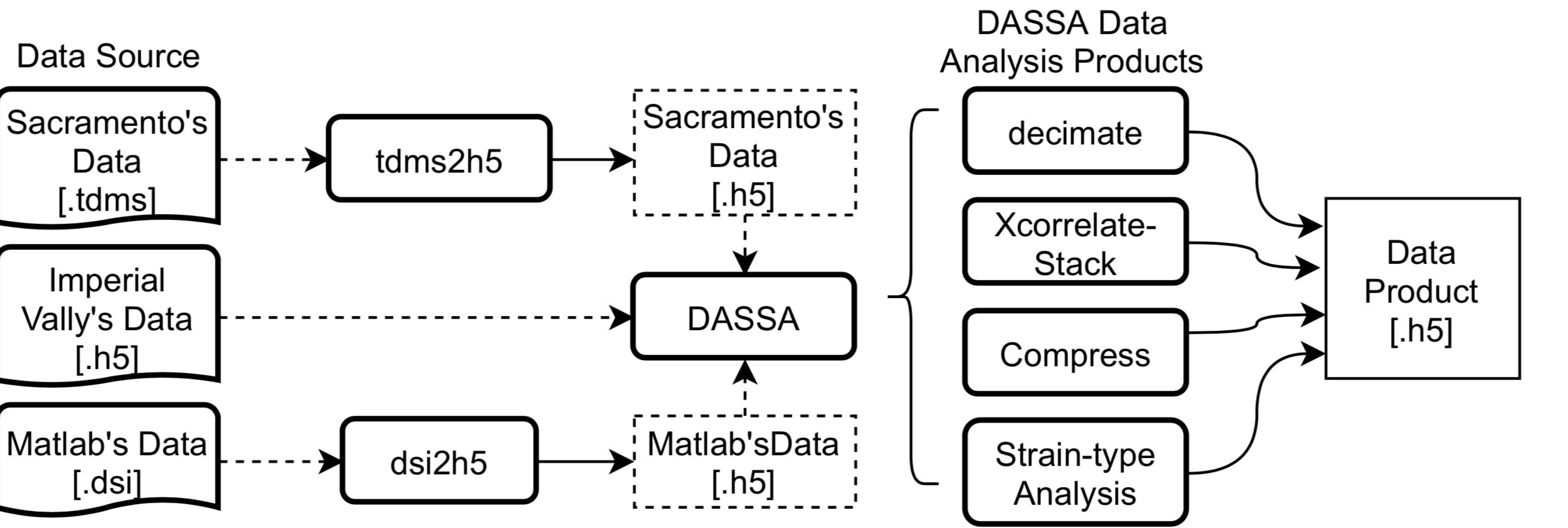


# Motivation

- A need for data provenance in HPC scientific workflows
  - Workflow & scientific data reusability
  - Scientific data lineage tracking
  - Workflow optimization
- Challenges
  - Little **quantitative** understanding or specification of scientists' diverse needs
  - The variety of the workflow characteristics (e.g., different I/O interfaces and file formats)

**Example: DASSA, an acoustic sensing workflow [1].** DASSA takes input dataset in multiple different formats (i.e., ‘tdms’, ‘h5’ and ‘dsi’) and uniformly converts them to intermediate ‘h5’ files in its first step. It then processes the intermediate data with four different data analysis products and gives the final data product. In DASSA workflow, the scientists want to know, given a final data product, what’s the original input data (backward lineage)



- A gap between the HPC data provenance and existing solutions
  - Vague provenance model
  - Incompatible provenance software on HPC environment
  - Limited transparency

Software	Type	Base Model	Environment	Language	Transparent?
Karma	Middleware	OPM	Cloud	Java	Yes
Komadu	Provenance system	W3C PROV-DM	Cloud	Java	No
Tanerna	Workflow system	W3C PROV-DM	Desktop	Java	Hybrid
ProvLake	Provenance system	W3C PROV-DM	Cloud	Python	No

Table 1: Existing Provenance Software for Scientific Workflows

# Methodology

- Case studies on realistic workflows and the end users' provenance needs
- A provenance model for HPC data provenance
- A provenance framework
  - Provenance Tracking for capturing diverse I/O operations
  - Provenance Storage for persisting the captured provenance information as standard RDF triples [2]
  - User Engine for querying and visualizing provenance

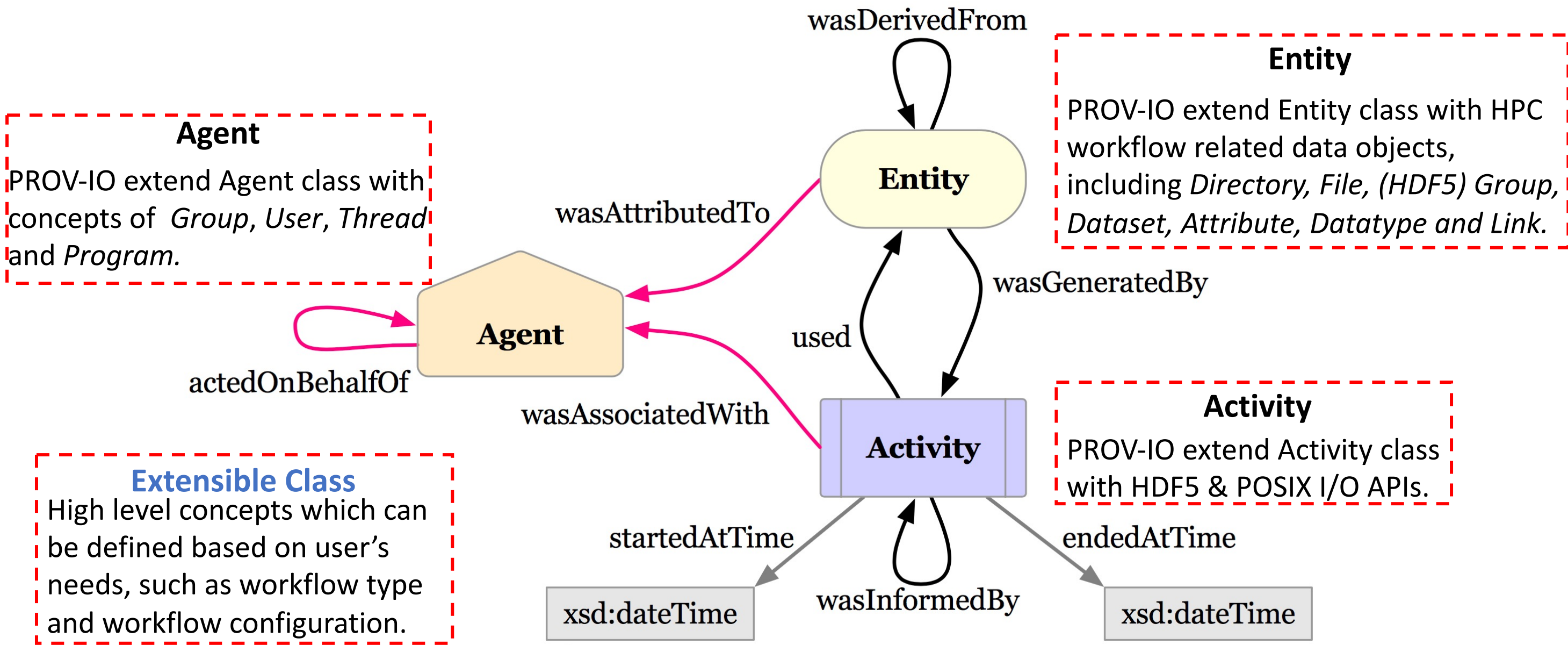
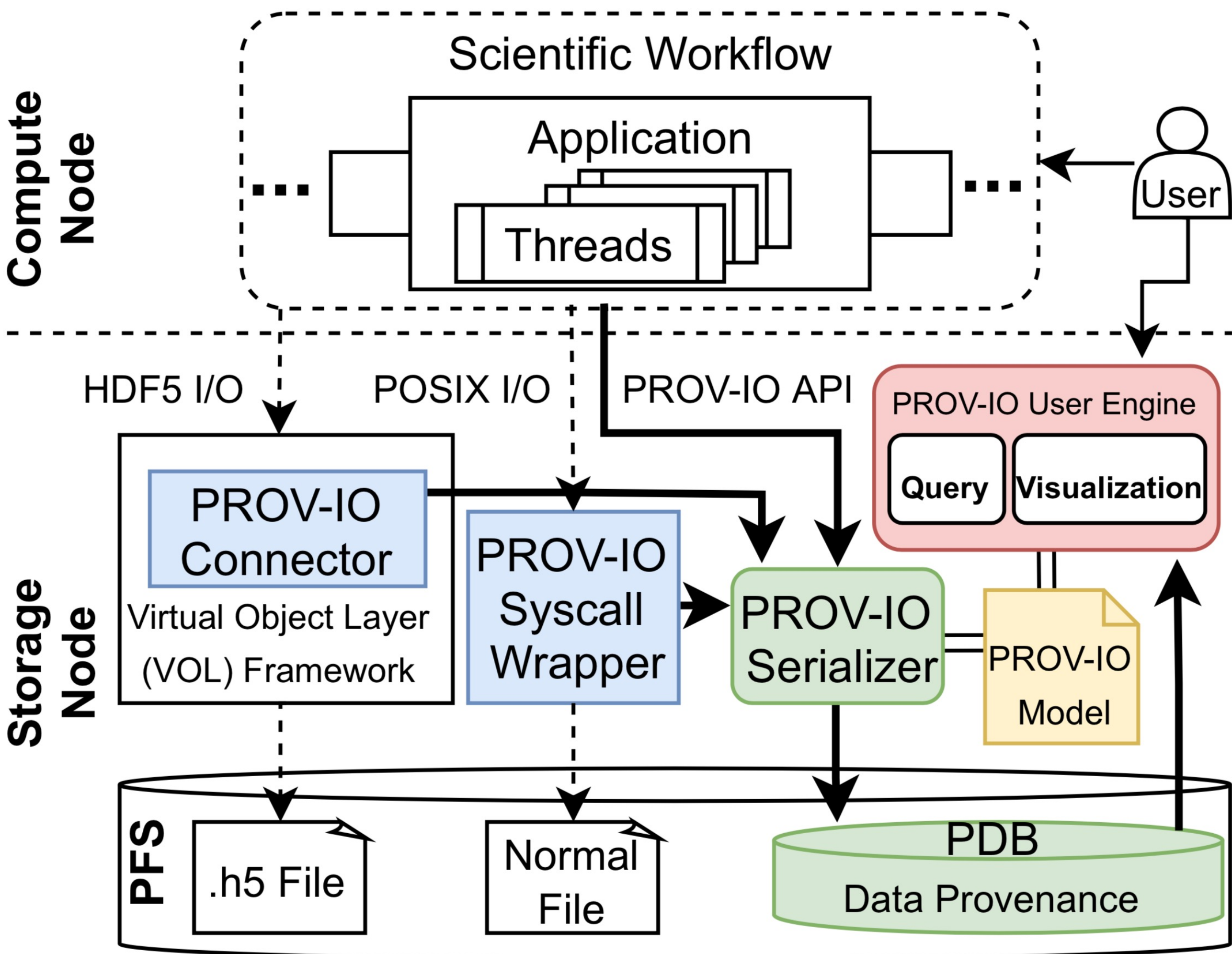
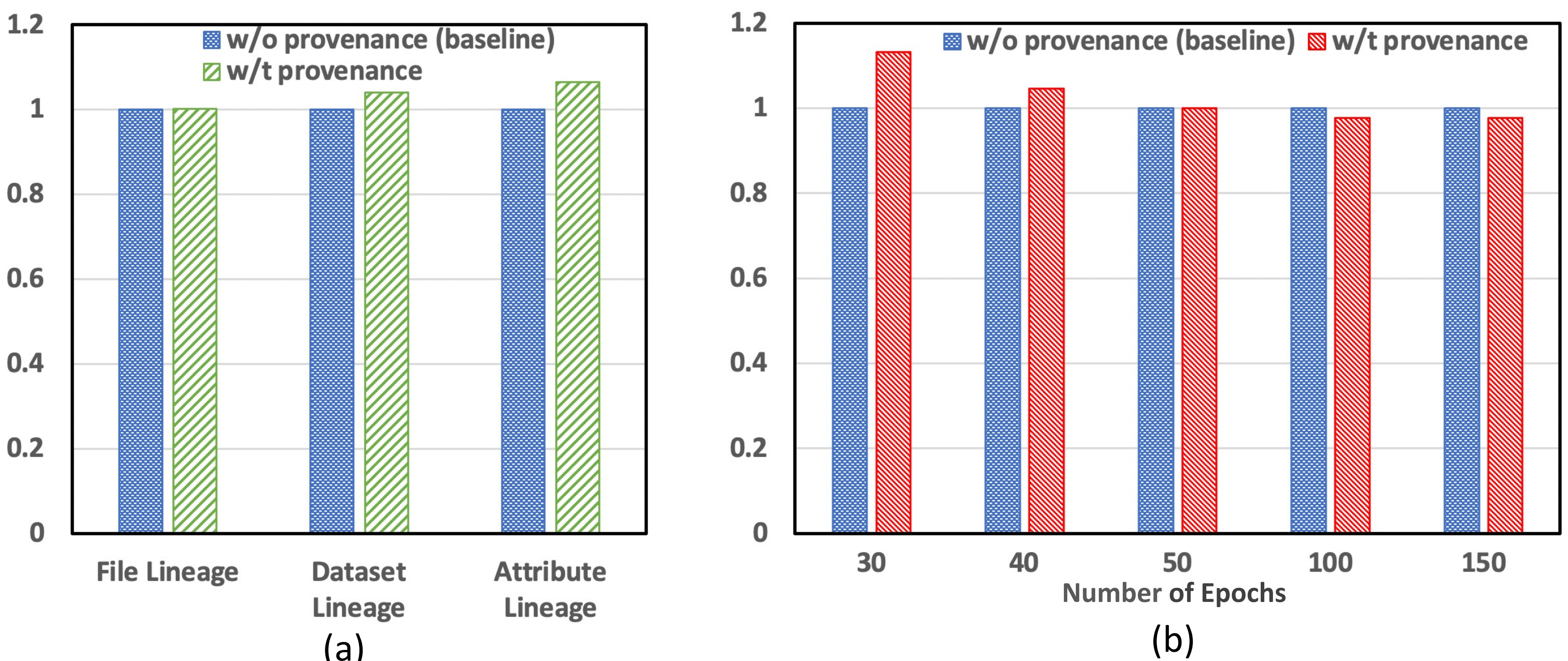


Figure 1: PROV-IO Model. An extension of W3C PROV-DM



**Figure 2: PROV-IO framework.** The scientific workflow’s I/O operations (e.g. , HDF5 I/O and POSIX I/O) from the compute node are redirected to PROV-IO framework on the storage node. PROV-IO lets I/O calls pass through and only captures provenance information. It then serializes provenance to the provenance database (PDB) which can be queried and visualized

# Preliminary Results



**Figure 3: Performance of Provenance Tracking.** (a) Tracking performance of DASSA. (b) Tracking performance of Top Reco. In the two figures, tracking performance is measured with normalized workflow completion time

- A maximum provenance tracking overhead of 5.2% In DASSA workflow, with the most fine-grained provenance level enabled (HDF5 Attribute Lineage)
- A maximum provenance tracking overhead of 13.2% in Top Reco workflow, with the training epoch of 30. The tracking overhead becomes negligible when applying more training epochs

# Future Work

- Fully implement PROV-IO Syscall Wrapper
- Improve PDB & User engine
  - Optimization existing database solutions and PROV-IO user APIs

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# References

[1] Bin Dong, Verónica R. Tribaldos, Xin Xing, Suren Byna, Jonathan Ajo-Franklin, Kesheng Wu. DASSA: Parallel DAS Data Storage and Analysis for Subsurface Event Detection. In Proceedings of IEEE International Parallel and Distributed Processing Symposium (IPDPS), 2020.

[2] Resource description framework. <https://www.w3.org/RDF/>