Computing Methods for Experimental Physics and Data Analysis

Introduction

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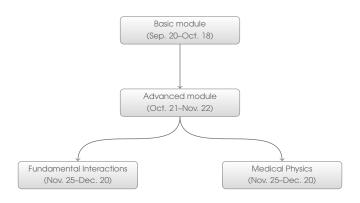
Goals and prerequisites

- - Automating repetitive tasks
 - > Python basics, standard library and scientific ecosystem
 - ▷ Collaborative code development and best practices
 - Algorithms and data structures

 - ▷ Specific tools for high-energy physics or medical physics
- ▷ This is not so much about Python or C++—it is about how to write code for effective data analysis
- ▷ Will I be a professional data scientist at the end of the semester?
 - No, but hopefully you'll be able to poke around and find the right tool for the job at hand
- - ▷ If you have ever programmed before that would be great!
- > This is our first round: we will adjust along the way



Basic structure of the course



- → Modularity and standard paths:

 - > 9 credits: basic + advanced + fundamental interactions



Basic module

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- Collaborative tools
- > Python basics
 - ▷ Coding conventions, structuring a package
 - ∨ Variables, native types, functions
 - ▷ The Python standard library
- Algorithms and data structures
 - ▷ Complexity and asymptotic running time
 - Python data structures and native algorithms
- - ▷ Classes, inheritance, composition
 - > Operator overload and emulation of Python builtin types
- ▷ The Python computing ecosystem
 - > numpy: arrays, functions, broadasting
 - Vectorization
 - Scipy: plotting and fitting
 - ▶ Pandas



Advanced module

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- Advanced code development
 - ▷ Unit testing, continuous integration, static analysis, documentation
- ▷ Advanced Python
 - ▷ Errors, exceptions, iterators and generators, decorators
 - ▷ Profiling and optimization
- - ▷ Computer architectures, memory, scaling laws, CPUs and GPUs
 - Parallel programming: concurrency and parallelism, threading in Python
- - Classification and regression: boosted decision trees and multilayer perceptrons
 - Deep learning: neural networks, the keras library
 - ▷ Supervised and unsupervised training, reinforcement learning
 - ▷ Tensorflow



Fundamental Interactions

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- - ▷ Coding style and organization, declaration of interfaces
 - Classes: constructors, virtual functions, private and public, abstract classes, inheritance
 - References, pointers, dynamic memory allocation, memory ownership, smart pointers
 - ▷ Templates, standard template library
- - Cuda and OpenCL
- > The ROOT data analysis framework
 - > ROOT toolkit
 - ▷ PyROOT, root-numpy, RDataFrame



Medical Physics

A. Retico

- Medical data processing and feature extraction (python/MATLAB)
 - > Tools for handling standard-format medical data (DICOM)
 - Data anonymization and visualization
 - ▷ Deriving features form images, image segmentation
 - Data quality control pipelines: outlier removal, dimensionality reduction
- ▷ Data analysis and classification (python/MATLAB)
 - Performance evaluations: figures of merit, cross-validation schemes, permutation test
 - Machine-learning and deep-learning tools for segmentation and classification
 - Data augmentation, transfer learning, retrieving localization information.



Logistics

Timetable and final exam

- ▷ Timetable: 6 hours a week

 - → Thursday, 08:30–11:30 (room O)
- - ▷ Tentative idea: 2 hour with slides and 2 hour at the computer
- ▷ "Lab": 2 hours on Thursday morning
 - → Attack a small project each week in a 2–3 hour session
 - ▷ (Bring your own laptop!)
- ▶ And, of course, the final exam
 - Development of a specific, reasonable-size software project (related to the topics covered in the course)
 - Two-page description of the project and source code made available in advance
 - ▷ Oral exam (starting from a presentation of the project)