

optimade-python-tools: a Python library for serving and consuming materials data via OPTIMADE APIs

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Summary

In recent decades, improvements in algorithms, hardware, and theory have enabled crystalline materials to be studied computationally at the atomistic level with great accuracy and speed. To enable dissemination, reproducibility, and reuse, many digital crystal structure databases have been created and curated, ready for comparison with existing infrastructure that stores structural characterizations (e.g., diffraction) of real crystals. These databases have been made available with bespoke application programming interfaces (APIs) to allow for automated and often open access to the underlying data. Such esoteric and specialized APIs incur maintenance and usability costs upon both the data providers and consumers, who may not be software specialists.

The [OPTIMADE API](#) specification ([Andersen et al., 2020, 2021](#)), released in July 2020, aimed to reduce these costs by designing a common API for use across a consortium of collaborating materials databases and beyond. Whilst based on the robust JSON:API standard ([Katz et al., 2015](#)), the OPTIMADE API specification presents several domain-specific features and requirements that can be tricky to implement for non-specialist teams. The repository presented here, `optimade-python-tools`, provides a modular reference server implementation and a set of associated tools to accelerate the development process for data providers, toolmakers and end-users.

Statement of need

In order to accommodate existing materials database APIs, the OPTIMADE API specification allows for flexibility in the specific data served, but enforces a simple yet domain-specific filter language on well-defined resources. However, this flexibility could be daunting to database providers, likely acting to increase the barrier to hosting an OPTIMADE REST API. `optimade-python-tools` aims to catalyse the creation of APIs from existing and new data sources by providing a configurable and modular reference server implementation for hosting materials data in an OPTIMADE-compliant way. The repository hosts the `optimade` Python package, which leverages the modern Python libraries `pydantic` ([Colvin & others, 2021](#)) and

FastAPI (Ramírez & others, 2021) to specify the data models and API routes defined in the OPTIMADE API specification, additionally providing a schema following the OpenAPI format (Miller et al., 2021). Two storage back-ends are supported out of the box, with full filter support for databases that employ the popular MongoDB or Elasticsearch frameworks.

Functionality

The modular functionality of `optimade` can be broken down by the different stages of a user query to the reference server. Consider the following query URL to an OPTIMADE API, which should filter for any crystal structures in the database with a composition that consists of any three elements in a 1:1:1 ratio:

```
https://example.org/v1/structures?filter=chemical_formula_anonymous="ABC"
```

1. After routing the query to the appropriate `/structures/` endpoint adhering to version v1 of the specification, the filter string `chemical_formula_anonymous="ABC"` is tokenized and parsed into an abstract tree by a `FilterParser` object using the Lark parsing library (Shinan & others, 2021) against the formal grammar defined by the specification.
2. The abstract tree is then transformed by a `FilterTransformer` object into a database query specific to the configured back-end for the server. This transformation can include aliasing and custom transformations such that the underlying database format can be accommodated.
3. The results from the database query are then de-serialized by `EntryResourceMapper` objects into the OPTIMADE-defined data models and finally re-serialized into JSON before being served to the user over HTTP.

Beyond this query functionality, the package also provides:

- A fuzzy implementation validator that performs HTTP queries against remote or local OPTIMADE APIs, with test queries and expected responses generated dynamically based on the data served at the introspective `/info/` endpoint of the API implementation.
- Entry “adapters” that can convert between OPTIMADE-compliant entries and the data models of popular Python libraries used widely in the materials science community: `pymatgen` (Ong et al., 2013), `ASE` (Larsen et al., 2017), `AiiDA` (Huber et al., 2020), and `JARVIS` (Choudhary et al., 2020).

Use cases

The package is currently used in production by three major data providers for materials science data:

- The Materials Project (Jain et al., 2013) uses `optimade-python-tools` alongside their existing API (Ong et al., 2015) and MongoDB database, providing access to highly-curated density-functional theory calculations across all known inorganic materials. `optimade-python-tools` handles filter parsing, database query generation and response validation by running the reference server implementation with minimal configuration.

- 81 ■ NOMAD (Ghiringhelli et al., 2017) uses `optimade-python-tools` as a library to extend
82 its existing web app with OPTIMADE API routes. It uses the Elasticsearch implementa-
83 tion to filter millions of structures from published first-principles calculations provided by
84 users and other projects. NOMAD also uses the filtering module in its own API to expose
85 the OPTIMADE filter language in the user-centric web interface search bar. NOMAD
86 uses a released version of `optimade-python-tools` and all necessary customization
87 can be realized via configuration and sub-classing.
- 88 ■ Materials Cloud (Talirz et al., 2020) uses `optimade-python-tools` as a library to
89 provide an OPTIMADE API entry to archived computational materials studies, created
90 with the AiiDA (Huber et al., 2020) Python framework and published through their
91 archive. In this case, each individual study and archive entry has its own database
92 and separate API entry. The Python classes within the `optimade` package have been
93 extended to make use of AiiDA and its underlying PostgreSQL storage engine.

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