

# 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.

→ but Artificial Intelligence may be changing all of that...

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

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

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

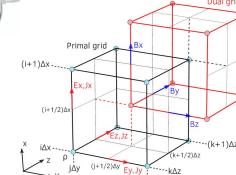
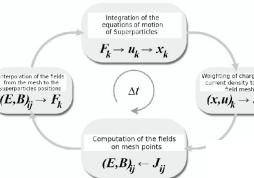
$$\frac{\partial E}{\partial t} + \nabla \cdot [(E + P^*)\mathbf{v} - \mathbf{B}(\mathbf{B} \cdot \mathbf{v})] = 0$$

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

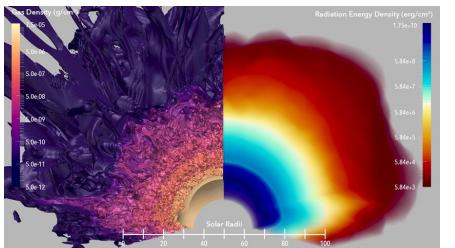
$$P^* = P + \frac{\mathbf{B} \cdot \mathbf{B}}{2}$$

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

# Algorithms and Numerical Methods



## Analysis and Interpretation



## Computer code

# Computational physics is multidisciplinary

Physics

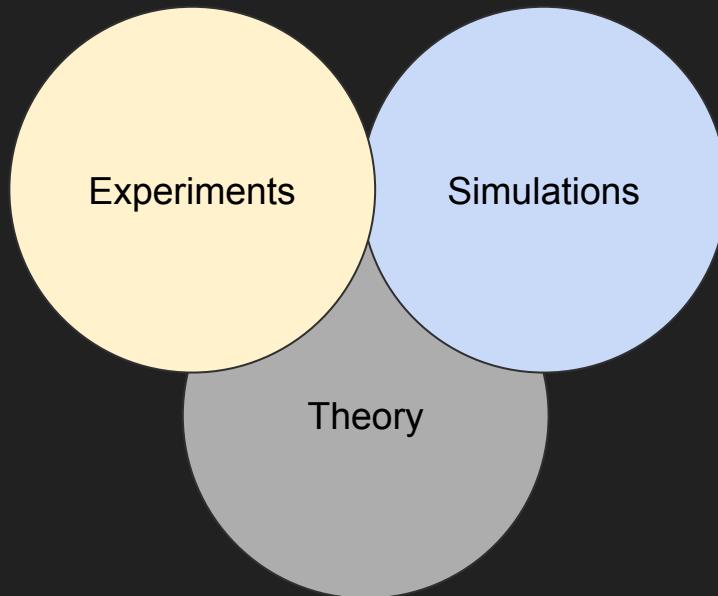
A large white rounded rectangle contains three smaller gray rounded rectangles. The top one is labeled 'Physics'. The bottom-left one is labeled 'Applied Mathematics'. The bottom-right one is labeled 'Computer Science'. All three are arranged horizontally.

```
graph TD; A[Physics] --- B[Applied Mathematics]; A --- C[Computer Science]
```

Applied  
Mathematics

Computer Science

# 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
- 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 eigenvalues
- ....

# Numerical methods and algorithms

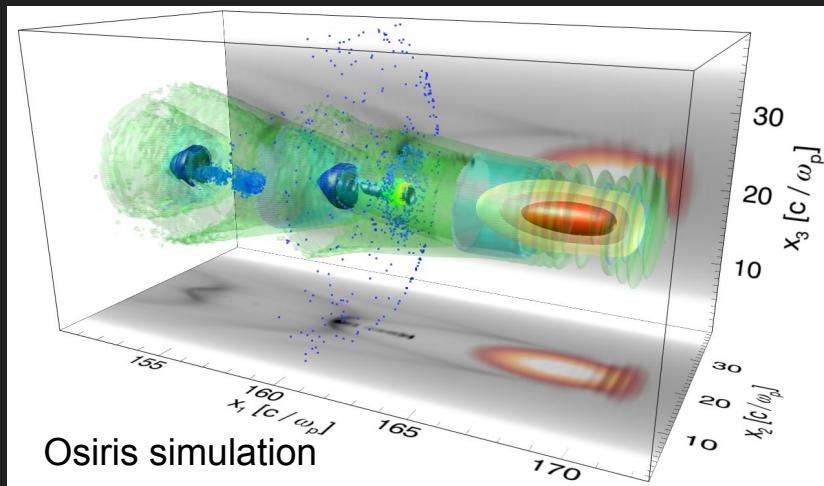
"Simple" building blocks are needed to construct complex numerical simulations

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  $6 \times 10^{23}$  particles
- **Finite representation of numbers and approximation of equations → errors**
  - A good scientist needs to know the limitations of computational results
- ....

This 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, using 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
  - *but can be too slow for large computations*
- 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 → jupyterLab

*"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/>

The screenshot shows the JupyterLab 2.0 interface. On the left is a sidebar with sections for Files, Running, Commands, Cell Tools, and Tabs. The Files section shows notebooks like Data.ipynb, Fasta.ipynb, Julia.ipynb, and Lorenz.ipynb. The Running section shows R.ipynb, iris.csv, lightning.json, and lorenz.py. The Commands section shows R.ipynb, iris.csv, lightning.json, and lorenz.py. The Cell Tools section shows sliders for sigma (10.0), beta (2.67), and rho (28.00). The Tabs section shows tabs for Lorenz.ipynb, Terminal 1, Console 1, Data.ipynb, README.md, and Python 3. The main area has a central notebook cell with the text: "In [4]: from lorenz import solve\_lorenz t, x\_t = solve\_lorenz(N=10)". To the right of the cell is an Output View showing a 3D plot of the Lorenz attractor, and below it is a code editor window titled "lorenz.py" containing the source code for generating the plot.

```
def solve_lorenz(N=10, max_time=4.0, sigma=10.0, beta=8./3, rho=28.0):
    """Plot a solution to the Lorenz differential equations."""
    fig = plt.figure()
    ax = fig.add_axes([0, 0, 1, 1], projection='3d')
    ax.axis('off')

    # prepare the axes limits
    ax.set_xlim((-25, 25))
    ax.set ylim((-35, 35))
    ax.set_zlim(5, 55)

    def lorenz_deriv(x_y_z, t0, sigma=sigma, beta=beta, rho=rho):
        """Compute the time-derivative of a Lorenz system."""
        x, y, z = x_y_z
        return [sigma * (y - x), x * (rho - z) - y, x * y - beta * z]

    # Choose random starting points, uniformly distributed from -15 to 15
    np.random.seed(1)
    x0 = -15 + 30 * np.random.random((N, 3))
```

# Organization of the course

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

12 classes (two of them are dedicated to exams)

- Classes are a mix of theory and practical work
- Practical work is done on a computer using jupyter notebooks
- Classwork is carried out in pairs
  - work ***must*** be submitted before the following class (i.e Thursday before midnight)

Course evaluations is based on

- Written (on computer) mid-semester exam 40%
- Classwork/homework and participation in class 20%
- Oral presentation of a mini-project 40%

**Computers are unreliable...**

**bring a USB key to make a backup, upload to the cloud (or learn to use GIT), use your own computer**

# Course content

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

# 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 (jupyterLab)
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: 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: **#numerical2024!**

# 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 there is nothing wrong with that, but you must say it (cite it).

Yet, you are expected to at least try first and produce original work.

Using a LLM (chatGPT, Gemini, copilot,...) to help you coding is fine.

**But you will not have access to it in the exam...so do not rely on it too much**

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

# 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

- x 1000s to millions cores
  - hundreds of TB of memory
  - hundreds of PB of disk
- They all run Linux-based operating systems



Fugaku supercomputer (Japan)



Barcelona supercomputing centre

# 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, ....

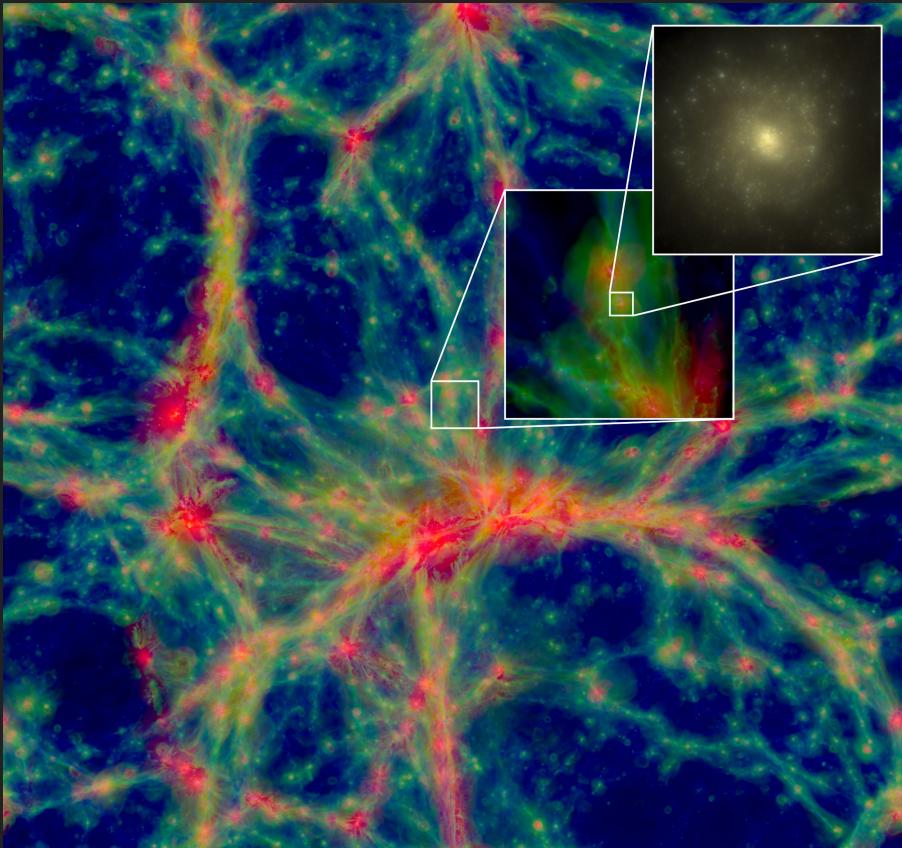
# Examples of numerical simulations

# Cosmological simulations

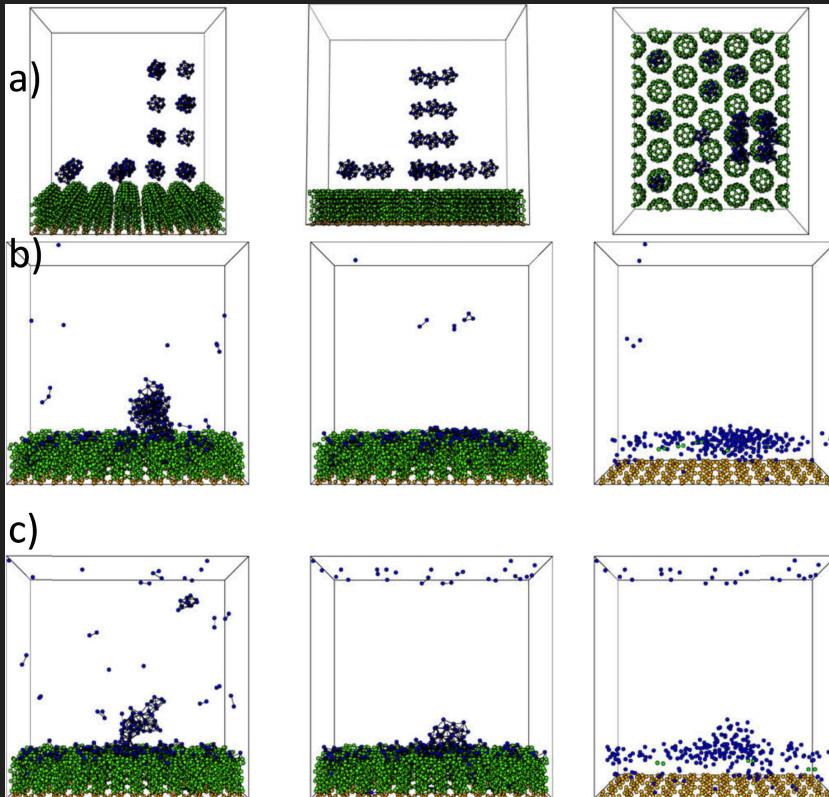
The EAGLE simulation is one of the largest cosmological hydrodynamical simulations ever, using nearly 7 billion particles to model the physics.

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.

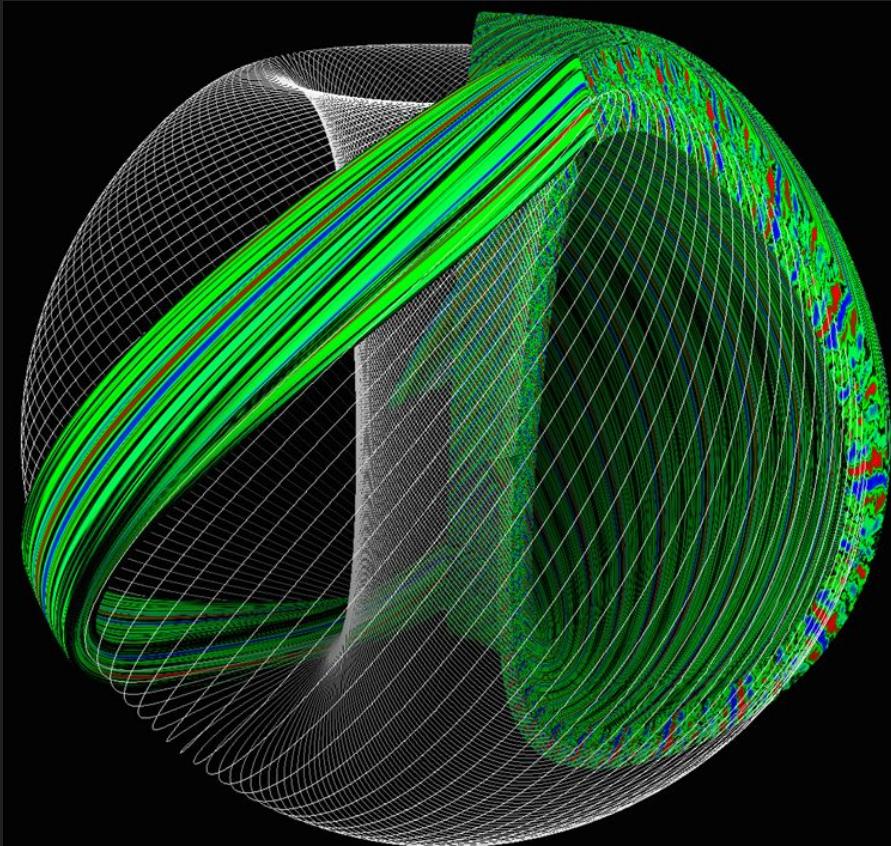


# Vapor deposition of Ni and Au clusters on vertically aligned carbon nanotubes.



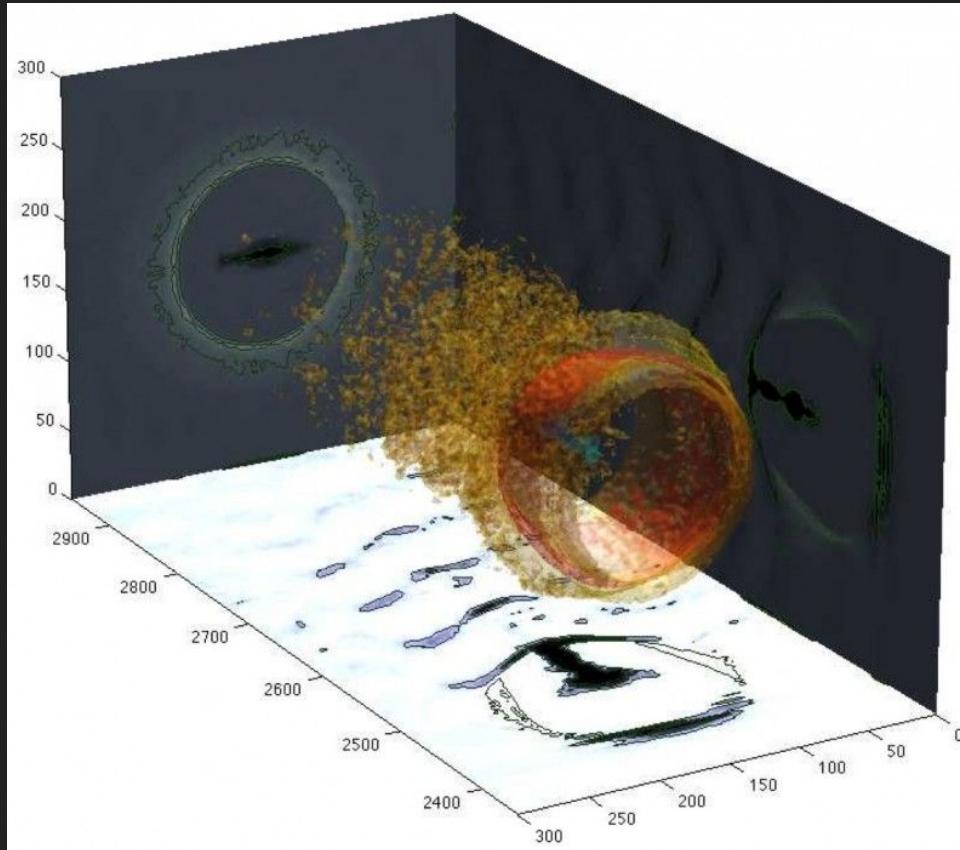
Molecular dynamics simulations of supported metal nanocatalyst formation  
by plasma sputtering  
Pascal Brault and Erik C.Neyts

# 5-D gyrokinetic turbulence simulations in a tokamak



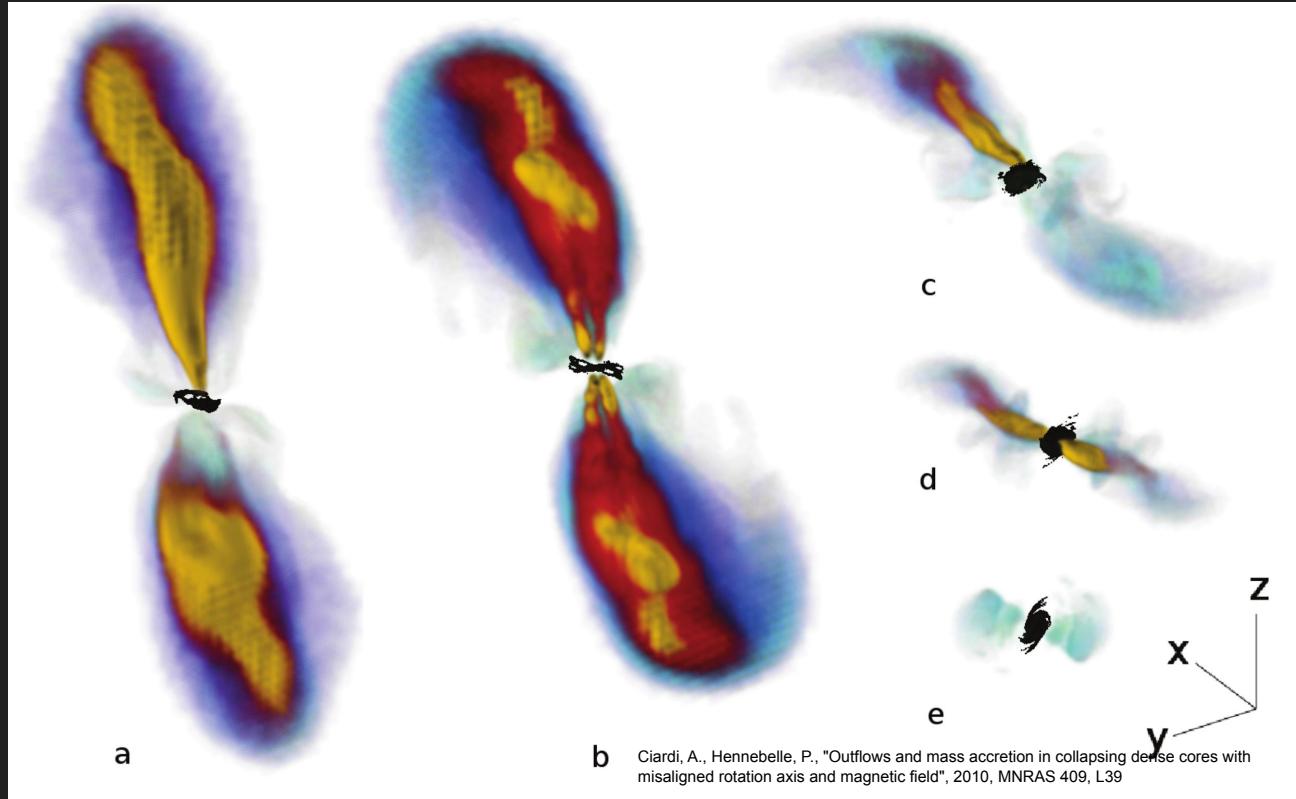
<https://w3.pppl.gov/~hammett/viz/viz.html>

# Ultra high intensity laser-plasma interaction

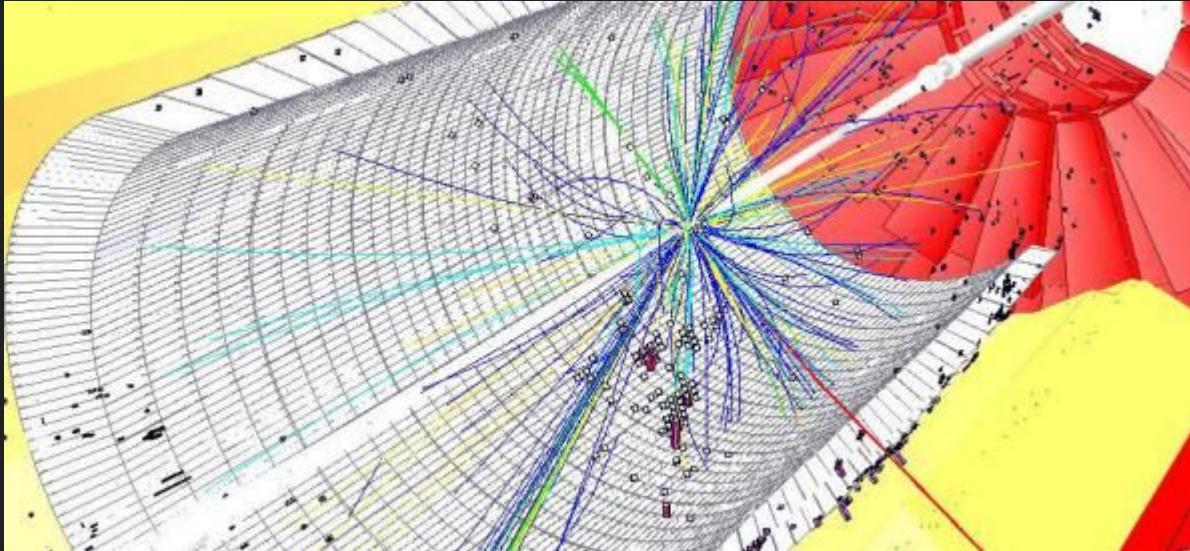


<https://cuos.engin.umich.edu/researchgroups/hfs/research/theory-and-computation/>  
The bubble formed in the wake of the HERCULES pulse, according to a 3D Particle-In-Cell simulation using the OSIRIS 2.0 framework

# Magnetohydrodynamic simulations of star formation



# Simulations of high-energy particle collisions



<http://lhcathome.web.cern.ch/projects/test4theory/high-energy-physics-simulations>