

#### Numerical Analysis

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# 1, 2025

#### Course Overview, Numerical Differentiation

- Course Overview
- ► Finite differences and truncation error
- Finite differences and rounding errors
- ► Richardson's extrapolation
- Method of undefind coefficients
- Stencil, nonuniform mesh, curvilinear mesh
- Compact finite differences
- Estimating expressions with finite differences
- ► Finite differences in multidimension
- Differentiation matrix, convolution, mask, mollifier, smoothing kernel
- Application to edge detection
- ► Q & A

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  - apply numerical methods to selected practical problems

### Numerical Analysis, Course Overview related

#### **QUESTIONS?**

- ▶ which slide? please give slide number
- which theorem?
- which example?
- ▶ which ...?

- Polynomial  $P_n(x) = \sum_{i=0}^n a_i x^i, x, a_i \in \mathbb{R}, i = 0, 1, ..., n$
- Trigonometric  $T(x) = \sum_{i=0}^{n} (a_i \sin(x) + b_i \cos(x)), x, a_i \in \mathbb{R}, i = 0, 1, ..., n$
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- ► Thomas algorithm for tridiagonal matrices almost perfect why?

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## Functions related

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  - ► Can I apriori estimate accuracy with respect to *h*?

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### Example 1.3

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$$f'(x) = \frac{2\sin\left(\frac{\sqrt{x^2 + x}}{\cos x - x}\right)\cos\left(\frac{\sqrt{x^2 + x}}{\cos x - x}\right)\left[\frac{2x + 1}{2\sqrt{x^2 + 1}(\cos x - x)} + \frac{\sqrt{x^2 + 1}(\sin x + 1)}{(\cos x - x)^2}\right]}{\sin\left(\frac{\sqrt{x^2 + x}}{\cos x - x}\right)\cos\left(\frac{\sqrt{x} - 1}{\sqrt{x^2 + 1}}\right)\left[\frac{1}{2\sqrt{x^2 + x^2}} - \frac{x(\sqrt{x} - 1)}{(x^2 + 1)^{3/2}}\right]}{\sin^2\left(\frac{\sqrt{x} - 1}{\sqrt{x^2 + 1}}\right)}.$$

- f'(0.25) = -9.066698770
- $f'(x) \approx f'_{centralFD}(x) = \frac{f(x+h) f(x-h)}{2h}$
- $h = 0.005, f'_{centralFD}(0.25) = -9.067464295$
- $|f'_{centralFD}(0.25) f'(0.25)| \approx 7.7 \cdot 10^{-4}$
- ► Conclusions:

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- ► Conclusions: fast? accurate enough? low cost?

#### Definition 1.4

Taylor's formula

$$f(x_0+h)=f(x_0)+hf'(x_0)+\frac{h^2}{2}f''(x_0)+...+\frac{h^n}{n!}f^{(n)}(x_0)+O(h^{n+1})$$

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Taylor's formula

$$f(x_0 + h) = f(x_0) + hf'(x_0) + \frac{h^2}{2}f''(x_0) + \dots + \frac{h^n}{n!}f^{(n)}(x_0) + O(h^{n+1})$$

## Example 1.5

**>** 

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## Example 1.5

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

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

Conclusions:

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- Conclusions:
  - Formula is first order accurate wrt h
  - ► For estimating first order derivative *f* should be twice continuously differentiable
- Question: what if f is continuously differentiable only?

## Finite differences for second order derivatives

#### Definition 1.6

Taylor's formula

$$f(x_0 + h) = f(x_0) + hf'(x_0) + \frac{h^2}{2}f''(x_0) + \frac{h^3}{3!}f^{(3)}(x_0) + \frac{h^4}{4!}f^{(4)}(x_0) + \dots$$

$$+ \frac{h^n}{n!}f^{(n)}(x_0) + O(h^{n+1})$$

$$f(x_0 - h) = f(x_0) - hf'(x_0) + \frac{h^2}{2}f''(x_0) - \frac{h^3}{3!}f^{(3)}(x_0) + \frac{h^4}{4!}f^{(4)}(x_0) + \dots$$

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## Finite differences for second order derivatives

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### Example 1.7

**>** 

$$\frac{f(x_0+h)-2f(x_0)+f(x_0-h)}{h^2}=f''(x_0)+\frac{2h^2}{4!}f^{(4)}(x_0)+O(h^4)$$

### Finite differences for second order derivatives

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#### Example 1.7

$$\frac{f(x_0+h)-2f(x_0)+f(x_0-h)}{h^2}=f''(x_0)+\frac{2h^2}{4!}f^{(4)}(x_0)+O(h^4)$$

Question: what is accuracy of  $\frac{f(x_0)-2f(x_0+h)+f(x_0+2h)}{h^2}$  at  $x_0$ ?

#### Finite differences related

#### **QUESTIONS?**

- which slide? please give slide number
- which theorem?
- which example?
- ▶ which ...?

#### Definition 1.8

Taylor's formula

$$f(x+h) = f(x) + hf'(x) + \frac{h^2}{2}f''(x) + \frac{h^3}{3!}f^{(3)}(x_0) + \frac{h^4}{4!}f^{(4)}(x) + \frac{h^5}{5!}f^{(5)}(x) + \dots + \frac{h^n}{n!}f^{(n)}(x_0) + O(h^{n+1})$$

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► 
$$f'(x) = \frac{4f'_{centralFD}(x,h) - f'_{centralFD}(x,2h)}{3} + O(h^4)$$
  
►  $f'(x) = \frac{-f(x+2h) + f(x+h) - 8f(x-h) + f(x-2h)}{12h} + O(h^4)$ 

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Conclusions

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- Questions
  - ▶ Using Richardson's extrapolation can you obtain FD formula having accuracy  $O(h^{2k})$ , k > 2
  - ► Is Richardson's extrapolation valid for computing second order derivatives?

### Richardson's extrapolation related

#### **QUESTIONS?**

- ▶ which slide? please give slide number
- which theorem?
- which example?
- ▶ which ...?

Considered approaches can be generalized in the following algorithm

### Algorithm 1.9

Finding undetermined coefficients

$$f'(x) \approx \frac{1}{h} \sum_{i=1}^{m} c_i f(x + p_i h) \tag{1}$$

Considered approaches can be generalized in the following algorithm

### Algorithm 1.9

Finding undetermined coefficients

1. Consider formula

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 (1)

2.  $c_i, p_i, i = 1, 2, ..., n$  are undetermined; define target stencil and set  $p_i$ 

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- 5. Consider equation

$$f'(x) - \frac{1}{h} \sum_{i=1}^{m} c_i f(x + p_i h) = O(h^k)$$
 (2)

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1. Consider formula

$$f'(x) \approx \frac{1}{h} \sum_{i=1}^{m} c_i f(x + p_i h) \tag{1}$$

- 2.  $c_i, p_i, i = 1, 2, ..., n$  are undetermined; define target stencil and set  $p_i$
- 3. Expand  $f(x + p_i h)$  in Taylor's series
- 4. Substitute  $f(x + p_i h)$  by its Taylor's series in (1)
- 5. Consider equation

$$f'(x) - \frac{1}{h} \sum_{i=1}^{m} c_i f(x + p_i h) = O(h^k)$$
 (2)

6. From (2) extract system of equations wrt unknown coefficients by means of grouping them with  $h^0$ ,  $h^1$ ,  $h^2$ , ... etc.

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Considered approaches can be generalized in the following algorithm

### Algorithm 1.9

Finding undetermined coefficients

1. Consider formula

$$f'(x) \approx \frac{1}{h} \sum_{i=1}^{m} c_i f(x + p_i h) \tag{1}$$

- 2.  $c_i, p_i, i = 1, 2, ..., n$  are undetermined; define target stencil and set  $p_i$
- 3. Expand  $f(x + p_i h)$  in Taylor's series
- 4. Substitute  $f(x + p_i h)$  by its Taylor's series in (1)
- 5. Consider equation

$$f'(x) - \frac{1}{h} \sum_{i=1}^{m} c_i f(x + p_i h) = O(h^k)$$
 (2)

6. From (2) extract system of equations wrt unknown coefficients by means of grouping them with  $h^0$ ,  $h^1$ ,  $h^2$ , ... etc.

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#### Undetermined coefficients related

#### **QUESTIONS?**

- which slide? please give slide number
- which theorem?
- which example?
- ▶ which ...?

Q & A