# Introduction to numerical astrophysics and python

Kristina Kislyakova University of Vienna, 08.10.2025

### **Schedule**

```
• 08.10.2025
                 Lecture #1 (Introduction)
• 15.10.2025
                 Lecture #2 (Visualization)
• 22.10.2025
                 Lecture #3 (Root finding)
• 29.10.2025
                 Lecture #4 (Interpolation and approximation)
• 05.11.2025
                 Lecture #5 (Derivation and integration)
• 19.11.2025
                 Lecture #6 (ODEs 1)
• 26.11.2025
                 Lecture #7 (ODEs 2)
• 03.12.2025
                 Lecture #8 (Fitting models to data)
• 10.12.2025
                 Lecture #9 (MCMC + sampling)
• 17.12.2025
                 Lecture #10 (DSMC, small lecture + no exercise)

    Winter Break

• 07.01.2026
                 Lecture #11 (FFT)
• 14.01.2026
                 Lecture #12 (Parallelisation)
                 Lecture #13 (Machine learning)
• 21.01.2026
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•	10.12.2025	Lecture #9 (MCMC + sampling)

Lecture #10 (DSMC, small lecture + **no exercise**)

#### **EXAMS:**

12.11.2025	Exam 1 (1st opportunity)
18.11.2025	Exam 1 (2nd opportunity)

28.01.2026 Exam 2 (1st opportunity) 04.02.2026 Exam 2 (2nd opportunity)

The students can select **only one date for each exam.** Registration for the exams will be organised on Moodle and announced in advance.

#### Winter Break

• 17.12.2025

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#### **Mini-projects:**

The deadline for mini-projects is **February**, **15th.** Mini-projects and exercises should be uploaded to Github.

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### **Exercises**

- Exercises take place in Seminarraum 1 (SE1) every Tuesday (see schedule on Moodle and u:find)
- The topic follows the previous lecture
- The next exercise becomes available on Wednesday after the lecture

#### **EXERCISE GROUPS:**

12.11.2025 Group 1 (Alina Böcker)

18.11.2025 Group 2 (Kristina Kislyakova)

#### **Registration:**

Registration for exercise groups opens after this lecture on October 8th at 12:00 and ends on October 10th at 23:59.

### **Exercises**

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- The topic follows the previous lecture
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Swapping exercise groups: if you fail to register for the exercise group you wish to visit, please find a student from the other group willing to swap a place with you. Then write me an email with both of you in cc, and I will change the group assignment for both of you

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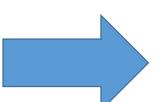
### This lecture

- Basic concepts in numerical astronomy
- A few words about different languages
- Useful software
- Introduction to python
- Introduction to Git/Github

## **Concepts**

### Why to use numerical methods at all?

Bessel functions of the first and second kind



$$1) \int x J_0(x) dx = x J_1(x)$$

2) 
$$\int x^2 J_0(x) dx = x^2 J_1(x) + x J_0(x) - \int J_0(x) dx$$

3) 
$$\int x^m J_0(x) dx = x^m J_1(x) + (m-1)x^{m-1} J_0(x) - (m-1)^2 \int x^{m-2} J_0(x) dx$$

4) 
$$\int \frac{J_0(x)}{x^2} dx = J_1(x) - \frac{J_0(x)}{x} - \int J_0(x) dx$$

5) 
$$\int \frac{J_0(x)}{x^m} dx = \frac{J_1(x)}{(m-1)^2 x^{m-2}} - \frac{J_0(x)}{(m-1)x^{m-1}} - \frac{1}{(m-1)^2} \int \frac{J_0(x)}{x^{m-2}} dx$$

6) 
$$\int J_1(x) dx = -J_0(x)$$

7) 
$$\int x J_1(x) dx = -x J_0(x) + \int J_0(x) dx$$

8) 
$$\int x^m J_1(x) dx = -x^m J_0(x) + m \int x^{m-1} J_0(x) dx$$

9) 
$$\int \frac{J_1(x)}{x} dx = -J_1(x) + \int J_0(x) dx$$

10) 
$$\int \frac{J_1(x)}{x^m} dx = -\frac{J_1(x)}{mx^{m-1}} + \frac{1}{m} \int \frac{J_0(x)}{x^{m-1}} dx$$

11) 
$$\int x^{\nu} J_{\nu-1}(x) dx = x^{\nu} J_{\nu}(x)$$

12) 
$$\int x^{-v} J_{v+1}(x) dx = -x^{-v} J_{v}(x)$$

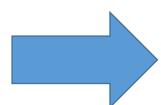
13) 
$$\int x^m J_{\nu}(x) dx = -x^m J_{\nu-1}(x) + (m+\nu-1) \int x^{m-1} J_{\nu-1}(x) dx$$

14) 
$$\int x J_{\nu}(\alpha x) J_{\nu}(\beta x) dx = \frac{x \left[ \alpha J_{\nu}(\beta x) J_{\nu}'(\alpha x) - \beta J_{\nu}(\alpha x) J_{\nu}'(\beta x) \right]}{\beta^2 - \alpha^2}$$

15) 
$$\int x J_{v}^{2}(\alpha x) = \frac{x^{2}}{2} \left[ J_{v}'(\alpha x) \right]^{2} + \frac{x^{2}}{2} \left( 1 - \frac{v^{2}}{\alpha^{2} x^{2}} \right) \left[ J_{v}(\alpha x) \right]^{2}$$

### Why to use numerical methods at all?

Bessel functions of the first and second kind



Analytical methods can be too complex or too time consuming

$$1) \quad \int x J_0(x) \, dx = x J_1(x)$$

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**ODEs and PDEs** 
$$\frac{df}{dt} = t^2 + 1 \qquad \frac{\partial \rho}{\partial t} + \frac{\partial (\rho v)}{\partial r} = 0$$

Ordinary and Partial Differential Equations are very important in most large numerical tasks. They usually require very different methods for solving.

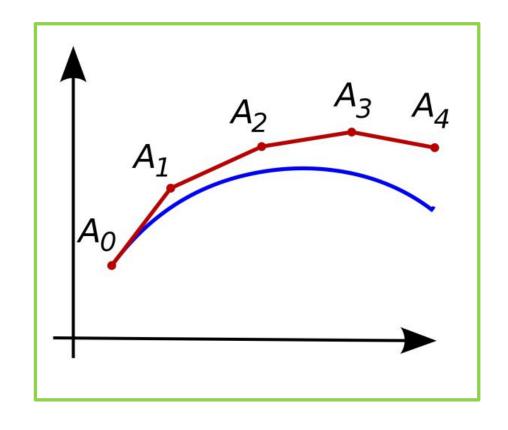
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Ordinary and Partial Differential Equations are very important in most large numerical tasks. They usually require very different methods for solving.

We will discuss the solution for ODEs in the lectures dedicated to them. One can often use similar or the same solvers for PDEs.

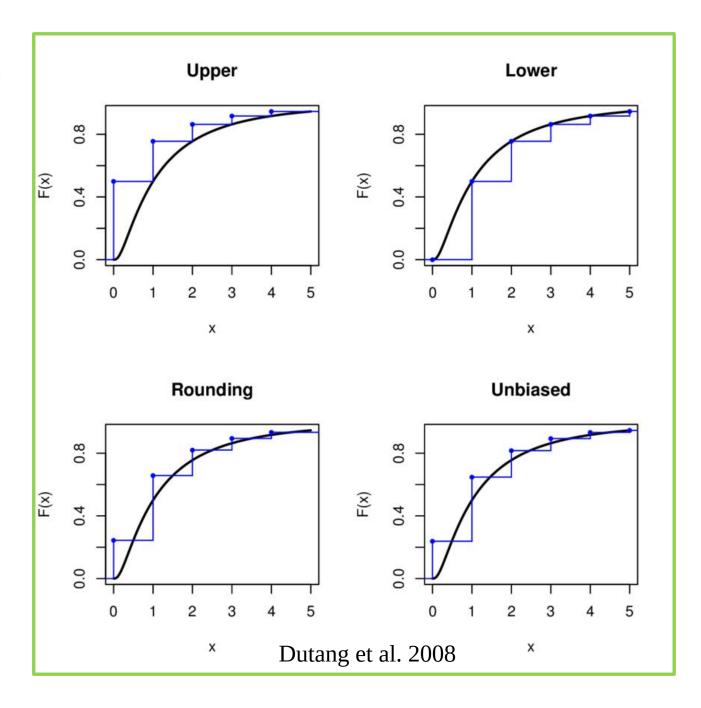
### **Initial conditions**

In many methods, before iteration can start, a guess of the solution is necessary. Quite often this guess will need to be close to the true solution.



### **Discretisation:**

the process of turning a continuous problem into a discrete problem. For example, you might break down space into smaller regions, often called 'cells'.



### **Errors**, for instance:

- round-off error due to finite accuracy of computers
- discretization error because discretization is only approximate

$$(1/3 = 0.333333....)$$

$$\frac{\partial y}{\partial x} \approx \frac{y_2 - y_1}{x_2 - x_1}$$

truncation error

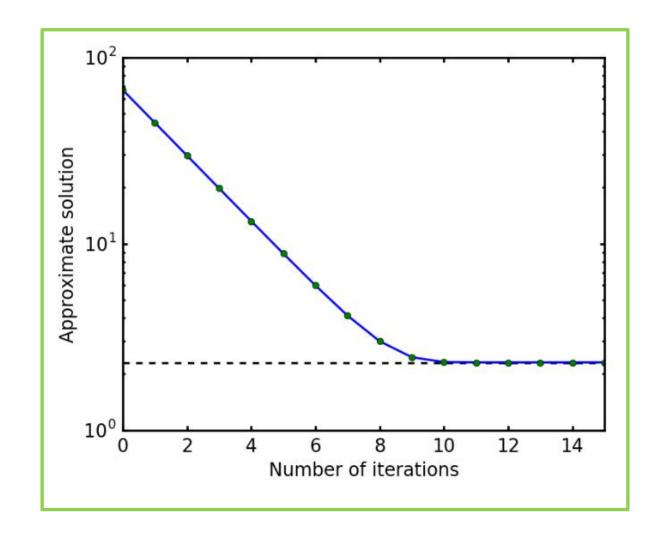
$$e^x = 1 + x + rac{x^2}{2!} + rac{x^3}{3!} + rac{x^4}{4!} + \cdots$$

### Convergence

You should stop iterating once the approximation stops to change with each iteration.

### **Steady state**

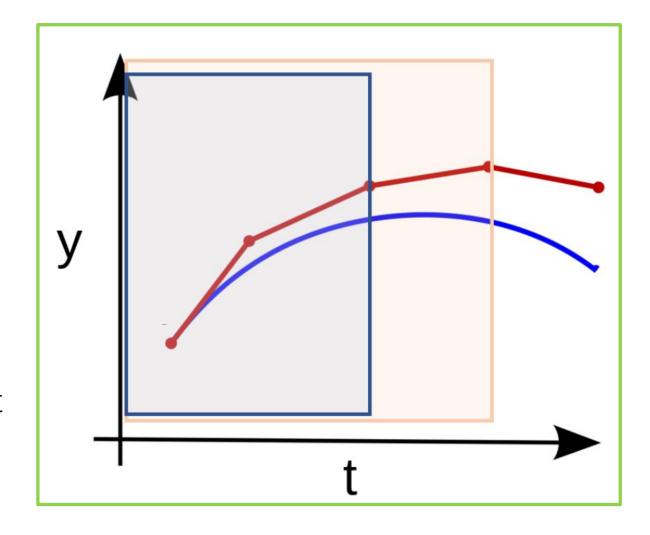
The system has reached the final result and stopped changing



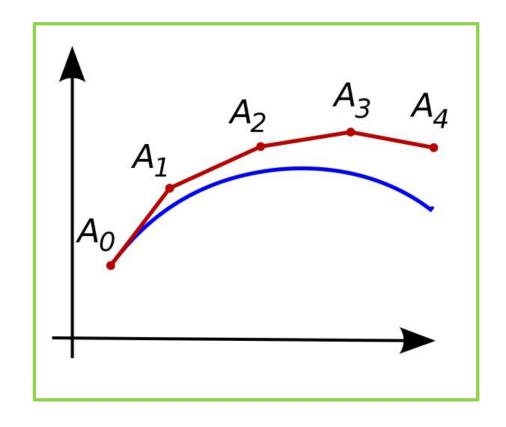
# **Explicit and implicit** methods

The implicit method uses the slope of the next unknown time-step, and therefore must be solved numerically.

The explicit method uses the slope at a known time-step to find the slope at the next time-step, so it can be determined directly



Iteration is the process of repeating a calculation many times. Most numerical methods use iteration in some form.



### **Initial value problem**

Often you need a very educated guess about your solution *in advance* 

INITIAL CONDITIONS: 
$$y_0 = y(t_0)$$

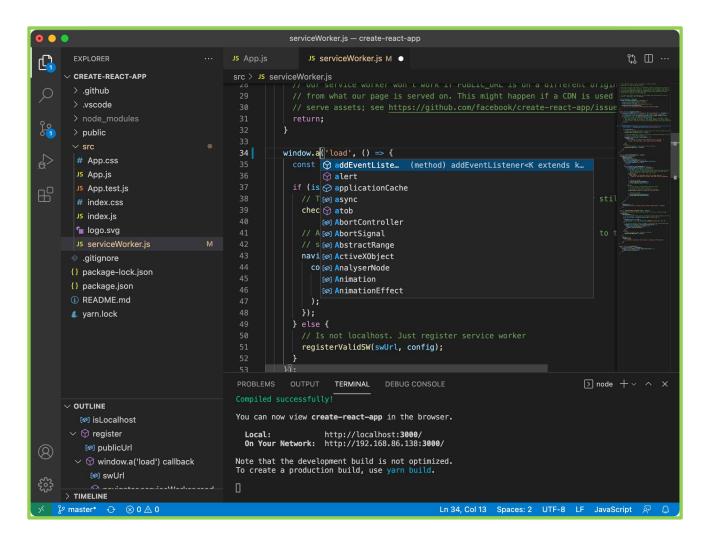
PHYSICAL MODEL:  $\frac{dy}{dt} = f(y, t)$ 

DESIRED:  $y(t)$ 

### Useful software / tools

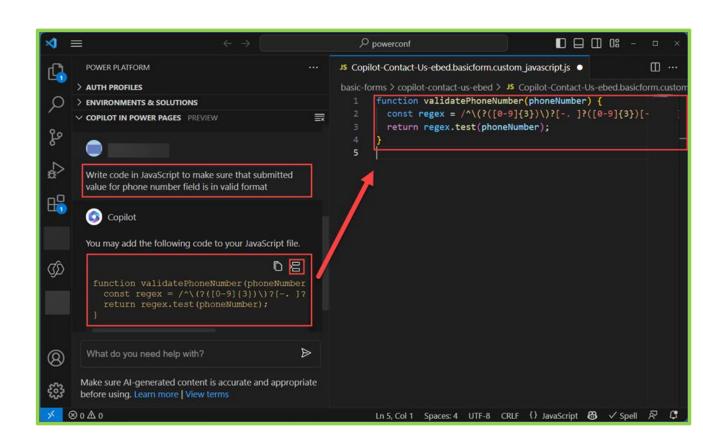
### Visual studio code

- Free to use
- Plugins. So many plugins!
- Integratable with github, co-pilot, etc.
- Support of multiple languages



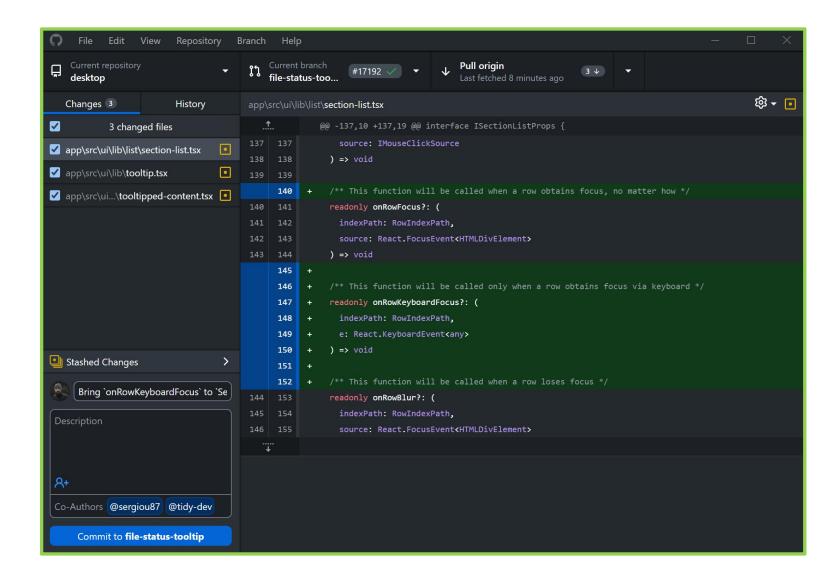
### Visual studio code & Co-pilot

- Co-P is not free to use, but there is an (old, limited) version available
- Integratable with VSC
- Support of multiple languages
- Existing solutions at your fingertips



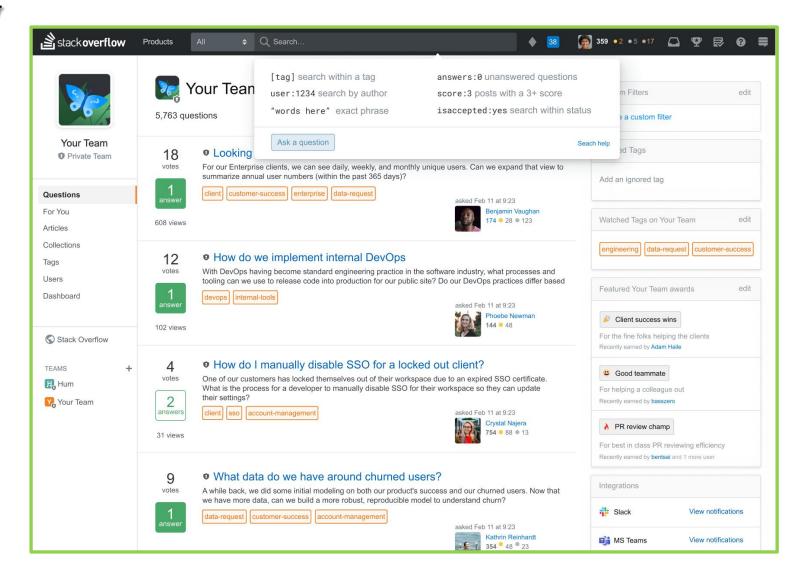
### **Github**

- Free to use
- Version control
- Code available everywhere
- Easy to share



### **Stack Overflow**

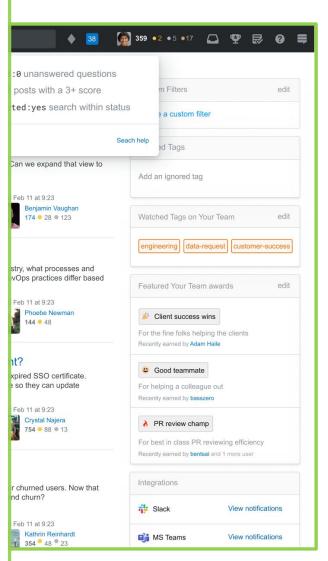
- Free to use
- The power of the hive mind



### Stack Overflow

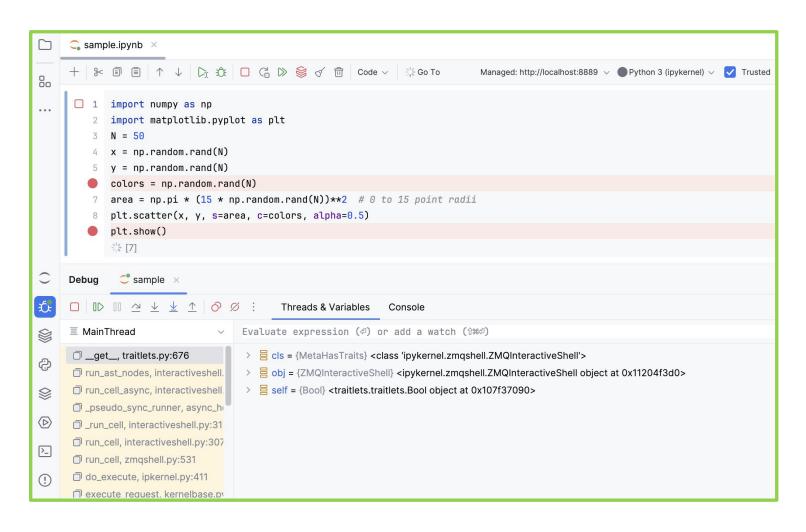
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### Python-specific: Jupyter notebook

- Includes debugging tools
- Generally easy to use



### A quick note on different languages

### ~Two big groups:

### • C, C++, Fortran:

- Compilation needed
- Advantages: fast!!! Should be used for large calculations.
- Disadvantages: can be slower to program with,no built-in plotting procedures

### • Python, IDL, R:

- Interpreted → no compilation
- Advantages: often very convenient to program with built-in procedures; plotting is easy
- Disadvantages: slow, really really slow

### **Examples**

Lecture1\_MainExample.py
Lecture1\_MainExample\_Classes.py
Intro\_to\_python-JupyterNotebook.pdf
Intro\_to\_python.ipynb
Python syntax hints.ipynb

A great lecture on Git and Github: https://people.irisa.fr/Anthony.Baire/git/git-for-beginners-handout.pdf

Also see the files on Github uploaded to Moodle