Beyond repositories

Enabling actionable FAIR open data reuse services in particle physics

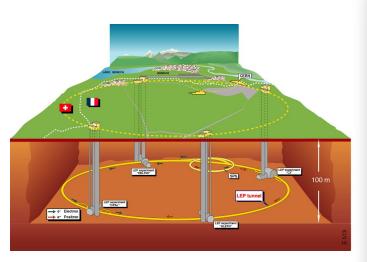
Diego Rodríguez

@diego_delemos

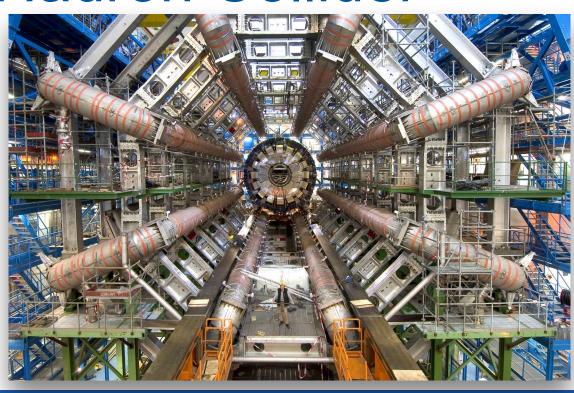
Tibor Šimko, Sünje Dallmeier-Tiessen, Sebastian Feger, Pamfilos Fokianos, Dinos Kousidis, Artemis Lavasa, Rokas Mačiulaitis, Jan Okraska, Diego Rodríguez, Anna Trzcińska, Ioannis Tsanaktsidis, Stephanie van de Sandt

CERN, Switzerland

CERN Hadron Collider

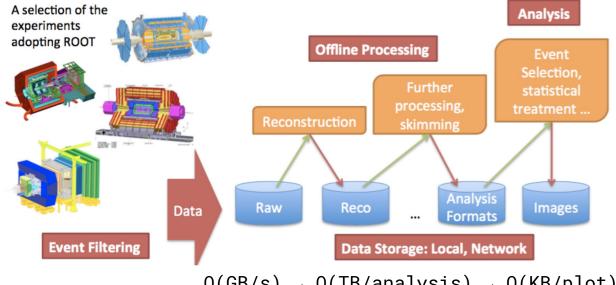


http://cds.cern.ch/record/842153





HEP data analyses

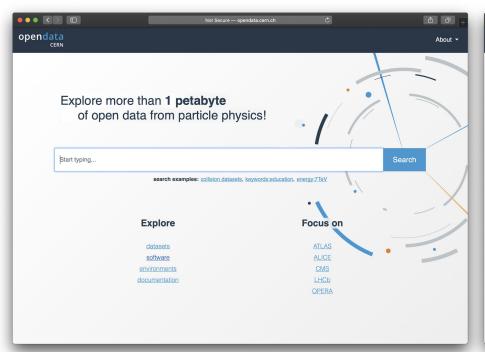


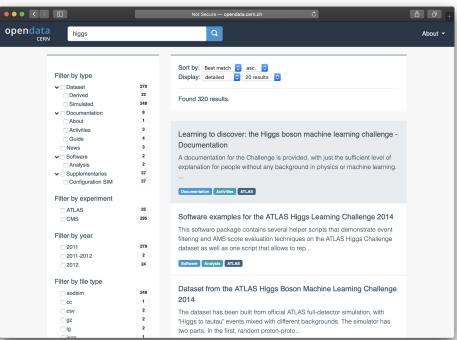
 $O(GB/s) \rightarrow O(TB/analysis) \rightarrow O(KB/plot)$

D. Krücker et al https://indico.desy.de/indico/event/18343



CERN Open Data portal

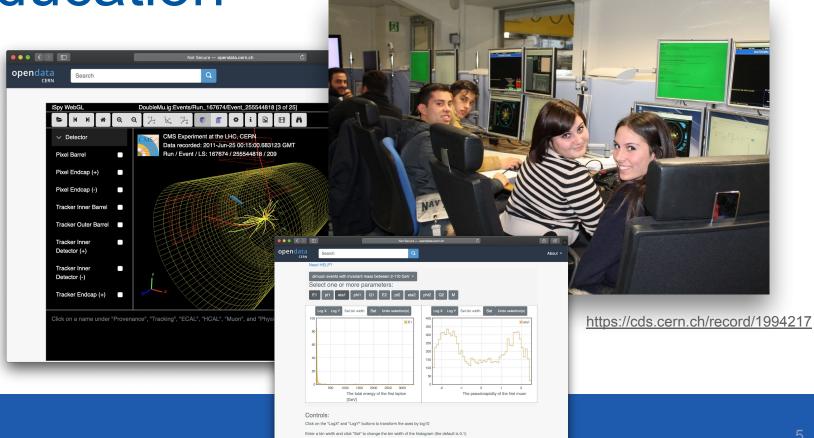




http://opendata.cern.ch/



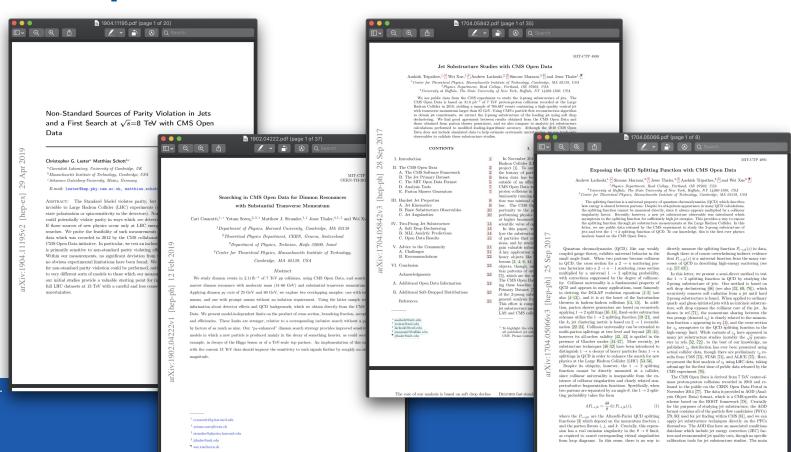
Education



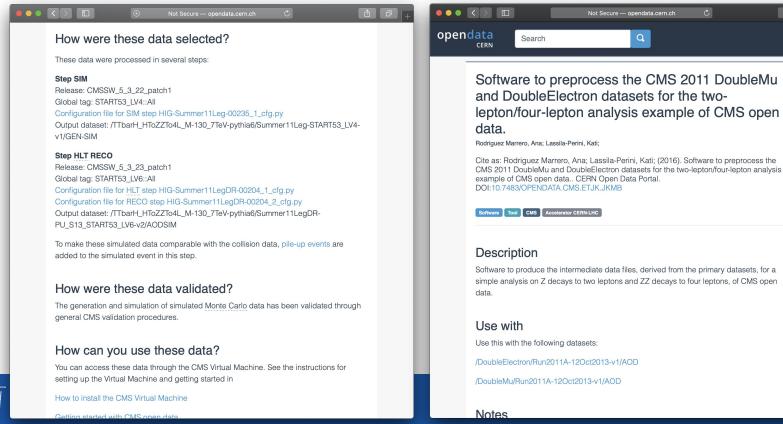
Click on the histogram and move to select a region along the x axis. Click "Undo selection(s)" to return to the original range.



Independent research



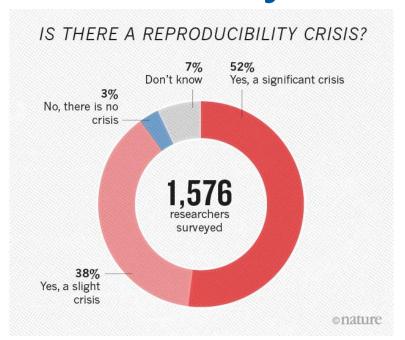
Tracking data provenance

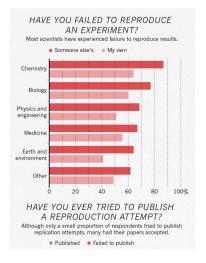




About -

Reproducibility crisis





https://www.nature.com/news/1-500-scientists-lift-the-lid-on-reproducibility-1.19970



How FAIR are we?

Findable Interoperable Reusable



✓ Accessible ✓



where is the data?

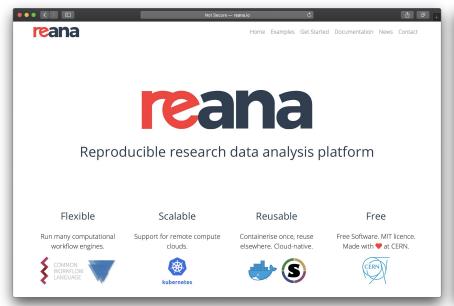
where is the code?

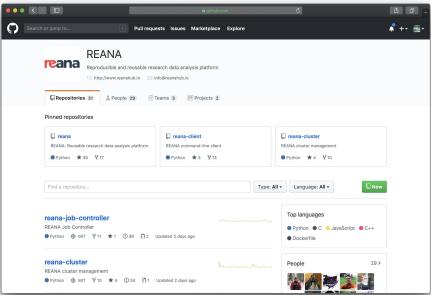
what is the **environment**?

what is the workflow?



Reproducible analyses





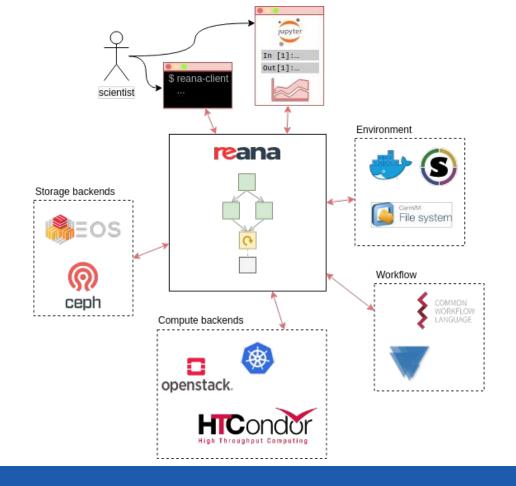
http://reana.io/

https://github.com/reanahub



REANA Architecture

- Cloud-native application
- Extensible
 - Storage backends
 - Compute backends
 - Container technologies
 - Workflow engines







data + code + environment + workflow =

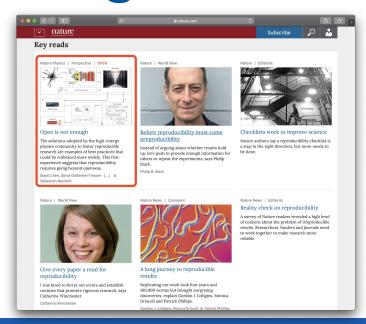


Conclusions

- successful open data and reproducible research example in particle physics
- beyond repositories: actionable data via runnable examples
- FAIR ≠ open, FAIR should start early
- "top-down" approach: funding agencies, best practices
- "bottom-up" activities: building useful tools for scientists



Open is not enough





Open is not enough

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The solutions adopted by the high-energy physics community to foster reproducible research are examples of best practices that could be embraced more widely. This first experience suggests that reproducibility requires going beyond openness.

vasive goals across research communities, political circles and funding bodies1-3. The understanding is that open and reproducible research practices enable scientific reuse, accelerating future projects and discoveries in any discipline. In the struggle to take concrete steps in pursuit of these aims there has been much make research products and scientific results open quickly.

Although these are laudable and necessary first steps, they are not sufficient to bring about the transformation that would allow us to reap the benefits of open and reproducible research. It is time to move beyond the rhetoric and the trust in quick fixes and start designing and implementing tools to power a more profound change.

Our own experience from opening up vast volumes of data is that openness cannot simply be tacked on as an afterthought at the end of the scientific endeavour. In addition, openness alone does not guarantee reproducibility or reusability, so it should not be purwhich need to be captured throughout the usual iterative and closed research lifecycle, ready for a timely open release with the results.

Thus, we argue that having the reuse of research results as a goal requires the adoption of new research practices during the data analysis process. Such practices need to be tailored to the needs of each given discipline with its particular research environment, culture and idiosyncrasies. Services and tools should be developed products are more likely to be useful when shared openly.

In tackling the challenge of enabling reusable research, we keep these ideas as our guiding light when putting changes into selves are intrinsically complex due to the large data volume and practice in our community-high-energy physics (HEP). Here, we illustrate our approach, particularly through our work at CERN, and present our community's requirements and rationale. We hope that the explanation of our challenges and solutions will tion on the analysis methods is maintained, the complexity of the

pen science and reproducible research have become per- flows for reproducible and reusable research more widely in other

Approaching reproducibility and reuse in HEP

To set the stage for the rest of this piece, we first construct a more nuanced spectrum in which to place the various challenges facing discussion and awareness-raising, often accompanied by a push to HEP, allowing us to better frame our ambitions and solutions. We choose to build on the descriptions introduced by Carole Goble⁴ and Lorena A. Barba5 shown in Table 1.

These concepts assume a research environment in which multiple labs have the equipment necessary to duplicate an experiment, which essentially makes the experiments portable. In the particle physics context, however, the immense cost and complexity of the experimental set-up essentially make the independent and complete replication of HEP experiments unfeasible and unhelpful. HEP experiments are set up with unique capabilities, often being the only facility or instrument of their kind in the world; they are also constantly being upgraded to satisfy requirements for higher sued as a goal in itself. Focusing on data is also not enough: it needs energy, precision and level of accuracy. The experiments at the Large to be accompanied by software, workflows and explanations, all of Hadron Collider (LHC) are prominent examples. It is this uniqueness that makes the experimental data valuable for preservation so that it can be later reused with other measurements for comparison. confirmation or inspiration.

Our considerations here really begin after gathering the data. This means that we are more concerned with repeating or verifying the computational analysis performed over a given dataset rather than with data collection. Therefore, in Table 2 we present a variawith the idea of meshing seamlessly with existing research procedures, encouraging the pursuit of reusability as a natural part of ment in which 'experimental set-up' refers to the implementation researchers' daily work (Fig. 1). In this way, the generated research of a computational analysis of a defined dataset, and a 'lab' can be thought of as an experimental collaboration or an analysis group.

In the case of computational processes, physics analyses themalgorithms involved. In addition, the analysts typically study more than one physics process and consider data collected under different running conditions. Although comprehensive documentastimulate discussions around the practical implementation of work- software implementations often hides minute but crucial details,





