

# Numerical Differentiation

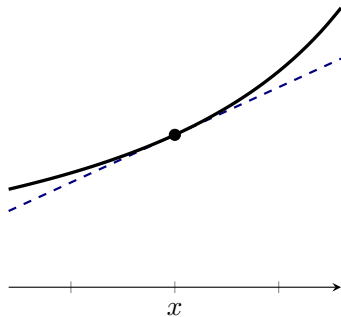
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# Introduction

# Difference Quotients to Approximate Slopes



Let  $f$  be a smooth function. Analytically, the derivative is calculated by

$$D\{f\}(x) = f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h},$$

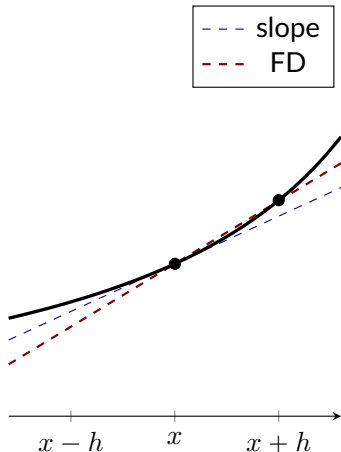
which represents the slope of the line tangent to the graph of  $f$  at  $x$ .

# Difference Quotients to Approximate Slopes (cont')

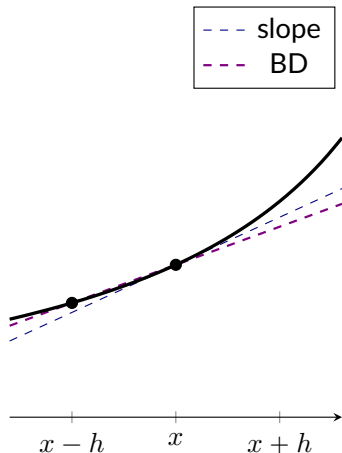
Since the definition relies on  $h$  approaching 0, choosing a small, fixed value for  $h$  approximates  $f'(x)$ .

- 1st-order forward difference

$$D_h^{[1f]}\{f\}(x) = \frac{f(x+h) - f(x)}{h}$$



# Difference Quotients to Approximate Slopes (cont')



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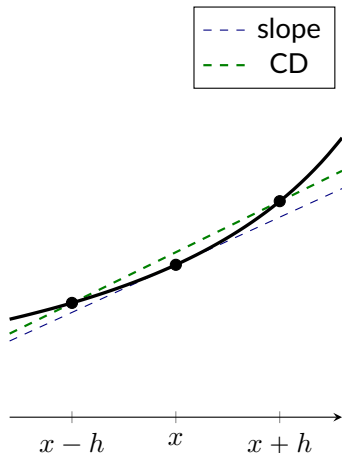
- 1st-order forward difference

$$D_h^{[1f]}\{f\}(x) = \frac{f(x+h) - f(x)}{h}$$

- 1st-order backward difference

$$D_h^{[1b]}\{f\}(x) = \frac{f(x) - f(x-h)}{h}$$

# Difference Quotients to Approximate Slopes (cont')



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- 1st-order forward difference

$$D_h^{[1f]}\{f\}(x) = \frac{f(x+h) - f(x)}{h}$$

- 1st-order backward difference

$$D_h^{[1b]}\{f\}(x) = \frac{f(x) - f(x-h)}{h}$$

- 2nd-order centered difference

$$D_h^{[2c]}\{f\}(x) = \frac{f(x+h) - f(x-h)}{2h}$$

## Difference Quotients to Approximate Slopes (cont')

The three approximation formulas presented above are examples of so-called **finite difference formulas**.

### Note

The terms *first-order* and *second-order* refer to how quickly the approximation converges to the actual value of  $f'(x)$  as  $h$  approaches 0, not to the order of differentiation. More on this later.



# Interpolation and Difference Formulas

For simplicity of notation, let's set  $x = 0$ .

Observe that

- The forward difference formula is simply the slope of the (secant) line through the two points  $(0, f(0))$  and  $(h, f(h))$ .
- Similarly, the backward difference formula is simply the slope of the (secant) line through the two points  $(0, f(0))$  and  $(h, f(h))$ .

Think:

- slope  $\Leftrightarrow$  derivative
- line through two points  $\Leftrightarrow$  interpolant

A natural extension of this perspective is to think of the centered difference formula as the derivative of the quadratic interpolant of the three points  $(-h, f(-h))$ ,  $(0, f(0))$ , and  $(h, f(h))$ .

# Interpolation and Difference Formulas (cont')

**Exercise 1.** Show that the quadratic function

$$q(x) = \frac{x(x-h)}{2h^2}f(-h) - \frac{x^2-h^2}{h^2}f(0) + \frac{x(x+h)}{2h^2}f(h).$$

interpolates  $(-h, f(-h))$ ,  $(0, f(0))$ , and  $(h, f(h))$ .

**Exercise 2.** Show that  $q'(0) = D_h^{[2c]}\{f\}(0)$ .

## Interpolation and Difference Formulas (cont')

In principle, once nodes are determined, a finite difference formula can be derived by:

*Interpolate the given function values, then differentiate the interpolant exactly.*

Some commonly used difference formulas are provided, without derivation, in the next slide.

# Common Difference Formulas

Type	Order	Notation	Formula
Forward	1	$D_h^{[1f]}\{f\}(x)$	$\frac{f(x+h) - f(x)}{h}$
	2	$D_h^{[2f]}\{f\}(x)$	$\frac{-3f(x) + 4f(x+h) - f(x+2h)}{2h}$
Backward	1	$D_h^{[1b]}\{f\}(x)$	$\frac{f(x) - f(x-h)}{h}$
	2	$D_h^{[2b]}\{f\}(x)$	$\frac{3f(x) - 4f(x-h) + f(x-2h)}{2h}$
Centered	2	$D_h^{[2c]}\{f\}(x)$	$\frac{f(x+h) - f(x-h)}{2h}$
	4	$D_h^{[4c]}\{f\}(x)$	$\frac{f(x-2h) - 8f(x-h) + 8f(x+h) - f(x+2h)}{12h}$

# Higher Derivatives

# Convergence of Difference Formulas

# Convergence of Difference Formulas

- All finite difference formulas introduced converge as  $h \rightarrow 0$ .
- But there are difficulties in implementing this limiting calculation numerically, so we need to work with  $h > 0$  which would yield acceptable accuracy.
- To address this kind of issues, we need to understand how the accuracy of difference formulas increases as  $h \rightarrow 0$ .
- In other words, we need to study how *quickly* the error  $D^{[\cdot]} \{f\}(x) - f'(x)$  diminishes as  $h \rightarrow 0$ .
- The main tool for the analysis is the Taylor series.

# First-Order Difference Formulas

The formula  $D_h^{[1f]}\{f\}$  is said to be a **first-order** method because

$$D_h^{[1f]}\{f\}(x) - f'(x) = \underbrace{\frac{1}{2}f''(x)h}_{\text{leading error}} + O(h^2).$$

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**Derivation.** Use the Taylor series of  $D_h^{[1f]}\{f\}$  at  $x$ :

$$\begin{aligned} D_h^{[1f]}\{f\}(x) - f'(x) &= \frac{f(x+h) - f(x)}{h} - f'(x) \\ &= \frac{1}{h} \left[ \cancel{f(x)} + f'(x)h + \frac{f''(x)}{2}h^2 + O(h^3) - \cancel{f(x)} \right] - f'(x) \\ &= \left[ \cancel{f'(x)} + \frac{f''(x)}{2}h + O(h^2) \right] - \cancel{f'(x)} \\ &= \frac{f''(x)}{2}h + O(h^2). \end{aligned}$$



## Second-Order Difference Formulas

The formula  $D_h^{[2c]}\{f\}$  is said to be a **second-order** method because

$$D_h^{[2c]}\{f\}(x) - f'(x) = \underbrace{\frac{1}{6}f'''(x)h^2}_{\text{leading error}} + O(h^4).$$

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**Derivation.** Exercise.

## Remarks

- The difference  $D_h^{[\cdot]} \{f\}(x) - f'(x)$  is called the **truncation error**. This term is derived from the idea that the finite difference formula has to truncate the series representation and thus cannot be exactly correct for all functions.
- When using a first-order formula, cutting  $h$  in half should reduce the error in the approximation of  $f'(x)$  by about half, when  $h$  is sufficiently small.
- In general, when using an  $n$ th-order formula, reducing  $h$  by a factor of two should cut the error in  $f'(x)$  estimate by a factor of about  $2^n$ , when  $h$  is sufficiently small.
- The notation  $O(f(n))$  is commonly used to describe the temporal or spatial complexity of an algorithm. In that context, a  $O(n^2)$  algorithm is much worse than a  $O(n)$  algorithm. (Remember flop counts for LU factorization.) However, when referring to error, a  $O(h^2)$  algorithm is **better** than a  $O(h)$  algorithm because it means that the accuracy improves faster as  $h$  decreases.

Comparison of convergence.

# Determining Optimal $h$

The difference formulas are inherently ill-conditioned as  $h \rightarrow 0$  due to catastrophic cancellation when implemented in floating point arithmetic. As an example, consider the numerical evaluation<sup>1</sup> of the centered difference formula:

$$\begin{aligned}\hat{D}_h^{[2c]}\{f\}(x) &= \frac{\widehat{f(x+h)} - \widehat{f(x-h)}}{2h} \\ &= \frac{f(x+h)(1+\epsilon_+) - f(x-h)(1+\epsilon_-)}{2h}, \quad \text{for some } |\epsilon_{\pm}| \leq \frac{1}{2} \boxed{\text{eps}} \\ &= \frac{f(x+h) - f(x-h)}{2h} + \frac{f(x+h)\epsilon_+ - f(x-h)\epsilon_-}{2h} \\ &= \left[ f'(x) + \frac{1}{6}f'''(x)h^2 + O(h^4) \right] + \frac{(f(x) + O(h))\epsilon_+ - (f(x) + O(h))\epsilon_-}{2h} \\ &= f'(x) + \frac{1}{6}f'''(x)h^2 + \frac{\epsilon_+ - \epsilon_-}{2h}f(x) + O(\boxed{\text{eps}}) + O(h^4).\end{aligned}$$

So the error is bounded from above by:

$$\left| \hat{D}_h^{[2c]}\{f\}(x) - f'(x) \right| \leq \frac{1}{6} |f'''(x)| h^2 + \frac{\boxed{\text{eps}}}{2h} |f(x)| + O(\boxed{\text{eps}}) + O(h^4).$$

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<sup>1</sup>For simplicity, we only consider round-off errors arising in the evaluation of  $f$ .

## Determining Optimal $h$ (cont')

Ignoring the last two terms, the error bound consists of two parts:

$$|\text{error}| \lesssim \underbrace{\frac{1}{6} |f'''(x)| h^2}_{\text{truncation error}} + \underbrace{\frac{\boxed{\text{eps}}}{2h} |f(x)|}_{\text{round-off error}} = \alpha h^2 + \frac{\beta}{h} \boxed{\text{eps}} =: g(h).$$

Using calculus, one can find that  $g(h)$  is minimized when

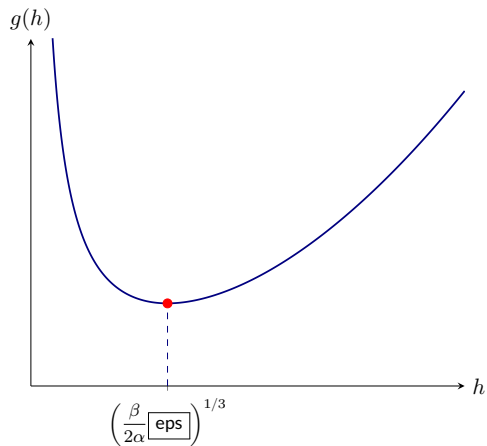
$$h = \left( \frac{\beta}{2\alpha} \boxed{\text{eps}} \right)^{1/3} = \left( \frac{3}{2} \left| \frac{f(x)}{f'''(x)} \right| \right)^{1/3} \boxed{\text{eps}}^{1/3} = O(\boxed{\text{eps}}^{1/3}),$$

in which case

$$\left| \hat{D}_h^{[2c]} \{f\}(x) - f'(x) \right| \lesssim \left( \frac{9}{32} |f^2(x) f'''(x)| \right)^{1/3} \boxed{\text{eps}}^{2/3} = O(\boxed{\text{eps}}^{2/3}).$$

**Exercise.** Repeat the analysis above to determine the optimal  $h$  for a first-order accurate difference formula. How about a general  $n$ th-order method?

## Determining Optimal $h$ (cont')



The effect of round-off error on numerical differentiation. For sufficiently small  $h$ , the error is dominated by rounding error.

# Richardson Extrapolation

**Set-up.** Let  $V_h$  be a numerical approximation to the analytical value  $V$  such that

$$V_h = V + \underbrace{c_1 h^{p_1} + c_2 h^{p_2} + c_3 h^{p_3} + \cdots}_{\text{discretization error}},$$

where  $p_1 < p_2 < p_3 < \cdots$  are positive integers and  $c_1, c_2, c_3, \dots$  are constant. Here  $h$  is some discretization of the analytical calculation, and as  $h \rightarrow 0$ ,  $V_h \rightarrow V$ .

**Goal.** Construct a higher-order accurate method approximating  $V$  out of a lower-order accurate method  $V_h$ .

**Procedure.** (*Richardson extrapolation*) Form a suitable linear combination of  $V_h$  and  $V_{h/2}$  approximating  $V$  which removes the leading error term  $h^{p_1}$ .

$$\text{Result: } \frac{2^{p_1}}{2^{p_1} - 1} V_{h/2} - \frac{1}{2^{p_1} - 1} V_h = V + O(h^{p_2})$$

# Richardson Extrapolation: Illustration of Method

Write down

$$V_h = V + c_1 h^{p_1} + c_2 h^{p_2} + c_3 h^{p_3} + \dots \quad (1)$$

$$V_{h/2} = V + c_1 \left(\frac{h}{2}\right)^{p_1} + c_2 \left(\frac{h}{2}\right)^{p_2} + c_3 \left(\frac{h}{2}\right)^{p_3} + \dots \quad (2)$$

Multiplying (1) by  $a$  and (2) by  $b$  and summing up, we obtain

$$aV_h + bV_{h/2} = \underbrace{(a+b)}_{=1} V + c_1 \underbrace{\left[a + \frac{b}{2^{p_1}}\right]}_{=0} h^{p_1} + c_2 \left[a + \frac{b}{2^{p_2}}\right] h^{p_2} + \dots$$

Find  $a$  and  $b$  satisfying

$$\begin{cases} a + b = 1 \\ a + \frac{b}{2^{p_1}} = 0 \end{cases} \implies \begin{cases} a = -\frac{1}{2^{p_1} - 1} \\ b = \frac{2^{p_1}}{2^{p_1} - 1} \end{cases}$$



## Another Derivation of 2nd-Order Forward Difference

**Exercise.** Derive  $D_h^{[2f]}\{f\}(x)$  using Richardson extrapolation on  $V_h = D_h^{[1f]}\{f\}(x)$ .