

The MCL-MAP

A platform for accelerated materials design based on active learning

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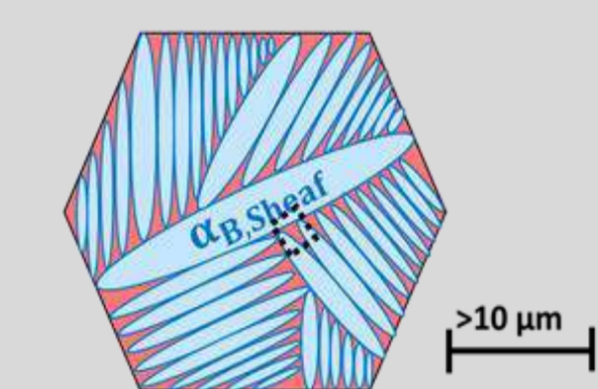
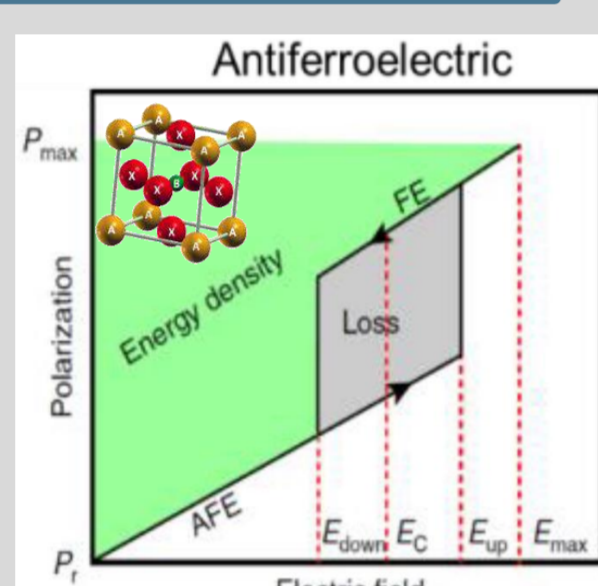
Abstract

The MCL-MAP is a materials acceleration platform (MAP) based at Materials Center Leoben (MCL), aiming at radically new possibilities for optimizing and discovering high-performance materials. This is achieved by integrating physical modeling and machine learning (ML) in a hybrid modeling approach, where process-structure-property relationships are optimized using an Active Learning Loop (ALL). In more detail, the MCL-MAP includes hard- and software for the platform backbone, a database with FAIR curated data, a framework for running physical modeling and Bayesian optimization algorithms, and a large number of software services for tasks such as interactive analytics, data visualization, data exploration and execution of modeling pipelines. Moreover, it includes a tool for integration of literature that enables search, semi-automated extraction of metadata and digitization of data that are then integrated with the original work. MCL-MAP follows a modular architecture implemented on a container-based infrastructure making it easy to extend and adapt it for future developments. As a proof of concept, the MCL-MAP is currently focusing on two use cases: Bainitic steels with optimized mechanical performance and perovskite-based dielectrics with exceptional energy storage capability. It will be available to cover other materials of interest in the future.

Use cases

Perovskite (anti-)ferroelectrics: Optimization of energy density of capacitor materials for power electronics (e-mobility, power conversion, etc.)

Bainitic steels: Increase of strength and elongation at fracture (automotive, rails and train wheels, seismic anchors etc.)



The MCL-MAP

Software services

Interactive analytics & visualization
Data exploration; modeling pipelines,
Literature tool

Framework for physics based
modeling and Bayesian optimization

FAIR database

Modular architecture:

Container-based infrastructure → easy to extend and adapt for future developments

Platform backbone

Platform Architecture

Platform Website

Data management (Metalnx) Literature tool Tools for data analytics Jupyterhub Other tools (ML, etc.)

Platform backbone

Foundational services

Data store (Metalnx) Authentication APIs

Extensible services

Low-level services

iRODS Data Keycloak Auth Kubernetes Docker

Components

Hardware resources

Proxmox

Foundational capabilities

Computation resources Data storage Data bases

FAIR database

- iRODS based data storage, ArrangoDB for metadata.
- Unique identifiers for data, use/development of standards for data and metadata.
- Includes data from literature, material production, characterization and simulation.
- Interoperability with other databases and platforms in progress.

Software services

Jupyterhub: Script-based analysis, visualization, etc.

Web-based user interfaces for interactive analytics and visualization (streamlit).

Modeling and analysis pipelines (Elyra).

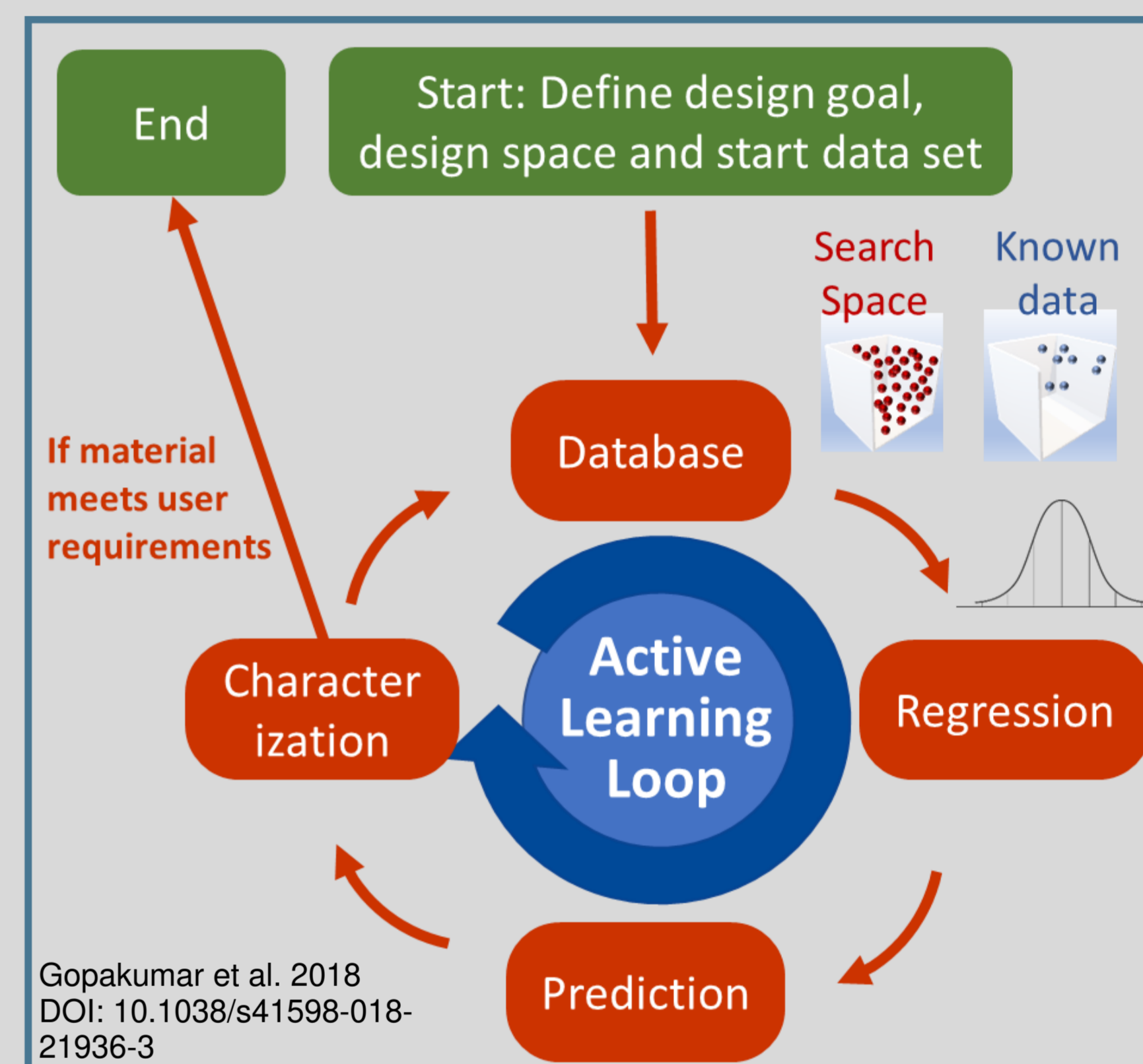
Use case specific analysis tools:

- Bainitic steels: Automated detection of Bainite start temperature from dilatometry; austenite fraction from evaluation of XRD peak ratio and dilatation.
- Perovskite dielectrics: Analysis of Raman spectra (combining exp + DFT)

Literature tool

- Automated keyword search/crawler.
- Semi-automated digitation of figures.
- Digitized data and metadata (chemistry, processing) stored together with original papers.
- Search for data and metadata on same footing as for bibliographic data.

Active Learning Loop (ALL)



Framework for physics based modeling and Bayesian optimization

- The MCL-MAP integrates physics based models, materials data and tools for Bayesian optimization → ecosystem for the ALL

Bainitic steels:

- Multi-dimensional parameter space including composition and heat treatment
- Hybrid model combining physics-based and ML approaches (e.g. Gaussian Process Regression).
- Bayesian inference for choosing best parameters for modeling the bainite start temperature.
- Optimization of Pareto front of yield strength and uniform elongation.

Perovskite (anti-)ferroelectrics:

- Probabilistic hybrid models under development.

Summary

- MCL-MAP enables active-learning based materials optimization.
- Integration of platform backbone and web services, FAIR data, physics-based, ML models.
- It is being developed by MCL, Know Center Graz and Montanuniversität Leoben together with use case specific partners.

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