

TDT4127 Programming and Numerics

Week 44

Solving ordinary differential equations

- More dimensions, stability and accuracy

Some facts about the exam

- **November 30, 09:00-13:00.**
 - Check location at **studentweb** (may not be available yet)
- The exam will be **digital**
 - Same as the auditorium exercises
 - <https://innsida.ntnu.no/wiki/-/wiki/norsk/digital+eksamen>
- Theory questions will have **multiple choice** answers
 - Mostly numerics, possibly some programming related
- Theory questions and numerical exercise(s) will be similar to those in **Auditorium exercise 2**
- You do **not** need to remember formulas for the numerics
 - But it will of course *help* if you are familiar with them!

Learning goals

- Goals
 - Solving **ordinary differential equations**
 - Analysis of algorithms:
 - *Explicit Euler method*
 - *Implicit Euler method*
 - *Heun's method*
 - Stability and accuracy
- Curriculum
 - Exercise set 9
 - But only in the interpretation of results



Numerical methods for ODEs

- Last week: **explicit Euler**, **implicit Euler**, **Heun's method**

- These schemes numerically solve the ODE

$$\dot{x}(t) = f(x(t), t)$$

- The **explicit Euler** method:

$$x^{j+1} = x^j + hf(x^j, t_j)$$

- The **implicit Euler** method:

$$x^{j+1} = x^j + hf(x^{j+1}, t_{j+1})$$

- **Heun's method:**

$$s^{j+1} = x^j + hf(x^j, t_j)$$
$$x^{j+1} = x^j + \frac{h}{2} \left(f(x^j, t_j) + f(s^{j+1}, t_{j+1}) \right)$$

Treating ODEs in many dimensions

- In more than one dimensions, equations are **vectorized**

$$\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), t)$$

- The methods are exactly the same, **but with vectors**
- The **explicit Euler** method:

$$\mathbf{x}^{j+1} = \mathbf{x}^j + h\mathbf{f}(\mathbf{x}^j, t_j)$$

- The **implicit Euler** method:

$$\mathbf{x}^{j+1} = \mathbf{x}^j + h\mathbf{f}(\mathbf{x}^{j+1}, t_{j+1})$$

- **Heun's method:**

$$\mathbf{s}^{j+1} = \mathbf{x}^j + h\mathbf{f}(\mathbf{x}^j, t_j)$$

$$\mathbf{x}^{j+1} = \mathbf{x}^j + \frac{h}{2} \left(\mathbf{f}(\mathbf{x}^j, t_j) + \mathbf{f}(\mathbf{s}^{j+1}, t_{j+1}) \right)$$

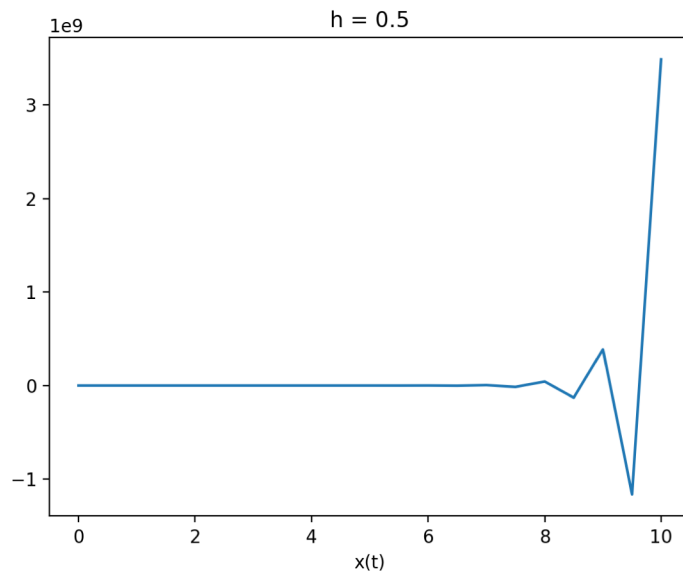
- Implementation difference: **vector addition. Numerical solver in several dimensions** for the **implicit Euler** method

Stability

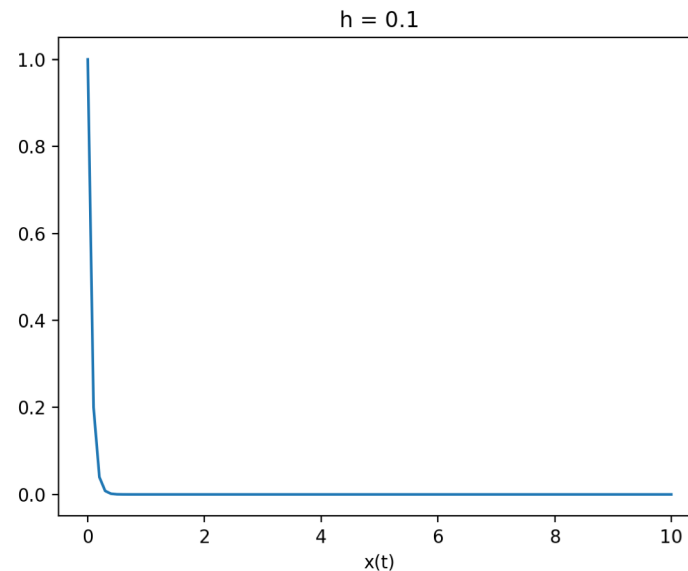
- When solving an ODE numerically, we need to choose an appropriate time **step size** h (i.e. **number of steps** N).
 - Too small h → takes long time to compute solution
 - Too large h → inaccurate (**bad**) or unstable (**worse**) solutions
- What is meant by **stability** and **instability**?
 - Instability is when the solution «blows up» when it's not supposed to
 - Stability is when it doesn't
- Test example: Apply the numerical method to the ODE
$$\dot{x}(t) = -\lambda x(t), \quad \lambda > 0, \quad x(0) = x_0$$
- This equation has the **strictly decreasing** solution
$$x(t) = x_0 e^{-\lambda t}$$

Numerical instability example

- Below: explicit Euler applied with various h to the ODE
 $\dot{x}(t) = -8x(t), \quad \lambda = 8, \quad x(0) = 1$



Left: **instability.**

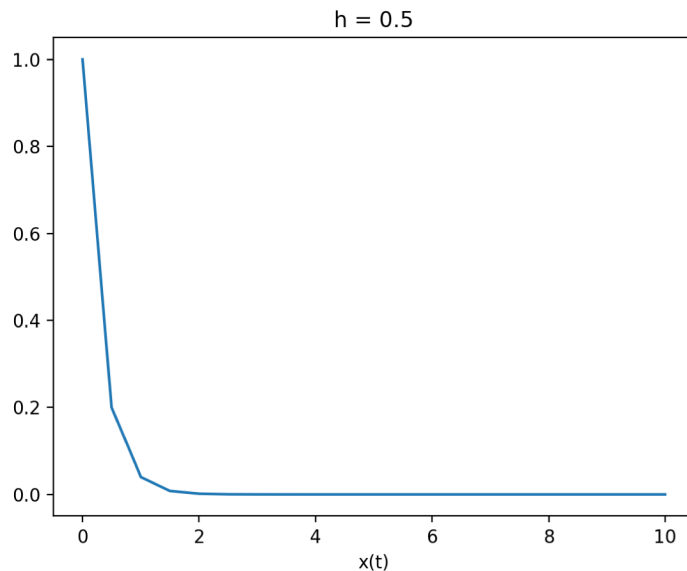


Right: **stability**

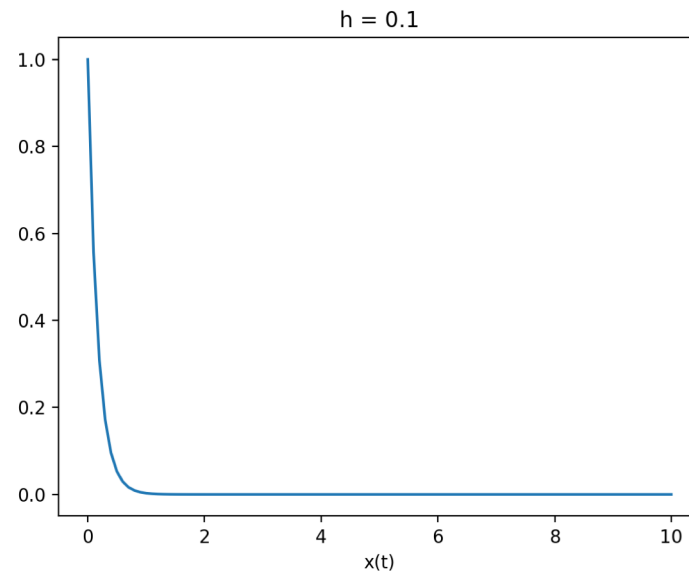
Numerical stability example

- Below: **implicit Euler** applied with various h to the ODE

$$\dot{x}(t) = -8x(t), \quad \lambda = 8, \quad x(0) = 1$$



Left: **stability.**



Right: **stability**

Numerical stability explained

- So what happens? Compare solutions for this particular ODE:

$$\dot{x}(t) = -\lambda x, \quad \lambda > 0, \quad x(0) = x_0$$

- **Explicit Euler:**

$$x_{n+1} = (1 - \lambda h)x_n = (1 - \lambda h)^2 x_{n-1} = \dots = (1 - \lambda h)^n x_0$$

- **Blows up** if $|1 - \lambda h| > 1$, i.e. if $\lambda h > 2$.
 - Must take $h < 2/\lambda$! This can be restrictive.

- **Implicit Euler:**

$$x_{n+1} = x_n - \lambda h x_{n+1}$$
$$x_{n+1} = \frac{1}{1 + \lambda h} x_n = \left(\frac{1}{1 + \lambda h} \right)^2 x_{n-1} = \dots = \left(\frac{1}{1 + \lambda h} \right)^n x_0$$

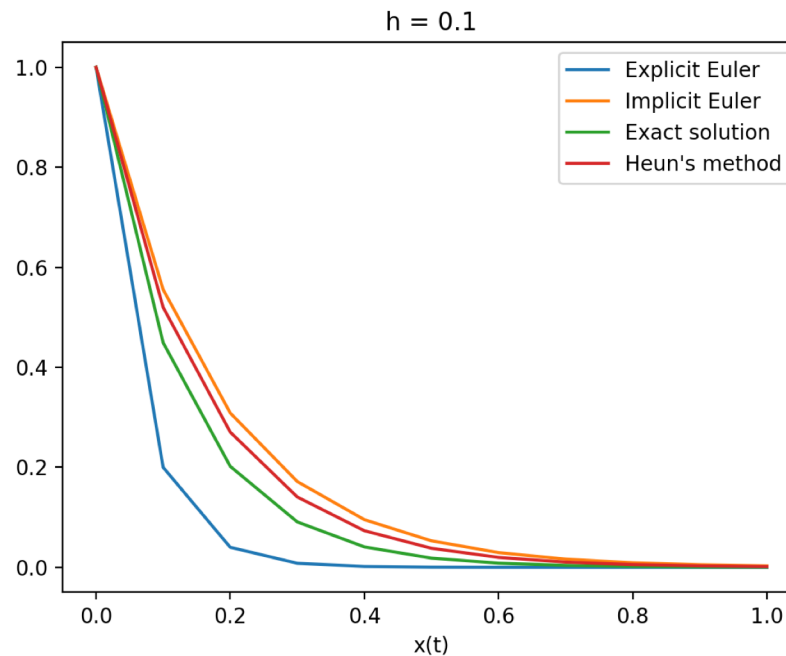
- Always decreasing, no matter the value of λ or h

Stability summarized

- Intuition: *different methods have different stability properties.*
 - Some methods are stable for larger or even **all** h
 - There is a large literature on this for *Runge-Kutta* methods
 - **Implicit** methods are typically **more stable** than explicit ones
 - No **explicit** method is stable for **all** h
 - All methods work with **small enough** h
 - But it may mean a restrictively very long computation time
- What about **Heun's method**?
 - **More stable** than **explicit Euler**, but it is **explicit** and has restrictions
- **Practical tips:** If you see unwanted blow-up or oscillations, try a smaller h first, before switching to another solver.

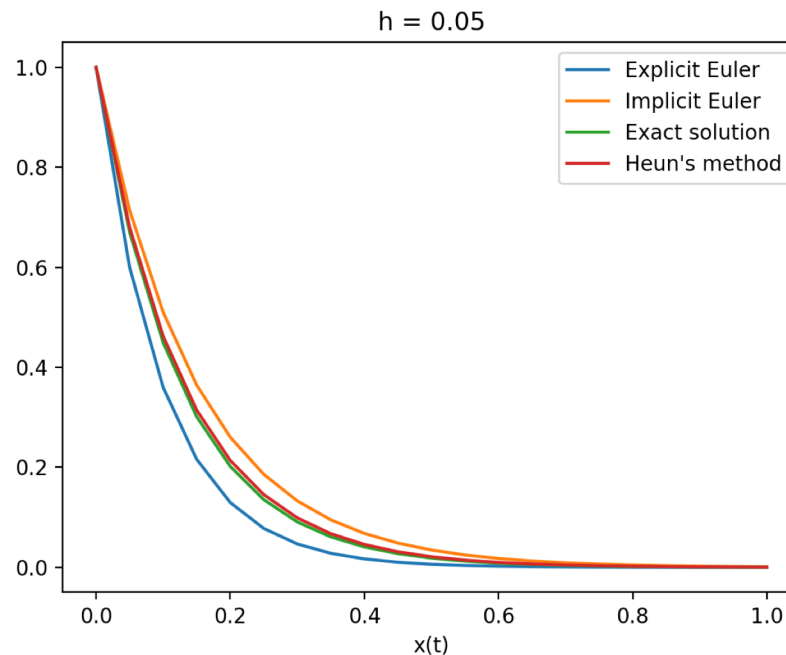
Accuracy

- Assuming our solutions don't blow up, the next question is: how **accurate** are they?
- Accuracy is a function of h



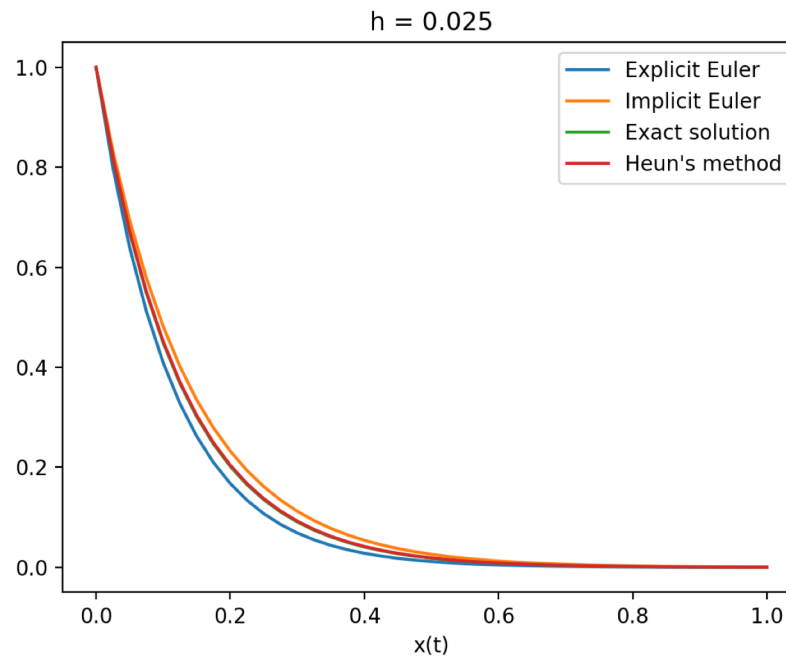
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Accuracy

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- Accuracy is a function of h



Accuracy

- Accuracy is measured in terms of the **global error**

$$e_j = |x(t_j) - x_j|, \quad t_j = jh$$

- A method is said to be of **order** p if for some constant C_j ,

$$e_j \leq h^p C_j$$

- **Explicit** and **implicit Euler** are both of **order 1**:

$$e_j^{\text{Euler}} \leq h C_j, \quad C_j = \frac{e^{L(t_j - t_0)} - 1}{L}$$

- **Heun's method** is of order 2:

$$e_j^{\text{Heun}} \leq h^2 C_j, \quad C_j = \frac{e^{L(t_j - t_0)} - 1}{L}$$

- Note: These C_j estimates are **very** conservative and should not be used in practice. ***The order is what's important.***

Accuracy

- **Practical consequence:** *order p methods improve errors by a factor 2^p when halving the step size.*
- The table below demonstrates the orders by considering the **global error** at $T = 1$ for the test problem from before.
- **explicit/implicit Euler:** $p = 1$, **Heun's method:** $p = 2$

| h | Explicit Euler | Implicit Euler | Heun's method |
|--------|----------------------|-----------------------|-----------------------|
| 0.01 | $9.62 \cdot 10^{-5}$ | $11.91 \cdot 10^{-5}$ | $30.54 \cdot 10^{-7}$ |
| 0.005 | $5.09 \cdot 10^{-5}$ | $5.66 \cdot 10^{-5}$ | $7.38 \cdot 10^{-7}$ |
| 0.0025 | $2.61 \cdot 10^{-5}$ | $2.76 \cdot 10^{-5}$ | $1.82 \cdot 10^{-7}$ |

- **High-order methods** gain a lot from smaller time steps!
 - Methods with **order 4** are often used!

Summary of ODE solvers

- **Explicit/implicit** methods
 - **Explicit**: Can calculate directly. **Explicit Euler** and **Heun's method**
 - **Implicit**: Need to solve an equation per step. **Implicit Euler**
- Multi-stage
 - Methods can have more than one **stage**. **Heun's method**
- Stability
 - Does the method blow up when applied to $\dot{x}(t) = -\lambda x$?
 - **Explicit** methods are **unstable** for too **large step sizes** h
 - **Implicit** methods are generally **more stable**
- Accuracy
 - A method is **order** p accurate if $e_j \leq h^p C_j$.
 - **implicit/explicit** Euler methods are **order 1**
 - **Heun's method** is **order 2**
 - **Can construct** integration methods of even **higher order**

Next weeks

- Three lectures left
 - **Adaptive Simpson** next week (November 9)
 - Last regular lecture
 - **Repetition** and **exam prep** on November 16 and November 23!
 - I will go through the numerics from **auditorium exercise 2** in detail
 - **Suggest other topics** you want me to cover
 - Otherwise, I'll pick them myself

Questions?