

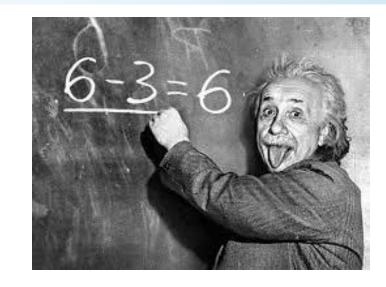


Collaborative data science: High Energy Physics and beyond

Marcin Sieprawski Head of Big Data Lab, Software Mind



- Not a High Energy Physics talk
 - * only a part of the picture of data analysis in HEP (last stage of analysis)



- * HEP as an excellent use case for ScienceMesh distributed Data Science environments
 - * Present the tools we provide to facilitate HEP research
 - * Why it is relevant in a wider perspective
 - * can be transferred to other sciences and business
 - * Get feedback





- Using particle accelerators to smash particles together at high speeds in the search for new particles
 - * Identifying events of interest
 - * Analysing data streams from detectors
- Name The Large Hadron Collider (LHC)
 - * analysing the myriad of particles
- * These experiments are run by collaborations of scientists from institutes all over the world



Image: Simulation of a particle shower resulting from a particle collision. © CERN



HEP: data challenges, collaborative data science

- * LHC produces unprecedented volumes of data
 - * Raw stream from detectors: 600TB per second, or 50 000 PB per day
 - * The raw data per event ~1MB, 600 million events per second
 - * Storing only fraction of this (total volume of all stored data: 350PB)
 - * Filtering of events/data need to be smart and fast!
 - * Large distributed infrastructure for transferring and large-scale processing
- * Constant innovation in tools and methods for analytics.
 - * Data streams from LHC increase with each upgrade
- Distributed teams of scientists from institutes all over the world
 a variety of storage systems and processing tools.
- *One of the challenges: providing tools in this distributed environment for effective collaboration in Data Science

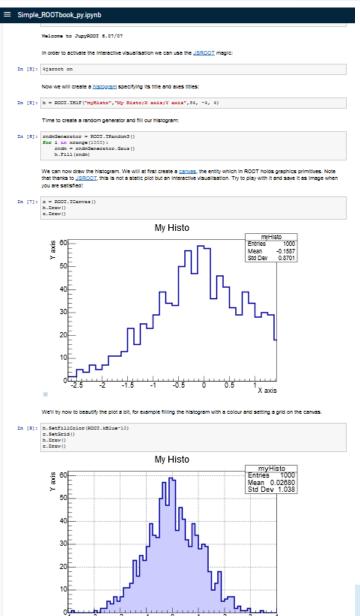


- *In main stream since 2012: Harvard Business Review
 - * "Data Scientist: The Sexiest Job of the 21st Century"
- In business: using data to increase competitive advantage
- * Data analytics approaches
 - * Data Warehouse, Business Intelligence, Data Mining ...
- * Origin of Data Science:
 - * using methods "always" used in science to analyse data
 - Most of those methods were invented for physics
 - * HEP pionieers in using many approaches (ag. Machine Learning)



Jupyter Notebook

- * A free, open-source, interactive web tool: a computational notebook
 - * combine software code, computational output, explanatory text and multimedia resources in a single document
 - * rapid uptake, an enthusiastic community of user-developers
 - Python / R / more
 - * 10 milion public Jupyter Notebooks on GitHub (December 2020)
 - * 2.5M in September 2018, 200k in 2015
- * "de facto" platform used by data scientists to build interactive applications and to tackle big data and AI problems
- * Replacing Business Intelligence tools





Jupyter Notebook: SWAN service @ CERN



Jupyter Notebook in HEP: SWAN galery

https://swan-gallery.web.cern.ch/



Gallery

Basic Examples

ROOT Primer

Accelerator Complex

Beam Dynamics

Machine Learning

Apache Spark

Outreach

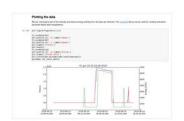
AWAKE

Accelerator Complex

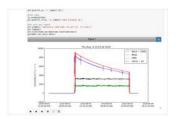
This gallery shows examples of machine studies relative to the CERN accelerators' complex.



LHC Page1



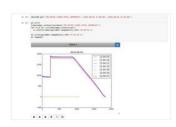
Experiments' Luminosities



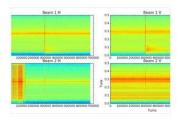
PyTimber Tutorial



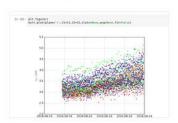
SPS Intensity



LHC BBQ Example



BSRT Example

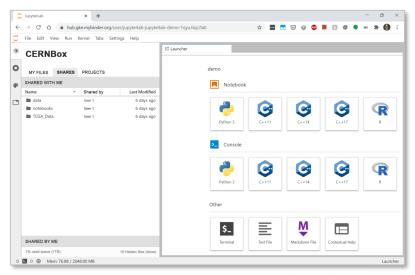






JupyterLab extension (Cs3Api4Lab)

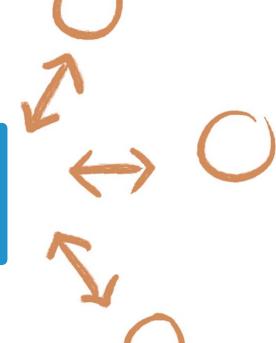
♣ Integration with ScienceMesh IOP (CS3 APIS)











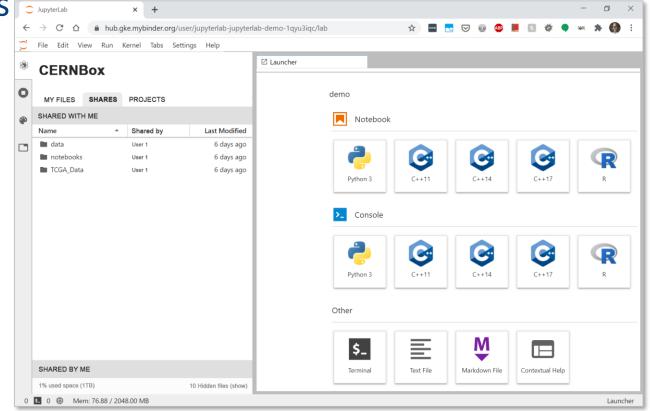






JupyterLab extension (Cs3Api4Lab): Frontend

- * Full client in Lab
- * File browser share functionalities
 - Shared by/with tab
 - Sharing buttons
 - * Entries in the context menus
 - * Pop-up windows: file information and sharing status
 - * Account info
- * File browsing





JupyterLab extension (Cs3Api4Lab): Backend

- * Replaces ContentsManager and Checkpoints
- * REST endpoints for integration with the frontend:
 - * API for content operations
 - * API for checkpoints operations (todo)
 - * API for share operations
- * Connecting IOP: gRPC (CS3 APIs)







Cs3Api4Lab (backend)

Jupyterlab







Distributed Data Science environments

Done (since CS3 conference)

- New implementation of file browser
- User information (look-up users, share with a user by name)
- Locking mechanism for concurent updating of notebooks

Current

- JupyterLab 3
- Concurrent updating of notebooks
- Unification share APIs (shares, OCM shares)

Next

- Stabilisation/tests
 - * Testing with multiple remote instances
 - OCM
- Mount the file system, to allow local access from the kernel

MVP

github.com/sciencemesh/cs3api4lab_check it out

- Accessing shared files and folders
- Sharing
- Info panel (sharing info)

*ready
**ready







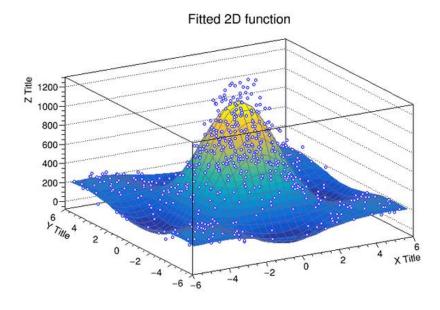








- * Data analytics play an increasing role in all types of research
- Distributed data science environments => all fields of study
- * A more effective collaboration between scientific institutions
- * Business: develop new products in all sectors
 - * Finance, IoT, SmartCities, energy and many others
- * Gartner Critical Capabilities for Data Science and Machine Learning Platforms
 - * (13 March 2021)
 - * By 2023, 30% of organizations will harness the collective intelligence of their analytics communities, outperforming competitors that rely solely on centralized analytics or self-service.
 - * By 2024, 70% of enterprises will use cloud and cloud-based AI infrastructure to operationalize AI, thereby significantly alleviating concerns about integration and upscaling.







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