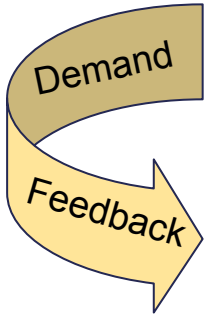
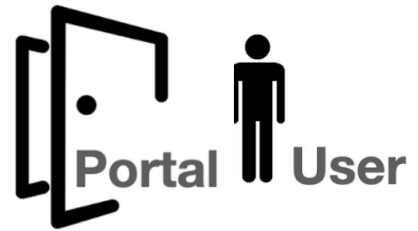
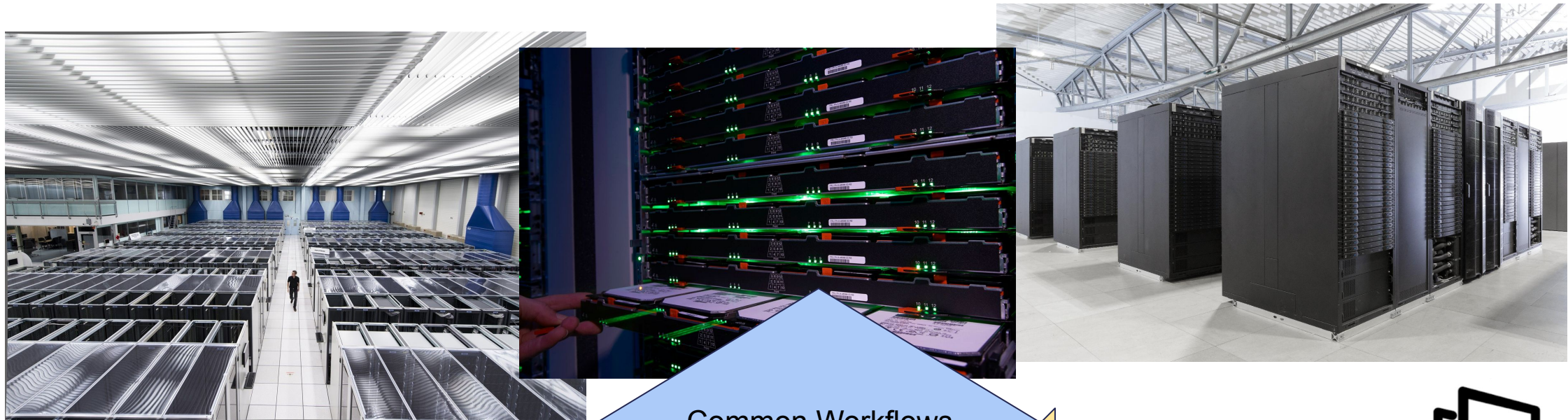


Task Area 3: Data Transformations

Provision of tools for parallel processing of huge datasets on heterogeneous computing resources



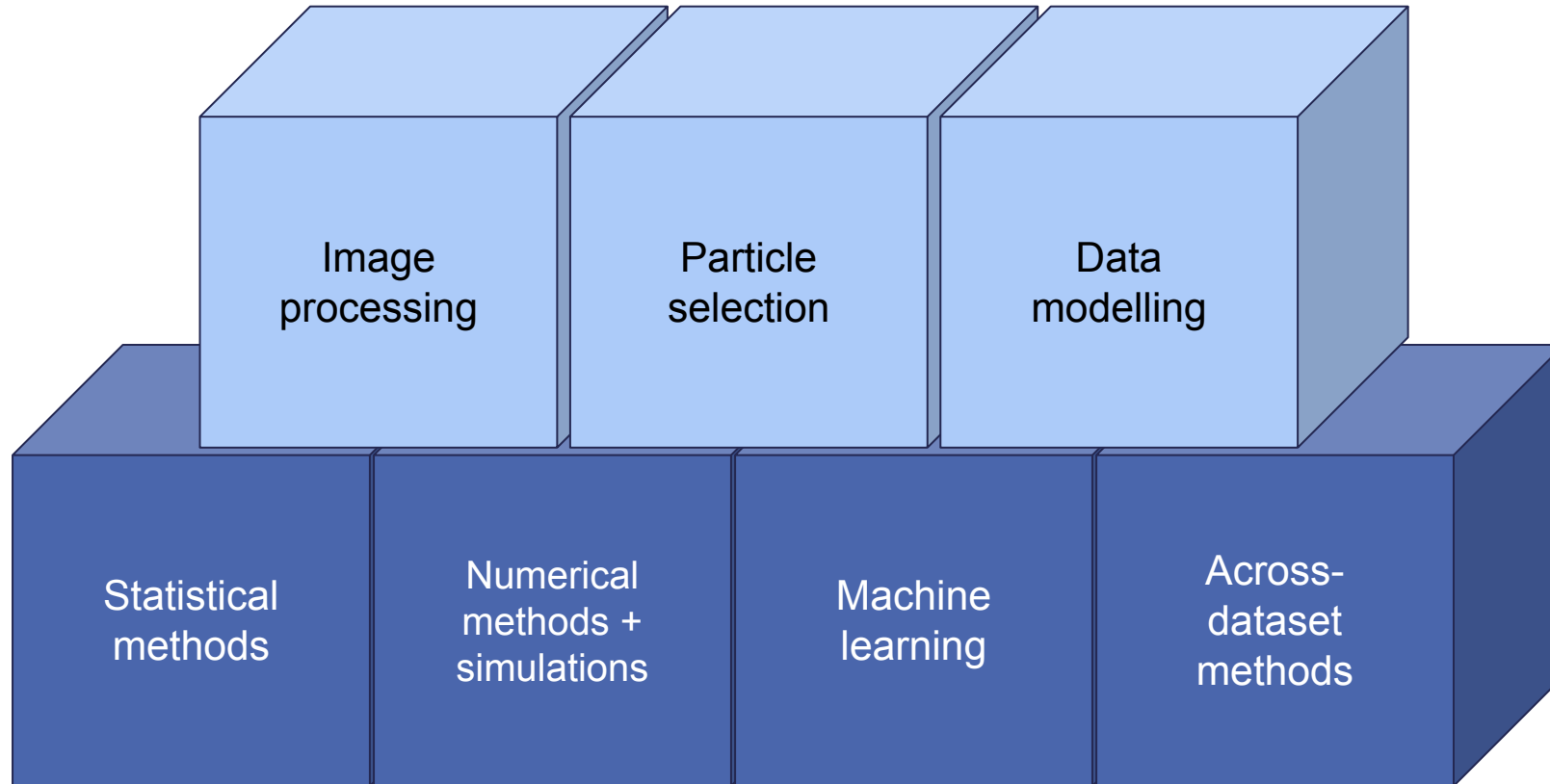
External Tools

Own Tools

New Tools

TA 3: Common Tools

Commonalities among multiple science fields



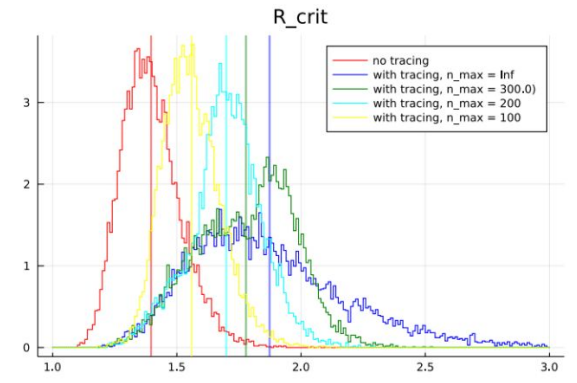
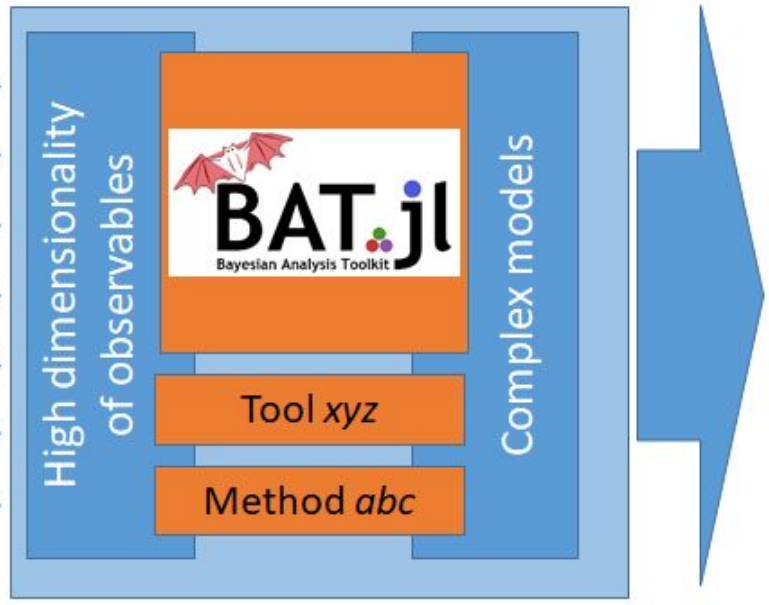
TA 3 / WP1: Statistical methods

Methods and tools for analyzing large data sets and complex models

Large raw-data sets

- Astro
- Particle
- Astroparticle
- Hadron and nuclear
- ...
- Physics
- ...

Statistical tools



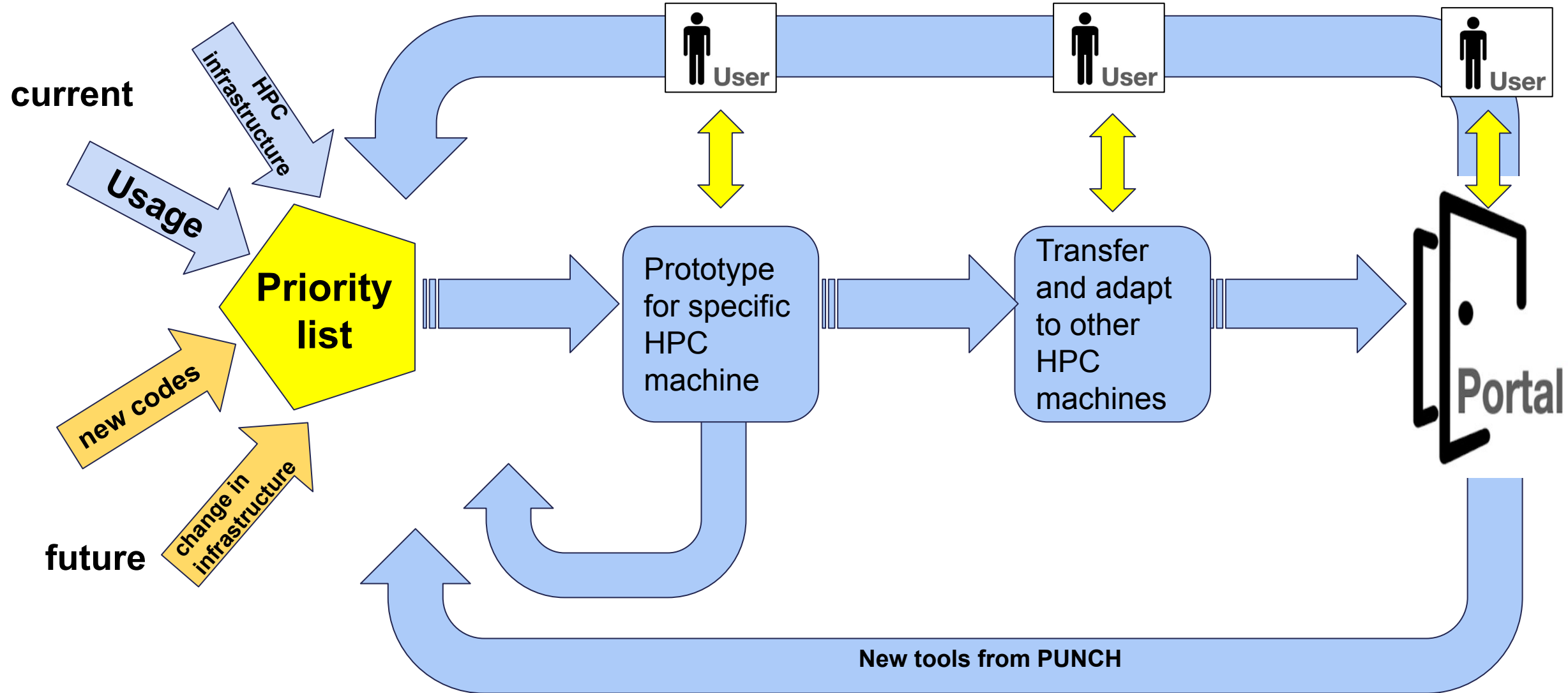
Interpretable and visualized results

Deliverables:

- **D-TA3-WP1-1 (30 Sep 2026):** Statistical inference in the limit of large datasets and highly parallel computing.
- **D-TA3-WP1-2 (30 Sep 2026):** Integration of a broad set of statistical methods; further development of a subset of methods into a common set of cross-community tools.

TA 3 / WP2: Numerical Methods and Simulations

Provision of tools optimized for simulations on heterogeneous computing resources



TA 3 / WP3: Machine Learning

Automated tools for machine learning on large datasets

Machine learning is a transformative technology, showing and promising gains for many aspects of PUNCH science

WP3 will focus on two aspects:

WP3.1: Framework for AutoML on scientific data based on the PUNCH domain:

- Fully automate pre-processing and training workflows to develop machine learning as a service for scientific data:
 - Find and benchmark algorithms on datasets from different domains
 - Transfer learning based on successful architectures
 - Automated model selection and hyperparameter selection
- Previous work: Physics Data for Machine Learning (pd4ml)
<https://github.com/erum-data-idt/pd4ml/>, arXiv:2107.00656

WP3.2: Tools and solutions for distributed learning using very large datasets:

- Scalable solutions for very large datasets:
 - efficient parallel training on partitions depending on distribution of data
 - combination (ensembling) of classifiers

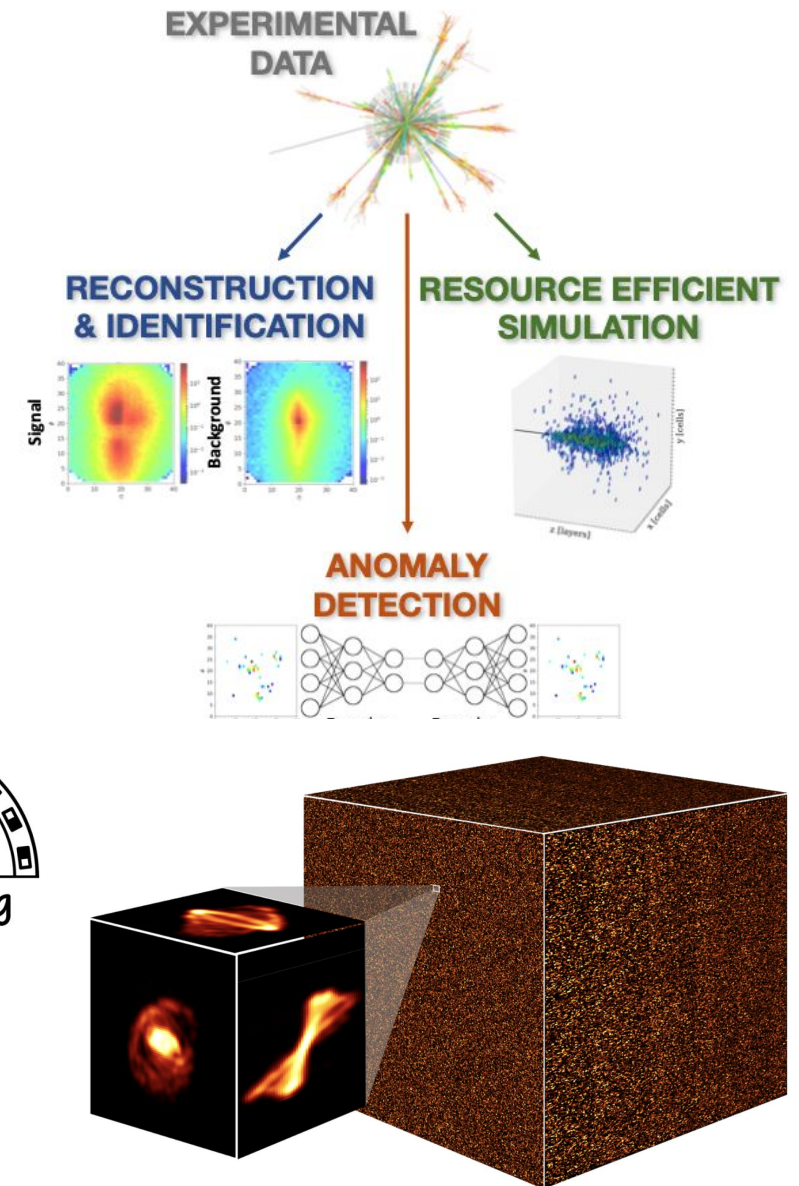


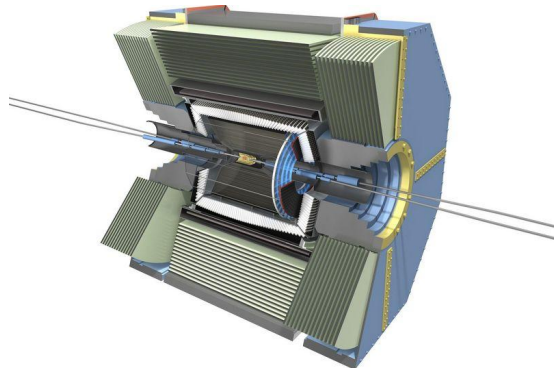
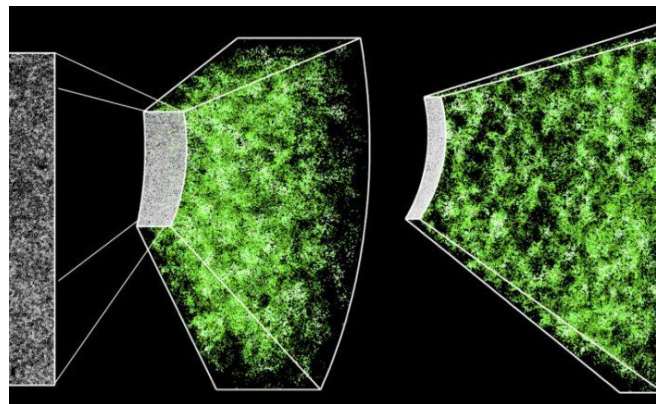
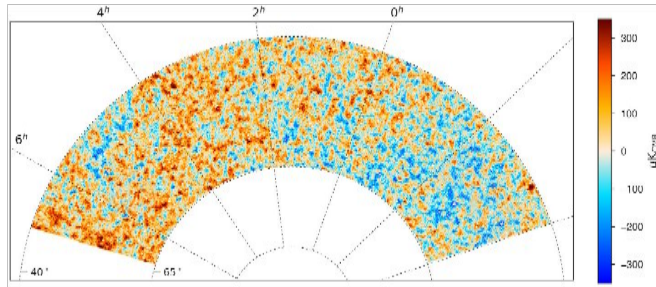
Illustration of ongoing machine learning developments

<https://sdc2.astronomers.skatelescope.org/>

The simulated datacube, before noise and instrumental effects are added. Covering a sky area of 20 square degrees and featuring nearly a quarter of a million galaxies, the cube represents an SKA observation of neutral hydrogen – or "HI" – emission.

TA 3 / WP4: Analyses Across Datasets

Methods for exploiting the full potential of data from multiple sources



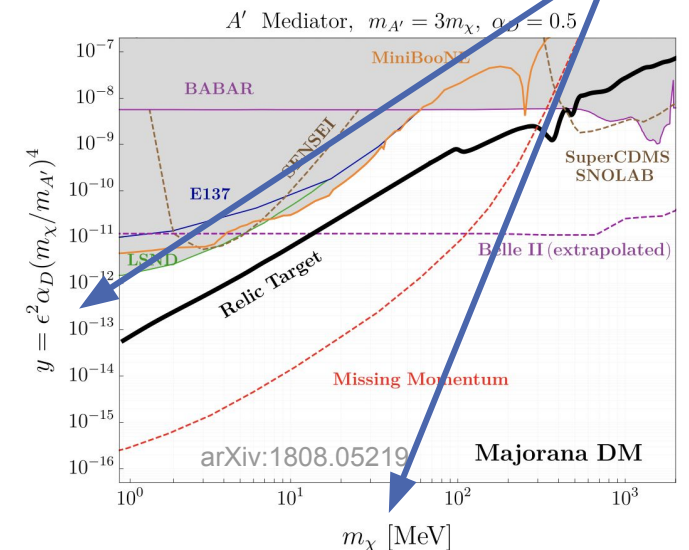
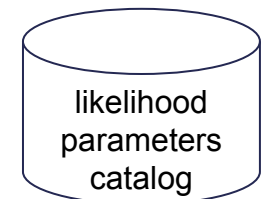
Enable joint analyses at pixel/event level

Transformation from low to high level data

Common analysis framework

- Data format converters
- Tools to support workflows executing across multiple archive nodes
- Library of workflow templates

Common definition of parameters



Common likelihood interface to implementations as

- Function
- Histogram
- MCMC
- Calculated on demand from data

TA 3: Organization

TA3:

- Monthly meetings on first Tuesday at 9:00
- punch4nfdi-ta3@desy.de
- Contact: Thomas.Kuhr@lmu.de, mbrueggen@hs.uni-hamburg.de

WP1 (statistical methods):

- punch4nfdi-ta3-wp1@desy.de
- Contact: kevin.kroeninger@tu-dortmund.de, Joseph.Mohr@physik.lmu.de

WP2 (numerical methods and simulations):

- punch4nfdi-ta3-wp2@desy.de
- Contact: s.pfalzner@fz-juelich.de, tilo.wettig@ur.de

WP3 (machine learning):

- punch4nfdi-ta3-wp3@desy.de
- Contact: gregor.kasieczka@cern.ch, mbrueggen@hs.uni-hamburg.de

WP4 (methods for analyses across datasets):

- punch4nfdi-ta3-wp4@desy.de
- Contact: Joseph.Mohr@Physik.lmu.de, Thomas.Kuhr@lmu.de

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