

# Introuction To numerical methods

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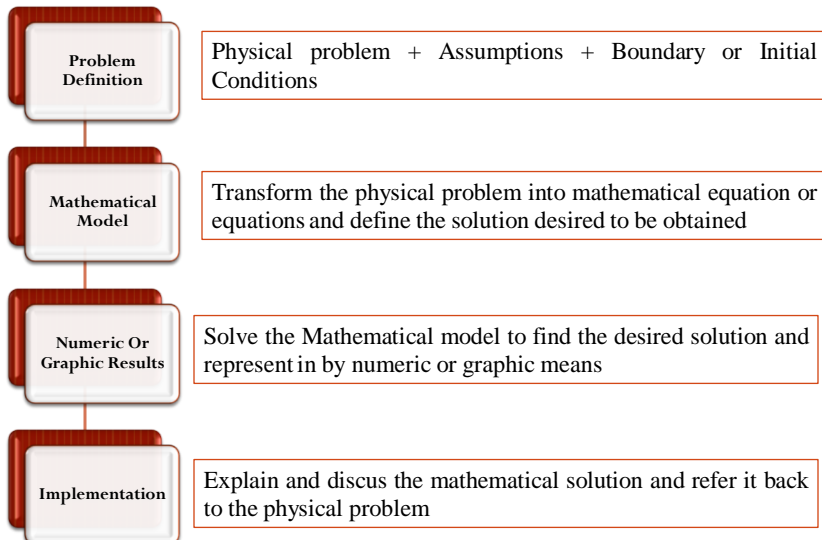
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## Introduction to numerical methods

- **Numerical methods** are those in which the mathematical problem is reformulated so it can be solved by arithmetic operations.
- **We use** numerical methods when:
  - Geometry is very **complex**
  - There is no way to find **exact solution**
- **Mathematical model** general form:

$$\text{Dependent Variable} = f \left( \text{Independent Variable}, \text{Parameters}, \text{Forcing Function} \right)$$

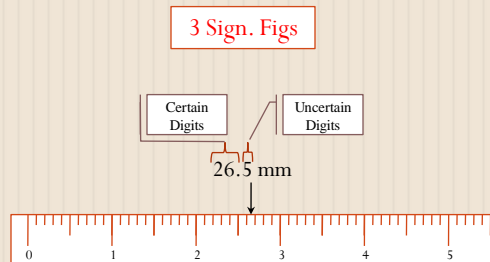
## Introduction to numerical methods



## Introduction to numerical methods

### Significant Figures

- **Significant Figures**: are the digits of a number we can use with confidence.
- **Significant Figures** = number of **certain** digits + **one digit**
- **Examples**:



## Introduction to numerical methods

### Significant Figures

#### • Special Cases:

- 2.55 & 25.5 & 255 & 2550 & 25500 have 3 Sig. Figs.

Representation :  $2.55 \times 10^1$  &  $2 \times 10^2$  &  $3 \times 10^3$  &  $4 \times 10^4$

- 2.55 & 0.255 & 0.0255 & 0.00255 have 3 Sig. Figs.

Representation :  $2.55 \times 10^{-1}$  &  $2.55 \times 10^{-2}$  &  $2.55 \times 10^{-3}$  &  $2.55 \times 10^{-4}$

## Introduction to numerical methods

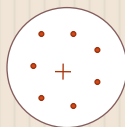
### Accuracy and Precision

- **Accuracy** refers to how closely a computed or measured value agree with the true value.
- **Precision** refers to how closely individual computed or measured value agree with each other.
- **Bias** (inaccuracy) is the opposite of accuracy. **Systematic deviation from true value**
- **Uncertainty** (imprecision) is the opposite of precision. **Magnitude of scatter.**
- **Example:**

Inaccurate & imprecise



accurate & imprecise

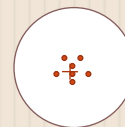


Inaccurate & precise



Numerical methods must be accurate & precise

accurate & precise



## Introduction to numerical methods

### Errors Definitions

- **Error** represents the deviation from the truth or true value
- There are two types of error may appear when numerical methods are used: **Round – off** error and **Truncation** error.
- Both errors can be represented mathematically by:

$$E_t = \text{True Value} - \text{Approximation}$$

$$E_t = \text{True error}$$

$$\varepsilon_t = \left| \frac{\text{True Value} - \text{Approximation}}{\text{True Value}} \right| \times 100\%$$

$$\varepsilon_t = \text{relative true error}$$

$$E_a = \text{Approximate error}$$

$$E_a = \text{Current Approximation} - \text{Previous Approximation}$$

$$\varepsilon_a = \text{relative Approximate error}$$

$$\varepsilon_a = \left| \frac{\text{Current Approximation} - \text{Previous Approximation}}{\text{Current Approximation}} \right| \times 100\%$$

## Introduction to numerical methods

### Round – off error

- **Round – off error:** Results from chopping off some significant figures. This type of error is found in computer processes.

Example:

$$\pi = 3.14159265\dots$$

Round – off error

- Computer limitations in representation of numeral quantity create such error.
- Computer reduce such problem by rearranging number presentation :  $25560000 \rightarrow 2.556 \times 10^7$

## Introduction to numerical methods

### Truncation error

• **Truncation error:** Results from using approximation methods instead of exact mathematical procedures to find the solution

**Example:**

**Assume**  $f(x) = x^2 + x + 1$ , then  $f(5) = 5^2 + 5 + 1 = 31$

**Assume** you used the 1<sup>st</sup> order **Taylor series** for example to find the value of  $f(5)$  and the results was:

$f(5) = f(4) + f'(4)(5-4) = 30$  **assume**  $x_i = 4$ ,  $x_{i+1} = 5$

**Error** =  $31 - 30 = 1$  (this is a truncation error)

## Truncation Error

### Taylor series

❖ **Taylor series formula:**

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)h^2}{2!} + \frac{f'''(x_i)h^3}{3!} + \dots + \frac{f^{(n)}(x_i)h^n}{n!} + R_n$$

$$R_n = \frac{f^{(n+1)}(\xi)h^{n+1}}{(n+1)!} \text{ : the remainder of Taylor series}$$

Where:

$h = x_{i+1} - x_i$  (step size)

$x_i$  : initial independent value

$x_{i+1}$  : next independent value

$\xi$  : is a value lies somewhere between  $x_{i+1}$  and  $x_i$

$n$  : Taylor series approximation order

## Truncation Error

### Taylor series

#### ❖ Example#1:

If  $f(x) = 5x^4 + 3x^3 + 6x^2 + 2x + 5$  and  $x_i = 0$   $x_{i+1} = 1$

Then the Taylor series expansion for

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)h^2}{2!} + \frac{f'''(x_i)h^3}{3!} + \frac{f^4(x_i)h^4}{4!}$$

❖ Note: when applying Taylor series to estimate polynomials, the number of terms that gives the exact solution equals the order of approximation and the rest term will be equal zero (ie.  $f^{(n+1)} = 0$ )

## Truncation Error

### Taylor series

#### ❖ Example#1(cont):

$$f(x) = 5x^4 + 3x^3 + 6x^2 + 2x + 5 \quad x_i = 0 \text{ \& } x_{i+1} = 1$$

$$f(x_i = 0) = 5 \quad f(x_{i+1} = 1) = 21$$

$$f^{(1)}(x) = 20x^3 + 9x^2 + 12x + 2 \quad f^{(1)}(0) = 2 \quad 0^{\text{th}} \text{ order: } f(1) = f(0) = 5$$

$$f^{(2)}(x) = 60x^2 + 18x + 12 \quad f^{(2)}(0) = 12 \quad 1^{\text{st}} \text{ order: } f(1) = f(0) + f^{(1)}(0)h = 7$$

$$f^{(3)}(x) = 120x + 18 \quad f^{(3)}(0) = 18 \quad 2^{\text{nd}} \text{ order: } f(1) = f(0) + f^{(1)}(0)h + \frac{f^{(2)}(0)h^2}{2!} = 13$$

$$f^{(4)}(x) = 120 \quad f^{(4)}(0) = 120 \quad 3^{\text{rd}} \text{ order: } f(1) = f(0) + f^{(1)}(0)h + \frac{f^{(2)}(0)h^2}{2!} + \frac{f^{(3)}(0)h^3}{3!} = 16$$

$$4^{\text{th}} \text{ order: } f(1) = f(0) + f^{(1)}(0)h + \frac{f^{(2)}(0)h^2}{2!} + \frac{f^{(3)}(0)h^3}{3!} + \frac{f^{(4)}(0)h^4}{4!} = 21$$

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)h^2}{2!} + \frac{f'''(x_i)h^3}{3!} + \frac{f^4(x_i)h^4}{4!}$$

## Truncation Error

### Taylor series

#### ❖ Example#1(cont):

$$f(x_{i+1} = 1) = 21$$

$$0^{\text{th}} \text{ order} : \varepsilon_t = \frac{21-5}{21} x 100\% = 76.19\%$$

$$1^{\text{st}} \text{ order} : \varepsilon_t = \frac{21-7}{21} x 100\% = 66.67\%$$

$$2^{\text{nd}} \text{ order} : \varepsilon_t = \frac{21-13}{21} x 100\% = 38.09\%$$

$$3^{\text{rd}} \text{ order} : \varepsilon_t = \frac{21-16}{21} x 100\% = 23.81\%$$

$$4^{\text{th}} \text{ order} : \varepsilon_t = \frac{21-21}{21} x 100\% = 0\%$$

Note : as the order of Taylor approximation increase, the accuracy increase too

## Truncation Error

### Taylor series

#### ❖ Example#2:

$$f(x) = \cos(x) \quad \text{and} \quad x_i = \pi/4 \quad x_{i+1} = \pi/3$$

Solution:

Order (n)	$f^{(n)}(x)$	$f(\pi/3)$	$\varepsilon_t$
0	$\cos(x)$	0.707106781	41.4
1	$-\sin(x)$	0.521986659	4.4
2	$-\cos(x)$	0.497754491	0.449
3	$\sin(x)$	0.499869147	$2.62 \times 10^{-2}$
4	$\cos(x)$	0.500007551	$1.51 \times 10^{-3}$
5	$-\sin(x)$	0.500000304	$6.08 \times 10^{-5}$
6	$-\cos(x)$	0.499999988	$2.44 \times 10^{-6}$

## Truncation Error

### Taylor series reminder

❖ Taylor series reminder:

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)h^2}{2!} + \dots + \frac{f^{(n)}(x_i)h^n}{n!} + \underbrace{\frac{f^{(n+1)}(\xi)h^{n+1}}{(n+1)!}}_{R_n}$$

❖ Assume 0<sup>th</sup> order T.S:

$$f(x_{i+1}) = f(x_i), \quad R_n = f'(x_i)h + \frac{f''(x_i)h^2}{2!} + \frac{f'''(x_i)h^3}{3!} + \dots$$

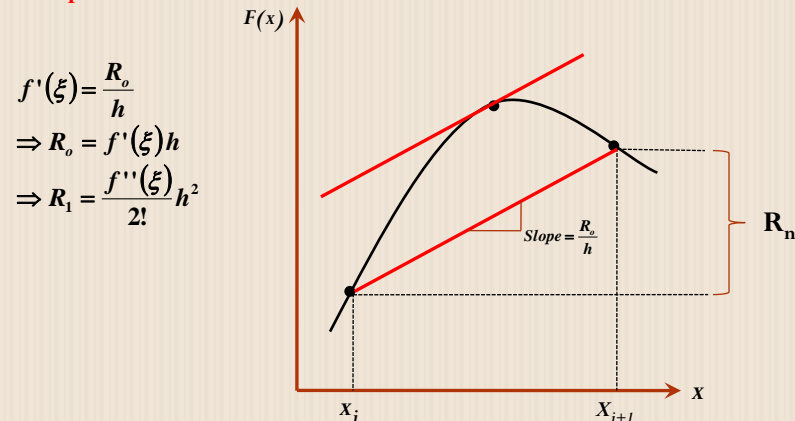
❖ It is inconvenient to deal with the remainder as infinite series. So, let us start with the 1<sup>st</sup> term of the remainder:  $R_n = f'(x_i)h$

❖ derivative mean – value theorem states that if a function  $f(x)$  and its derivative are continuous over an interval from  $x_i$  and  $x_{i+1}$ , then there exists at least one point on the function has a slope, designated by  $f'(\xi)$ , that is parallel to the line joining  $f(x_i)$  and  $f(x_{i+1})$ .

## Truncation Error

### Taylor series reminder

❖ Graphical Presentation





## Truncation Error

Using Taylor series to estimate truncation error

**Example:**

Let us take the 1<sup>st</sup> order approximation in Taylor series:

$$f(x_{i+1}) = f(x_i) + f'(x_i)(x_{i+1} - x_i) + R_1$$

Rearrange the terms :

$$f'(x_i) = \underbrace{\frac{f(x_{i+1}) - f(x_i)}{(x_{i+1} - x_i)}}_{\text{First-order Approximation}} + \underbrace{\frac{R_1}{(x_{i+1} - x_i)}}_{\text{Truncation Error}}$$

Note that the 1<sup>st</sup> part is used to approximate the 1<sup>st</sup> derivative in numerical methods

## Truncation Error

Using Taylor series to estimate truncation error

**Example<sup>(cont)</sup>:**

The reminder represents the truncation error and equal to

$$\frac{R_1}{(x_{i+1} - x_i)} = \frac{f''(\xi)}{2!} (x_{i+1} - x_i)$$

Or:

$$\frac{R_1}{(x_{i+1} - x_i)} = O(x_{i+1} - x_i)$$

$O(x_{i+1} - x_i)$ : error order notification and it means that the error is in the order of h so, if h is halve, then the error is halve too. In similar way, if the error was in order  $O(h^2)$ , its mean if the step h is halve the error will be reduced to 1/4<sup>th</sup> of its previous value.

## Numerical Differentiation

### Forward Difference

- ❖ **First derivative**

- ❖ **Formula derivation**

- ❖ Assume we take the 1<sup>st</sup> order Taylor series terms

$$f(x_{i+1}) = f(x_i) + f'(x_i)(x_{i+1} - x_i) + R_1$$

- ❖ Rearrange to separate the first derivative:

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{(x_{i+1} - x_i)} + O(x_{i+1} - x_i) = \frac{\Delta f_i}{h} + O(h)$$

- ❖  $\Delta f_i$ : 1<sup>st</sup> forward difference.

- ❖  $\Delta f_i/h$ : 1<sup>st</sup> finite forward divided difference.

## Numerical Differentiation

### Backward Difference

- ❖ **First derivative**

- ❖ **Formula derivation**

- ❖ Expand 2<sup>nd</sup> order Taylor series terms backward to get:

$$f(x_{i-1}) = f(x_i) - f'(x_i)(h) + \frac{f''(x_i)}{2!}h^2 + \dots$$

- ❖ Truncate above equation after the 1<sup>st</sup> term and rearrange to separate the first derivative:

$$f'(x_i) = \frac{f(x_i) - f(x_{i-1})}{(h)} + O(x_{i+1} - x_i) = \frac{\nabla f_i}{h} + O(h)$$

- ❖  $\nabla f_i$ : 1<sup>st</sup> Backward difference.

- ❖  $\nabla f_i/h$ : 1<sup>st</sup> finite backward divided difference.

## Numerical Differentiation

### Centered Difference

#### ❖ First derivative

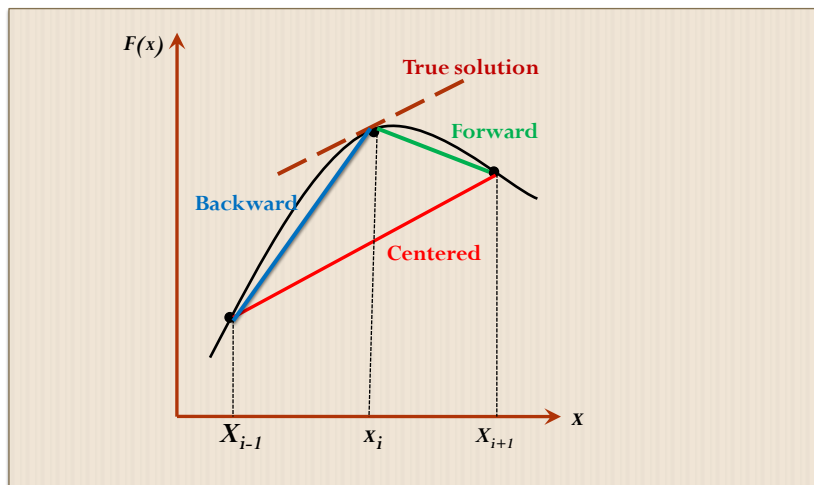
#### ❖ Formula derivation

$$f(x_{i+1}) = f(x_i) + f'(x_i)(h) + \frac{f''(x_i)}{2!}h^2 + \dots \quad f(x_{i-1}) = f(x_i) - f'(x_i)(h) + \frac{f''(x_i)}{2!}h^2 + \dots$$

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_{i-1})}{2h} - \frac{f^{(3)}(x_i)}{6}h^2 + \dots = \frac{f(x_{i+1}) - f(x_{i-1})}{2h} + O(h^2)$$

## Numerical Differentiation

### Graphical representation



## Examples

### Taylor series

#### ❖ Example #1

##### ❖ **Given information**

❖ assume you have the following functions :

$$a. f(x) = \frac{x^4 - x^3 + 1}{x^2 - 1}$$

$$b. f(x) = \sqrt{x^4 - x^3 + 1}$$

$$c. f(x) = \frac{x^4 - x^3 + 1}{\sqrt{x^3 + 1}}$$

❖ Find the value of  $f(2)$  using Taylor series 3<sup>rd</sup> order expansion. Let  $h=0.5$  and find the relative true error ( $\epsilon_i$ )

## Examples

### Taylor series

#### ❖ Example #1

##### ❖ **Solution**

❖ Taylor series 3<sup>rd</sup> order expansion (n=3) is obtained as:

$$f(x_{i+1}) = f(x_i) + f^{(1)}(x_i)h + \frac{f^{(2)}(x_i)h^2}{2!} + \frac{f^{(3)}(x_i)h^3}{3!}$$

❖ The exact solution is found as:

$$a. f(2) = \frac{2^4 - 2^3 + 1}{2^2 - 1} = 3$$

$$b. f(2) = \sqrt{2^4 - 2^3 + 1} = 3$$

$$c. f(2) = \frac{2^4 - 2^3 + 1}{\sqrt{2^3 + 1}} = 3$$

## Examples

### Taylor series

#### ❖ Example #1

##### ❖ **Solution**

$$❖ a. f_1(2) = f_1(1.5) + f_1'(1.5) * 0.5 + \frac{f_1''(1.5) * 0.5^2}{2!} + \frac{f_1'''(1.5) * 0.5^3}{3!} = 2.5008$$

$$b. f_2(2) = f_2(1.5) + f_2'(1.5) * 0.5 + \frac{f_2''(1.5) * 0.5^2}{2!} + \frac{f_2'''(1.5) * 0.5^3}{3!} = 2.9944$$

$$c. f_3(2) = f_3(1.5) + f_3'(1.5) * 0.5 + \frac{f_3''(1.5) * 0.5^2}{2!} + \frac{f_3'''(1.5) * 0.5^3}{3!} = 3.0027$$

function	$f^{(1)}(1.5)$	$f^{(2)}(1.5)$	$f^{(3)}(1.5)$	$f(2)$
a	0.2400	9.8080	0.2400	2.5008
b	2.0587	2.9046	2.0587	2.9944
c	2.2359	4.5990	2.2359	3.0027

## Examples

### Finite divided difference

#### ❖ Example #2

##### ❖ **Given information**

❖ assume you have the following functions :

$$a. f(x) = \frac{x^4 - x^3 + 1}{x^2 - 1}$$

$$b. f(x) = \sqrt{x^4 - x^3 + 1}$$

$$c. f(x) = \frac{x^4 - x^3 + 1}{\sqrt{x^3 + 1}}$$

❖ Find the value of  $f'(2)$  using forward, backward and centered finite divided difference

❖ Assume  $x_{j-1} = 1.5$   $x_j = 2$   $x_{j+1} = 2.5$

## Examples

### Finite divided difference

❖ **Example #2**

❖ **Solution**

❖ **forward** finite divided difference:

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{(x_{i+1} - x_i)}$$

❖ Assume:  $x_j = 2$   $x_{j+1} = 2.5$

$$a. f'(2) = \frac{f(2.5) - f(2)}{2.5 - 2} = 3.3095$$

$$b. f'(2) = \frac{f(2.5) - f(2)}{2.5 - 2} = 3.8869$$

$$c. f'(2) = \frac{f(2.5) - f(2)}{2.5 - 2} = 5.9869$$

## Examples

### Finite divided difference

❖ **Example #2**

❖ **Solution**

❖ **backward** finite divided difference:

$$f'(x_i) = \frac{f(x_i) - f(x_{i-1})}{(h)}$$

❖ Assume:  $x_{j-1} = 1.5$   $x_j = 2$

$$a. f'(2) = \frac{f(2) - f(1.5)}{2.5 - 2} = 1.7000$$

$$b. f'(2) = \frac{f(2) - f(1.5)}{2.5 - 2} = 2.7213$$

$$c. f'(2) = \frac{f(2) - f(1.5)}{2.5 - 2} = 3.4303$$

## Examples

### Finite divided difference

#### ❖ Example #2

#### ❖ **Solution**

#### ❖ **centered** finite divided difference:

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_{i-1}))}{(2h)}$$

#### ❖ Assume: $x_{i-1} = 1.5$ $x_{i+1} = 2.5$

$$a. f'(2) = \frac{f(2.5) - f(1.5)}{2(2.5 - 2)} = 2.5048$$

$$b. f'(2) = \frac{f(2.5) - f(1.5)}{2(2.5 - 2)} = 3.3041$$

$$c. f'(2) = \frac{f(2.5) - f(1.5)}{2(2.5 - 2)} = 4.7086$$