

E-TASC 2nd Meeting - Garching

A Scalable Production Framework for Multimachine Databases

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This work has been carried out within the framework of the EUROfusion Consortium, funded by the European Union via the Euratom Research and Training Programme (Grant Agreement No 101052200 — EUROfusion). Views and opinions expressed are however those of the author(s) only and do not necessarily reflect those of the European Union or the European Commission. Neither the European Union nor the European Commission can be held responsible for them.



Multimachine Database: a challenge

Fusion experiments generate vast, heterogeneous, multi-modal data across devices with different diagnostics, naming conventions, and legacy systems.

This fragmentation is a **major bottleneck** for:

Downstream tasks — AI/ML training, control-related tasks, simulations, etc.

Scientific exploitation — Cross-machine statistics, scaling laws, extrapolation to ITER

Reproducibility — Provenance tracking, versioned workflows, FAIR compliance

Collaborative validation — Expert consensus at scale, standardized labeling and metadata templates

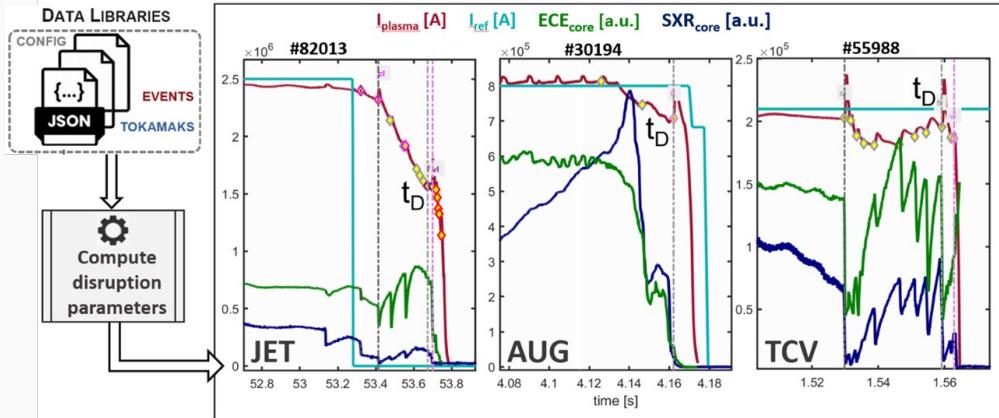
What the Community Needs

- Standardized data curation tools — not ad hoc per-device processing
- Interoperable data mapping tools
- Scalable infrastructure for heterogeneous data at production scale
- End-to-end provenance and metadata for reproducibility
- Tools for collaborative validation orchestrated across a mixture of physics experts
- AI-ready labeled datasets with quantified uncertainty



Multi-machine databases, main use cases

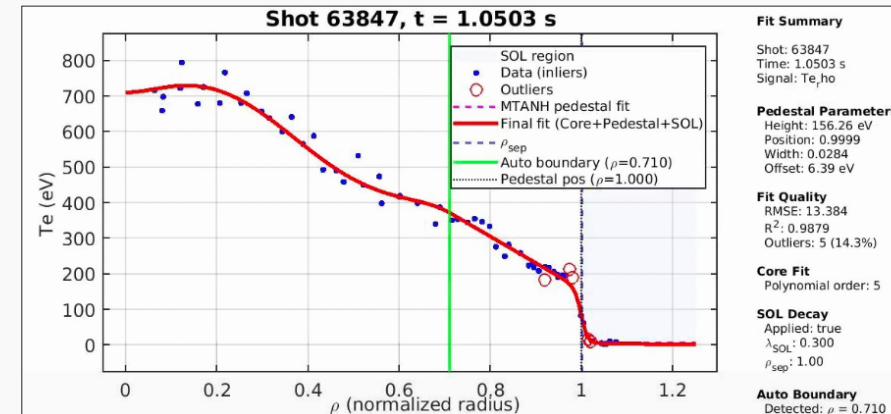
Disruption Database



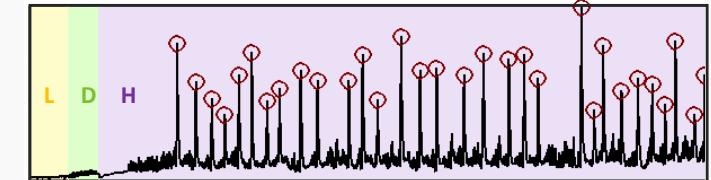
Validated multi-machine disruption database covering TCV, JET-ILW, AUG, MAST ~ several thousands of disruptions validated.



Pedestal database



L-H transition database



Conf. state database of ~1k discharges including JET & TCV experiments. Data-driven models with calibrated uncertainty (JET-PPF)

<https://zenodo.org/records/14996664>

<https://zenodo.org/records/16631053>

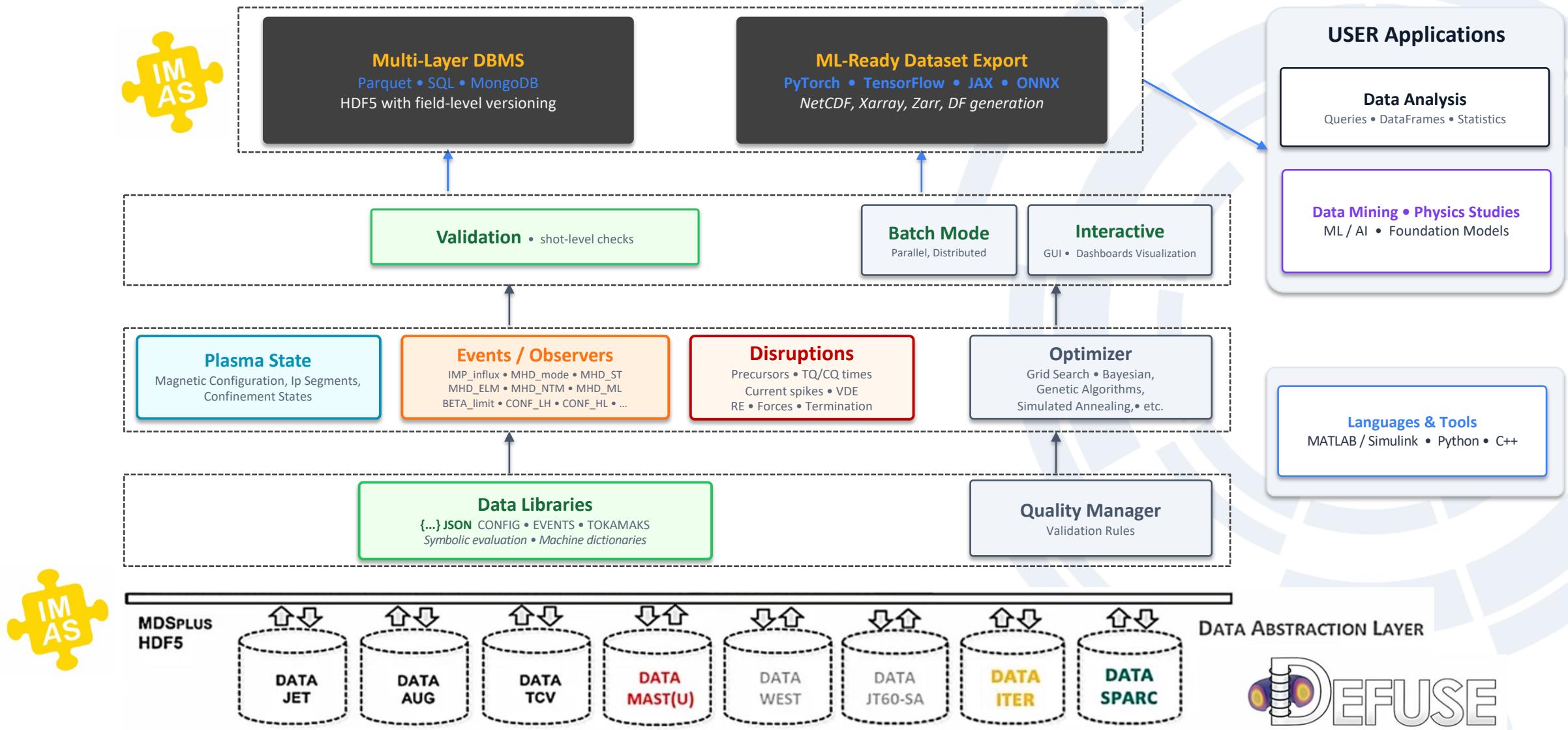
Validation for EUROfusion WPTE

campaigns & IMAS-ready for
Gateway storage.

Automated initial fit and outlier
rejection based on Gaussian Process
Regression (GPR) and principled
Uncertainty Quantification (UQ)



DEFUSE a scalable framework for fusion experiment





Data libraries in DEFUSE



TOKAMAKS

```
► AUG {4}
► JET {5}
► MAST {5}
▼ TCV {5}
  ► DATA_dictionary {333}
  ► DATA_processing {5}
  ► DIAG_dictionary {22}
  Data_access : local
  Data_format : hdf5
► Dictionaries_info {6}
► Table_DATA {35}
► Release {5}
version : 20
```

- **Template DATA dictionary**
main source reference for SQL & MongoDB
- **Symbolic evaluation**
dynamic and abstract representation and transformation of data.
- **Data mapping**
Custom plugins (IMAS, etc.)

Data Dictionary

```
▼ BETAN {8}
  Name : BETAN
  Format : 1D
  Description : normalized toroidal beta
  Units : a.u.
  ErrorBar : null
  ▼ Origin {3}
    ▼ variables {4}
      0 : BZERO
      1 : a_minor
      2 : IPLA
      3 : BETAT
    ▼ coefficients {1}
      k : 100000000
    ▼ expressions {1}
      0 : k * abs((BETAT ./ IPLA) .* BZERO .* a_minor)
  IMAS : /equilibrium/time_slice(itime)/global_quantities/beta_normal
  MachineID : 3
```



Data libraries in DEFUSE



• Dictionaries and Catalogs

encoding the description of disruption characteristic parameters, off-normal events & causes, and plasma state (main discharge segments)

EVENTS

```
▶ Catalog {27}  
▶ Detectors {4}  
▶ Discharge {8}  
▶ Disruptions {5}  
▶ Release {5}  
version : 2
```

Detectors & Observer

```
Detectors {4}  
  info_tag : dictionary for detector main parameter settings  
  ▶ Dictionary {12}  
  ▶ Catalog_with_dictionary {40}  
    ▶ Fields {37}  
    ▶ DUMMY {12}  
      label : DUMMY  
      machine : JET  
      Type : offNormal  
      ComparisonMethod : Equal  
      Threshold : 1  
      timeRef : null  
      AssertionTime : 1  
      minDelta : 1  
      maxDelta : 1  
      t_offset : 1  
      diag : value  
      model : physics-based
```



Database entries

```
_id: ObjectId('670733328bdb57fe44beb38b')
machine : "'TCV'"
shot_number : "no63847"
► Disruptions : Object
► Discharge : Object
► Events : Object
► Detectors : Object
► validation_TAG : Object
► git_TAG : Object
```

```
_id: ObjectId('670733328bdb57fe44beb38b')
machine : "'TCV'"
shot_number : "no63847"
► Disruptions : Object
  tD : 1.73620009
  ► CQ_rates_tD : Object
  ► CQ_currents_tD : Array (10)
    CS_amplitude_tD : 16.2989712
  ► CQ_delta : Object
  ► CQ_S_delta : Object
  ► CQ_SL_delta : Object
  ► IP_control_loss : Object
  ► TQ : Object
  ► Mitigation : Array (empty)
  ► CS_times : Array (4)
  ► CS_values : Array (4)
  ► CS_amplitudes : Array (4)
  ► VDE : Object
  ► RE : Array (empty)
► Discharge : Object
► Events : Object
► Detectors : Object
► validation_TAG : Object
► git_TAG : Object
```



Database entries

```
_id: ObjectId('670733328bdb57fe44beb38b')
machine : "'TCV'"
shot_number : "no63847"
► Disruptions : Object
► Discharge : Object
► Events : Object
► Detectors : Object
► validation_TAG : Object
► git_TAG : Object
```

```
_id: ObjectId('670733328bdb57fe44beb38b')
machine : "'TCV'"
shot_number : "no63847"
► Disruptions : Object
▼ Discharge : Object
  ▼ IP_segments : Object
    ► Ramp_up : Array (2)
    ► Flat_top : Array (2)
    ► Ramp_down : Array (2)
  □ Flat_top_actual : Array (2)
    0: 0.1051
    1: 1.73880017
  ► Ramp_up_actual : Array (2)
  ► Ramp_down_actual : Array (2)
  ► Mat_phases : Array (4)
    IPmean : 170000
    IPLAmean : -171988.203
    RDrate_mean : -3212493.5
  ▼ MAGN_conf : Object
    ▼ ILSNlow : Object
      time : 0.444000095
    ► SNlowIL : Object
  ▼ CONF_states : Object
    ▼ LD : Object
      time : 0.7065
    ► DH : Object
    ► HD : Object
    ► DL : Object
  ► Events : Object
  ► Detectors : Object
  ► validation_TAG : Object
  ► git_TAG : Object
```



Database entries

```
_id: ObjectId('670733328bdb57fe44beb38b')
machine : "'TCV'"
shot_number : "no63847"
► Disruptions : Object
► Discharge : Object
► Events : Object
► Detectors : Object
► validation_TAG : Object
► git_TAG : Object
```

```
_id: ObjectId('670733328bdb57fe44beb38b')
machine : "'TCV'"
shot_number : "no63847"
► Disruptions : Object
► Discharge : Object
► Events : Object
► Detectors : Object
  ► IP0 : Object
  ► ML1 : Object
  ► VDE0 : Object
  ► RMD0 : Object
  ► RC1 : Object
  ► ST1 : Object
  ▼ ELM0 : Object
    ▼ time : Array (2)
      0: 0.7517
      1: 0.4461
    tag : "TCV_ELM0_runLog_09-Oct-2024 11:58:28"
  ► P_LH0 : Object
  ► P_LH1 : Object
  ► A_RM0 : Object
  ▼ N10 : Object
    ▼ time : Array (2)
      ▼ 0: Array (3)
        0: 0.0500000082
        1: 1.5044
        2: 1.70924
      □ 1: Array (3)
        0: 0.01728
        1: 0.07454
        2: 0.02698
    tag : "TCV_N10_runLog_09-Oct-2024 11:58:28"
  ► validation_TAG : Object
  ► git_TAG : Object
```



Collaborative Environment for Validation

Interactive Tools for the Community



Why Collaborative?

Mixture of physics experts

Validation requires device-specific domain expertise. No single person can label all events accurately — orchestrated collaborative review is essential.

Interactive validation

Automated detectors provide initial labels; experts validate and correct through interactive GUIs and SQL-baked dashboards for advanced visualization.

User annotation & review

Multiple experts contribute annotations on specific discharges. Iterative refinement with versioning ensures long-term database quality and traceability.

Integration with DMP-IMAS catalogues

Web-based dashboards and advanced visualization utilities can be integrated with DMP-IMAS catalogues



Standards & Procedures

Data Curation & labeling at scale

Scalable infrastructure for heterogeneous data at production scale;
Standardized data curation tools — not ad hoc per-device processing

Standardized Metadata

Enforce consistent metadata schemas across devices. Machine descriptions, diagnostic specs, and processing parameters must be interoperable and self-documenting.

Validation Orchestration

Enable orchestrated validation from a mixture of physics experts — device-specific knowledge with cross-machine consistency checks. Not a single-group effort.

Database Governance

Standardize versioning, backup, and maintenance procedures. Long-term sustainability requires clear governance, access policies, and community-driven data stewardship.

Uncertainty & Reproducibility & AI

WUQ & End-to-End Uncertainty

Data-driven models must propagate uncertainty from experimental measurements, through processing pipelines, to final labels and predictions. Calibrated uncertainty is non-negotiable.

Reproducibility in Science

Track full data provenance: from raw signals through processing workflows to curated datasets. Versioned pipelines, atomic workflow snapshots, persistent identifiers (DOIs).

Open & FAIR Science

Open-source tooling, public repositories, standardized outputs. Broad community reuse, citation, and contribution. Lower barriers for new devices and research groups.

AI-assisted workflows

AI-assisted curation, automated pipeline development, and LLM-based tools can dramatically accelerate database development, scientific exploitation, and code generation.



Backup slides



INGEST Automated loaders from MDSplus / HDF5 / IMAS / UDA with caching

VALIDATE Quality assessment: gaps, saturation, outliers, SNR, ...

PROCESS Resampling, rescaling, feature extraction with full provenance

LABEL Automated + expert-validated labels for system states, events, etc.

EXPORT ML-ready dataset and efficient data-loaders

Labeling & Curation Strategy

Hybrid model-assisted labeling and validation

Plasma state and events detection framework (MHD-modes, sawteeth, ELMs, thermal collapses, off-normal events) provides automated labels. Physics-based & data-driven models with configs encoded in data libraries for reproducibility.

Expert validation through GUIs

GUIs for interactive review and correction. Confinement states, L-H transitions, disruption event chains validated by domain experts (automated labeling will never be perfect, but we can flag what is “suspicious”).

Scalable batch execution

Parallel computation with checkpoint-based batch processing. support. Performance: few tens of seconds for full shot characterization.

Standardized output formats

Standardized dataframes with labels. HDF5 with rich metadata. Full traceability from raw signal to final label, including the entire data processing pipeline.



CI/CD Pipelines

GitLab CI/CD with Jenkins integration. Automated build, test (~80% coverage target), deployment. Pre-commit validation blocks regressions.

Multi-Layer DBMS

SQL (SQLite) relational model for IMAS mapping with role-based auth. NoSQL (MongoDB) for events & plasma state. JSON1 for nested structures.

Parallel Processing

Configurable workers for batch execution. Checkpoint-based with merge & recovery. Target: <30s/shot for full disruption characterization.

Multi-Format I/O

HDF5 (IMAS backend), Parquet/Xarray (analytics), NetCDF. Automated table generation: DATA, COORD, LABELS, NORM variants per shot.

Versioning & Provenance

Shot-level validation with rollback & logging. Role-based auth. Field-level provenance tracking. Unit testing framework with regression detection.

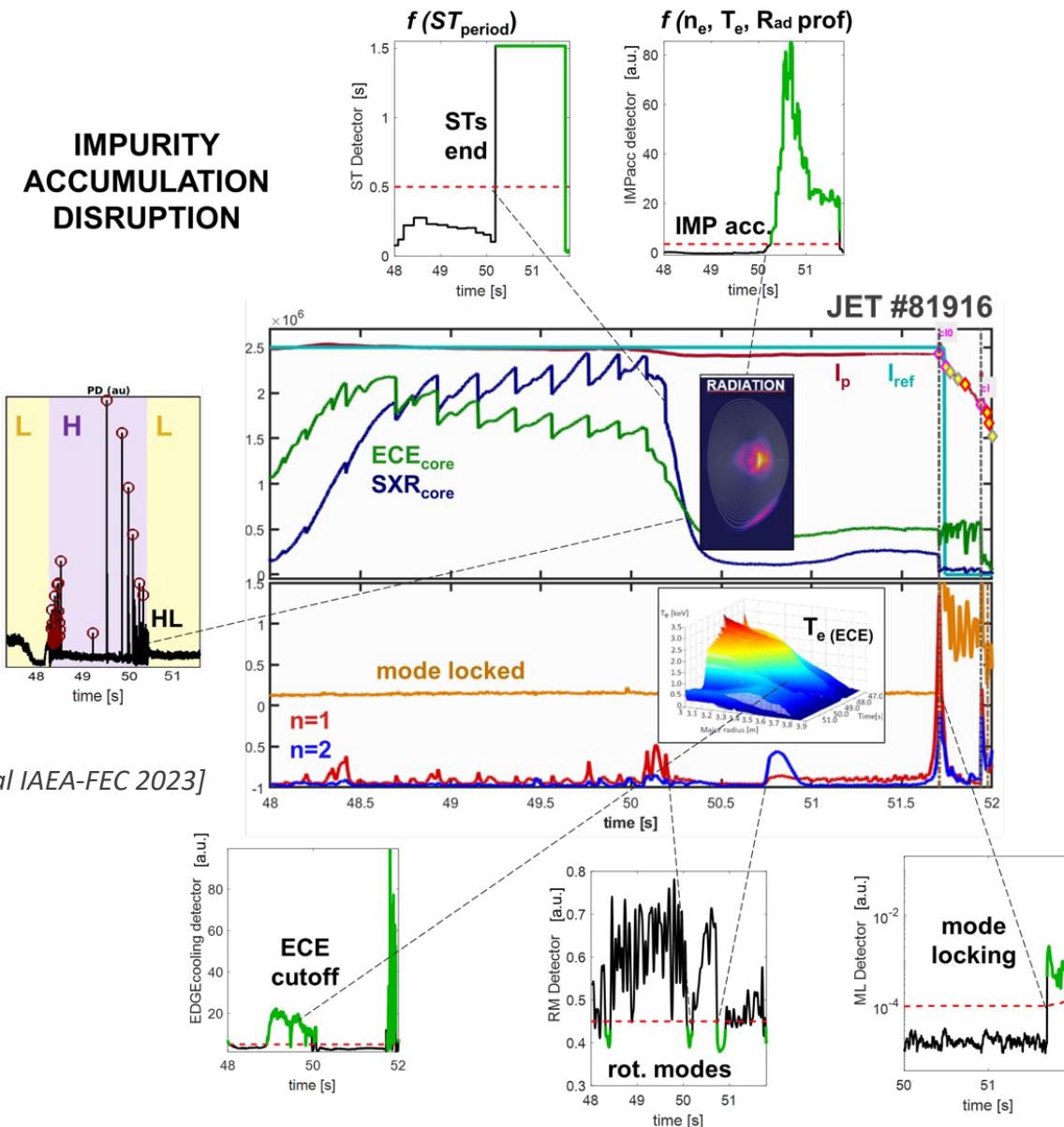
Gateway Migration

Stack migrating to new EUROfusion Gateway. 20–50 TB storage. IMAS-compliant backend. Web service interfaces for collaborative access.

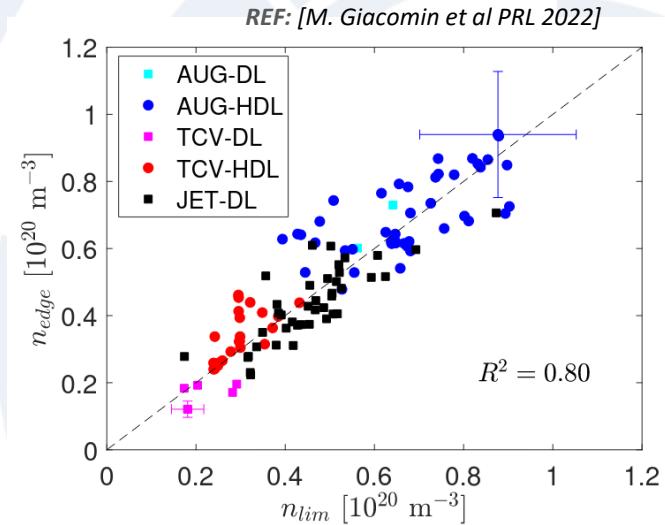


EUROfusion DDB: Events and Modelling

IMPURITY ACCUMULATION DISRUPTION



Density Limit First-Principles Scaling



1. P_{SOL} dependence of n_{eLIM}
2. $n_{eLIM}^{\text{ITER}} \simeq 2 \cdot n_{eGW}^{\text{ITER}}$