

"Showing Joey..."

$$f[x_1, x_2, x_3] = \frac{f[x_2, x_3] - f[x_1, x_2]}{x_3 - x_1} \quad \begin{cases} \text{apparent} \\ \text{asymmetry in} \\ \text{the defn.} \end{cases}$$
$$= \frac{\frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1}}{x_3 - x_1}$$
$$= \frac{(x_2 - x_1)(f(x_3) - f(x_2)) - (x_3 - x_2)(f(x_2) - f(x_1))}{(x_3 - x_1)(x_3 - x_2)(x_2 - x_1)}$$
$$= \frac{x_3(f(x_1) - f(x_2)) + x_2(f(x_3) - f(x_1)) + x_1(f(x_2) - f(x_3))}{(x_1 - x_2) * (x_3 - x_1) * (x_2 - x_3)}$$

Consider $f[x_2, x_1, x_3] =$

$$= \frac{x_3(f(x_2) - f(x_1)) + x_1(f(x_3) - f(x_2)) + x_2(f(x_1) - f(x_3))}{(x_2 - x_1) * (x_3 - x_2) * (x_1 - x_3)}$$

Changed the sign in the denominator!

but changed the sign in the numerator too!

$$f[x_1, x_2, x_3] = f[x_2, x_1, x_3]$$

In fact all permutations of the arguments give the same thing! (2)

Any permutation is made up of pair-wise transpositions:

$$x_1 \ x_2 \ x_3$$

$$x_1 \ x_3 \ x_2$$

$$x_3 \ x_1 \ x_2$$

$$x_3 \ x_2 \ x_1$$

$$x_2 \ x_3 \ x_1$$

$$x_2 \ x_1 \ x_3$$

The
logarithmic
video

Theorem 5.6 (Invariance Theorem)

$f(x_1, \dots, x_n)$ is invariant
under all permutations!

(3)

In numerical differentiation one may be inclined to make h smaller & smaller, hoping for better accuracy.

However, we run into problems because of round off as we make h too small.

Let's define

$$\tilde{f}(x_0) \approx f(x_0) \quad (\text{due to round off})$$

we compute \tilde{f}

$$\begin{aligned} e(x_0) &= \tilde{f}(x_0) - f(x_0) \\ e(x_0+h) &= \tilde{f}(x_0+h) - f(x_0+h) \end{aligned} \quad \left. \begin{array}{l} \text{either} \\ \text{round off} \\ \text{error.} \end{array} \right.$$

Assume ϵ represents a bound on round off error.

$$E = \left| f'(x_0) - \frac{\tilde{f}(x_0+h) - \tilde{f}(x_0)}{h} \right| \quad \left(\begin{array}{l} \text{error in} \\ \text{the derivative} \end{array} \right)$$

$$= \left| f'(x_0) - \frac{f(x_0+h) - e(x_0+h) - (f(x_0) - e(x_0))}{h} \right|$$

$$= \left| f'(x_0) - \frac{f(x_0+h) - f(x_0)}{h} + \frac{e(x_0+h) - e(x_0)}{h} \right|$$

$$\begin{aligned}
 &= \left| \frac{h}{2} f''(s) + \frac{e(x_0+h) - e(x_0)}{h} \right| \\
 &\leq \left| \frac{h}{2} f''(s) \right| + \left| \frac{e(x_0+h) - e(x_0)}{h} \right| \\
 &\leq \frac{h}{2} M + \frac{2\varepsilon}{h}
 \end{aligned}$$

↑
 upper bound on $|f''(x)|$; assume $h > 0$

Objective: minimize this bond!

This is a function, we want
 to minimize it: 1st + 2nd
derivatives!

$$B(h) = \frac{hM}{2} + \frac{2\varepsilon}{h}$$

$$B'(h) = \frac{M}{2} - \frac{2\varepsilon}{h^2} = 0$$

demand

for an extremum

$$h^2 = \frac{4\varepsilon}{M} \rightarrow \left| h = 2\sqrt{\frac{\varepsilon}{M}} \right|$$

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So now we know how to compute derivatives using Taylor. The linear algebra approach is really nice: find weights w_1 & w_2

$$w_1 f(x+h) + w_2 f(x) = f'(x) + \text{error}$$

$$w_1 f(x+h) = w_1 (f(x)) + w_1 h f'(x) + w_1 \frac{h^2}{2} f''(\xi)$$

$$+ w_2 f(x) = w_2 f(x)$$

$$f'(x) = (w_1 + w_2) f(x) + (w_1 h) f'(x) + w_1 \frac{h^2}{2} f''(\xi) + \text{error}$$

$$\begin{aligned} \therefore w_1 + w_2 &= 0 \\ w_1 h &= 1 \end{aligned} \quad \begin{bmatrix} 1 & 1 \\ h & 0 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

linear system

What about higher derivatives?

$$\begin{aligned}
 f''(x) &\simeq \frac{f'(x+h) - f'(x)}{h} \quad \left(\text{sees symmetric } \right) \\
 &\simeq \underbrace{\frac{f(x+h) - f(x)}{h} - \frac{f(x) - f(x-h)}{h}}_h \\
 &= \boxed{\frac{f(x+h) - 2f(x) + f(x-h)}{h^2}} \quad \text{symmetry}
 \end{aligned}$$

Let's get the error of τ_{\pm} formula using Taylor series:

$$f(x+h) = f(x) + hf'(x) + \frac{h^2}{2}f''(x) + \frac{h^3}{3!}f'''(x) + \frac{h^4}{4!}f^{(iv)}(\xi)$$

$$f(x) = f(x)$$

$$f(x-h) = f(x) - hf'(x) + \frac{h^2}{2}f''(x) - \frac{h^3}{3!}f'''(x) + \frac{h^4}{4!}f^{(iv)}(\xi_-)$$

$$\begin{aligned}
 f''(x) &= \frac{f(x+h) - 2f(x) + f(x-h)}{h^2} + \underbrace{\frac{h^2}{4!} (f^{(iv)}(\xi_+) + f^{(iv)}(\xi_-))}_{\frac{h^2}{12} f^{(iv)}(\xi)}
 \end{aligned}$$

by τ_2 IVT.

Error: $\mathcal{O}(h^2)$