

Utilizing the Elephant and NetworkUnit frameworks within the Collaboratory for an HPC-enabled workflow

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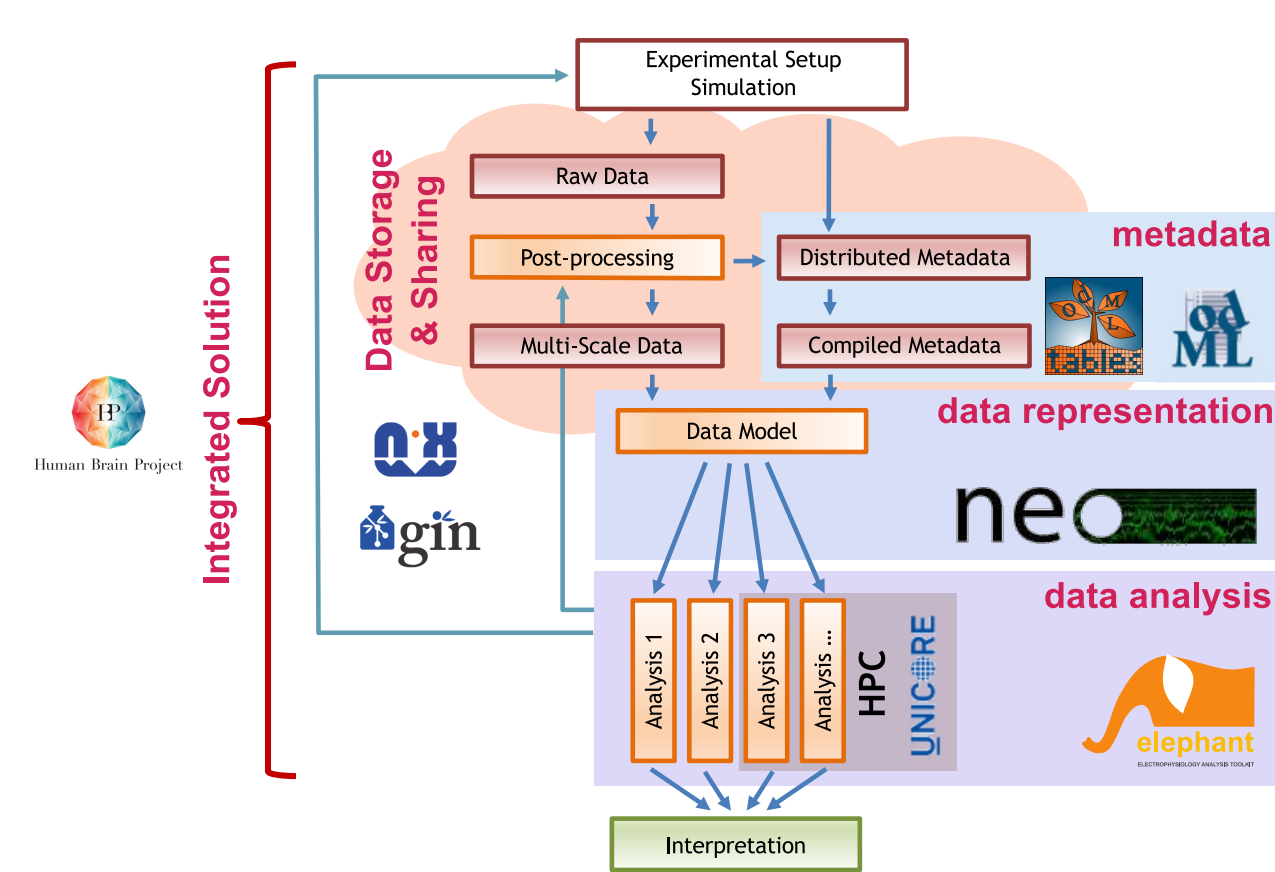
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Introduction



The need for **reproducible research** is a topic of intense discussion in the neurosciences. In the context of **data analysis**, we develop the Electrophysiology Analysis Toolkit (**Elephant**, [1]) as a central resource to provide tested and validated reference implementations of common analysis methods for activity data. However, reproducibility also requires such tools to be embedded in **collaborative, holistic workflows** [2] providing **clear, traceable** analysis steps from data acquisition to publication.

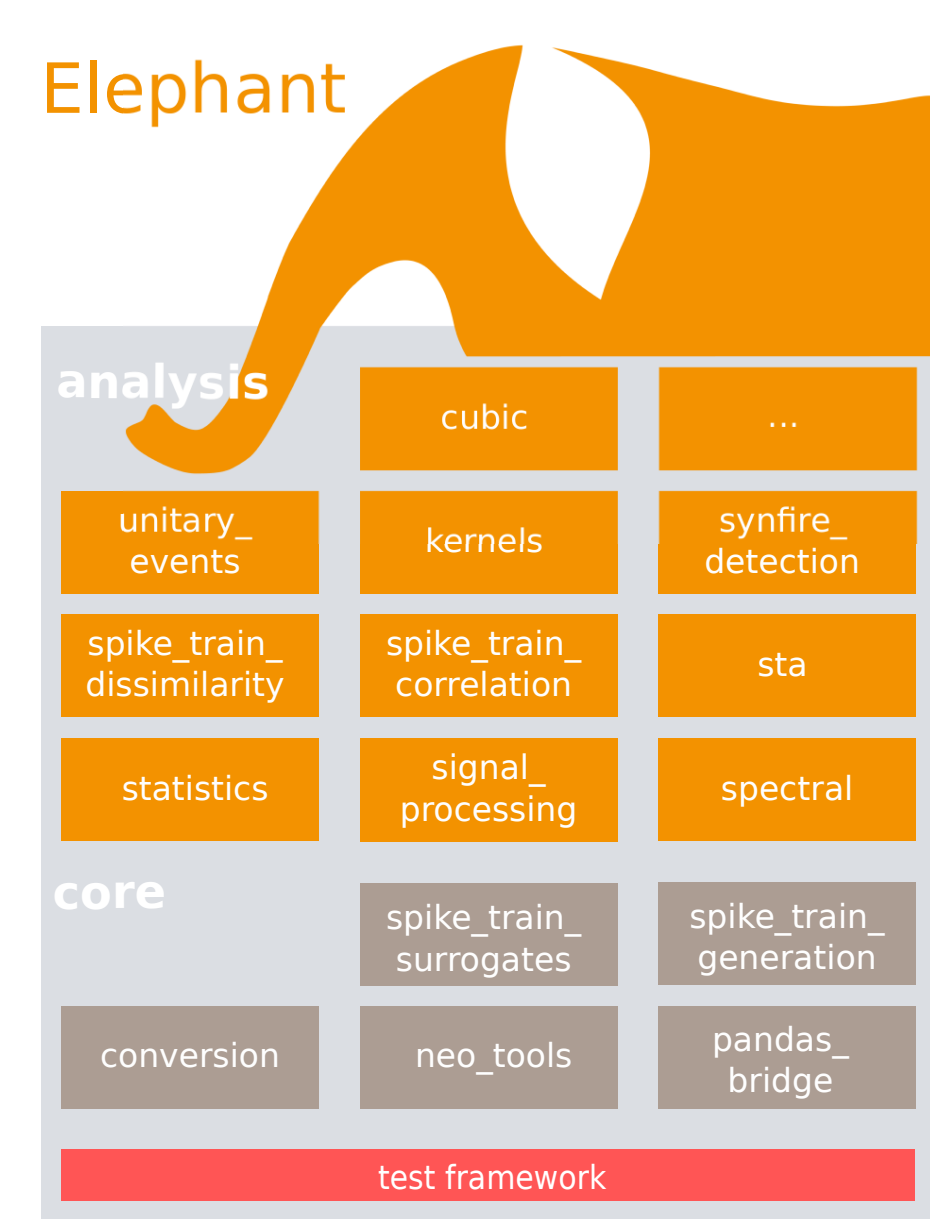


Here, we showcase how **Elephant** is integrated into an analysis workflow running on the **Collaboratory**, reproducing work in [3]. The workflow consists of complementary open-source tools and services [4] for: metadata management (**odML** and **odMLtables**, [5,6,7]), data query (**HBP Knowledgegraph**, [8]), data versioning (**gin**, [9]), data storage (**nix**, [10]), data handling (**Neo**, [11]), and usage of high performance computing (HPC) using **Unicorn** [12]. Finally, we outline how these building blocks, combined with generic tools (e.g., version control systems), can be assembled into formalized workflows to form a **blueprint for performing collaborative work** including access to high-performance computing.

Analysis using Elephant

Elephant is a community-centered, open-source software package that provides components for the analysis of multi-scale electrophysiological data (e.g., spike trains, local field potentials) from experiments and neuronal simulations, focusing on:

- ▶ methods for the analysis of parallel recordings
- ▶ correlative features of brain dynamics
- ▶ bridging different scales of observation



Summary

The presented analysis workflow...

- ▶ ... **combines** several public, community-centered software tools to achieve a reproducible analysis.
- ▶ ... is suitable for **collaborative work** between laboratories by use of the **HBP Collaboratory**.
- ▶ ... provides a comprehensible data flow **independent of the data format** using the **Neo** library.
- ▶ ... leads the way towards the implementation of **future analysis workflows** based on the **Elephant** library.



Find further resources:

<http://python-elephant.org>
<https://github.com/NeuralEnsemble/elephant>

References

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- [6] Zehl, L. et al. (2016). Handling metadata in a neurophysiology laboratory. Frontiers in Neuroinformatics, 10, p.26.
- [7] <https://github.com/INM-6/python-odmltables>
- [8] <https://www.humanbrainproject.eu/en/explore-the-brain/search>
- [9] <https://web.gin.g-node.org>
- [10] <https://github.com/G-Node/nix/wiki>
- [11] Garcia S, et al. (2014). Neo: an object model for handling electrophysiology data in multiple formats. Frontiers in Neuroinformatics, 8, 10. DOI:10.3389/fninf.2014.0001
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- [13] <https://collab.humanbrainproject.eu/#/collab/5185>
- [14] Brochier, T. et al. "Massively parallel recordings in macaque motor cortex during an instructed delayed reach-to-grasp task." Scientific data 5 (2018): 180055. DOI:10.1038/sdata.2018.55
- [15] https://web.gin.g-node.org/INT/multielectrode_grasp
- [16] Gutzen et al. (sub.) "Reproducible neural network simulations: statistical methods for model validation on the level of network activity data"
- [17] <https://github.com/INM-6/NetworkUnit>
- [18] <https://github.com/scidash/scinuit>

Analysis workflow on the HBP Collaboratory

The **Collaboratory** infrastructure of the Human Brain Project hosts the overall workflow of the project, required tools, central data management, data search, HPC access, and the ability for interactive work.

The Collab shown here [13] reproduces main findings of [3] using a Collab-centric workflow. The work concerns the analysis of spatio-temporal beta activity in motor cortex of macaque during a reach-to-grasp task [14].

The core of the project consists of a **Jupyter notebook** in the Collab, an associated **git repository** at github, and the datasets stored in an external **git-annex** based data store.

Jupyter notebook in the Collaboratory:

- Central control instance to orchestrate and display the analysis workflow
- Construction of the environment
- Visualization of the results
- Documentation of the analysis process

Git repository:

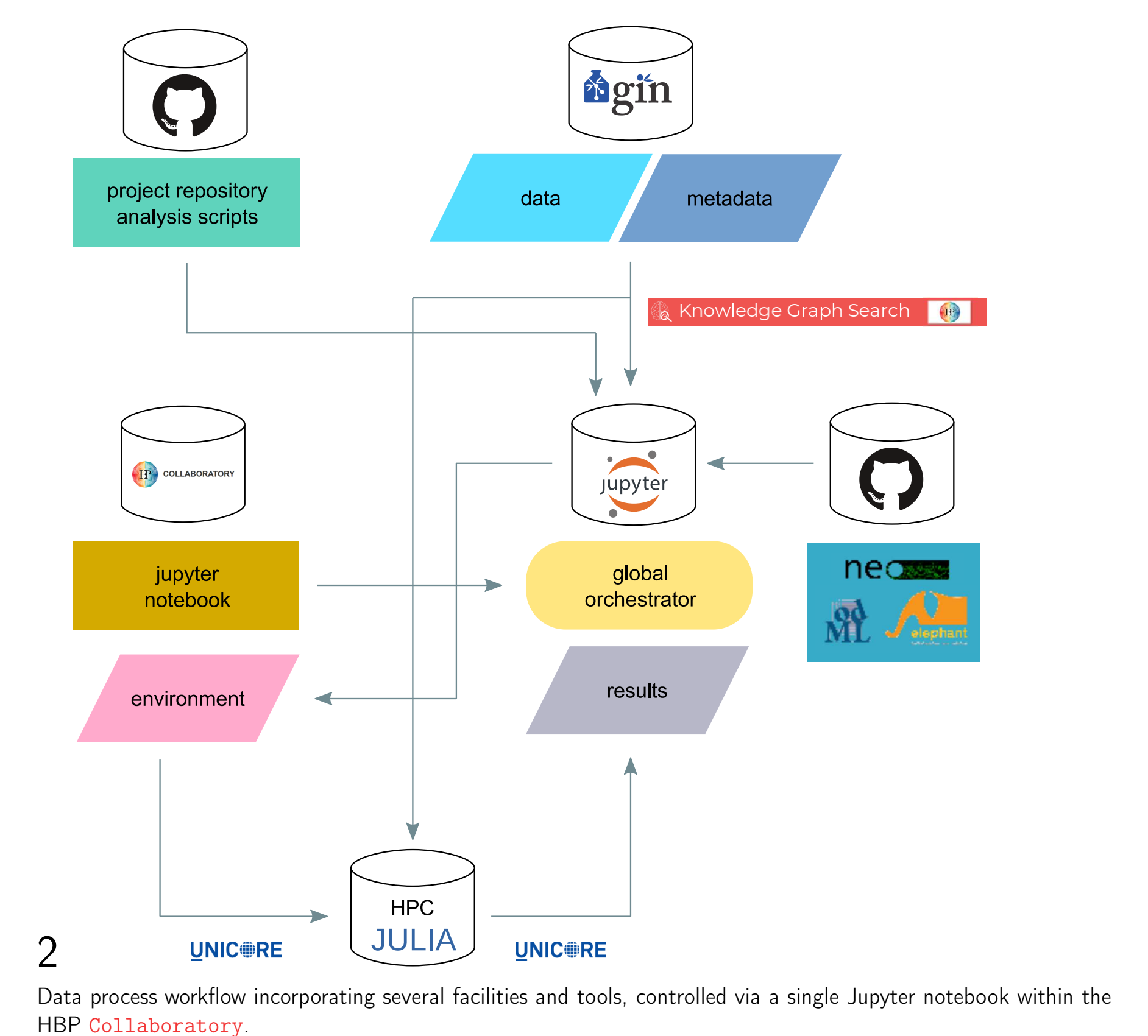
- Analysis scripts are developed off-line and retained, version-controlled and shared in a git repository
- Jupyter notebook synchronizes the repository to the Jupyter notebook container
- Repository contains dependency file for software installations and data requirements

Unicorn:

- Delegates analysis to an HPC resource (JULIA)
- Stage-in installs all software requirements
- Transfers environment to JULIA via Collab storage
- Executes the corresponding scripts on JULIA
- Transfers results back to the Jupyter notebook

Elephant/Neo:

- Base libraries for performing analysis work
- Analysis functions are chained together in the project analysis scripts
- Data is read via the blackrockIO file connector provided by the **Neo** library



odML/odMLtables:

- Metadata is read from associated **odML** file
- Selected metadata are used as annotations to the created **Neo** object

gin:

- Experimental data are published on the **gin** repository under version control

Knowledgegraph:

- Data are registered and located using the **HBP Knowledgegraph** service and the **pyxus** API

Validation with NetworkUnit

Validation is the process of establishing confidence in a model by quantitatively testing whether its prediction accuracy is within an acceptable agreement to its system of interest.

Network-level validation evaluates the model simulation on the level of the network activity as opposed to the complementary approach of validating on a single-cell level.

Model-to-model validation compares models (or their implementations) for consistency, cross-validation, simulator evaluation, or quantification of model developments. [16]

The Python module **NetworkUnit** [17] is based on the **SciUnit** [18] and **Elephant**, and provides a formalized framework as well as a battery of standardized tests for network-level validation.

- ▶ Models are matched to appropriate tests via 'capabilities'.
- ▶ New or variation of tests can be easily derived from a range of base tests.
- ▶ Tests can be adapted to also compare multiple models.
- ▶ Test scores are annotated with their provenance.

Standardization → Reproducibility

Modularization → Versatility

Formalization → Understandability

