

# From physics to insights: an overview of CERN computing pipelines

*Vincenzo Eduardo Padulano (CERN EP-SFT)*

Valencia, November 2025

Founded in 1954  
Biggest research centre for  
High Energy Physics (HEP)



Accelerating Science

A collaboration of many  
member states, including  
**Spain!**



A collaboration of many member states, including Spain!





# LHC: Large Hadron Collider

The largest scientific  
experiment to date

Circumference:  
**27 Km**

Depth:  
**-175 m ~ -50 m**

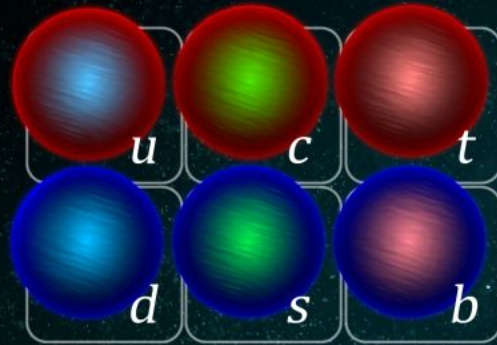
Working temperature:  
**-273.1 C°**

Raw physics data:  
**1000 TB/s**

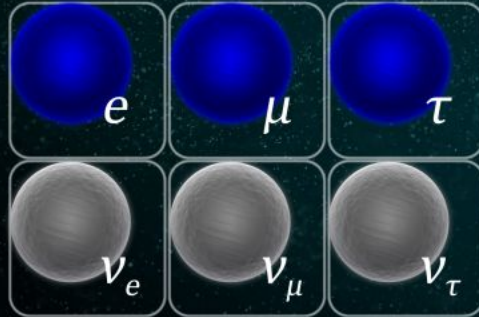




Research at CERN is focused on the Standard Model, defining how the building blocks of the universe interact



Quarks



Leptons



Higgs boson

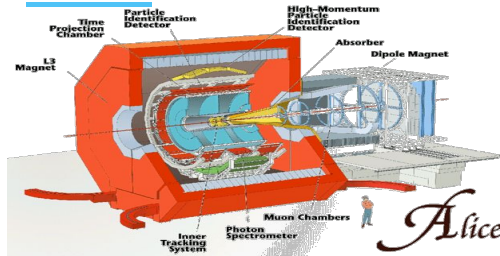


Forces

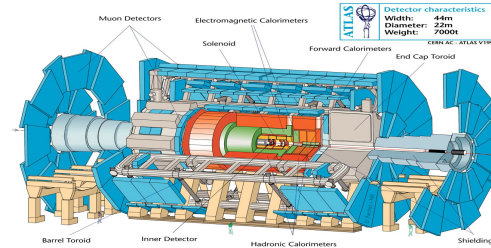


## Large particle detectors at CERN

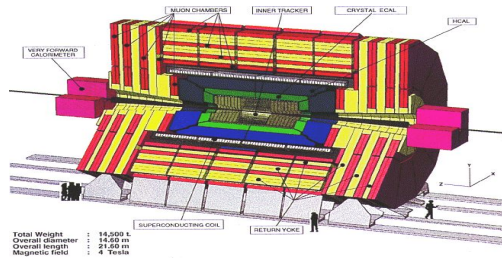
### ALICE



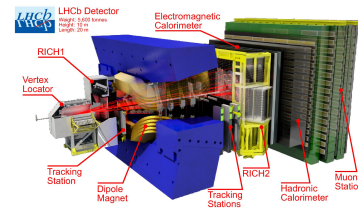
### ATLAS



### CMS



### LHCb







... but also large collaborations spread worldwide



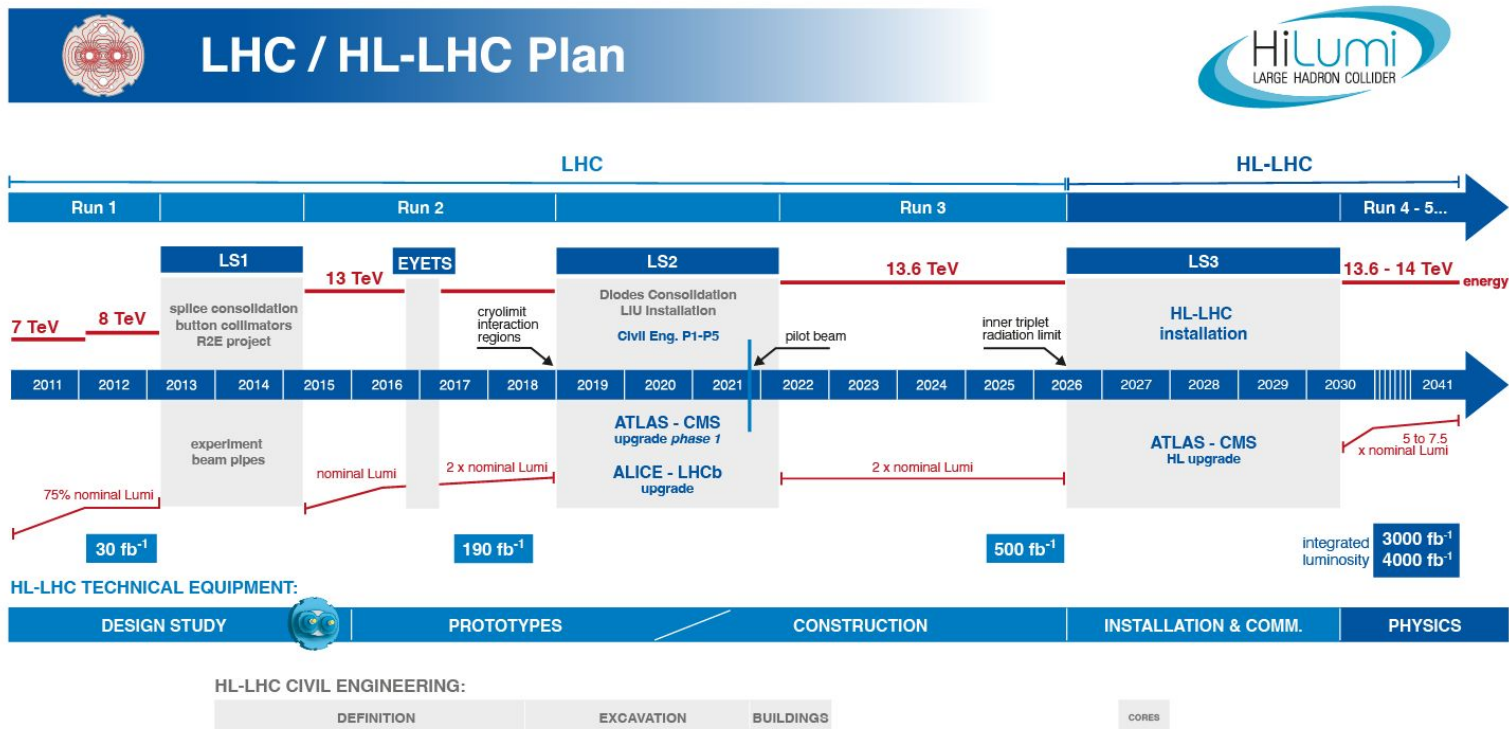
[Source](#)



[Source](#)



# The LHC project schedule

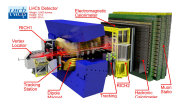
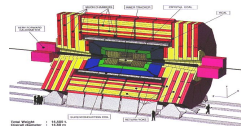
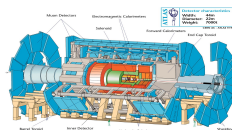
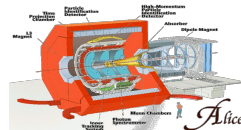






# Data lifecycle at the LHC

Data Collection  
1 PB/s



Trigger  
systems



Raw  
data



Simulated  
data

Reconstruction

RECO

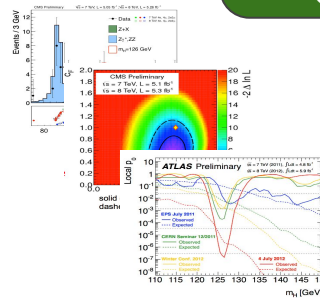
Processing,  
skimming

Analysis  
formats

Offline Processing  
several times a year  
100 PB/y

Data Analysis  
 $O(10)$  GB  $\rightarrow$  exploration  
 $O(10)$  TB  $\rightarrow$  full-scale

Event selection,  
statistical  
treatment...





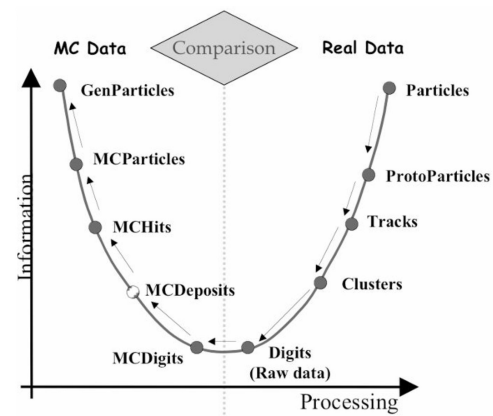
- ▶ Data acquisition: HEP data are statistically independent particle collisions (events)
- ▶ Analysis at experiments requires the highest possible number of collision events
  - We are looking for very, very rare events!
- ▶ One possible approach: more protons in the collisions
  - Thus, heavier computations
  - But! higher chance of uninteresting collisions





# Data lifecycle at the LHC

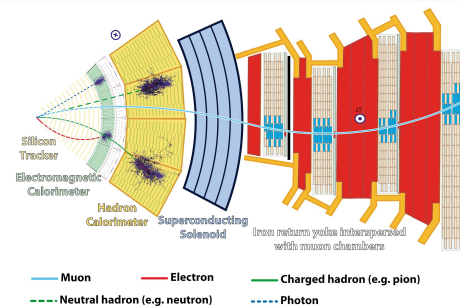
- ▶ Reconstruction (RECO): transform RAW data into understandable things
  - muons, electrons, photons...
- ▶ Further processing to create slimmer datasets
  - Usually called AOD, Analysis Datasets



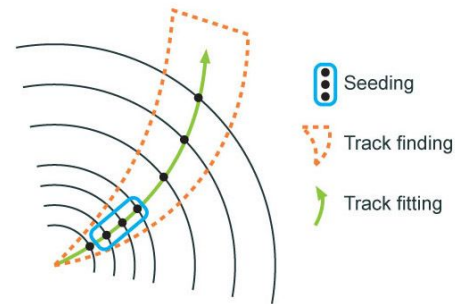


# From RAW to analysis

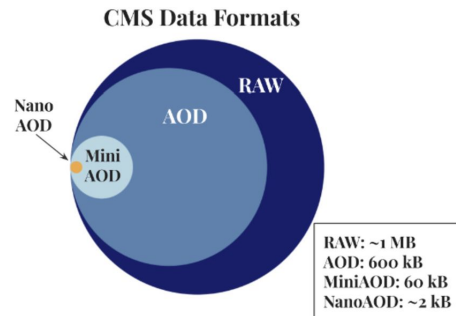
- ▶ Reconstruction is done with many different specialized software
  - Depends on experiment (e.g. [CMSSW](#))
- ▶ Has seen wide accelerator usage
  - [Example from CMS](#)
- ▶ AODs then see various skimming
  - experiment-dependent



[Source](#)



[Source](#)



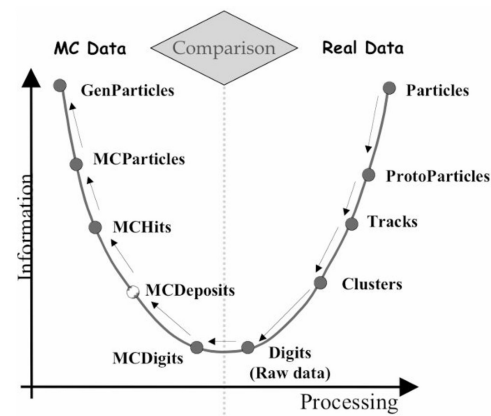
[Source](#)





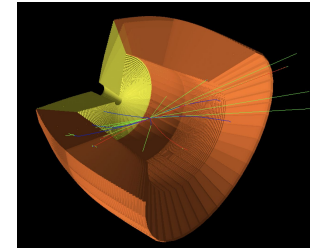
# Data lifecycle at the LHC

- ▶ Generation (GEN): Create physics events (through Monte Carlo generators)
  - A set of particles
- ▶ Simulation: simulate response of detector based on generated events
  - Compare real physics collision with simulation from theory

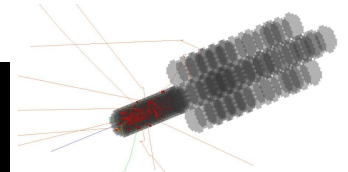




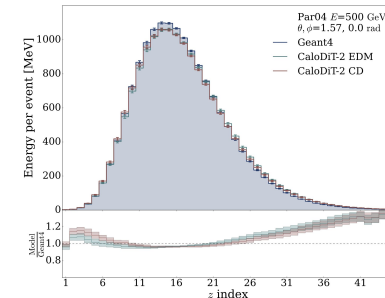
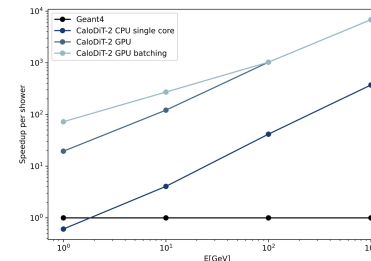
- ▶ There are various **event generator** software
  - [Pythia](#), [MadGraph](#), [SHERPA](#), ...
- ▶ [GEANT4](#): **state-of-the-art** software used for **full simulation** of particle transport through matter
- ▶ **Fast simulation**
  - Use a mix of ML techniques to run simulations
  - **Less precise** than full simulation with Geant4, **but faster**



[Source](#)



[Source](#)

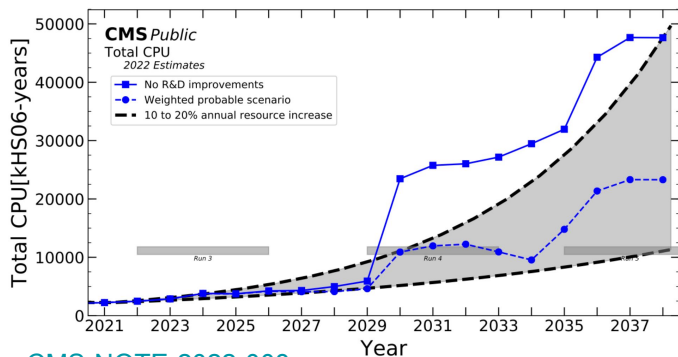


[Raikwar et al.](#)

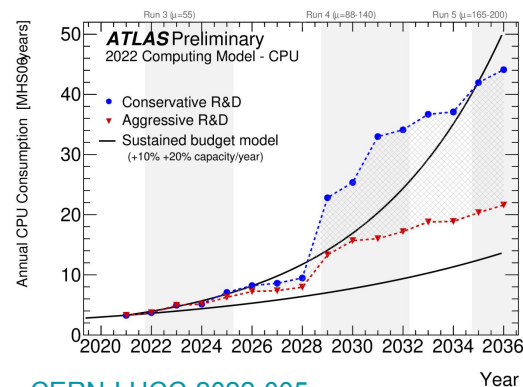


# Computing challenges

- ▶ More than **2 EB** physics data collected and stored so far
- ▶ That's only 10% of the total dataset, **90% with HL-LHC**
- ▶ **No software R&D** means drastic **reduction** of future physics programme



[CMS-NOTE-2022-008](#)



[CERN-LHCC-2022-005](#)





# A computing dilemma

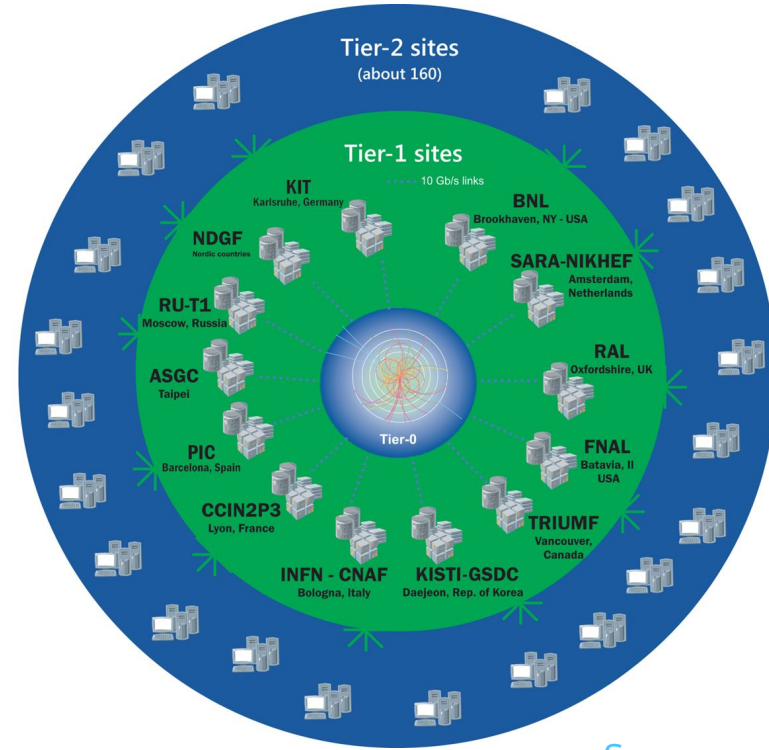
- ▶ Very large quantities of data collected
- ▶ Different experiments with different schemas and processing workflows
- ▶ Collaborations spread among research groups worldwide

**How?**



# The Worldwide LHC Computing Grid

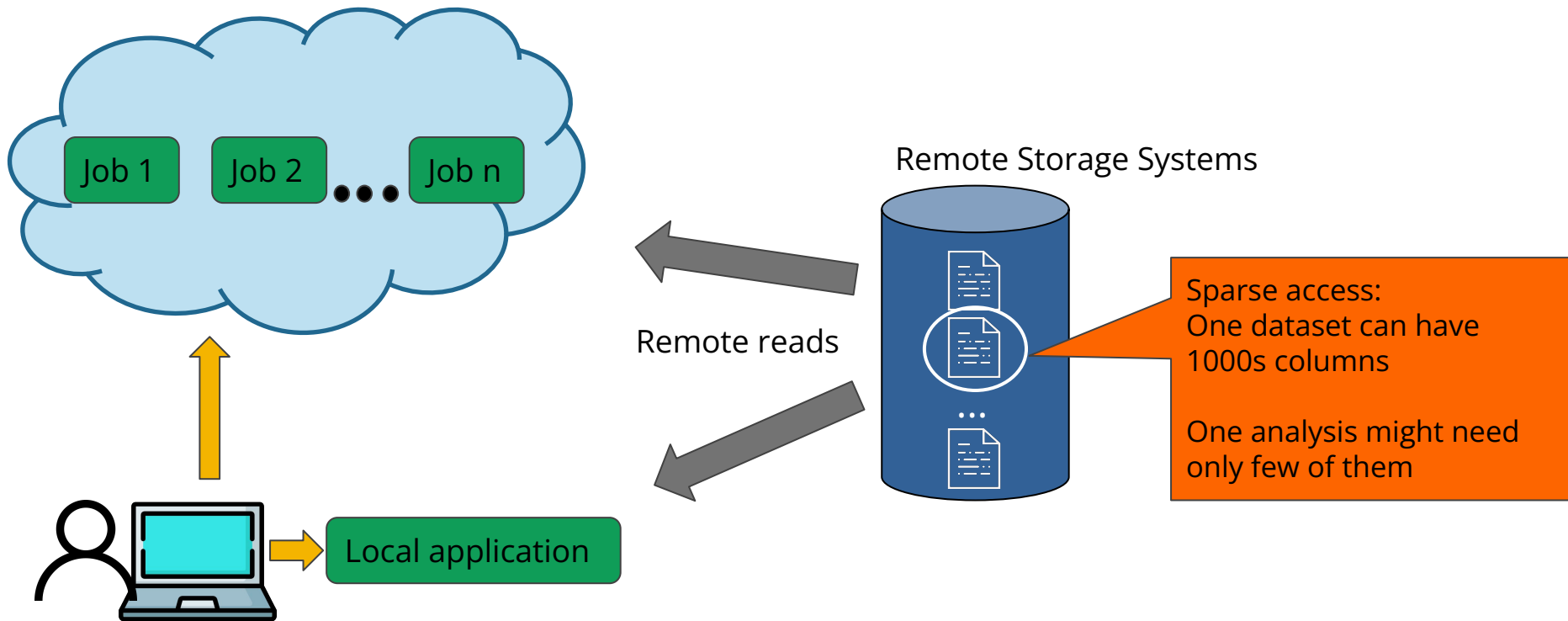
- ▶ Tier layers (Tier 0 at CERN)
- ▶ From large computing facilities to small university clusters
- ▶ All connected together.
- ▶ Physicists tap into the grid for practically all the steps of the lifecycle



[Source](#)



# Data access in HEP data analysis



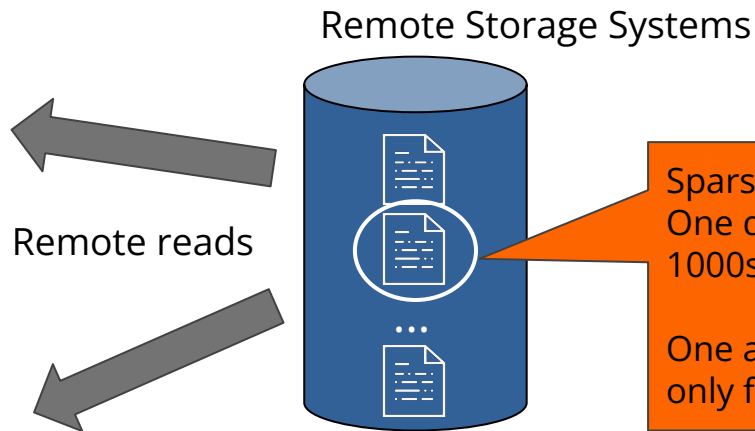
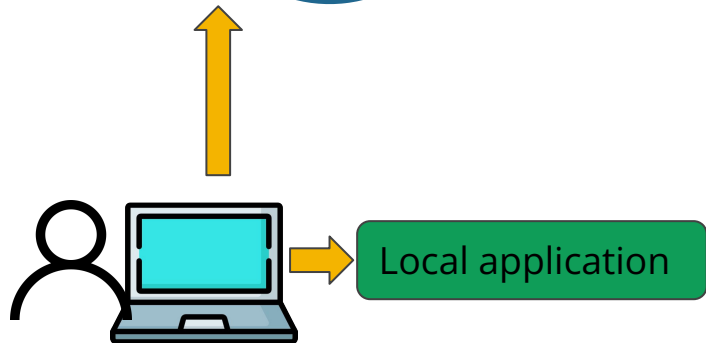
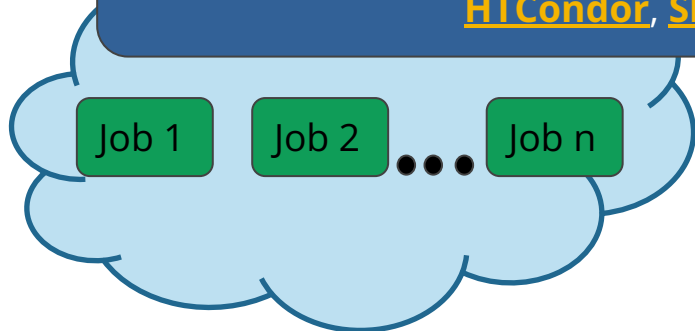




# Data access in HEP data analysis

Many ways to access resources and schedule,  
but definitely batch systems are the most widely used

HTCondor, Slurm

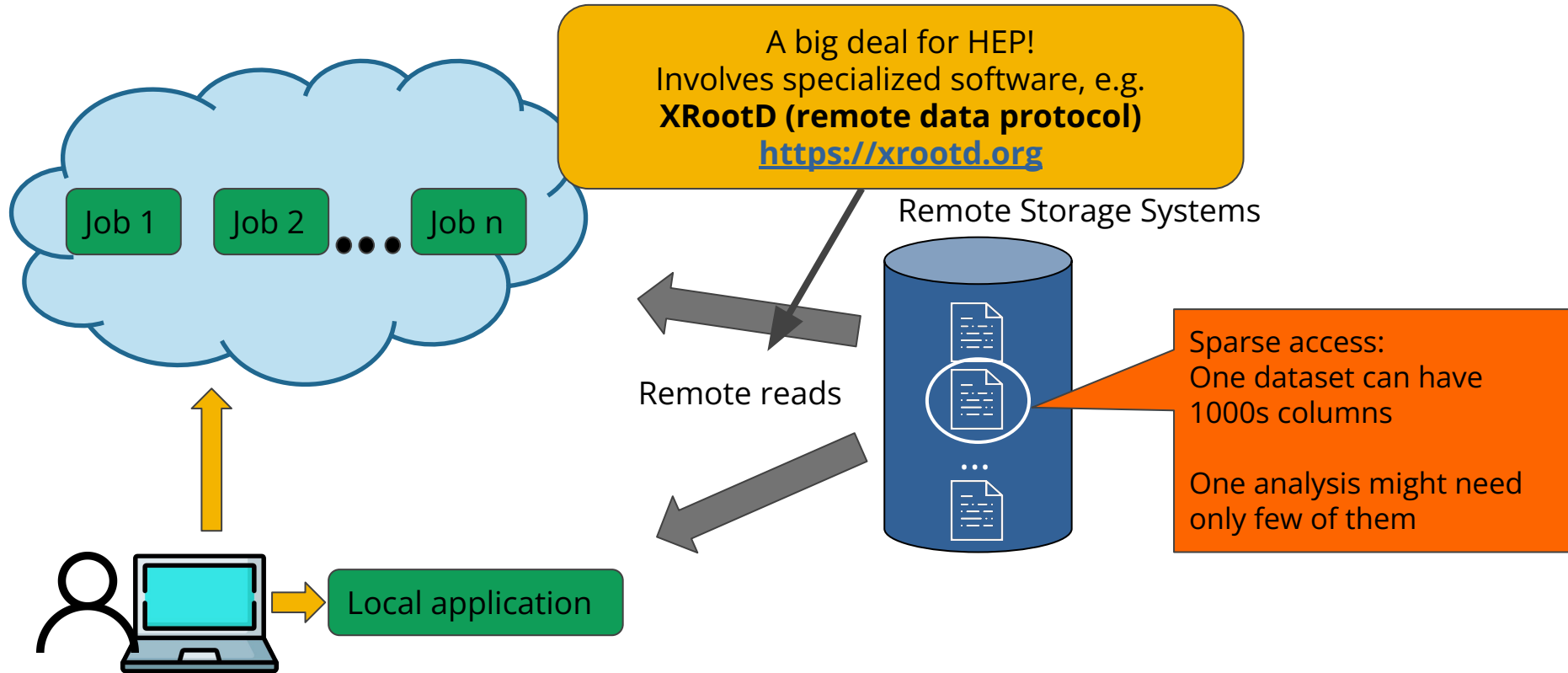


Sparse access:  
One dataset can have  
1000s columns

One analysis might need  
only few of them

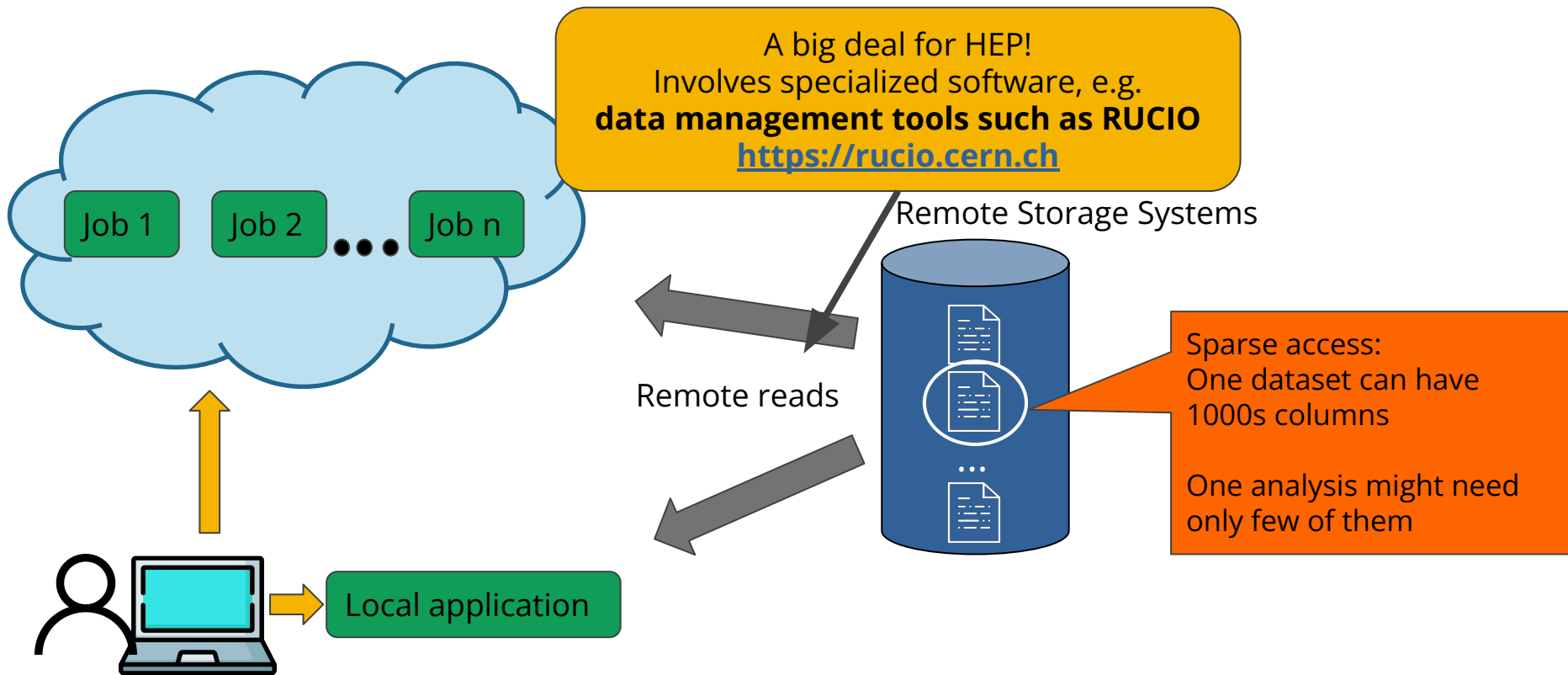


# Data access in HEP data analysis





# Data access in HEP data analysis

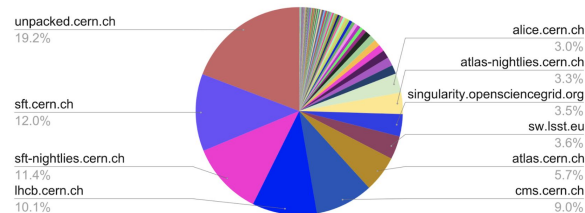
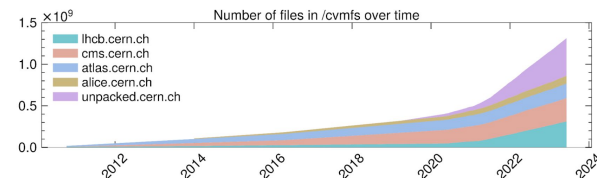






# And what about software access?

- ▶ CVMFS: The **standard** HEP **software distribution** tool
- ▶ A shared filesystem with lazy on-demand access to files (libraries, scripts etc.)
- ▶ Aggressive caching for maximum performance
- ▶ Used by institutes worldwide



<https://cernvm.cern.ch/fs/>

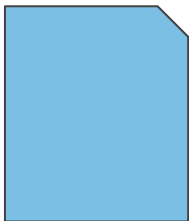


# HEP data processing



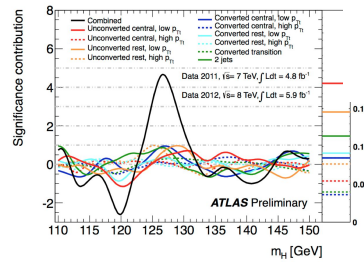
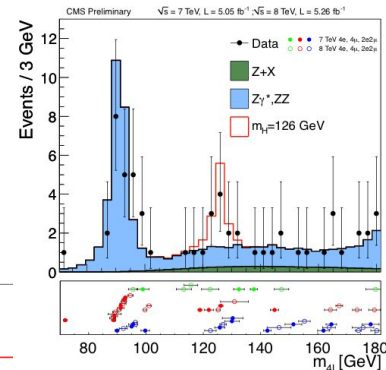
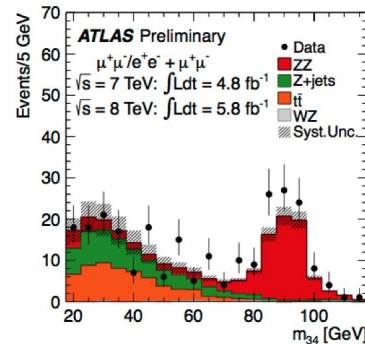
# High Energy Physics analysis

Gather accelerator data  
already stored in analysis  
formats



Perform analysis:

- Select interesting events
- Transform physics measurements with complex functions
- Aggregate into relevant statistics







# Various actors involved

## Analysts:

- The end users
- (Mostly) physicists affiliated with some CERN experiment
- Interested in minimizing time between starting the application and seeing the final results
  - a.k.a “time to plot”

## Framework developers:

- Develop software for the analysts
  - experiment-specific ([CMSSW](#), [ATHENA](#))
  - generic ([ROOT](#), [GEANT4](#))
- Striving for performance, user friendliness and HEP feature needs

## Site admins:

- Responsible for computing clusters and infrastructures
- Worry about the software stack, resilience, hardware scalability



# ROOT, a protagonist in HEP computing

- ▶ Storage, processing, analysis and visualisation of scientific data
- ▶ Widely adopted in **High Energy Physics** and in other scientific and industrial fields
  - **2 EB** data stored in ROOT data format
  - Fits and parameters estimations for discoveries (e.g. the Higgs)
  - Thousands of ROOT plots in scientific publications



[root.cern](https://root.cern)



# ROOT: An Open International Collaboration



## Open-source and Open-development

- ▶ [On GitHub](#), LGPL 2.1
- ▶ PR-based model with a public review process
- ▶ Already 89 unique contributors in 2025!

## Open-planning

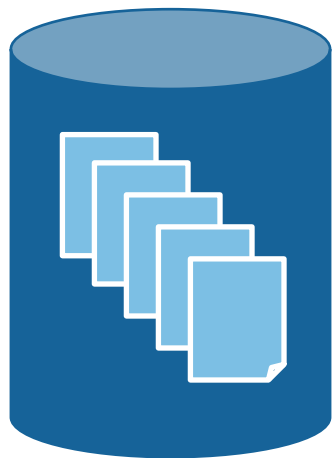
- ▶ The [Program of Work \(PoW\)](#) can be influenced with active engagement and contributions!
- ▶ Quarterly **public** reports to check progress according to experiments' inputs → [link to Q3 2025](#)

## Core ROOT team





# The HEP Dataset



dataset stored in one  
or multiple files



px	py	pz	e

Column:  
**physics entity**



Row:  
**collision event**



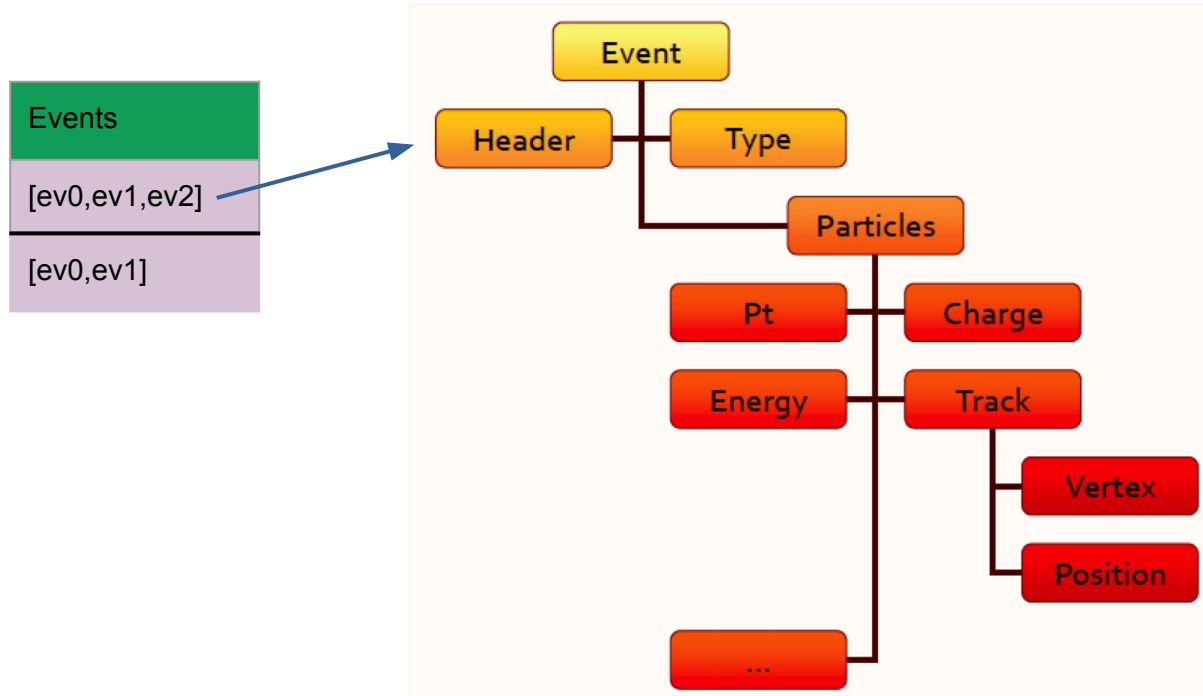
Can be nested  
structures







A relational DB, arbitrarily **nested**, with **varying sizes** per each DB entry  
All of this **in each file**





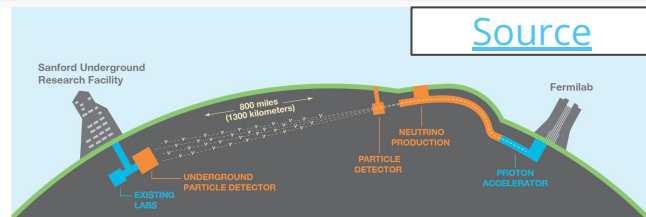
# The HEP data format, today...

**TTree** is the data **format** for **HEP data** used in production by **all** LHC **experiments**:

- ▶ Columnar
- ▶ Extremely flexible
  - Write simple types, collections, arbitrary user-defined types
- ▶ Scalable
  - More than **2 EB** written so far



- ▶ Next-Generation Collider Experiments: **HL-LHC, DUNE**
  - **10-30** times the computing requirements
  - Single events in the **multi-gigabyte** range for DUNE, heavy-ion experiments
  - Real **analysis challenge** depends on several factors: number of events, analysis complexity, number of reruns, etc.



- ▶ Full exploitation of **modern storage hardware**
  - Ultra fast networks and **SSDs**: 10GB/s per device reachable (HDD: 250MB/s)
  - Flash storage is **inherently parallel** → asynchronous, parallel I/O key
  - Heterogeneous computing hardware → **GPU** should be able to load data directly from **SSD**, e.g. to feed ML pipeline
  - Distributed storage systems move from **POSIX** to **object stores**

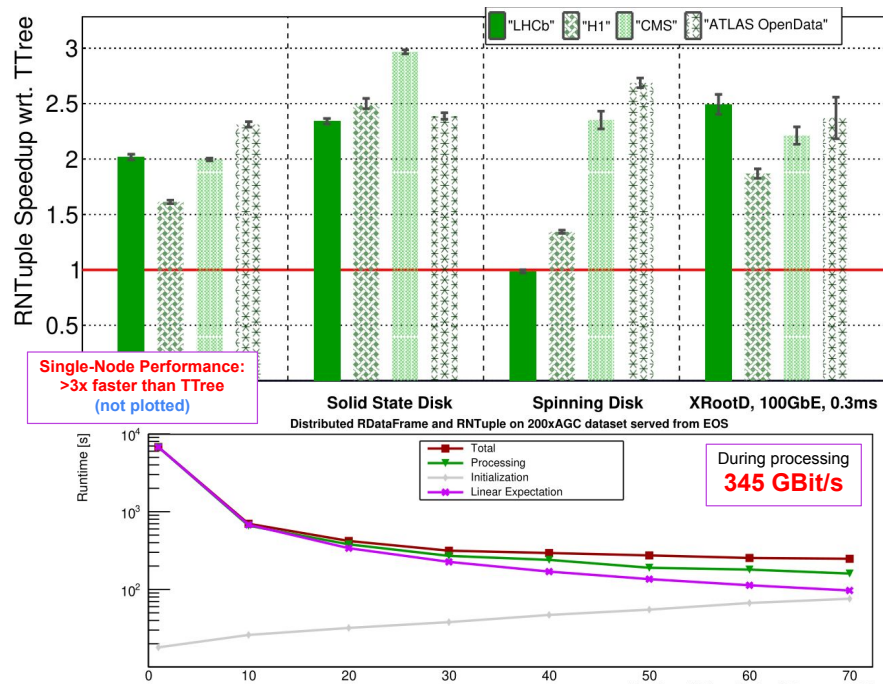




## RNTuple

Redesigned **I/O** subsystem, based on **25+ years** of **TTree** experience

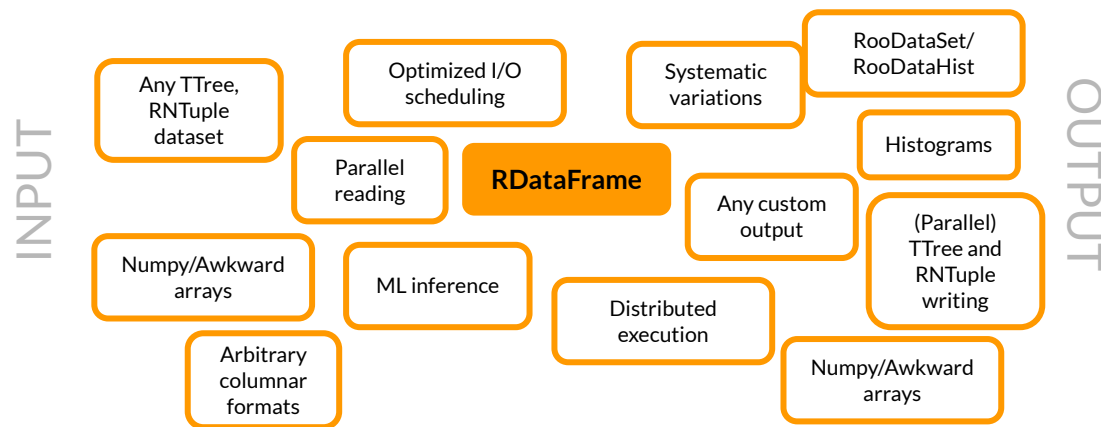
- ▶ **Less** disk and CPU **usage**
- ▶ **Efficient** support of **modern hardware**
- ▶ Transparent **file-less** storage
- ▶ **Covering** all of today's **TTree** use cases
- ▶ **Binary format** defined in a [dedicated specification](#)





# RDataFrame: ROOT's declarative interface for analysis

- ▶ ROOT's high level **interface** for analysis



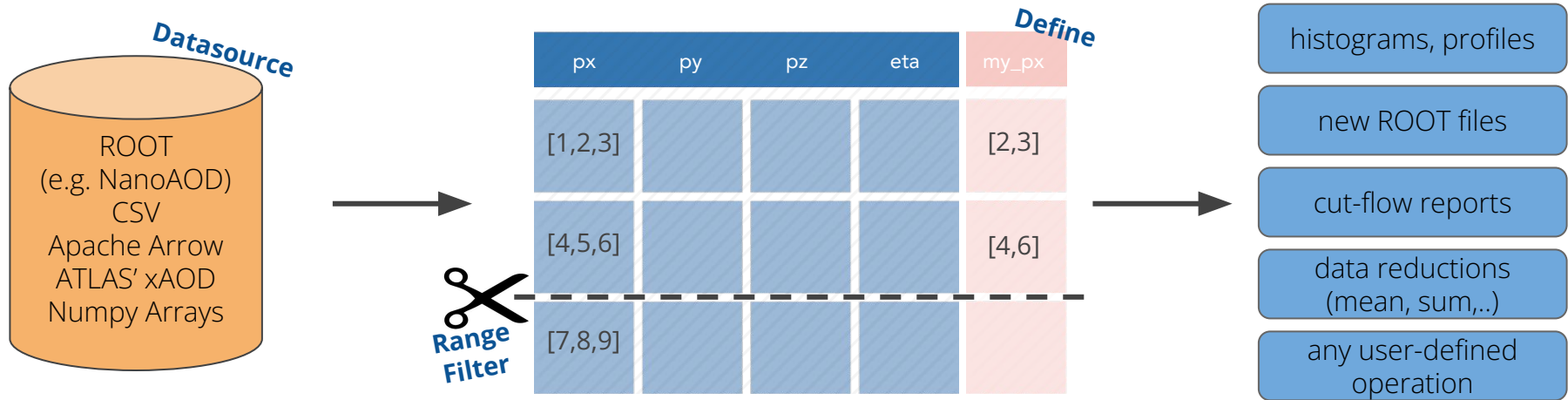
- ▶ Supports native parallel execution
- ▶ multi-threaded and distributed
- ▶ Exploit CMS compute, today







# RDataFrame: ROOT's declarative interface for analysis





# A small RDataFrame analysis

```
import ROOT
df = ROOT.RDataFrame('Events', 'Run2012BC_DoubleMuParked_Muons.root')

df_mass = df.Filter('nMuon == 2')\
             .Filter('Muon_charge[0] != Muon_charge[1]')\
             .Define('Dimuon_mass', 'InvariantMass(pt, eta, phi, mass)')

h = df_mass.Histo1D(
    ('Dimuon_mass', 'Dimuon mass;m_{#mu#mu} (GeV);N_{Events}', 30000, 0.25, 300), 'Dimuon_mass')

h.Draw()
```



# A small RDataFrame analysis

```
import ROOT
df = ROOT.RDataFrame('Events', 'Run2012BC_DoubleMuParked_Muons.root')

df = df.Filter('pt > 15 && eta < 1.4 && muon_charge[0] != Muon_charge[1]')\
      .Filter('InvariantMass(pt, eta, phi, mass)')

h = df.Histo1D('Dimuon_mass', 'Dimuon mass;m_{#mu#mu} (GeV);N_{Events}', 30000, 0.25, 300), 'Dimuon_mass')

h.Draw()
```

Open the dataset



# A small RDataFrame analysis

```
import ROOT
df = ROOT.RDataFrame('Events', 'Run2012BC_DoubleMuParked_Muons.root')

df_mass = df.Filter('nMuon == 2')\
            .Filter('Muon_charge[0] != Muon_charge[1]')\
            .Define('Dimuon_mass', 'InvariantMass(pt, eta, phi, mass)')

h = df_mass.Histo1D('Dimuon_mass', 30000, 0.25, 300), 'Dimuon_mass')
h.Draw()
```

## Analysis steps:

1. Select events with exactly two muons
2. Select muons with opposite charge
3. Compute invariant mass



# A small RDataFrame analysis

```
import ROOT
df = ROOT.RDataFrame('Events', 'Run2012BC_DoubleMuParked_Muons.root')

df_mass = df.Filter('nMuon == 2')\
             .Filter('Muon_charge[0] != Muon_charge[1]')\
             .Define('Dimuon_mass', 'InvariantMass(pt, eta, phi, mass)')

h = df_mass.Histo1D(
    ('Dimuon_mass', 'Dimuon mass;m_{#mu#mu} (GeV);', 'Dimuon mass; m_{\mu\mu} (GeV)'))
h.Draw()
```

Aggregate result into histogram





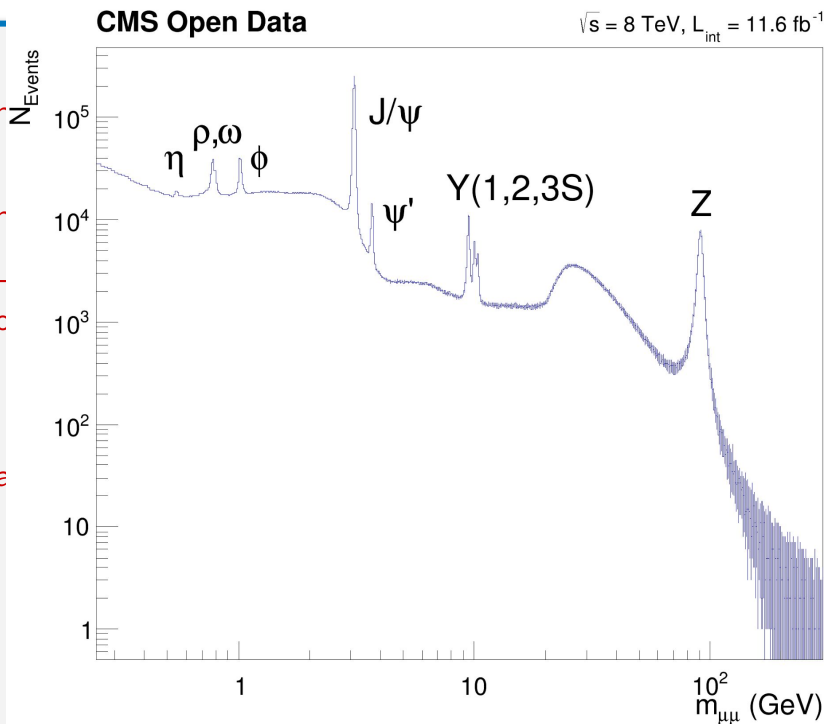
# A small RDataFrame analysis

```
import ROOT
df = ROOT.RDataFrame('Events')

df_mass = df.Filter('nMuon > 0')
df_mass.Filter('Muon_pt > 10')
df_mass.Define('Dimuon_mass', 'sqrt((Muon_pt1 - Muon_pt2)**2 + (Muon_eta1 - Muon_eta2)**2)')

h = df_mass.Histo1D(
    ('Dimuon_mass', 'Dimuon_mass'))

h.Draw()
```



'Dimuon\_mass')

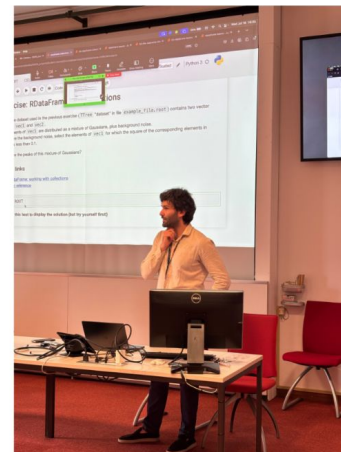
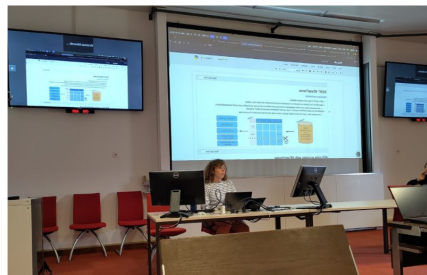


# Outreach and collaborations



We teach ROOT data analysis:

- At CERN, for the summer students
  - Around 200 students per year
- Online on various occasions
- At universities, on demand in collaboration with interested groups



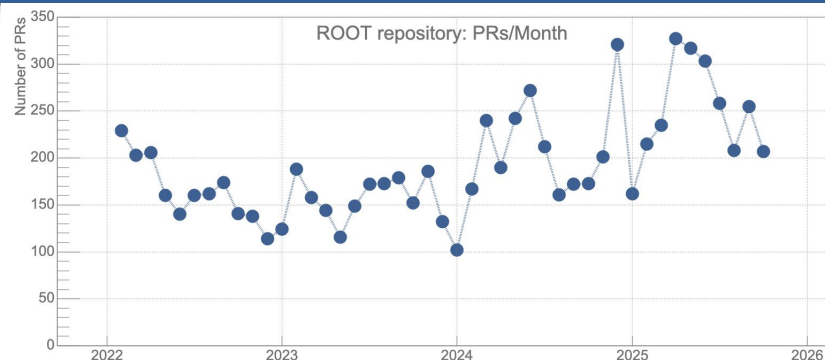


# Collaborate with Us

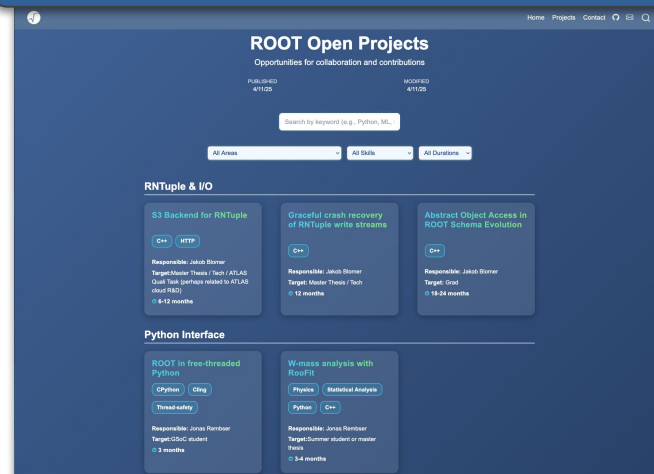
- ▶ ~100 unique contributors per year, 10% of the commits in `24 authored outside of the core dev team
- ▶ **A single, motivated individual can make the difference, irrespective of own level of experience!**

## How to start?

- ▶ We mark the easiest issues in the [GH tracker](#) with **good first issue**
- ▶ Selecting from the [list of open projects](#)
- ▶ [Simple contribution guidelines](#) are available



[http://root.cern/open\\_projects](http://root.cern/open_projects)





- ▶ CERN is a leading laboratory for High Energy Physics
- ▶ A rich scientific programme, with hard computing challenges
- ▶ ROOT is a tool for storage, processing, analysis and visualisation of physics data
  - A protagonist in the software landscape of the field
- ▶ Collaboration is possible and encouraged!
  - Contact us if you're interested in one of our student projects