

# Numerical Analysis

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## Lecture 10: Numerical Differentiation

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# Outline

- 1 Numerical Differentiation
- 2 General Derivative Approximation Formulas
- 3 Three-point Derivative Approximation Formulas
- 4 Numerical Approximations to Higher Derivatives
- 5 Round-Off Error Instability
- 6 Richardson's Extrapolation

# Introduction to Numerical Differentiation

## Approximating a Derivative

- The derivative of the function  $f$  at  $x_0$  is

$$f'(x_0) = \lim_{h \rightarrow 0} \frac{f(x_0 + h) - f(x_0)}{h}$$

- This formula gives an obvious way to generate an approximation to  $f'(x_0)$ ; simply compute

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

for small values of  $h$ . Although this may be obvious, it is not very successful, due to our old nemesis round-off error.

- But it is certainly a place to start.

# Introduction to Numerical Differentiation

## Approximating a Derivative (cont'd)

- To approximate  $f'(x_0)$ , suppose first that  $x_0 \in (a, b)$ , where  $f \in C^2[a, b]$ , and that  $x_1 = x_0 + h$  for some  $h \neq 0$  that is sufficiently small to ensure that  $x_1 \in [a, b]$ .
- We construct **the first Lagrange polynomial**  $P_{0,1}(x)$  for  $f$  determined by  $x_0$  and  $x_1$ , with its error term:

$$\begin{aligned} f(x) &= P_{0,1}(x) + \frac{(x-x_0)(x-x_1)}{2!} f''(\xi(x)) \\ &= \frac{f(x_0)(x-x_0-h)}{-h} + \frac{f(x_0+h)(x-x_0)}{h} + \frac{(x-x_0)(x-x_0-h)}{2!} f''(\xi(x)) \end{aligned}$$

for some  $\xi(x)$  between  $x_0$  and  $x_1$ .

# Introduction to Numerical Differentiation

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

Differentiating gives

$$\begin{aligned}f'(x) &= \frac{f(x_0 + h) - f(x_0)}{h} + D_x \left[ \frac{(x - x_0)(x - x_0 - h)}{2!} f''(\xi(x)) \right] \\&= \frac{f(x_0 + h) - f(x_0)}{h} + \frac{2(x - x_0) - h}{2} f''(\xi(x)) \\&\quad + \frac{(x - x_0)(x - x_0 - h)}{2} D_x(f''(\xi(x)))\end{aligned}$$

Deleting the terms involving  $\xi(x)$  gives

$$f'(x) \approx \frac{f(x_0 + h) - f(x_0)}{h}$$

# Introduction to Numerical Differentiation

$$f'(x) \approx \frac{f(x_0 + h) - f(x_0)}{h}$$

## Approximating a Derivative (cont'd)

- One difficulty with this formula is that we have no information about  $D_x(f''(\xi(x)))$ , so the truncation error cannot be estimated.
- When  $x$  is  $x_0$ , however, the coefficient of  $D_x(f''(\xi(x)))$  is 0, and the formula simplifies to

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

# Introduction to Numerical Differentiation

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

## Forward-Difference and Backward-Difference Formulate

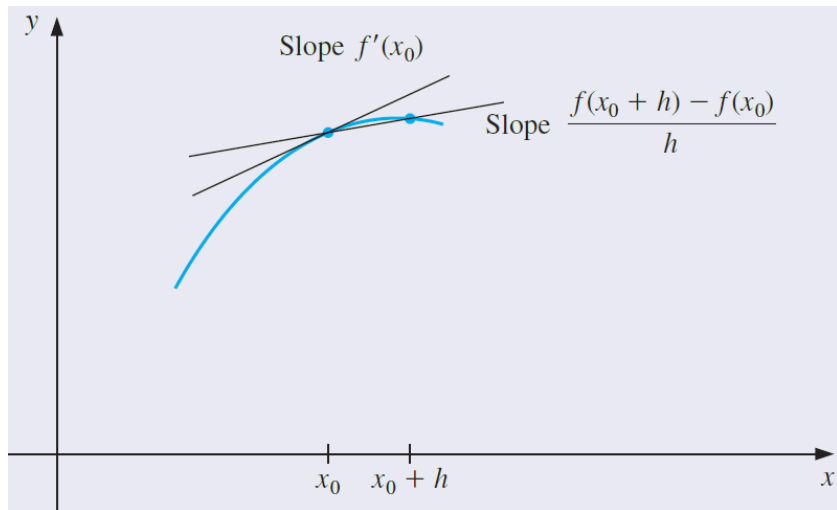
- For small values of  $h$ , the difference quotient

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

can be used to approximate  $f'(x_0)$  with **an error bounded by  $M|h|/2$** , where  $M$  is a bound on  $|f''(x)|$  for  $x$  between  $x_0$  and  $x_0 + h$ .

- This formula is known as the **forward-difference formula** if  $h > 0$  and the **backward-difference formula** if  $h < 0$ .

# Introduction to Numerical Differentiation





# Introduction to Numerical Differentiation

## Example 1: $f(x) = \ln x$

Use the forward-difference formula to approximate the derivative of  $f(x) = \ln x$  at  $x_0 = 1.8$  using  $h = 0.1$ ,  $h = 0.05$ , and  $h = 0.01$ , and determine bounds for the approximation errors.

# Introduction to Numerical Differentiation

## Example 1: $f(x) = \ln x$

Use the forward-difference formula to approximate the derivative of  $f(x) = \ln x$  at  $x_0 = 1.8$  using  $h = 0.1$ ,  $h = 0.05$ , and  $h = 0.01$ , and determine bounds for the approximation errors.

## Solution (1/3)

The forward-difference formula

$$\frac{f(1.8 + h) - f(1.8)}{h}$$

with  $h = 0.1$  gives

$$\frac{\ln 1.9 - \ln 1.8}{0.1} = \frac{0.64185389 - 0.58778667}{0.1} = 0.5406722$$

# Introduction to Numerical Differentiation

## Solution (2/3)

Because  $f''(x) = -1/x^2$  and  $1.8 < \xi < 1.9$ , a bound for this approximation error is

$$\frac{|hf''(\xi)|}{2} = \frac{|h|}{2\xi^2} < \frac{0.1}{2(1.8)^2} = 0.0154321$$

The approximation and error bounds when  $h = 0.05$  and  $h = 0.01$  are found in a similar manner and the results are shown in the following table.

# Introduction to Numerical Differentiation

## Solution (3/3)

$h$	$f(1.8 + h)$	$\frac{f(1.8 + h) - f(1.8)}{h}$	$\frac{ h }{2(1.8)^2}$
0.1	0.64185389	0.5406722	0.0154321
0.05	0.61518564	0.5479795	0.0077160
0.01	0.59332685	0.5540180	0.0015432

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# General Derivative Approximation Formulas

## Method of Construction

- To obtain general derivative approximation formulas, suppose that  $\{x_0, x_1, \dots, x_n\}$  are  $(n + 1)$  distinct numbers in some interval  $I$  and that  $f \in C^{n+1}(I)$ .
- From the interpolation error theorem, we have

$$f(x) = \sum_{k=0}^n f(x_k) L_k(x) + \frac{(x - x_0) \cdots (x - x_n)}{(n + 1)!} f^{(n+1)}(\xi(x))$$

for some  $\xi(x)$  in  $I$ , where  $L_k$  denotes the  $k$ th Lagrange coefficient polynomial for  $f$  at  $x_0, x_1, \dots, x_n$ .

# General Derivative Approximation Formulas

$$f(x) = \sum_{k=0}^n f(x_k) L_k(x) + \frac{(x - x_0) \cdots (x - x_n)}{(n + 1)!} f^{(n+1)}(\xi(x))$$

## Method of Construction (Cont'd)

Differentiating this expression gives

$$\begin{aligned} f'(x) = & \sum_{k=0}^n f(x_k) L'_k(x) + D_x \left[ \frac{(x - x_0) \cdots (x - x_n)}{(n + 1)!} \right] f^{(n+1)}(\xi(x)) \\ & + \frac{(x - x_0) \cdots (x - x_n)}{(n + 1)!} D_x [f^{(n+1)}(\xi(x))] \end{aligned}$$

# General Derivative Approximation Formulas

$$f'(x) = \sum_{k=0}^n f(x_k) L'_k(x) + D_x \left[ \frac{(x-x_0) \cdots (x-x_n)}{(n+1)!} \right] f^{(n+1)}(\xi(x)) \\ + \frac{(x-x_0) \cdots (x-x_n)}{(n+1)!} D_x [f^{(n+1)}(\xi(x))]$$

## Method of Construction (Cont'd)

We again have a problem estimating the truncation error unless  $x$  is one of the numbers  $x_j$ . In this case, the term multiplying  $D_x[f^{(n+1)}(\xi(x))]$  is 0, and the formula becomes

$$f'(x_j) = \sum_{k=0}^n f(x_k) L'_k(x_j) + \frac{f^{(n+1)}(\xi(x_j))}{(n+1)!} \prod_{k=0, k \neq j}^n (x_j - x_k)$$

which is called an  **$(n+1)$ -point formula** to approximate  $f'(x_j)$ .



# General Derivative Approximation Formulas

$$f'(x_j) = \sum_{k=0}^n f(x_k) L'_k(x_j) + \frac{f^{(n+1)}(\xi(x_j))}{(n+1)!} \prod_{k=0, k \neq j}^n (x_j - x_k)$$

## Comment on the $(n+1)$ -point Formula

- In general, using more evaluation points produces greater accuracy, although the number of functional evaluations and growth of round-off error discourages this somewhat.
- The most common formulas are those involving three and five evaluation points.

We first derive some useful three-point formulas and consider aspects of their errors.

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# Three-point Formulas

## Important Building Blocks

Since

$$L_0(x) = \frac{(x - x_1)(x - x_2)}{(x_0 - x_1)(x_0 - x_2)}$$

we obtain

$$L'_0(x) = \frac{2x - x_1 - x_2}{(x_0 - x_1)(x_0 - x_2)}$$

In a similar way, we find that

$$L'_1(x) = \frac{2x - x_0 - x_2}{(x_1 - x_0)(x_1 - x_2)}$$

$$L'_2(x) = \frac{2x - x_0 - x_1}{(x_2 - x_0)(x_2 - x_1)}$$

# Three-point Formulas

## Important Building Blocks (Cont'd)

Using these expressions for  $L'_j(x)$ ,  $1 \leq j \leq 2$ , the  $n + 1$ -point formula

$$f'(x_j) = \sum_{k=0}^n f(x_k) L'_k(x_j) + \frac{f^{(n+1)}(\xi(x_j))}{(n+1)!} \prod_{k=0, k \neq j}^n (x_j - x_k)$$

becomes for  $n = 2$ :

$$\begin{aligned} f'(x_j) = & f(x_0) \left[ \frac{2x_j - x_1 - x_2}{(x_0 - x_1)(x_0 - x_2)} \right] + f(x_1) \left[ \frac{2x_j - x_0 - x_2}{(x_1 - x_0)(x_1 - x_2)} \right] \\ & + f(x_2) \left[ \frac{2x_j - x_0 - x_1}{(x_2 - x_0)(x_2 - x_1)} \right] + \frac{1}{6} f^{(3)}(\xi(x_j)) \prod_{k=0, k \neq j}^2 (x_j - x_k) \end{aligned}$$

for each  $j = 0, 1, 2$ , where  $\xi_j = \xi_j(x)$ .

# Three-point Formulas

$$\begin{aligned} f'(x_j) = & f(x_0) \left[ \frac{2x_j - x_1 - x_2}{(x_0 - x_1)(x_0 - x_2)} \right] + f(x_1) \left[ \frac{2x_j - x_0 - x_2}{(x_1 - x_0)(x_1 - x_2)} \right] \\ & + f(x_2) \left[ \frac{2x_j - x_0 - x_1}{(x_2 - x_0)(x_2 - x_1)} \right] + \frac{1}{6} f^{(3)}(\xi(x_j)) \prod_{k=0, k \neq j}^2 (x_j - x_k) \end{aligned}$$

## Assumption

The 3-point formulas become especially useful if the nodes are equally spaced, that is, when

$$x_1 = x_0 + h \text{ and } x_2 = x_0 + 2h, \text{ for some } h \neq 0$$

We will assume equally-spaced nodes throughout the remainder of this section.

# Three-point Formulas

$$\begin{aligned} f'(x_j) = & f(x_0) \left[ \frac{2x_j - x_1 - x_2}{(x_0 - x_1)(x_0 - x_2)} \right] + f(x_1) \left[ \frac{2x_j - x_0 - x_2}{(x_1 - x_0)(x_1 - x_2)} \right] \\ & + f(x_2) \left[ \frac{2x_j - x_0 - x_1}{(x_2 - x_0)(x_2 - x_1)} \right] + \frac{1}{6} f^{(3)}(\xi(x_j)) \prod_{k=0, k \neq j}^2 (x_j - x_k) \end{aligned}$$

## Three-Point Formulas (1/3)

With  $x_j = x_0$ ,  $x_1 = x_0 + h$ , and  $x_2 = x_0 + 2h$ , the general 3-point formula becomes

$$f'(x_0) = \frac{1}{h} \left[ -\frac{3}{2}f(x_0) + 2f(x_1) - \frac{1}{2}f(x_2) \right] + \frac{h^2}{3} f^{(3)}(\xi_0)$$

# Three-point Formulas

$$\begin{aligned} f'(x_j) = & f(x_0) \left[ \frac{2x_j - x_1 - x_2}{(x_0 - x_1)(x_0 - x_2)} \right] + f(x_1) \left[ \frac{2x_j - x_0 - x_2}{(x_1 - x_0)(x_1 - x_2)} \right] \\ & + f(x_2) \left[ \frac{2x_j - x_0 - x_1}{(x_2 - x_0)(x_2 - x_1)} \right] + \frac{1}{6} f^{(3)}(\xi(x_j)) \prod_{k=0, k \neq j}^2 (x_j - x_k) \end{aligned}$$

## Three-Point Formulas (2/3)

Doing the same for  $x_j = x_1$  gives

$$f'(x_1) = \frac{1}{h} \left[ -\frac{1}{2}f(x_0) + \frac{1}{2}f(x_2) \right] - \frac{h^2}{6} f^{(3)}(\xi_1)$$

# Three-point Formulas

$$\begin{aligned} f'(x_j) = & f(x_0) \left[ \frac{2x_j - x_1 - x_2}{(x_0 - x_1)(x_0 - x_2)} \right] + f(x_1) \left[ \frac{2x_j - x_0 - x_2}{(x_1 - x_0)(x_1 - x_2)} \right] \\ & + f(x_2) \left[ \frac{2x_j - x_0 - x_1}{(x_2 - x_0)(x_2 - x_1)} \right] + \frac{1}{6} f^{(3)}(\xi(x_j)) \prod_{k=0, k \neq j}^2 (x_j - x_k) \end{aligned}$$

## Three-Point Formulas (3/3)

... and for  $x_j = x_2$ , we obtain

$$f'(x_2) = \frac{1}{h} \left[ \frac{1}{2} f(x_0) - 2f(x_1) + \frac{3}{2} f(x_2) \right] + \frac{h^2}{6} f^{(3)}(\xi_2)$$



# Three-point Formulas

## Three-point Formulas: Further Simplification

Since  $x_1 = x_0 + h$ , and  $x_2 = x_0 + 2h$ , these formulas can also be expressed as

$$f'(x_0) = \frac{1}{h} \left[ -\frac{3}{2}f(x_0) + 2f(x_0 + h) - \frac{1}{2}f(x_0 + 2h) \right] + \frac{h^2}{3}f^{(3)}(\xi_0)$$

$$f'(x_0 + h) = \frac{1}{h} \left[ -\frac{1}{2}f(x_0) + \frac{1}{2}f(x_0 + 2h) \right] - \frac{h^2}{6}f^{(3)}(\xi_1)$$

$$f'(x_0 + 2h) = \frac{1}{h} \left[ \frac{1}{2}f(x_0) - 2f(x_0 + h) + \frac{3}{2}f(x_0 + 2h) \right] + \frac{h^2}{3}f^{(3)}(\xi_2)$$

As a matter of convenience, the variable substitution  $x_0$  for  $x_0 + h$  is used in the middle equation to change this formula to an approximation for  $f'(x_0)$ . A similar change,  $x_0$  for  $x_0 + 2h$ , is used in the last equation.

# Three-point Formulas

## Three-point Formulas: Further Simplification (Cont'd)

This gives three formulas for approximating  $f'(x_0)$ :

$$f'(x_0) = \frac{1}{2h} [-3f(x_0) + 4f(x_0 + h) - f(x_0 + 2h)] + \frac{h^2}{3}f^{(3)}(\xi_0)$$

$$f'(x_0) = \frac{1}{2h} [-f(x_0 - h) + f(x_0 + h)] - \frac{h^2}{6}f^{(3)}(\xi_1)$$

$$f'(x_0) = \frac{1}{2h} [f(x_0 - 2h) - 4f(x_0 - h) + 3f(x_0)] + \frac{h^2}{3}f^{(3)}(\xi_2)$$

Finally, note that the last of these equations can be obtained from the first by simply replacing  $h$  with  $-h$ , so there are actually only two formulas.

# Three-point Formulas

## Three-Point Endpoint Formula

$$f'(x_0) = \frac{1}{2h} [-3f(x_0) + 4f(x_0 + h) - f(x_0 + 2h)] + \frac{h^2}{3} f^{(3)}(\xi_0)$$

where  $\xi_0$  lies between  $x_0$  and  $x_0 + 2h$ .

## Three-Point Midpoint Formula

$$f'(x_0) = \frac{1}{2h} [f(x_0 + h) - f(x_0 - h)] - \frac{h^2}{6} f^{(3)}(\xi_1)$$

where  $\xi_1$  lies between  $x_0 - h$  and  $x_0 + h$ .

# Three-point Formulas

$$(1) \quad f'(x_0) = \frac{1}{2h} [-3f(x_0) + 4f(x_0 + h) - f(x_0 + 2h)] + \frac{h^2}{3} f^{(3)}(\xi_0)$$

$$(2) \quad f'(x_0) = \frac{1}{2h} [f(x_0 + h) - f(x_0 - h)] - \frac{h^2}{6} f^{(3)}(\xi_1)$$

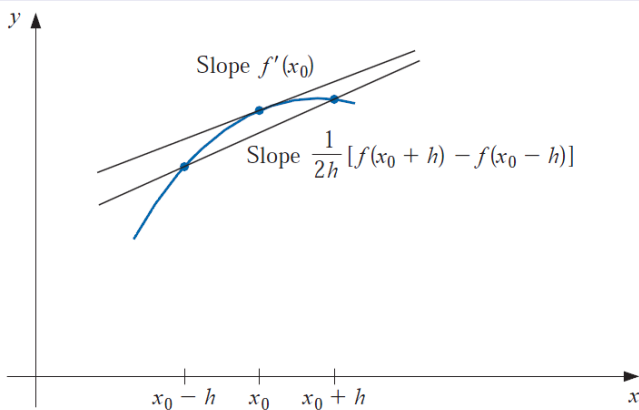
## Comments

- Although the errors in both equations are  $O(h^2)$ , the error in (2) is approximately half the error in (1).
- This is because (2) uses data on both sides of  $x_0$  and (1) uses data on only one side.
- Note also that  $f$  needs to be evaluated at only two points in (2), whereas in (1) three evaluations are needed.

# Three-point Formulas

$$f'(x_0) = \frac{1}{2h} [f(x_0 + h) - f(x_0 - h)] - \frac{h^2}{6} f^{(3)}(\xi_1)$$

where  $\xi_1$  lies between  $x_0 - h$  and  $x_0 + h$ .



# Five-Point Formulas

## Five-Point Endpoint Formula

$$f'(x_0) = \frac{1}{12h} [-25f(x_0) + 48f(x_0 + h) - 36f(x_0 + 2h) \\ + 16f(x_0 + 3h) - 3f(x_0 + 4h)] + \frac{h^4}{5}f^{(5)}(\xi)$$

where  $\xi$  lies between  $x_0$  and  $x_0 + 4h$ .

## Five-Point Midpoint Formula

$$f'(x_0) = \frac{1}{12h} [f(x_0 - 2h) - 8f(x_0 - h) + 8f(x_0 + h) - f(x_0 + 2h)] + \frac{h^4}{30}f^{(5)}(\xi)$$

where  $\xi$  lies between  $x_0 - 2h$  and  $x_0 + 2h$ .

# Three-Point vs. Five-Point

## Example

Three-Point and Five-Point formulas to approximate  $f(x) = xe^x$  at  $x = 2.0$ .

- Three-point endpoint with  $h = 0.1$ :  $1.35 \times 10^{-1}$
- Three-point endpoint with  $h = -0.1$ :  $1.13 \times 10^{-1}$
- Three-point midpoint with  $h = 0.1$ :  $-6.16 \times 10^{-2}$
- Three-point midpoint with  $h = 0.2$ :  $-2.47 \times 10^{-1}$
- Five-point midpoint with  $h = 0.1$ :  $1.69 \times 10^{-4}$

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# Numerical Approximations to Higher Derivatives

## Illustrative Method of Construction

Expand a function  $f$  in a third Taylor polynomial about a point  $x_0$  and evaluate at  $x_0 + h$  and  $x_0 - h$ . Then

$$f(x_0 + h) = f(x_0) + f'(x_0)h + \frac{1}{2}f''(x_0)h^2 + \frac{1}{6}f'''(x_0)h^3 + \frac{1}{24}f^{(4)}(\xi_1)h^4$$

and

$$f(x_0 - h) = f(x_0) - f'(x_0)h + \frac{1}{2}f''(x_0)h^2 - \frac{1}{6}f'''(x_0)h^3 + \frac{1}{24}f^{(4)}(\xi_{-1})h^4$$

where  $x_0 - h < \xi_{-1} < x_0 < \xi_1 < x_0 + h$

# Numerical Approximations to Higher Derivatives

$$f(x_0 + h) = f(x_0) + f'(x_0)h + \frac{1}{2}f''(x_0)h^2 + \frac{1}{6}f'''(x_0)h^3 + \frac{1}{24}f^{(4)}(\xi_1)h^4$$

$$f(x_0 - h) = f(x_0) - f'(x_0)h + \frac{1}{2}f''(x_0)h^2 - \frac{1}{6}f'''(x_0)h^3 + \frac{1}{24}f^{(4)}(\xi_{-1})h^4$$

## Illustrative Method of Construction (Cont'd)

Adding these equations, the terms involving  $f'(x_0)$  cancel

$$f(x_0 + h) + f(x_0 - h) = 2f(x_0) + f''(x_0)h^2 + \frac{1}{24} [f^{(4)}(\xi_1) + f^{(4)}(\xi_{-1})] h^4$$

Solving this equation for  $f''(x_0)$  gives

$$f''(x_0) = \frac{1}{h^2} [f(x_0 - h) - 2f(x_0) + f(x_0 + h)] - \frac{h^2}{24} [f^{(4)}(\xi_1) + f^{(4)}(\xi_{-1})]$$

# Numerical Approximations to Higher Derivatives

$$f''(x_0) = \frac{1}{h^2} [f(x_0 - h) - 2f(x_0) + f(x_0 + h)] - \frac{h^2}{24} [f^{(4)}(\xi_1) + f^{(4)}(\xi_{-1})]$$

## Illustrative Method of Construction (Cont'd)

Suppose  $f^{(4)}$  is continuous on  $[x_0 - h, x_0 + h]$ . Since  $\frac{1}{2} [f^{(4)}(\xi_1) + f^{(4)}(\xi_{-1})]$  is between  $f^{(4)}(\xi_1)$  and  $f^{(4)}(\xi_{-1})$ , the Intermediate Value Theorem implies that a number  $\xi$  exists between  $\xi_1$  and  $\xi_{-1}$ , and hence in  $(x_0 - h, x_0 + h)$ , with

$$f^{(4)}(\xi) = \frac{1}{2} [f^{(4)}(\xi_1) + f^{(4)}(\xi_{-1})]$$

This permits us to rewrite the formula in its final form:

# Numerical Approximations to Higher Derivatives

$$f''(x_0) = \frac{1}{h^2} [f(x_0 - h) - 2f(x_0) + f(x_0 + h)] - \frac{h^2}{24} [f^{(4)}(\xi_1) + f^{(4)}(\xi_{-1})]$$

## Second Derivative Midpoint Formula

$$f''(x_0) = \frac{1}{h^2} [f(x_0 - h) - 2f(x_0) + f(x_0 + h)] - \frac{h^2}{12} f^{(4)}(\xi)$$

for some  $\xi \in [x_0 - h, x_0 + h]$ .

Note: If  $f^{(4)}$  is continuous on  $[x_0 - h, x_0 + h]$ , then it is also bounded, and the approximation is  $O(h^2)$ .

# Numerical Approximations to Higher Derivatives

## Example

Values for  $f(x) = xe^x$  are given in the following table:

$x$	1.8	1.9	2.0	2.1	2.2
$f(x)$	10.889365	12.703199	14.778112	17.148957	19.855030

Use the second derivative midpoint formula approximate  $f''(2.0)$ .

# Numerical Approximations to Higher Derivatives

## Example

The data permits us to determine two approximations for  $f''(2.0)$ . Using the formula with  $h = 0.1$  gives

$$\frac{1}{0.01} [f(1.9) - 2f(2.0) + f(2.1)] =$$

$$100[12.703199 - 2(14.778112) + 17.148957] = 29.593200$$

and using the formula with  $h = 0.2$  gives

$$\frac{1}{0.04} [f(1.8) - 2f(2.0) + f(2.2)] =$$

$$25[10.889365 - 2(14.778112) + 19.855030] = 29.704275$$

The exact value is  $f''(2.0) = 29.556224$ . Hence the actual errors are  $-3.70 \times 10^{-2}$  and  $-1.48 \times 10^{-1}$ , respectively.

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- 6 Richardson's Extrapolation

# Round-Off Error Instability

## Concept of Total Error

- It is particularly important to pay attention to round-off error when approximating derivatives.
- To illustrate the situation, let us examine the three-point midpoint formula:

$$f'(x_0) = \frac{1}{2h} [f(x_0 + h) - f(x_0 - h)] - \frac{h^2}{6} f^{(3)}(\xi_1)$$

more closely.

- Suppose that in evaluating  $f(x_0 + h)$  and  $f(x_0 - h)$ , we encounter round-off errors  $e(x_0 + h)$  and  $e(x_0 - h)$ .



# Round-Off Error Instability

## Concept of Total Error (Cont'd)

- Then our computations actually use the values  $\tilde{f}(x_0 + h)$  and  $\tilde{f}(x_0 - h)$ , which are related to the true values  $f(x_0 + h)$  and  $f(x_0 - h)$  by

$$f(x_0 + h) = \tilde{f}(x_0 + h) + e(x_0 + h), \quad f(x_0 - h) = \tilde{f}(x_0 - h) + e(x_0 - h)$$

- The total error in the approximation,

$$f'(x_0) - \frac{\tilde{f}(x_0 + h) - \tilde{f}(x_0 - h)}{2h} = \frac{e(x_0 + h) - e(x_0 - h)}{2h} - \frac{h^2}{6} f^{(3)}(\xi_1)$$

is due both to round-off error, the first part, and to truncation error.

# Round-Off Error Instability

$$f'(x_0) - \frac{\tilde{f}(x_0 + h) - \tilde{f}(x_0 - h)}{2h} = \frac{e(x_0 + h) - e(x_0 - h)}{2h} - \frac{h^2}{6} f^{(3)}(\xi_1)$$

## Concept of Total Error (Cont'd)

If we assume that the round-off error  $e(x_0 \pm h)$  are bounded by some number  $\epsilon > 0$  and that the third derivative of  $f$  is bounded by a number  $M > 0$ , then

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

# Round-Off Error Instability

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

## Concept of Total Error (Cont'd)

- To reduce the truncation error,  $h^2M/6$ , we need to reduce  $h$ .
- But as  $h$  is reduced, the round-off error  $\epsilon/h$  grows.
- In practice, then, it is seldom advantageous to let  $h$  be too small because, in that case, the round-off error will dominate the calculations.

# Round-Off Error Instability

## Example

Consider using the values in the following table

$x$	$\sin x$	$x$	$\sin x$
0.800	0.71736	0.901	0.78395
0.850	0.75128	0.902	0.78457
0.880	0.77074	0.905	0.78643
0.890	0.77707	0.910	0.78950
0.895	0.78021	0.920	0.79560
0.898	0.78208	0.950	0.81342
0.899	0.78270	1.000	0.84147

to approximate  $f'(0.900)$ , where  $f(x) = \sin x$ . The true value is  $\cos 0.900 = 0.62161$ .

# Round-Off Error Instability

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

## Solution (1/4)

The formula

$$f'(0.900) \approx \frac{f(0.900 + h) - f(0.900 - h)}{2h}$$

with different values of  $h$ , gives the approximations in the following table.

# Round-Off Error Instability

## Solution (3/4)

$h$	Approximation to $f'(0.900)$	Error
0.001	0.62500	0.00339
0.002	0.62250	0.00089
0.005	0.62200	0.00039
0.010	0.62150	-0.00011
0.020	0.62150	-0.00011
0.050	0.62140	-0.00021
0.100	0.62055	-0.00106

The optimal choice for  $h$  appears to lie between 0.005 and 0.05.

# Round-Off Error Instability

## Solution (3/4)

We can use calculus to verify that a minimum for

$$e(h) = \frac{\epsilon}{h} + \frac{h^2}{6}M,$$

occurs at  $h = \sqrt[3]{3\epsilon/M}$ , where

$$M = \max_{x \in [0.800, 1.000]} |f'''(x)| = \max_{x \in [0.800, 1.000]} |\cos x| = \cos 0.8 \approx 0.69671.$$

Because values of  $f$  are given to five decimal places, we will assume that the round-off error is bounded by  $\epsilon = 5 \times 10^{-6}$ .

# Round-Off Error Instability

## Solution (4/4)

Therefore, the optimal choice of  $h$  is approximately

$$h = \sqrt[3]{3\epsilon/M} = \sqrt[3]{\frac{3(0.000005)}{0.69671}} \approx 0.028,$$

which is consistent with the results in the earlier table.

- In practice, we cannot compute an optimal  $h$  to use in approximating the derivative, since we have no knowledge of the third derivative of the function.
- But we must remain aware that reducing the step size will not always improve the approximation.



# Round-Off Error Instability

## Concluding Remarks

- We have considered only the round-off error problems that are presented by the three-point midpoint formula, but similar difficulties occur with all the differentiation formulas.
- The reason can be traced to the need to divide by a power of  $h$ .
- Division by small numbers tends to exaggerate round-off error, and this operation should be avoided if possible.
- In the case of numerical differentiation, we cannot avoid the problem entirely, although the higher-order methods reduce the difficulty.

# Round-Off Error Instability

## Concluding Remarks

- As approximation methods, numerical differentiation is **unstable**, since the small values of  $h$  needed to reduce truncation error also cause the round-off error to grow.
- This is the first class of unstable methods we have encountered, and these techniques would be avoided if it were possible.
- However, in addition to being used for computational purposes, the formulas are needed for approximating the solutions of ordinary and partial-differential equations.

# Outline

- 1 Numerical Differentiation
- 2 General Derivative Approximation Formulas
- 3 Three-point Derivative Approximation Formulas
- 4 Numerical Approximations to Higher Derivatives
- 5 Round-Off Error Instability
- 6 Richardson's Extrapolation

# Richardson's Extrapolation

## Generating the Extrapolation Formula

- To see specifically how we can generate the extrapolation formulas, consider the  $O(h)$  formula for approximating  $M$

$$M = N_1(h) + K_1h + K_2h^2 + K_3h^3 + \dots$$

- The formula is assumed to hold for all positive  $h$ , so we replace the parameter  $h$  by half its value.
- Then we have a second  $O(h)$  approximation formula

$$M = N_1\left(\frac{h}{2}\right) + K_1\frac{h}{2} + K_2\frac{h^2}{4} + K_3\frac{h^3}{8} + \dots$$

# Richardson's Extrapolation

$$M = N_1(h) + K_1h + K_2h^2 + K_3h^3 + \dots$$

$$M = N_1\left(\frac{h}{2}\right) + K_1\frac{h}{2} + K_2\frac{h^2}{4} + K_3\frac{h^3}{8} + \dots$$

## Generating the Extrapolation Formula (Cont'd)

Subtracting the first from twice the second eliminates the term involving  $K_1$  and gives

$$M = N_1\left(\frac{h}{2}\right) + \left[N_1\left(\frac{h}{2}\right) - N_1(h)\right] - \frac{K_2}{2}h^2 - \frac{3K_3}{4}h^3 - \dots$$

# Richardson's Extrapolation

$$M = N_1\left(\frac{h}{2}\right) + \left[N_1\left(\frac{h}{2}\right) - N_1(h)\right] - \frac{K_2}{2}h^2 - \frac{3K_3}{4}h^3 - \dots$$

## Generating the Extrapolation Formula (Cont'd)

- Define

$$N_2(h) = N_1\left(\frac{h}{2}\right) + \left[N_1\left(\frac{h}{2}\right) - N_1(h)\right]$$

- Then the above equation is an  $O(h^2)$  approximation formula for  $M$ :

$$M = N_2(h) - \frac{K_2}{2}h^2 - \frac{3K_3}{4}h^3 - \dots$$

# Richardson's Extrapolation

Example:  $f(x) = \ln x$

- In an earlier example, we used the forward-difference method

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

with  $h = 0.1$  and  $h = 0.05$  to find approximations to  $f'(1.8)$  for  $f(x) = \ln x$ .

- Assume that this formula has truncation error  $O(h)$  and use extrapolation on these values to see if this results in a better approximation.

# Richardson's Extrapolation

## Solution

Using the forward-difference method

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

we find that

$$\text{with } h = 0.1 : f'(1.8) \approx 0.5406722$$

$$\text{with } h = 0.05 : f'(1.8) \approx 0.5479795$$

This implies that

$$N_1(0.1) = 0.5406722 \quad \text{and} \quad N_1(0.05) = 0.5479795$$



# Richardson's Extrapolation

$$N_2(h) = N_1\left(\frac{h}{2}\right) + \left[N_1\left(\frac{h}{2}\right) - N_1(h)\right]$$

## Solution

- Extrapolating these results gives the new approximation

$$\begin{aligned} N_2(0.1) &= N_1(0.05) + (N_1(0.05) - N_1(0.1)) \\ &= 0.5479795 + (0.5479795 - 0.5406722) \\ &= 0.555287 \end{aligned}$$

- The  $h = 0.1$  and  $h = 0.05$  results were found to be accurate to within  $1.5 \times 10^{-2}$  and  $7.7 \times 10^{-3}$ , respectively.
- Because  $f'(1.8) = 1/1.8 = 0.\bar{5}$ , the extrapolated value is accurate to within  $2.7 \times 10^{-4}$ .

# Richardson's Extrapolation

## When can be extrapolation applied?

Extrapolation can be applied whenever the truncation error for a formula has the form

$$\sum_{j=1}^{m-1} K_j h^{\alpha_j} + O(h^{\alpha_m})$$

for a collection of constants  $K_j$  and when  $\alpha_1 < \alpha_2 < \dots < \alpha_m$ .

# Richardson's Extrapolation

## More accuracy

$O(h^2)$	$O(h^4)$	$O(h^6)$	$O(h^8)$
<b>1:</b> $N_1(h)$			
<b>2:</b> $N_1(\frac{h}{2})$	<b>3:</b> $N_2(h)$		
<b>4:</b> $N_1(\frac{h}{4})$	<b>5:</b> $N_2(\frac{h}{2})$	<b>6:</b> $N_3(h)$	
<b>7:</b> $N_1(\frac{h}{8})$	<b>8:</b> $N_2(\frac{h}{4})$	<b>9:</b> $N_3(\frac{h}{2})$	<b>10:</b> $N_4(h)$

# Richardson's Extrapolation

## Ensuring accuracy

- Each column beyond the first in the extrapolation table is obtained by a simple averaging process, so the technique can produce high-order approximations with minimal computational cost.
- However, as  $k$  increases, the round-off error in  $N_1(h/2^k)$  will generally increase because the instability of numerical differentiation is related to the step size of  $h/2^k$ .
- Also, the higher-order formulas depending increasingly on the entry to their immediate left in the table, which is the reason we recommend comparing the final diagonal entries to ensure accuracy.

# Assignment

## Assignment

- Reading assignment: Chap 4.1-4.2