

# INTRODUCTION TO NUMERICAL ANALYSIS

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# 8. NUMERICAL DIFFERENTIATION

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- 8.10 Numerical Partial Differentiation

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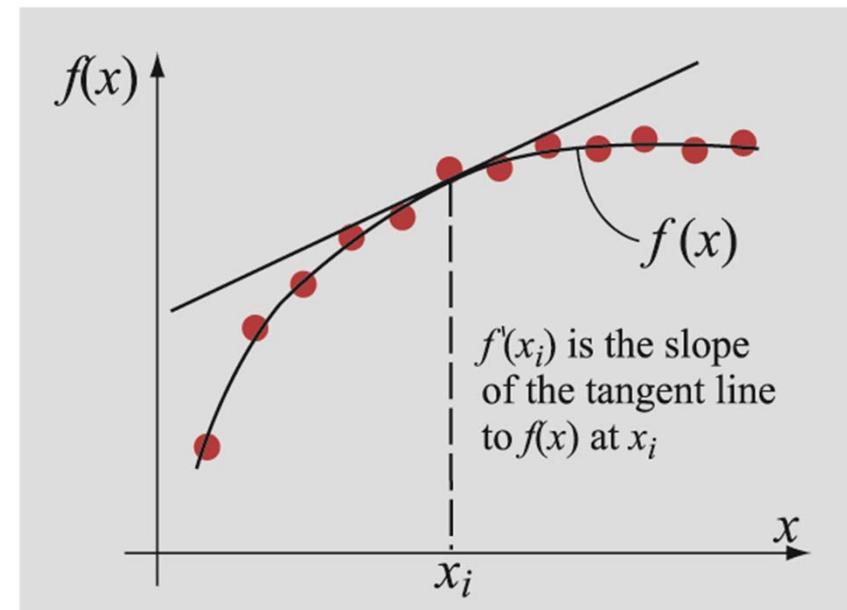
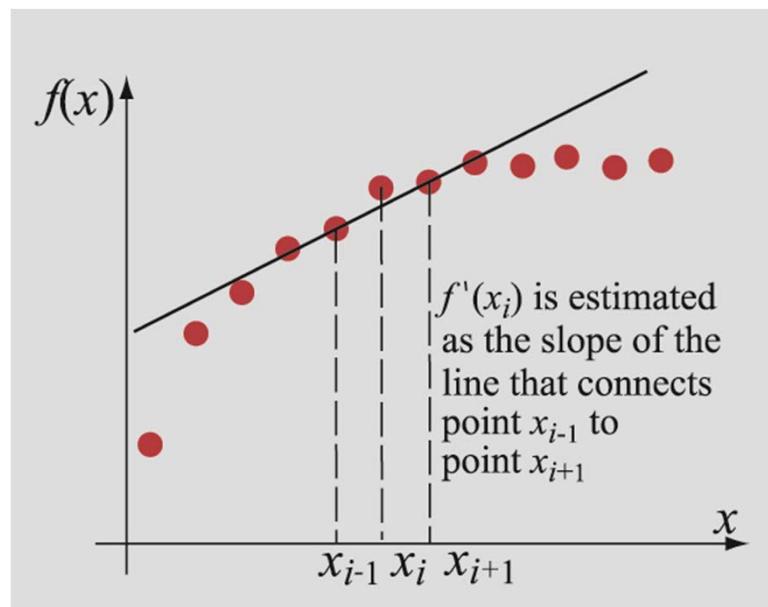
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### ❖ The need for numerical differentiation

- The function to be differentiated can be given as an analytical expression or as a set of discrete points (tabulated data).
- When the function is given as a simple mathematical expression, the derivative can be determined analytically.
- When analytical differentiation of the expression is difficult or not possible, numerical differentiation has to be used.
- When the function is specified as a set of discrete points, differentiation is done by using a numerical method.
- Numerical differentiation also plays an important role in some of the numerical methods used for solving differential equations.

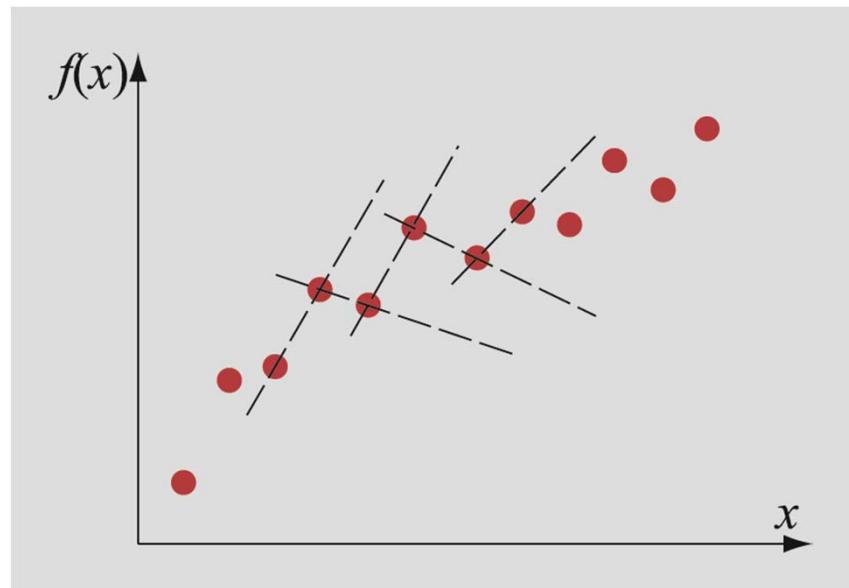
### ❖ Approaches to numerical differentiation

- - Derivative at a point  $x_i \Rightarrow$  based on the value of points in the neighborhood of  $x_i$
- Approximate analytical expression
  - Analytical expression that can be easily differentiated
  - Derived by using curve fitting
  - Derivative at a point  $x_i \Rightarrow$  analytical differentiation



### ❖ Noise and scatter in the data points

- Scatter in data
  - Experimental error, uncertainties in the measurement (electrical noise)
- Simplest form of finite difference approximation
  - Large variations !
  - Better results with ***higher-order formulas*** of finite difference approximation
    - Ex) four, five and seven-point finite difference formulas
  - Curve fitting can smooth out the data before the differentiation

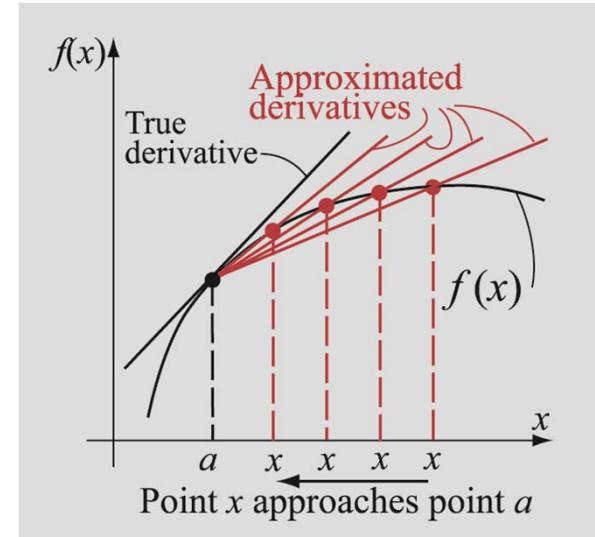


## 8.2 Finite Difference Approximation of the Derivative

- ❖ The derivative  $f'(x)$  of a function  $f(x)$  at  $x = a$

$$\frac{df(x)}{dx} \Big|_{x=a} = f'(a) = \lim_{x \rightarrow a} \frac{f(x)-f(a)}{x-a}$$

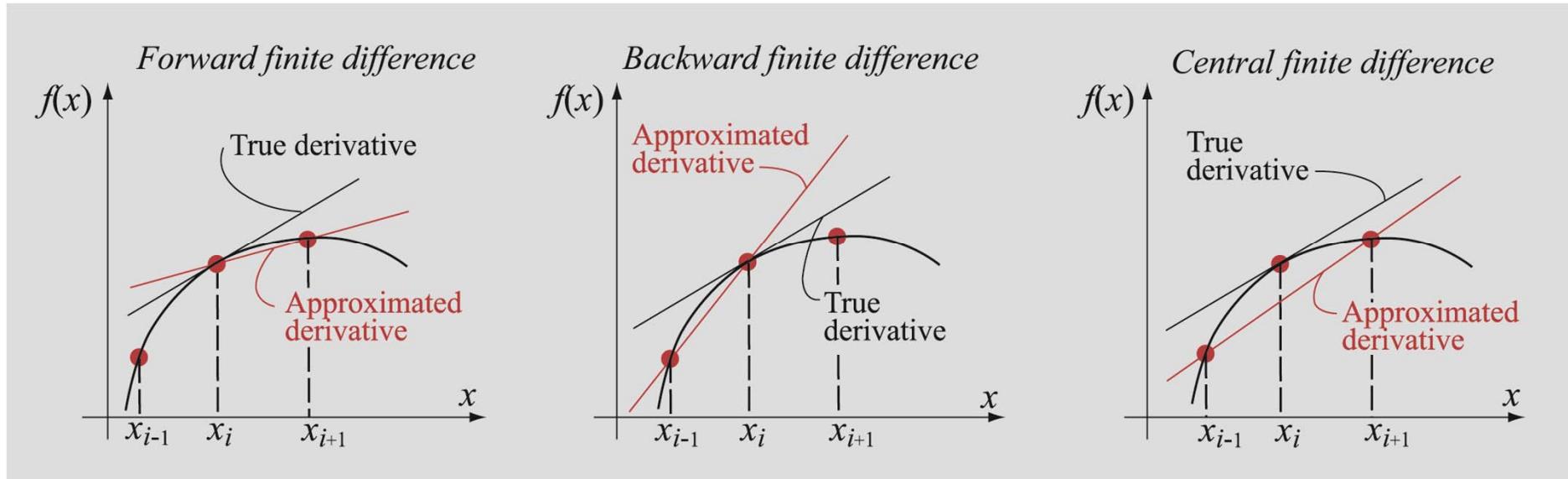
- Various finite difference formulas
  - 
  - 
  -



Estimated derivative: value of the slope of the line that connects the two points

## 8.2 Finite Difference Approximation of the Derivative

### ❖ Finite difference approximation of derivative



## 8.2 Finite Difference Approximation of the Derivative

### ❖ Example 8-1: Comparing numerical and analytical differentiation

Consider the function  $f(x) = x^3$ . Calculate its first derivative at point  $x = 3$  numerically with the forward, backward, and central finite difference formulas and using:

- (a) Points  $x = 2$ ,  $x = 3$ , and  $x = 4$ .
- (b) Points  $x = 2.75$ ,  $x = 3$ , and  $x = 3.25$ .

Compare the results with the exact (analytical) derivative.

$x:$	2	3	4
$f(x):$	8	27	64

*Forward finite difference:*

$$\frac{df}{dx} \Big|_{x=3} = \frac{f(4)-f(3)}{4-3} = \frac{64-27}{1} = 37 \quad \text{error} = \left| \frac{37-27}{27} \cdot 100 \right| = 37.04\%$$

*Backward finite difference:*

$$\frac{df}{dx} \Big|_{x=3} = \frac{f(3)-f(2)}{3-2} = \frac{27-8}{1} = 19 \quad \text{error} = \left| \frac{19-27}{27} \cdot 100 \right| = 29.63\%$$

*Central finite difference:*

$$\frac{df}{dx} \Big|_{x=3} = \frac{f(4)-f(2)}{4-2} = \frac{64-8}{2} = 28 \quad \text{error} = \left| \frac{28-27}{27} \cdot 100 \right| = 3.704\%$$

## 8.2 Finite Difference Approximation of the Derivative

### ❖ Example 8-1: Comparing numerical and analytical differentiation

Consider the function  $f(x) = x^3$ . Calculate its first derivative at point  $x = 3$  numerically with the forward, backward, and central finite difference formulas and using:

- (a) Points  $x = 2$ ,  $x = 3$ , and  $x = 4$ .
- (b) Points  $x = 2.75$ ,  $x = 3$ , and  $x = 3.25$ .

Compare the results with the exact (analytical) derivative.

$x:$	2.75	3	3.25
$f(x):$	$2.75^3$	$3^3$	$3.25^3$

*Forward finite difference:*

$$\frac{df}{dx} \Big|_{x=3} = \frac{f(3.25) - f(3)}{3.25 - 3} = \frac{3.25^3 - 27}{0.25} = 29.3125$$

$$\text{error} = \left| \frac{29.3125 - 27}{27} \right| \cdot 100 = 8.565 \quad \%$$

*Backward finite difference:*

$$\frac{df}{dx} \Big|_{x=3} = \frac{f(3) - f(2.75)}{3 - 2.75} = \frac{27 - 2.75^3}{0.25} = 24.8125$$

$$\text{error} = \left| \frac{24.8125 - 27}{27} \right| \cdot 100 = 8.102 \quad \%$$

*Central finite difference:*

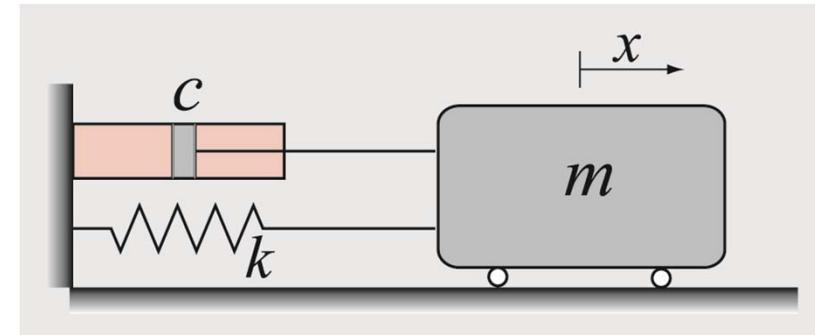
$$\frac{df}{dx} \Big|_{x=3} = \frac{f(3.25) - f(2.75)}{3.25 - 2.75} = \frac{3.25^3 - 2.75^3}{0.5} = 27.0625$$

$$\text{error} = \left| \frac{27.0625 - 27}{27} \right| \cdot 100 = 0.2315 \quad \%$$

## 8.2 Finite Difference Approximation of the Derivative

### ❖ Example 8-2: Damped vibrations

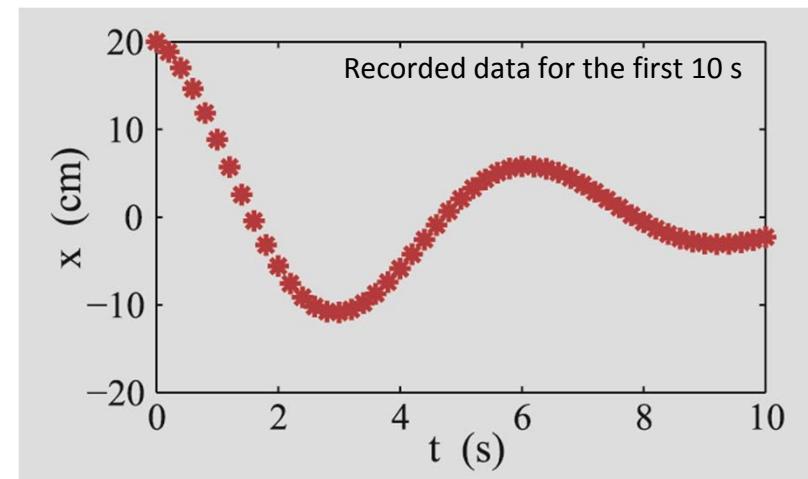
- A block of mass  $m$ 
  - Attached to a spring ( $k$ ) and a dashpot ( $c$ )
- Data record for the block position: 5 Hz
  - (a) Calculate velocity at time  $t = 5$  and  $t = 6$
  - (b) Write a user-defined MATLAB function that calculates the velocity: `dx=derivative(x,y)`  
first point: forward difference , last point: backward difference  
all the other: central difference



$t$ (s)	4.0	4.2	4.4	4.6	4.8	5.0	5.2
$x$ (cm)	-5.87	-4.23	-2.55	-0.89	0.67	2.09	3.31

$t$ (s)	5.4	5.6	5.8	6.0	6.2	6.4	6.6
$x$ (cm)	-4.31	5.06	5.55	5.78	5.77	5.52	5.08

$t$ (s)	6.8	7.0	7.2	7.4	7.6	7.8	8.0
$x$ (cm)	4.46	3.72	2.88	2.00	1.10	0.23	-0.59



## 8.2 Finite Difference Approximation of the Derivative

### ❖ Example 8-2: Damped vibrations

(a) Calculate velocity at time  $t = 5$  and  $t = 6$

The velocity is calculated by using Eq. (8.7):

$$\text{for } t = 5 \text{ s:} \quad \frac{dx}{dt}\Big|_{x=5} = \frac{f(5.2) - f(4.8)}{5.2 - 4.8} = \frac{3.31 - 0.67}{0.4} = 6.6 \text{ cm/s}$$

$$\text{for } t = 6 \text{ s:} \quad \frac{dx}{dt}\Big|_{x=5} = \frac{f(6.2) - f(5.8)}{6.2 - 5.8} = \frac{5.77 - 5.55}{0.4} = 0.55 \text{ cm/s}$$

(b) Write a user-defined MATLAB function that calculates the velocity

```
function dx = derivative(x,y)

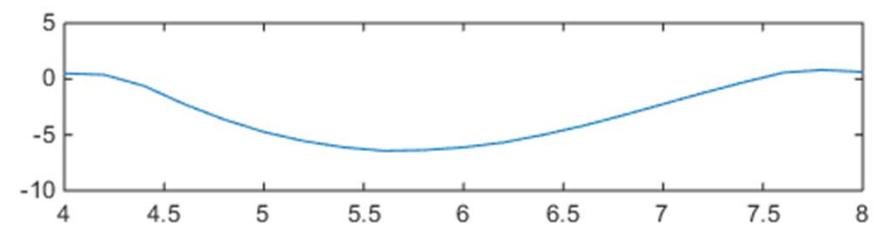
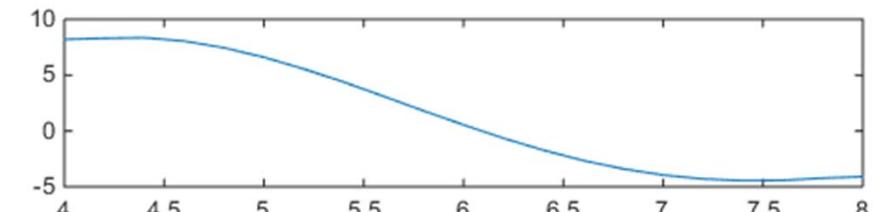
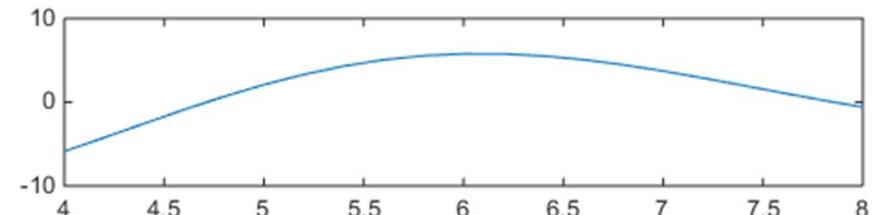
n = length(x);
dx(1)=(y(2)-y(1))/(x(2)-x(1));
for i=2:n-1
    dx(i)=(y(i+1)-y(i-1))/(x(i+1)-x(i-1));
end
dx(n)=(y(n)-y(n-1))/(x(n)-x(n-1));
```

## 8.2 Finite Difference Approximation of the Derivative

### ❖ Example 8-2: Damped vibrations

- (b) Write a user-defined MATLAB function that calculates the velocity.
- (c) Make a plot of the displacement, velocity, and acceleration, versus time for  $4 \leq t \leq 8$  s

```
clear all
t = 4:0.2:8;
x = [-5.87 -4.23 -2.55 -0.89 0.67 2.09 3.31 4.31 5.06 ...
       5.55 5.78 5.77 5.52 5.08 4.46 3.72 2.88 2.00 1.10 0.23 -0.59];
vel = derivative(t,x)
acc = derivative(t,vel);
subplot (3,1,1)
plot(t,x)
subplot (3,1,2)
plot(t,vel)
subplot (3,1,3)
plot(t,acc)
```



## 8.3 Finite Difference Formulas Using Taylor Series Expansion

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### ❖ Using Taylor series expansion for approximating derivatives

- Many finite difference formulas can be derived by using Taylor series expansion.
- It also provides an estimate for the truncation error.
- First derivative
- Second derivative

### ❖ Finite difference formulas of first derivative

- For the case where the points are equally spaced
- Two-point forward difference formula for first derivative
  - $f(x_{i+1})$  : approximated by a Taylor series,  $h = x_{i+1} - x_i$

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

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

$$x_i \leq \xi \leq x_{i+1}$$

Remainder theorem !

## 8.3 Finite Difference Formulas Using Taylor Series Expansion

### ❖ Finite difference formulas of first derivative

- Two-point forward difference formula for first derivative

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

■

- The magnitude of the truncation error is not really known.

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

## 8.3 Finite Difference Formulas Using Taylor Series Expansion

### ❖ Finite difference formulas of first derivative

- Two-point backward difference formula for first derivative

- $f(x_{i+1})$  : approximated by a Taylor series,  $h = x_i - x_{i-1}$

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

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

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

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

## 8.3 Finite Difference Formulas Using Taylor Series Expansion

### ❖ Finite difference formulas of first derivative

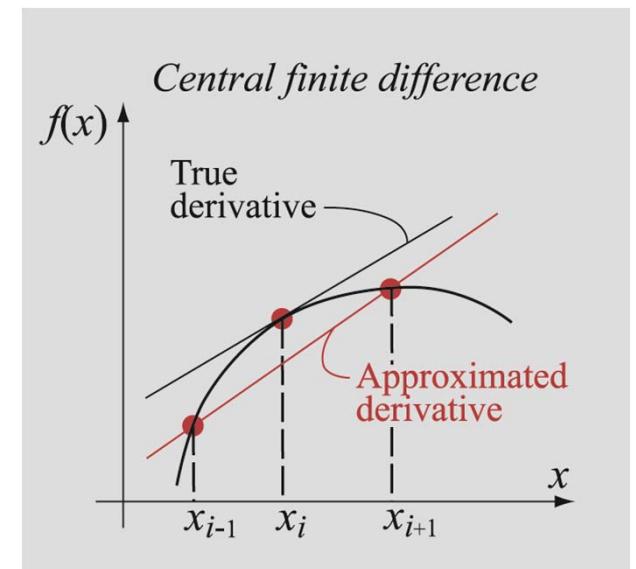
- Two-point central difference formula for first derivative (for equal spacing)

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)}{2!}h^2 + \frac{f'''(\xi_1)}{3!}h^3$$

$$f(x_{i-1}) = f(x_i) - f'(x_i)h + \frac{f''(x_i)}{2!}h^2 - \frac{f'''(\xi_2)}{3!}h^3$$

$$f(x_{i+1}) - f(x_{i-1}) = 2f'(x_i)h + \frac{f'''(\xi_1)}{3!}h^3 + \frac{f'''(\xi_2)}{3!}h^3$$

- Second order accuracy !
- More accurate than the forward and backward differencing



## 8.3 Finite Difference Formulas Using Taylor Series Expansion

### ❖ Finite difference formulas of first derivative

- Three-point forward/backward difference formula for first derivative (for equal spacing)
  - Central difference: second order accurate, but useful only for interior points
- Three-point forward difference
  - $x_i, x_{i+1}, x_{i+2}$  with uniform space ( $h = x_{i+2} - x_{i+1} = x_{i+1} - x_i$ )

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)}{2!}h^2 + \frac{f'''(\xi_1)}{3!}h^3$$

$$f(x_{i+2}) = f(x_i) + f'(x_i)2h + \frac{f''(x_i)}{2!}(2h)^2 + \frac{f'''(\xi_2)}{3!}(2h)^3$$

$$+ \frac{4f'''(\xi_1)}{3!}h^3 - \frac{f'''(\xi_2)}{3!}(2h)^3$$

$$f'(x_i) =$$

$$O(h^2)$$

Second order accuracy !

## 8.3 Finite Difference Formulas Using Taylor Series Expansion

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### ❖ Finite difference formulas of first derivative

#### ● Three-point backward difference

- $x_i, x_{i-1}, x_{i-2}$  with uniform space ( $h = x_{i+2} - x_{i+1} = x_{i+1} - x_i$ )

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

Second order accuracy !

## 8.3 Finite Difference Formulas Using Taylor Series Expansion

### ❖ Example 8-3: Comparing numerical and analytical differentiation

Consider the function  $f(x) = x^3$ . Calculate the first derivative at point  $x = 3$  numerically with the three-point forward difference formula, using:

- (a) Points  $x = 3$ ,  $x = 4$ , and  $x = 5$ .
- (b) Points  $x = 3$ ,  $x = 3.25$ , and  $x = 3.5$ .

Compare the results with the exact value of the derivative, obtained analytically.

- (a) The points used for numerical differentiation are:

$$x: \quad 3 \quad 4 \quad 5$$

$$f(x): \quad 27 \quad 64 \quad 125$$

Using Eq. (8.24), the derivative using the three-point forward difference formula is:

$$f'(3) = \frac{-3f(3) + 4f(4) - f(5)}{2 \cdot 1} = \frac{-3 \cdot 27 + 4 \cdot 64 - 125}{2} = 25 \quad \text{error} = \left| \frac{25 - 27}{27} \right| \cdot 100 = 7.41 \%$$

- (b) The points used for numerical differentiation are:

$$x: \quad 3 \quad 3.25 \quad 3.5$$

$$f(x): \quad 27 \quad 3.25^3 \quad 3.5^3$$

Using Eq. (8.24), the derivative using the three points forward finite difference formula is:

$$f'(3) = \frac{-3f(3) + 4f(3.25) - f(3.5)}{2 \cdot 0.25} = \frac{-3 \cdot 27 + 4 \cdot 3.25^3 - 3.5^3}{0.5} = 26.875$$
$$\text{error} = \left| \frac{26.875 - 27}{27} \right| \cdot 100 = 0.46 \%$$

Ex 8.1:

37.04%

8.56%

## 8.3 Finite Difference Formulas Using Taylor Series Expansion

### ❖ Finite Difference Formulas for the Second Derivative

- Central difference
- One-sided forward difference
- One-sided backward difference
- Three-point central difference formula for the second derivative

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)}{2!}h^2 + \frac{f'''(x_i)}{3!}h^3 + \frac{f^{(4)}(\xi_1)}{4!}h^4$$

$$x_i \leq \xi_1 \leq x_{i+1}$$

$$f(x_{i-1}) = f(x_i) - f'(x_i)h + \frac{f''(x_i)}{2!}h^2 - \frac{f'''(x_i)}{3!}h^3 + \frac{f^{(4)}(\xi_2)}{4!}h^4$$

$$x_{i-1} \leq \xi_2 \leq x_i$$

$$= (+) + \frac{f^{(4)}(\xi_1)}{4!}h^4 + \frac{f^{(4)}(\xi_2)}{4!}h^4$$

Second order truncation error

## 8.3 Finite Difference Formulas Using Taylor Series Expansion

### ❖ Finite Difference Formulas for the Second Derivative

- Five-point central difference formula for the second derivative

- 4<sup>th</sup> order accurate

$$f''(x_i) = \frac{-f(x_{i-2}) + 16f(x_{i-1}) - 30f(x_i) + 16f(x_{i+1}) - f(x_{i+2})}{12h^2} + O(h^4)$$

- Three-point forward difference formula for the second derivative

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)}{2!}h^2 + \frac{f'''(\xi_1)}{3!}h^3 \quad \text{x2}$$

$$f(x_{i+2}) = f(x_i) + f'(x_i)2h + \frac{f''(x_i)}{2!}(2h)^2 + \frac{f'''(\xi_2)}{3!}(2h)^3$$

$$f''(x_i) = \quad + O(h)$$

## 8.3 Finite Difference Formulas Using Taylor Series Expansion

### ❖ Finite Difference Formulas for the Second Derivative

- Three-point backward difference formula for the second derivative

$$f(x_{i-1}) = f(x_i) - f'(x_i)h + \frac{f''(x_i)}{2!}h^2 - \frac{f'''(\xi_i)}{3!}h^3 \quad \text{x2}$$

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

$$f''(x_i) = \quad \quad \quad + O(h)$$

## 8.3 Finite Difference Formulas Using Taylor Series Expansion

### ❖ Example 8-4: Comparing numerical and analytical differentiation

Consider the function  $f(x) = \frac{2^x}{x}$ . Calculate the second derivative at  $x = 2$  numerically with the three-point central difference formula using:

(a) Points  $x = 1.8$ ,  $x = 2$ , and  $x = 2.2$ .

(b) Points  $x = 1.9$ ,  $x = 2$ , and  $x = 2.1$ .

Compare the results with the exact (analytical) derivative.

$$f''(x) = \frac{2^x [\ln(2)]^2}{x} - \frac{2 \cdot 2^x \ln(2)}{x^2} + \frac{2 \cdot 2^x}{x^3}$$

$$f''(2) = 0.574617$$

(a) **0.57748177389232**

$$\text{error} = \frac{0.577482 - 0.574617}{0.574617} \cdot 100 = 0.4986 \%$$

(b) **0.57532441566441**

$$\text{error} = \frac{0.575324 - 0.574617}{0.574617} \cdot 100 = 0.1230 \%$$

## 8.4 Summary of FD Formulas for Numerical Differentiation

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First Derivative		
Method	Formula	Truncation Error
Two-point forward difference	$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{h}$	$O(h)$
Three-point forward difference	$f'(x_i) = \frac{-3f(x_i) + 4f(x_{i+1}) - f(x_{i+2})}{2h}$	$O(h^2)$
Two-point backward difference	$f'(x_i) = \frac{f(x_i) - f(x_{i-1})}{h}$	$O(h)$
Three-point backward difference	$f'(x_i) = \frac{f(x_{i-2}) - 4f(x_{i-1}) + 3f(x_i)}{2h}$	$O(h^2)$
Two-point central difference	$f'(x_i) = \frac{f(x_{i+1}) - f(x_{i-1})}{2h}$	$O(h^2)$
Four-point central difference	$f'(x_i) = \frac{f(x_{i-2}) - 8f(x_{i-1}) + 8f(x_{i+1}) - f(x_{i+2})}{12h}$	$O(h^4)$

## 8.4 Summary of FD Formulas for Numerical Differentiation

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<i>Second Derivative</i>		
<b>Method</b>	<b>Formula</b>	<b>Truncation Error</b>
Three-point forward difference	$f''(x_i) = \frac{f(x_i) - 2f(x_{i+1}) + f(x_{i+2})}{h^2}$	$O(h)$
Four-point forward difference	$f''(x_i) = \frac{2f(x_i) - 5f(x_{i+1}) + 4f(x_{i+2}) - f(x_{i+3})}{h^2}$	$O(h^2)$
Three-point backward difference	$f''(x_i) = \frac{f(x_{i-2}) - 2f(x_{i-1}) + f(x_i)}{h^2}$	$O(h)$
Four-point backward difference	$f''(x_i) = \frac{-f(x_{i-3}) + 4f(x_{i-2}) - 5f(x_{i-1}) + 2f(x_i)}{h^2}$	$O(h^2)$
Three-point central difference	$f''(x_i) = \frac{f(x_{i-1}) - 2f(x_i) + f(x_{i+1})}{h^2}$	$O(h^2)$
Five-point central difference	$f''(x_i) = \frac{-f(x_{i-2}) + 16f(x_{i-1}) - 30f(x_i) + 16f(x_{i+1}) - f(x_{i+2})}{12h^2}$	$O(h^4)$

## 8.4 Summary of FD Formulas for Numerical Differentiation

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<i>Third Derivative</i>		
<b>Method</b>	<b>Formula</b>	<b>Truncation Error</b>
Four-point forward difference	$f'''(x_i) = \frac{-f(x_i) + 3f(x_{i+1}) - 3f(x_{i+2}) + f(x_{i+3})}{h^3}$	$O(h)$
Five-point forward difference	$f'''(x_i) = \frac{-5f(x_i) + 18f(x_{i+1}) - 24f(x_{i+2}) + 14f(x_{i+3}) - 3f(x_{i+4})}{2h^3}$	$O(h^2)$
Four-point backward difference	$f'''(x_i) = \frac{-f(x_{i-3}) + 3f(x_{i-2}) - 3f(x_{i-1}) + f(x_i)}{h^3}$	$O(h)$
Five-point backward difference	$f'''(x_i) = \frac{3f(x_{i-4}) - 14f(x_{i-3}) + 24f(x_{i-2}) - 18f(x_{i-1}) + 5f(x_i)}{2h^3}$	$O(h^2)$
Four-point central difference	$f'''(x_i) = \frac{-f(x_{i-2}) + 2f(x_{i-1}) - 2f(x_{i+1}) + f(x_{i+2})}{2h^3}$	$O(h^2)$
Six-point central difference	$f'''(x_i) = \frac{f(x_{i-3}) - 8f(x_{i-2}) + 13f(x_{i-1}) - 13f(x_{i+1}) + 8f(x_{i+2}) - f(x_{i+3})}{8h^3}$	$O(h^4)$

## 8.4 Summary of FD Formulas for Numerical Differentiation

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<i>Fourth Derivative</i>		
<b>Method</b>	<b>Formula</b>	<b>Truncation Error</b>
Five-point forward difference	$f^{(iv)}(x_i) = \frac{f(x_i) - 4f(x_{i+1}) + 6f(x_{i+2}) - 4f(x_{i+3}) + f(x_{i+4})}{h^4}$	$O(h)$
Six-point forward difference	$f^{(iv)}(x_i) = \frac{3f(x_i) - 14f(x_{i+1}) + 26f(x_{i+2}) - 24f(x_{i+3}) + 11f(x_{i+4}) - 2f(x_{i+5})}{h^4}$	$O(h^2)$
Five-point backward difference	$f^{(iv)}(x_i) = \frac{f(x_{i-4}) - 4f(x_{i-3}) + 6f(x_{i-2}) - 4f(x_{i-1}) + f(x_i)}{h^4}$	$O(h)$
Six-point backward difference	$f^{(iv)}(x_i) = \frac{-2f(x_{i-5}) + 11f(x_{i-4}) - 24f(x_{i-3}) + 26f(x_{i-2}) - 14f(x_{i-1}) + 3f(x_i)}{h^4}$	$O(h^2)$
Five-point central difference	$f^{(iv)}(x_i) = \frac{f(x_{i-2}) - 4f(x_{i-1}) + 6f(x_i) - 4f(x_{i+1}) + f(x_{i+2})}{h^4}$	$O(h^2)$
Seven-point central difference	$f^{(iv)}(x_i) = \frac{f(x_{i-3}) + 12f(x_{i-2}) - 39f(x_{i-1}) + 56f(x_i) + 39f(x_{i+1}) + 12f(x_{i+2}) - f(x_{i+3})}{6h^4}$	$O(h^4)$

## 8.5 Differentiation Formulas using Lagrange Polynomials

### ❖ Lagrange polynomials

#### ● First derivative

- Two-point central, three-point forward, three-point backward difference formulas

$$f(x) = \frac{(x-x_{i+1})(x-x_{i+2})}{(x_i-x_{i+1})(x_i-x_{i+2})} y_i + \frac{(x-x_i)(x-x_{i+2})}{(x_{i+1}-x_i)(x_{i+1}-x_{i+2})} y_{i+1} + \frac{(x-x_i)(x-x_{i+1})}{(x_{i+2}-x_i)(x_{i+2}-x_{i+1})} y_{i+2}$$

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

- $x_i \rightarrow x$

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

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

$$h = x_{i+1} - x_i = x_{i+2} - x_{i+1}$$

## 8.5 Differentiation Formulas using Lagrange Polynomials

### ❖ Lagrange polynomials

#### ● First derivative

- Two-point central, three-point forward, three-point backward difference formulas

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

- $x_{i+1} \rightarrow x$

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

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

## 8.5 Differentiation Formulas using Lagrange Polynomials

### ❖ Lagrange polynomials

#### ● First derivative

- Two-point central, three-point forward, three-point backward difference formulas

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

- $x_{i+2} \rightarrow x$

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

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

## 8.5 Differentiation Formulas using Lagrange Polynomials

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### ❖ Lagrange polynomials

#### ● Important features

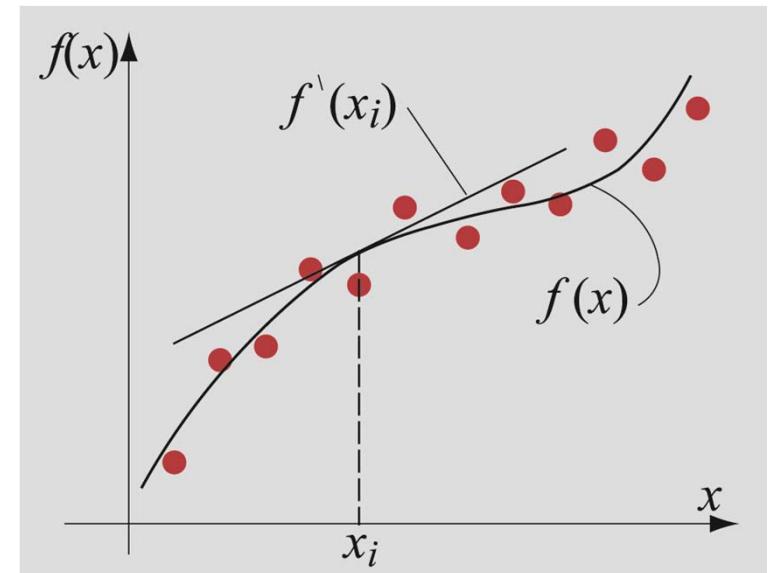
$$f'(x) = \frac{2x-x_{i+1}-x_{i+2}}{(x_i-x_{i+1})(x_i-x_{i+2})} y_i + \frac{2x-x_i-x_{i+2}}{(x_{i+1}-x_i)(x_{i+1}-x_{i+2})} y_{i+1} + \frac{2x-x_i-x_{i+1}}{(x_{i+2}-x_i)(x_{i+2}-x_{i+1})} y_{i+2}$$

- 
- It can be used for calculating the value of the
- With more points, higher order derivatives can be derived by Lagrange polynomials.
- Use of Lagrange polynomials to
  
- Taylor series provides

## 8.6 Differentiation using Curve Fitting

### ❖ Differentiation of data

- A set of discrete points
- Approximate with an analytical function
  - Easily differentiated
- Differentiate the approximate function
- Data of a non-linear relationship
  - Least square curve fitting
  - Ex) exponential function, power function, a combination of a non-linear functions



## 8.7 Use of MATLAB Built-in Functions for Numerical Differentiation

### ❖ Diff command

- Calculates the differences between adjacent elements of a vector

`d = diff(x)`

d is a vector with the differences  
between elements of x:

$$d = [(x_2 - x_1), (x_3 - x_2), \dots, (x_n - x_{n-1})]$$

x is a vector:

$$: [x_1, x_2, x_3, x_4, \dots, x_{n-1}, x_n]$$

One element shorter than x

- Derivative

`diff(y) ./ diff(x)`

$$\left[ \frac{(y_2 - y_1)}{(x_2 - x_1)}, \frac{(y_3 - y_2)}{(x_3 - x_2)}, \frac{(y_4 - y_3)}{(x_4 - x_3)}, \dots, \frac{(y_n - y_{n-1})}{(x_n - x_{n-1})} \right]$$

- Two-point forward difference formula

## 8.7 Use of MATLAB Built-in Functions for Numerical Differentiation

### ❖ Diff command

- High-order derivatives

**d = diff(x, n)**

- $n$  : number of times that `diff` is applied recursively
- $\text{diff}(x, 2) = \text{diff}((\text{diff}(x)))$
- With  $n$ -element vector  $(x_1, \dots, x_n)$ 
  - $\text{diff}(x)$ :  $n-1$  elements:  $(x_2 - x_1, \dots, x_n - x_{n-1})$
  - $\text{diff}(x, 2)$ :  $n-2$  elements:  $((x_3 - x_2) - (x_2 - x_1), \dots, (x_n - x_{n-1}) - (x_{n-1} - x_{n-2}))$   
 $((x_3 - 2x_2 + x_1), \dots, (x_n - 2x_{n-1} + x_{n-2}))$
- Three-point forward difference
  - $\text{diff}(x, 2)/h^2$

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

## 8.7 Use of MATLAB Built-in Functions for Numerical Differentiation

### ❖ `polyder` command

`dp = polyder(p)`

`dp` is a vector with the coefficients of the polynomial that is the derivative of the polynomial `p`.

`p` is a vector with the coefficients of the polynomial that is differentiated.

$$f(x) = 4x^3 + 5x + 7$$

`p = [4 0 5 7]`

`df = polyder(p)`

`df = [12 0 5]`

## 8.8 Richardson's Extrapolation

### ❖ Richardson's extrapolation

- Method for calculating a more accurate approximation of a derivative from two less accurate approximations of that derivative

$$D = D(h) + k_2 h^2 + k_4 h^4$$

$D(h)$ : approximated derivative

x4       $k_2 h^2$  and  $k_4 h^4$ : error terms

-

$$D = \frac{1}{3} \left( 4D\left(\frac{h}{2}\right) - D(h) \right) + O(h^4)$$

/3

Calculated with an error  $O(h^2)$

## 8.8 Richardson's Extrapolation

### ❖ Richardson's extrapolation

- Direct derivation from a FD formula

$$f(x_i + h) = f(x_i) + f'(x_i)h + \frac{f''(x_i)}{2!}h^2 + \frac{f'''(x_i)}{3!}h^3 + \frac{f^{iv}(x_i)}{4!}h^4 + \frac{f^v(\xi_1)}{5!}h^5$$

$$f(x_i - h) = f(x_i) - f'(x_i)h + \frac{f''(x_i)}{2!}h^2 - \frac{f'''(x_i)}{3!}h^3 + \frac{f^{iv}(x_i)}{4!}h^4 - \frac{f^v(\xi_2)}{5!}h^5 \quad (-)$$

$$f(x_i + h) - f(x_i - h) = 2f'(x_i)h + 2\frac{f'''(x_i)}{3!}h^3 + \frac{f^v(\xi_1)}{5!}h^5 + \frac{f^v(\xi_2)}{5!}h^5$$

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

- Repeat with  $h/2$

$$f'(x_i) = \frac{f(x_i + h/2) - f(x_i - h/2)}{2(h/2)} - \frac{f'''(x_i)}{3!}\left(\frac{h}{2}\right)^2 + O(h^4)$$

$$f'(x_i) = \frac{f(x_i + h/2) - f(x_i - h/2)}{h} -$$

## 8.8 Richardson's Extrapolation

### ❖ Richardson's extrapolation

- Direct derivation from a FD formula

$$f'(x_i) = \frac{f(x_i+h) - f(x_i-h)}{2h} - \frac{f'''(x_i)}{3!} h^2 + O(h^4)$$

$$f'(x_i) = \frac{f(x_i+h/2) - f(x_i-h/2)}{h} - \frac{f'''(x_i)}{4 \cdot 3!} h^2 + O(h^4) \quad (\times 4)$$

$$4f'(x_i) = 4 \left[ \frac{f(x_i+h/2) - f(x_i-h/2)}{h} - \frac{f'''(x_i)}{3!} h^2 \right] + O(h^4)$$

$$f'(x_i) = \frac{1}{3} \left[ 4 \frac{f(x_i+h/2) - f(x_i-h/2)}{h} - \frac{f(x_i+h) - f(x_i-h)}{2h} \right] + O(h^4)$$

First derivative calculated with two-point central difference formula, Eq. (8.20), with error  $O(h^2)$  for points with spacing of  $h/2$ .

First derivative calculated with two-point central difference formula, Eq. (8.20), with error  $O(h^2)$  for points with spacing of  $h$ .

## 8.8 Richardson's Extrapolation

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### ❖ Richardson's extrapolation

- Higher-order extrapolation

- Two approximated derivatives with different spacing ( $h$  and  $h/2$ )
- 4<sup>th</sup> order accuracy
- 6<sup>th</sup> order approximation

$$D = \frac{1}{15} \left( 16D\left(\frac{h}{2}\right) - D(h) \right) + O(h^6)$$

$$D = D(h) + k_1 h^4 + k_2 h_6$$

$$D = D\left(\frac{h}{2}\right) + k_1 \left(\frac{h}{2}\right)^4 + k_2 \left(\frac{h}{2}\right)^6$$

## 8.8 Richardson's Extrapolation

### ❖ Using Richardson's extrapolation in differentiation

Use Richardson's extrapolation with the results in Example 8-4 to calculate a more accurate approximation for the derivative of the function  $f(x) = \frac{2^x}{x}$  at the point  $x = 2$ .  $f''(2) = 0.574617$   
Compare the results with the exact (analytical) derivative.

for  $h = 0.2$ ,  $f''(2) = 0.577482$ . The error in this approximation is 0.5016 %.

for  $h = 0.1$ ,  $f''(2) = 0.575324$ . The error in this approximation is 0.126 %.

$$D = \frac{1}{3} \left( 4D\left(\frac{h}{2}\right) - D(h) \right) + O(h^4) = \frac{1}{3} (4 \cdot 0.575324 - 0.577481) = 0.574605$$

$$\text{error} = \frac{0.574605 - 0.5746}{0.5746} \cdot 100 = 0.00087 \quad \%$$

## 8.9 Error in Numerical Differentiation

### ❖ Truncation error (or discretization error)

- Smaller error with smaller  $h$
- For a set of discrete data points (the spacing is fixed)
  - The truncation error cannot be reduced.
  - A smaller truncation error can be obtained by using a FD formula with higher-order truncation error.
- $h$  can be made arbitrarily small
  - Round-off error can even grow as  $h$  is made smaller and smaller.

$$f(x) = e^x$$

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_{i-1})}{2h} - 2 \frac{f'''(\xi)}{3!} h^2 \quad f'(0) = \frac{e^h - e^{-h}}{2h} - 2 \frac{f'''(\xi)}{3!} h^2$$

$$f'(0) = \frac{e^h + R_1 - e^{-h} - R_2}{2h} - 2 \frac{f'''(\xi)}{3!} h^2 = \frac{e^h - e^{-h}}{2h} + \boxed{\frac{R_1 - R_2}{2h}} - 2 \frac{f'''(\xi)}{3!} h^2$$

## 8.10 Numerical Partial Differentiation

### ❖ Partial differentiation

- 2-D and 3-D problem
- Transient condition
- Rate of change of the value of the function with respect to this variable, while all the other variables are kept constant

$$\frac{\partial f(x,y)}{\partial x} \Big|_{\substack{x=a \\ y=b}} = \lim_{x \rightarrow a} \frac{f(x,b) - f(a,b)}{x-a}$$

$$\frac{\partial f(x,y)}{\partial y} \Big|_{\substack{x=a \\ y=b}} = \lim_{y \rightarrow b} \frac{f(a,y) - f(a,b)}{y-b}$$

- Two-point forward difference

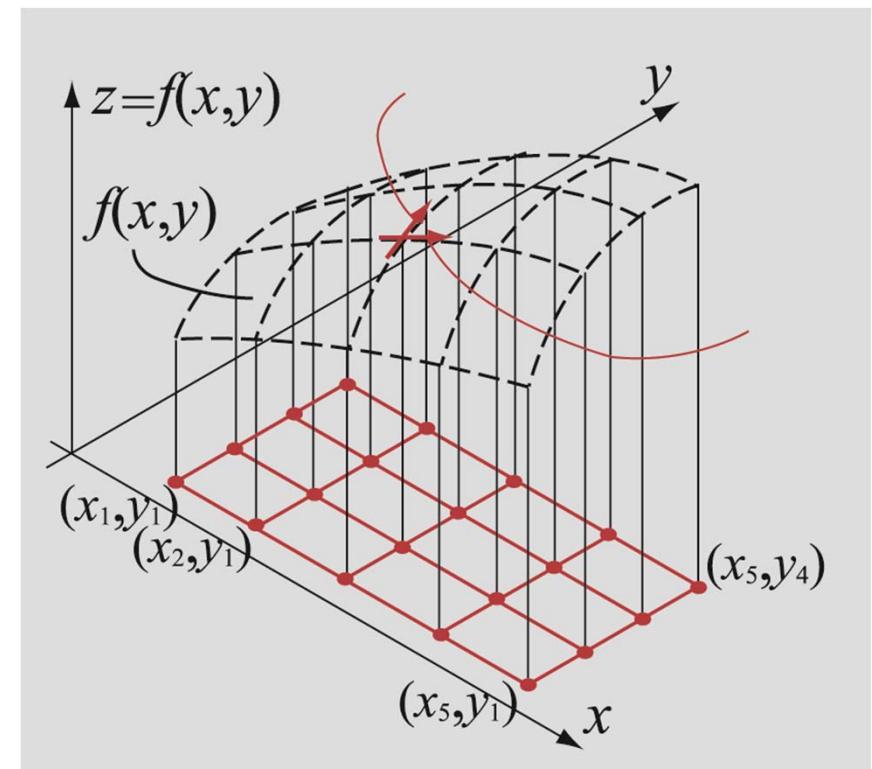
$$\frac{\partial f}{\partial x} \Big|_{\substack{x=x_i \\ y=y_i}} =$$

$$h_x = x_{i+1} - x_i$$
$$h_y = y_{i+1} - y_i$$

$$\frac{\partial f}{\partial y} \Big|_{\substack{x=x_i \\ y=y_i}} =$$

$$T(x, y, z, t)$$

$$\frac{\partial T(x, y, z, t)}{\partial z}$$



## 8.10 Numerical Partial Differentiation

### ❖ Partial differentiation

- Two-point backward and central difference

$$\frac{\partial f}{\partial x} \Big|_{\substack{x=x_i \\ y=y_i}} = \frac{f(x_i, y_i) - f(x_{i-1}, y_i)}{h_x} \quad \frac{\partial f}{\partial y} \Big|_{\substack{x=x_i \\ y=y_i}} = \frac{f(x_i, y_i) - f(x_i, y_{i-1})}{h_y}$$

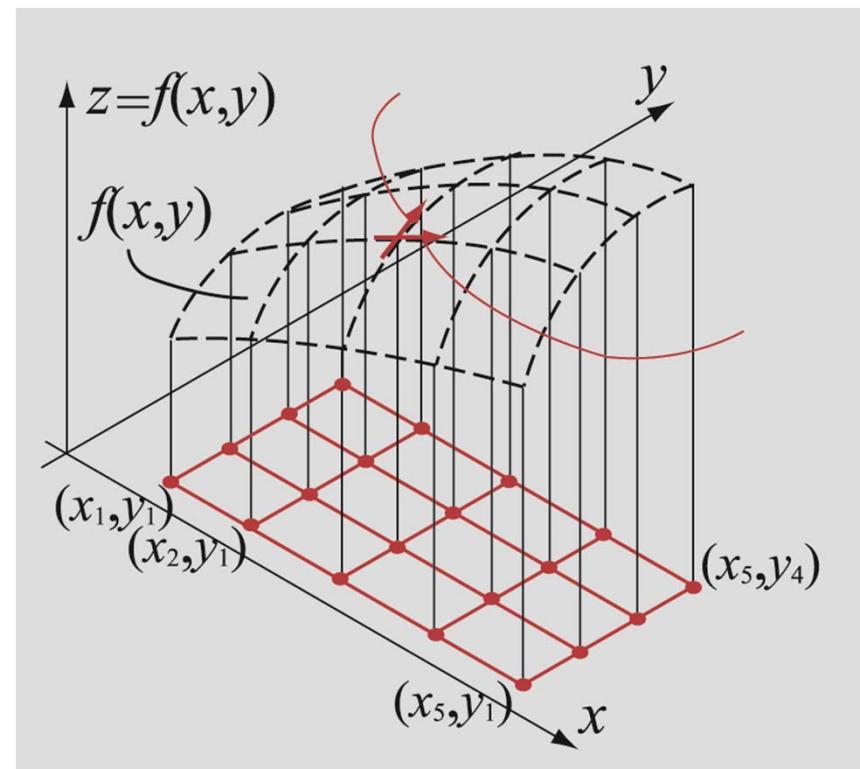
$$\frac{\partial f}{\partial x} \Big|_{\substack{x=x_i \\ y=y_i}} = \quad \frac{\partial f}{\partial y} \Big|_{\substack{x=x_i \\ y=y_i}} =$$

### ❖ Second derivatives

- Three-point central difference formula

$$\frac{\partial^2 f}{\partial x^2} \Big|_{\substack{x=x_i \\ y=y_i}} = \frac{f(x_{i-1}, y_i) - 2f(x_i, y_i) + f(x_{i+1}, y_i)}{h_x^2}$$

$$\frac{\partial^2 f}{\partial y^2} \Big|_{\substack{x=x_i \\ y=y_i}} = \frac{f(x_i, y_{i-1}) - 2f(x_i, y_i) + f(x_i, y_{i+1})}{h_y^2}$$



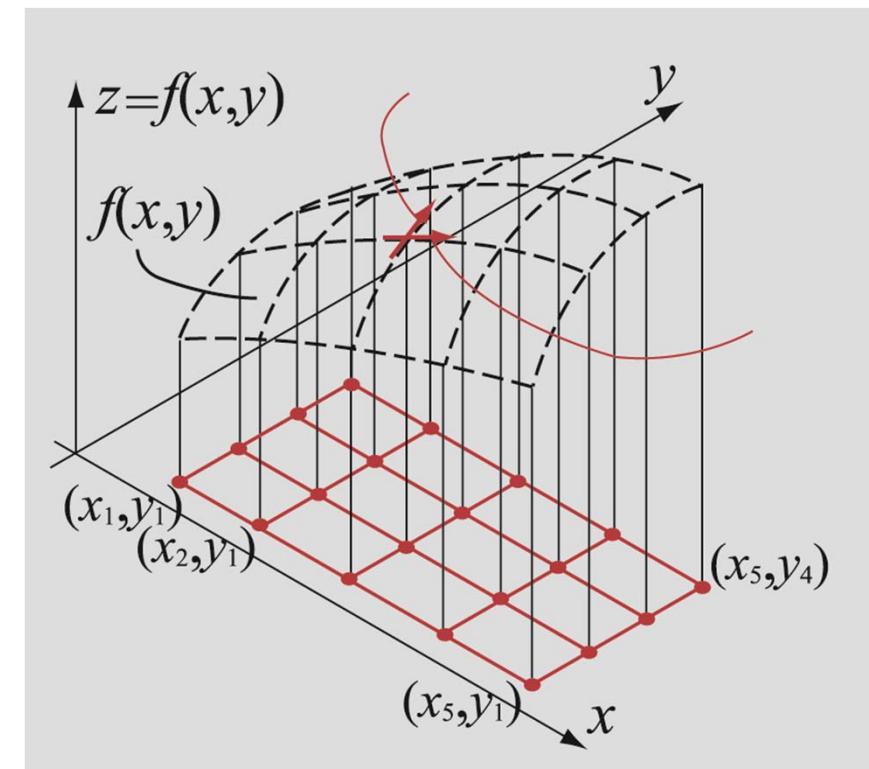
## 8.10 Numerical Partial Differentiation

### ❖ Second derivatives (mixed derivative)

$$\frac{\partial^2 f}{\partial x \partial y} = \frac{\partial}{\partial y} \left( \frac{\partial f}{\partial x} \right) = \frac{\partial}{\partial x} \left( \frac{\partial f}{\partial y} \right)$$

- Obtained from the first-order finite difference formula for the first derivative

$$\left. \frac{\partial^2 f}{\partial x \partial y} \right|_{\substack{x=x_i \\ y=y_i}} =$$



## 8.10 Numerical Partial Differentiation

### ❖ Example 8-7: Numerical partial differentiation

The following two-dimensional data for the  $x$  component of velocity  $u$  as a function of the two coordinates  $x$  and  $y$  is measured from an experiment:

	$x = 1.0$	$x = 1.5$	$x = 2.0$	$x = 2.5$	$x = 3.0$
$y = 1.0$	163	205	250	298	349
$y = 2.0$	228	291	361	437	517
$y = 3.0$	265	350	448	557	676

- (a) Using central difference approximations, calculate  $\partial u / \partial x$ ,  $\partial u / (\partial y)$ ,  $\partial^2 u / \partial y^2$ , and  $\partial^2 u / \partial x \partial y$  at the point  $(2, 2)$ .

$$\frac{\partial u}{\partial x} \Bigg|_{\substack{x=x_i \\ y=y_i}} = \frac{u(x_{i+1}, y_i) - u(x_{i-1}, y_i)}{2h_x} = \frac{u(2.5, 2) - u(1.5, 2)}{2 \cdot 0.5} = \frac{437 - 291}{1} = 146$$

$$\frac{\partial u}{\partial y} \Bigg|_{\substack{x=x_i \\ y=y_i}} = \frac{u(x_i, y_{i+1}) - u(x_i, y_{i-1})}{2h_y} = \frac{u(2, 3) - u(2, 1)}{2 \cdot 1} = \frac{448 - 250}{2} = 99$$

$$\frac{\partial^2 u}{\partial y^2} \Bigg|_{\substack{x=x_i \\ y=y_i}} = \frac{u(x_i, y_{i-1}) - 2u(x_i, y_i) + u(x_i, y_{i+1})}{h_y^2} = \frac{250 - (2 \cdot 361) + 448}{1^2} = -24$$

## 8.10 Numerical Partial Differentiation

### ❖ Example 8-7: Numerical partial differentiation

The following two-dimensional data for the  $x$  component of velocity  $u$  as a function of the two coordinates  $x$  and  $y$  is measured from an experiment:

	$x = 1.0$	$x = 1.5$	$x = 2.0$	$x = 2.5$	$x = 3.0$
$y = 1.0$	163	205	250	298	349
$y = 2.0$	228	291	361	437	517
$y = 3.0$	265	350	448	557	676

- Using central difference approximations, calculate  $\partial u / \partial x$ ,  $\partial u / (\partial y)$ ,  $\partial^2 u / \partial y^2$ , and  $\partial^2 u / \partial x \partial y$  at the point  $(2, 2)$ .
- Using a three-point forward difference approximation, calculate  $\partial u / \partial x$  at the point  $(2, 2)$ .
- Using a three-point forward difference approximation, calculate  $\partial u / \partial y$  at the point  $(2, 1)$ .

$$\left. \frac{\partial^2 u}{\partial x \partial y} \right|_{\substack{x=x_i \\ y=y_i}} = \frac{[u(x_{i+1}, y_{i+1}) - u(x_{i-1}, y_{i+1})] - [u(x_{i+1}, y_{i-1}) - u(x_{i-1}, y_{i-1})]}{2h_x \cdot 2h_y}$$

$$= \frac{[u(2.5, 3) - u(1.5, 3)] - [u(2.5, 1) - u(1.5, 1)]}{2 \cdot 0.5 \cdot 2 \cdot 1} = \frac{[557 - 350] - [298 - 205]}{2 \cdot 0.5 \cdot 2 \cdot 1} = 57$$

## 8.10 Numerical Partial Differentiation

### ❖ Example 8-7: Numerical partial differentiation

The following two-dimensional data for the  $x$  component of velocity  $u$  as a function of the two coordinates  $x$  and  $y$  is measured from an experiment:

	$x = 1.0$	$x = 1.5$	$x = 2.0$	$x = 2.5$	$x = 3.0$
$y = 1.0$	163	205	250	298	349
$y = 2.0$	228	291	361	437	517
$y = 3.0$	265	350	448	557	676

- (b) Using a three-point forward difference approximation, calculate  $\partial u / \partial x$  at the point  $(2,2)$ .  
(c) Using a three-point forward difference approximation, calculate  $\partial u / \partial y$  at the point  $(2,1)$ .

$$\begin{aligned}\left. \frac{\partial u}{\partial x} \right|_{\substack{x=x_i \\ y=y_i}} &= \frac{-3u(x_i, y_i) + 4u(x_{i+1}, y_i) - u(x_{i+2}, y_i)}{2h_x} = \\ &= \frac{-3u(2, 2) + 4u(2.5, 2) - u(3.0, 2)}{2 \cdot 0.5} = \frac{-3 \cdot 361 + 4 \cdot 437 - 517}{2 \cdot 0.5} = 148\end{aligned}$$

$$\begin{aligned}\left. \frac{\partial u}{\partial y} \right|_{\substack{x=x_i \\ y=y_i}} &= \frac{-3u(x_i, y_i) + 4u(x_i, y_{i+1}) - u(x_i, y_{i+2})}{2h_y} = \\ &= \frac{-3u(2, 1) + 4u(2, 2) - u(2, 3)}{2 \cdot 1} = \frac{-3 \cdot 250 + 4 \cdot 361 - 448}{2 \cdot 1} = 123\end{aligned}$$