CEGM1000 Modelling, Uncertainty and Data for Engineers

Week 1.2 Numerical modelling (Fundamentals)

Ronald Brinkgreve, Anna Störiko

Based on a previous version from Jaime Arriaga and the rest of the MUDE team







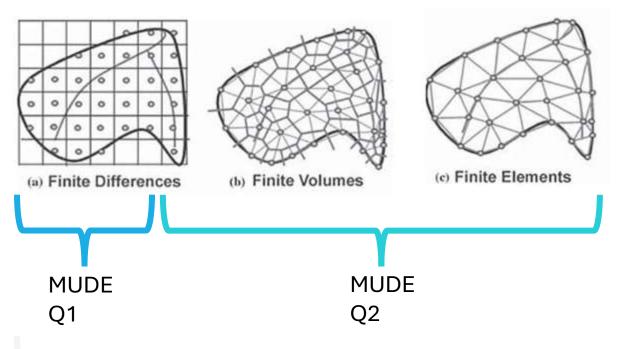
Learning objectives

At the end of this lecture, you should be able to

- Discuss the relevance of numerical modelling keeping in mind its pitfalls
- Use Taylor series to find approximations of derivatives and its accuracy
- Apply numerical integration methods to functions using pen and paper.



Methods



Other methods:

- Based on characteristics
- Boundary elements
- Physics-informed neural networks



Q3, Q4 and beyond (depending on your track)

(e.g. CIEM2110 Numerical Modelling in Geotechnical Engineering)

What relationship do you **expect** to get with numerical models?

Developer / creator	
	0%
User (occasionally)	
	0%
Super user (daily)	
	0%
Decision maker	
	0%

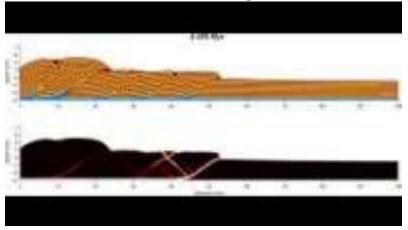
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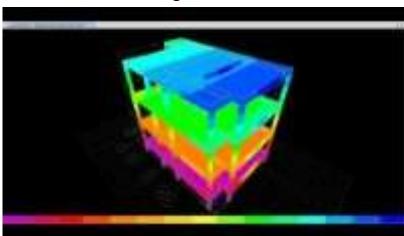
RESULTS SLIDE

Some applications

Mountain building



Building Deformation



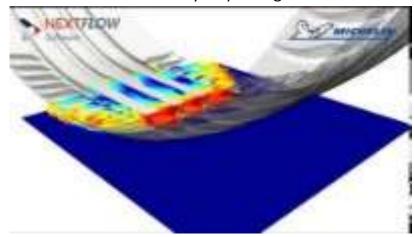
Dam Break Simulation

Garrett Apuzen-Ito. (2016, Nov 16). *Numerical model of mountain building* [Video]. Youtube. https://www.youtube.com/watch?v=HUn8lzdDmfk

ONKAR CHAUHAN. (2018, Jan 7). 3D Animation of deformation of Building after analysis in ETABS [Video]. Youtube. https://www.youtube.com/watch?v=RJZRtdINSms

XC ENGINEERING. (2016, Mar 17). Damn break simulation with FLOW-3D [Video]. Youtube. https://www.youtube.com/watch?v=3q8EY4zBf3w

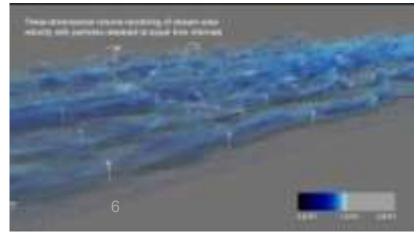
Tire - hydroplaning



Nextflow Software. (2019, Mar 14). *Tyre Hydroplanning simulation* [Video].

Youtubehttps://www.youtube.com/watch?v=0sVCOn_hoGU

Large eddy simulation of a Wind Farm



Physics of Fluids Group University of Twente. (2016, October 27). *Large eddy simulation of a Wind Farm- Explanatory clip* [Video]. Youtube. https://www.youtube.com/watch?v=qEtcCiln-0Q

Which simulation do you **feel** is more reliable?

)%
)%
)%
)%
)%
)

Which simulation do you **feel** is more reliable?



A short discussion about numerical models

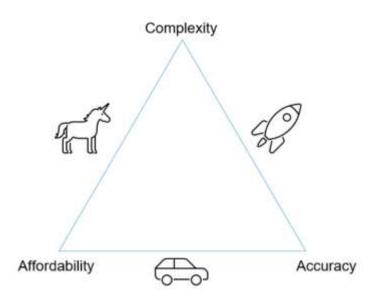
Some Pros

- 1. Design checks
- 2. Future scenarios
- 3. Non-existent situations
- 4. Some experiments cannot be scaled

Some Cons

- 1. May give credibility to incorrect results
- 2. Good modellers are scarce
- 3. Can take a lot of effort

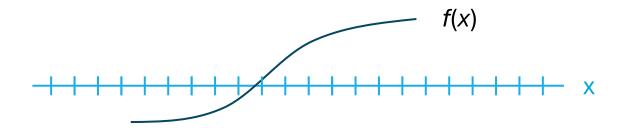
"All models are wrong, but some are useful" (after George Box)



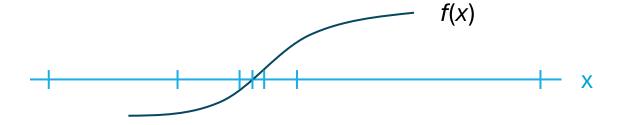
"When all you have is a hammer, everything looks like a nail" (after Abraham Kaplan)

Root finding: Numerical solutions for f(x) = 0 (which x?)

 Stepping method (fixed interval)



2. Bisection method



f(x)

3. Newton-Raphson

(taking advantage of derivatives)

Differential equations – ODEs, PDEs

Ice growth
First order ODE →

Beam deformation
Second order ODE →

1D consolidation
Second order PDE →

Navier Stokes 3D
System on non-linear PDEs →

$$ho_{ice}rac{dh_{ice}}{dt}=-k_{ice}rac{T_{water}-T_{air}}{h_{ice}}$$

We often approximate the derivatives numerically!

$$\frac{dh_{ice}}{dt} \approx ?$$

11

Contents

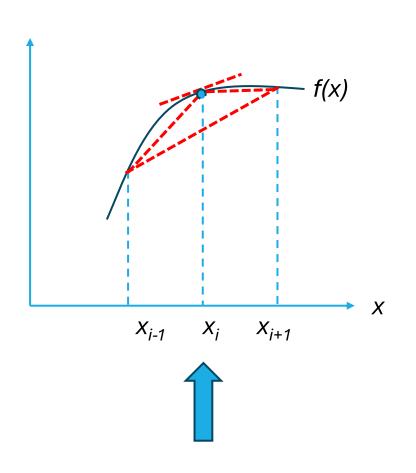
- Numerical Derivatives
- Taylor Series Expansion
- Numerical Integration

Numerical Derivatives

Definition
$$\Rightarrow$$
 $\left. \frac{df}{dx} \right|_{x_0} = f'(x_0) = \lim_{x \to x_0} \frac{f(x) - f(x_0)}{x - x_0}$

Numerically
$$igoplus f'(x_0) pprox rac{f(x) - f(x_0)}{\Delta x}, ext{where } \Delta x = x - x_0$$

More than one way to approximate derivative



$$\left. ext{forward} \, rac{df}{dx}
ight|_{x_i} pprox rac{f(x_{i+1}) - f(x_i)}{x_{i+1} - x_i}
ight.$$

$$\left. ext{backward } \frac{df}{dx} \right|_{x_i} pprox rac{f(x_i) - f(x_{i-1})}{x_i - x_{i-1}}$$

$$\left. ext{central} \left. rac{df}{dx}
ight|_{x_i} pprox rac{f(x_{i+1}) - f(x_{i-1})}{x_{i+1} - x_{i-1}}
ight.$$

Taylor Series Expansions

• The exact value of any function f(x) around $x = x_0$ can be calculated using an infinite number of terms.

$$f(x) = f(x_0) + f'(x_0) \frac{(x - x_0)}{1!} + f''(x_0) \frac{(x - x_0)^2}{2!} + f'''(x_0) \frac{(x - x_0)^3}{3!} + \dots + f^{(n)}(x_0) \frac{(x - x_0)^n}{n!}$$

The symbol! means factorial (in Dutch: faculteit): $3! = 1 \times 2 \times 3 = 6$

Taylor Series Expansions

• Compute the Taylor Series Expansion of $f(x) = \sin(x)$ around $x_0 = 0$ with third order accuracy

$$f(x) \approx f(x_0) + f'(x_0) \frac{(x - x_0)}{1!} + f''(x_0) \frac{(x - x_0)^2}{2!} + f'''(x_0) \frac{(x - x_0)^3}{3!}$$

$$f(0) = \sin(0) \qquad \qquad f(0) = 0$$

$$f'(0) = \cos(0) \qquad \qquad f'(0) = 1$$

$$f''(0) = -\sin(0) \qquad \qquad f''(0) = 0$$

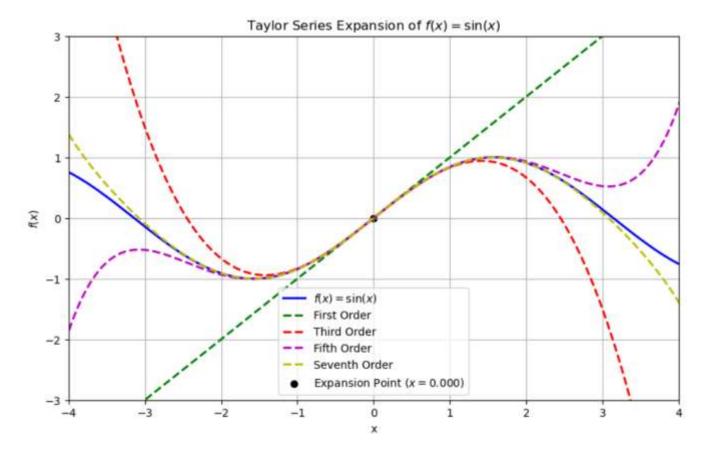
$$f'''(0) = -\cos(0) \qquad \qquad f'''(0) = -1$$

$$\rightarrow f(x) \approx 0 + (1) \frac{(x - 0)}{1} + 0 + (-1) \frac{(x - 0)^3}{6} = x - \frac{x^3}{6}$$

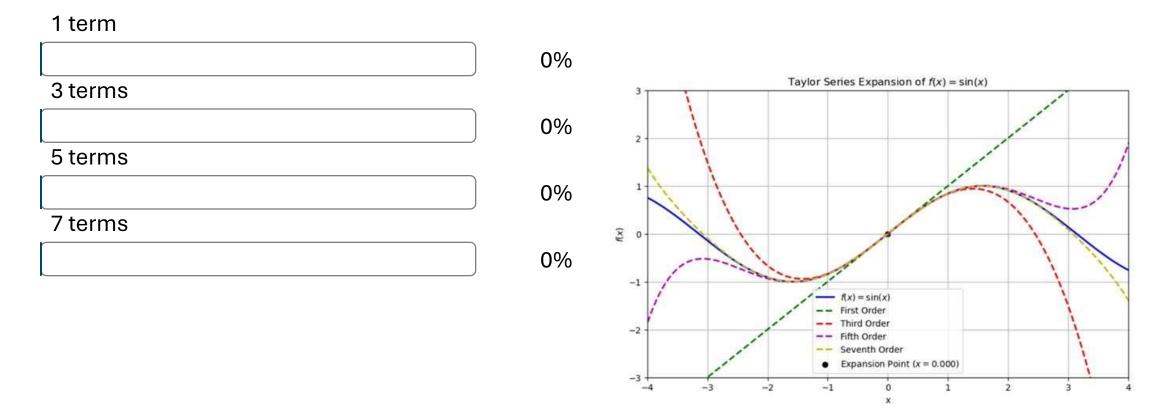
Taylor Series Expansions

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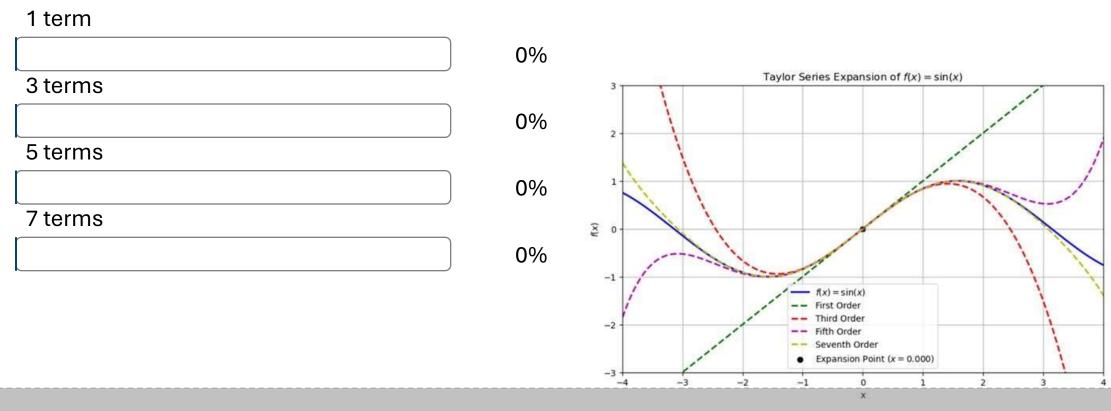
$$\sin(x) \approx x - \frac{x^3}{6}$$



Which approximation gives the best result at x=10?

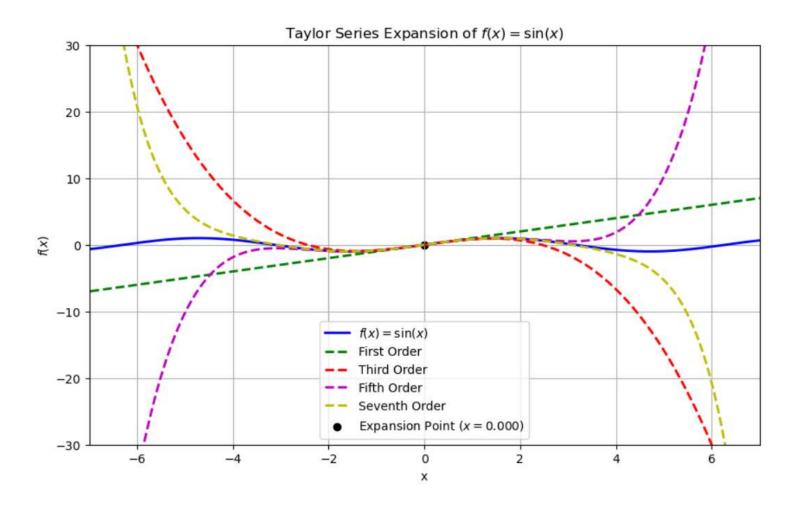


Which approximation gives the best result at x=10?



RESULTS SLIDE

- The more terms used; the solution will be more accurate near x_0
- The farther from x_0 ; the error increases



Taylor Series Expansions to get the derivatives

- Basic derivatives
- Higher order Finite Difference
- Second derivative Finite Difference

Taylor expansion – 1 or 2 variables

• Taylor series for a function of one variable y=f(x):

$$f(x) \approx f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2!}(x - x_0)^2 + \dots + \frac{f^n(x_0)}{n!}(x - x_0)^n$$

Second degree Taylor Polynomial of a function of two variables, f(x,y)

$$f(x,y) \approx f(x_0, y_0) + f'_x(x_0, y_0)(x - x_0) + f'_y(x_0, y_0)(y - y_0) + \frac{f''_x(x_0, y_0)}{2}(x - x_0)^2 + \frac{f''_y(x_0, y_0)}{2}(y - y_0)^2 + f''_{xy}(x_0, y_0)(x - x_0)(y - y_0)$$

Taylor expansion – Derivation of expression for first derivative

$$f(x) \approx f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2!}(x - x_0)^2 + \dots + \frac{f^n(x_0)}{n!}(x - x_0)^n$$
 (TSE)

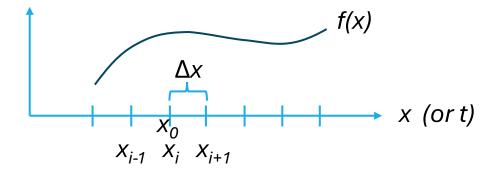
Based on TSE, consider a regular interval $(x-x_0) = \Delta x$ (hence, $x = x_0 + \Delta x$): Derive an expression for the $O(\Delta x)$ -accurate first derivative of function f(x) around x_0

$$f(x_0 + \Delta x) = \dots?$$

$$f(x_0 + \Delta x) = f(x_0) + f'(x_0) \Delta x \pm O(\Delta x^2)$$

$$f'(x_0) = \frac{f(x_0 + \Delta x) - f(x_0)}{\Delta x} \pm O(\Delta x)$$

$$f'(x_0) = \frac{f(x_0) - f(x_0 - \Delta x)}{\Delta x} \pm O(\Delta x)$$



Forward Difference method

Backward Difference method

Taylor expansion – Derivation of expression for higher-order derivative

$$f(x) \approx f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2!}(x - x_0)^2 + \dots + \frac{f^n(x_0)}{n!}(x - x_0)^n$$
 (TSE)

Derive an expression for the $O(\Delta x^2)$ -accurate first derivative of function f(x) around x_0

$$f(x_0 + \Delta x) = f(x_0) + f'(x_0)\Delta x + \frac{f''(x_0)}{2}(\Delta x)^2 \pm O(\Delta x^3)$$
$$f(x_0 - \Delta x) = f(x_0) - f'(x_0)\Delta x + \frac{f''(x_0)}{2}(\Delta x)^2 \pm O(\Delta x^3)$$

$$f(x_0 + \Delta x) - f(x_0 - \Delta x) = 2 f'(x_0) \Delta x \pm O(\Delta x^3)$$

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

Central Difference method

Taylor expansion - Derivation of

$$f(x) \approx f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2!}(x - x_0)$$

Derive an expression for the second deriv

$$f(x_0 + \Delta x) = f(x_0) + f'(x_0)\Delta x + \frac{f''}{f'}$$

$$f(x_0) = f(x_0)$$

$$f(x_0 - \Delta x) = f(x_0) - f'(x_0)\Delta x + \frac{f''}{f'}$$

$$f(x_0 + \Delta x) - 2f(x_0) + f(x_0 - \Delta x) =$$

First define a TSE for a point two steps away from x_i :

$$f(x_i+2\Delta x)=f(x_i)+2\Delta x f'(x_i)+rac{(2\Delta x)^2}{2!}f''(x_i)+\mathcal{O}(\Delta x^3).$$

Now multiply the TSE by two for a point one step away from x_i :

$$2f(x_i+\Delta x)=2f(x_i)+2\Delta x f'(x_i)+rac{2\Delta x^2}{2!}f''(x_i)+\mathcal{O}(\Delta x^3).$$

By subtracting the first expression from the second one the first derivative disappears:

$$f(x_i+2\Delta x)-2f(x_i+\Delta x)=-f(x_i)+rac{2\Delta x^2}{2!}f''(x_i)+\mathcal{O}(\Delta x^3).$$

By solving for f'' we obtain the **forward** expression:

$$f''(x_i) = rac{f(x_i + 2\Delta x) - 2f(x_i + \Delta x) + f(x_i)}{\Delta x^2} + \mathcal{O}(\Delta x).$$

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

Taylor Series Expansions to get the derivatives

- Different numerical approximations of the derivative can be found using TSE
- Using TSE, we also find the error order.
 FD/BD are first-order accurate, CD second-order accurate.
- You can find more accurate approximations by using more points
- The approximation of the second derivative requires at least one more point of information

Numerical Integration

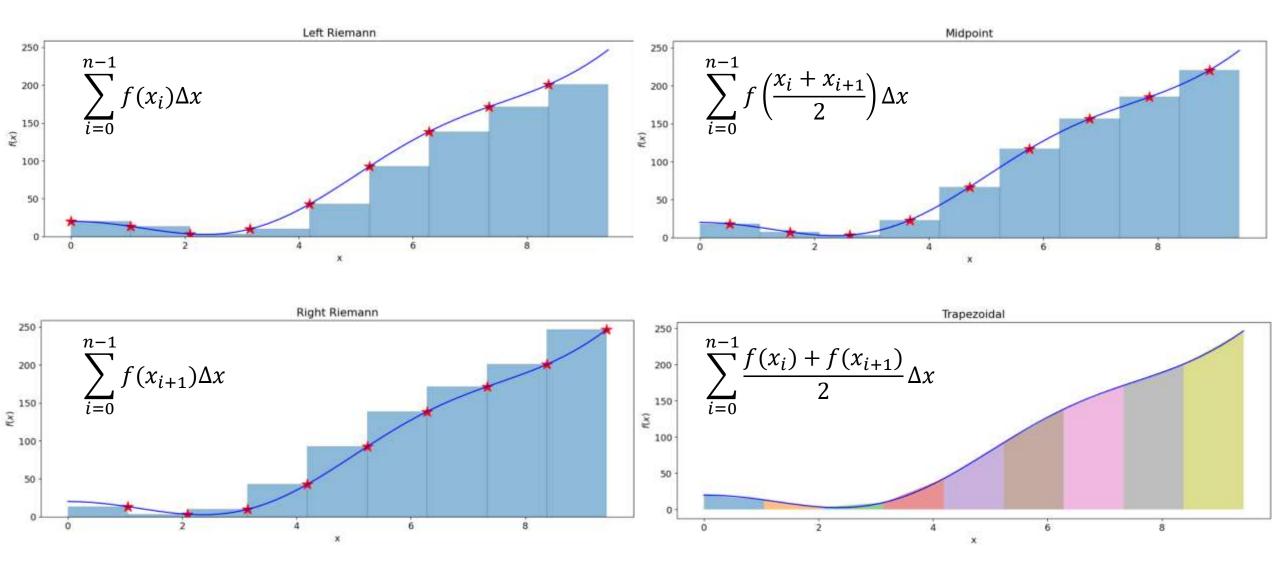
Numerical Integration

Numerical integration : a technique used to approximate the value of a defined integral, when it is not possible to obtain the exact value. $I = \int_{-b}^{b} f(x)dx$

Numerical integration rules:

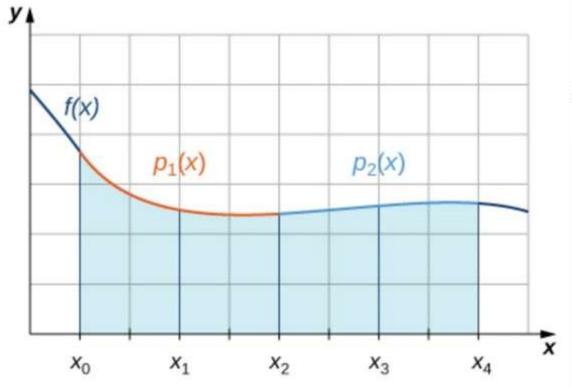
- Left Riemann
- Right Riemann
- Midpoint rule
- Trapezoidal rule
- •

Numerical Integration Methods: $I \approx$



Numerical Integration Methods: $I \approx$

Simpson's rule:



$$pprox \sum_{i=1}^{n/2} rac{f(x_{2i-2}) + 4f(x_{2i-1}) + f(x_{2i})}{6} 2\Delta x$$

Works only for equally spaced intervals

Numerical Integration Rules

Left Riemann Error
$$= |\int_a^b f(x) dx - \sum_{i=0}^{n-1} f(x_i) \Delta x|$$

Elaboration using TSE gives:

Left Riemann Error
$$=\left|ar{f}'(b-a)\Delta x/2
ight|$$
 therefore $\mathcal{O}(\Delta x)$

Right Riemann Error
$$=\left|ar{f}'(b-a)\Delta x/2
ight|$$
 therefore $\mathcal{O}(\Delta x)$

Midpoint Error
$$=\left|ar{f}''(b-a)\Delta x^2/2
ight|$$
 therefore $\mathcal{O}(\Delta x^2)$

Trapezoidal Error
$$=\left|ar{f}''(b-a)\Delta x^2/2
ight|$$
 therefore $\mathcal{O}(\Delta x^2)$

Simpsons Error
$$=\left|ar{f}''''(b-a)\Delta x^4/2
ight|$$
 therefore $\mathcal{O}(\Delta x^4)$

Consider the Standard Gumbel Probability Density Function (PDF) for a stochastic variable: $f(x) = e^{-(x+e^{-x})}$

Calculate the Cumulative Distribution (CDF) for x = 0, 2 and 4 by numerical integration

Theoretical solution for CDF:

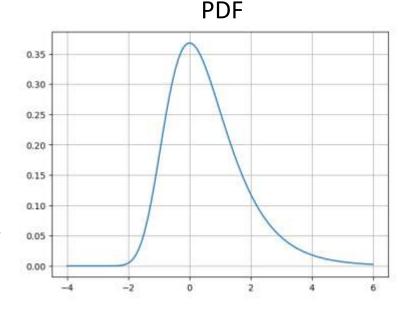
$$F(x) = \int_{-\infty}^{x} f(x) dx = e^{-e^{-x}}$$

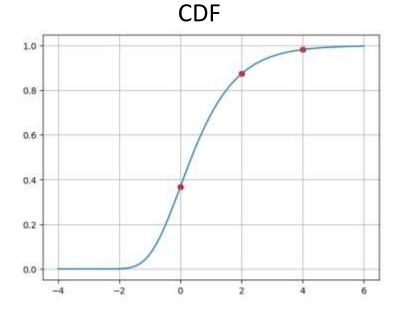
So, for target points:

x = 0: F(0) = 0.36787944

x = 2: F(2) = 0.87342302

x = 4: F(4) = 0.98185107



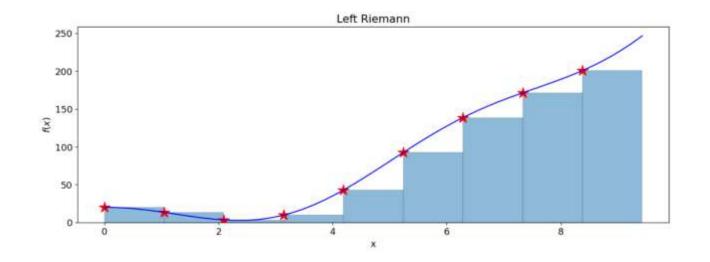


Numerical solutions for Left and Right Riemann, Midpoint and Trapezoidal rule (next slides):

$$F(x) = \int_{-\infty}^{x} f(x) \ dx$$

Left Riemann:

$$\sum_{i=0}^{n-1} f(x_i) \Delta x$$



```
Y1 = np.zeros(nsteps+1)

dx = 10. / nsteps
x1 = np.linspace(-4, 6, num=nsteps+1)
for i in range(nsteps):
     Y1[i+1] = Y1[i] + f(x1[i]) * dx
```

nsteps = 100

Target points (x=0, 2, 4):

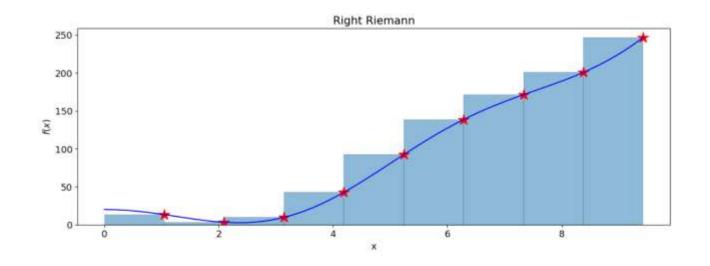
Left Riemann rule: [0.34948542 0.8674276 0.9809372] Analytical solution: [0.36787944 0.11820495 0.01798323]

$$F(x) = \int_{-\infty}^{x} f(x) \ dx$$

Right Riemann:

$$\sum_{i=0}^{n-1} f(x_{i+1}) \Delta x$$

or
$$\sum_{i=1}^{n} f(x_i) \Delta x$$



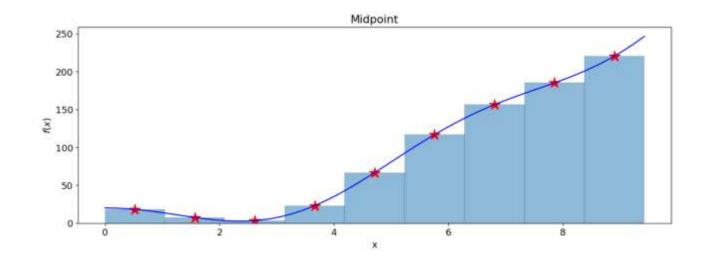
Target points (x=0, 2, 4):

Right Riemann rule: [0.38627336 0.8792481 0.98273553] Analytical solution: [0.36787944 0.11820495 0.01798323]

$$F(x) = \int_{-\infty}^{x} f(x) \, dx$$

Midpoint rule:

$$\sum_{i=0}^{n-1} f\left(\frac{x_i + x_{i+1}}{2}\right) \Delta x$$



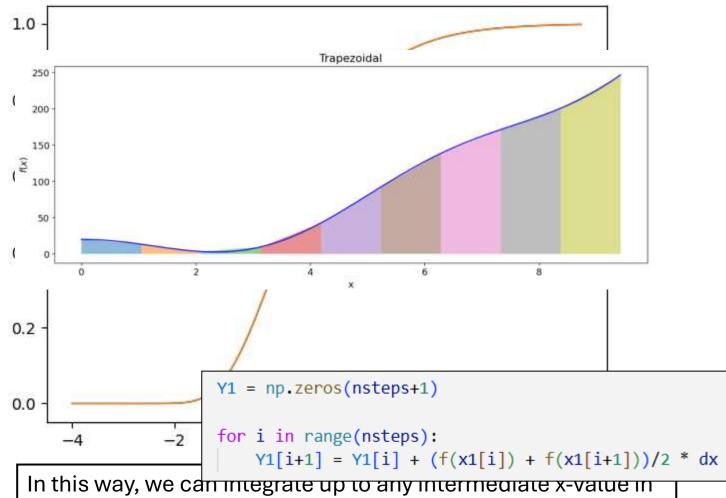
Target points (x=0, 2, 4):

Midpoint rule: [0.36787949 0.8734656 0.98185843] Analytical solution: [0.36787944 0.11820495 0.01798323]

$$F(x) = \int_{-\infty}^{x} f(x) dx$$

Trapezoidal rule:

$$\sum_{i=0}^{n-1} \frac{f(x_i) + f(x_{i+1})}{2} \Delta x$$



In this way, we can integrate up to any intermediate x-value in a range to reproduce the entire CDF by numerical integration

Target points (x=0, 2, 4):

Trapezoidal rule: [0.36787939 0.87333785 0.98183636] Analytical solution: [0.36787944 0.11820495 0.01798323]

Differential equations – ODEs, PDEs

Ice growth
First order ODE →

$$ho_{ice}rac{dh_{ice}}{dt}=-k_{ice}rac{T_{water}-T_{air}}{h_{ice}}$$

Beam deformation
Second order ODE →

$$\frac{d^2v}{dz^2} = \frac{-1}{EI} \left(-\frac{qz^2}{2} + qLz - \frac{qL^2}{2} \right)$$

1D consolidation
Second order PDE →

$$\frac{\partial p}{\partial t} = c_v \frac{\partial^2 p}{\partial z^2}$$

In Q1, we will NOT further consider PDEs!



1 000 1

How many constraints do the following equations need?

0%

0%

0%

0%

0%

0%

0%

0%

1 and 1
1 and 2
2 and 1
1 and 3
3 and 1
2 and 2
2 and 3
3 and 2

$$\rho_{ice} \frac{dh_{ice}}{dt} = -k_{ice} \frac{T_{water} - T_{air}}{h_{ice}}$$

and

$$\frac{d^2v}{dz^2} = \frac{-1}{EI} \left(-\frac{qz^2}{2} + qLz - \frac{qL^2}{2} \right)$$

How many constraints do the following equations need?

1 and 1			
		0%	11 (7) (7)
1 and 2			dh_{ice} , $T_{water} - T_{air}$
		0%	$\rho_{ice} = -\kappa_{ice}$
2 and 1			at n_{ice}
		0%	
1 and 3			and
		0%	and
3 and 1			12 120X 27 PM
		0%	$d^2 = 1 \left(a \pi^2 - a I^2 \right)$
2 and 2			$\frac{a \cdot v - 1}{a \cdot v} - \frac{qz}{a \cdot v} + aIz - \frac{qL}{a \cdot v}$
		0%	$dz^2 = EI \begin{pmatrix} 2 & qLz & 2 \end{pmatrix}$
2 and 3			$u_2 E_1 z$
		0%	
3 and 2	BEOLI		

RESULISSLIDE

Initial Value Problem

$$\frac{dy}{dt} = -y + e^{-t}$$



How many solutions do exist for this initial value problem?

How many solutions exist for this initial value problem?

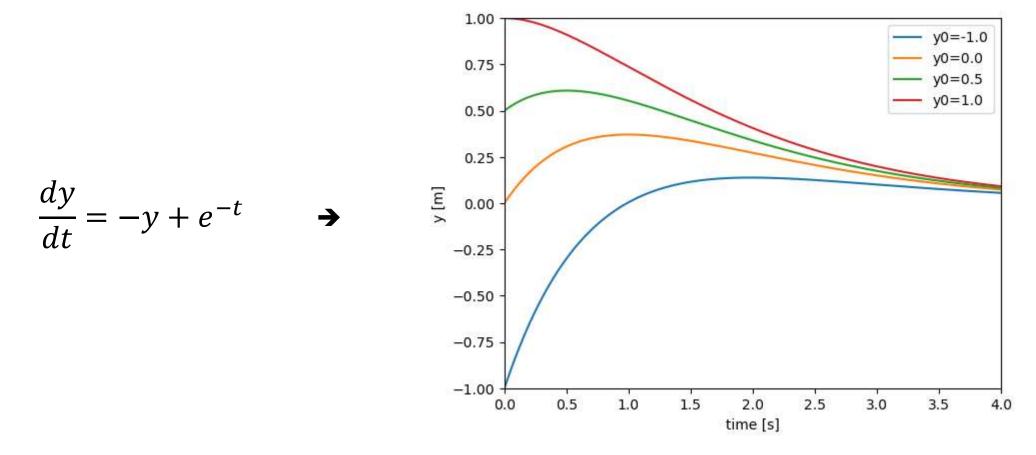
U	
	0%
1	
	0%
2	,
	0%
3	
	0%
5	
	0%
Infinite	
	0%
	*

How many solutions exist for this initial value problem?

0	
	0%
1	
	0%
2	
	0%
3	_
	0%
5	_
	0%
Infinite	_
	0%

RESULTS SLIDE

Initial Value Problem



Any ODE that has a time dependency, no matter the order, has a solution completely dependent on the initial value!

Summary

- Numerical solutions are used in all kinds of situations
 - Root finding (f(x) = 0)
 - Solving differential equations
 - Numerical integration
- Taylor Series Expansion
 - Derivatives (FD, BD, CD)
 - The relevance of taking higher-order terms into account
 - Accuracy order
- Numerical integration:
 - Left Riemann
 - Right Riemann
 - Midpoint
 - Trapezoidal rule
 - Simpson's rule
- Differential equations (to be continued next week):
 - Initial value problems
 - Boundary value problems

What you can expect further this week...

- Programming Assignment 1.2:
 - Working with GitHub
 - Creating reports with MarkDown
 - Programming: Fundamentals, visualizing a matrix, creating subplots, list comprehension, filling a matrix
- Wednesday 10:45-12:30: Workshop 1.2:
 - Integrating a Probability Density Function (PDF) into a Cumulative Distribution Function (CDF)
 - Integrating and differentiating an earthquake signal
- Friday 9:45-12:30: Group Assignment 1.2:
 - Integrating salt concentration in a river
 - TS approximation of a function
 - TSE approach to function derivatives and accuracy





Ronald Brinkgreve, Anna Störiko, Jaime Arriaga r.b.j.brinkgreve@tudelft.nl