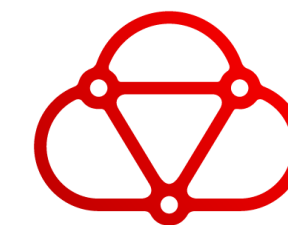


PSI

Center for Scientific Computing,
Theory and Data



AiiDA



MATERIALSCLOUD



AiiDAlab

Beyond FAIR data: FAIR and reproducible workflows with AiiDA, Materials Cloud and AiiDAlab



Giovanni Pizzi, PSI

EnhanceR symposium, 7 November 2024

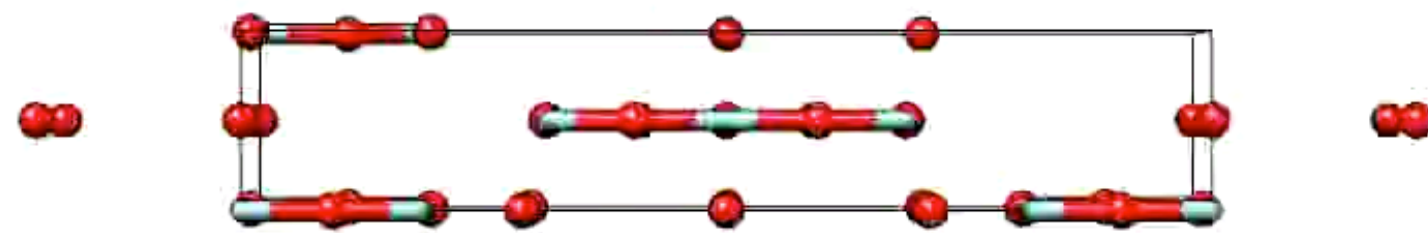
My group's role (“Materials Software and Data”)



Between research and research software engineering (RSE)

- 11 people in my group, **~half are RSE**
- Myself: physicist. Most people in the group (both students/postdocs and RSE): **physics/chemistry/mat. science background**
- **Goal:** develop tools that enable:
 - **efficient** high-throughput research
 - automated simulations with **reproducibility**
 - **robust automated workflows** with minimal input/knowledge required
 - **seamless access** to advanced workflows by “non-experts” (e.g. via GUIs)

Scientific goal: Compute materials properties with supercomputers



Aim: Compute properties for all of them
(and even new, invented ones)
and **discover novel functional materials**



Challenges in high-throughput HPC



Workflow automation

- Need tools to define complex workflows with advanced error handling
- An automated, robust and scalable engine to run the workflows

Data management

- Data should be stored reliably and efficiently
- Stored data should be interoperable and queryable

Reproducibility

- All produced data should be reproducible by storing the full provenance

Further challenges to make it FAIR

FAIR data

- Make it easy for users to generate (and publish) FAIR data with **minimal effort for researchers**
- Can we leverage ontologies/semantic annotations to make data **machine-interoperable**?

Beyond FAIR data: FAIR workflows

- Can we make workflows **interoperable**? (e.g. between codes)
- Can we make workflows and advanced simulation methods **accessible**?

Reproducible simulations and interoperable workflows



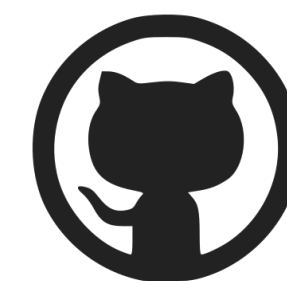
- COMPUTATIONAL SCIENCE INFRASTRUCTURE
- FOR HIGH THROUGHPUT WORKFLOWS
- WITH FULL DATA PROVENANCE



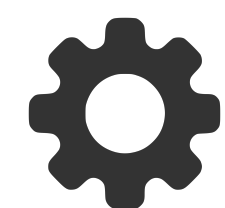
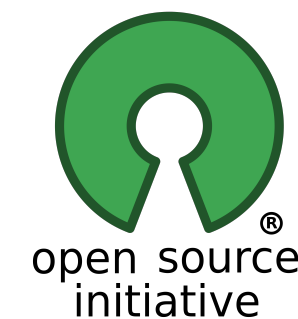
Language: implemented and API in python

License: MIT open source <http://www.aiida.net/>

Source: <https://github.com/aiidateam/aiida-core>

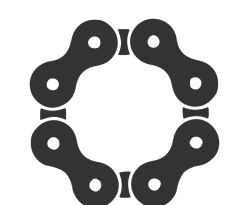


MIT LICENSED



Scalable workflow engine: **robustness**

Automated full data provenance: **reproducibility**



Built-in support for HPC: **performance**

Flexible plugin system: **interoperability**

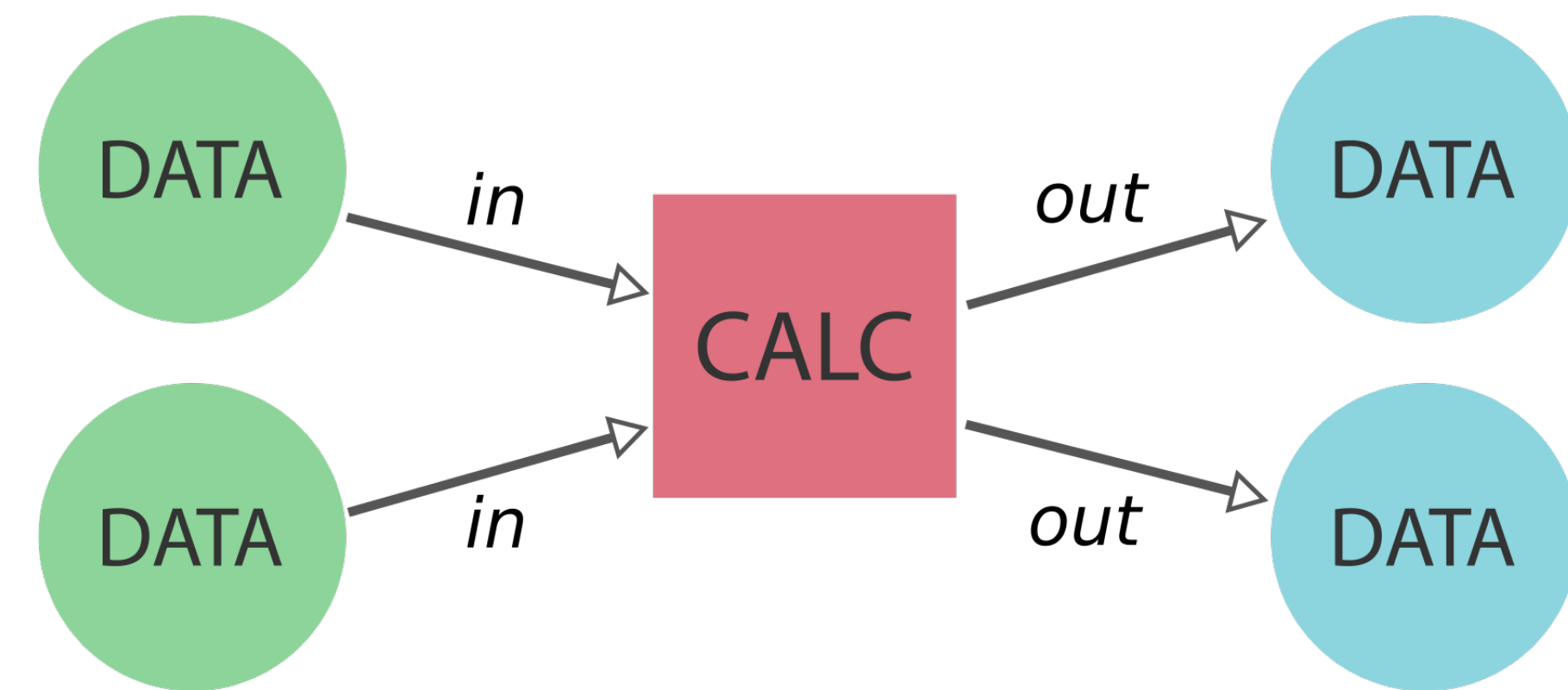


G. Pizzi et al., Comp. Mat. Sci. 111, 218-230 (2016)

S.P. Huber et al., Scientific Data 7, 300 (2020)

Simple recipe

- Store data transformations or '**calculations**'
- Store its **inputs** and their metadata
- Store its **outputs** and their metadata
- Most **crucially** store the **inter-connections**

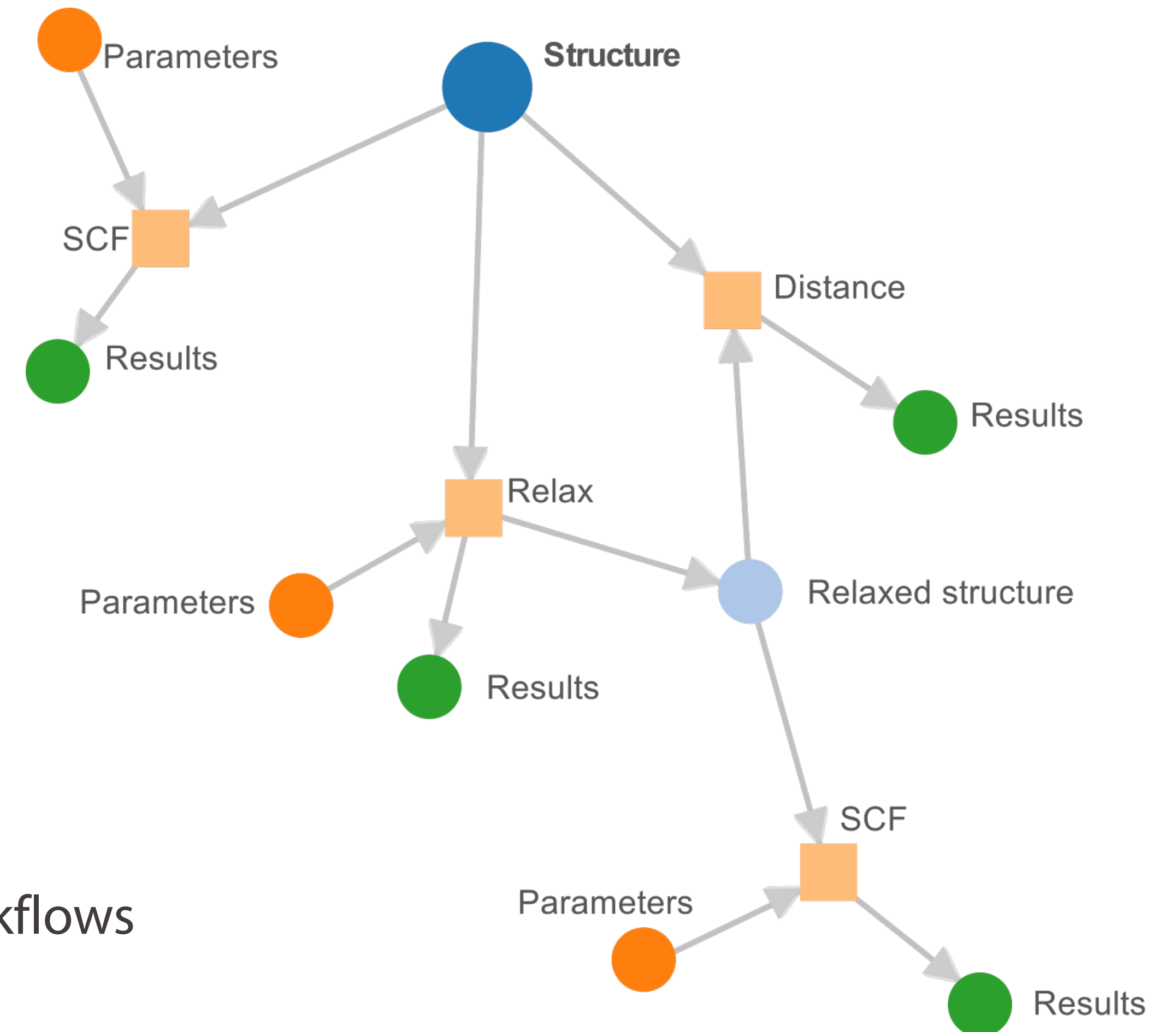


Simple recipe

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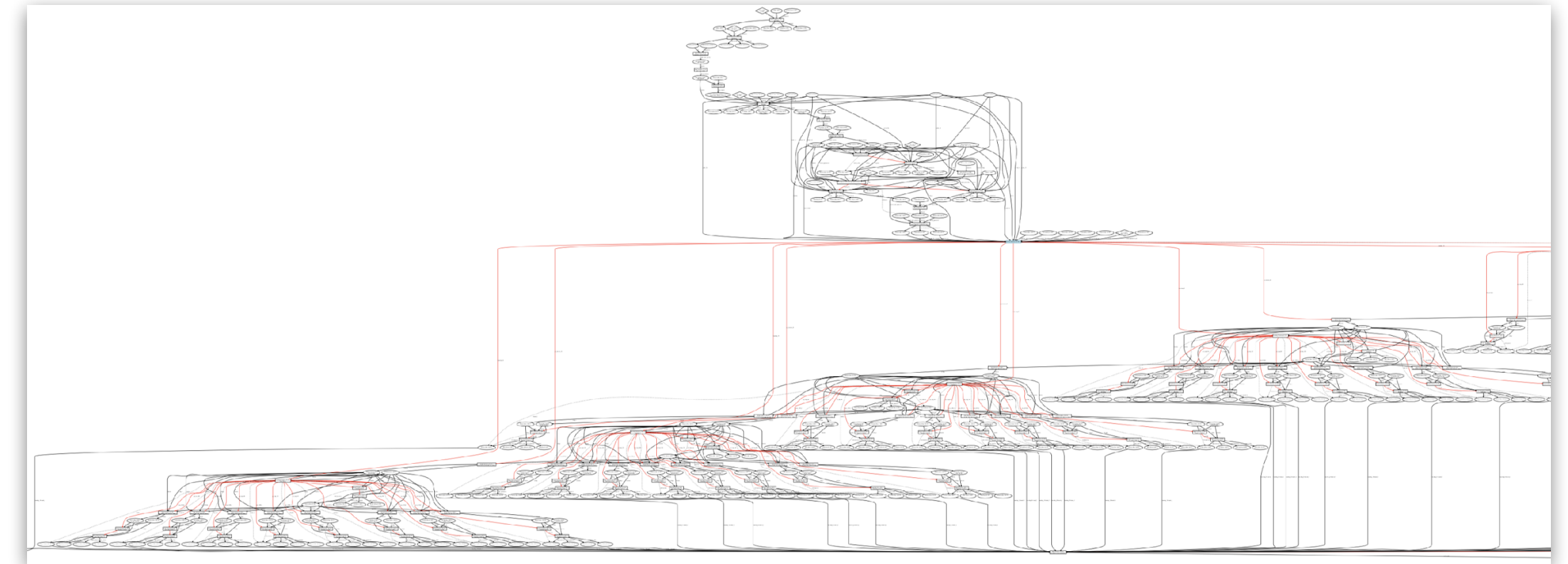
Provenance graphs

- When data gets reused, a directed graph is created
- That quickly grow in complexity even for "simple" workflows



Simple recipe

- Store data transformations or '**calculations**'
- Store its **inputs** and their metadata
- Store its **outputs** and their metadata
- Most **crucially** store the **inter-connections**

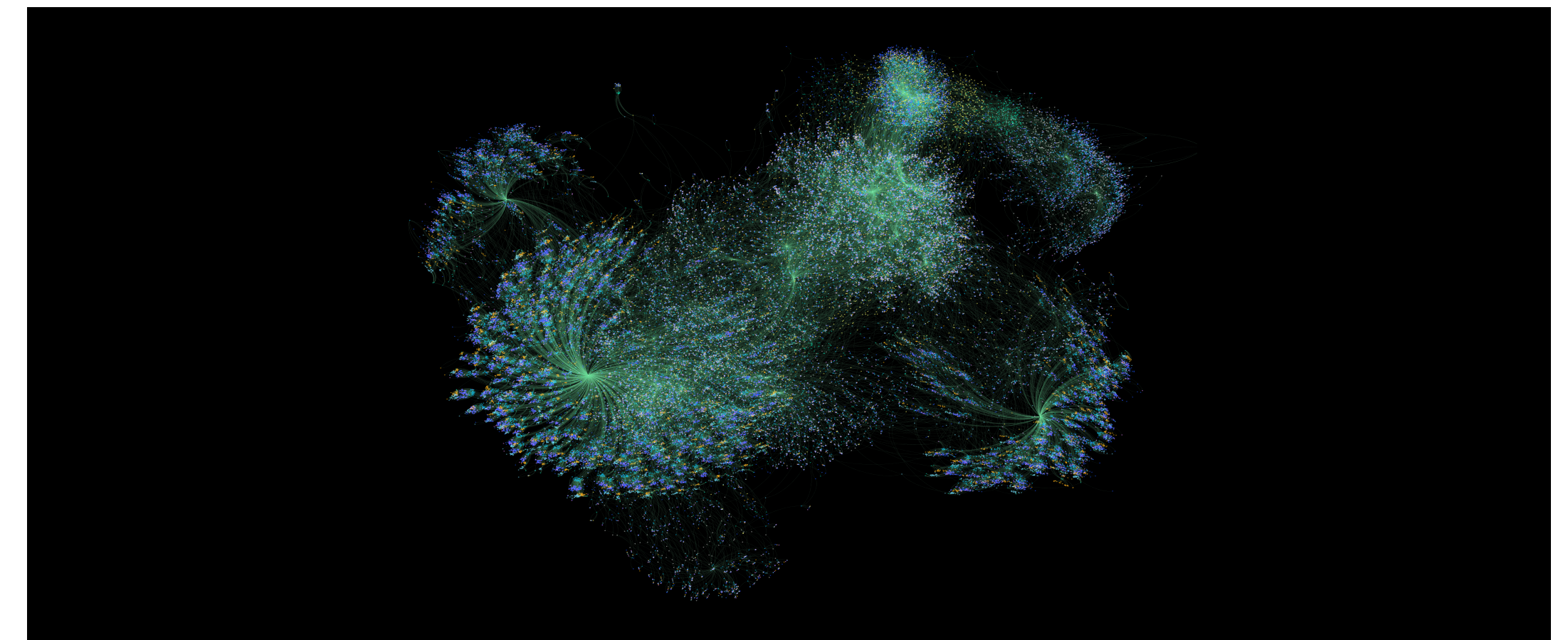


Molecular dynamics study of Lithium in a solid electrolyte

Graph requirements

- Needs to be automated
- Needs to be stored *as data is created*

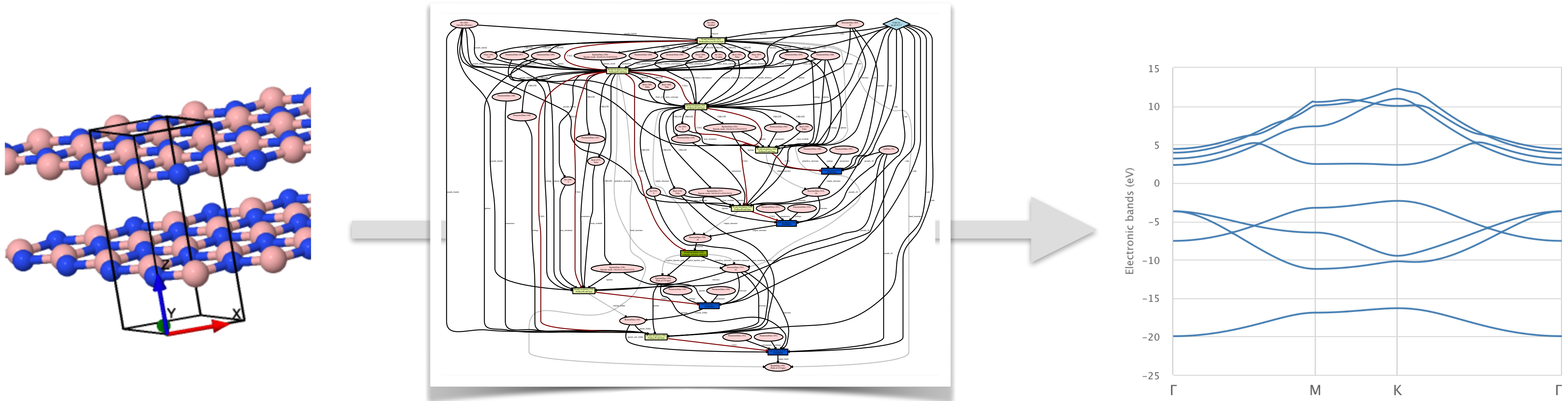
Complexity grows quickly even for simple workflows and is impossible to reconstruct *a posteriori*



Graphical representation of actual AiiDA database

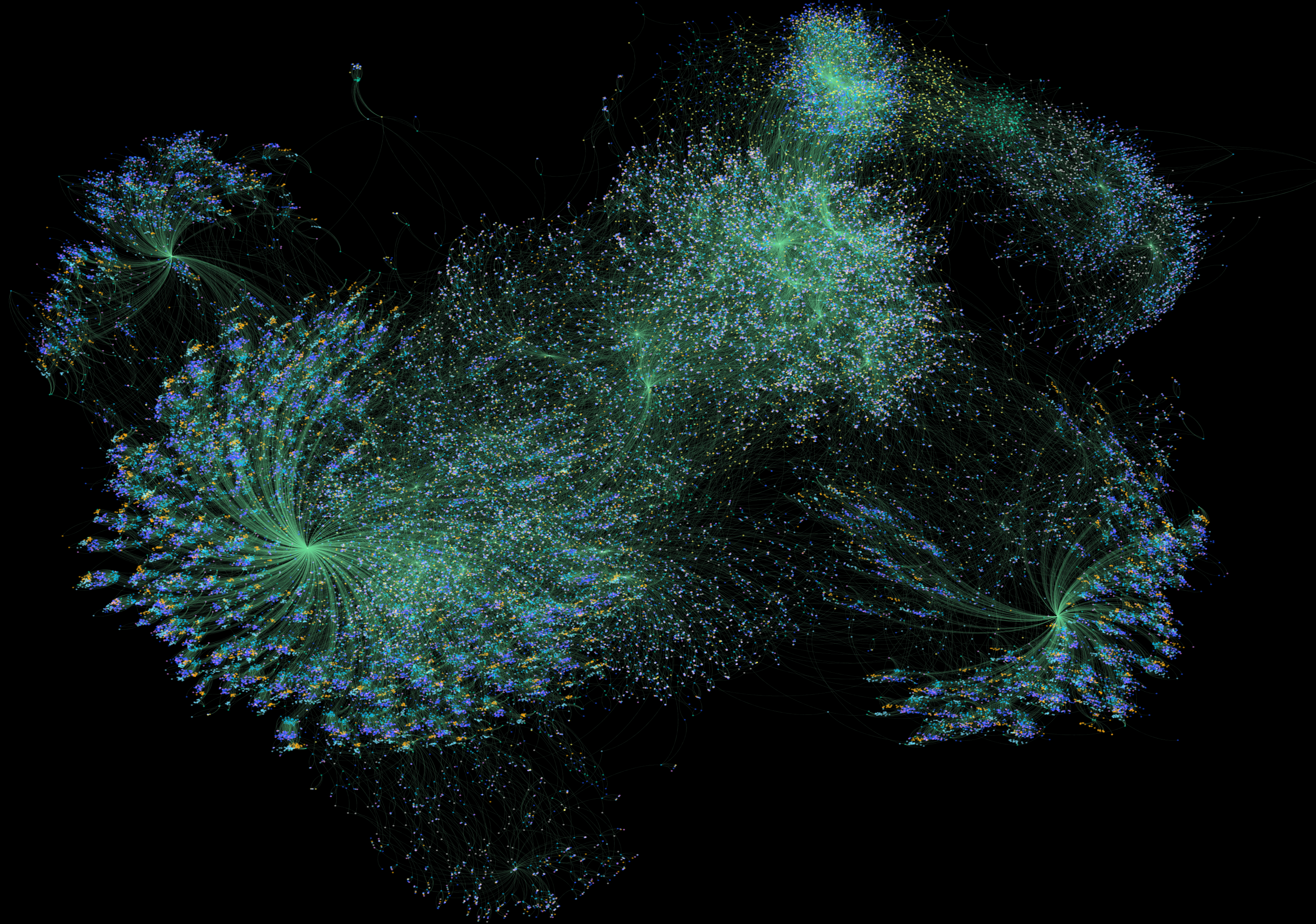
Workflows to generate data

- Given a material, we compute advanced quantities: often non-trivial, result of a complex workflow

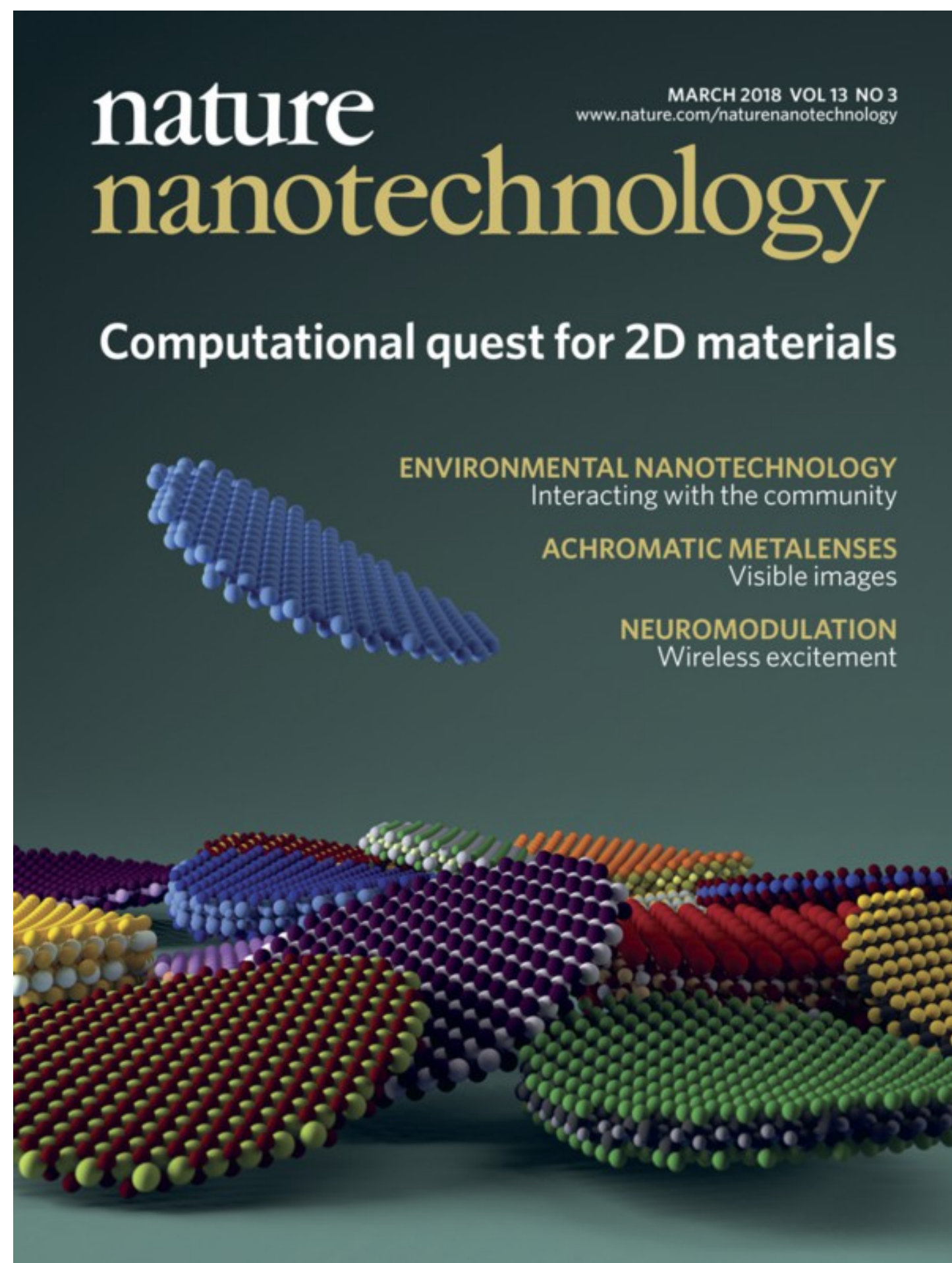


- AiiDA provenance graph: *log of “what happened in the past”*:
reproduce that single specific workflow execution
- AiiDA workflow engine**: python interface to encode complex scientific workflows

FAIR (workflow) data sharing



Making all data open (and reproducible, and FAIR)



N. Mounet et al., Nat. Nanotech. 13, 246 (2018)
D. Campi et al., ACS Nano 17 11268 (2023)

AiiDA used to obtain **MC2D**, a large database of exfoliable 2D materials

Automated workflows to perform 100'000+ DFT simulations, leading to >1800 novel 2D materials

All data open and FAIR on Materials Cloud Archive

N. Mounet et al., Materials Cloud Archive 2020.158 (2020), doi: 10.24435/materialscloud:az-b2
D. Campi et al., Materials Cloud Archive 2022.84 (2022), doi: 10.24435/materialscloud:36-nd

Materials Cloud two-dimensional crystals database (MC2D)

DOI 10.24435/materialscloud:az-b2
DOI 10.24435/materialscloud:36-nd

Compound: BN

Get data from REST API

Available space groups for this formula: P-6m2

Drag to rotate, scroll to zoom, right-click for other
Download structure:

Double-click to toggle interactions on and off
(This feature is not available on iPad and iPhone)

Formula: BN
Spacegroup: P-6m2
Pointgroup: -6m2
Prototype: BN
Band gap [eV]: 4.7
Abundance: $8.6 \cdot 10^{-6}$

Magnetic properties:
Magnetic State: non-magnetic
Tot. Magnetization [$\mu\text{B}/\text{cell}$]: N/A
Abs. Magnetization [$\mu\text{B}/\text{cell}$]: N/A

Binding Energies:
DF2-C09 Binding energy [$\text{meV}/\text{\AA}^2$]: 19.4
(From parent ICSD 186248)
rVV10 Binding energy [$\text{meV}/\text{\AA}^2$]: 24.4
(From parent ICSD 186248)

Delta in interlayer distance (vdW vs revPBE):
 $\Delta_{\text{DF2}} [\%]$: 45.4
(From parent ICSD 186248)

Supercell: 3 3 1 UPDATE
RESET 3x3x1 CELL

Camera: X Y Z

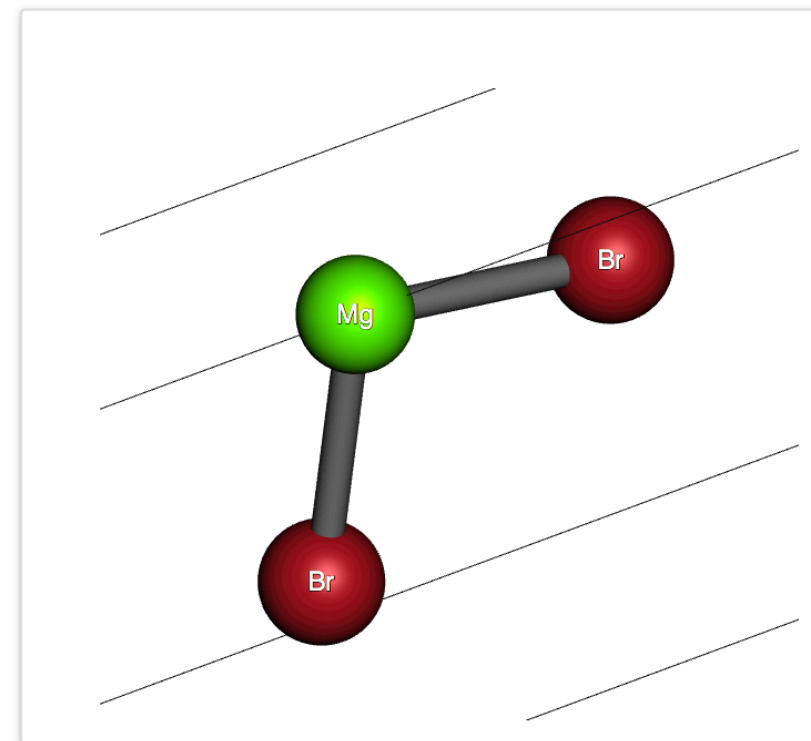
<http://mc2d.materialscloud.org>

FAIR data sharing (MC2D): Materials Cloud Archive, Discover, Explore



DISCOVER

Compound: MgBr_2



Info and properties
[See definitions...](#)

Formula: MgBr_2
Spacegroup: P-3m1
Pointgroup: -3m
Prototype: CdI2
Band gap [eV]: 4.8

Magnetic properties:
Magnetic State: non-m
Tot. Magnetization [μ_B]
Abs. Magnetization [μ_B]

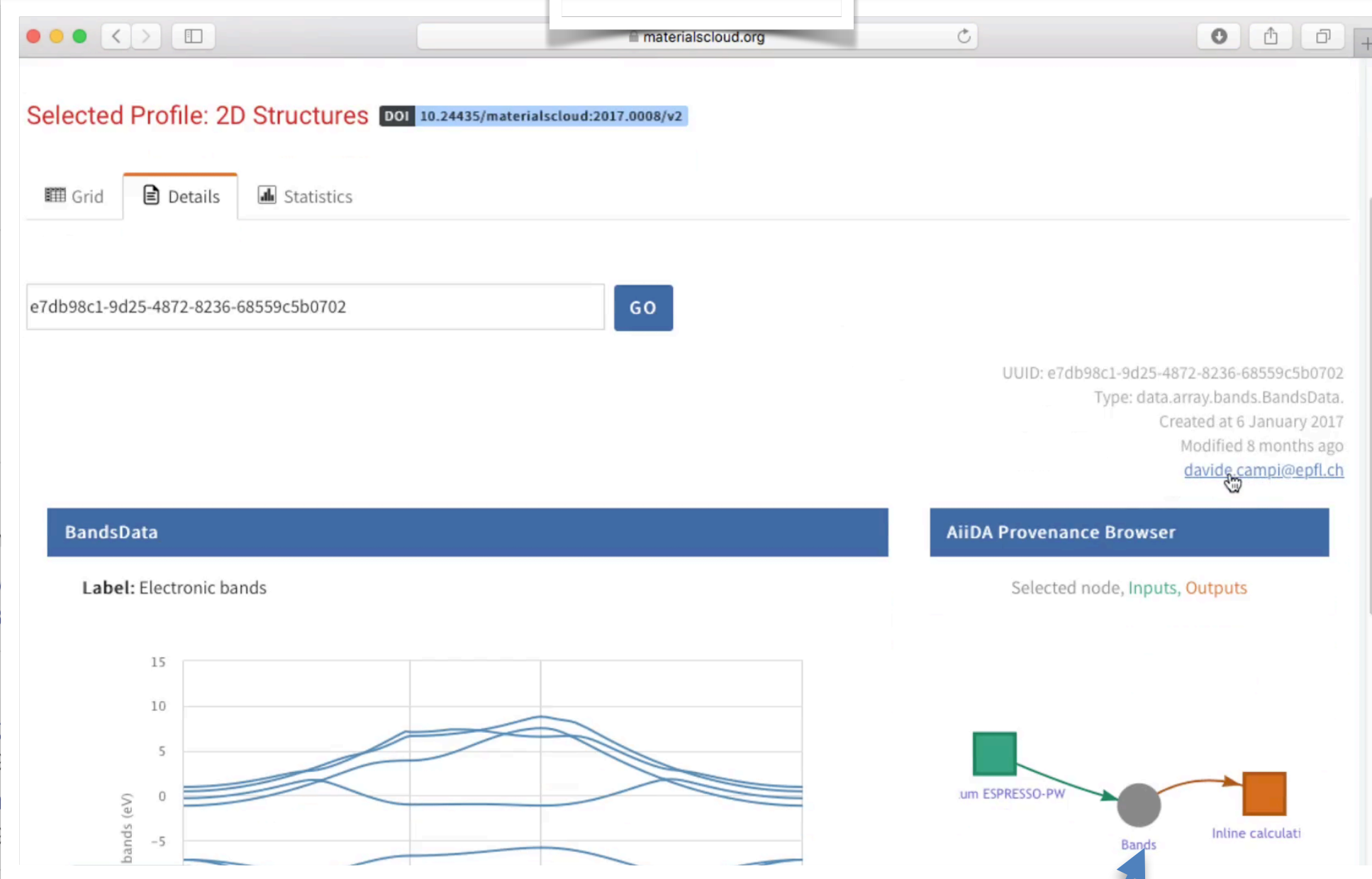
Binding Energies:
DF2-C09 Binding energy
(From parent COD 9009107)
rVV10 Binding energy [μ_B]
(From parent COD 9009107)

Delta in interlayer distance (vdW vs revPBE):
 Δ_{DF2} [%]: 17.1 (From parent COD 9009107)
 Δ_{rVV10} [%]: 18.3 (From parent COD 9009107)

Band structure



EXPLORE



UUID links to jump to the provenance graph in the EXPLORE section

Browse the full AiiDA provenance graph (inputs, outputs, ...) at any level

<http://mc2d.materialscloud.org>

ARCHIVE



materialscloud:2020.158

Two-dimensional materials from high-throughput computational exfoliation of experimentally known compounds

Nicolas Mounet^{1*}, Marco Gibertini¹, Philippe Schwaller¹, Davide Campi¹, Andrius Merkys^{1,2}, Antimo Marrazzo¹, Thibault Sohier¹, Ivano E. Castelli¹, Andrea Cepellotti¹, Giovanni Pizzi¹, Nicola Marzari^{1*}

¹ Theory and Simulation of Materials (THEOS), and National Centre for Computational Design and Discovery of Novel Materials (MARVEL), École Polytechnique Fédérale de Lausanne, CH-1015 Lausanne, Switzerland
² Vilnius University Institute of Biotechnology, Sauletekio al. 7, LT-10257 Vilnius, Lithuania

* Corresponding authors emails: nicolas.mounet@epfl.ch, nicola.marzari@epfl.ch

DOI: 10.24435/materialscloud:az-b2 [version v4]
Publication date: Dec 02, 2020

How to cite this record

Nicolas Mounet, Marco Gibertini, Philippe Schwaller, Davide Campi, Andrius Merkys, Antimo Marrazzo, Thibault Sohier, Ivano E. Castelli, Andrea Cepellotti, Giovanni Pizzi, Nicola Marzari, *Two-dimensional materials from high-throughput computational exfoliation of experimentally known compounds*, Materials Cloud Archive 2020.158 (2020), doi: 10.24435/materialscloud:az-b2.

N. Mounet et al., Materials Cloud Archive 2020.158 (2020),
doi: 10.24435/materialscloud:az-b2

Research data repository: Materials Cloud Archive



materialscloud:2017.0008/v3

DOIs assigned

Two-dimensional materials from high-throughput computational exfoliation of experimentally known compounds

Nicolas Mounet^{1*}, Marco Gibertini¹, Philippe Schwaller¹, Davide Campi¹, Andrius Merkys^{1,2}, Antimo Marrazzo¹, Thibault Sohier¹, Ivano E. Castelli¹, Andrea Cepellotti¹, Giovanni Pizzi¹, Nicola Marzari^{1*}

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² Vilnius University Institute of Biotechnology, Sauletekio al. 7, LT-10257 Vilnius, Lithuania

* Corresponding authors emails: nicolas.mounet@epfl.ch, nicola.marzari@epfl.ch

DOI 10.24435/materialscloud:2017.0008/v3 [version v3]
Publication date: Apr 03, 2019

How to cite this record

Nicolas Mounet, Marco Gibertini, Philippe Schwaller, Davide Campi, Andrius Merkys, Antimo Marrazzo, Thibault Sohier, Ivano E. Castelli, Andrea Cepellotti, Giovanni Pizzi, Nicola Marzari, *Two-dimensional materials from high-throughput computational exfoliation of experimentally known compounds*, Materials Cloud Archive **2017.0008/v3** (2019), doi: [10.24435/materialscloud:2017.0008/v3](https://doi.org/10.24435/materialscloud:2017.0008/v3).

Description

Two-dimensional (2D) materials have emerged as promising candidates for next-generation electronic and optoelectronic applications. Yet, only a few dozens of 2D materials have been successfully synthesized or exfoliated. Here, we search for novel 2D materials that can be easily exfoliated from their parent compounds. Starting from 108423 unique, experimentally known three-dimensional compounds we identify a subset of 5619 that appear layered according to robust geometric and bonding criteria. High-throughput calculations using van-der-Waals density-functional theory, validated against experimental structural data and calculated random-phase-approximation binding energies, allow to identify 1825 compounds that are either easily or potentially exfoliable. In particular, the subset of 1036 easily exfoliable cases provides novel structural prototypes and simple ternary compounds as well as a large portfolio of materials to search for optimal properties. For a subset of 258 compounds we explore vibrational, electronic, magnetic, and topological properties, identifying 56 ferromagnetic and antiferromagnetic systems, including half-metals and half-semiconductors. This archive entry contains the database of 2D materials (structural parameters, band structures, binding energies, phonons for the subset of the 258 easily exfoliable materials with less than 6 atoms, structures and binding energies for the remaining 1567 materials) together with the provenance of all data and calculations as stored by AiiDA.

Materials Cloud sections using this data

- Select 2d materials via interactive periodic table and view their properties (with links to provenance)
- Explore interface providing access to the full database

Files

File name	Size	Description
2d_materials.tar.gz	113.0 MiB	We provide 258 two-dimensional crystal structures (lattice vectors, atomic species and positions), exfoliated from three-dimensional experimental crystal structures. The structures were relaxed at the DFT-PBE level. Together with each structure, a set of materials properties is also given (at the DFT-PBE level): chemical formula, spacegroup, structural prototype, magnetic state, magnetization, band-gap, electronic bands, and phonon



Export
Dublin Core JSON

Recommended repository
by Nature's journal **Scientific Data**,
the EU **Open Research Europe**, and
SNSF

SCIENTIFIC
DATA



Research and Innovation

Open Research Europe

Indexed by **Google Dataset Search**
and EUDAT/EOSC's **B2FIND**;
Registered on [FAIRsharing.org](https://www.fairsharing.org) and
re3data.org

Currently using (extended
version of) **CERN's Invenio v3**

Now **migrating to**
InvenioRDM v12

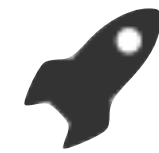
Direct links to
Discover & Explore

Data (and metadata)
guaranteed to be online
for at least 10 years after
deposition

Accessible simulation capabilities



FAIR sharing in AiiDA beyond data: codes, plugins and workflows



Calculation



Data



Parsers



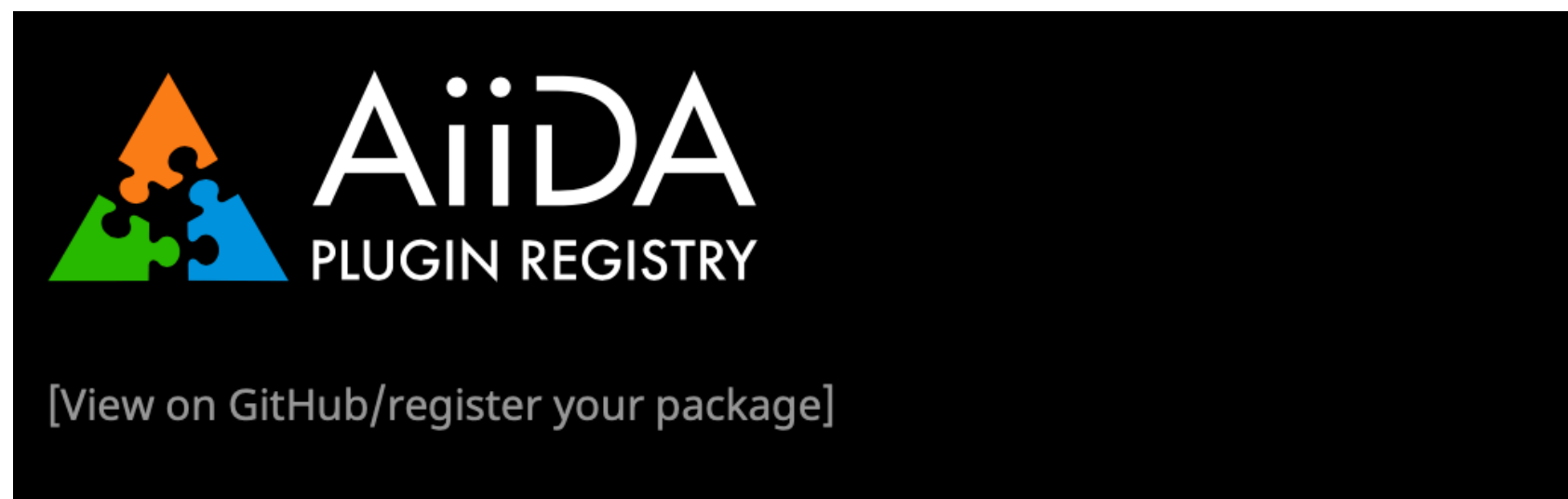
Transport and
scheduler



Workflows



Importers &
exporters

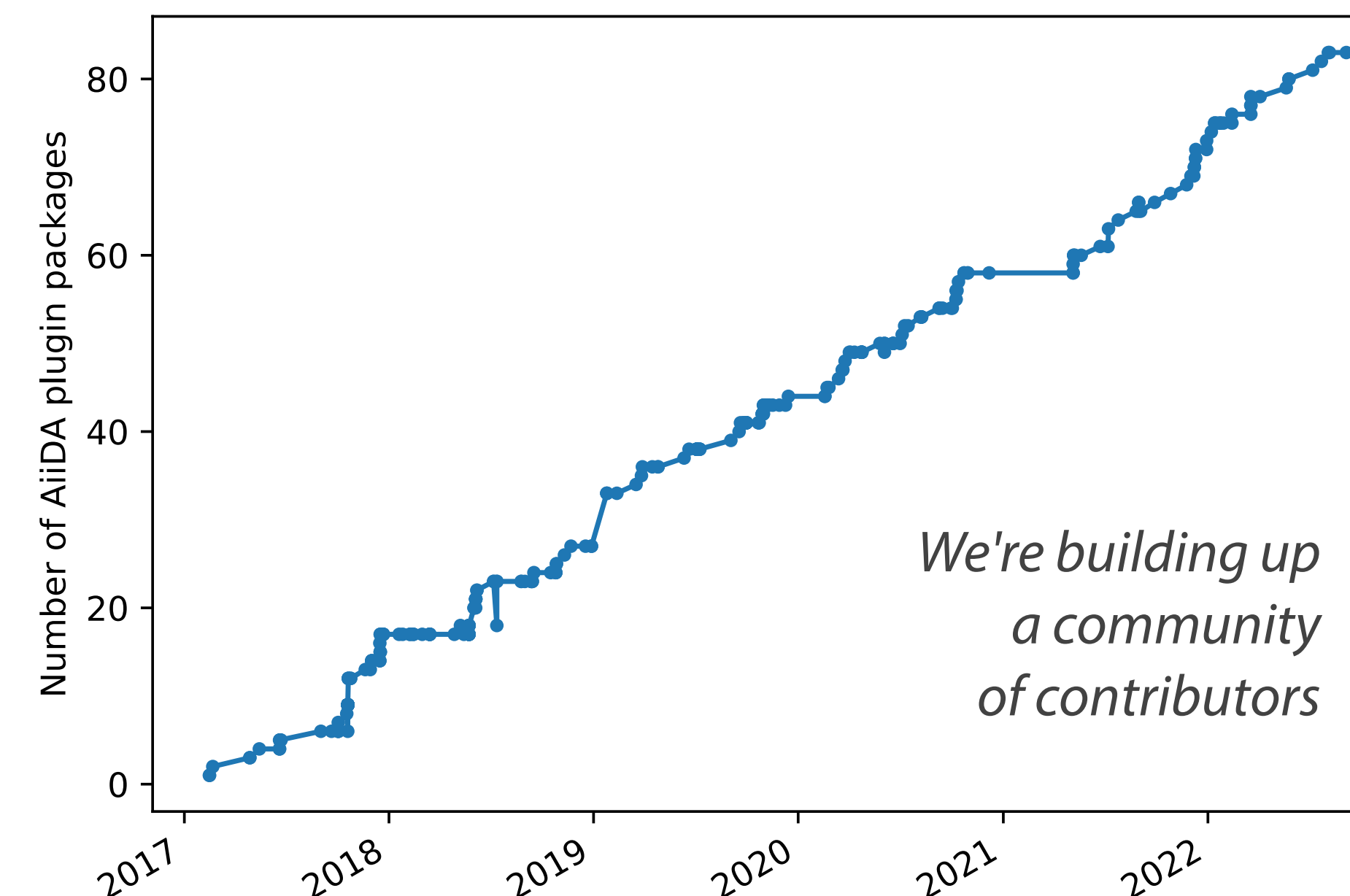


Registered plugin packages: 97

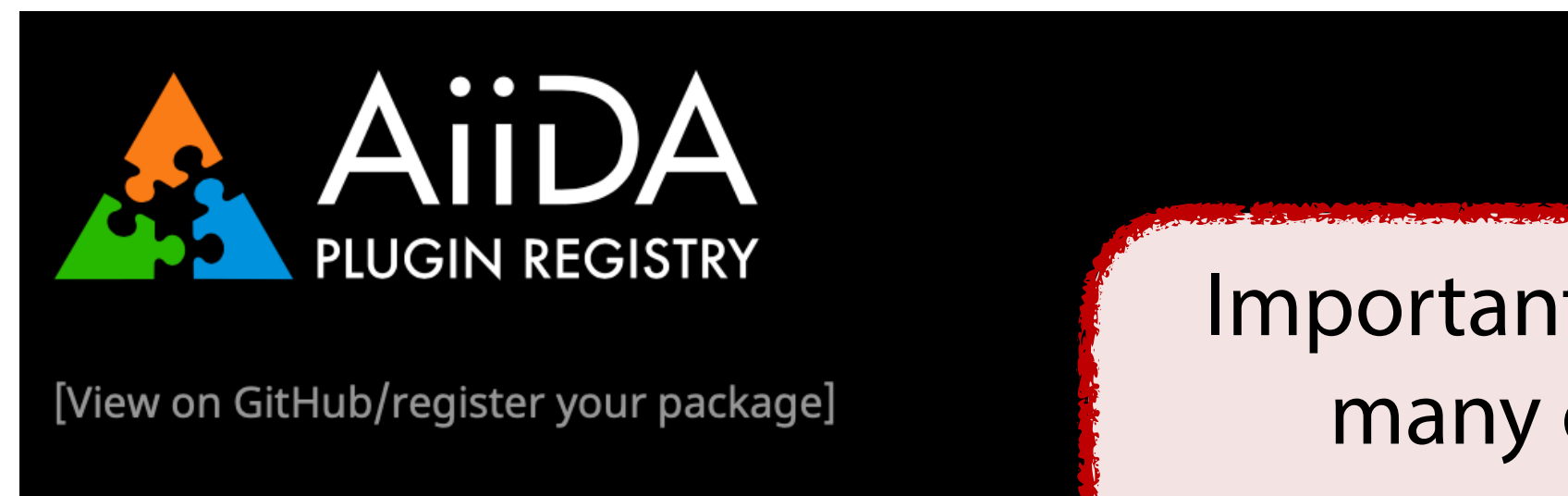
Calculations	164 plugins in 66 packages
Parsers	145 plugins in 67 packages
Data	122 plugins in 38 packages
Workflows	211 plugins in 50 packages
Console scripts	33 plugins in 17 packages
Other	110 plugins in 33 packages

<https://aiidateam.github.io/aiida-registry/>

- Plugins collected in the AiiDA plugin registry
- **160+ codes currently supported**, 200+ workflows
- Many are **community-contributed**



FAIR sharing in AiiDA beyond data: codes, plugins and workflows



- Plugins collected in the AiiDA plugin registry
- **Not yet fully supported**, 200+ workflows
- **Community-contributed**

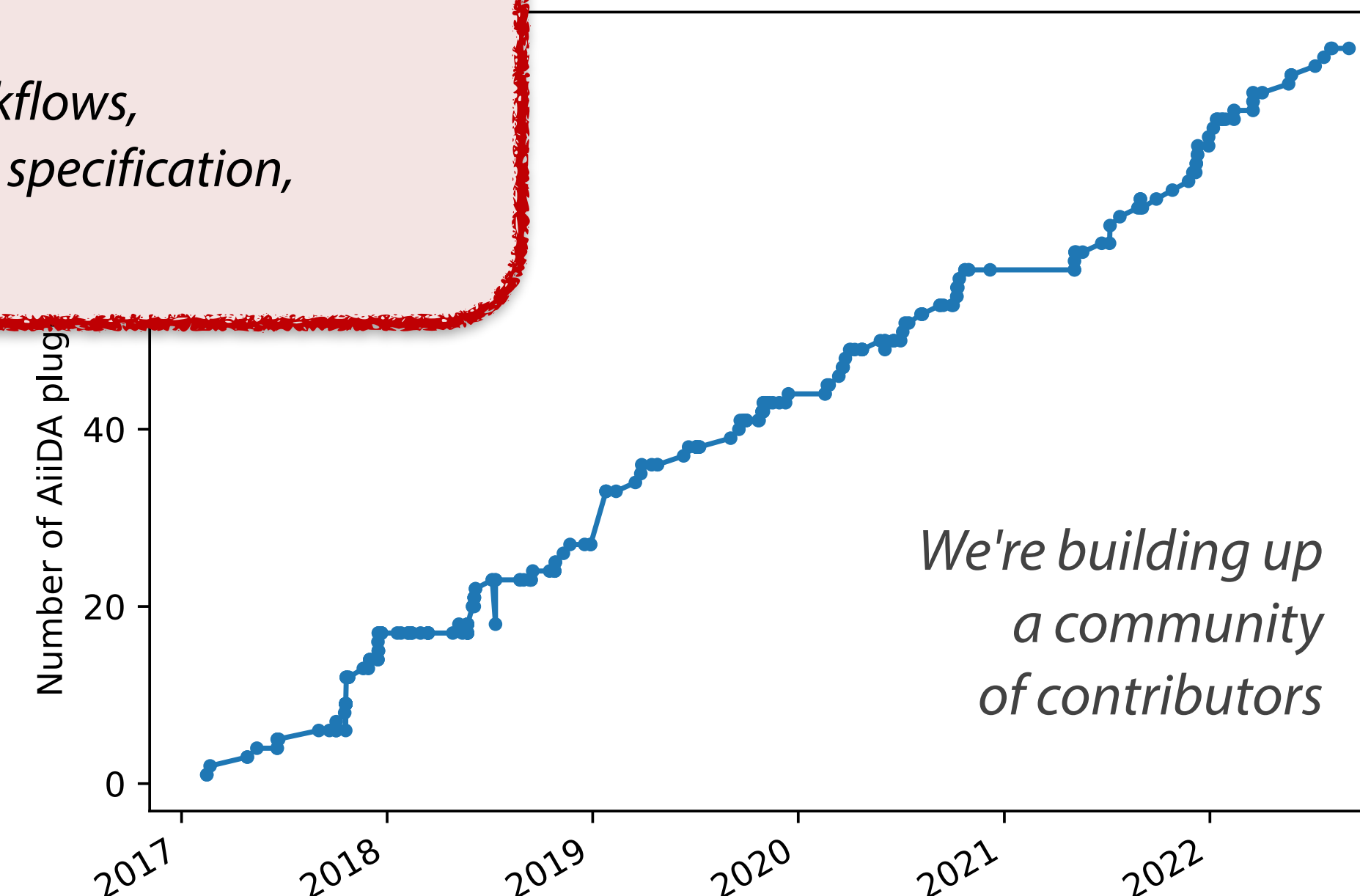
Important "requirement" to support many codes, but **not enough!**

*No turn-key workflows,
no unique input/output specification,
...*

Registered plugin packages: 97

Calculations	164 plugins in 66 packages
Parsers	145 plugins in 67 packages
Data	122 plugins in 38 packages
Workflows	211 plugins in 50 packages
Console scripts	33 plugins in 17 packages
Other	110 plugins in 33 packages

<https://aiidateam.github.io/aiida-registry/>



The need for turn-key solutions



Like in a car: drive without needing to know how the engine works

- Engines are "**robust**"
- Just **turn the key** and drive
- I need a **driving license**, but I don't need to learn again if I change the brand of my car

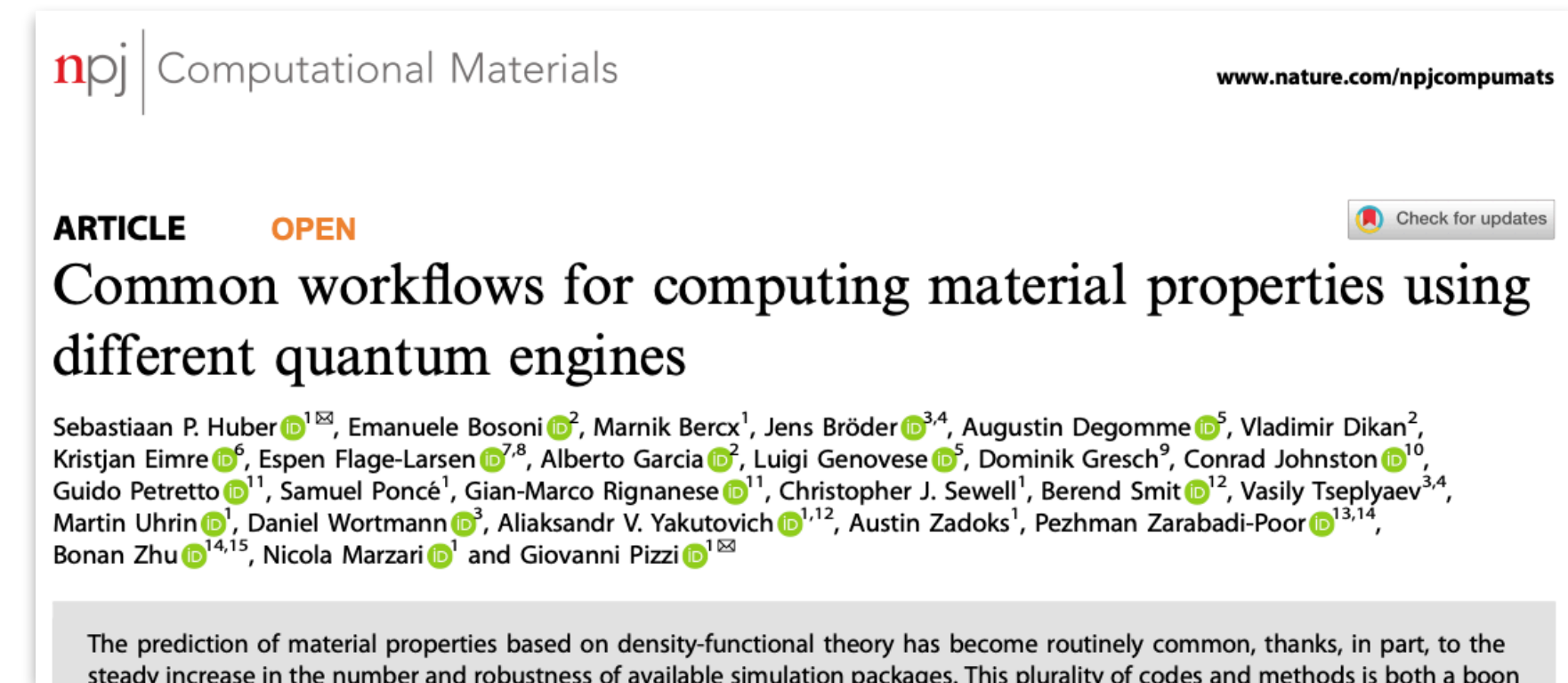
Interoperable workflows:

AiiDA common workflow interfaces (ACWF)

As a non-expert, be able to ask

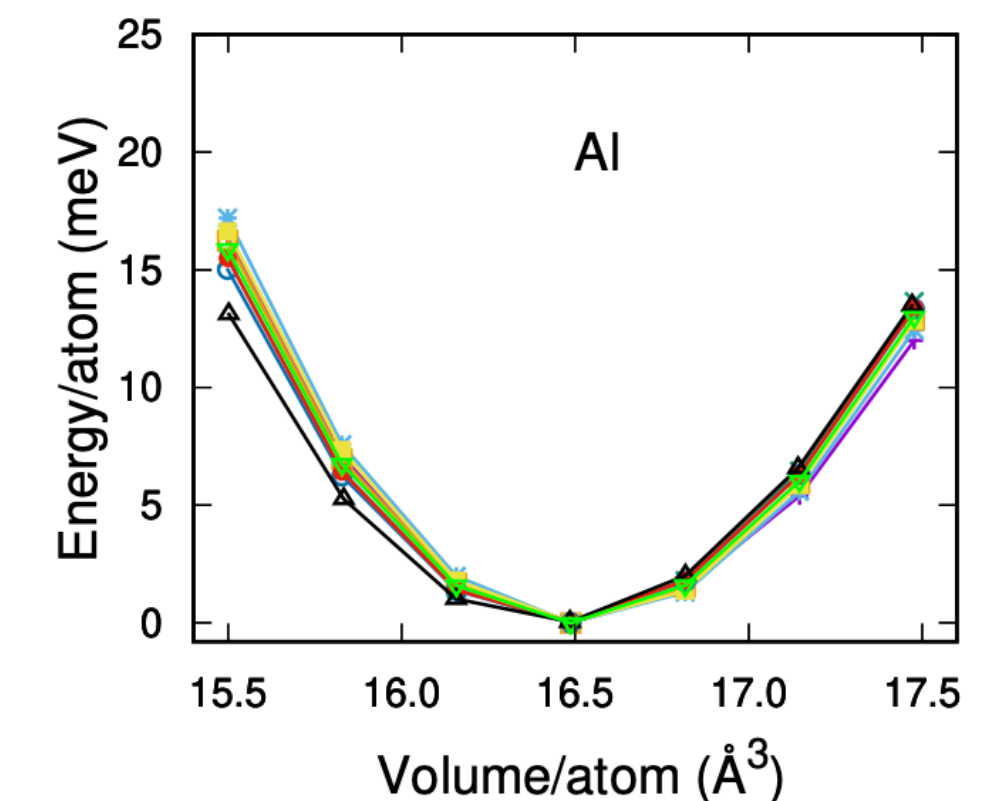
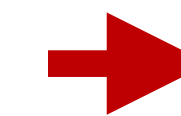
"Please run an **equation of state** with code **[Quantum ESPRESSO|SIESTA|VASP|...]** on the **XXX** supercomputer, using **YY** nodes, and **automatically choose numerical parameters** to get converged results."

- As an expert:
 - adapt the automatic parameters, if needed
 - check details of already-run simulations (by someone else): via provenance tracking



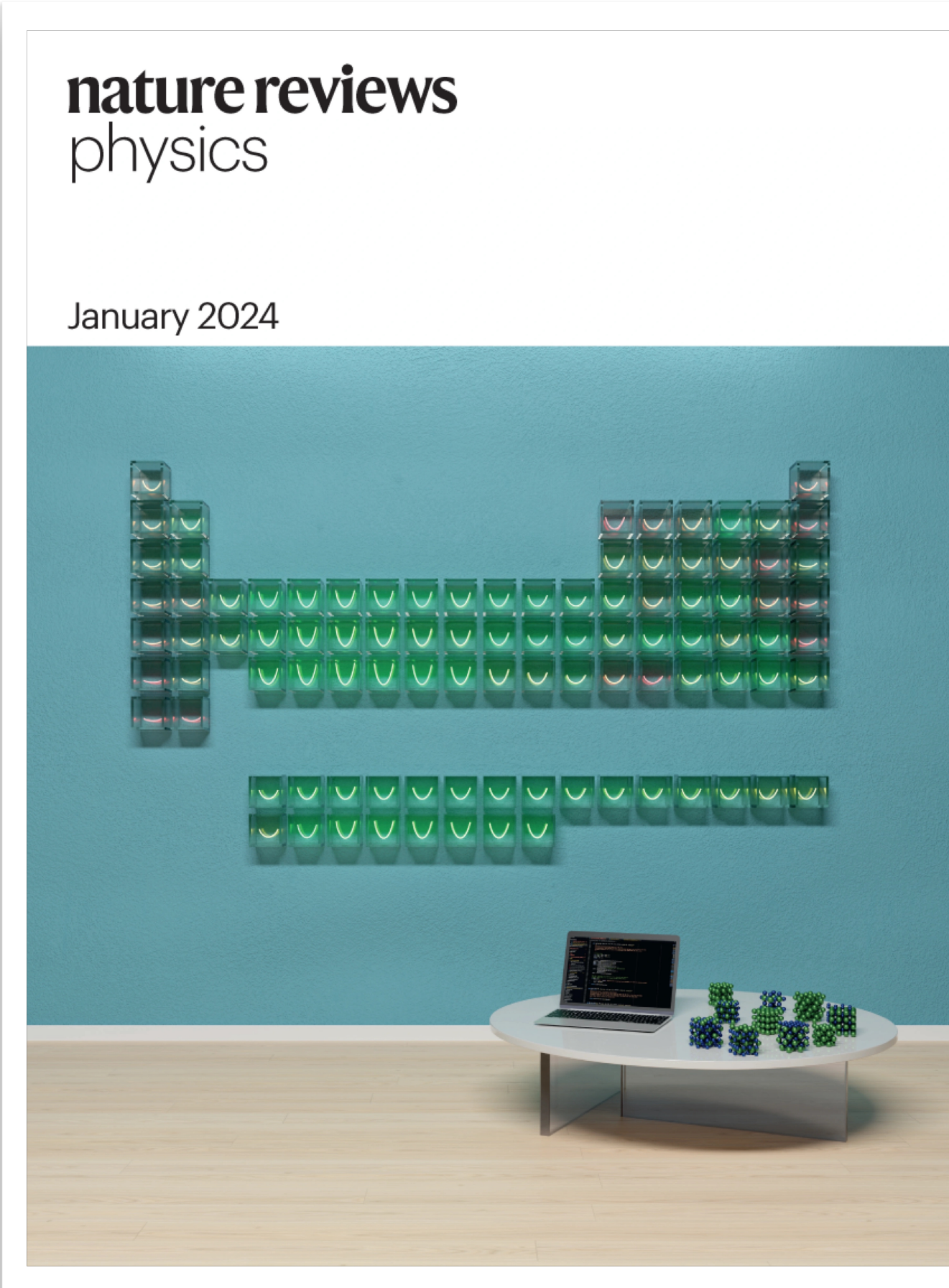
S. P. Huber et al., npj Comput. Mater. 7, 136 (2021)

\$ aiiida-common-workflows launch eos siesta --structure=Al --protocol=precise



<https://github.com/aiidateam/aiida-common-workflows/>

Enabling code verification via AiiDA common workflows



nature reviews physics

https://doi.org/10.1038/s42254-023-00655-3

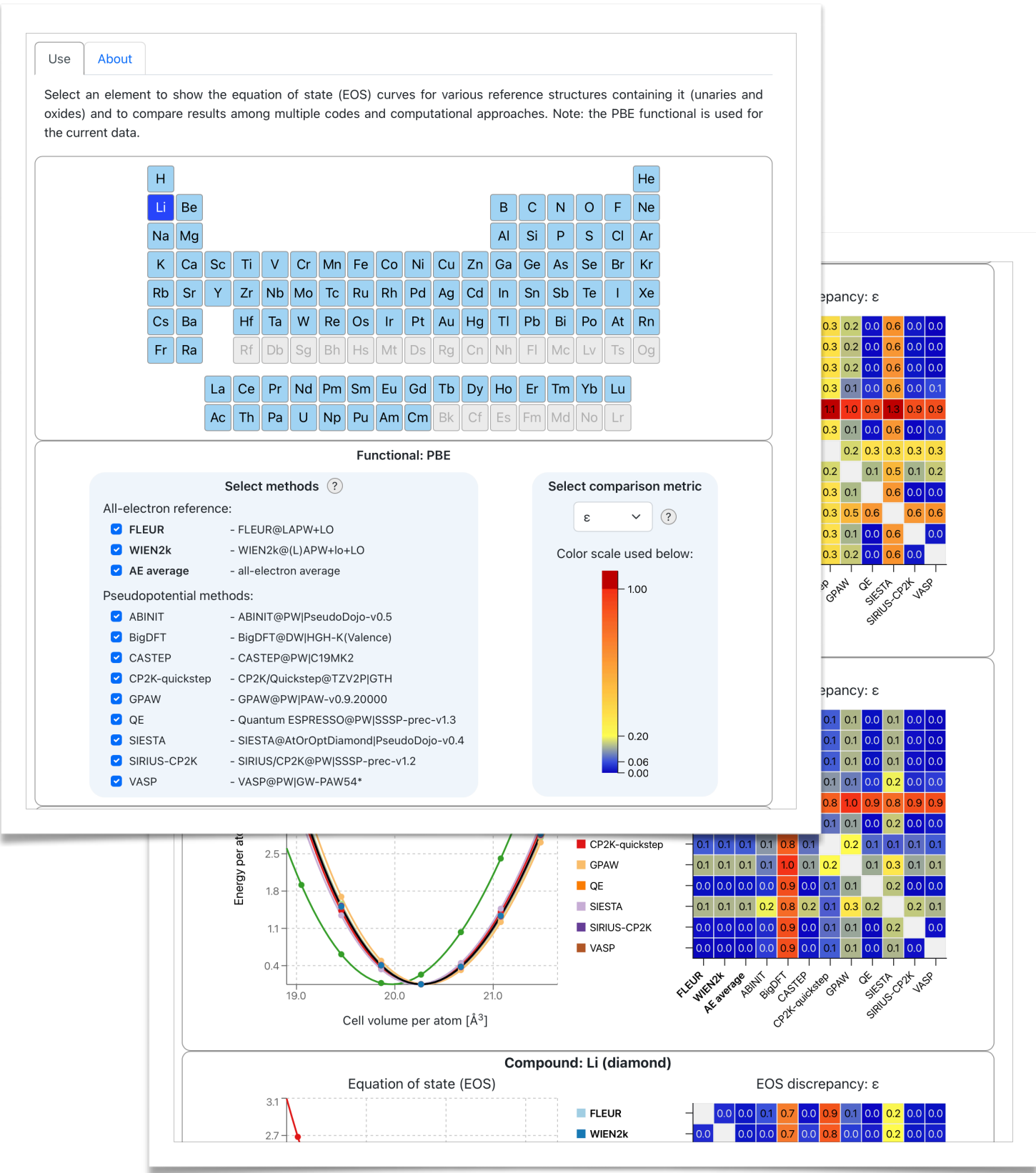
Expert recommendation

Check for updates

How to verify the precision of density-functional-theory implementations via reproducible and universal workflows

Emanuele Bosoni¹, Louis Beal², Marnik Bercx³, Peter Blaha⁴, Stefan Blügel⁵, Jens Bröder^{5,6}, Martin Callsen^{7,8,9}, Stefaan Cottenier^{7,8}, Augustin Degomme², Vladimir Dikan¹, Kristjan Eimre³, Espen Flage-Larsen^{10,11}, Marco Fornari¹², Alberto Garcia¹, Luigi Genovese², Matteo Giantomassi¹³, Sebastiaan P. Huber^{3,14}, Henning Janssen¹⁵, Georg Kastlunger¹⁵, Matthias Krack¹⁶, Georg Kresse^{17,18}, Thomas D. Kühne^{19,20}, Kurt Lejaeghere^{8,21}, Georg K. H. Madsen⁴, Martijn Marsman^{17,18}, Nicola Marzari^{3,16}, Gregor Michalíček⁵, Hossein Mirhosseini²², Tiziano M. A. Müller²³, Guido Petretto¹³, Chris J. Pickard^{24,25}, Samuel Poncé¹³, Gian-Marco Rignanese¹³, Oleg Rubel²⁶, Thomas Ruh^{4,8,27}, Michael Sluydts^{7,8,28}, Danny E. P. Vanpoucke^{7,29}, Sudarshan Vijay¹⁵, Michael Wolloch^{17,18}, Daniel Wortmann⁵, Aliaksandr V. Yakutovich³⁰, Jusong Yu^{3,16}, Austin Zadoks³, Bonan Zhu^{31,32} & Giovanni Pizzi^{3,16}

E. Bosoni *et al.*, Nat. Rev. Phys. 6, 45 (2024)



<http://acwf-verification.materialscloud.org>

Comparison of 11 codes and computational approaches
(*algorithms, basis sets, pseudopotentials, ...*)

Robustness is important (for code users)



Many efforts in the community: **more efficient, faster codes, new architectures** (GPUs, ...)

What about robustness? (guarantee to converge to solution): crucial, but (probably?) gets less attention

Robustness is important (for code users)

Many efforts in the community: **more efficient, faster codes, new architectures** (GPUs, ...)

What about robustness? (guarantee to converge to solution): crucial, but (probably?) gets less attention

Question: Would you buy a car that:

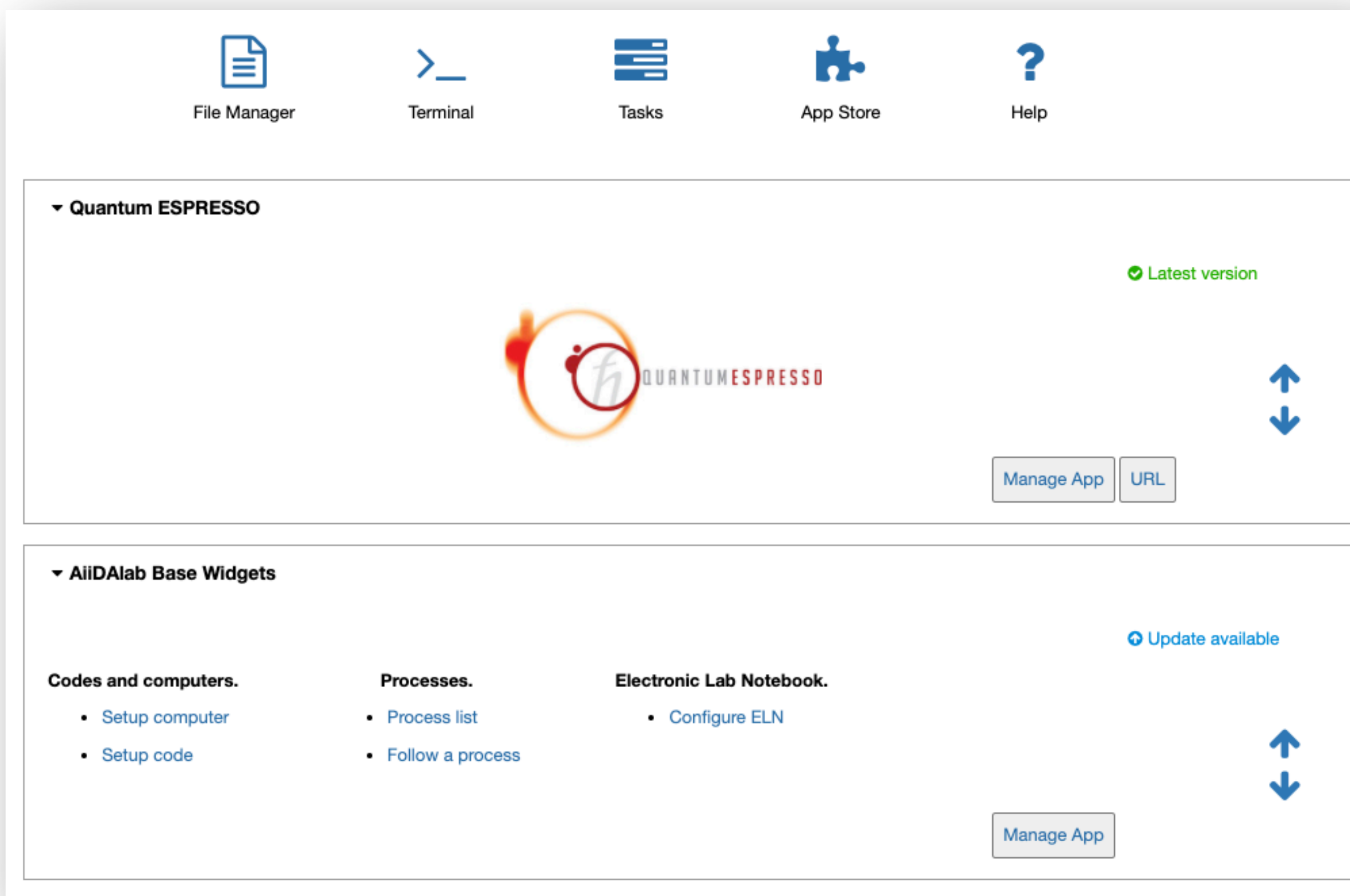
- costs 1/3; consumes 1/3;
- but randomly stops in the middle of the road 1/3 of your trips?
(Every other day!)





<https://www.aiidalab.net>

A. V. Yakutovich et al., Comp. Mat. Sci. 188, 110165 (2021)



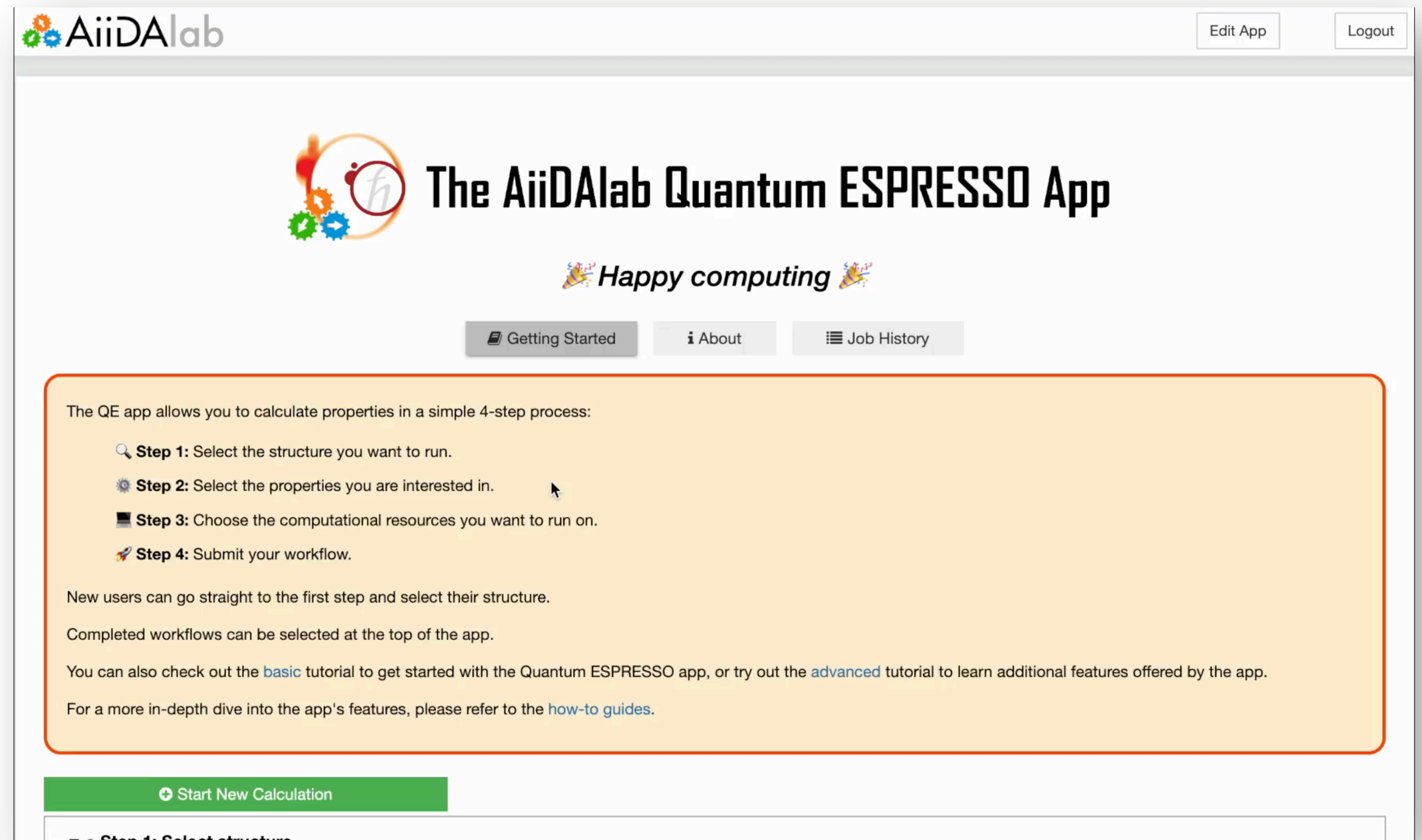
- **Jupyter(Hub)-based**, with AiiDA pre-configured
- **AppMode** to hide input cells, only show outputs as “web apps”
- Installable “Apps” with 1 click from an App Store
- Provide access to custom GUIs, making robust workflows accessible
- One example: **Quantum ESPRESSO app**

AiiDALab Quantum ESPRESSO app

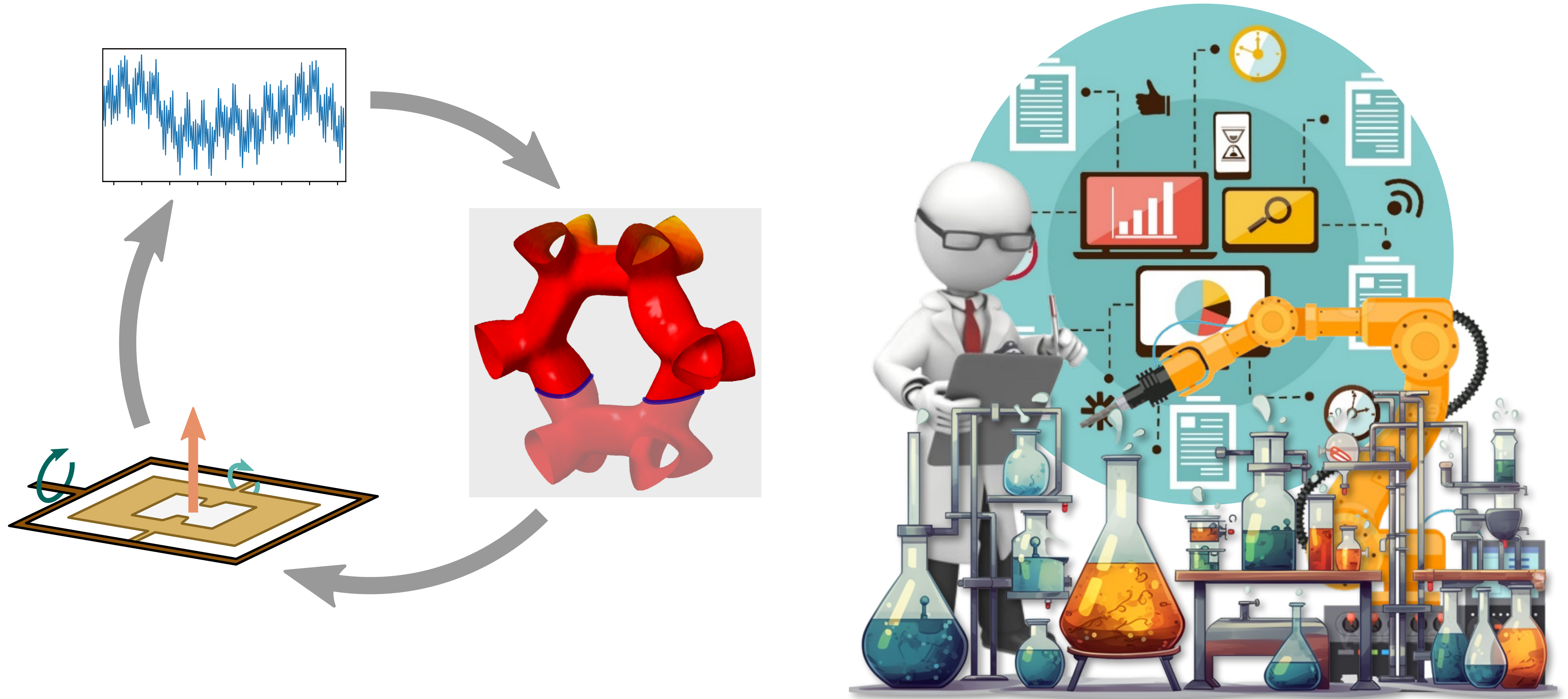


Choice of workflows/plugins driven by: existence of robust workflows, broad interest of community (PSI, Empa, ...), ...

<https://www.aiidalab.net>



Towards (FAIR) autonomous laboratories



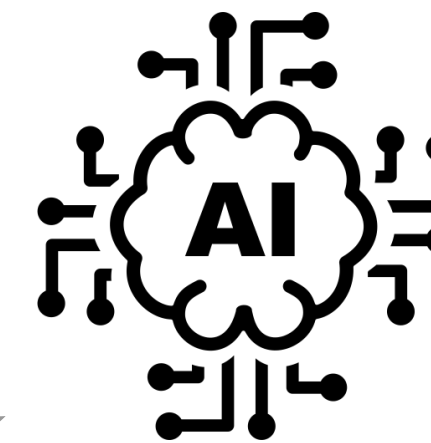
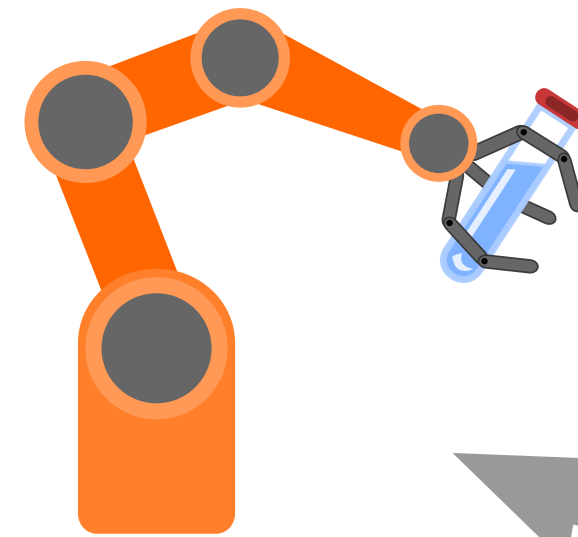
Enabling “FAIR-by-design” autonomous laboratories

The future of computational materials science:

Automated simulations
via robust workflows



Robotic
experiments

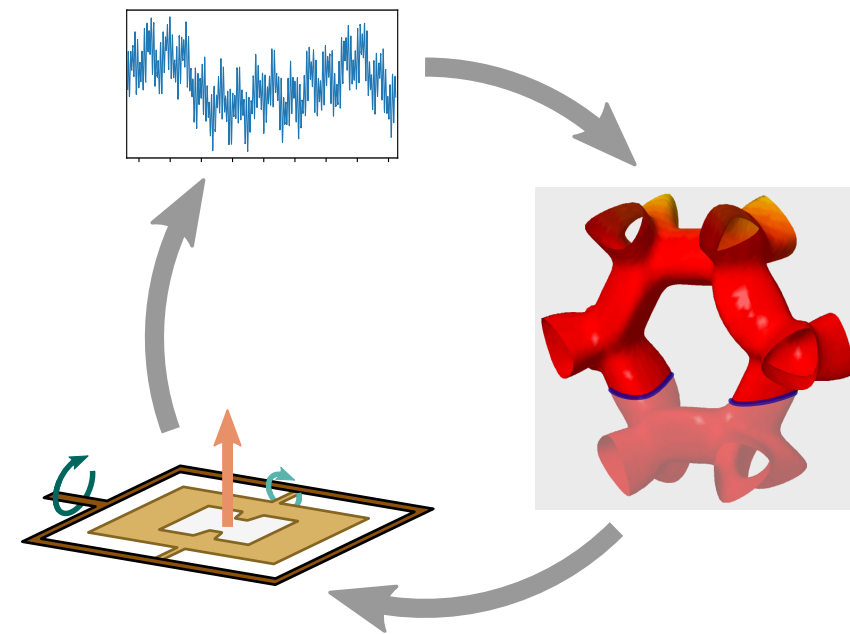


Artificial intelligence/ML
for autonomous decisions

Goal: **open-science platform for accelerated materials research**
enabling **seamless FAIR data collection/sharing “by design”**

“Discover” with autonomous workflows

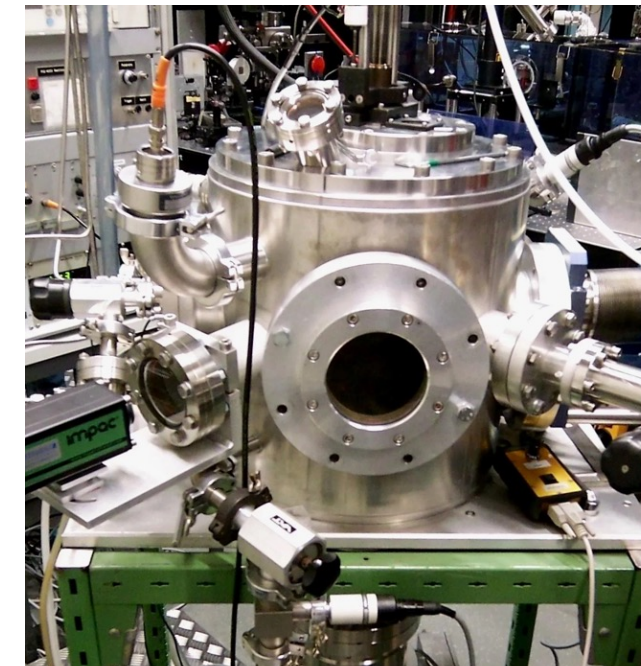
Ongoing projects:



Accurate Fermi-surface mapping (Shubnikov-de Haas)

Collaboration with Philip Moll
(MPG, Germany)

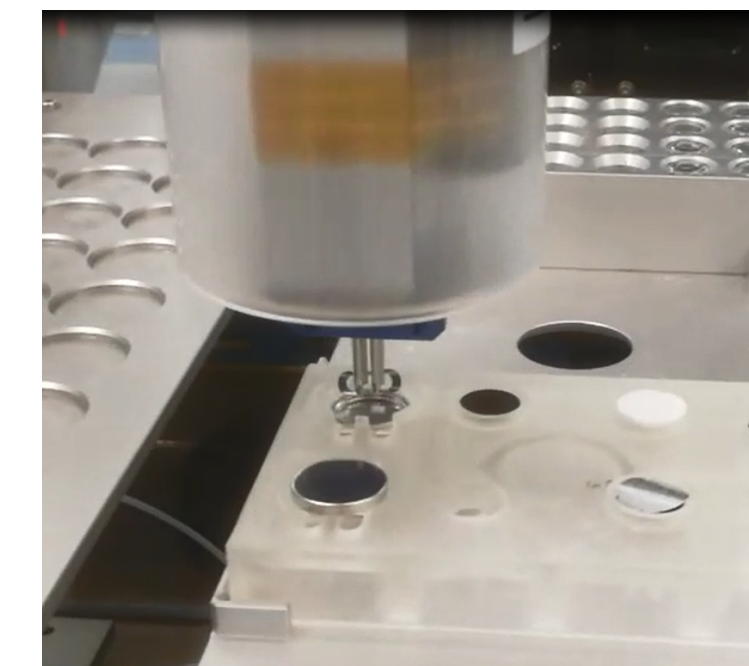
Goal: accelerate experiments
integrating with robotic 2-axis
SdH equipment



Pulsed laser deposition (autonomous crystal growth)

Collaboration with
Nikita Shepelin (PSI CNM)

Goal: accelerate identification
of ideal parameters for
crystalline growth



Battery assembly and testing

Collaboration within
BIG-MAP/Battery2030+

M. Vogler et al. **Matter** 6,
2647(2023); **ChemRxiv**
chemrxiv-2024-vfq1n (2024)

Collaboration with
Corsin Battaglia (Empa)

P. Kraus, ..., GP, **J. Mater. Chem. A**
12, 10773 (2024)

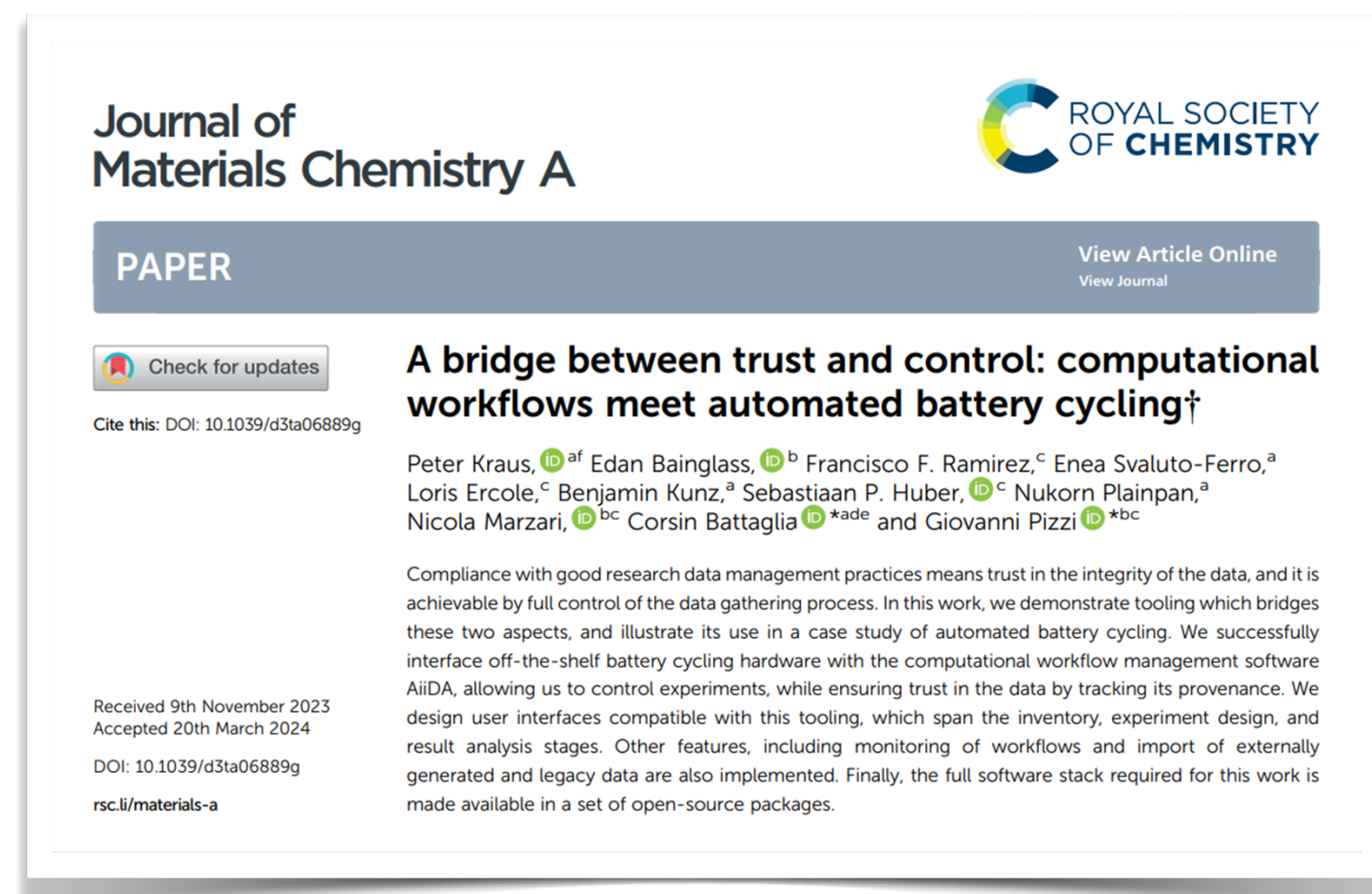
Goals: 1. accelerate optimization of
batteries (e.g. end-of-life)
2. Develop platform for autonomous
orchestration and data management

From experiment orchestration to FAIR digital twins

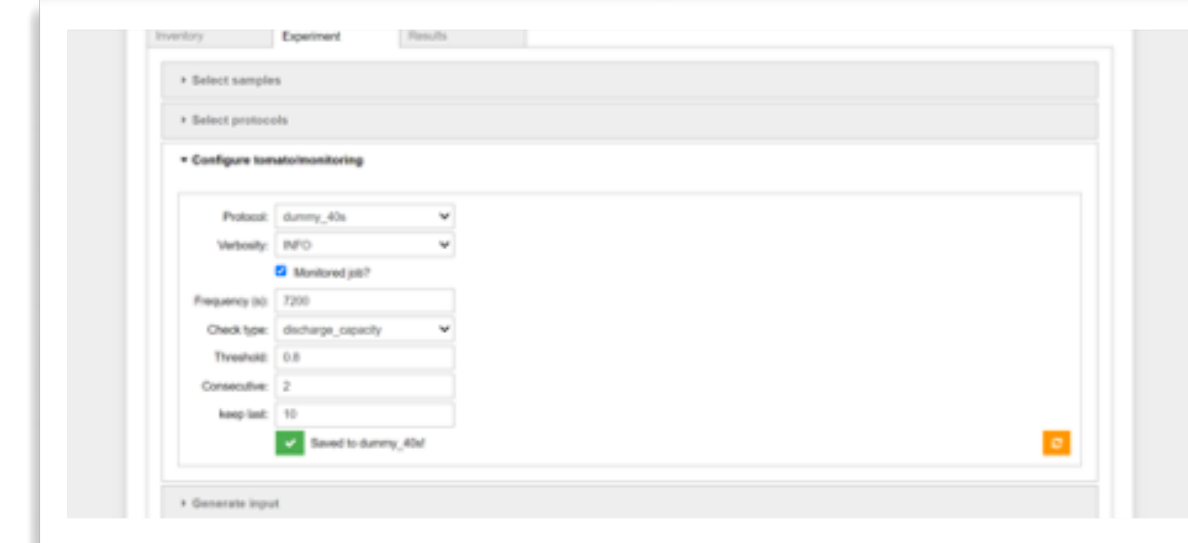
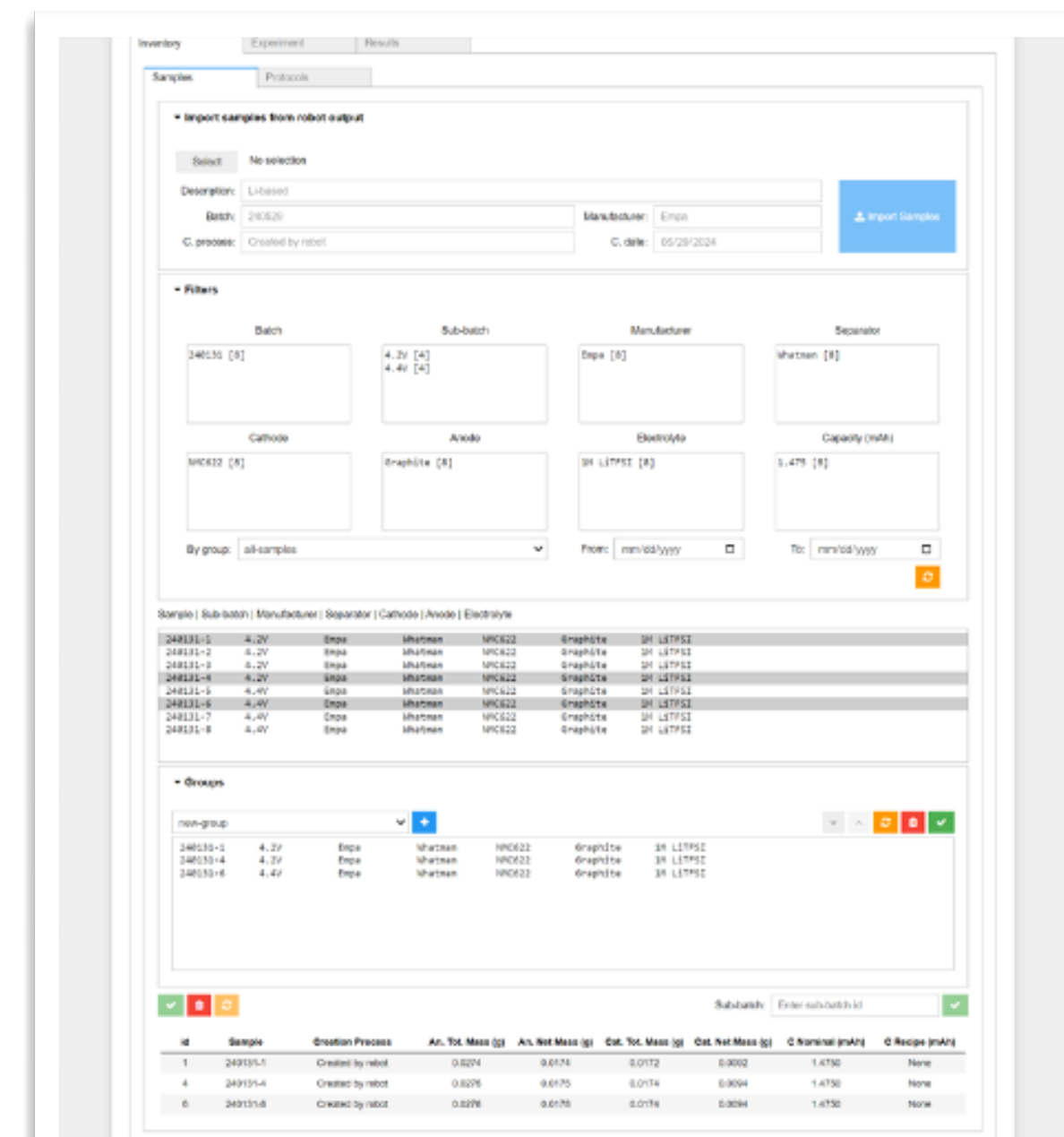


- **AiiDALab-Aurora integration**

- Collaboration with Battaglia's lab (Empa)
- Automated battery experiments, orchestrated via *AiiDA* (+ *tomato* [1])
- *AiiDALab GUI*: batch experiment submission + data analysis



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(co-funded by BIG-MAP)



From experiment orchestration to FAIR digital twins



- AiiDA

- Colla

- Auto
orch

- AiiDA
subn

Next step: integration with Bayesian algorithms for autonomous optimization of battery electrolyte formulation

Ultimate goal (Research): accelerate design of improved batteries

Ultimate goal (Platform): Seamless data interoperability (experiments and simulations) and autonomous platform “FAIR-by-design”

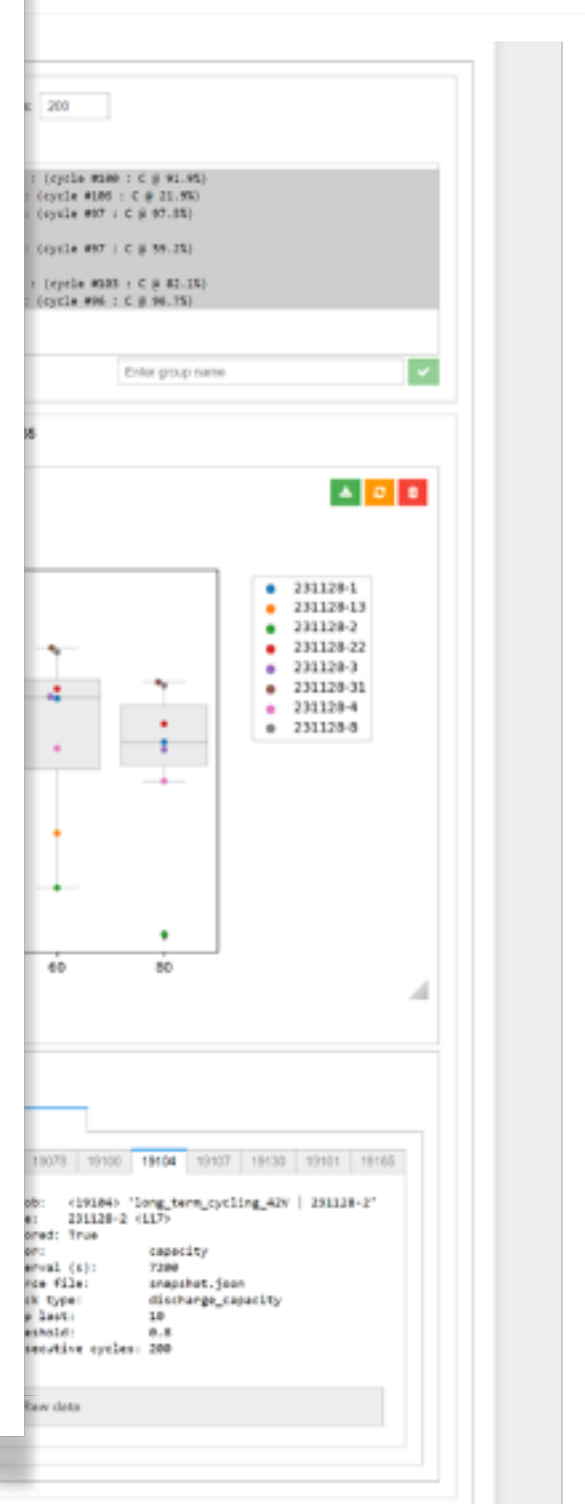
See more on the PREMISE website

<https://ord-premise.org/deliverables/>



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- Many (but not all!) researchers **care about maximizing output** (papers/citations/faster tasks)
- **Tool availability** (ELNs, workflow managers, ...) (if well designed!) **can boost FAIR adoption**
- Design **tools around researchers' needs**:
 - Not all researchers are professional coders: **make it easy, clarify immediate benefits**
 - Clarify **target audience** (to you and stakeholders)! Limit use cases to avoid “failure”
E.g. *simulation experts vs experimentalists*; but also target technical goals:
performance vs. easy-of-use vs. reproducibility vs. dynamic workflows vs. ...
 - If you have a team: balance RSEs with background/**experience** in the field (important!) with **professional developers**, and encourage mutual learning
 - **RSE role** still far from being consolidated: tension between “doing research” and robust software development (also for funding)

Acknowledgements: funding



SNSF NCCR “MARVEL”

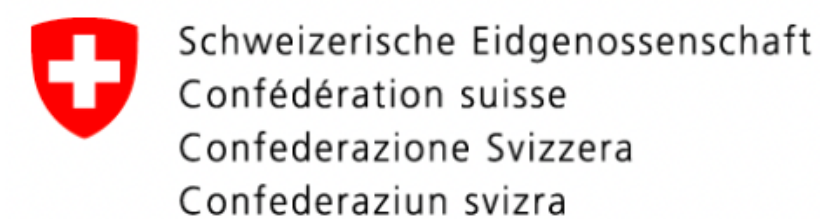
Discovery of new materials via simulations and dissemination of curated data



H2020 Centre of Excellence “MaX”

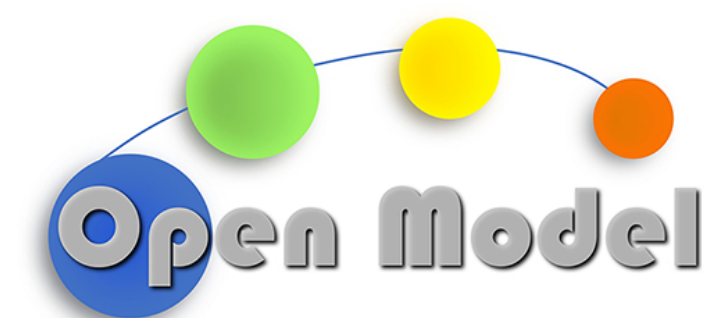
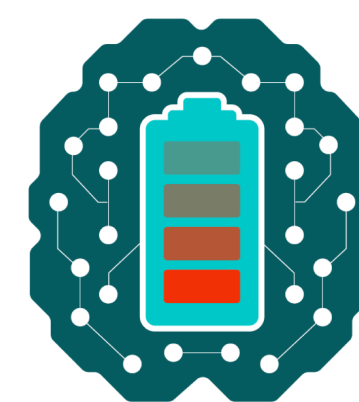
Scaling towards exascale machines and high-throughput efficiency

Moreover:



State Secretariat for Education,
Research and Innovation SERI

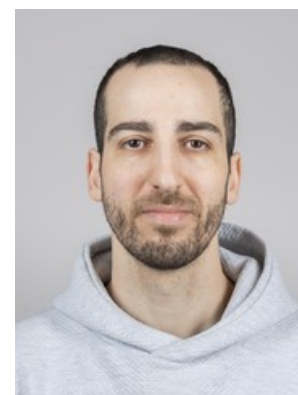
BIG-MAP



Acknowledgements: AiiDA, AiiDALab, Materials Cloud teams



The current AiiDA, AiiDALab
and Materials Cloud teams



Edan
Bainglass
(PSI)



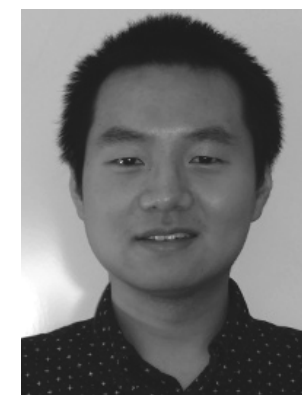
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(EPFL)



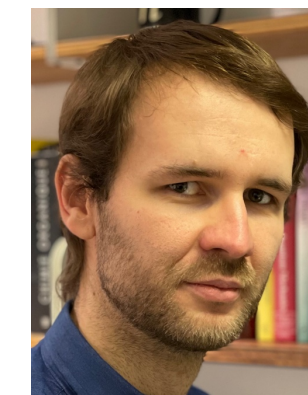
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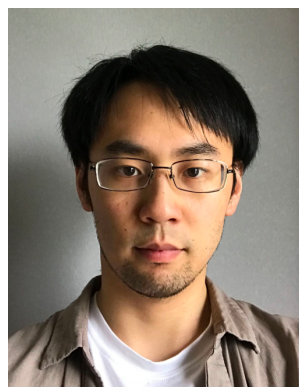
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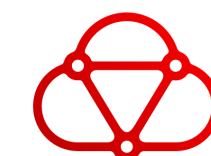
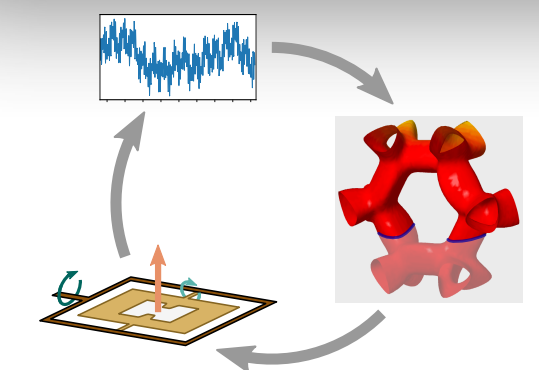
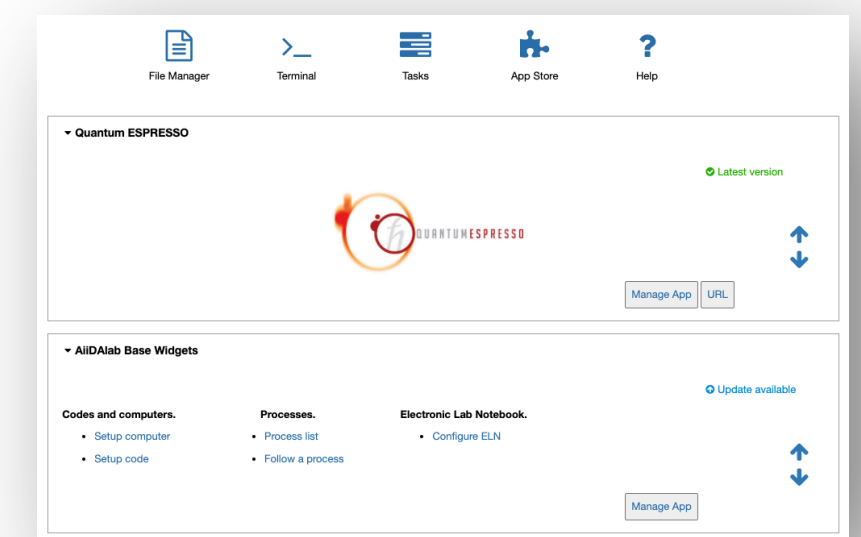
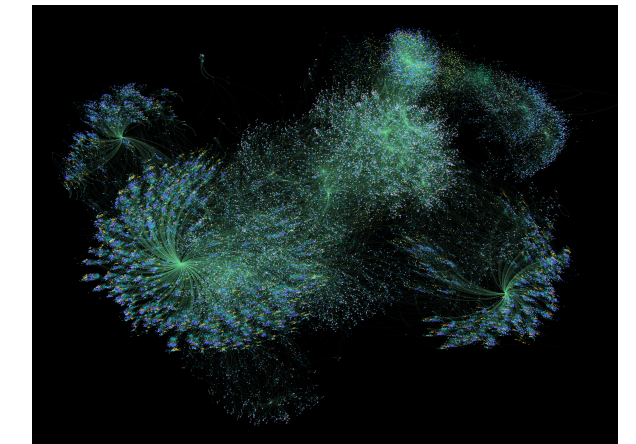
And all former team members, who actively contributed with ideas and code
to make these platforms what they are today:

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- Tiziano Müller
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- Riccardo Sabatini
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- Berend Smit
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- Spyros Zoupanos
- ...

Summary



- AiiDA+Materials Cloud: **automated simulations** and **FAIR** access to simulation data
- **Plugin ecosystem + common workflow interfaces** for code interoperability
- **Robustness** is crucial for users (not only performance)
- Beyond FAIR data to **FAIR workflows**: robust turn-key workflows, **made accessible** via AiiDAlab GUIs
- Ontologies can enable **FAIR-by-design autonomous labs**



MATERIALSCLOUD

