

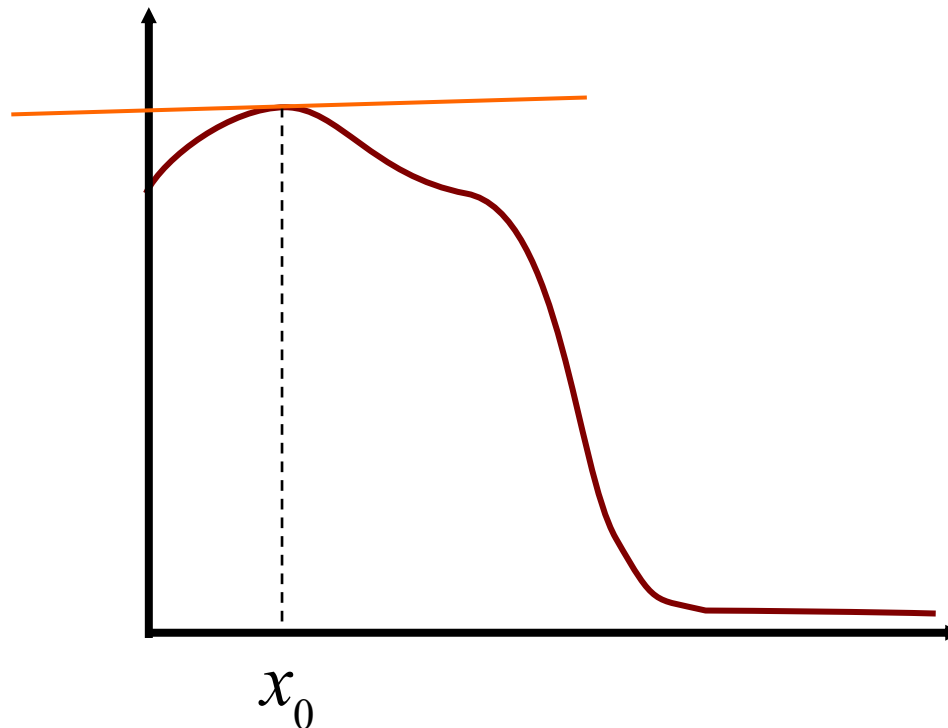
Numerical differentiation

Recall: derivatives

- Function $f: \mathbb{R} \rightarrow \mathbb{R}$, $x_0 \in \mathbb{R}$

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

- Interpretation: it is the slope of the tangent line at x_0



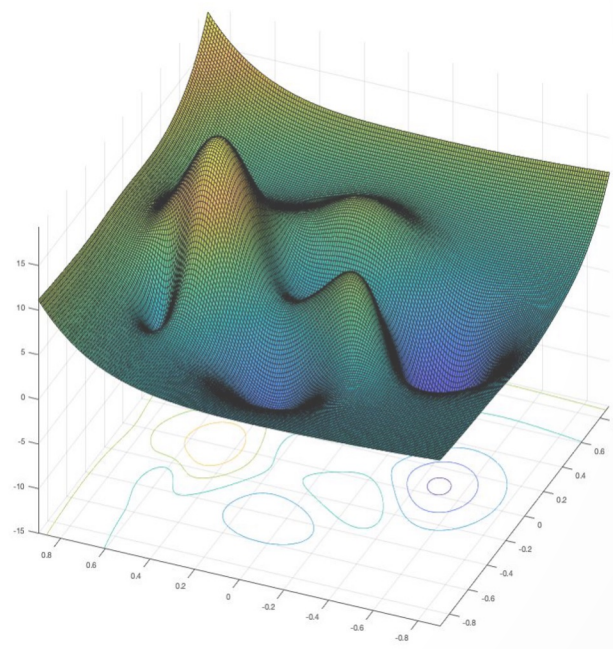
Why do we need numerical derivatives?

- Basic operation in other numerical problems:
 - Solving differential equations (both ODEs and PDEs)
 - Numerical optimization
- We need to compute

- Gradient of f : $g_k = \frac{\partial f}{\partial x_k}$

- Jacobian of $\mathbf{f}$: $J_{jk} = \frac{\partial f_j}{\partial x_k}$

- Hessian of f : $H_{jk} = \frac{\partial^2 f}{\partial x_j \partial x_k}$



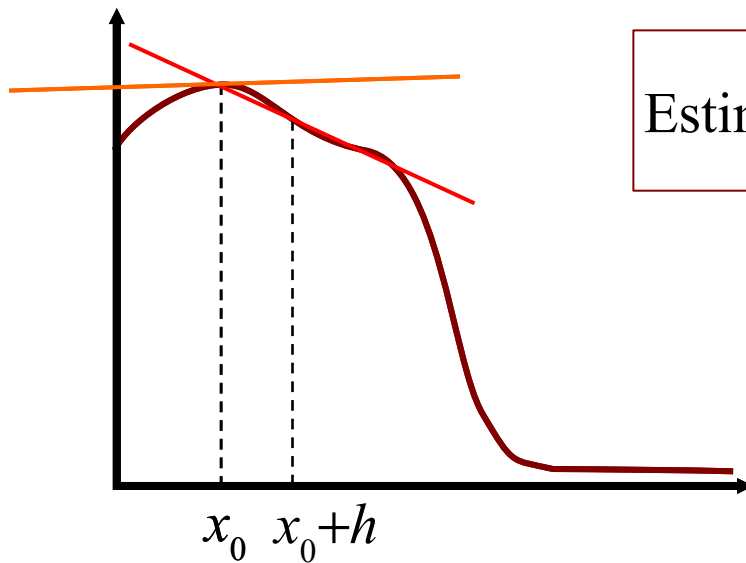
What we care about when computing derivatives numerically?

- Accuracy of our numerical estimate, and how we can improve it
- Computational complexity (how many times our function f needs to be evaluated to obtain the derivative with a desired accuracy)

Forward differences (FDF)

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

- Numerical differentiation: the limit cannot be achieved, so we have to approximate: $h > 0$ is fixed (and small)



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

- This natural approximation is called **Forward Differences (FDF)**

Interpretations of finite differences

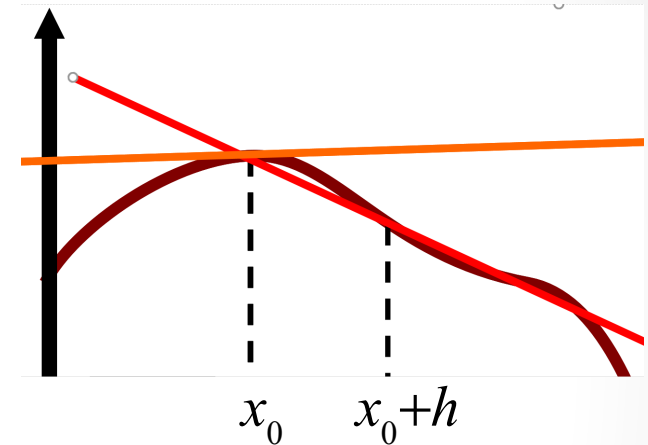
- Mean-value theorem: if a function $f: R \rightarrow R$ is continuous on $[a, b]$, and differentiable on (a, b) , then there exists $c \in (a, b)$

such that:

$$f'(c) = \frac{f(b) - f(a)}{b - a}$$

- Another interpretation: since

$$f(b) - f(a) = \int_a^b f'(x) dx$$



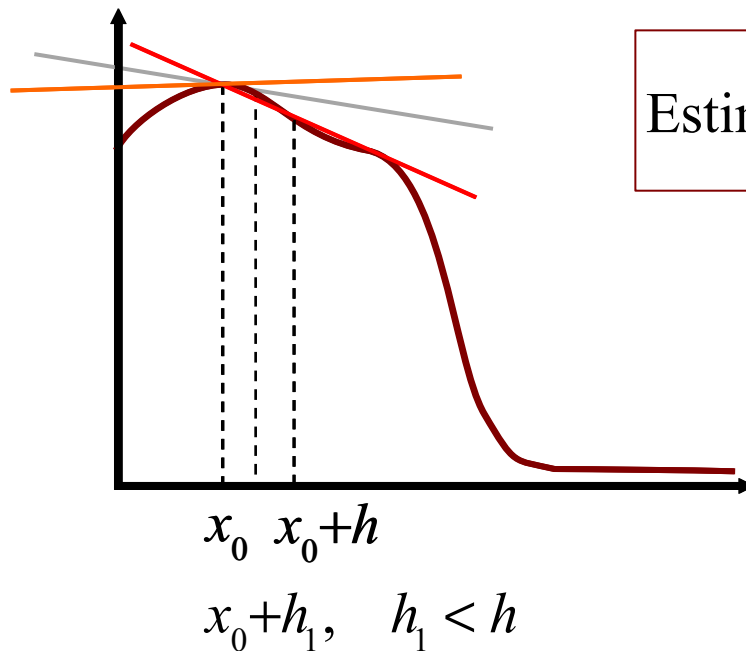
- in the case of FDF we have $b = x_0 + h$, $a = x_0$, and if we divide by h

$$\frac{f(x_0 + h) - f(x_0)}{h} = \frac{1}{h} \int_{x_0}^{x_0 + h} f'(x) dx$$

- i.e. our estimate is the average value of f' over $(x_0, x_0 + h)$

Forward differences (FDF)

- Smaller h ---> better approximation



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

Forward differences: example

- Let us test this: for $f(x)=e^{(5*x)}$ compute numerically the value of the first derivative at $x_0 = 0.2$ using different values of h :

h	df/dx	Error
1.0e-01	17.63407	4.04266328
1.0e-02	13.93693	0.34551982
1.0e-03	13.62544	0.03403522
1.0e-04	13.59481	0.00339842
1.0e-05	13.59175	0.00033979
1.0e-06	13.59144	0.00003398
1.0e-07	13.59141	0.00000340
1.0e-08	13.59141	0.00000037
1.0e-09	13.59141	0.00000026
1.0e-10	13.59141	0.00000114
1.0e-11	13.59139	0.00001447
1.0e-12	13.59091	0.00050296

How fast FDF approaches $f'(x_0)$?

Truncation
error $\longrightarrow$

$$\left| f'(x_0) - \frac{f(x_0 + h) - f(x_0)}{h} \right| = O(? h)$$

Reminder: big O notation

For $f, g : R \rightarrow R$ we can write:

$$f(x) = O(g(x)) \quad \text{as } x \rightarrow \infty$$

if and only if (iff) there exist positive C and x_0 such that

$$|f(x)| \leq C|g(x)| \quad \text{for } \forall x \geq x_0$$

Similarly, we can write

$$f(x) = O(g(x)) \quad \text{as } x \rightarrow a$$

iff there exist there exist positive C and e such that

$$|f(x)| \leq C|g(x)| \quad \text{for } |x - a| < e$$

Taylor series:

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

$$\Omega = O(h^n) \Rightarrow \exists C \in R, : |\Omega| \leq C|h|^n$$

How fast FDF approaches $f'(x_0)$?

Truncation
error $\longrightarrow$

$$\left| f'(x_0) - \frac{f(x_0 + h) - f(x_0)}{h} \right| = O(? h)$$

- We assume that $f: R \rightarrow R$ is C^2 , i.e., 2x continuously differentiable function, $x_0 \in R$, $h > 0$, and $h < C_h$
 - using Taylor expansion we get

$$f(x_0 + h) = f(x_0) + f'(x_0)h + f''(\xi)\frac{h^2}{2!}, x_0 \leq \xi \leq x_0 + h$$

How fast FDF approaches $f'(x_0)$?

Truncation error $\longrightarrow$ $\left| f'(x_0) - \frac{f(x_0 + h) - f(x_0)}{h} \right| = O(? h)$

- We assume that $f: R \rightarrow R$ is C^2 , i.e., 2x continuously differentiable function, $x_0 \in R$, $h > 0$, and $h < C_h$

- using Taylor expansion we get

$$f(x_0 + h) = f(x_0) + f'(x_0)h + f''(\xi)\frac{h^2}{2!}, x_0 \leq \xi \leq x_0 + h$$

- so

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

- and

$$\left| f'(x_0) - \frac{f(x_0 + h) - f(x_0)}{h} \right| \leq \frac{h}{2} \max_{x_0 \leq x \leq x_0 + C_h} |f''(x)|$$

- i.e.

$$\left| f'(x_0) - \frac{f(x_0 + h) - f(x_0)}{h} \right| \leq Ch$$

Forward differences: example

- Therefore:

$$\left| f'(x_0) - \frac{f(x_0 + h) - f(x_0)}{h} \right| = O(h) \leftarrow \begin{array}{l} \text{10x smaller } h \rightarrow \text{the} \\ \text{FDF error 10x smaller} \end{array}$$

- Let us test this: for $f(x)=e^{(5*x)}$ compute numerically the value of the first derivative at $x_0 = 0.2$ using different values of h :

h	df/dx	Error
1.0e-01	17.63407	4.04266328
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1.0e-07	13.59141	0.00000340
1.0e-08	13.59141	0.00000037
1.0e-09	13.59141	0.00000026
1.0e-10	13.59141	0.00000114
1.0e-11	13.59139	0.00001447
1.0e-12	13.59091	0.00050296

Round-off error

Round-off error

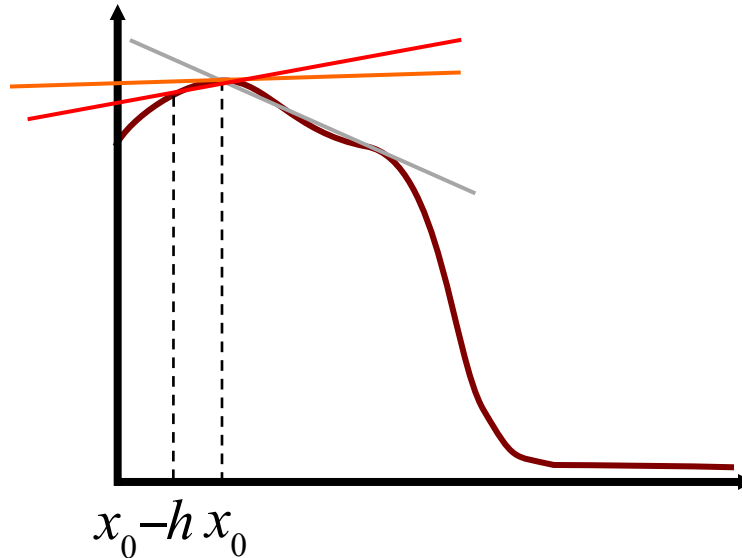
- Round-off error occurs due to a finite machine precision
- Matlab uses IEEE double precision standard -> the numbers are represented by 52 floating binary digits (or 15 floating decimal digits)
- Test this in Matlab:
 - `>> (2^52+1) - 2^52`
 - `>> (2^53+1) - 2^53`
- Using a finer grid (smaller h) reduces the truncation error
- It also increases the round-off error
- There is a trade-off between these two errors !

h	Error
1.0e-01	4.04266328
1.0e-02	0.34551982
1.0e-03	0.03403522
1.0e-04	0.00339842
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1.0e-09	0.00000026
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Backward differences (BDF)

- Use instead the point on the left-hand side

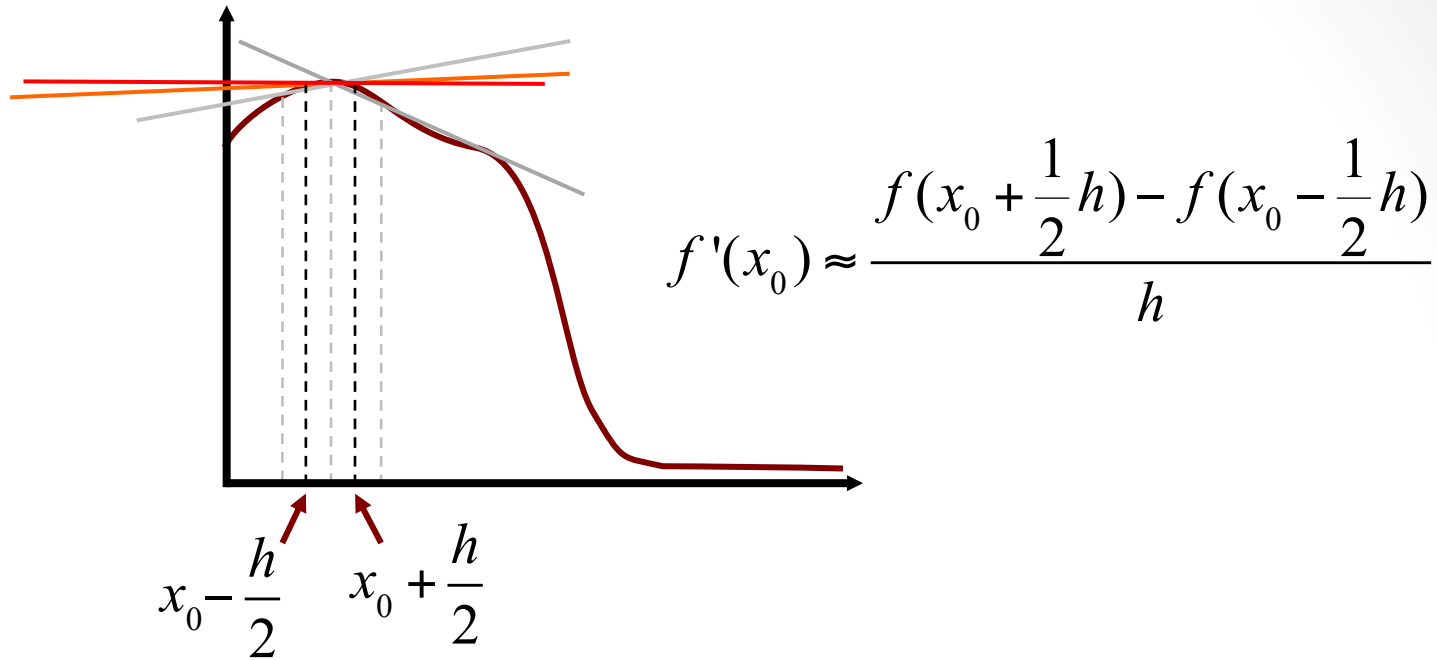
$$f'(x_0) \approx \frac{f(x_0) - f(x_0 - h)}{h}$$



- BDF has similar properties to FDF as h gets smaller, i.e.

$$\left| f'(x_0) - \frac{f(x_0) - f(x_0 - h)}{h} \right| = O(h)$$

Central differences (CDF)



- How fast CDF approaches the actual derivative as h gets smaller?

$$\left| f'(x_0) - \frac{f(x_0 + \frac{1}{2}h) - f(x_0 - \frac{1}{2}h)}{h} \right| = O(?h)$$

Central differences (CDF): truncation error

- We assume that $f: \mathbb{R} \rightarrow \mathbb{R}$ is C^3 , $x_0 \in \mathbb{R}$, $0 < h < C_h$, and we apply

Taylor's expansion:

$$f\left(x_0 + \frac{h}{2}\right) = f(x_0) + f'(x_0)\frac{h}{2} + f''(x_0)\frac{h^2}{8} + f'''(\xi)\frac{h^3}{8 \cdot 3!}, \quad x_0 \leq \xi \leq x_0 + \frac{h}{2}$$

$$f\left(x_0 - \frac{h}{2}\right) = f(x_0) - f'(x_0)\frac{h}{2} + f''(x_0)\frac{h^2}{8} - f'''(\theta)\frac{h^3}{8 \cdot 3!}, \quad x_0 - \frac{h}{2} \leq \theta \leq x_0$$

- subtract two expressions:

$$f\left(x_0 + \frac{h}{2}\right) - f\left(x_0 - \frac{h}{2}\right) = f'(x_0)h + \left(f'''(\xi)\frac{h^3}{8 \cdot 3!} + f'''(\theta)\frac{h^3}{8 \cdot 3!} \right)$$

- Subsequently, we get

$$\left| f'(x_0) - \left(f\left(x_0 + \frac{1}{2}h\right) - f\left(x_0 - \frac{1}{2}h\right) \right) / h \right| = f'''(\xi)\frac{h^2}{8 \cdot 3!} + f'''(\theta)\frac{h^2}{8 \cdot 3!}$$

$\Downarrow$

$$\left| f'(x_0) - \left(f\left(x_0 + \frac{1}{2}h\right) - f\left(x_0 - \frac{1}{2}h\right) \right) / h \right| \leq \frac{h^2}{8 \cdot 3} \max_{x_0 - C_h/2 \leq x \leq x_0 + C_h/2} |f'''(x)|$$

Central differences (CDF): example

$$\left| f'(x_0) - \frac{f(x_0 + \frac{1}{2}h) - f(x_0 - \frac{1}{2}h)}{h} \right| = O(h^2)$$

- 10x smaller h -> the error of CDF becomes 100x smaller
- Numerical test for $f(x)=e^{(5*x)}$ at $x_0 = 0.2$

h	df/dx	Error
1.0e-01	13.73343	0.14202027
1.0e-02	13.59282	0.00141582
1.0e-03	13.59142	0.00001416
1.0e-04	13.59141	0.00000014
1.0e-05	13.59141	0.00000000
1.0e-06	13.59141	0.00000000
1.0e-07	13.59141	0.00000000
1.0e-08	13.59141	0.00000003
1.0e-09	13.59141	0.00000063
1.0e-10	13.59141	0.00000330
1.0e-11	13.59139	0.00001447
1.0e-12	13.59091	0.00050296

Second derivatives

- Function $f: R \rightarrow R$ is C^2 and $x_0 \in R$

$$f''(x_0) = \lim_{h \rightarrow 0} \frac{f'(x_0 + h/2) - f'(x_0 - h/2)}{h}$$

- using CDF to estimate the 1st derivatives, i.e.,

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

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

- We finally get:

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

Partial derivatives

- Function $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ is C^1 and $x_0, y_0 \in \mathbb{R}$

$$\frac{\partial f}{\partial x}(x_0, y_0) \approx \frac{f\left(x_0 + \frac{1}{2}h, y_0\right) - f\left(x_0 - \frac{1}{2}h, y_0\right)}{h}$$

$$\frac{\partial f}{\partial y}(x_0, y_0) \approx \frac{f\left(x_0, y_0 + \frac{1}{2}h\right) - f\left(x_0, y_0 - \frac{1}{2}h\right)}{h}$$

- Exactly the same properties as CDF, i.e., no particularities

Improving estimates

- Several ways to improve the accuracy of the estimates:
 - Decrease the step size
 - Use higher-order expressions employing more points
 - Use FDF, BDF, CDF or other derivative estimates for two different h to compute a more accurate approximation
 - Richardson extrapolation

Richardson extrapolation

- The Taylor expansions (assuming that higher derivatives exist):

$$f(x_0 + \frac{h}{2}) = f(x_0) + f'(x_0)\frac{h}{2} + f''(x_0)\frac{h^2}{8} + f'''(x_0)\frac{h^3}{48} + f^{(4)}(x_0)\frac{h^4}{384} + f^{(5)}(x_0)\frac{h^5}{3840} + \dots$$

$$f(x_0 - \frac{h}{2}) = f(x_0) - f'(x_0)\frac{h}{2} + f''(x_0)\frac{h^2}{8} - f'''(x_0)\frac{h^3}{48} + f^{(4)}(x_0)\frac{h^4}{384} - f^{(5)}(x_0)\frac{h^5}{3840} + \dots$$

- if we subtract these two expressions, we get

$$f(x_0 + \frac{h}{2}) - f(x_0 - \frac{h}{2}) = f'(x_0)h + \frac{f'''(x_0)}{24}h^3 + f^{(5)}(x_0)\frac{h^5}{1920} + \dots$$

- Therefore

$$f'(x_0) = \frac{f(x_0 + \frac{h}{2}) - f(x_0 - \frac{h}{2})}{h} - \frac{f'''(x_0)}{24}h^2 - f^{(5)}(x_0)\frac{h^4}{1920} - \dots$$

$$f'(x_0) = \underbrace{\frac{f(x_0 + \frac{h}{2}) - f(x_0 - \frac{h}{2})}{h}}_{\varphi(h) \leftarrow \text{CDF}} - a_2 h^2 - a_4 h^4 - \boxed{R}$$

Remainder

Richardson extrapolation

- Idea: get rid of

$$f'(x_0) = \varphi(h) - a_2 h^2 - a_4 h^4 - R(h)$$

- For $h/2$ we can write

$$f'(x_0) = \varphi\left(\frac{h}{2}\right) - a_2 \left(\frac{h}{2}\right)^2 - a_4 \left(\frac{h}{2}\right)^4 - R\left(\frac{h}{2}\right)$$

- We obtain

$$-3f'(x_0) = \varphi(h) - 4\varphi\left(\frac{h}{2}\right) - \frac{3}{4}a_4 h^4 + 4R\left(\frac{h}{2}\right) - R(h)$$

- wherefrom we get

$$f'(x_0) = \frac{4}{3}\varphi\left(\frac{h}{2}\right) - \frac{1}{3}\varphi(h) + O(h^4)$$

- So 2x smaller h -> the error becomes 16x smaller

Richardson extrapolation:example

- Numerical test for $f(x)=e^{(5*x)}$ at $x_0 = 0.2$

h	Error (FDF)	Error (CDF)	Error (Richardson)
1.0e-01	4.04266328	0.14202027	-1.1081e-04
1.0e-02	0.34551982	0.00141582	-1.1061e-08
1.0e-03	0.03403522	0.00001416	-7.5318e-13
1.0e-04	0.00339842	0.00000014	1.1678e-11
1.0e-05	0.00033979	0.00000000	-2.1037e-10
1.0e-06	0.00003398	0.00000000	1.1811e-09
1.0e-07	0.00000340	0.00000000	-2.1171e-08
1.0e-08	0.00000037	0.00000003	1.4462e-07
1.0e-09	0.00000026	0.00000063	-5.5112e-07
1.0e-10	0.00000114	0.00000330	3.2977e-06
1.0e-11	0.00001447	0.00001447	4.4746e-05
1.0e-12	0.00050296	0.00050296	-5.0296e-04

Richardson extrapolation

- Idea: get rid of

$$f'(x_0) = \varphi(h) - a_2 h^2 - a_4 h^4 - R(h)$$

- For $h/2$ we can write

$$f'(x_0) = \varphi\left(\frac{h}{2}\right) - a_2 \left(\frac{h}{2}\right)^2 - a_4 \left(\frac{h}{2}\right)^4 - R\left(\frac{h}{2}\right)$$

- We obtain

$$-3f'(x_0) = \varphi(h) - 4\varphi\left(\frac{h}{2}\right) - \frac{3}{4}a_4 h^4 + 4R\left(\frac{h}{2}\right) - R(h)$$

- i.e.

$$f'(x_0) = \frac{4}{3}\varphi\left(\frac{h}{2}\right) - \frac{1}{3}\varphi(h) + O(h^4)$$

- So 10x smaller h -> the error becomes 10000x smaller
- Is it possible to cancel out the h^4 term?

Richardson extrapolation table

$D(0,0) = \varphi(h)$	
$D(1,0) = \varphi(h/2)$	$D(1,1)$



- Recall that

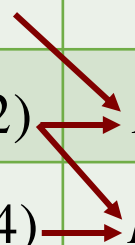
$$f'(x_0) = \frac{4}{3}\varphi\left(\frac{h}{2}\right) - \frac{1}{3}\varphi(h) + O(h^4) \Rightarrow f'(x_0) = \underbrace{\varphi\left(\frac{h}{2}\right) + \frac{1}{3}\left(\varphi\left(\frac{h}{2}\right) - \varphi(h)\right)}_{D(1,1)} + O(h^4)$$


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

$$D(1,1) = D(1,0) + \frac{1}{4^1 - 1}(D(1,0) - D(0,0))$$

Richardson extrapolation table

$D(0,0) = \varphi(h)$	
$D(1,0) = \varphi(h/2)$	$D(1,1)$
$D(2,0) = \varphi(h/4)$	$D(2,1)$



- Recall that

$$f'(x_0) = \frac{4}{3}\varphi\left(\frac{h}{2}\right) - \frac{1}{3}\varphi(h) + O(h^4) \Rightarrow f'(x_0) = \underbrace{\varphi\left(\frac{h}{2}\right) + \frac{1}{3}\left(\varphi\left(\frac{h}{2}\right) - \varphi(h)\right)}_{D(1,1)} + O(h^4)$$

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


$$D(1,1) = D(1,0) + \frac{1}{4^1 - 1}(D(1,0) - D(0,0))$$

Richardson extrapolation table

$D(0,0) = \varphi(h)$			
$D(1,0) = \varphi(h/2)$	$D(1,1)$		
$D(2,0) = \varphi(h/4)$	$D(2,1)$	$D(2,2)$	

- Recall that

$$f'(x_0) = \frac{4}{3}\varphi\left(\frac{h}{2}\right) - \frac{1}{3}\varphi(h) + O(h^4) \Rightarrow f'(x_0) = \underbrace{\varphi\left(\frac{h}{2}\right) + \frac{1}{3}\left(\varphi\left(\frac{h}{2}\right) - \varphi(h)\right)}_{D(1,1)} + O(h^4)$$

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

$$D(1,1) = D(1,0) + \frac{1}{4^1 - 1}(D(1,0) - D(0,0))$$

Richardson extrapolation table

m ↓

$D(0,0) = \varphi(h)$			
$D(1,0) = \varphi(h/2)$	$D(1,1)$		
$D(2,0) = \varphi(h/4)$	$D(2,1)$	$D(2,2)$	
$D(3,0) = \varphi(h/8)$	$D(3,1)$	$D(3,2)$	$D(3,3)$

↓ n →

$$D(n, m) = D(n, m-1) + \frac{1}{4^m - 1} [D(n, m-1) - D(n-1, m-1)]$$