

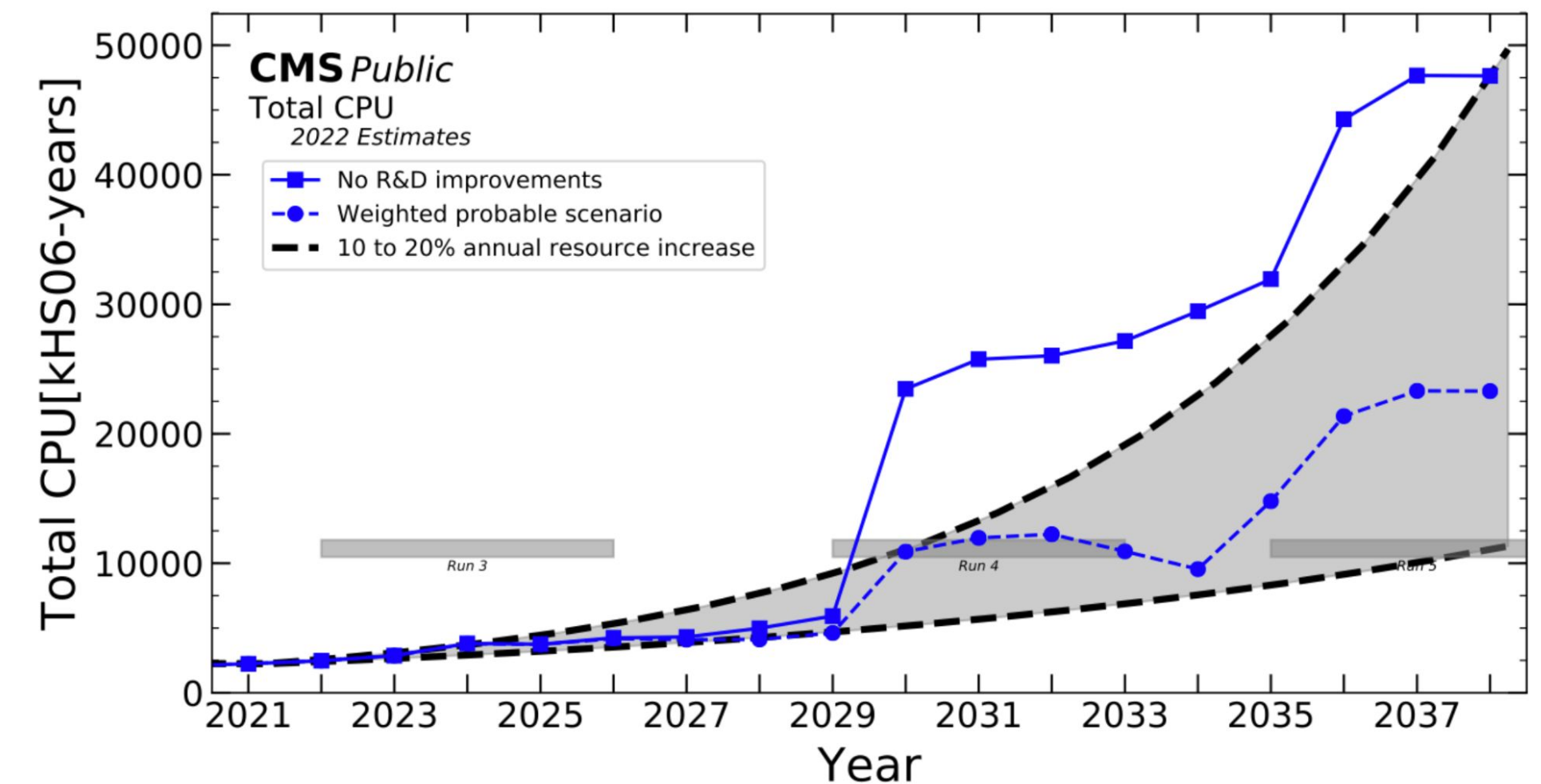
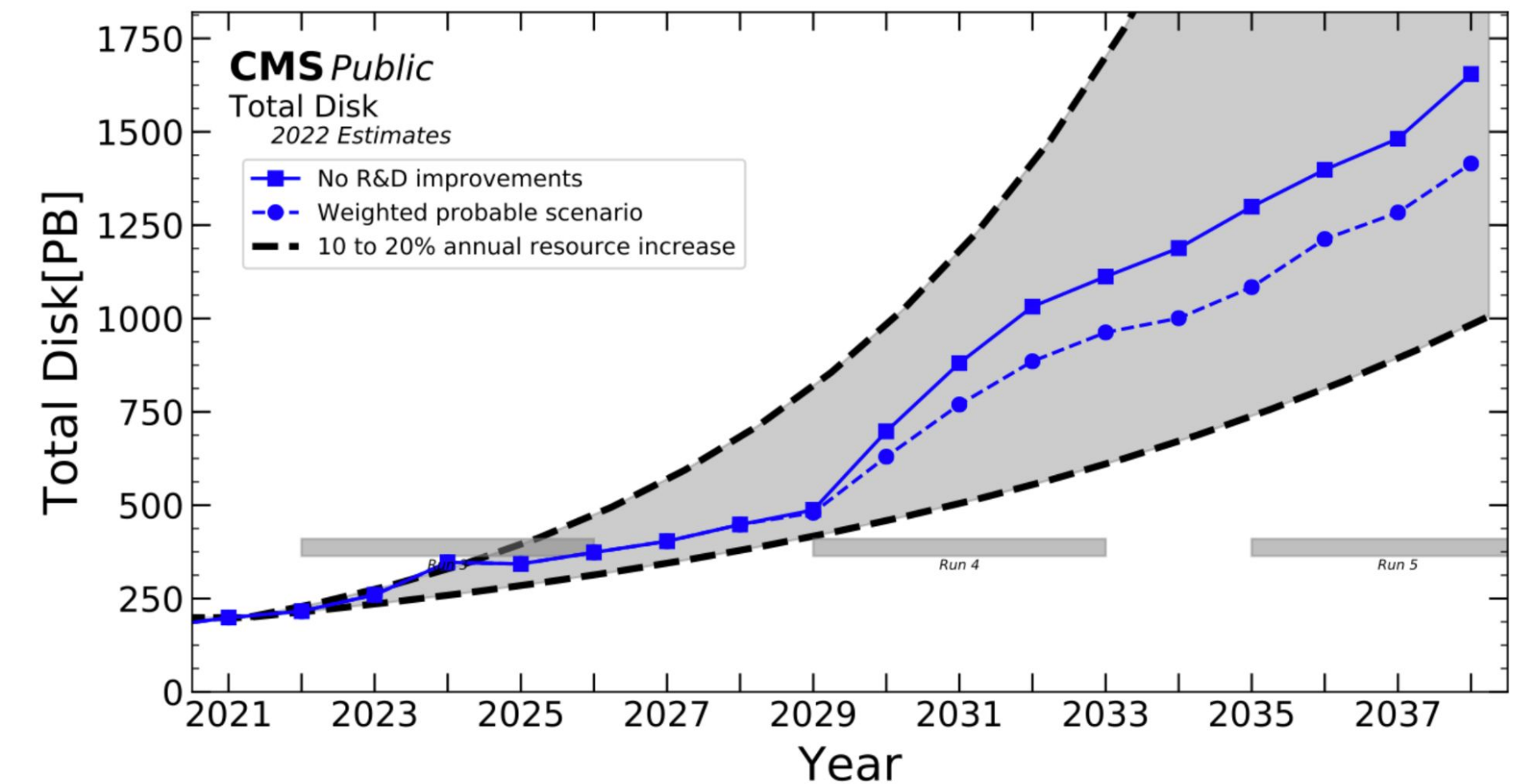
Coffea-casa analysis facility

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International Symposium on Grids & Clouds (ISGC) 2023

Preparing High Energy Physics (HEP) Analysis Facilities for HL-LHC

- Preparing for HL-LHC era starting in 2029
 - Two general-purpose detectors, **ATLAS** and the **Compact Muon Solenoid (CMS)**, are expected to **generate up to an exabyte each year** = about 1000 times as much as the data used to discover the Higgs boson in 2012
- **Community will need to come up with new computing model and integrate existing resources**



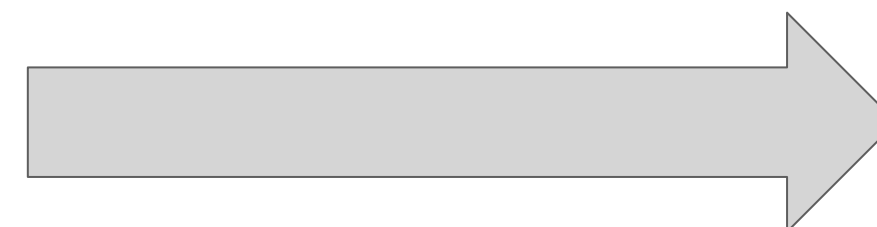
HEP Analysis facilities: what it is?

HEP Analysis Facilities are usually used for end-user analysis

- **People, software / services, and hardware** used to support analysis activities for an experiment.
- **Services which** includes:
 - Access to experimental data products.
 - Storage space for per-group or per-user data (often ntuples).
 - Access to significant computing resources.
- **Physics software:** ROOT and the growing Python-based ecosystem.
- **Computing hardware** (currently) looks like most of our computing facilities: CPUs and disks, maybe GPUs

B.Bockelman, Analysis Facilities for the HL-LHC, Snowmass

“How is this different than an existing facility, e.g. USCMS LPC @ Fermilab?”



Faster analysis and better end user experience

HEP Analysis facilities: requirements

- **Interactivity:** need to support both interactive analysis and batch mode
- **Low latency data access:** expected to facilitate low latency data access to achieve best performance (via various data access patterns)
- **Reusability:** should support extraction of user defined experiment data formats to migrate them onto laptops, desktops, workstations at home institutions or at home
- **Easy Deployment:** AF services expected to be deployed with industry standard platforms such as Kubernetes to facilitate easy deployment within a Tier-1s/Tier-2s/Tier-3s (easily co-locating next to existing resources)

Building blocks used for designing AFs

Modern authentication (AIM/OIDC), tokens, macaroons

Efficient data delivery and data management technologies

Columnar analysis and support new pythonic ecosystem

Modern deployment and integration techniques

Support for object storage

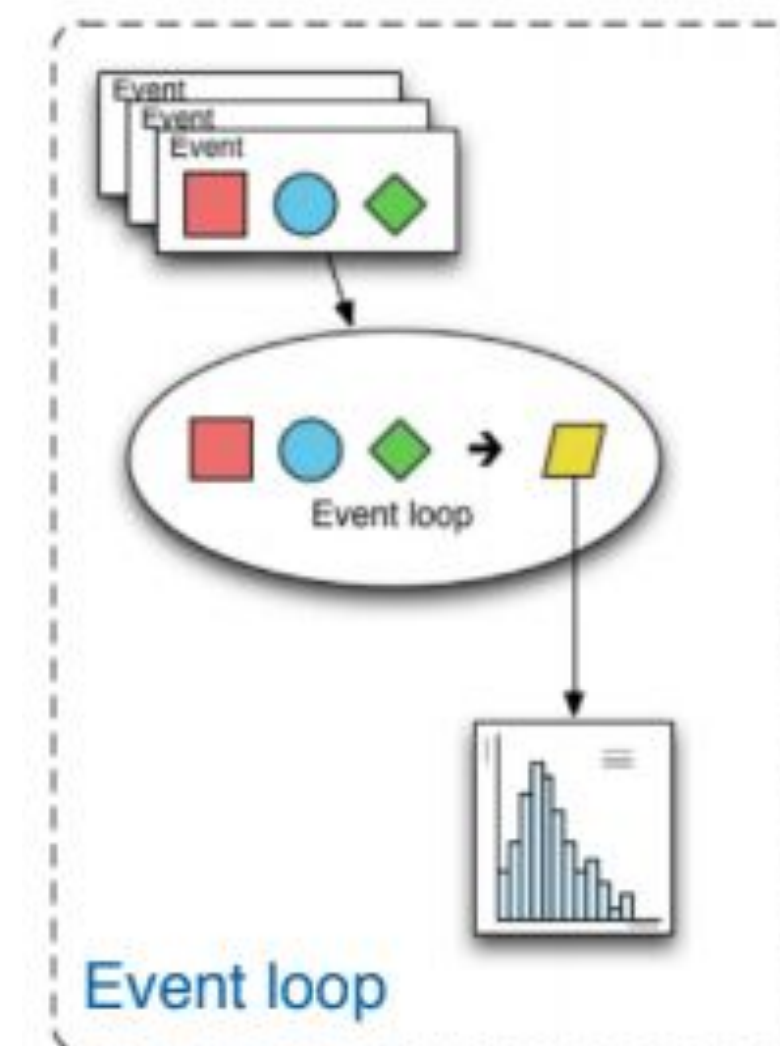
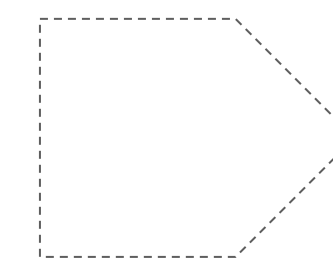
Efficient data caching solutions

Easy integration with existing HPC resources

Analysis facilities: columnar analysis frameworks

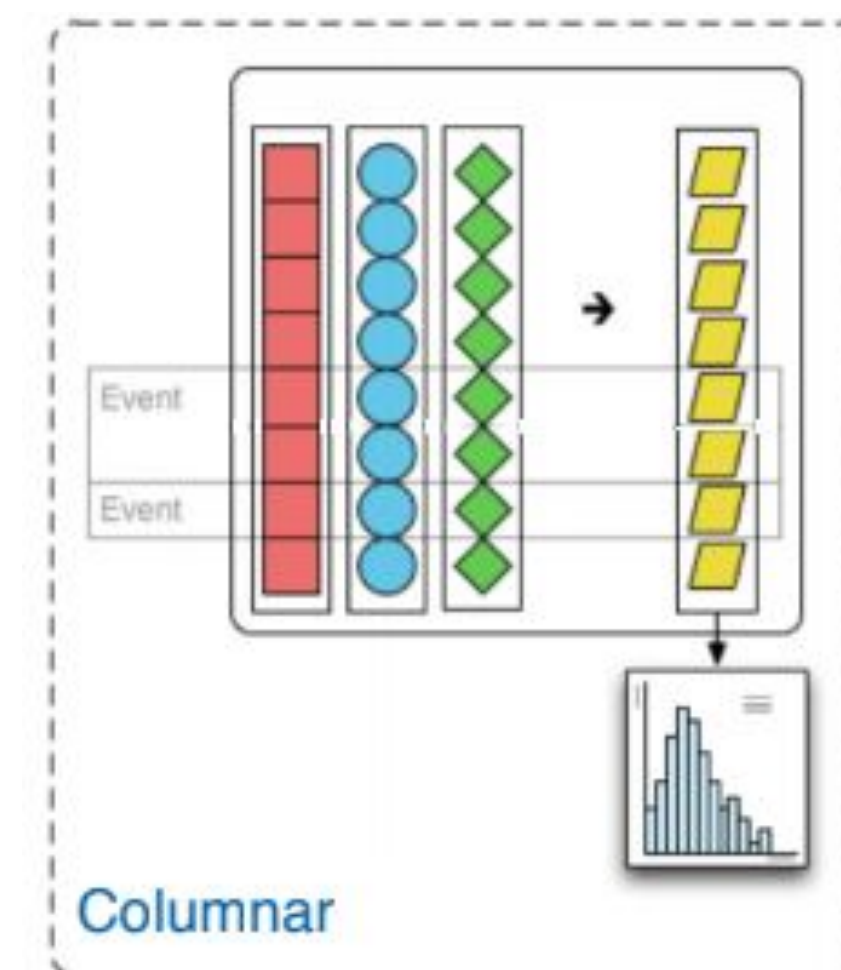
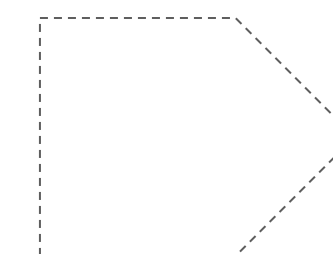
Traditionally physicist have done **row-wise analysis**

- Process one physics event at a time serially
- Used in scripts and batch jobs (and now also in notebooks)

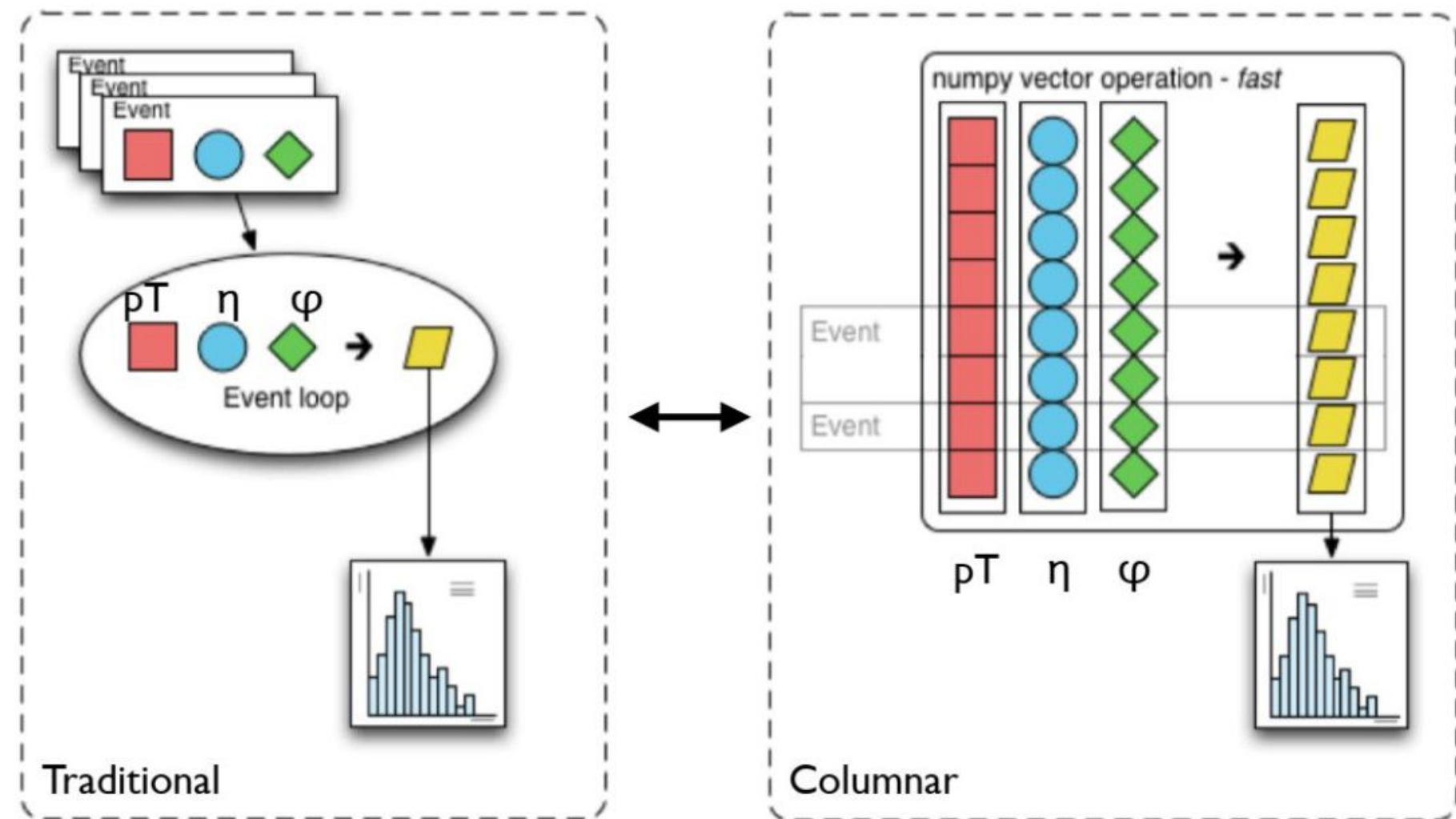


Trending towards **column-wise (tidy/big data) analysis**

- Low-latency access to CPU resources
- Fast access to a significant amount of disk



Building blocks: columnar analysis frameworks (coffea analysis framework)



User just needs to define a high-level wrapper around user analysis code: **the coffea processor** and coffea framework will take care of everything incl. **scaling-out**

New columnar data analysis concepts!

Distributed executors!



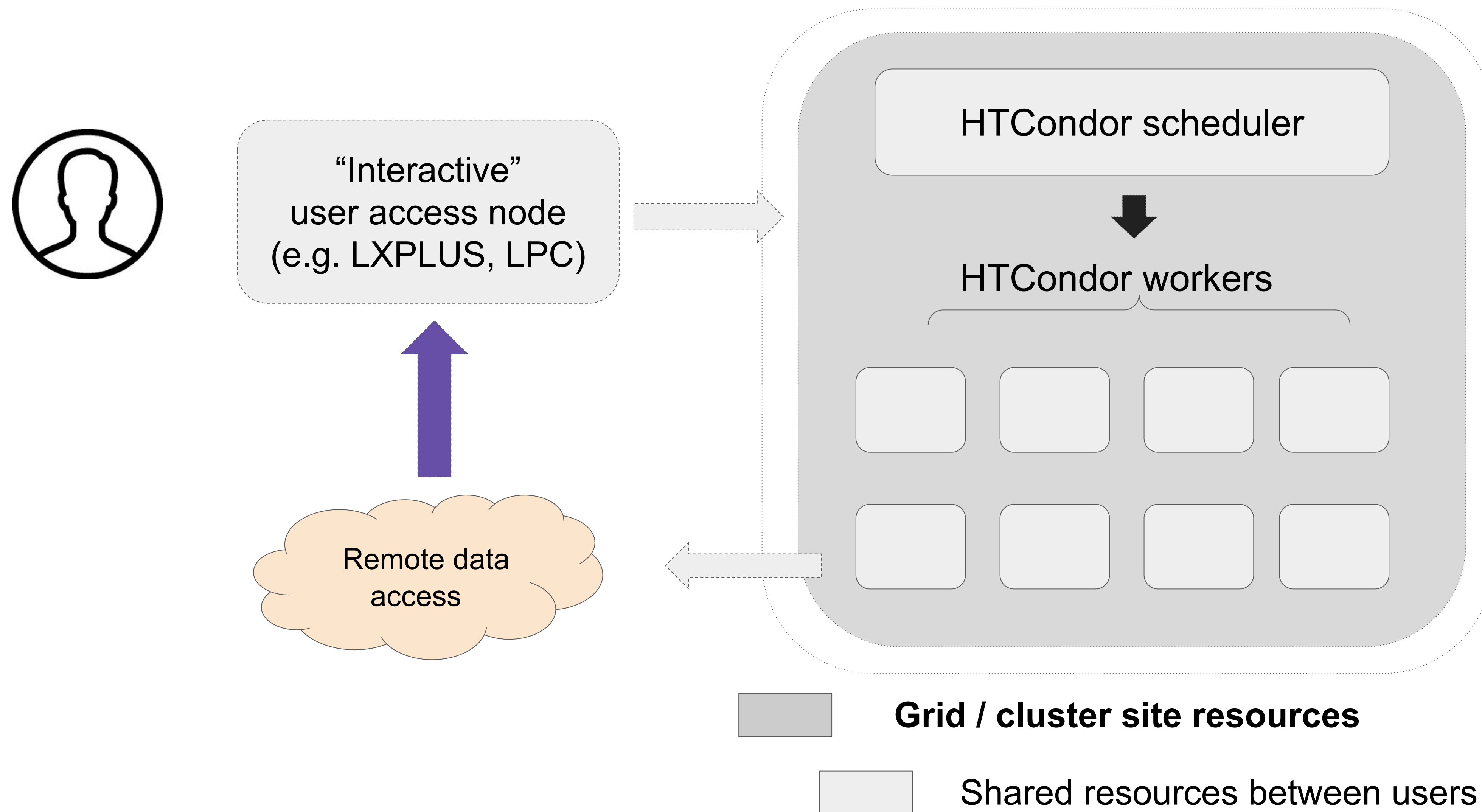
Smith, N., Gray, L., Cremonesi, M., Jayatilaka, B., Gutsche, O., Hall, A., ... & Pivarski, J. (2020). Coffea columnar object framework for effective analysis. In *EPJ Web of Conferences* (Vol. 245, p. 06012). EDP Sciences.

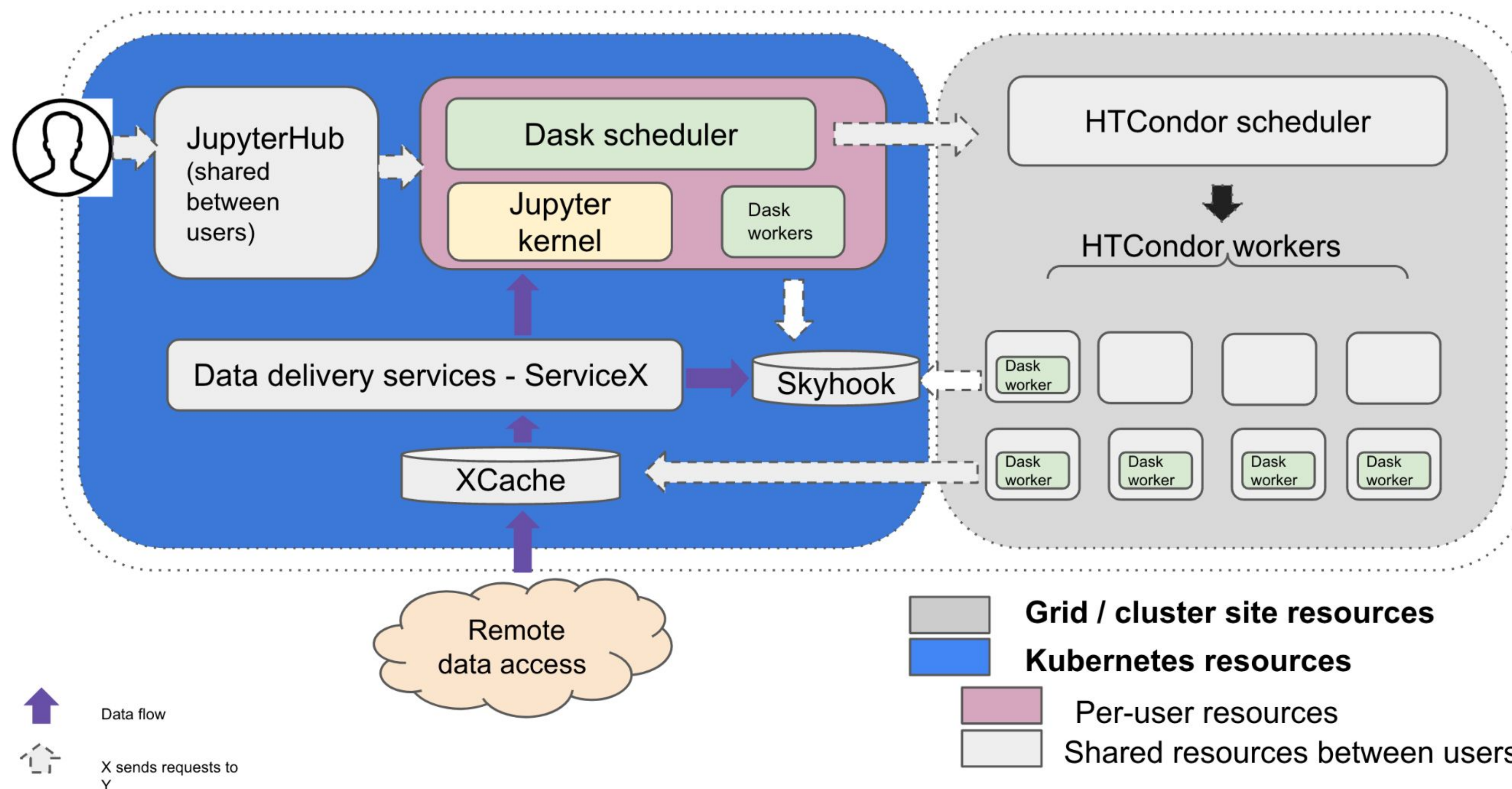
Contributors 51

+ 40 contributors

<https://github.com/CoffeaTeam/coffea>

Simplified diagram of hypothetical Analysis Facility currently used by user

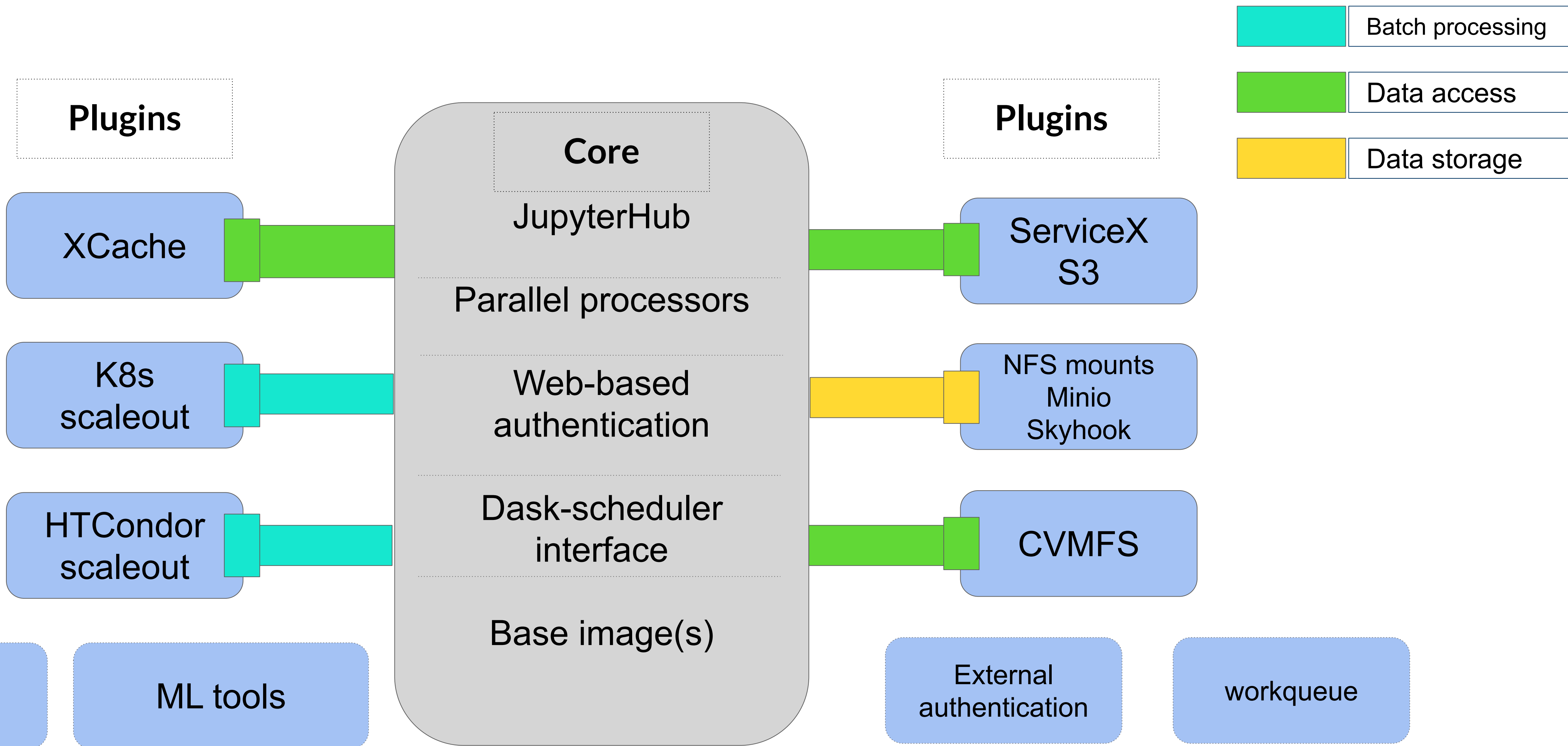




Coffea-casa facility @ UNL is co-located at US.CMS Tier-2 at University Nebraska-Lincoln and other instance is co-located at US.ATLAS Tier-3 at University UChicago

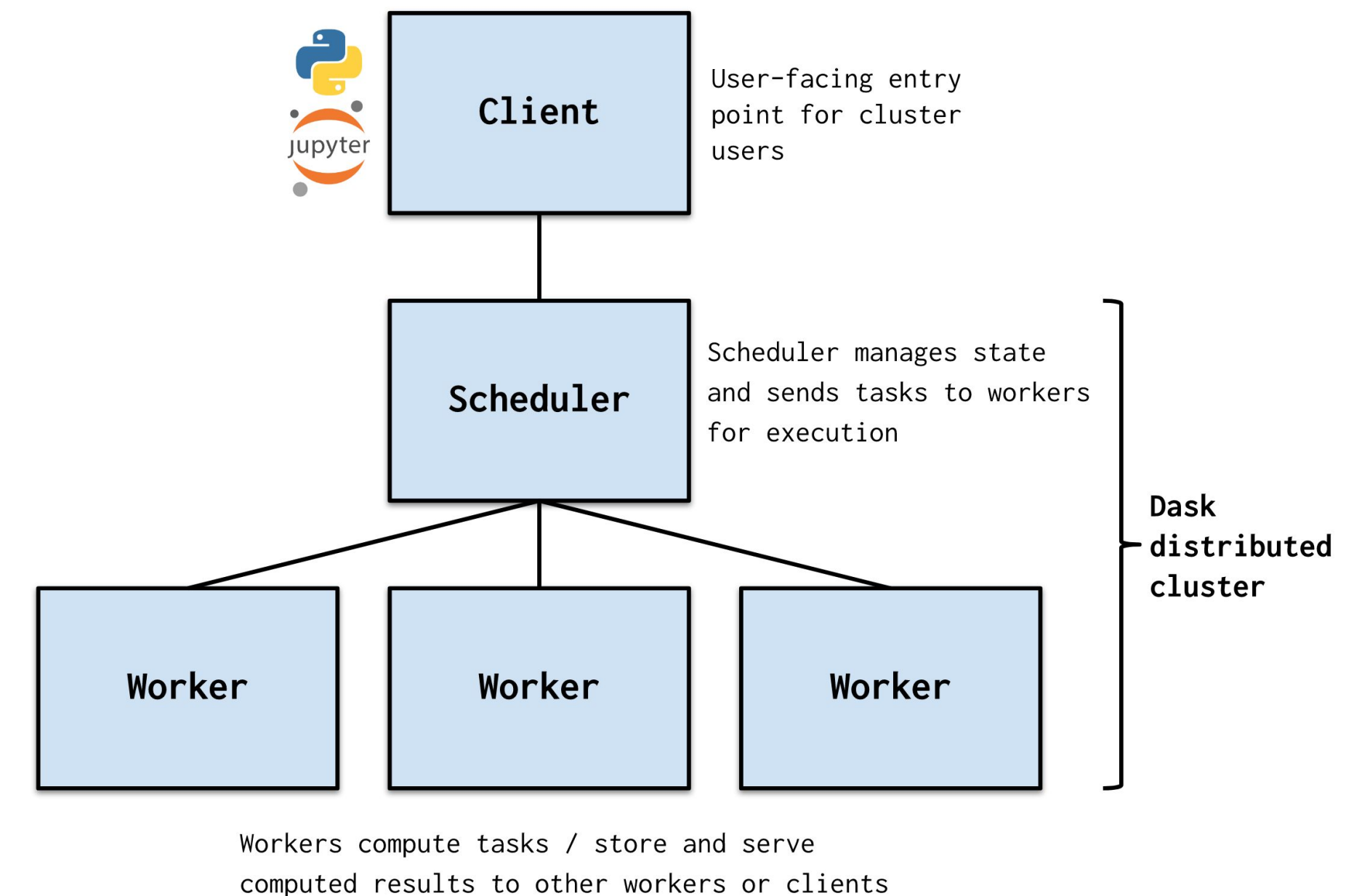


Coffea-casa AF components



Building blocks: Dask

- Dask provides a task-management computational framework in Python (based on the manager-worker paradigm)
- Integrates with HPC clusters, running a variety of schedulers including SLURM, LSF, SGE and HTCondor via “*dask-jobqueue*”
- **This allows us to create a user-level interactive system via queueing up in the batch system**



Dask can be used inside Jupyter or you can simply launch it through Jupyter and connect directly from your laptop

Building blocks: Batch scale-out

- ***CoffeaCasaCluster***: extending HTCondorCluster integration for Dask
 - To handle the customizations needed for the Coffea-casa environment, we developed the *CoffeaCasaCluster* object, an extension of ***Dask-jobqueue***'s *HTCondorCluster* object.
 - *CoffeaCasaCluster* ensures the Dask worker starts with the appropriate Docker container in the HTCondor batch system with appropriate configurations and with the firewall ports configured correctly.
- **Looking into new backends:**
 - *Workqueue* (<http://ccl.cse.nd.edu/software/workqueue/>)

Building blocks: AuthN & AuthZ

- **Authentication inside the system is independent of grid credentials**
- Coffea-casa facility uses **OpenID Connect (OIDC)**
 - CMS, ATLAS IAM for experiment specific facilities
 - CILogon with COmanage for Opendata facility
- **Token authentication**
 - Pre-generated token for authentication with HTCondor, required for Dask scale-out to the larger resources
 - Pre-generated data access token for authentication with a local XRootD server (XCache)
 - R&D on enabling tokens for ServiceX data delivery service
- **Other credentials**
 - Generated X.509 credentials (including a CA, host certificate, and user certificate) for use in Dask for TLS as well for user communication to Dask scheduler endpoint
 - Security: TLS enabled communication between workers and scheduler by default

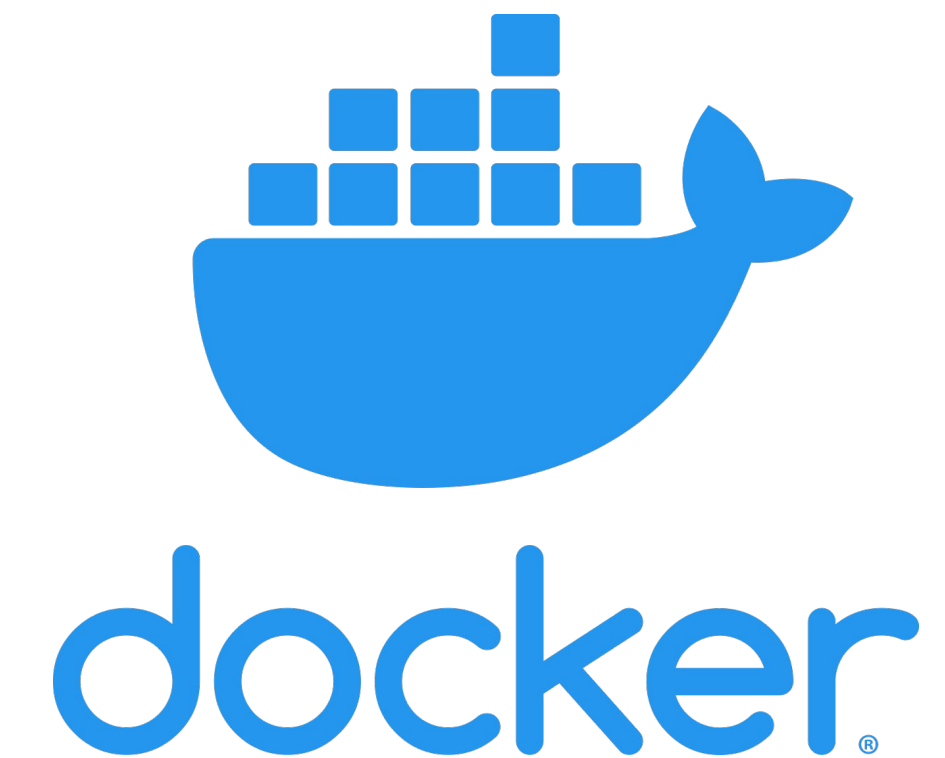
Building blocks: orchestration

For users:

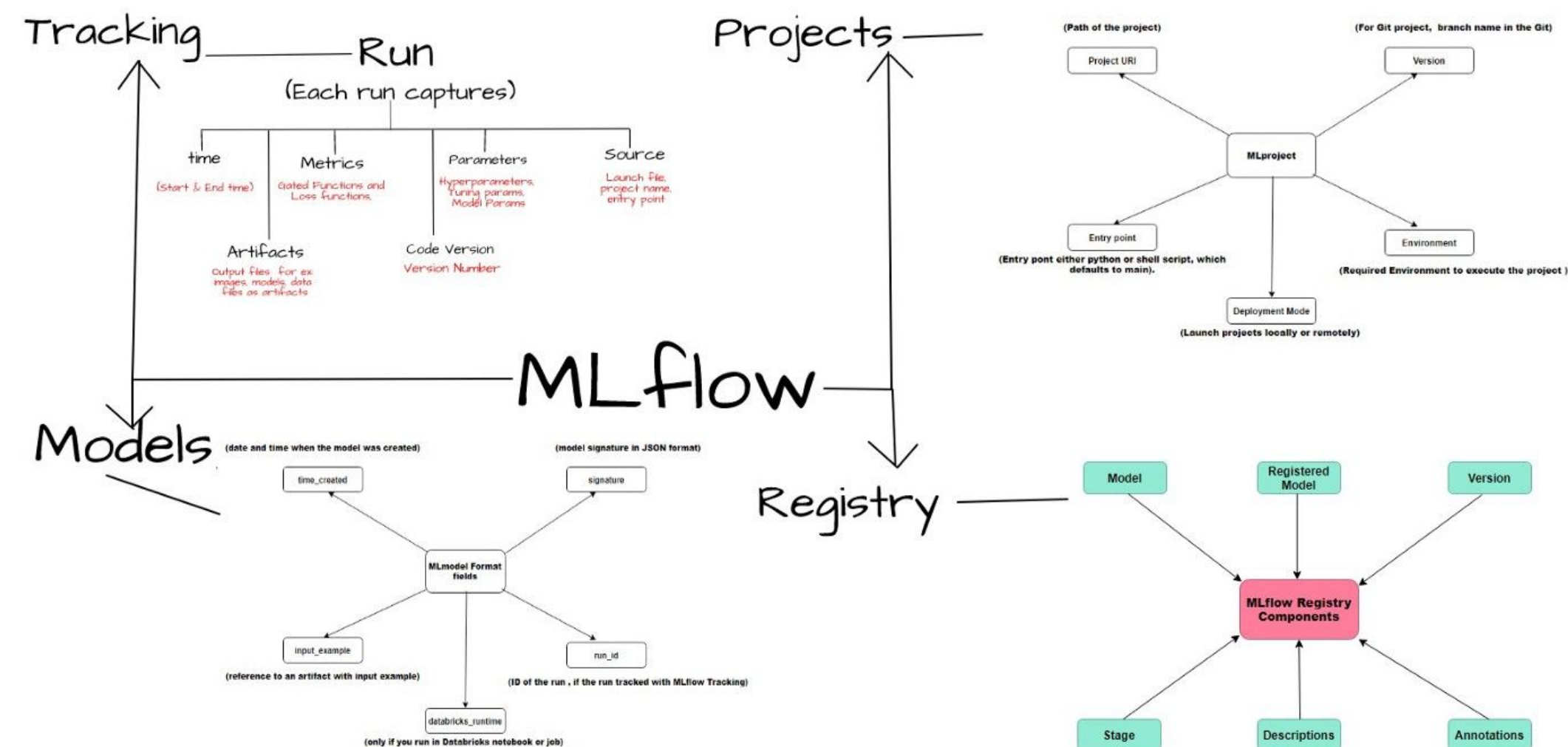
- Highly customized “analysis” Docker container(s)
- **Planning to add Binderhub support**
 - It will allow users to share reproducible interactive computing environments from their code repositories at coffea-casa

For developers:

- All features are **incorporated into a Helm chart** (Kubernetes packaging format)

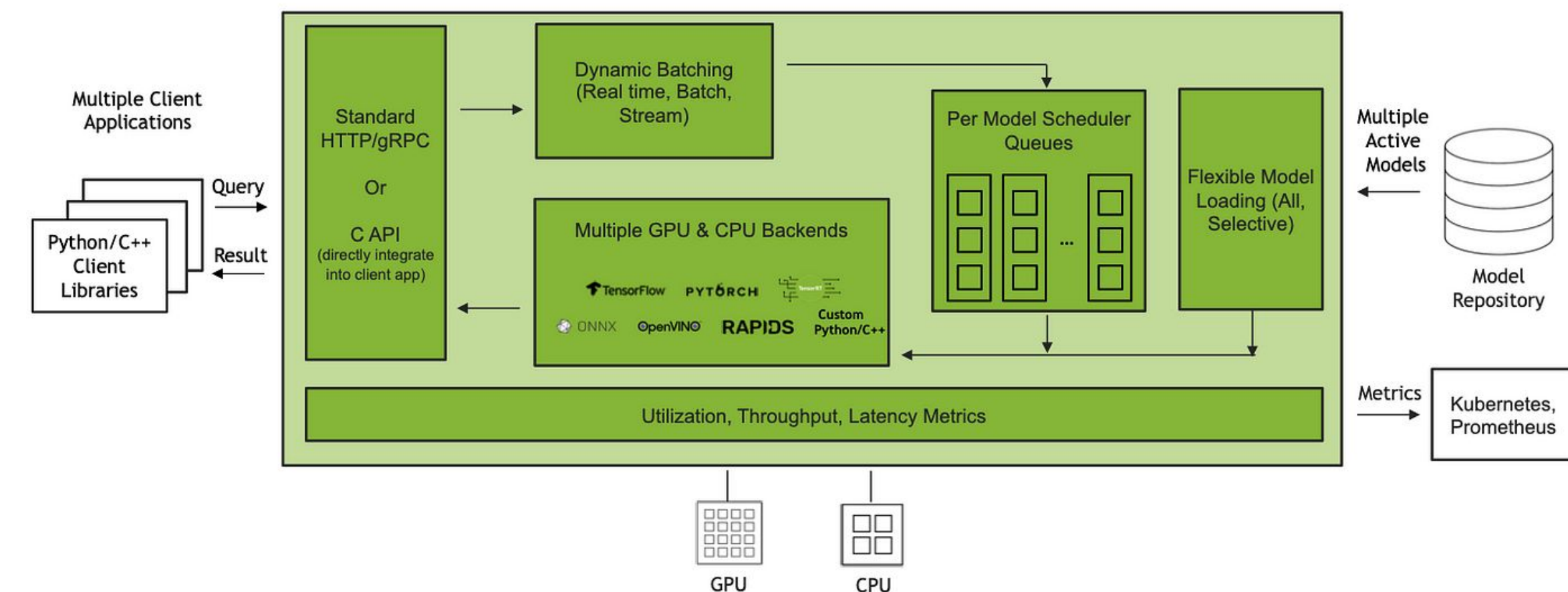


Building blocks: machine learning services and tools



NVIDIA TRITON INFERENCE SERVER ARCHITECTURE

Open-Source Software For Scalable, Simplified Inference Serving



MLflow

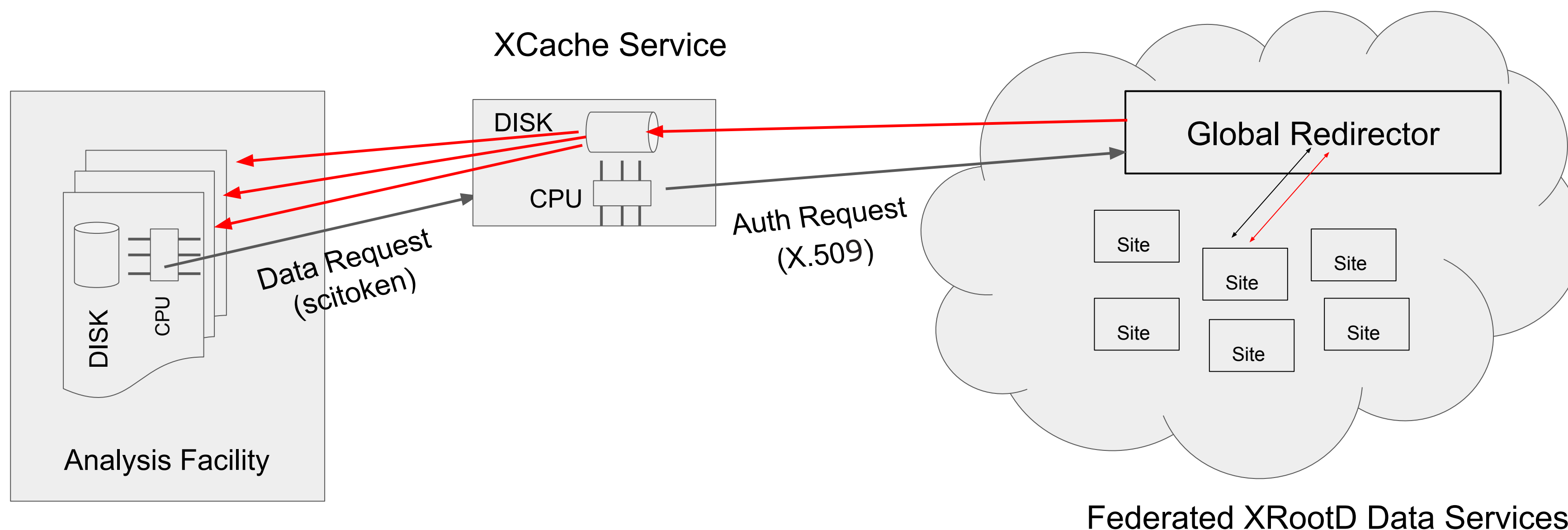
- Provide a central store to manage models (their versions and stage transitions)
- Allow packaging and re-deploying models
- Allows easily to reproduce code
- Provides easy tracking of ML experiments

Triton Inference Server

- **Support for various deep-learning (DL) frameworks**
- **Simultaneous execution**—Triton can run multiple instances of a model, or multiple models, concurrently, either on multiple GPUs or on a single GPU.
- **Dynamic scheduling and batching**
- **Model ensembles**

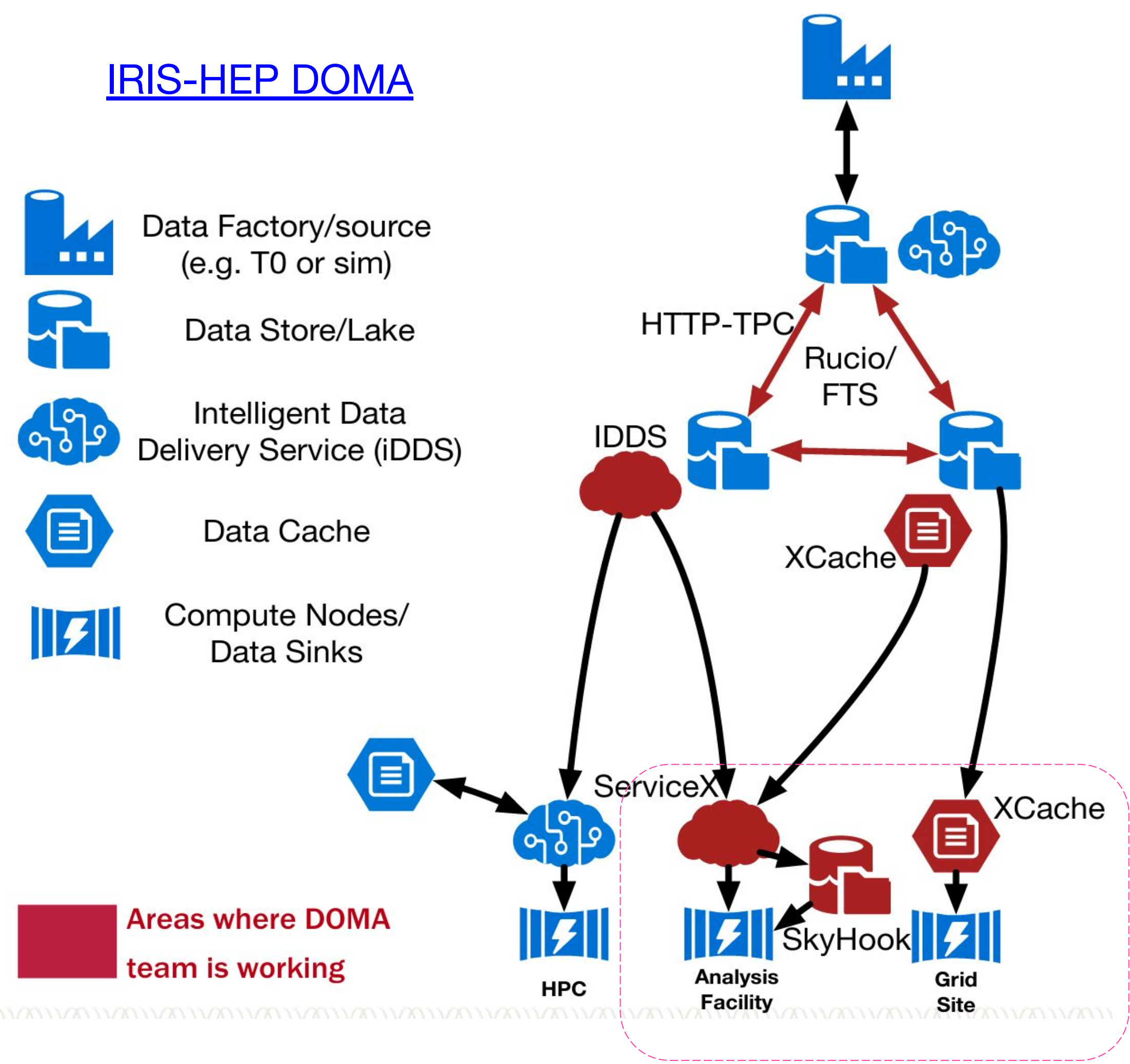
Building blocks: data access (e.g. XCache)

- For speeding up data access Nebraska Tier-2 hosts an **XCache service with 90TB of cache space**
- Access data hosted by an HEP experiment:
 - *no GSI credential within the facility, the auto-generated data access scitoken can be used to authenticate with an proxy service based on XRootD/XCache*

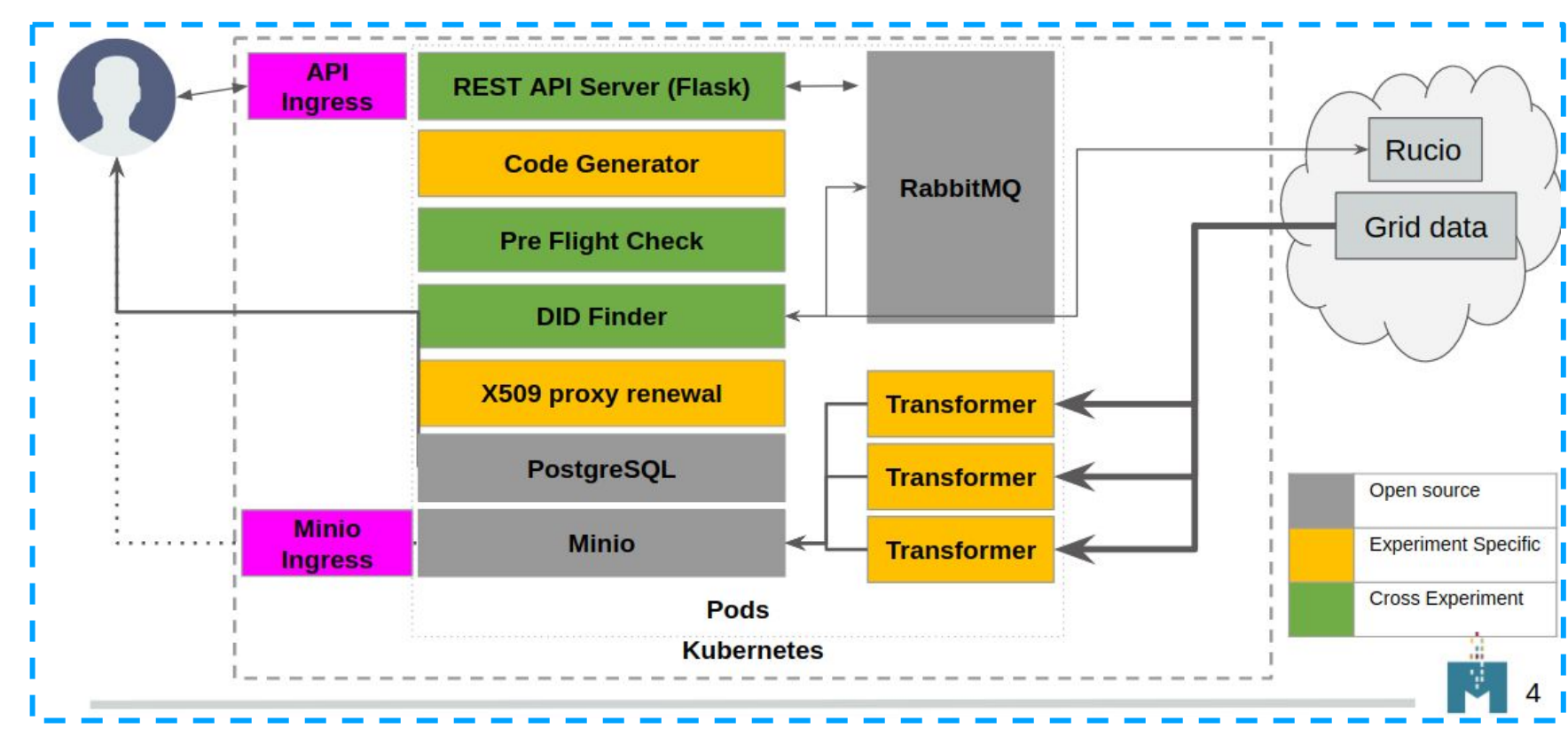


Building blocks: Data Organisation and Management and Access (DOMA)

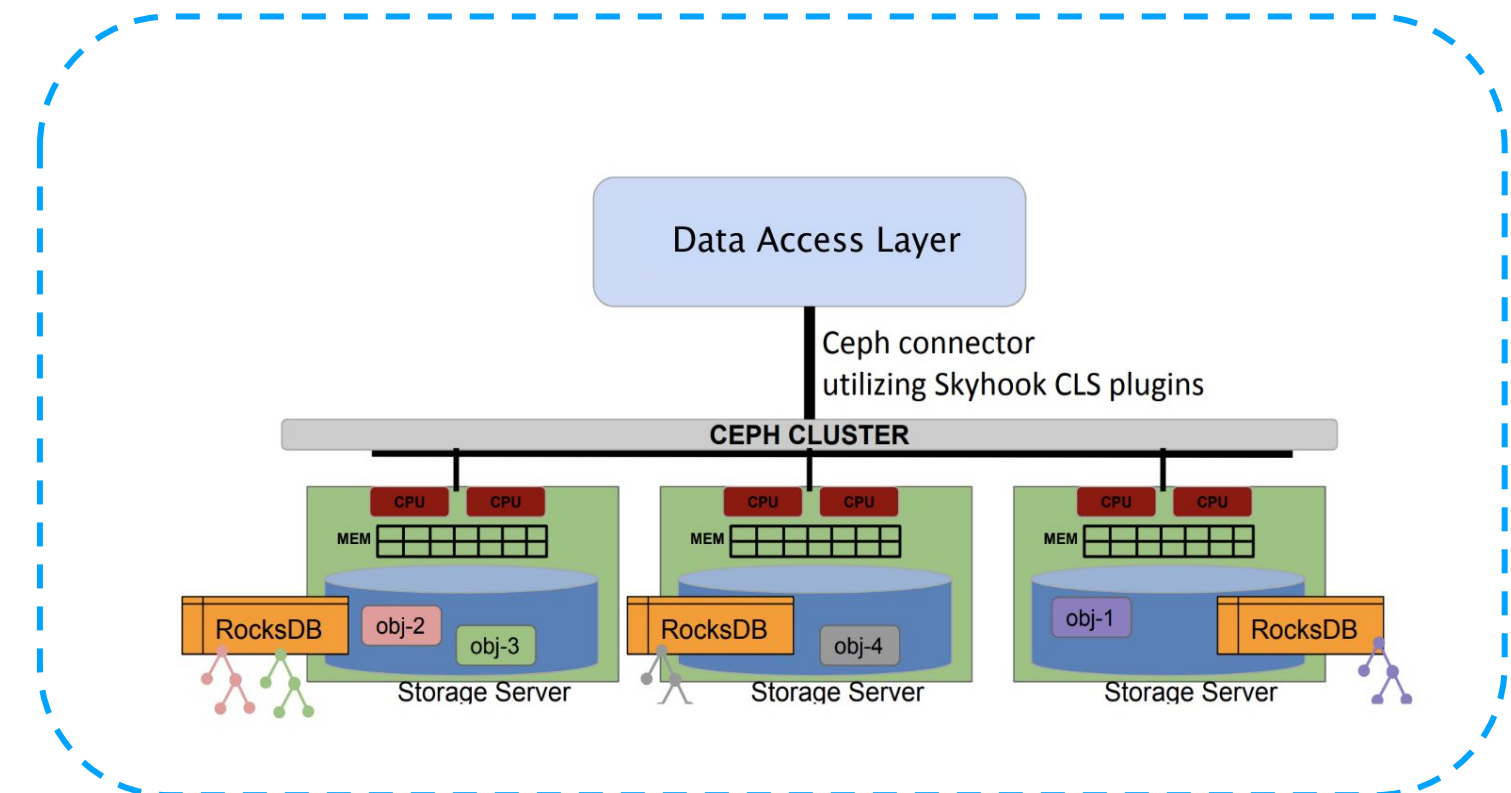
IRIS-HEP DOMA



Coffea-casa



ServiceX - data extraction and delivery delivery service for columnar analysis

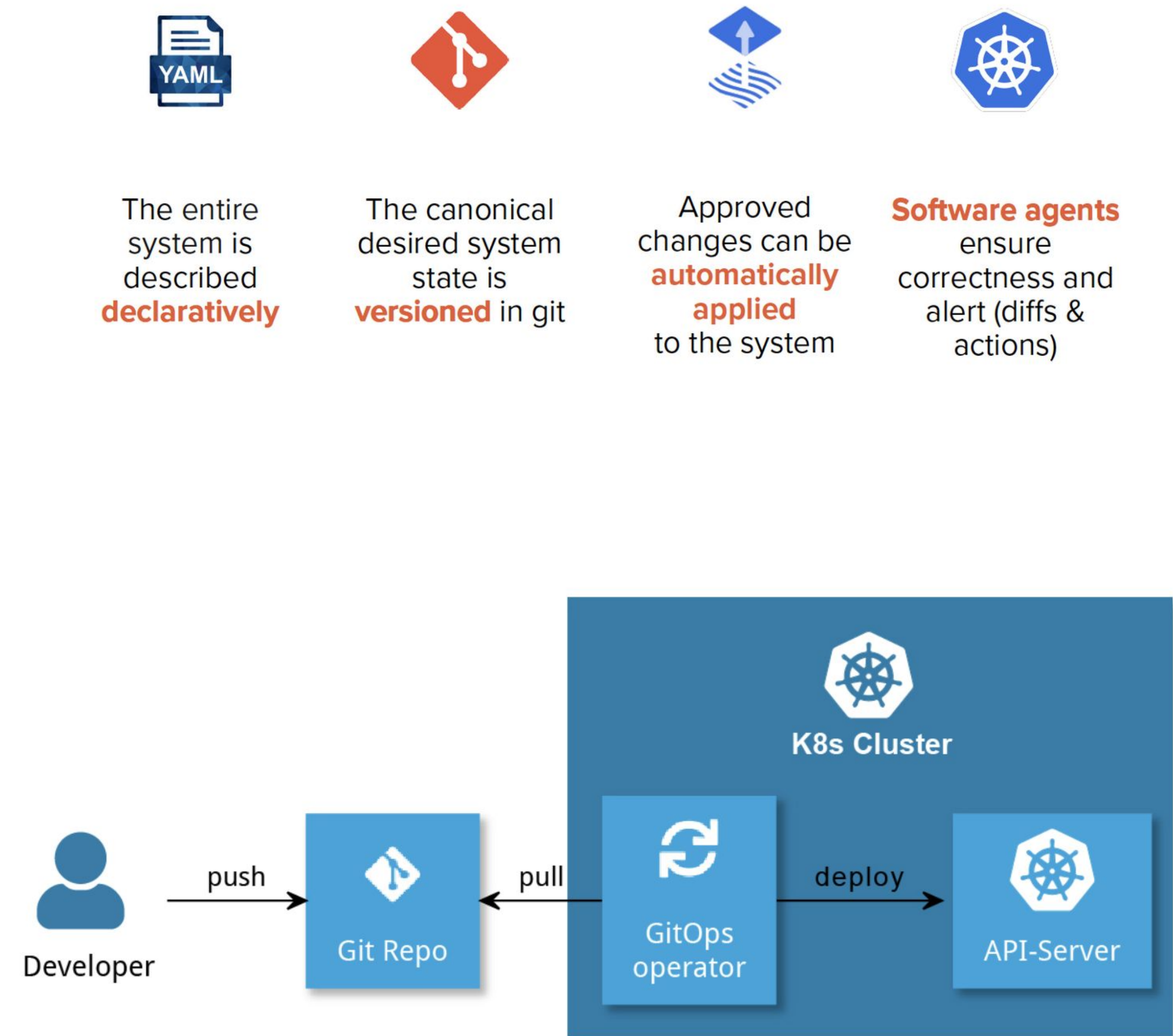


Skyhook - extension of the Ceph for the scalable storage of tables and for offloading common data management operations

New development strategies - GitOps

Principles of GitOps

- **GitOps** defined as a model for operating Kubernetes clusters or cloud-native applications (e.g. coffea-casa AF)
- **Concept: infrastructure-as-a-code**
 - Allow for rapid collaboration, better quality control, and automation (CD/CI).
 - AF is easily handled via a collaborative group of administrators in a deterministic manner.
 - Allows easily packages the core infrastructure (e.g., removing the site-specific passwords and secret keys) as a Helm chart.

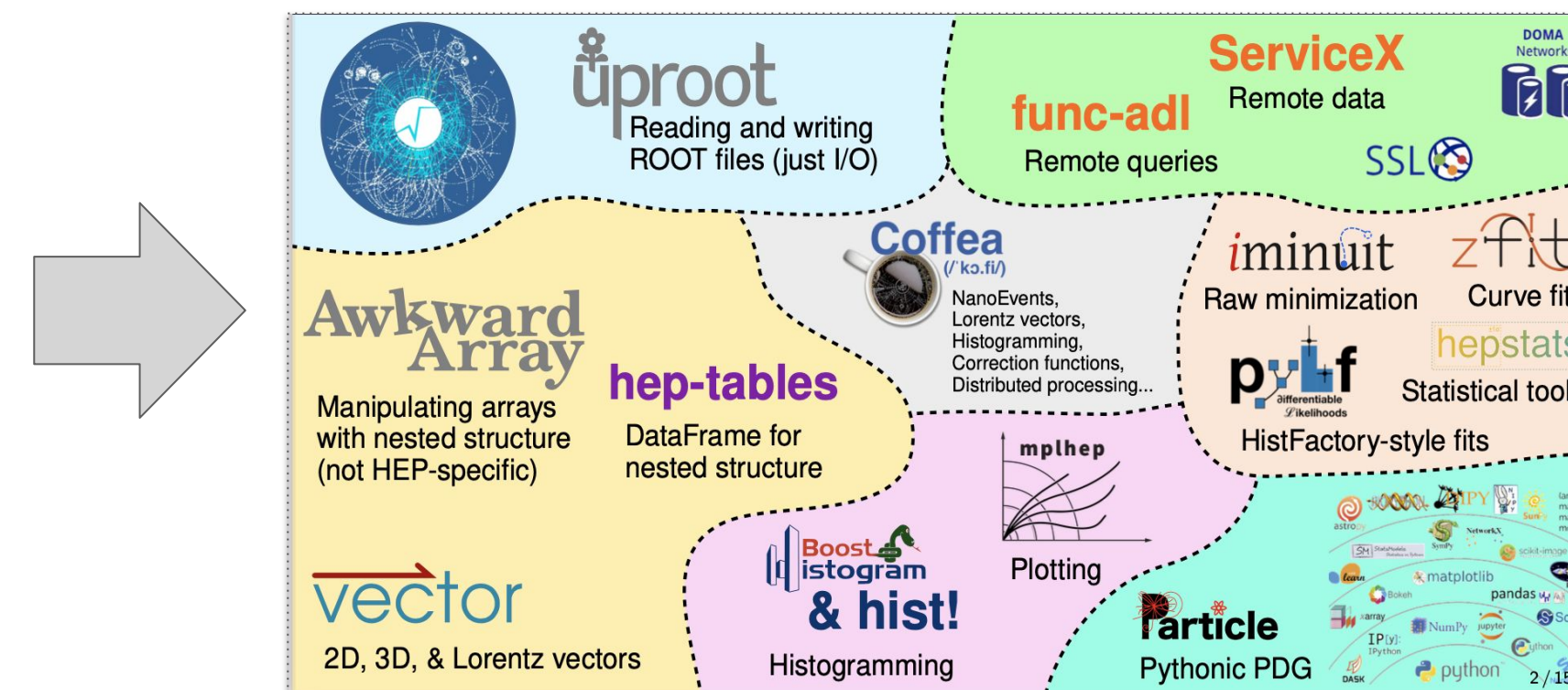


Analysis Grand Challenge

Motivation:

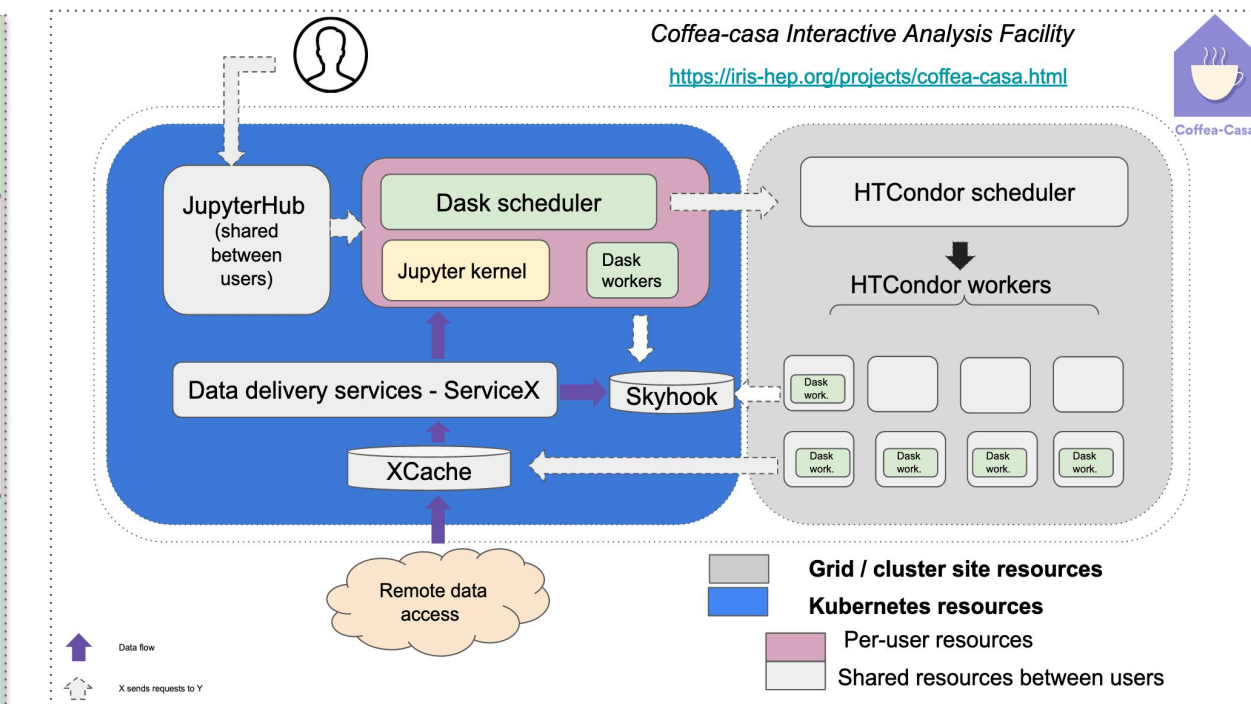
- Allow coping with HL-LHC data sizes by rethinking data pipeline
 - Evaluating the new Python analysis ecosystem and integrating a differentiable analysis pipeline
- Provide flexible, easy-to-use, low latency analysis facilities

Analysis Tools



Analysis Facilities

(coffea-casa AF or any other facility matching tech.requirements)



Execution of AGC analysis benchmark

Check the talk from A. Held “Physics analysis workflows and pipelines for the HL-LHC” on Thursday

Analysis Grand Challenge will be conducted during 2021–2023, leaving enough time for tuning software tools and services developed as a part of the IRIS-HEP ecosystem before the start-up of the HL-LHC and *organized together with the US LHC Operations programs, the LHC experiments and other partners.*

Thank you!
Q&A