

Imperial College  
London



# Introducing Students to Open-Source Research Codes

Solving PDEs in Python

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## Acknowledgements



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## Criteria to Evaluate Computational Tools

1. Availability: Are they free/low cost and available to users readily?
  2. Ease of use: Is there a steep learning curve / are resources available?
  3. Scalability: Can these tools be used for both small tasks on a laptop and large tasks on a HPC cluster?
  4. Learning Objectives: Is the course focused on theory or application?
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## PDE Solver Landscape

### Commercial Codes

- Often used in teaching and professional settings
- Extensive resources available
- Very expensive

### Open-Source (C++ etc.)

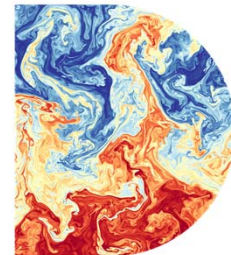
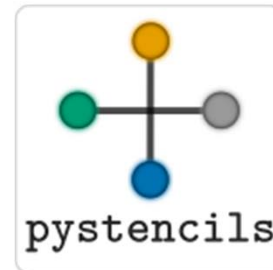
- Well-established in professional/research uses
- Limited resources available
- Steep learning curve
- Free

### DIY Codes

- Typically simple finite difference codes used for teaching purposes
  - Often scales poorly
  - Bespoke codes that are fast and scale are very inaccessible
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## Examples of PDE Solvers in Python

1. FEniCS: Finite Element Method (FEM)
2. FiPy: Finite Volume Method (FVM)
3. Dedalus: Spectral Methods
4. Pystencils: Finite Difference Method (FDM)



## Open-Source PDE Solvers in Python

- ✓ Very available due to their open-source nature
  - ✓ Easy to use Python front-ends
  - ✓ Often coupled to a fast backend written in C++ / compiled languages e.g., PETSc
  - ✓ Many discretization methods (FVM, FEM, FDM, Spectral Methods).
  - ✓ Adequate documentation and access to developers
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## Lesson Objective

- To introduce 2<sup>nd</sup>/3<sup>rd</sup> year students to PDE solver codes
    - Provide students the foundation to use such codes in their own research/work
    - Complement the teaching of advanced engineering math / transport
    - Introduce basic concepts in numerical methods
    - Establish some familiarity with coding
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## Coding Environment

- Anaconda was selected as the coding environment for multiple reasons
  - Convenient coding environment well suited for Windows
  - Comes with the Spyder IDE → Low barriers to entry for students
  - Good introduction to reproducible coding environments and containerization → Inculcate good practices





## Lesson Structure

- 4 exercises of increasing complexity are developed with progression as the guiding philosophy
  - FiPy was used as the main solver code as it is very intuitive
  - Extensive notes were provided for each exercise outlining the basic numerical methods concepts and mathematics (e.g. analytical solutions where available)
  - Students were advised to treat the provided code as boiler plate
  - Some exercises are further broken down into parts that explore specific aspects
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## Exercises

### Exercise 1: Laplace Equation

$$\nabla^2 \phi = 0$$

$$\phi(0, y) = \phi(1, y) = \phi(x, 0) = 0, \phi(x, 1) = 1$$

#### Objectives:

- Set up and run a simulation
- Visualize the results
- Benchmark the results to the analytical solution

### Exercise 2: Diffusion Equation

$$\frac{\partial c}{\partial t} = D \nabla^2 c$$

$$c(x = 0, t) = 1, c(x = 1, t) = 0, c(x, t = 0) = 0$$

#### Objectives:

- Explore simple timestepping schemes
- Appreciate the features of the transient solution
- Introduce source and sink terms



## Exercises

### Exercise 3: Advection-Diffusion

$$\frac{\partial c}{\partial \tilde{t}} + \frac{\partial c}{\partial \tilde{x}} = \frac{1}{\text{Pe}} \frac{\partial^2 c}{\partial \tilde{x}^2}$$

#### Objectives:

- Appreciate the mechanism of advective transport
- Appreciate the issues related to the numerical solution
- Implement coupled problems

### Exercise 4: Cahn-Hilliard Equation

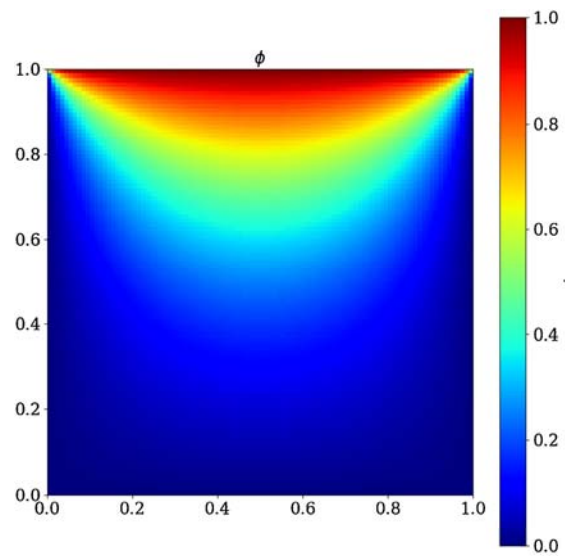
$$\tilde{\mu}_{AB} = \frac{\partial g}{\partial a} - \tilde{\kappa} \tilde{\nabla}^2 a$$

$$\frac{\partial a}{\partial \tilde{t}} = \tilde{\nabla} \cdot (a(1-a)\tilde{\nabla} \tilde{\mu}_{AB})$$

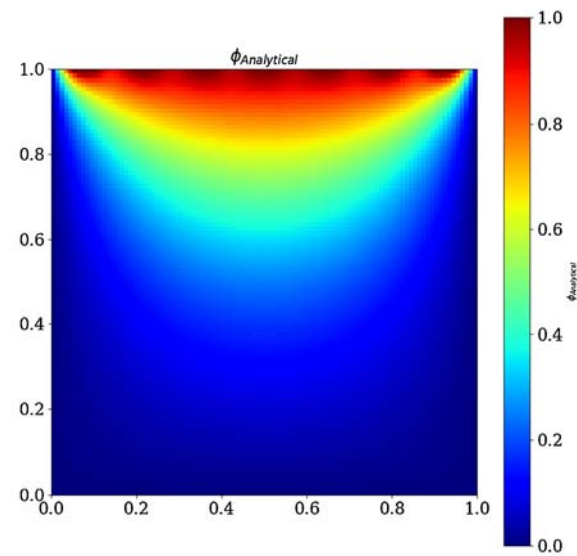
#### Objectives:

- Explore an advanced problem
- Advanced postprocessing in ParaView

## Exemplar Results

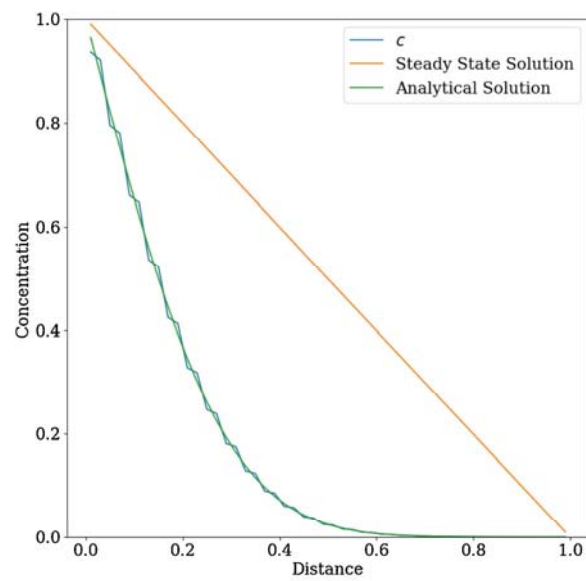


(a) Numerical solution

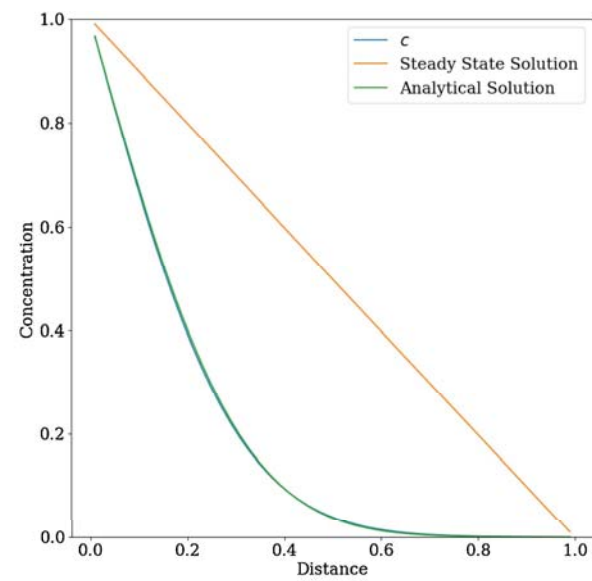


(b) Analytical solution truncated to the 7th term

## Exemplar Results

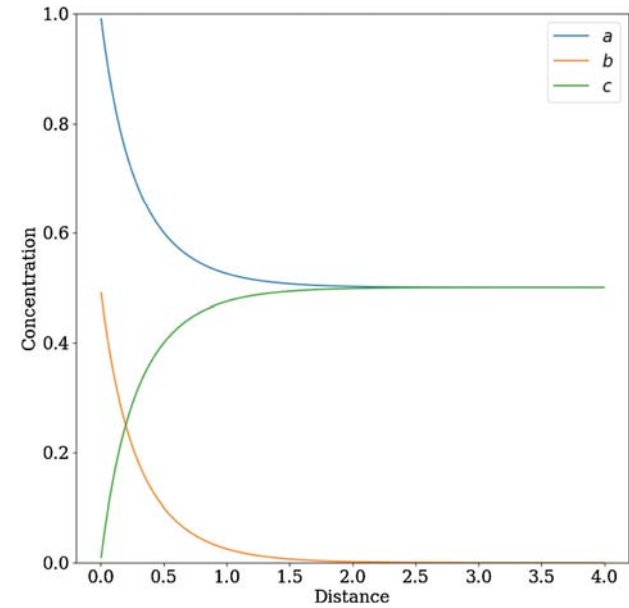
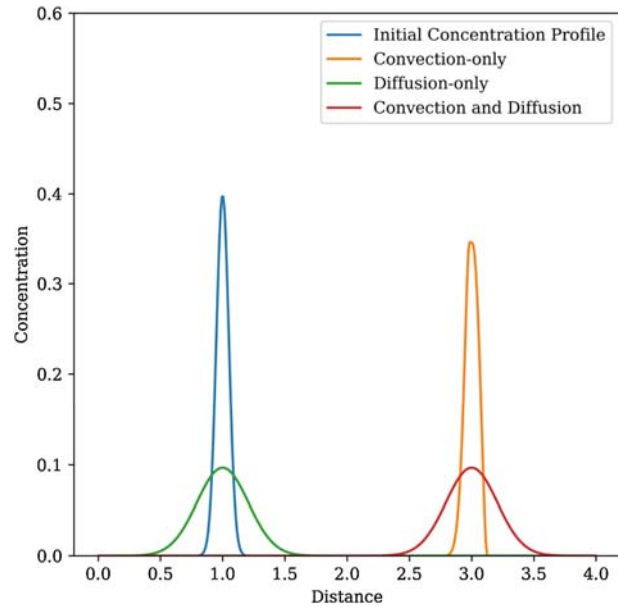


(a) Explicit forward Euler time-stepping with  $\Delta t = 0.0001998$

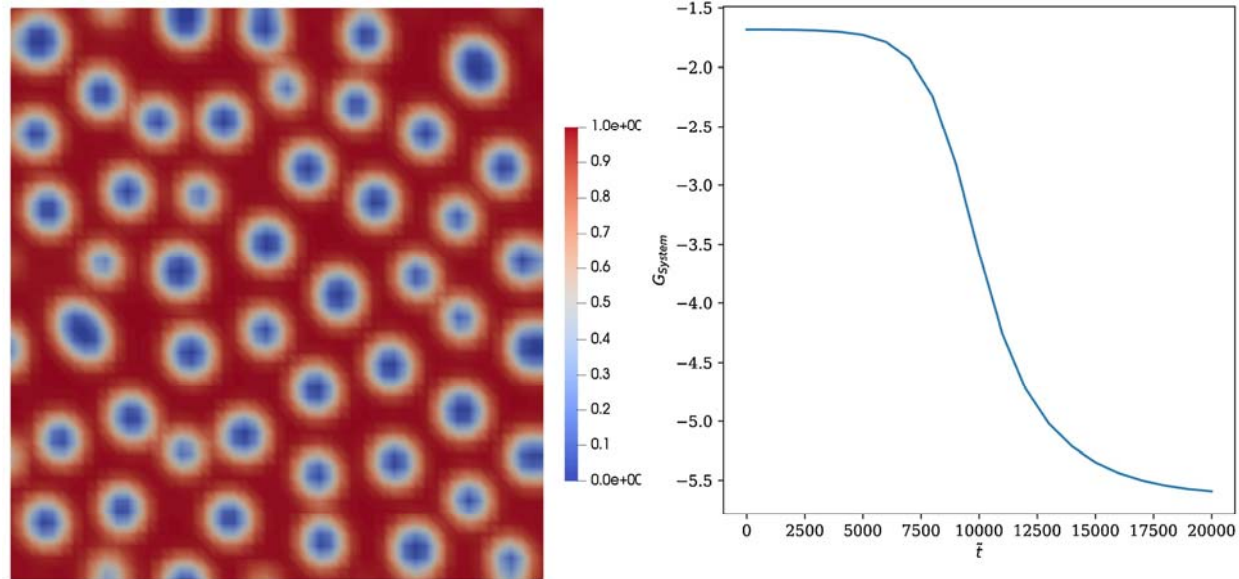


(b) Implicit backward Euler time-stepping with  $\Delta t = 0.0018$

## Exemplar Results



## Exemplar Results



(a) Polymer blend morphology @  $\bar{t} = 20000$ . The volume fraction colour bar is included for reference.

(b) Evolution of  $G_{\text{System}}$  over time, demonstrating how the demixing process decreases the total Gibbs energy.

## Student Perception and Feedback

Students valued the visualizations of various mathematical concepts made possible:

*“I felt the course explained PDEs better than we got taught it, because you can see it as it was happening and see how one thing affects the other. The graphs gave a more visual approach than when it was taught...”*

Students also indicated they would like to know more about the mechanics of FiPy and Python:

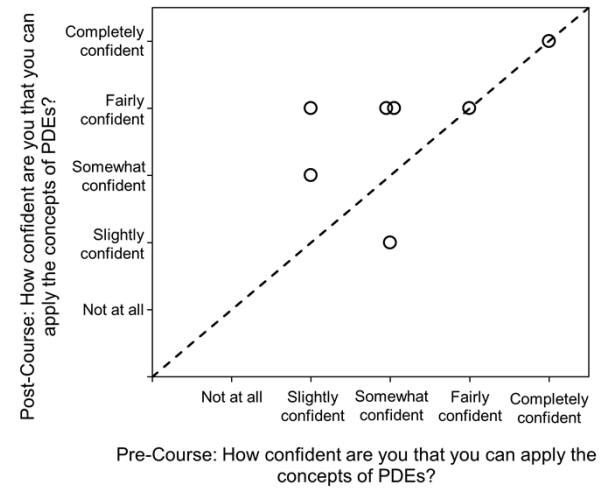
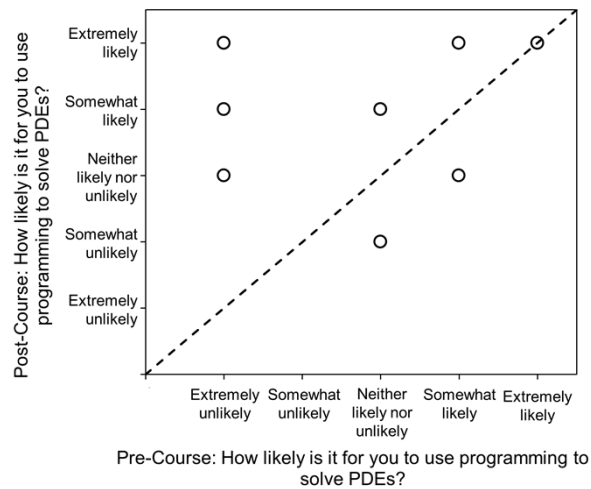
*“I was more thinking in the future. For example, if I want to use this myself to solve some problems. Then it would then be nice to also be introduced to how this whole thing is structured.”*

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## Student Perception and Feedback

The short course has also helped to improve students' self efficacy in programming and PDEs



## Conclusions

- A short course successfully introducing students to PDE solvers was developed.
  - Student feedback indicated they had an improved perception of Python and increased self-efficacy in advanced engineering math.
  - Educators can employ these tools to enhance their teaching
  - The course material is available at <https://github.com/pavanninguva/pde-Solver-Course>
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