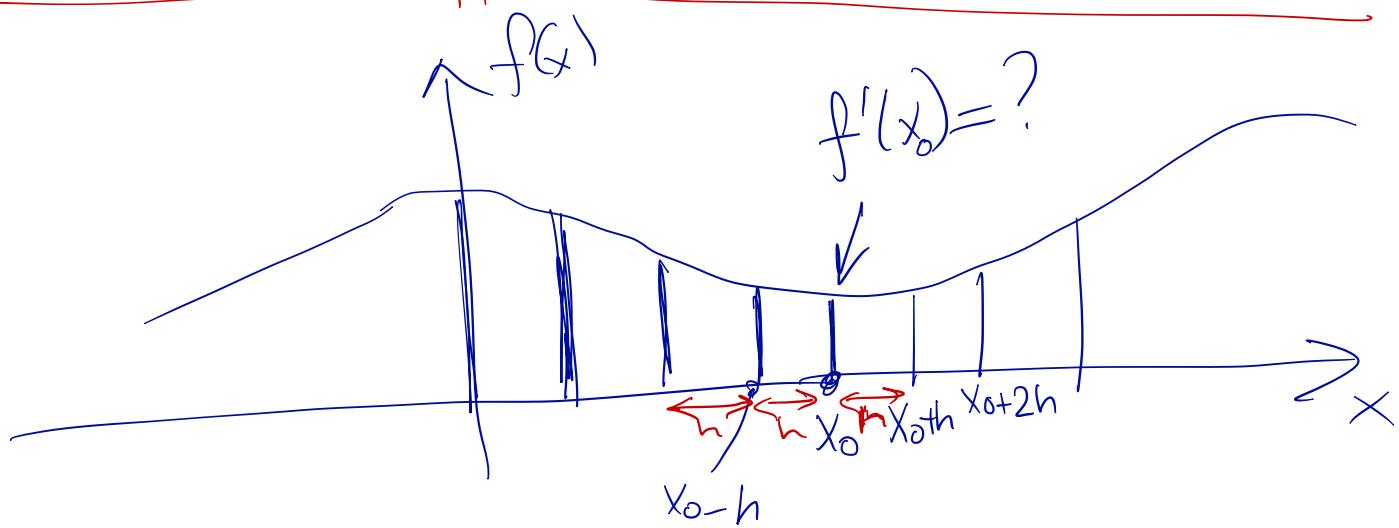
BLG 202E

Numerical Methods

Numerical Differentiation - Part 2

19.04.2016

Numerical Differentiation (Chap 14) - 3/4



① Taylor Series Method:

Forward difference

$O(h)$

← Truncation error

$$\text{T.S. expand} \quad f(x_0+h) = f(x_0) + h f'(x_0) + \dots$$

$$f(x_0) = \dots$$

$$= \text{- 3 pt centered formula. } O(h^2)$$

$$f(x_0+h) = \dots$$

$$f(x_0-h) = \dots$$

② Richardson Extrapolation: Used two methods of lower order at other grid distances (e.g. $2h$), to obtain a higher order method.

③ Used polynomial interpolation (today we continue w/ this)

$$f(x) \approx p_n(x)$$

$$f'(x) \approx p_n'(x)$$

Differentiation using Lagrange Interpolation:

$$p(x) = \sum_{j=0}^n f(x_j) L_j(x) ; \text{ Given interpolation points: } x_0, x_1, \dots, x_n$$

Recall

Lagrange polynomials:

$$L_j(x) = \frac{(x-x_0) \dots (x-x_{j-1})(x-x_{j+1}) \dots (x-x_n)}{(x_j-x_0) \dots (x_j-x_{j-1})(x_j-x_{j+1}) \dots (x_j-x_n)}$$

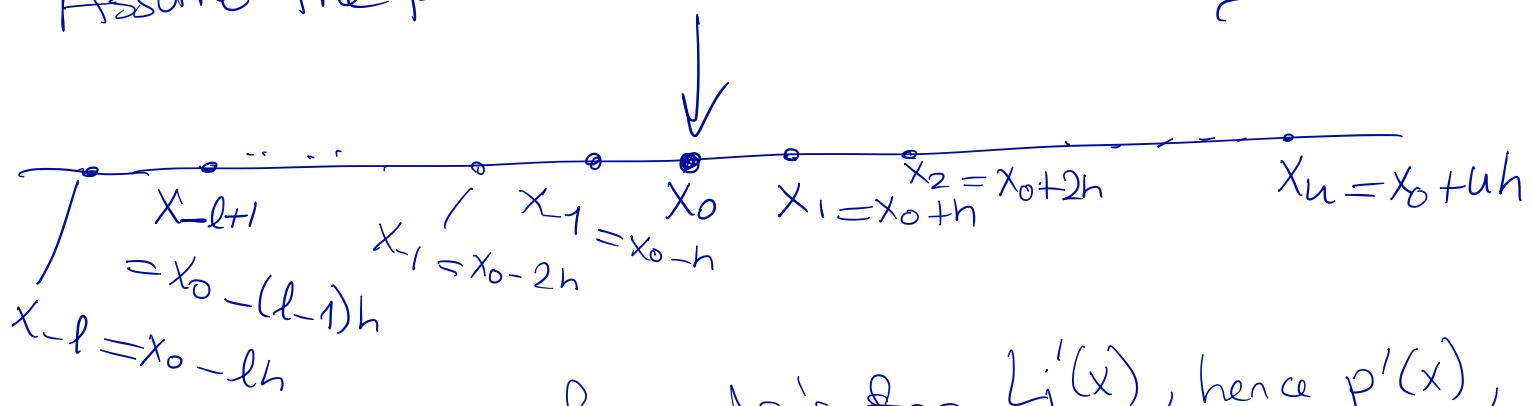
Take the derivative & substitute $x=x_0$

$$p'(x_0) = \sum_{j=0}^n f(x_j) L'_j(x)$$

Recall the points x_0, x_1, \dots, x_n are not required to be equidistant!

Differentiation using equidistant points:

Assume the points are distributed around $\underline{x_0}$:



Next, we derive formula's for $L'_j(x)$, hence $p'(x)$, to calculate $f'(x)$.

$$p'(x_0) = \sum_{j=-l}^u f(x_j) L_j'(x)$$

$L_j(x) = \frac{(x-x_{-l}) \dots (x-x_0) \dots (x-x_u)}{(x_j-x_{-l}) \dots (x_j-x_0) \dots (x_j-x_u)}$

Remember
 $(x-x_j)$
is not included

$$L_0(x) = \frac{(x-x_1)(x-x_2) \dots (x-x_u)(x-x_{-1}) \dots (x-x_{-l})}{(x_0-x_1)(x_0-x_2) \dots (x_0-x_u)(x_0-x_{-1}) \dots (x_0-x_{-l})}$$

$$L_0'(x_0) = \frac{1}{x_0-x_1} + \frac{1}{x_0-x_2} + \dots + \frac{1}{x_0-x_u} + \frac{1}{x_0-x_{-1}} + \dots + \frac{1}{x_0-x_{-l}}$$

$\underbrace{-h}_{-h}$ $\underbrace{\dots}_{2h}$ \dots $\underbrace{\dots}_{u-h}$ $\underbrace{\dots}_{h}$ \dots $\underbrace{x_0-x_{-l}}_{-l.h}$

$$L_0'(x_0) = \sum_{\substack{j=0 \\ k=-l \\ k \neq 0}}^u \frac{1}{x_0-x_k} = \frac{1}{h} \sum_{\substack{k=-l \\ k \neq 0}}^u \left(-\frac{1}{k} \right)$$

$$L_j'(x_0) = \frac{1}{x_j-x_0} \prod_{\substack{k=-l \\ k \neq 0 \\ k \neq j}}^u \frac{(x_0-x_k)}{(x_j-x_k)}$$

$\stackrel{x_0-(x_0+kh) = -kh}{=} \stackrel{\cancel{x_0}}{\cancel{x_0+kh}}$

$$= \frac{1}{jh} \prod_{k=-l}^u \frac{(-k)}{(j-k)} \triangleq a_j$$

$x_j-x_k = x_0+jh - (x_0+kh) = (j-k)h$

Exercise : derive this !

Let $a_j = h L_j'(x_0)$, $j = -l, \dots, u$ j is a term
indep of f and h . \rightarrow these weights can be calculated once & for all.

$$\Rightarrow p'(x_0) = \frac{1}{h} \sum_{j=-l}^u a_j f(x_j)$$

Q: How accurate is this formula (on the previous page)?

We'll use the polynomial interpolation error:

$$f(x) - p_n(x) = f[x_{-l}, x_{-l+1}, \dots, x_u, x] \prod_{k=-l}^u (x - x_k)$$

The error in the numerical differentiation formula

at $x = x_0$:

$$\begin{aligned}
 & \text{at } x = x_0 : \\
 f'(x) - p_n'(x) &= \frac{d}{dx} \left\{ f[x_{-l}, \dots, x_u, x] \right\}_{:= l}^{\frac{u}{l}(x-x_k)} \\
 (\text{evaluate at } x=x_0) &= \frac{d}{dx} \left[\dots \right] \cancel{l(x-x_k)} + f[x_{-l}, \dots, x_u, x_0], \\
 &\quad \text{at } x=x_0 \\
 &= \frac{d}{dx} \left\{ (x-x_{-l})(x-x_{-l+1}) \dots (x-x_0)(x-x_1) \dots (x-x_u) \right\}_{x=x_0} \\
 &= f[x_{-l}, \dots, x_u, x_0] \\
 &\quad \text{at } x=x_0 \\
 &\quad \text{use the theorem to}
 \end{aligned}$$

Use the theorem to rewrite the divided diff (p. 312)

$$\begin{aligned}
 &= \frac{f^{(n+1)}(\xi)}{(n+1)!} (x_0 - x_{-l}) (x_0 - x_{-l+1}) \dots (x_0 - x_{-1}) (x_0 - x_1) \dots \\
 &\quad \underbrace{(l \cdot h)}_{\text{l.h}} \quad \underbrace{((l-1) \cdot h)}_{(l-1) \cdot h} \quad \underbrace{h}_{h} \quad \underbrace{-h}_{-h} \quad \underbrace{(x_0 - x_u)}_{-u \cdot h} \\
 &= \left(\frac{f^{(n+1)}(\xi)}{(n+1)!} l! u! \right) \underbrace{h^n}_{h^n} \underbrace{(-1)^u}_{(-1)^u} \quad \xi \in [x_{-l}, x_u] \\
 &\quad \rightarrow O(h^n) \text{ for nub: .}
 \end{aligned}$$

$$\left| f'(x_0) - \frac{1}{h} \sum_{j=-l}^u a_j f(x_j) \right| \leq \frac{l! u!}{(n+1)!} \|f^{(n+1)}\|_{\infty} h^n$$

$= P_n'(x_0)$

Truncation error ↗

Ex: (14.5): Let $l = +1$, $u = 1$; calculate $f'(x_0)$
Given $f(x_{-1})$, $f(x_0)$, $f(x_1)$, $\Rightarrow n = 2$

$$x_{-1} = x_0 - h \quad x_0 \quad x_0 + h = x_1$$

We derived
Recall:

$$a_j = h L_j'(x_0)$$

$$L_j'(x_0) = \frac{1}{jh} \sum_{k=-l}^u \left(\frac{-k}{j-k} \right)$$

$k \neq 0$
 $k \neq j$

$$P_2'(x_0) = \frac{1}{h} \sum_{j=-l}^u a_j f(x_j)$$

(HW: write a matlab fn. to calculate a_j)

$$L_0'(x_0) = \frac{1}{h} (1 + (-1))$$

$$L_0'(x_0) = \frac{1}{h} \sum_{k=-l}^u \left(-\frac{1}{k} \right)$$

$$L_0'(x_0) = 0 \Rightarrow a_0 = 0$$

$$L_{-1}'(x_0) = \frac{1}{(-1)h} \left(\frac{-1}{-1-1} \right) = -\frac{1}{2h}$$

$$L_1'(x_0) = \frac{1}{h} \left(-\frac{(-1)}{1-(-1)} \right) = \frac{1}{2h}$$

$$\Rightarrow f'(x_0) = P_2'(x_0) = \frac{1}{h} (a_{-1} f(x_0-h) + a_1 f(x_0+h) + a_0 f(x_0))$$

$$f'(x_0) = \frac{f(x_0+h) - f(x_0-h)}{2h}$$

\Rightarrow The error is estimated by $f'(x_0) - p'_2(x_0)$

Recall: $f'(x_0) - p'_n(x_0) = \frac{f^{(n+1)}(\xi)}{(n+1)!} l! u! (-1)^u h^n$

$$f'(x_0) - p'_2(x_0) = -\frac{1! 1! h^2}{3!} f'''(\xi)$$

*compare
to the
result
of
T.S method
(Taylor Series)*

$$\Rightarrow O(h^2)$$

Exercise: Take $l=2, u=2,$

Derive the 1st derivative: (calculate)

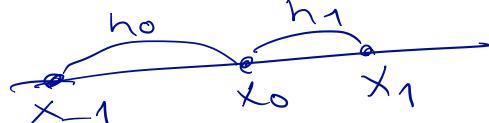
$$f'(x_0) \approx \frac{1}{12h} (f(x_0-2h) - 8f(x_0-h) + 8f(x_0+h) - f(x_0+2h))$$

Calculate truncation error: $e(h) = \frac{h^4}{30} f^{(4)}(\xi) \Rightarrow O(h^4)$

Compare to T.S. results!

Non-uniformly spaced Points: This is the major advantage of the polynomial differentiation, which is more complicated but useful.

Ex: (14.6): Suppose points $x_{-1} = x_0 - h_0, x_1 = x_0 + h_1$



Want to derive a 2nd order formula for $f'(x_0).$

An ad-hoc approach: $f'(x_0) = \frac{f(x_1) - f(x_{-1})}{h_0 + h_1}$

Instead, we'll use poly. differentiation \Rightarrow

naive
approach

$$\Rightarrow P_2'(x_0) = \sum_{j=-1}^1 f(x_j) L_j'(x_0)$$

$$L_0'(x_0) = ?$$

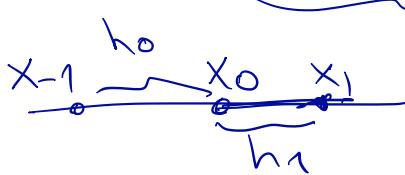
$$L_{-1}'(x_0) = ?, \quad L_1'(x_0) = ?$$

Recall:

$$L_0'(x_0) = \sum_{\substack{k=-l \\ k \neq 0}}^u \frac{1}{x_0 - x_k}$$

$$L_j'(x_0) = \frac{1}{x_j - x_0} \prod_{\substack{k=-l \\ k \neq j}}^u \frac{x_0 - x_k}{x_j - x_k}$$

$$L_0'(x_0) = \underbrace{\frac{1}{(x_0 - x_{-1})}}_{h_0} + \underbrace{\frac{1}{(x_0 - x_1)}}_{h_1} = \frac{1}{h_0} + \frac{(-1)}{h_1} = \frac{h_1 - h_0}{h_1 h_0}$$



$$; L_{-1}'(x_0) = \frac{1}{(x_0 - x_{-1})} \frac{(x_0 - x_1)}{(x_{-1} - x_1)} = -\frac{1}{h_0} \frac{h_1(-1)}{(h_0 + h_1)}$$

$$L_1'(x_0) = \frac{1}{(x_1 - x_0)} \frac{(x_0 - x_{-1})}{(x_1 - x_{-1})} = \frac{1}{h_1} \frac{h_0}{h_0 h_1}$$

$$f'(x_0) = f(x_0) L_0'(x_0) + f(x_{-1}) L_{-1}'(x_0) + f(x_1) L_1'(x_0)$$

$\hookrightarrow \approx P_2'(x_0)$

$$f'(x_0) = \frac{h_1 - h_0}{h_1 h_0} f(x_0) + \frac{1}{h_0 h_1} \left(\frac{h_0}{h_1} f(x_1) - \frac{h_1}{h_0} f(x_{-1}) \right)$$

eg: Insert $h_1 = h_0 \Rightarrow$ this becomes the familiar
3-pt centered formula

\Rightarrow Compare the naive approach to the principled approach in Table 14-2 :

$\sim O(h)$ $O(h^2)$

Formulas for Higher Order Derivatives:

$$f^{(k)}(x_0) \approx P^{(k)}(x_0) = \sum_{j=-l}^u f(x_j) L_j^{(k)}(x_0)$$

End of 14.3

14.4 Round-off errors \times Data errors in Differentiation. Numerical.

Ex 14.7 $f'(x_0) \approx \frac{f(x_0+h) - f(x_0-h)}{2h} = D_h$

\bar{D}_h : calculated value for D_h .

* Let $\bar{f}(x) = \text{fl}(f(x))$: floating pt approx
for $f(x)$ at each x .

$$\bar{f}(x) = f(x) + e_r(x)$$

$\underbrace{\qquad\qquad\qquad}_{\text{round-off error term.}}$

We'll assume that $|e_r(x)| \leq \varepsilon$ \Rightarrow depends on the rounding unit.

(Recall: $\left| \frac{x - \bar{x}}{|x|} \right| \leq n$) for the IEEE flpt.
 $n = 2^{-52}$

Consider the centered diff. formula: $\bar{D}_h = \frac{\bar{f}(x_0+h) - \bar{f}(x_0-h)}{2h}$

(we neglect the floating pt errors in subtraction & division operations)
for simplicity.

* Want to see how h (the spacing) affects the round-off error & truncation error & the interplay between the two errors,

$$\begin{aligned}
 \Rightarrow |\bar{D}_h - D_h| &= \left| \frac{\bar{f}(x_0+h) - \bar{f}(x_0-h)}{2h} - \frac{f(x_0+h) - f(x_0-h)}{2h} \right| \\
 &= \left| \frac{\text{er}(x_0+h) - \text{er}(x_0-h)}{2h} \right| \\
 &\leq \left| \frac{\text{er}(x_0+h)}{2h} \right| + \left| \frac{\text{er}(x_0-h)}{2h} \right| \\
 &\leq \frac{\varepsilon}{2h} \quad \left(\text{Note: The roundoff error } \text{er}(x) \text{ is not smooth, i.e. signs of } \text{er}(x_0+h) \text{ may differ from } \text{er}(x_0-h) \right)
 \end{aligned}$$

* The actual error in our approx:

$$\begin{aligned}
 |f'(x_0) - \bar{D}_h| &= |(\text{truncation error}) + (\text{round-off error})| \\
 &\leq |f'(x_0) - D_h| + |\bar{D}_h - D_h| \\
 \text{Recall } \text{truncation error} &= -\frac{h^2}{6} f'''(\xi), \quad \text{assume } |f'''(\xi)| \leq M \\
 &\Leftrightarrow \left| \frac{h^2 M}{6} + \frac{\varepsilon}{h} \right| \quad \left\{ \begin{array}{l} \text{truncation error bound} \\ \text{round-off error bound} \end{array} \right\} \quad \xi \in [x_0-h, x_0+h] \\
 \text{Derive an optimal } h: \text{ define } E(h) &\triangleq \frac{h^2 M}{6} + \frac{\varepsilon}{h} \Rightarrow E'(h) = 0 \\
 E(h) &\triangleq \frac{h^2 M}{6} + \frac{\varepsilon}{h} \Rightarrow E'(h) = 0
 \end{aligned}$$

$$\begin{aligned}
 \frac{dE}{dh} = \frac{h}{3} M - \frac{\varepsilon}{h^2} = 0 \Rightarrow h_* &= \left(\frac{3\varepsilon}{M} \right)^{1/3} \quad \text{optimal value} \\
 \text{(verify this is a minimizer } \frac{d^2 E(h_*)}{dh^2} > 0 \text{)}
 \end{aligned}$$

\Rightarrow Actual choice of h : depends on the rounding unit,
also on the order of truncation error, also on the derivative bounds.

$$(h_* = \left(\frac{3\epsilon}{M}\right)^{1/3})^{1/q+1}$$

$$|f'(x_0) - D_h| \leq \frac{h^2}{6} M + \frac{\epsilon}{ch}$$

in the round-off error
 h is in denominator

actual value of derivative

calculated value in the finite precision arithmetic

truncation error $\rightarrow O(h^2)$

— h should not be too close to η (error blow up!
(we don't know ϵ and M values)
rounding unit: too small!

\Rightarrow Rule of thumb: keep h well above $\eta^{(1/q+1)}$
for a method of accuracy q .

(Note $q=2$ for the above example)

Also Note:
— We want the truncation error to dominate the round-off error, which has an oscillating characteristic.

Summary: we've seen how the choice of h affects the truncation errors & round-off errors (their relative magnitudes).

Note that: As h is reduced, truncation error shrinks by h^2
while round-off error grows by $\frac{1}{h}$.

\therefore If h is very small, the round-off error term dominates, we don't want this to happen.
We've seen this in Fig. 14.2.

The measured error \approx truncation error + round-off errors
 $h \downarrow$ \downarrow \nearrow