

Numerical methods for physics

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Computational physics

Computational physics aims at finding numerical solution to problems for which a *quantitative theory* already exists.

What does a computational physicist do:

- develops/chooses a quantitative theory (physical model)
- develops/chooses appropriate algorithms (numerical analysis)
- implements the algorithms (computer code)
- runs *numerical simulations* and *interprets* the results

Physical model

$$\frac{\partial \rho}{\partial t} + \nabla \cdot [\rho \mathbf{v}] = 0$$

$$\frac{\partial (\rho \mathbf{v})}{\partial t} + \nabla \cdot [\rho \mathbf{v} \mathbf{v} - \mathbf{B} \mathbf{B} + \mathbf{P}^*] = 0$$

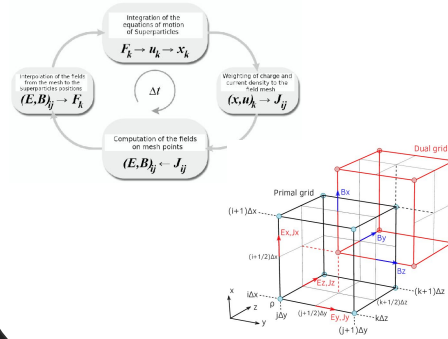
$$\frac{\partial E}{\partial t} + \nabla \cdot [(E + P^*) \mathbf{v} - \mathbf{B} (\mathbf{B} \cdot \mathbf{v})] = 0$$

$$\frac{\partial \mathbf{B}}{\partial t} - \nabla \times (\mathbf{v} \times \mathbf{B}) = 0$$

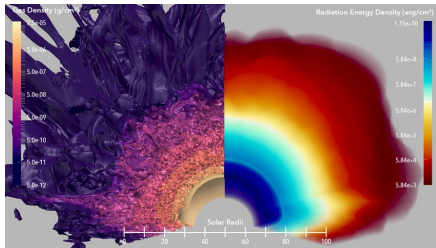
$$P^* = P + \frac{\mathbf{B} \cdot \mathbf{B}}{2}$$

$$E = P/(\gamma - 1) + \frac{\rho (\mathbf{v} \cdot \mathbf{v})}{2} + \frac{\mathbf{B} \cdot \mathbf{B}}{2}$$

Algorithms and Numerical Methods



Analysis and Interpretation



Computer code

```

-----BEGIN PROGRAM
use, saveObject
implicit none
call open_deviceIO
+
call newGrid
...
call close_deviceIO
end
-----END OF MAIN

subroutine useGrid
use, saveObject
implicit none
integer, parameter :: nsize=1024
type(deviceIO) :: dev1, dev2, dev3, dev4
real :: nsize1(1024), nsize2(1024), nsize3(1024), nsize4(1024)

call random_number(nsize1)
call random_number(nsize2)
call random_number(nsize3)
call random_number(nsize4)

dev1=allocate_dev('real', nsize1)
dev2=allocate_dev('real', nsize2)
dev3=allocate_dev('real', nsize3)
dev4=allocate_dev('real', nsize4)

call transfer_4(nsize1, dev1, time1)
call transfer_4(nsize2, dev2, time1)
call transfer_4(nsize3, dev3, time1)
call transfer_4(nsize4, dev4, time1)

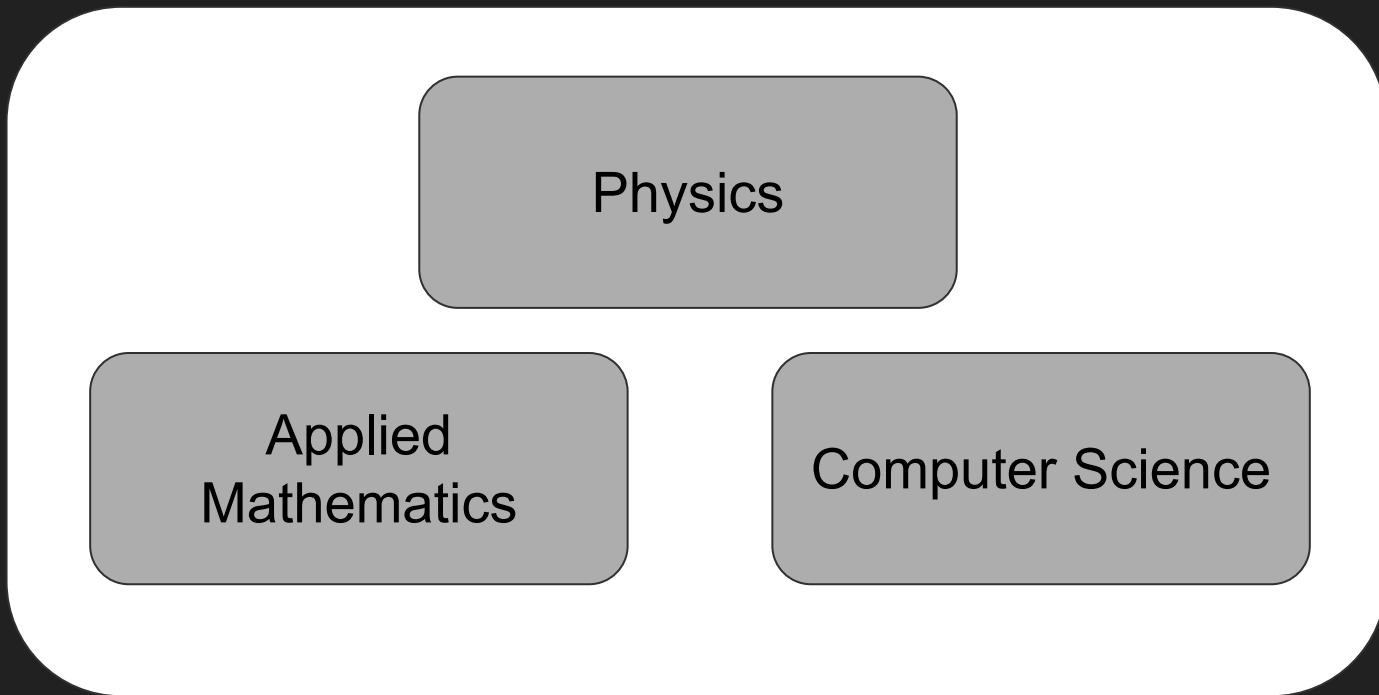
call transfer_4(nsize1, dev1, time2)
call transfer_4(nsize2, dev2, time2)
call transfer_4(nsize3, dev3, time2)
call transfer_4(nsize4, dev4, time2)

call transfer_4(nsize1, dev1, time3)
call transfer_4(nsize2, dev2, time3)
call transfer_4(nsize3, dev3, time3)
call transfer_4(nsize4, dev4, time3)

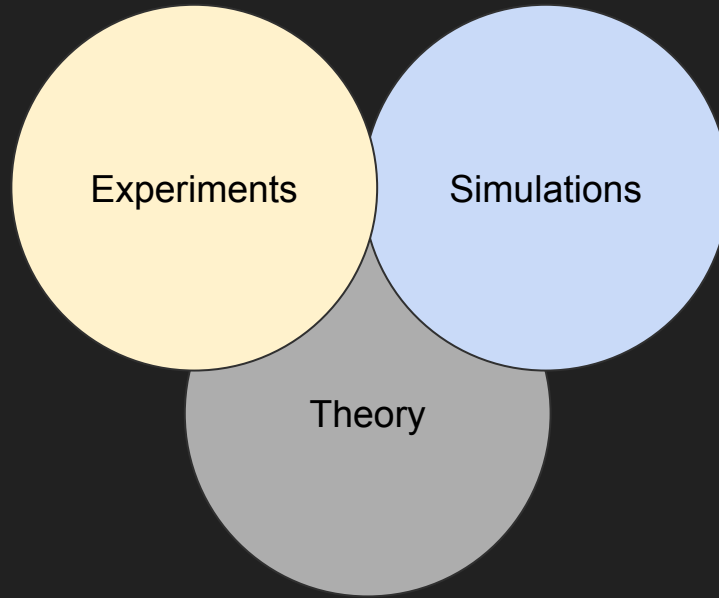
call transfer_4(nsize1, dev1, time4)
call transfer_4(nsize2, dev2, time4)
call transfer_4(nsize3, dev3, time4)
call transfer_4(nsize4, dev4, time4)

end subroutine useGrid
    
```

Computational physics is multidisciplinary



Theory - Experiments...and Simulations



Why computational physics?

Analytical solutions of physical problems are generally found for systems that are often too simplified.

The vast majority of realistic problems in physics are *complex* and require a numerical approach.

For example:

- Non-analytical solutions or no mathematical tools able to describe them
- Too many degrees of freedom: Vlasov equation in 3D/3V + time + electromagnetic fields
- Chaotic, stochastic, highly-non-linear....
- Solution is known but too many "particles" and need for fast computations

Numerical methods and algorithms

Depending on the physical problem, many numerical techniques exist for

- finding roots (i.e. $f(x) = 0$)
- system of linear equations
- ordinary and partial differential equations
- integration,
- data fitting
- discrete Fourier transform
- matrix eigenvalue
-

Numerical methods and algorithms

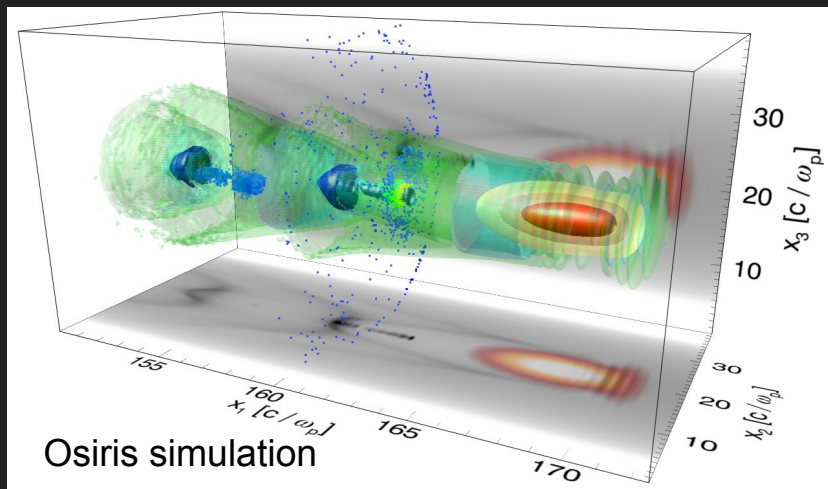
"Simple" building blocks are needed to construct complex numerical simulations

For example: Simulation of particle acceleration in laser-plasma interaction

→ interpolation/extrapolation

→ numerical integration

→ ...



Simulations can produce x 100
TB of output data that need to
be stored and processed
→ big data

Advantages and disadvantages of numerical calculations

- Fast
 - But often Nature is faster...no forecasting (to model 1 ns of laser interaction it takes a day)
- Large memory and data storage
 - Real systems are too big: 12 g of Carbon contain 6×10^{23} particles
- Finite representation of numbers and approximation of equations → errors
 - A good scientist needs to know the limitations of computational results
-

The Numerical Methods course

Objectives of the course

1. Learn basic and not-so-basic numerical methods and algorithms
2. Learn to implement different numerical methods and algorithms
3. Use computers to solve problems in physics and interpret the results

...and also:

- learn PYTHON, jupyter notebooks, ...
- work with others to solve problems

Python

- high-level, interpreted language
- very flexible, thousands of libraries
- standard choice in many R&D applications and beyond
- huge amounts of information and documentation online
 - *it can be daunting to find the correct/relevant information*

Check out the website: <https://www.python.org/>

List of python packages: <https://pypi.org/>

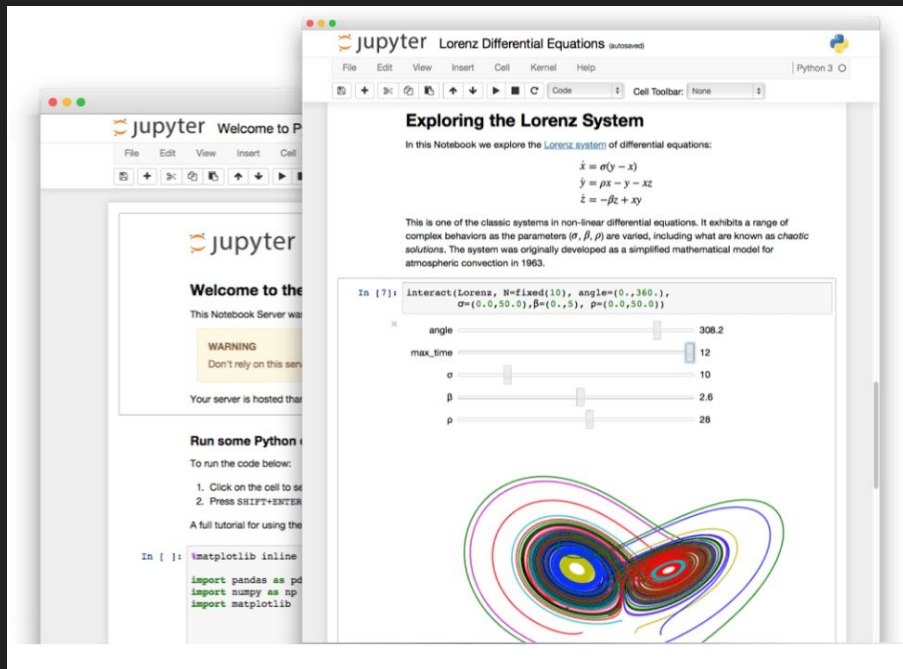
Jupyter notebook

"The Jupyter Notebook is an open-source web application that allows you to create and share documents that contain live code, equations, visualizations and narrative text."

Best way to learn is to try it!!

JupyterLab 2.0: Jupyter's
Next-Generation Notebook Interface

website: <https://jupyter.org/>



Organization of the course

website: <https://sites.google.com/view/paris-physics-master/home>

12 lectures of 5 hours

- Classes are split in ~ 2h theory and ~ 3h practical work
- Practical work is done on a computer using jupyter notebooks
- Mixture of in-person and virtual classes
- Classwork is done individually or in pairs (preferable)
 - work **must** be submitted before the following class (i.e Thursday before midnight)

Course note is based on

- continuous assessment 25% x 2 (50%)
- classwork notebooks (50%)

**Classes are in different rooms and computers are unreliable...
bring a USB key to make a backup (or learn to use GIT)**

Course content

Lecture 1 → Introduction and python language - Andrea

Lecture 2 → Roots finding, Interpolation - Andrea

Lecture 3 → Extrapolation and linear regression - Andrea

Lecture 4 → Integration - Andrea

Lecture 5 → Minimization / Optimization 1D - Sebastien

Lecture 6 → Multi-dimensional minimization / optimization - Sebastien

Lecture 7 → Ordinary Differential Equations - part I - Sebastien

Lecture 8 → Ordinary Differential Equations - part II - Sebastien

Lecture 9 → Partial Differential Equations - part I - Sebastien

Lecture 10 → Partial Differential Equations - part II - Sebastien + [CC 1](#)

Lecture 11 → PIC methods and Montecarlo methods I - Andrea

Lecture 12 → PIC methods and Montecarlo methods II - Andrea + [CC 2](#)

Workflow for the practical part of our classes

1. Download the course material from:
<https://sites.google.com/view/paris-physics-master/home>
2. Open the .ipynb files in jupyter notebook
3. WRITE YOUR NAME IN THE FIRST LINE OF THE NOTEBOOK
4. Do some great work and save
5. Upload the .ipynb file to the same website, where there is a dedicated link to upload

The note for the practical work is global, i.e. you will not get a note for each report.

password: PPM2020!

A little warning...

On the internet (or in this class) you can find
almost all the solutions to *almost* all the problems.

Using other people's work as if it was your own and without proper
acknowledgement is wrong!

Plagiarism is never acceptable and it will land you into serious troubles.

If you use other people's work/ideas just say it. There is nothing wrong with that.
Yet, you are expected to at least try first.

There is a fine line between copying a "snippet" of code and an entire program...if
in doubt, just ask us

The resources of computational physics

Hardware: personal computer

- single CPU multi core
- 16 GB of memory
- TB of disk



Hardware: workstation and servers

- 2 to 4 CPUs many cores + GPU
- hundreds of GB of memory
- x 10 TB of disk



Hardware: small local clusters

- x 10s of CPUs, x10s cores + few GPU
- hundreds of GB of memory
- x 10 TB of disk



Hardware: large clusters

- x 100s of CPUs, few 1000s cores + many GPU
- 1000s of GB of memory
- x 100 TB of disk



Hardware: High-performance super-computers

- x10000s to few million cores
- hundreds xTB of memory
- many PB of disk

They all run Linux-based operating systems



Oak Ridge National Lab.

IBM
Summit



Barcelona supercomputing centre

Application Performance	200 PF
Number of Nodes	4,608
Node performance	42 TF
Memory per Node	512 GB DDR4 + 96 GB HBM2
NV memory per Node	1600 GB
Total System Memory	>10 PB DDR4 + HBM2 + Non-volatile
Processors	2 IBM POWER9™ 9,216 CPUs 6 NVIDIA Volta™ 27,648 GPUs
File System	250 PB, 2.5 TB/s, GPFS™
Power Consumption	13 MW
Interconnect	Mellanox EDR 100G InfiniBand
Operating System	Red Hat Enterprise Linux (RHEL) version 7.4

Software: languages, libraries and visualization

- languages:
 - C++, Fortran, Python, ...
- softwares:
 - MAPLE, MATHEMATICA, ...
- libraries:
 - many libraries exist to perform numerical calculations (e.g. linear algebra, fitting, etc.)
 - MPI for parallel computing, openMP, ...
 - In this course we will make extensive use of numpy
- visualization
 - plotting, 2D and 3D visualization
 - VisIt, Paraview, Mayavi,
- version control and other utilities to keep track of your work and collaborate
 - git, gitkraken,

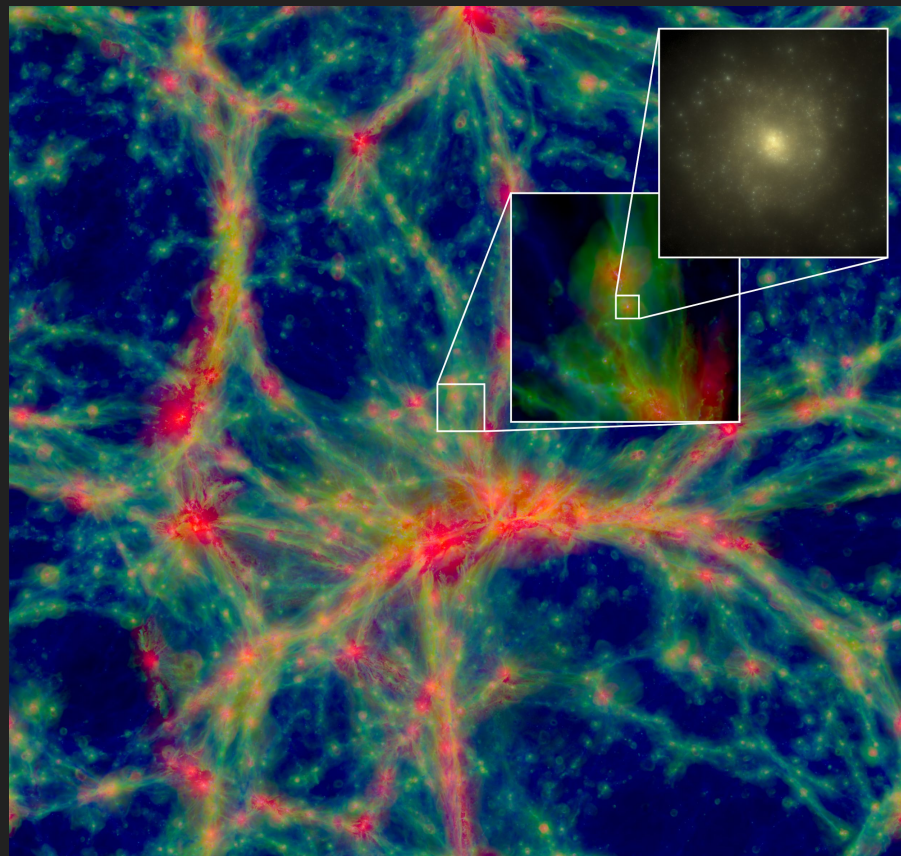
Examples of numerical simulations

Cosmological simulations

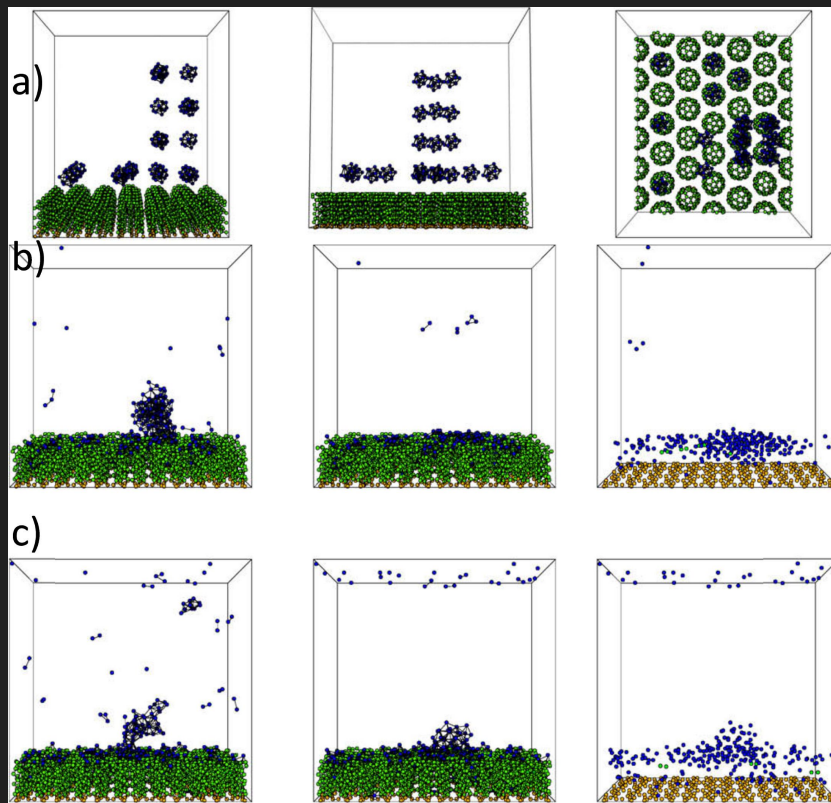
The EAGLE simulation is one of the largest cosmological hydrodynamical simulations ever, using nearly 7 billion particles to model the physics.

It took more than one and a half months of computer time on 4000 compute cores of the DiRAC-2 supercomputer in Durham.

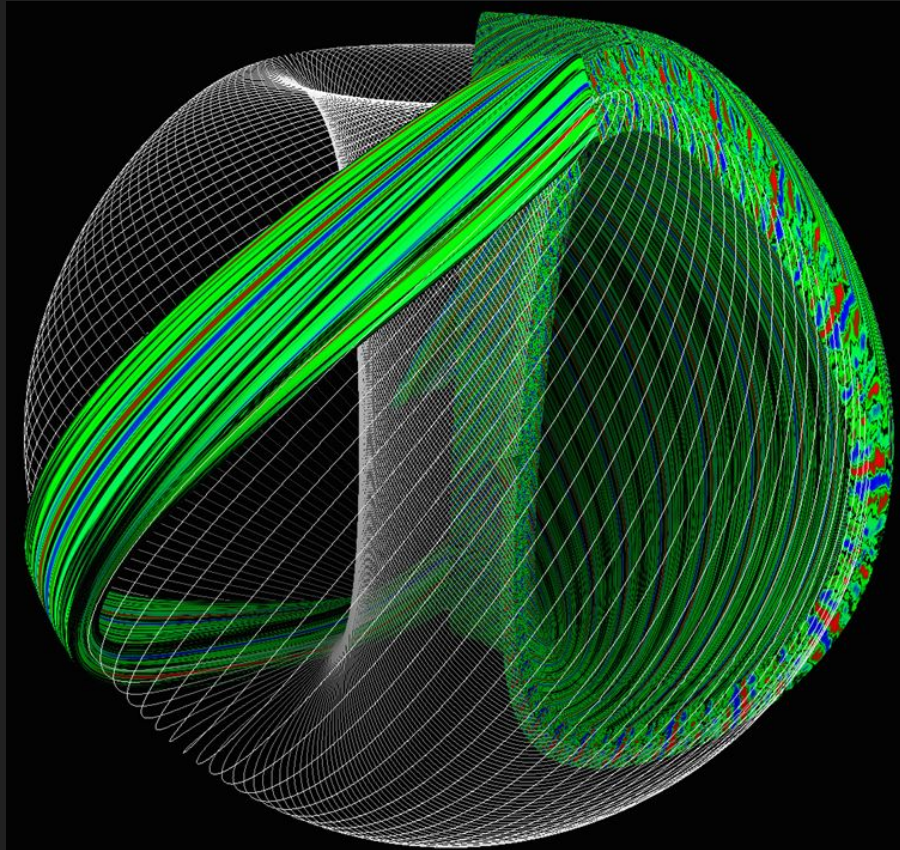
It was performed with a heavily modified version of the public GADGET-2 simulation code.



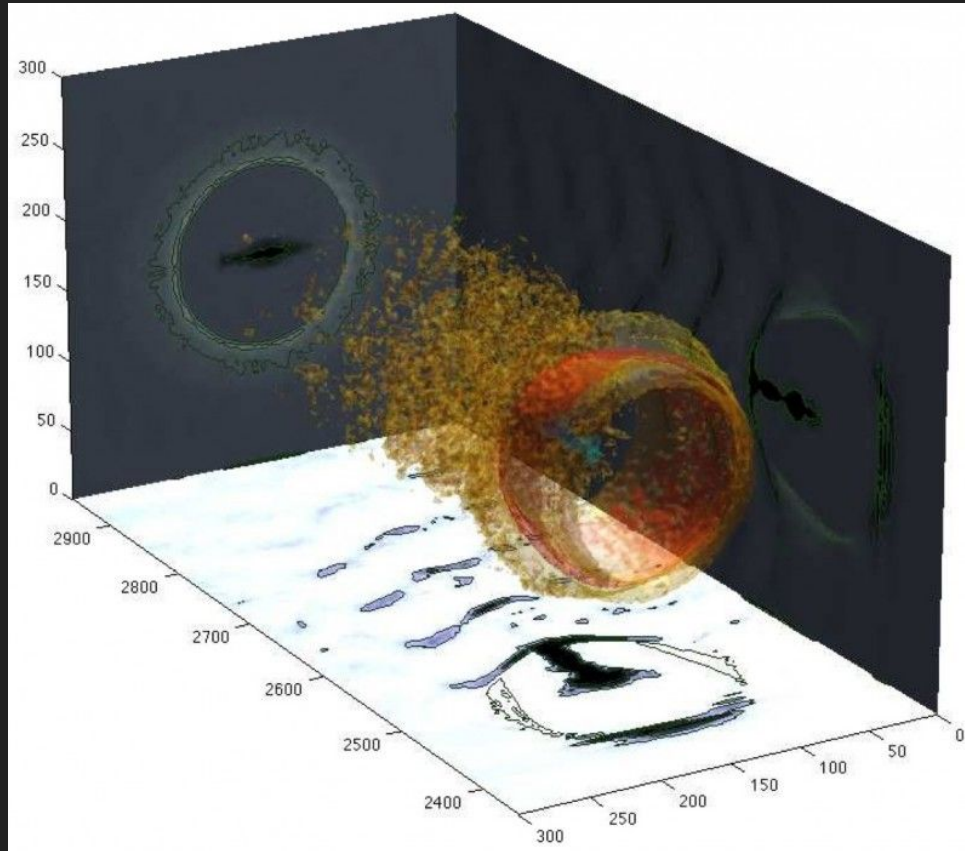
Vapor deposition of Ni and Au clusters on vertically aligned carbon nanotubes.



5-D gyrokinetic turbulence simulations in a tokamak

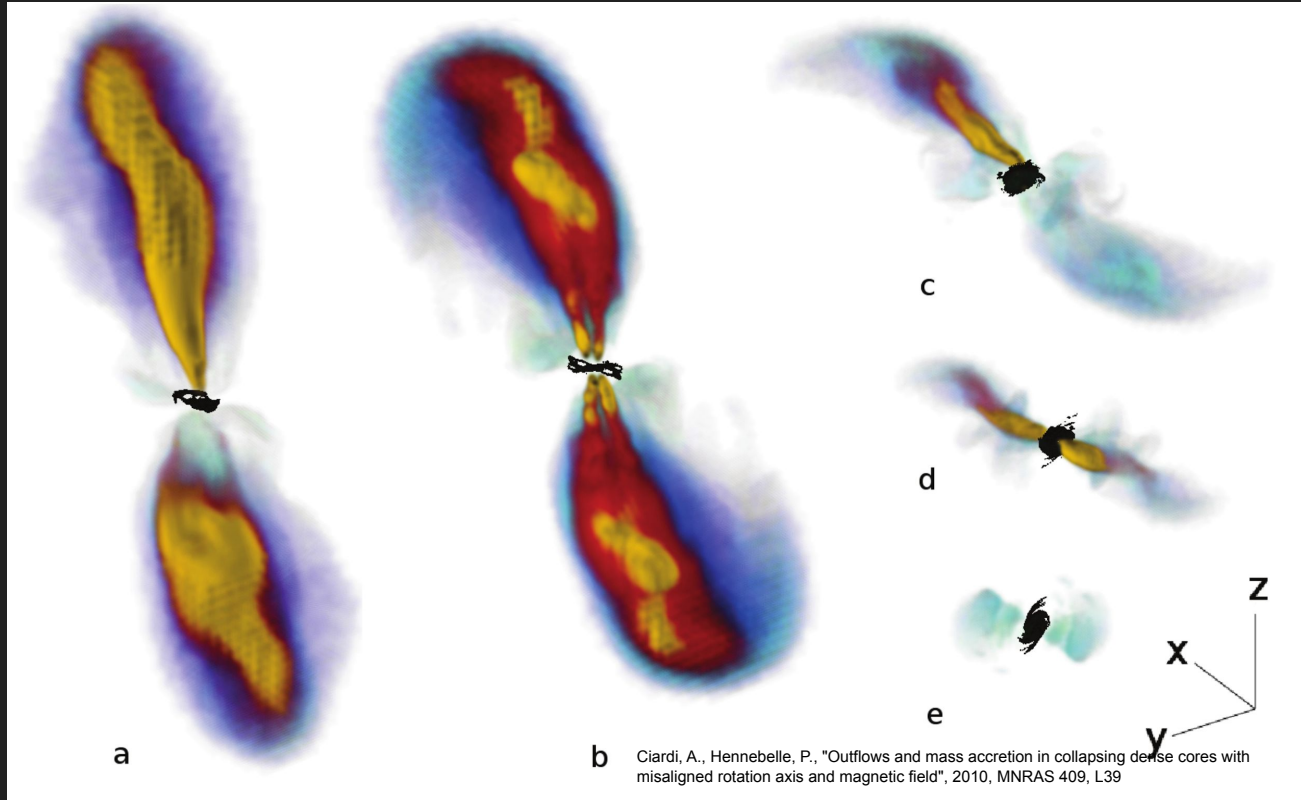


Ultra high intensity laser-plasma interaction



<https://cuos.engine.umich.edu/researchgroups/hfs/research/theory-and-computation/>
The bubble formed in the wake of the HERCULES pulse, according to a 3D Particle-In-Cell simulation using the OSIRIS 2.0 framework

Magnetohydrodynamic simulations of star formation



Simulations of high-energy particle collisions

