

































pynxtools: A Python framework for generating and validating NeXus files in experimental data workflows

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DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

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Editor: 

Submitted: 14 October 2025

Published: unpublished

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Summary

Scientific data across physics, materials science, and materials engineering often lacks adherence to FAIR principles (Barker et al., 2022; Jacobsen et al., 2020; M. D. Wilkinson et al., 2016; S. R. Wilkinson et al., 2025) due to incompatible instrument-specific formats and diverse standardization practices. pynxtools is a Python software development framework with a command line interface (CLI) that standardizes data conversion for scientific experiments in materials science to the NeXus format (Klosowski et al., 1997; Könnicke, 2006; Könnicke et al., 2015) across diverse scientific domains. NeXus defines data storage specifications for different experimental techniques through application definitions. pynxtools provides a fixed, versioned set of NeXus application definitions that ensures convergence and alignment in data specifications across, among others, atom probe tomography, electron microscopy, optical spectroscopy, photoemission spectroscopy, scanning probe microscopy, and X-ray diffraction. Through its modular plugin architecture pynxtools provides conversion of data and metadata from instruments and electronic lab notebooks to these unified definitions, while performing validation to ensure data correctness and NeXus compliance. pynxtools can be integrated directly into Research Data Management Systems (RDMS) to facilitate parsing and normalization. We detail one example for the RDM system NOMAD. By simplifying the adoption of NeXus, the framework enables true data interoperability and FAIR data management across multiple experimental techniques.

Statement of need

Achieving FAIR (Findable, Accessible, Interoperable, and Reproducible) data principles in experimental physics and materials science requires consistent implementation of standardized

data formats. NeXus provides comprehensive data specifications for structured storage of scientific data. `pynxtools` simplifies the use of NeXus for developers and researchers by providing guided workflows and automated validation to ensure complete compliance. Existing solutions (Jemian et al., 2025; Könnicke et al., 2024) provide individual capabilities, but none offers a comprehensive end-to-end workflow for proper NeXus adoption. `pynxtools` addresses this critical gap by providing a framework that enforces complete NeXus application definition compliance through automated validation, detailed error reporting for missing required data points, and clear implementation pathways via configuration files and extensible plugins. This approach transforms NeXus from a complex specification into a practical solution, enabling researchers to achieve true data interoperability without deep technical expertise in the underlying standards.

Dataconverter and validation

The dataconverter, core module of `pynxtools`, combines instrument output files and data from electronic lab notebooks into NeXus-compliant HDF5 files. The converter performs three key operations: extracting experimental data through specialized readers, validating against NeXus application definitions to ensure compliance with existence and format constraints, and writing valid NeXus/HDF5 output files.

The dataconverter provides a command-line interface (CLI) for generating NeXus files, supporting both built-in readers for general-purpose functionality and technique-specific reader plugins, which are distributed as separate Python packages.

For developers, the dataconverter provides an abstract reader class for building plugins that process experiment-specific formats and populate the NeXus specification. It passes a `Template`, a subclass of Python's dictionary, to the reader as a form to fill. The `Template` ensures structural compliance with the chosen NeXus application definition and organizes data by NeXus's required, recommended, and optional levels.

The dataconverter validates reader output against the selected NeXus application definition, checking for instances of required concepts, complex dependencies (like inheritance and nested group rules), and data integrity (type, shape, constraints). It validates required concepts, reporting errors for any violations, and issues warnings for invalid data, facilitating reliable and practical NeXus file generation.

All reader plugins are tested using the `pynxtools.testing` suite, which runs automatically via GitHub CI to ensure compatibility with the dataconverter, the NeXus specification, and integration across plugins.

The dataconverter includes `eln_mapper` that creates either a fillable YAML file or a NOMAD (Scheidgen et al., 2023) ELN schema based on a selected NeXus application definition.

NeXus reader and annotator

`read_nexus` enables semantic access to NeXus files by linking data items to NeXus concepts, allowing applications to locate relevant data without hardcoding file paths. It supports concept-based queries that return all data items associated with a specific NeXus vocabulary term. Each data item is annotated by traversing its group path and resolving its corresponding NeXus concept, included inherited definitions.

Items not part of the NeXus schema are explicitly marked as such, aiding in validation and debugging. Targeted documentation of individual data items is supported through path-specific annotation. The tool also identifies and summarizes the file's default plottable data based on the `NXdata` definition.

NOMAD integration

While `pynxtools` works independently, it can also be integrated directly into any Research Data Management System (RDMS). The package works as a plugin within the NOMAD platform (Draxl & Scheffler, 2019; Scheidgen et al., 2023) out of the box. This enables data in the NeXus format to be integrated into NOMAD's metadata model, making it searchable and interoperable with other data from theory and experiment. The plugin consists of several key components (so called entry points):

`pynxtools` extends NOMAD's data schema, known as `MetaInfo` (Ghiringhelli et al., 2017), by integrating NeXus definitions as a NOMAD Schema Package. This integration introduces NeXus-specific quantities and enables interoperability by linking to other standardized data representations within NOMAD. The `dataconverter` is integrated into NOMAD, making the conversion of data to NeXus accessible via the NOMAD GUI. The `dataconverter` also processes manually entered NOMAD ELN data in the conversion.

The NOMAD Parser module in `pynxtools` (`NexusParser`) extracts structured data from NeXus HDF5 files to populate NOMAD with `MetaInfo` object instances as defined by the `pynxtools` schema package. This enables ingestion of NeXus data directly into NOMAD. Parsed data is post-processed using NOMAD's Normalization pipeline. This includes automatic handling of units, linking references (including sample and instrument identifiers defined elsewhere in NOMAD), and populating derived quantities needed for advanced search and visualization.

`pynxtools` contains an integrated Search Application for NeXus data within NOMAD, powered by Elasticsearch (Elasticsearch B.V., 2025). This provides a search dashboard whereby users can efficiently filter uploaded data based on parameters like experiment type, upload timestamp, and domain- and technique-specific quantities. The entire `pynxtools` workflow (conversion, parsing, and normalization) is exemplified in a representative NOMAD Example Upload that is shipped with the package. This example helps new users understand the workflow and serves as a template to adapt the plugin to new NeXus applications.

Funding

The work is funded by the Deutsche Forschungsgemeinschaft (DFG, German Research Foundation) - project 460197019 (FAIRmat).

Acknowledgements

We acknowledge the following software packages our package depends on: `h5py` (Collette & al., 2008), `numpy` (Harris et al., 2020), `click` (The Pallets development team, 2014), `CFF` (Druskat et al., 2021), `xarray` (S. Hoyer & Hamman, 2017), (Stephan Hoyer et al., 2025), `pandas` (The pandas development team, 2020), `scipy` (McKinney, 2010), `lxml` (Behnel et al., 2005), `mergedeep` (Clarke, 2019), Atomic Simulation Environment (Larsen et al., 2017), `pint` (The Pint development team, 2012).

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