

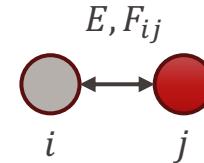
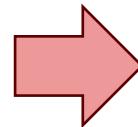
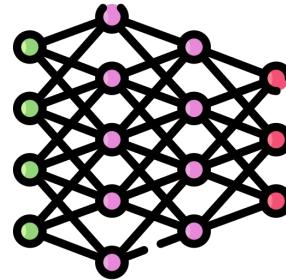
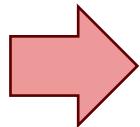
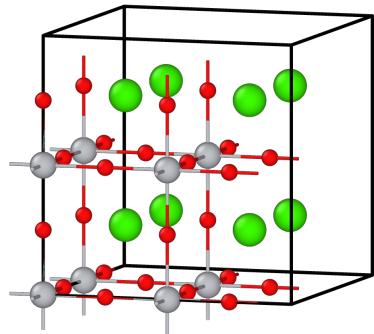
PET-MAD: A universal interatomic potential for advanced materials modeling

Arslan Mazitov

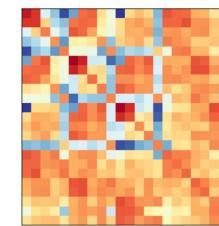
EPFL, Lausanne, Switzerland

Metatensor Workshop
June 13, 2025

Atomistic Machine Learning

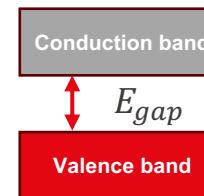


MLIP: energies,
forces, stresses



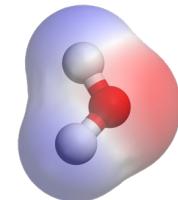
ML model

Energy ↑

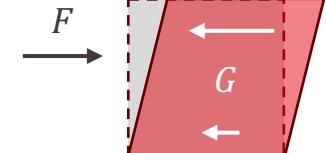


Electronic
properties

- Positions
- Cell vectors
- Chemical species

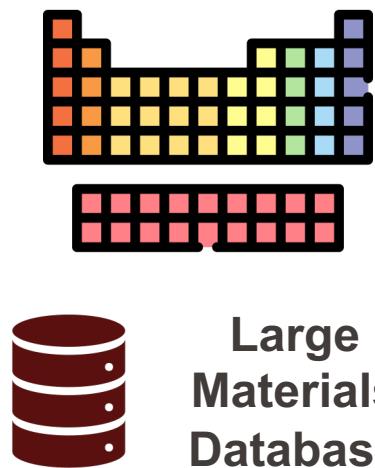


Charge
densities

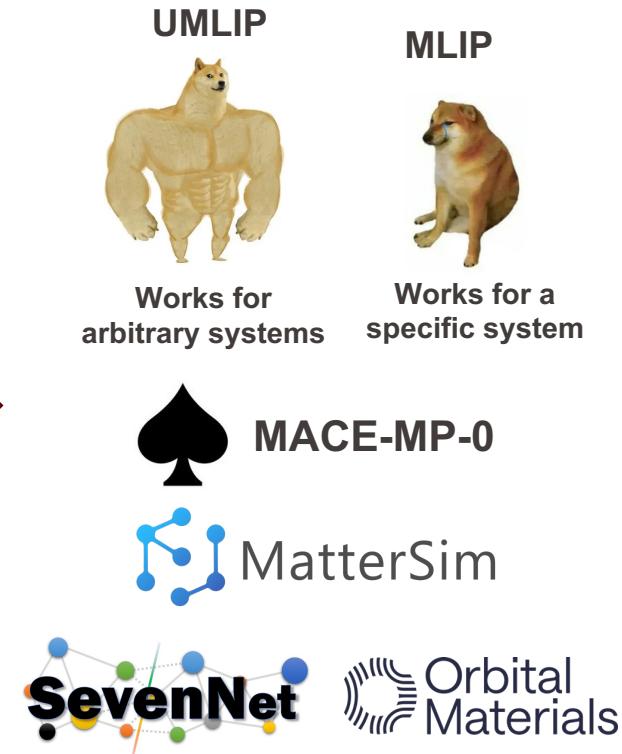


Mechanical
properties

Universal MLIPs



ML model



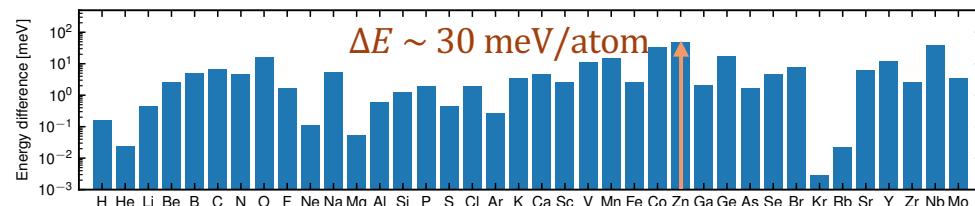
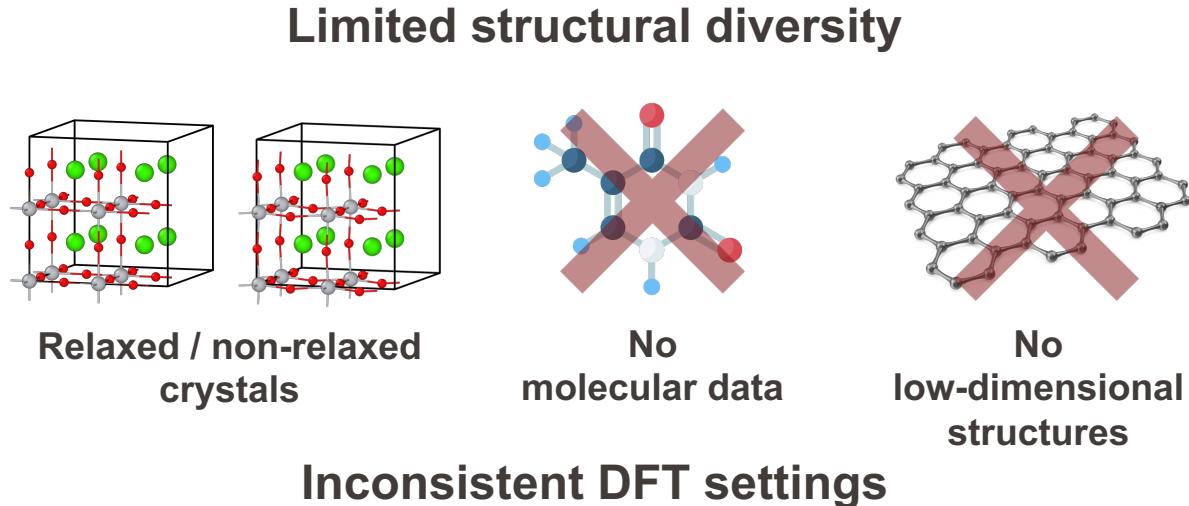
Examples of the
universal MLIPs

Are UMLIPs really universal?



Works for arbitrary
systems

... maybe
... maybe not



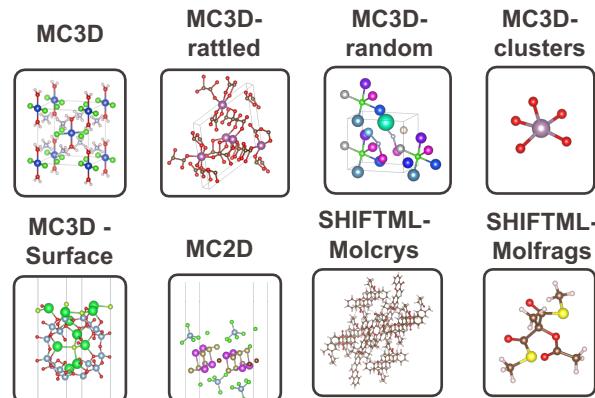
Energy difference between *consistent* and
default DFT settings



PET-MAD

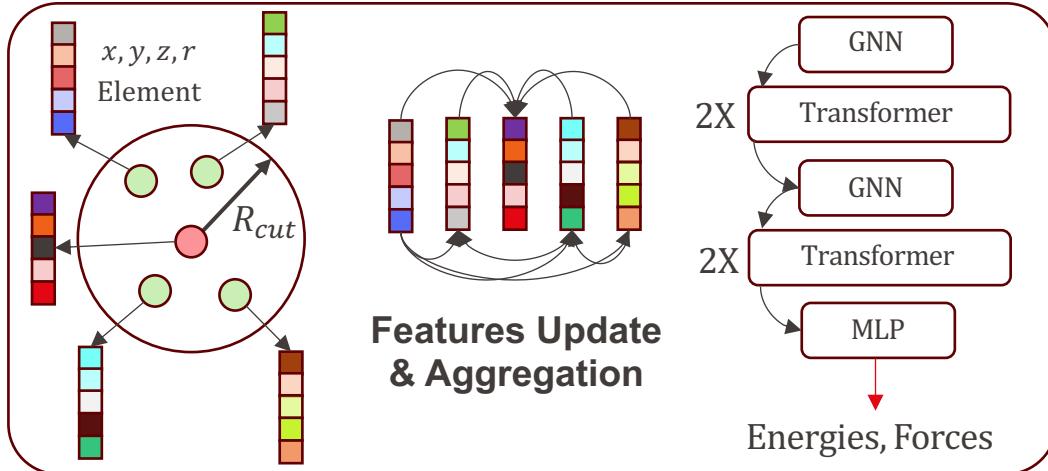
A UNIVERSAL INTERATOMIC POTENTIAL

Massive Atomic Diversity (MAD) Dataset



■ 8 classes of systems with 85 elements, enhanced diversity and consistent DFT

Point-Edge Transformer (PET)



Modern Transformer + GNN based architecture

Features of



PET-MAD

A UNIVERSAL INTERATOMIC POTENTIAL

- Infinite body-order
- Unconstrained architecture
- Learns **rotational equivariance** via Rotational Augmentations
- Has **non-conservative** (i.e. direct) **forces** and **stresses** prediction: 2x-3x speedup, yet a bit less accuracy
- Uncertainty **Quantification** and Uncertainty **Propagation** are available
- ASE and LAMMPS (with KOKKOS support) MD interfaces
- Fine-tuning capabilities

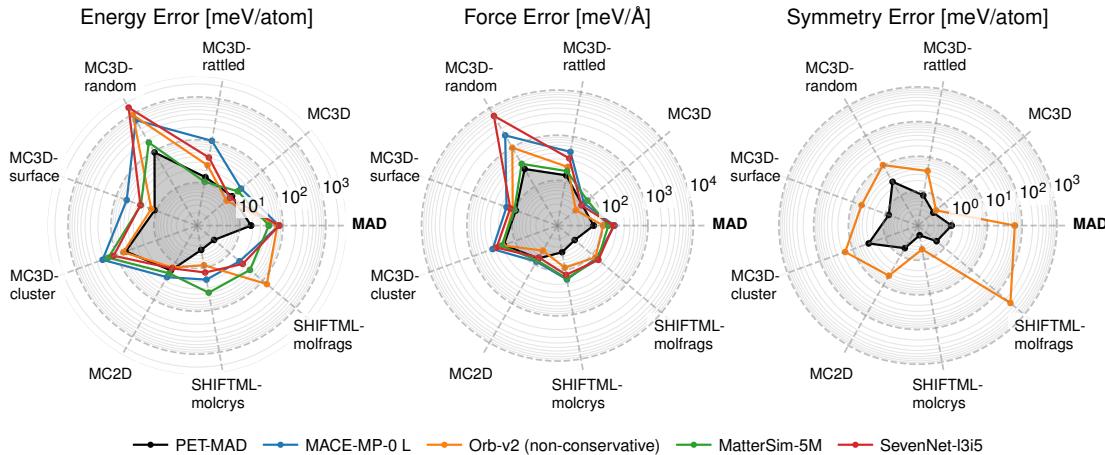


PET-MAD on benchmarks

Benchmarking on subsets of the MAD dataset against popular UMLIPS

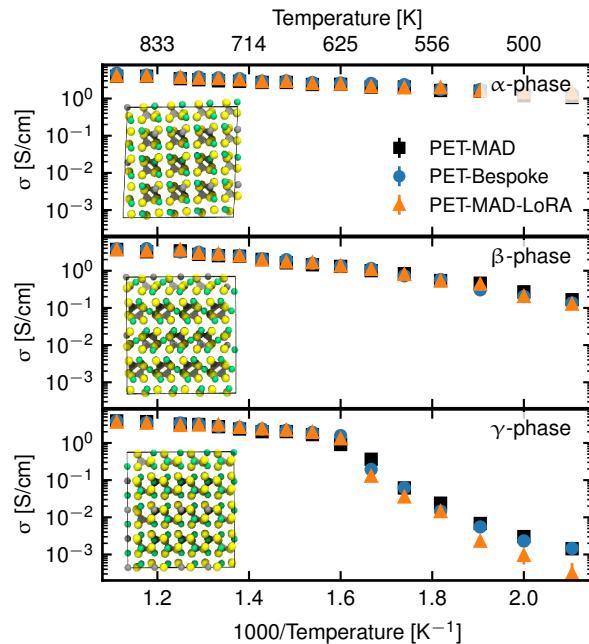
**WITH
CONSISTENT DFT**

Benchmarking on various atomistic datasets

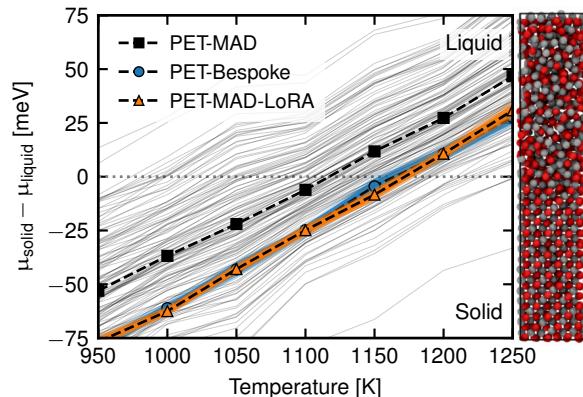


Dataset	PET-MAD	MACE-MP-0-L	MatterSim-5M	Orb-v2	SevenNet
MAD	18.2 63.2	81.6 181.5	47.3 133.7	74.4 103.0	82.1 173.5
MatBench	45.8 66.6	62.5 —	45.8 —	47.0 —	53.1 —
MPtrj	22.0 79.1	16.0 54.1	24.6 64.3	5.5 25.1	9.8 29.0
Alexandria	52.2 57.6	66.3 70.5	25.7 40.8	11.6 9.1	40.8 62.4
OC2020	17.6 114.6	82.8 167.8	30.9 118.1	20.6 104.3	46.4 157.5
SPICE	3.7 59.8	10.6 167.3	21.2 146.2	70.8 184.0	11.2 141.2
MD22	1.9 65.8	10.8 219.5	25.8 156.8	142.8 321.7	13.4 202.4

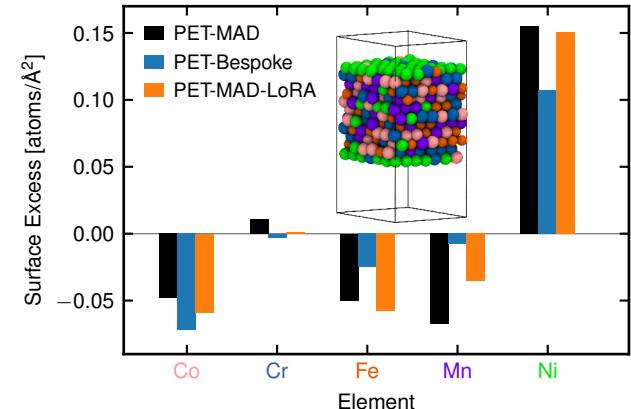
PET-MAD in applications



Ionic conductivity
in LiPS4

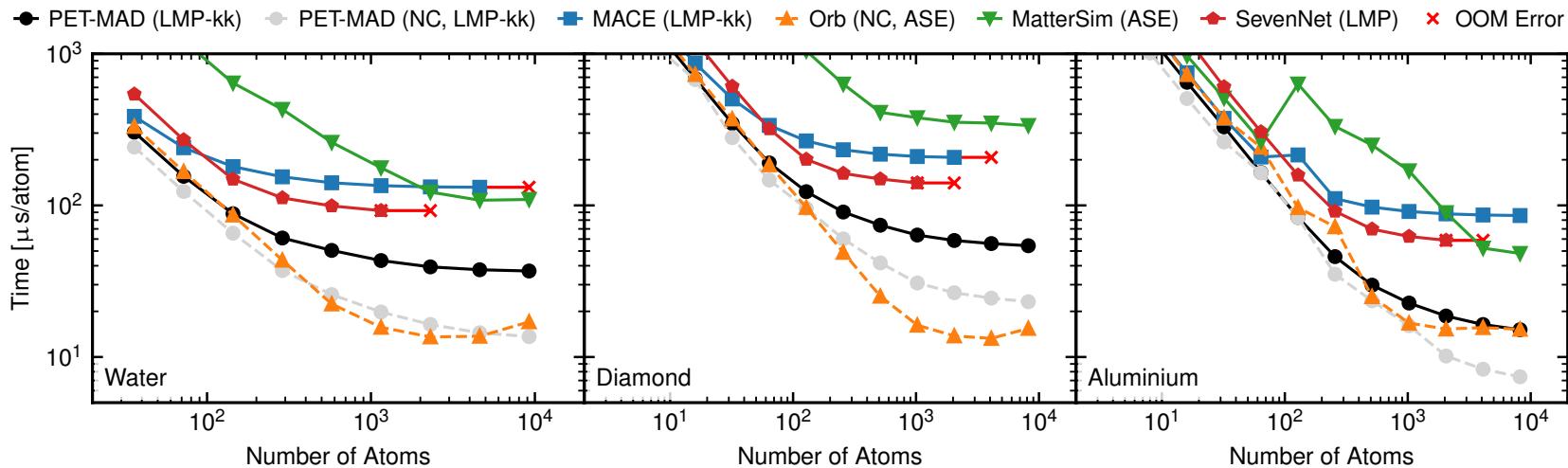


Melting point of GaAs



Surface segregation in
CoCrFeMnNi HEA

PET-MAD inference speed



PET-MAD delivers SOTA speed in various systems!

PET-MAD in Metatensor ecosystem

PET-MAD is trained using Metatrain



[checks passing](#) [codecov 91%](#) [documentation latest](#)

`metatrain` is a command line interface (cli) to train and evaluate atomistic models of various architectures. It features a common `yaml` option inputs to configure training and evaluation. Trained models are exported as standalone files that can be used directly in various molecular dynamics (MD) engines (e.g. `LAMMPS`, `i-PI`, `ASE` ...) using the [metatensor](#) atomistic interface.

The idea behind `metatrain` is to have a general hub that provide an homogeneous environment and user interface transforms every ML architecture in an end-to-end model that can be connected to an MD engine. Any custom architecture compatible with [TorchScript](#) can be integrated in `metatrain`, gaining automatic access to a training and evaluation interface, as well as compatibility with various MD engines.



[metatensor/metatrain](#)

Try PET-MAD!



PET-MAD

A UNIVERSAL INTERATOMIC POTENTIAL

PET-MAD: A Universal Interatomic Potential for Advanced Materials Modeling

This repository contains PET-MAD - a universal interatomic potential for advanced materials modeling across the periodic table. This model is based on the Point Edge Transformer (PET) model trained on the Massive Atomic Diversity (MAD) Dataset and is capable of predicting energies and forces in complex atomistic simulations.



[lab-cosmo/pet-mad](#)

Why Metatensor?

- Easy model training available via **Metatrain**
- Unified data interchange format via **Metatensor**
- Unified interface to atomistic simulation engines via **Metatomic**



<https://github.com/metatensor>

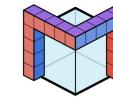


metatrain

[checks passing](#) [codecov 91%](#) [documentation latest](#)

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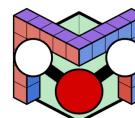


metatensor

[checks passing](#) [documentation latest](#) [codecov 87%](#)

Metatensor is a self-describing sparse tensor data format for atomistic machine learning and beyond; storing values and gradients of these values together. Think `numpy.ndarray` or `pytorch Tensor` equipped with extra metadata for atomic systems and other point clouds data. The core of this library is written in Rust and we provide API for C, C++, and Python.

The main class of metatensor is the `TensorMap` data structure, defining a custom block-sparse data format. If you are using metatensor from Python, we additionally provide a collection of mathematical, logical and other utility operations to make working with `TensorMaps` more convenient.

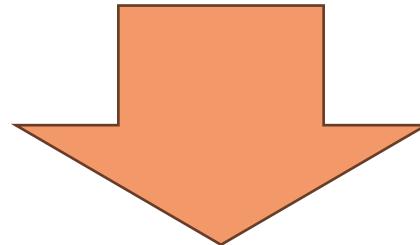


metatomic

[checks passing](#) [documentation latest](#) [codecov 82%](#)

`metatomic` is a library that defines a common interface between atomistic machine learning models, and atomistic simulation engines. Our main goal is to define and train models once, and then be able to re-use them across many different simulation engines (such as LAMMPS, GROMACS, etc.). We strive to achieve this goal without imposing any structure on the model itself, and to allow any model architecture to be used.

I tried PET-MAD, it didn't work on my data



Fine-tuning

Join the PET-MAD fine-tuning
session today at 13:30
in MXF 312!

Big thanks to our team!



Filippo



Matthias



Paolo



Davide



Guillaume



Sofia

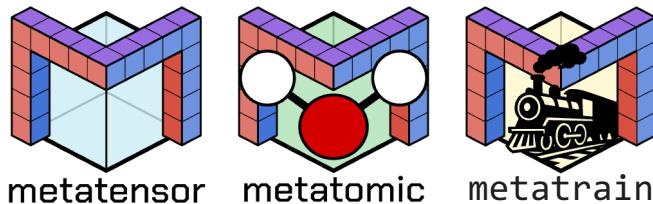
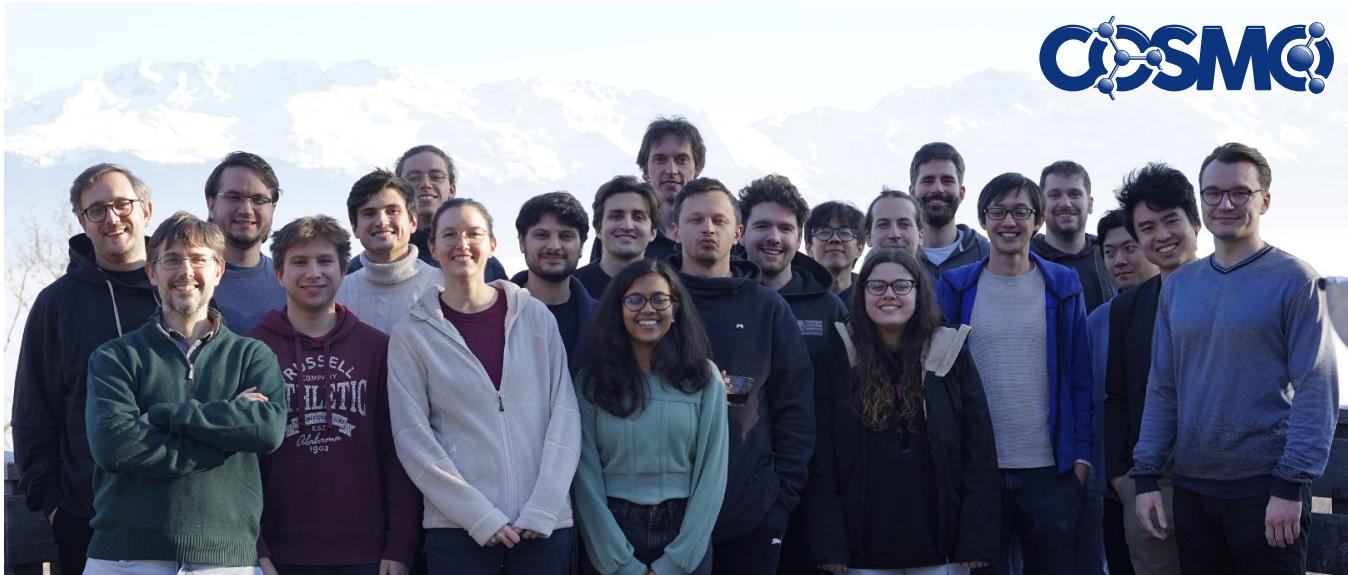


Sergey



Michele

Big thanks to our group!



- Software stack to meet the diverse needs of atomistic machine learning
- Developed with support from SNSF, ERC, and importantly: MARVEL



Useful links

PET-MAD arXiv
pre-print



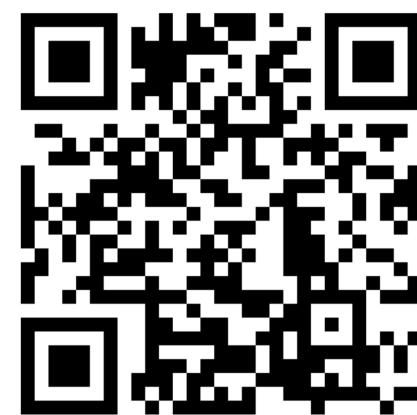
<https://arxiv.org/pdf/2503.14118.pdf>

PET-MAD repo



lab-cosmo/pet-mad

Metatensor
ecosystem



metatensor