

Automated high-throughput Wannierisation

Implementation of the entangled SCDM-k method in Quantum ESPRESSO and full automation with AiiDA workflows

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- (2) Cavendish Laboratory, Department of Physics, University of Cambridge, UK
 - (3) Departments of Materials and Physics, and the Thomas Young Centre for Theory and Simulation of Materials, Imperial College London, UK
- (4) Department of Materials, University of Oxford, Parks Road, Oxford OX1 3PH, UK

Hunting for projections

- Usually, code needs user to specify initial projections (guesses for the minimisation procedure)
- This needs a lot of chemical understanding and experience. Biggest challenges for new users, and very hard to automate

Some recent emails from the Wannier90 mailing list:

Dear Experts, How can I define the correct projection of particular material? [...]

Dear Sir,
I need to know the correct projection of
Graphene for a converged wannier
calculation. [...]

Dear Wannier Community, [...]

My question is how do I define three projections for the half-filled p states of the two As atoms?

Hunting for projections

- Usually, code needs user to specify initial projections (guesses for the minimisation procedure)
- This needs a lot of chemical understanding and experience. Biggest challenges for new users, and very hard to automate
- Recently: **SCDM** method (selected columns of the density matrix) [1,2] proposed, aiming at automatically finding Wannier functions

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Dear Wannier Community,
[...]
My question is how do I define three projections for the half-filled p states of the

two As atoms?

[1] Damle, A., Lin, L. & Ying, L. Compressed representation of Kohn–Sham orbitals via selected columns of the density matrix. Journal of Chemical Theory and Computation 11, 1463–1469 (2015).

[2] Damle, A. & Lin, L. Disentanglement via entanglement: A unified method for Wannier localization. *Multiscale Modeling & Simulation* 16, 1392–1410 (2018).

Aim and outcome

• Implementation of the SCDM method in Quantum ESPRESSO (pw2wannier90.x code)

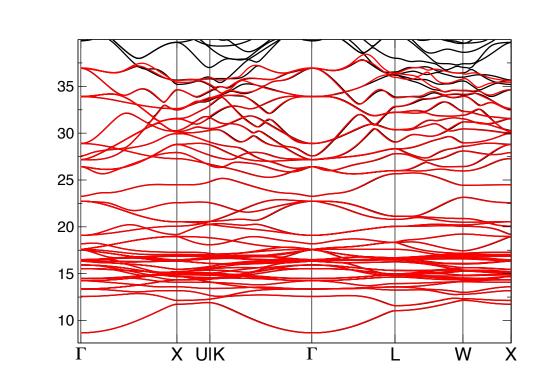


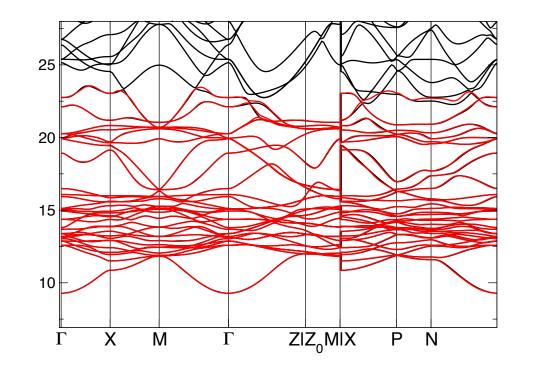
- Implementation of automatic workflows (within AiiDA) for fully automatic
 Wannierisation
- Validation on a set of ~200 crystalline materials (~80 insulators) covering the chemical space

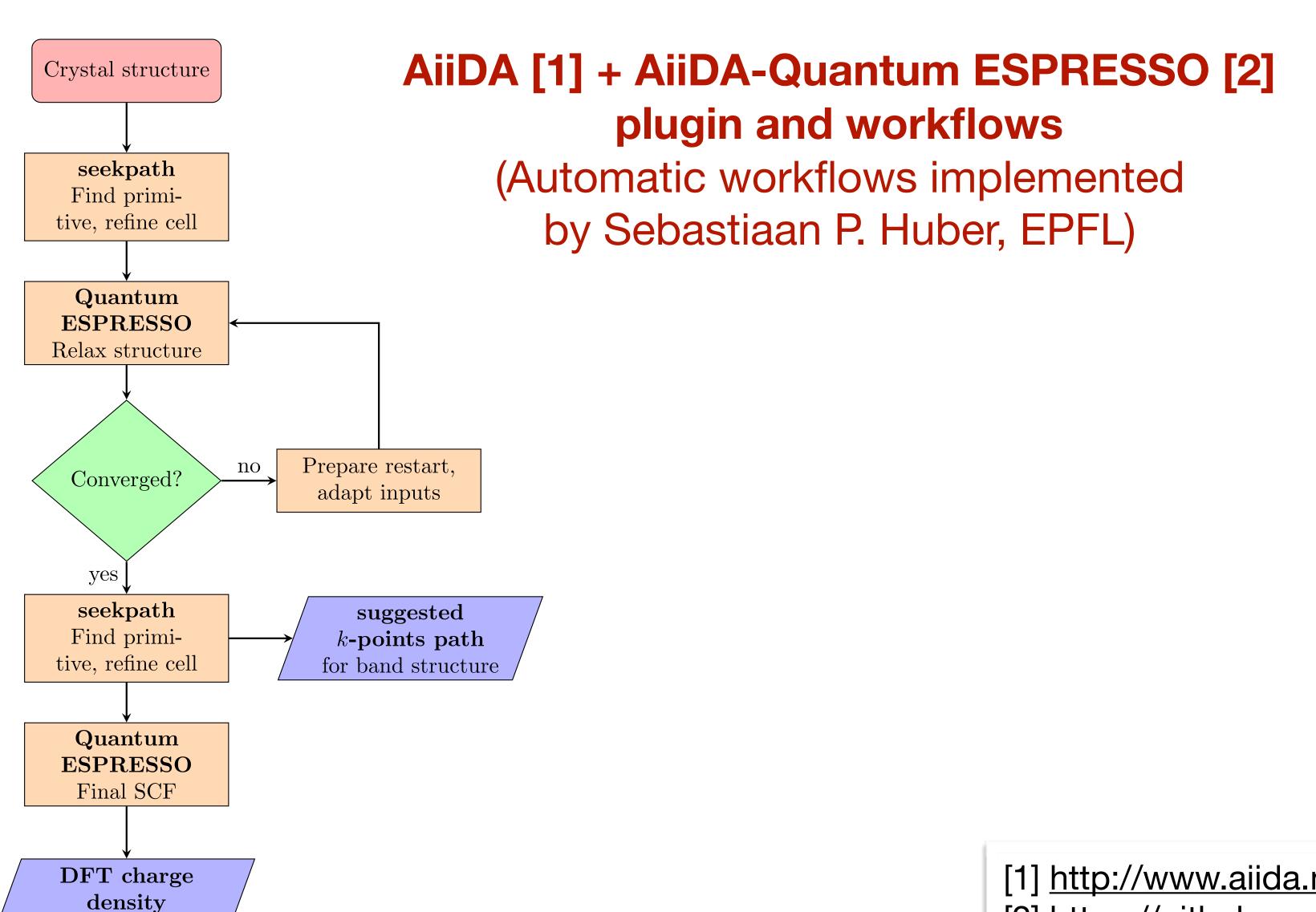


 N, μ, σ



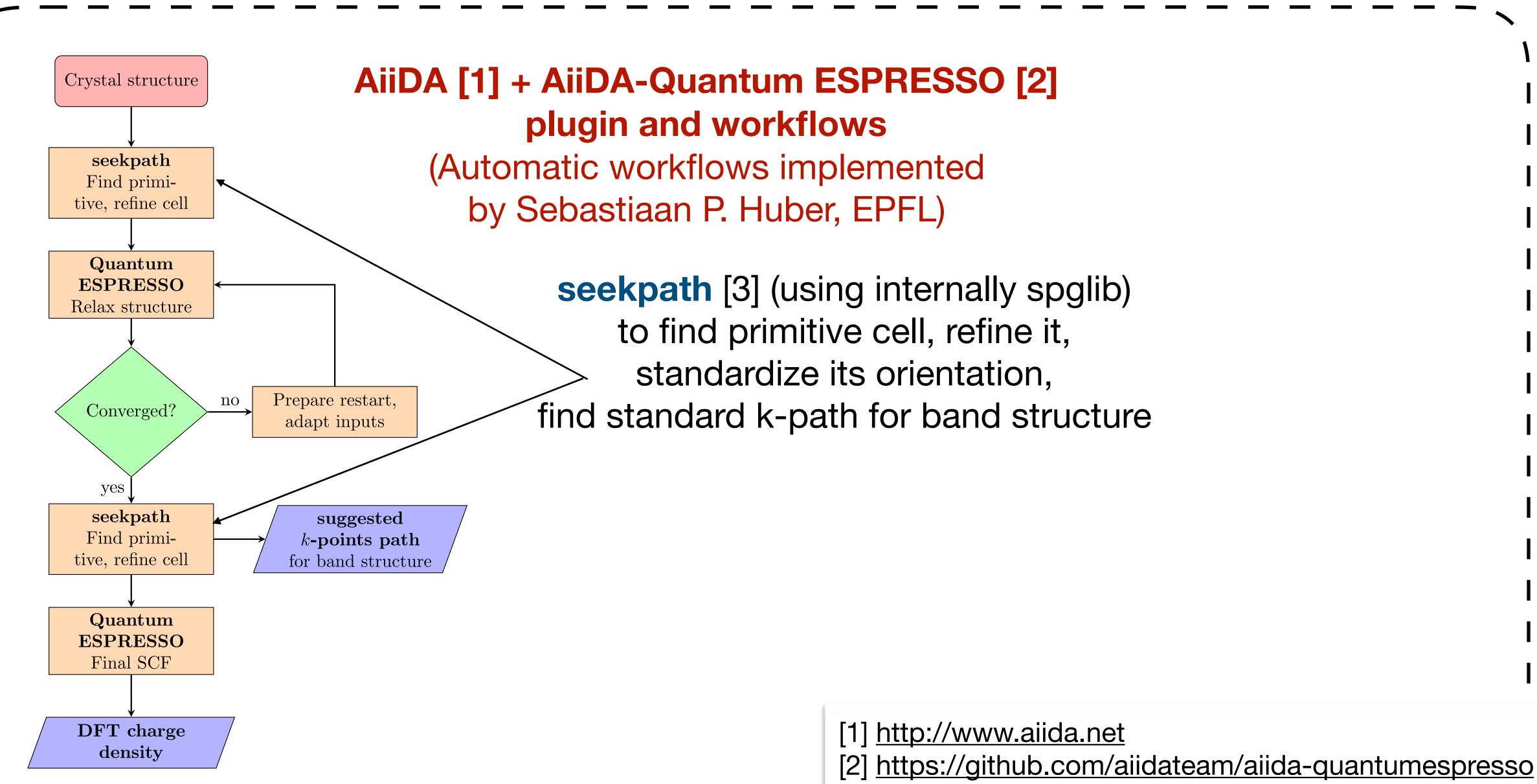


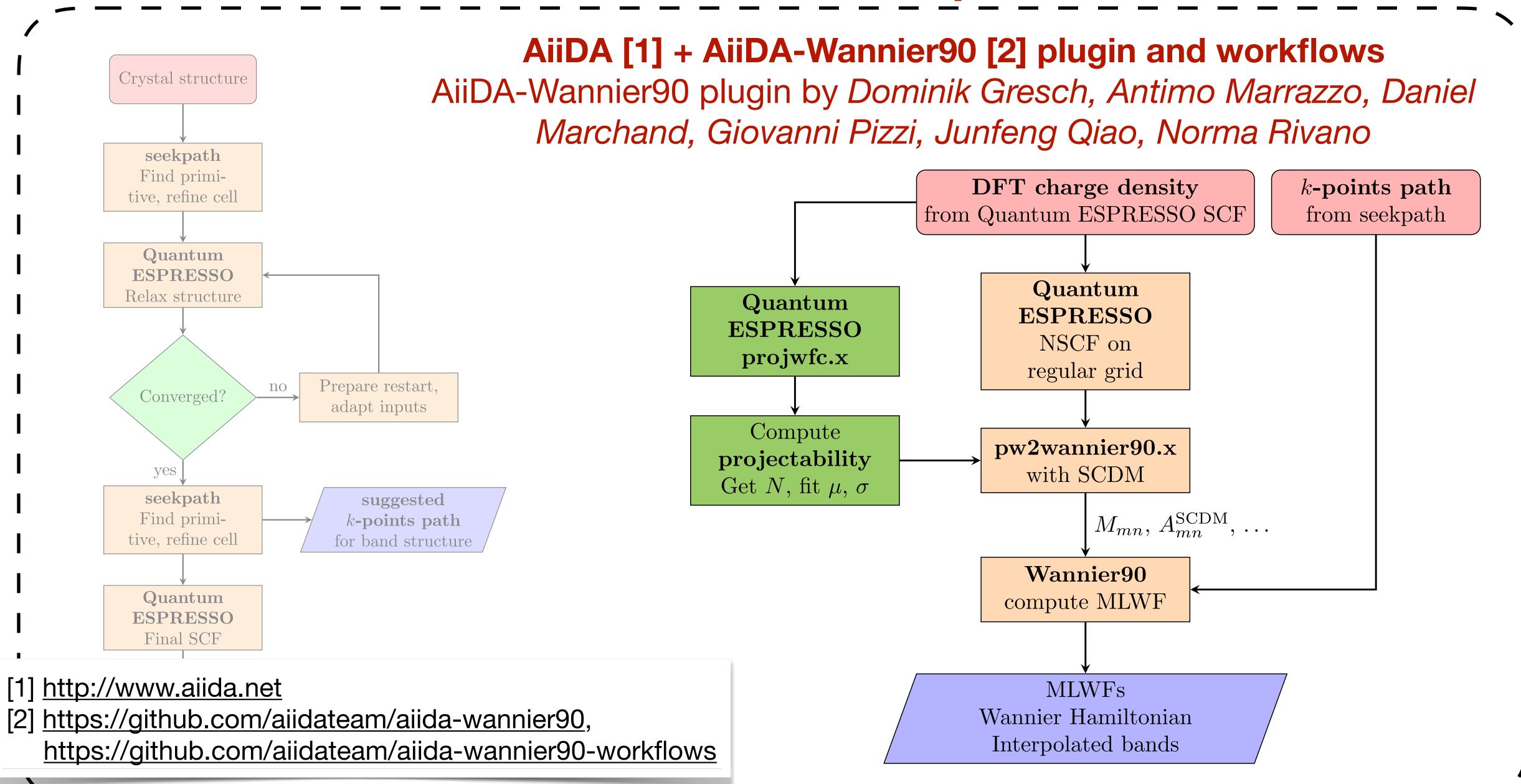


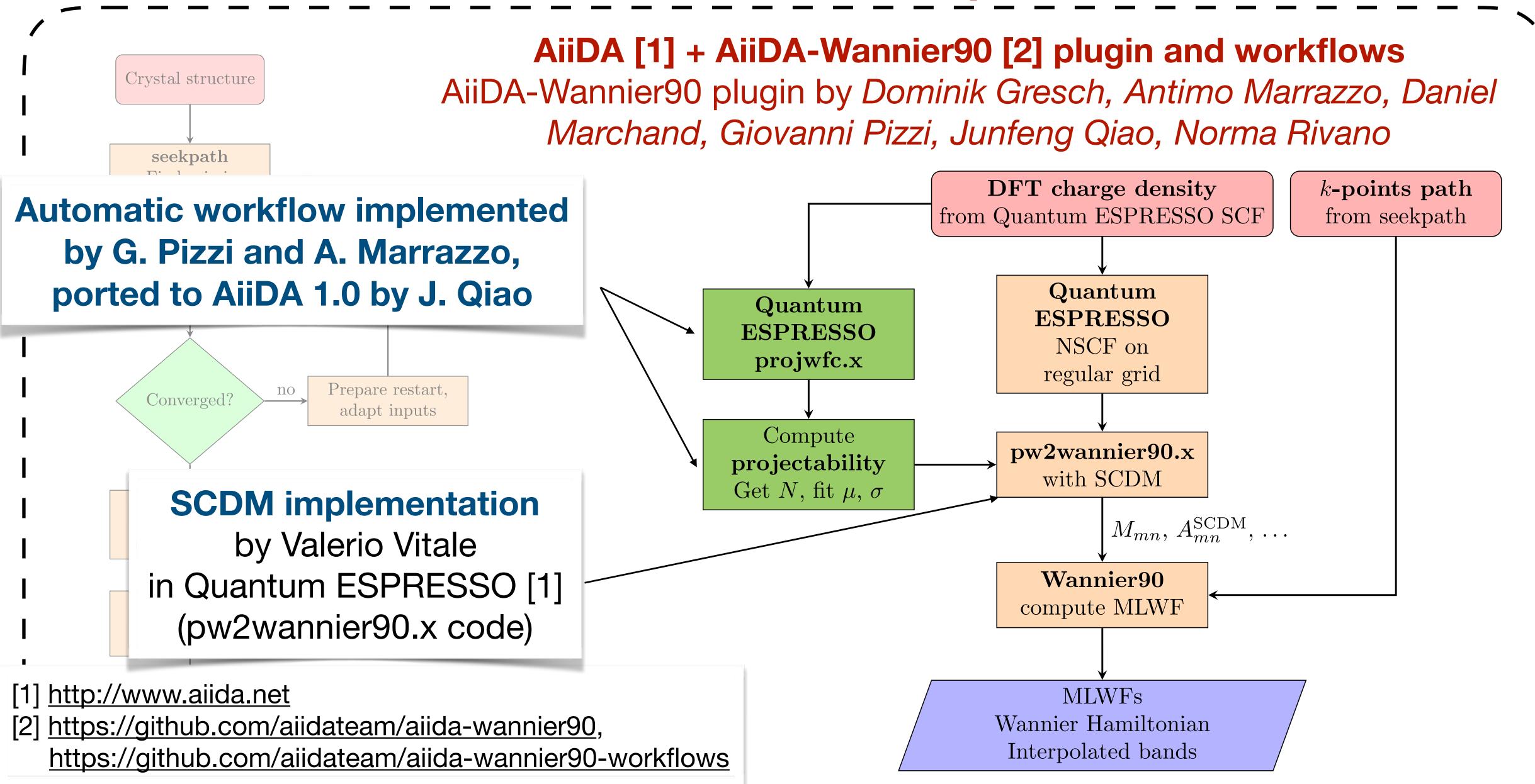


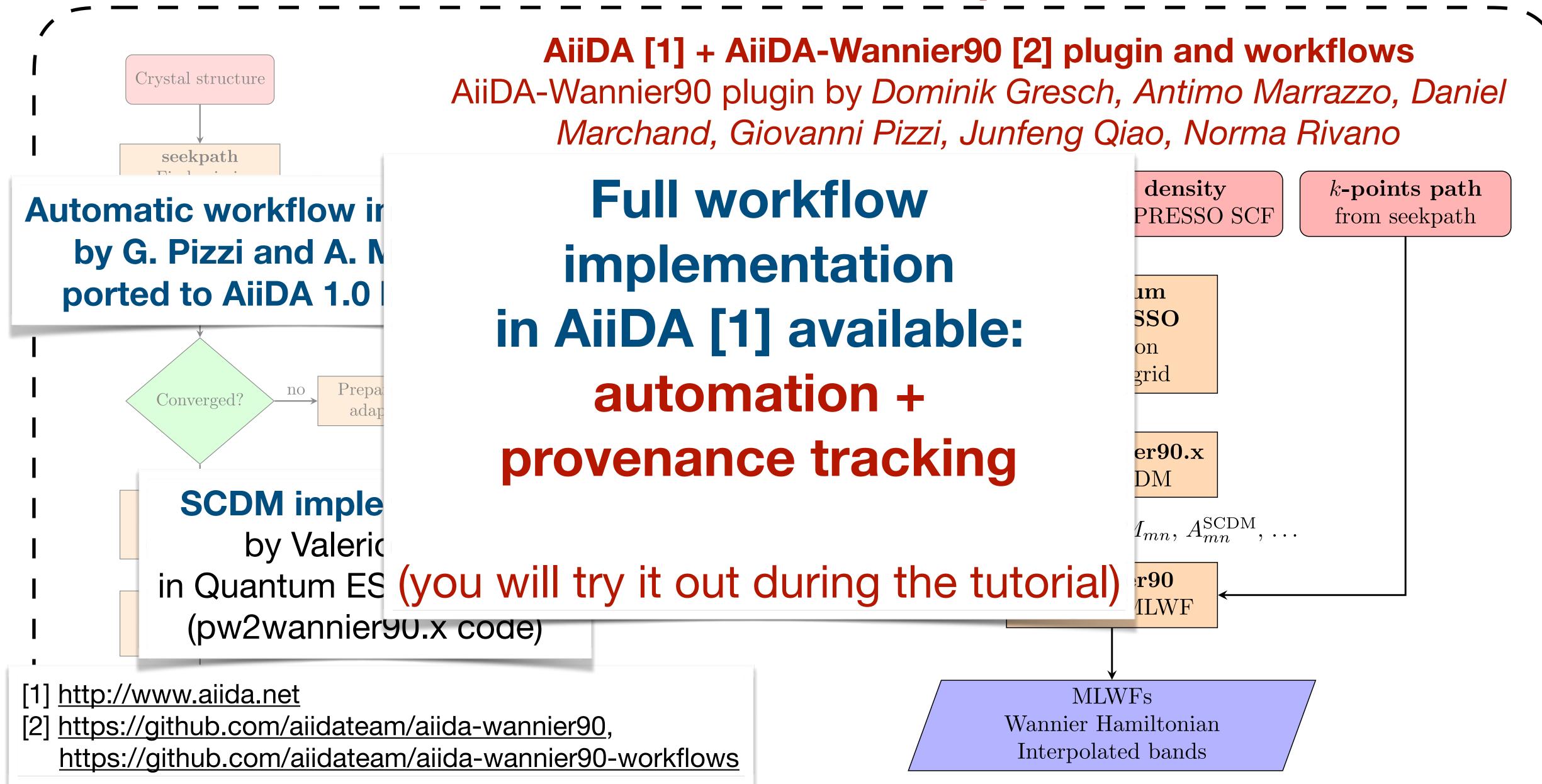
- [1] http://www.aiida.net
- [2] https://github.com/aiidateam/aiida-quantumespresso
- [3] https://www.materialscloud.org/work/tools/seekpath

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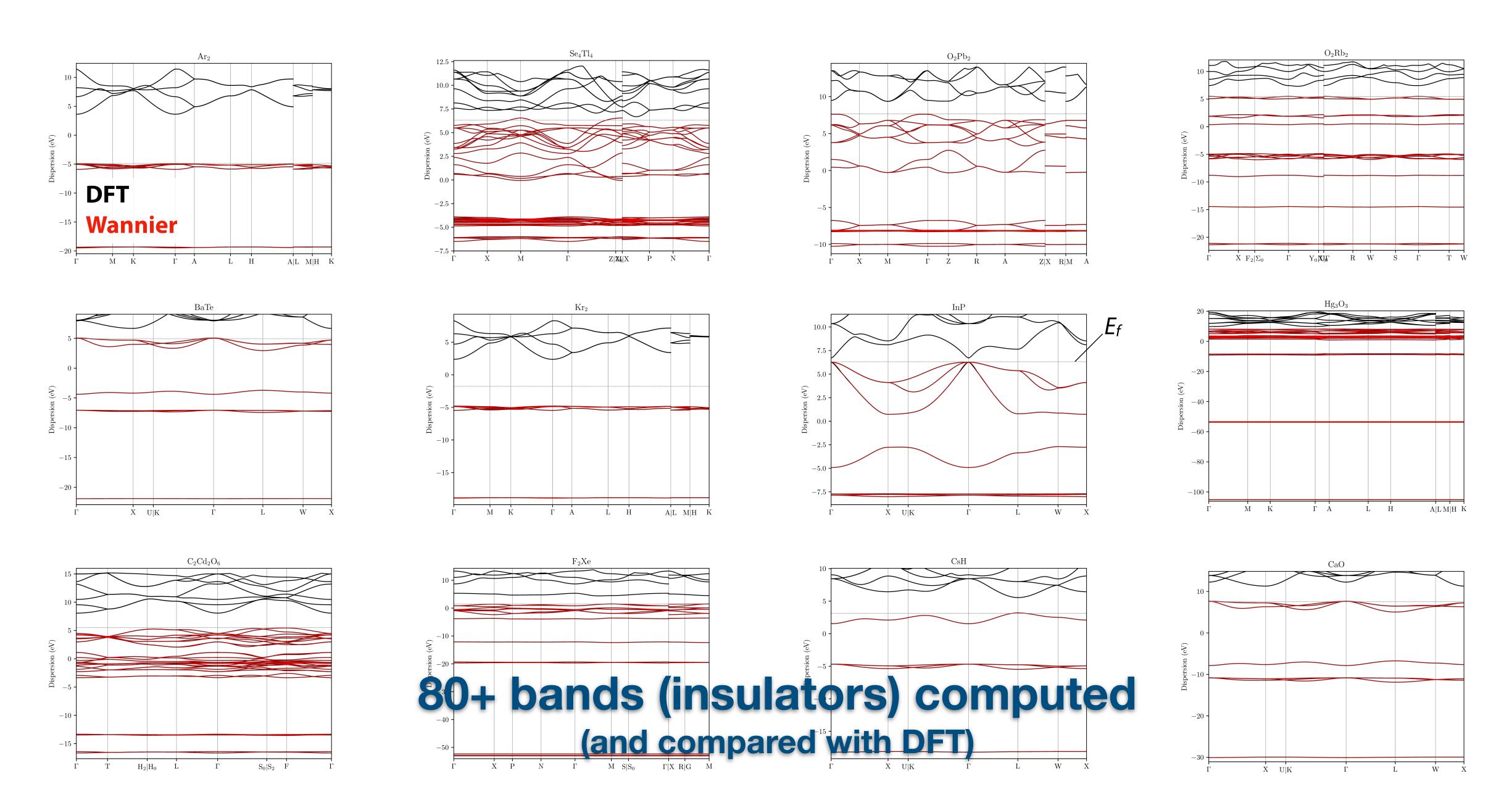








Insulators: (subset of the) results



Bands distance

 To assess quality of Wannierisation and interpolation: we define a **bands distance** (between DFT bands and interpolated bands)

$$\eta = \sqrt{\sum_{n\mathbf{k}} \left(\varepsilon_{n\mathbf{k}}^{\mathrm{DFT}} - \varepsilon_{n\mathbf{k}}^{\mathrm{Wan}}\right)^{2}}, \qquad \qquad \eta^{\mathrm{max}} = \max_{n\mathbf{k}} \left(\left|\varepsilon_{n\mathbf{k}}^{\mathrm{DFT}} - \varepsilon_{n\mathbf{k}}^{\mathrm{Wan}}\right|\right)$$

Average bands distance

Max bands distance

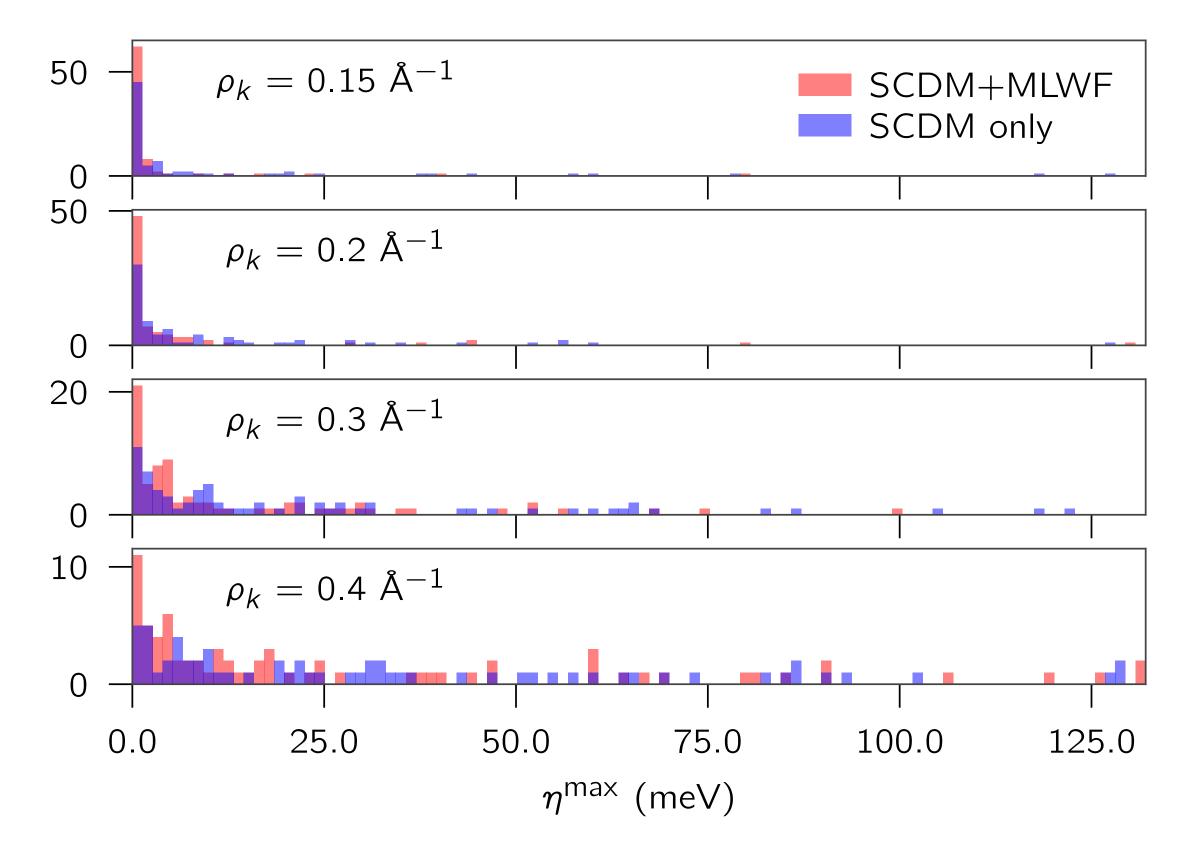
• Moreover, we want to assess the importance of the **density of** k**-points** in the NSCF/Wannierisation step We will use a linear density ρ_k in Å⁻¹

Comparison of SCDM + MLWF with SCDM only (80 insulators)

Max bands distance

$\rho_k = 0.15 \text{ Å}^{-1}$ 50 SCDM+MLWF 14x14x14 in bulk Si SCDM only 50 $\rho_k = 0.2 \, \text{\AA}^{-1}$ 11x11x11 in bulk Si 20 $\rho_k = 0.3 \, \text{\AA}^{-1}$ 7x7x7 in bulk Si $\rho_k = 0.4 \text{ Å}^{-1}$ 6x6x6 in bulk Si 0.0 2.5 5.0 15.0 10.0 12.5 20.0 η (meV)

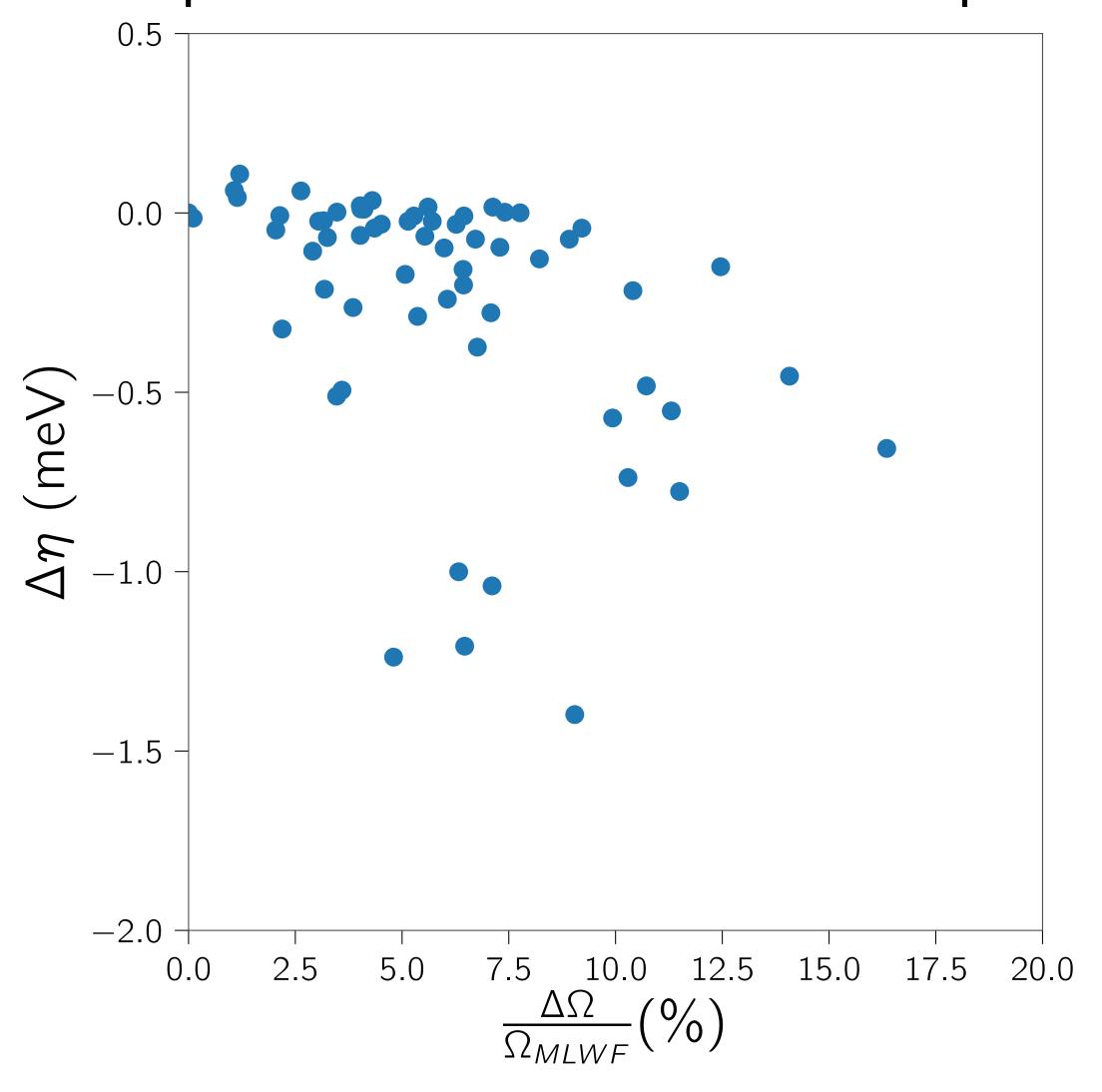
Average bands distance



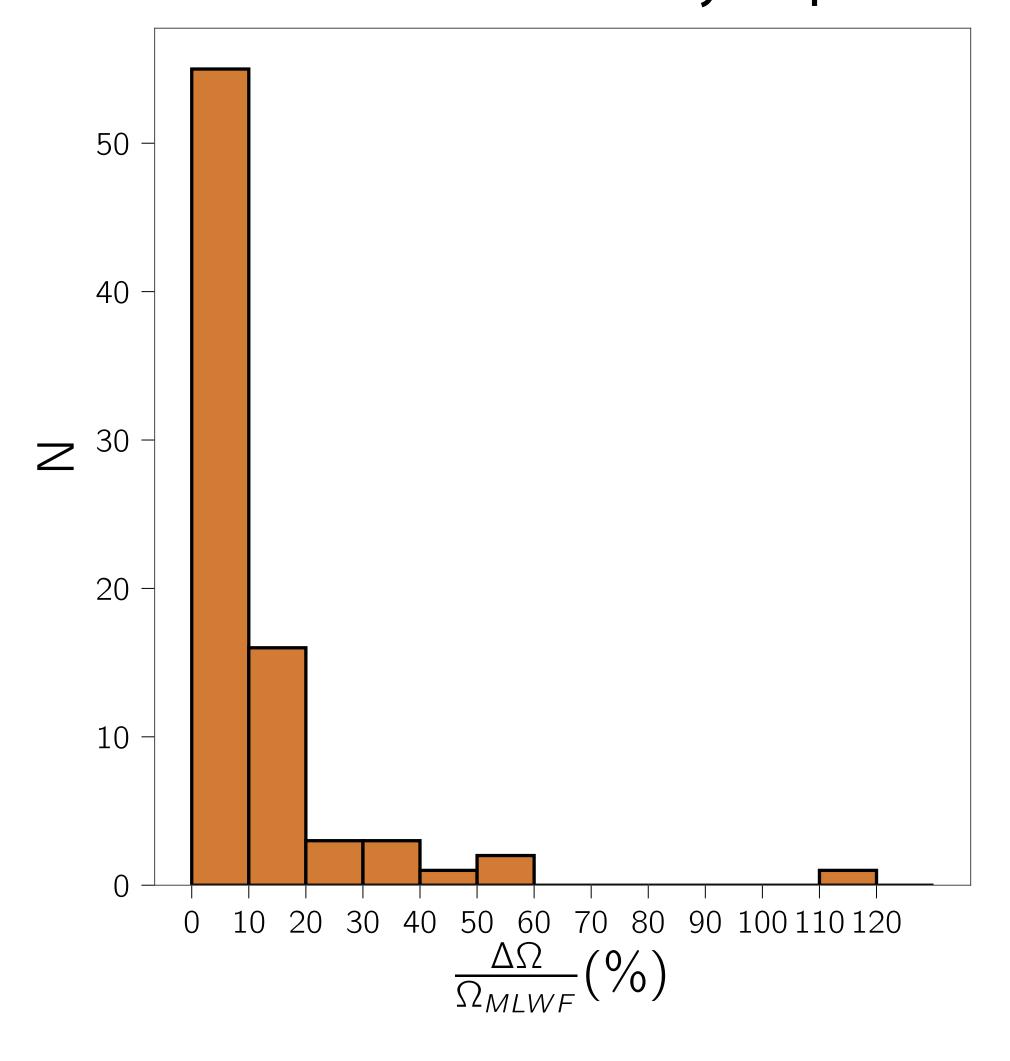
- Good results require a density of at least 0.2 Å-1 or more dense
- For insulators, SCDM-only already provides very good results; MLWF improves them
- In general, very small band distances (i.e. very good interpolation)

Comparison of SCDM + MLWF with SCDM only (80 insulators)

 Improvement of bands distance with MLWF on top of SCDM visible also from this plot



Spread reduction typically of ~10-20%:
 SCDM for insulators already is quite localised

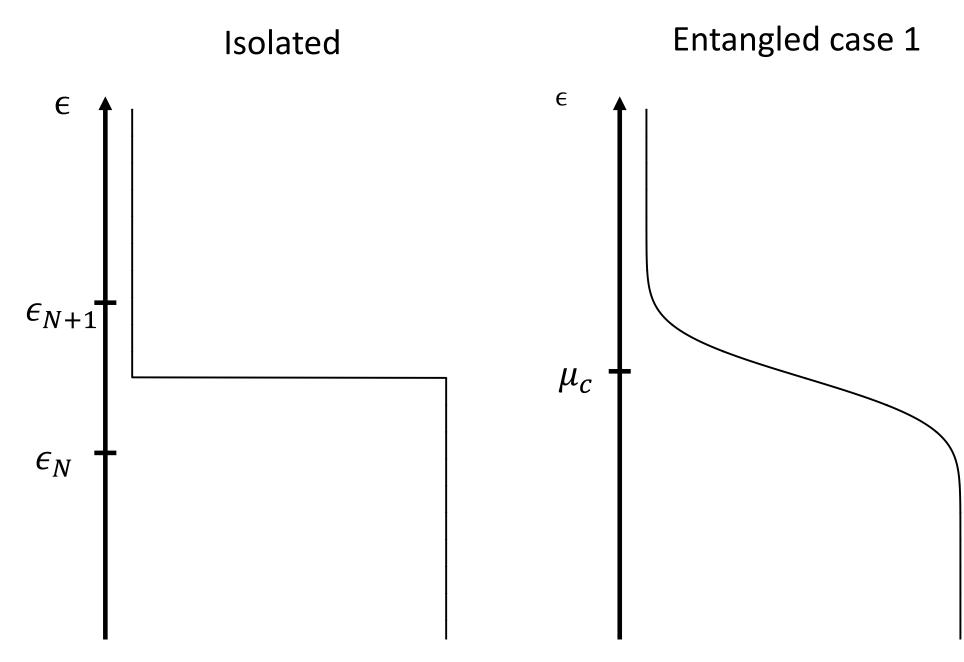


Entangled bands

• We consider (formally) all eigenstates, and give a weight in the quasi-density-matrix P

$$P = \sum_{i} |\psi_{i}\rangle f(\varepsilon_{i})\langle\psi_{i}| = f(H)$$

- *f:* smooth function of energy, selecting relevant states. If *f* is smooth: *P*(**r**,**r**') decays rapidly [2]
- We select the most N_w representative columns; procedure is analogous to isolated case



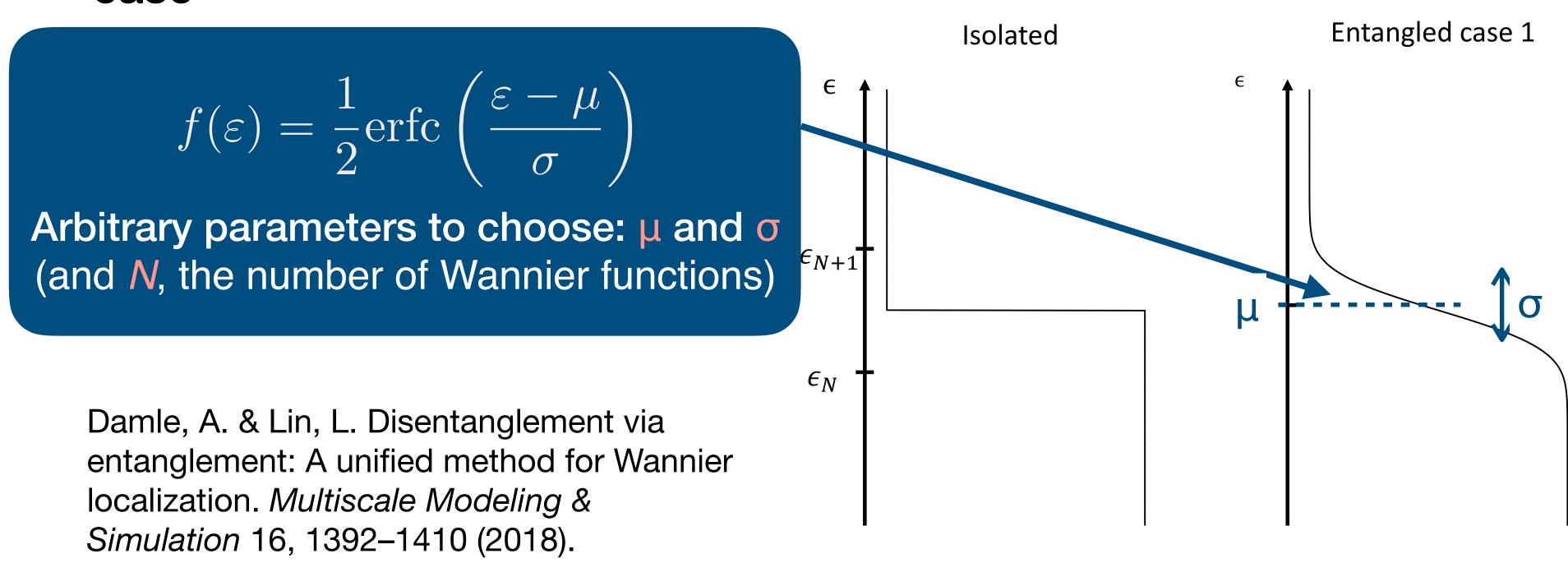
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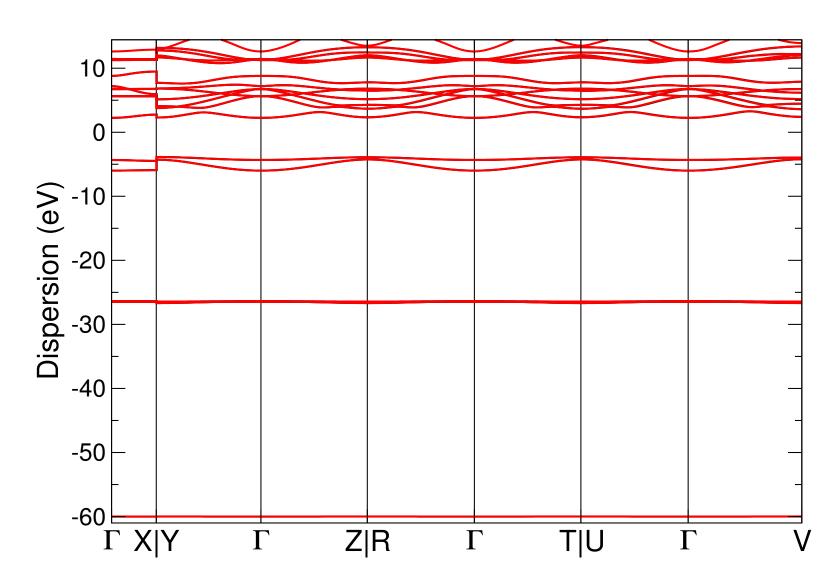
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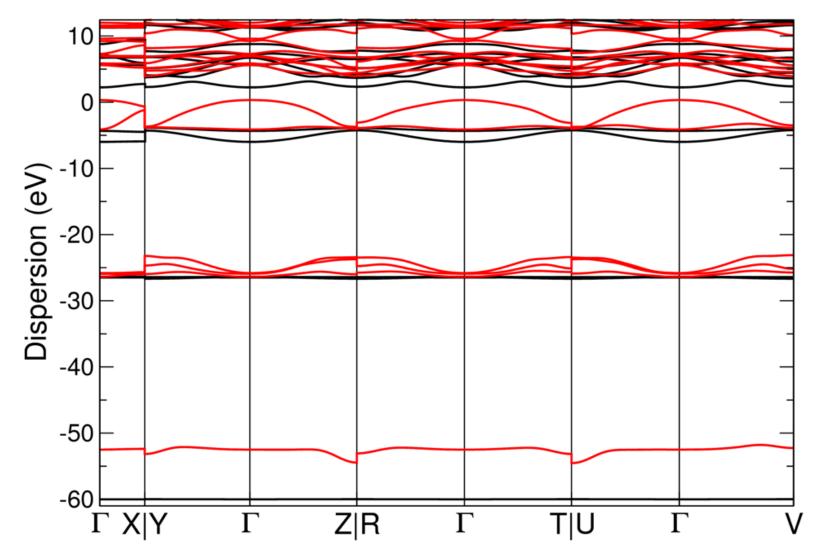


- The SCDM method does not suggest how to choose the μ and σ parameters (and neither the number N of Wannier functions)
- The choice cannot be arbitrary: "bad" values generate bad interpolations
- μ too small: not enough information on highenergy bands: QRCP will pick top states randomly
- μ too large: high-energy states (that we are not interested into) might have a large weight and QRCP might prefer to select them: interpolation tends to have higher energy than the actual bands

How to choose these parameters (automatically)?



"Correct" µ: excellent interpolation



Too large μ : bad interpolation

Important ingredient: Projectability

Copper

Orbitals: s,p,d (no nodes) + s,p (1 node)

 $p(|\psi_{n,\mathbf{k}}\rangle) = \sum_{i} |\langle o_i | \psi_{n,\mathbf{k}} \rangle|^2$

atomic orbitals o_i described in the pseudopotential

- For each band (n,k), it is the
 projection of that state on all
 the pseudo-atomic orbitals
 described in the pseudopotential
 file
- Easy to obtain from Quantum ESPRESSO's *projwfc.x*

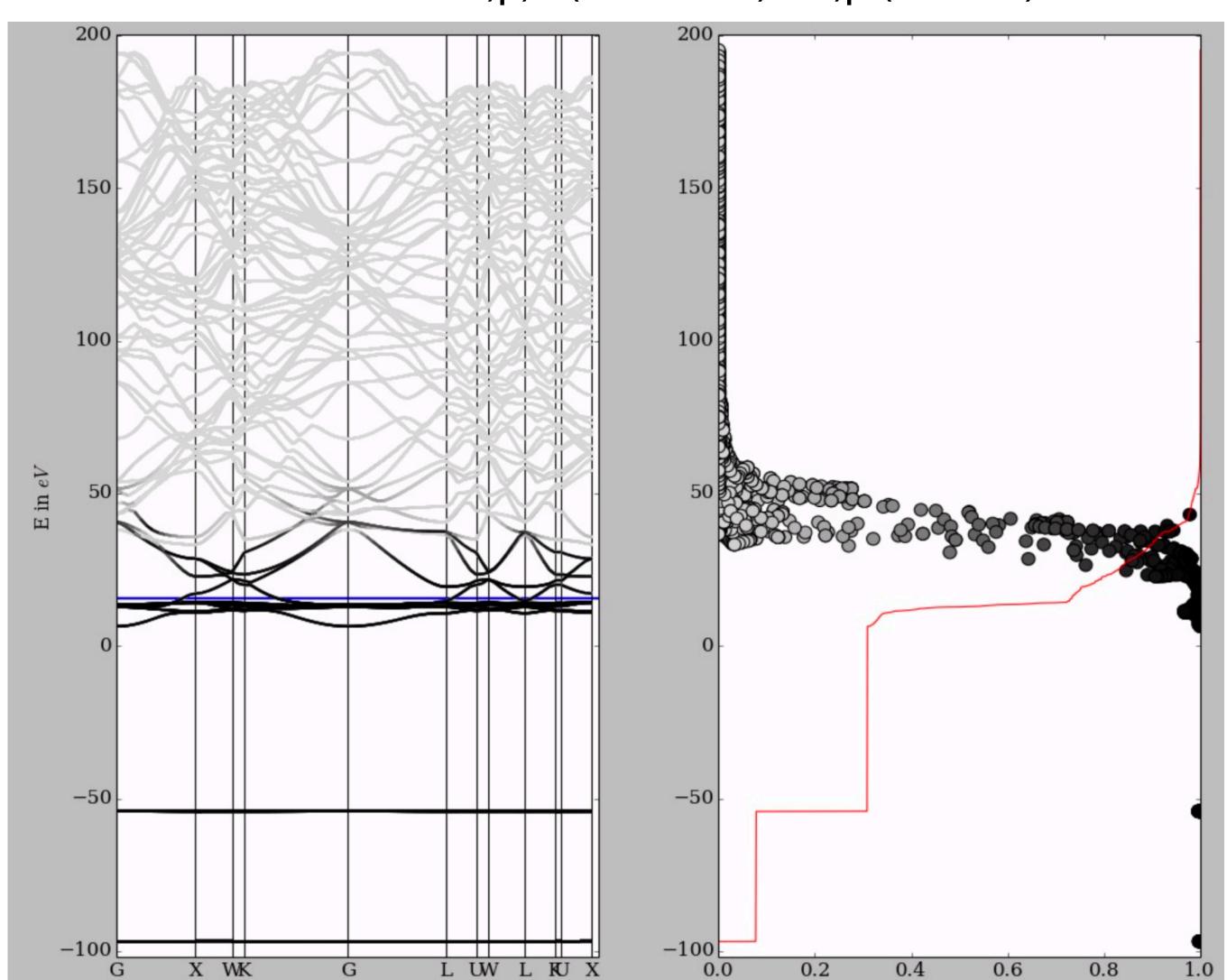


Image: courtesy of Daniel Marchand, EPFL

$$p(|\psi_{n,\mathbf{k}}\rangle) = \sum_{i} |\langle o_i | \psi_{n,\mathbf{k}} \rangle|^2$$

How to choose the parameters: Our recipe

- We aim at getting a good band interpolation for the low-lying bands
- 1: choose N as the number of atomic orbitals for which we have information in the pseudopotential file (see also Agapito et al., PRB 88, 165127 (2013))

• 2: compute the "projectability" of each state as the projection of each state on the subspace of the atomic orbitals o_i described in the pseudopotential:

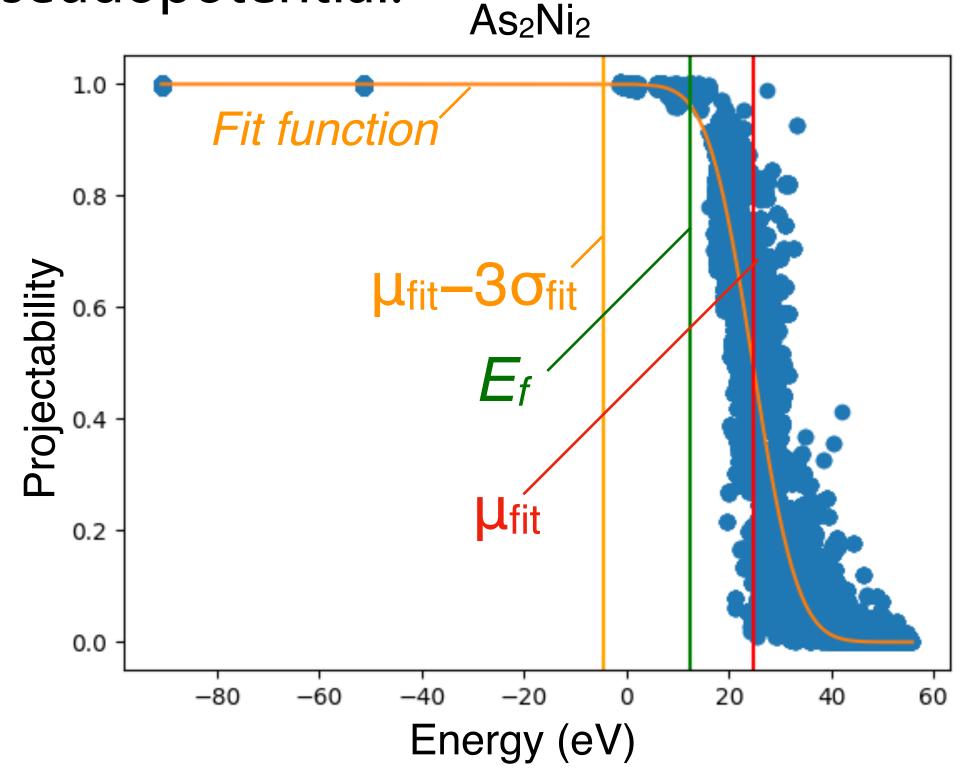
$$p(|\psi_{n,\mathbf{k}}\rangle) = \sum_{i} |\langle o_i | \psi_{n,\mathbf{k}} \rangle|^2$$

• 3: Fit the plot of the projectability vs. energy with

$$f(\varepsilon) = \frac{1}{2} \operatorname{erfc} \left(\frac{\varepsilon - \mu_{\operatorname{fit}}}{\sigma_{\operatorname{fit}}} \right)$$

• 4: choose the parameters μ and σ as follows

$$\mu=\mu_{\rm fit}-3\sigma_{\rm fit}; \qquad \sigma=\sigma_{\rm fit}$$
 V. Vitale, GP et al., arXiv:1909.00433, accepted in npj Comp. Mat.

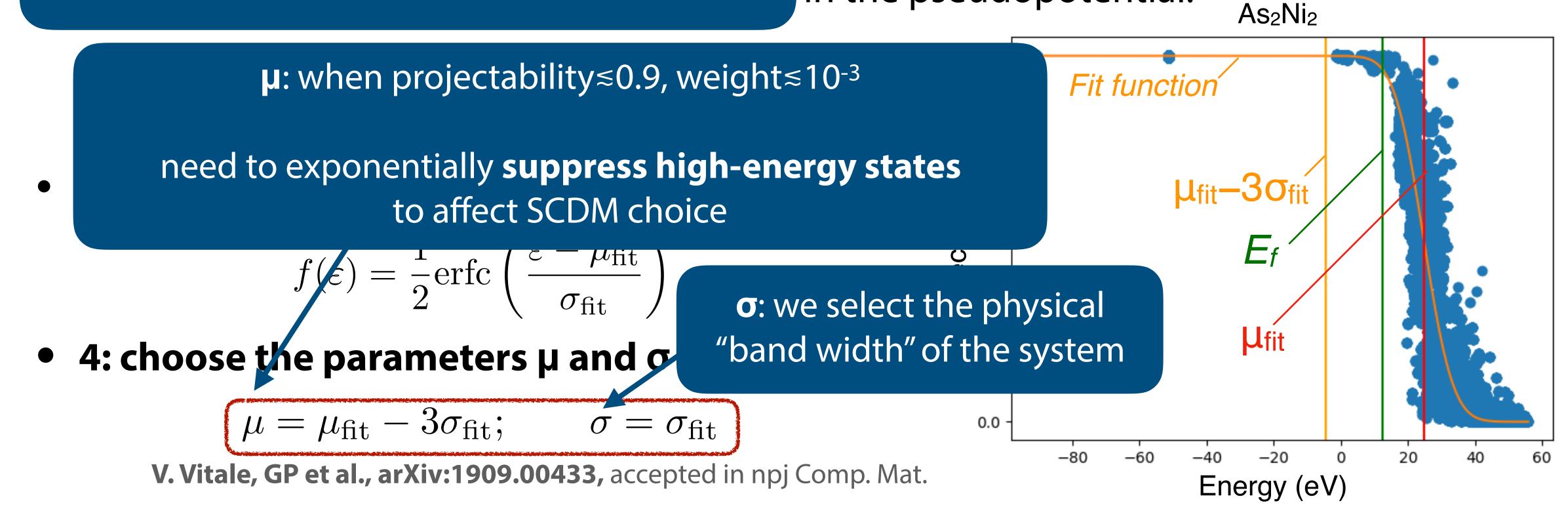


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N: we select the most relevant, low-lying bands, and we can compute the projectability

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µ: when projectability≤0.9, weight≤10-3

need to exponentially **suppress high-energy states** to affect SCDM choice

$$f(\varepsilon) = \frac{1}{2} \operatorname{erfc} \left(\frac{\varepsilon - \mu_{\text{fit}}}{\sigma_{\text{fit}}} \right)$$

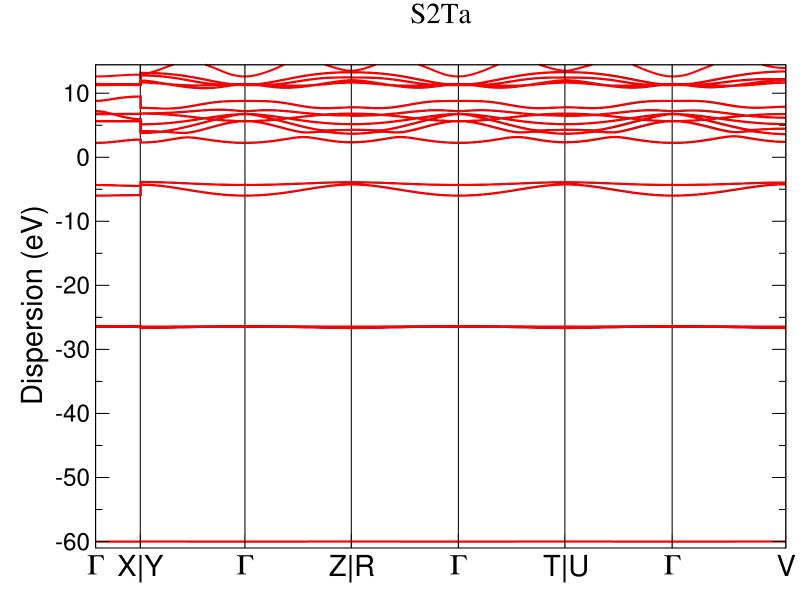
4: choose the parameters μ and σ

σ: we select the physi "band width" of the sys

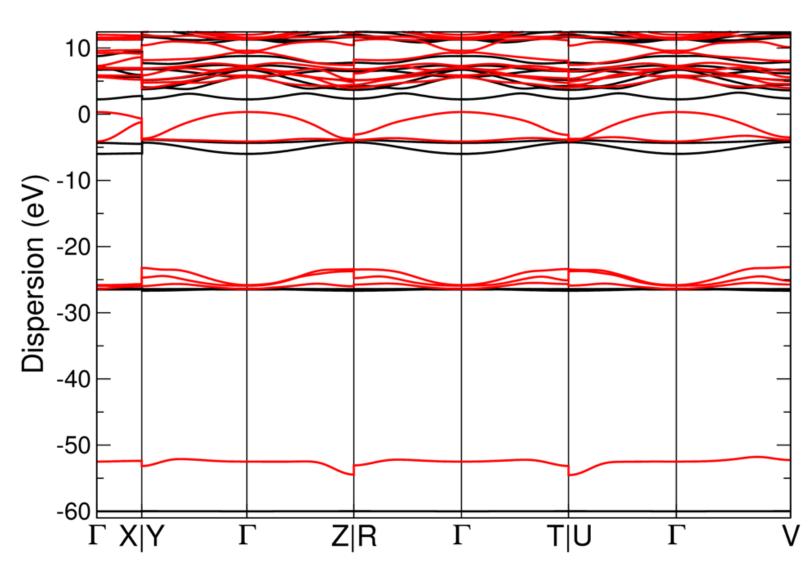
0.0 -

$$\mu = \mu_{\rm fit} - 3\sigma_{\rm fit}; \qquad \sigma = \sigma_{\rm fit}$$

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"Correct" µ: excellent interpolation



Too large, µ: bad interpolation

Bands distance for metals

• For metals, we introduce a smearing function, to consider bands only up to a given energy:

$$\eta = \sqrt{\frac{\sum_{n\mathbf{k}} \left(\varepsilon_{n\mathbf{k}}^{\mathrm{DFT}} - \varepsilon_{n\mathbf{k}}^{\mathrm{Wan}}\right)^{2} \tilde{f}_{n\mathbf{k}}}{\sum_{n\mathbf{k}} \tilde{f}_{n\mathbf{k}}}},$$

Average bands distance

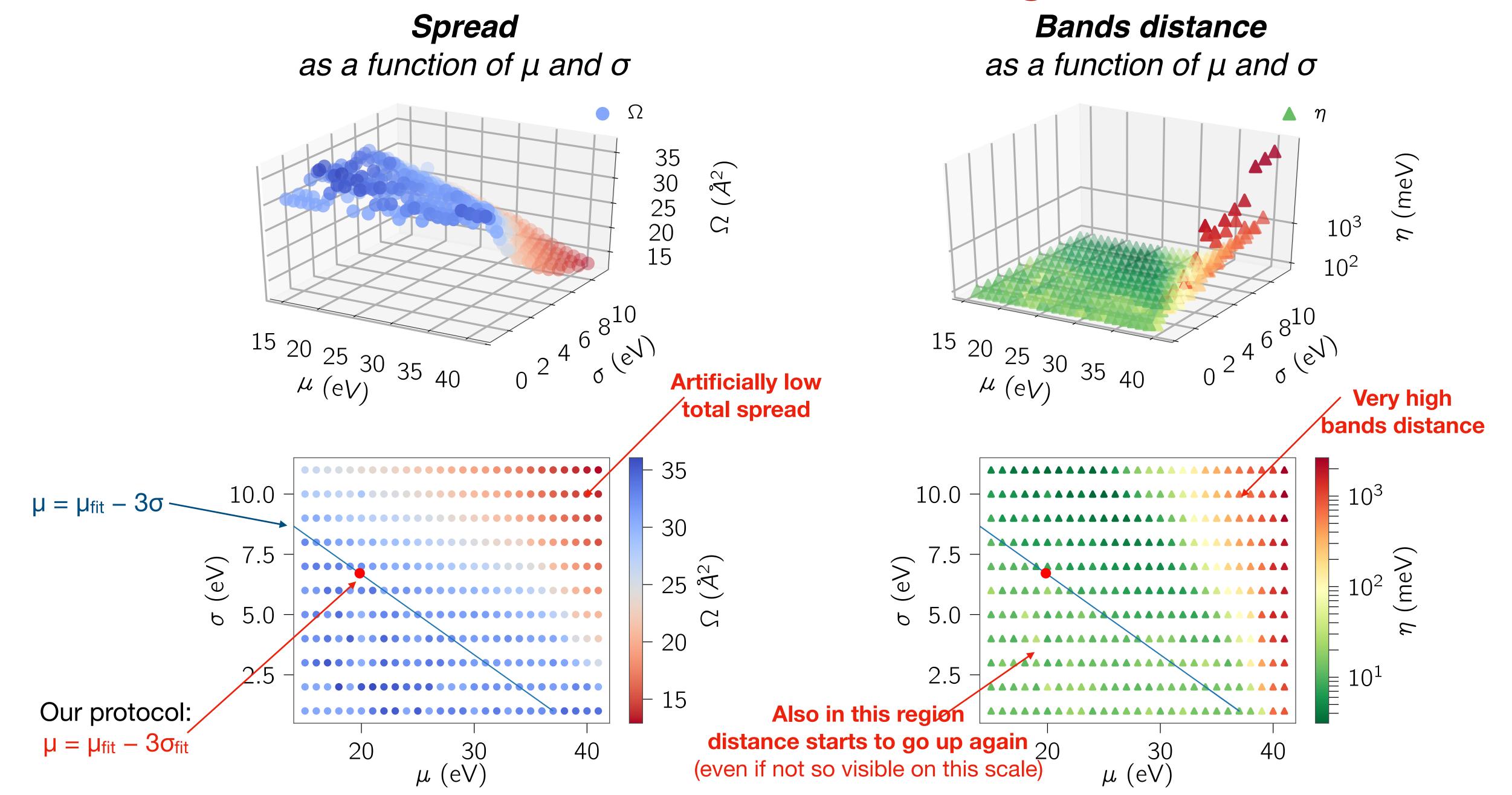
$$\eta^{\max} = \max_{n\mathbf{k}} \left(\tilde{f}_{n\mathbf{k}} | \varepsilon_{n\mathbf{k}}^{\text{DFT}} - \varepsilon_{n\mathbf{k}}^{\text{Wan}} | \right)$$

Max bands distance

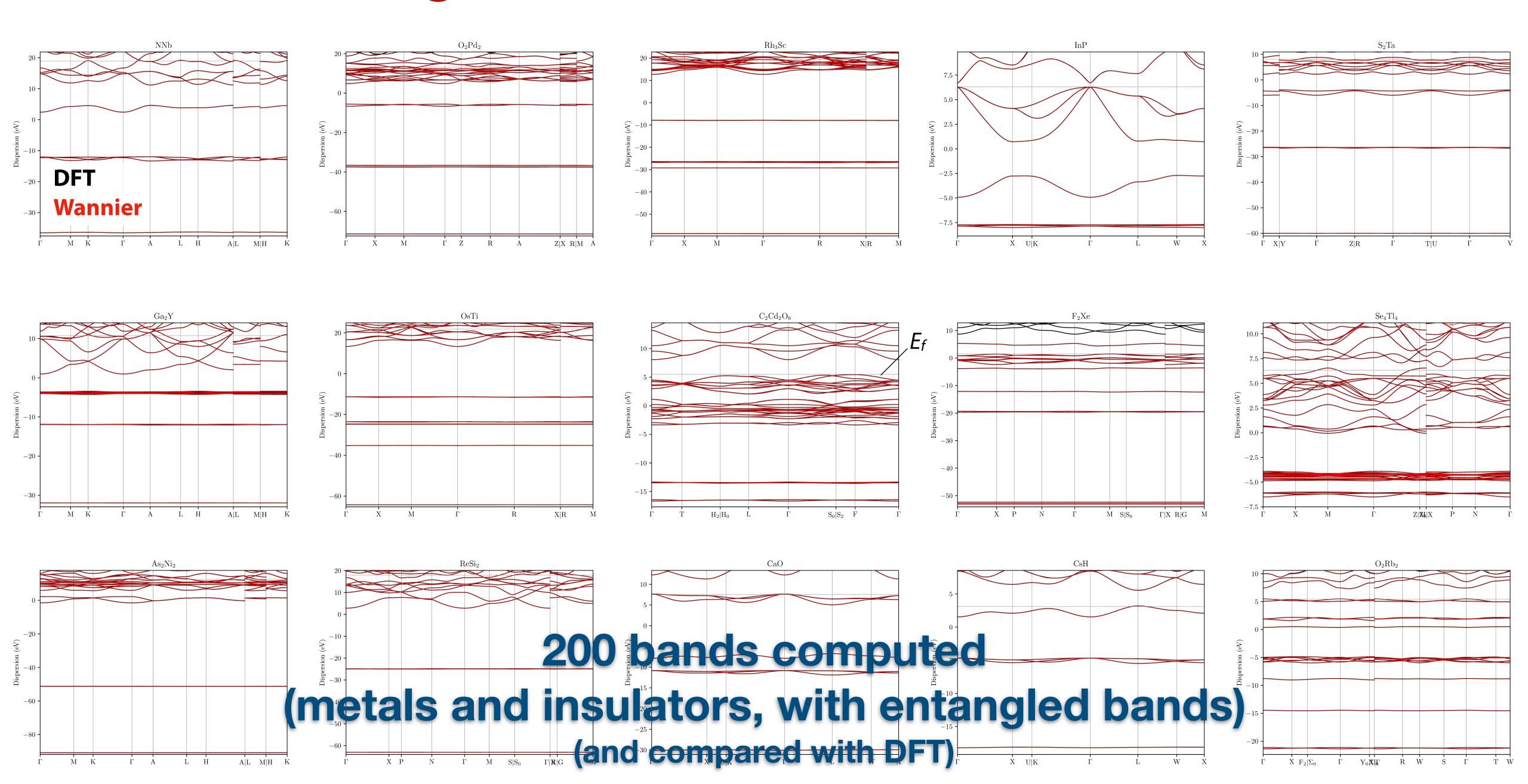
• Smearing function:
$$\tilde{f}_{n\mathbf{k}} = \sqrt{f_{n\mathbf{k}}^{\mathrm{DFT}}(\nu,\tau)f_{n\mathbf{k}}^{\mathrm{Wan}}(\nu,\tau)}$$

• Typically, the "fake" chemical potential \mathbf{v} is set e.g. 1eV above the Fermi level (and we choose $\mathbf{\tau}=0.1\text{eV}$)

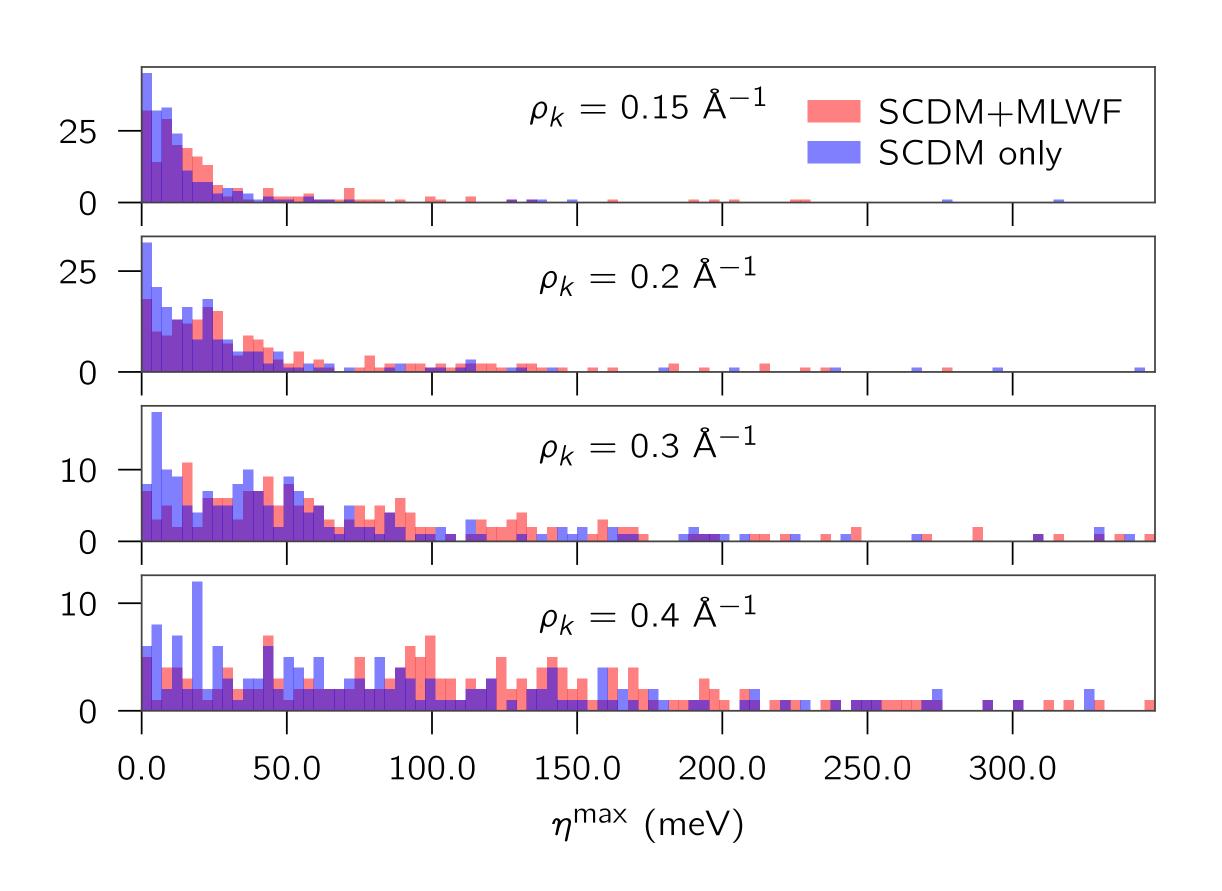
Parameter choice validation: tungsten (W)

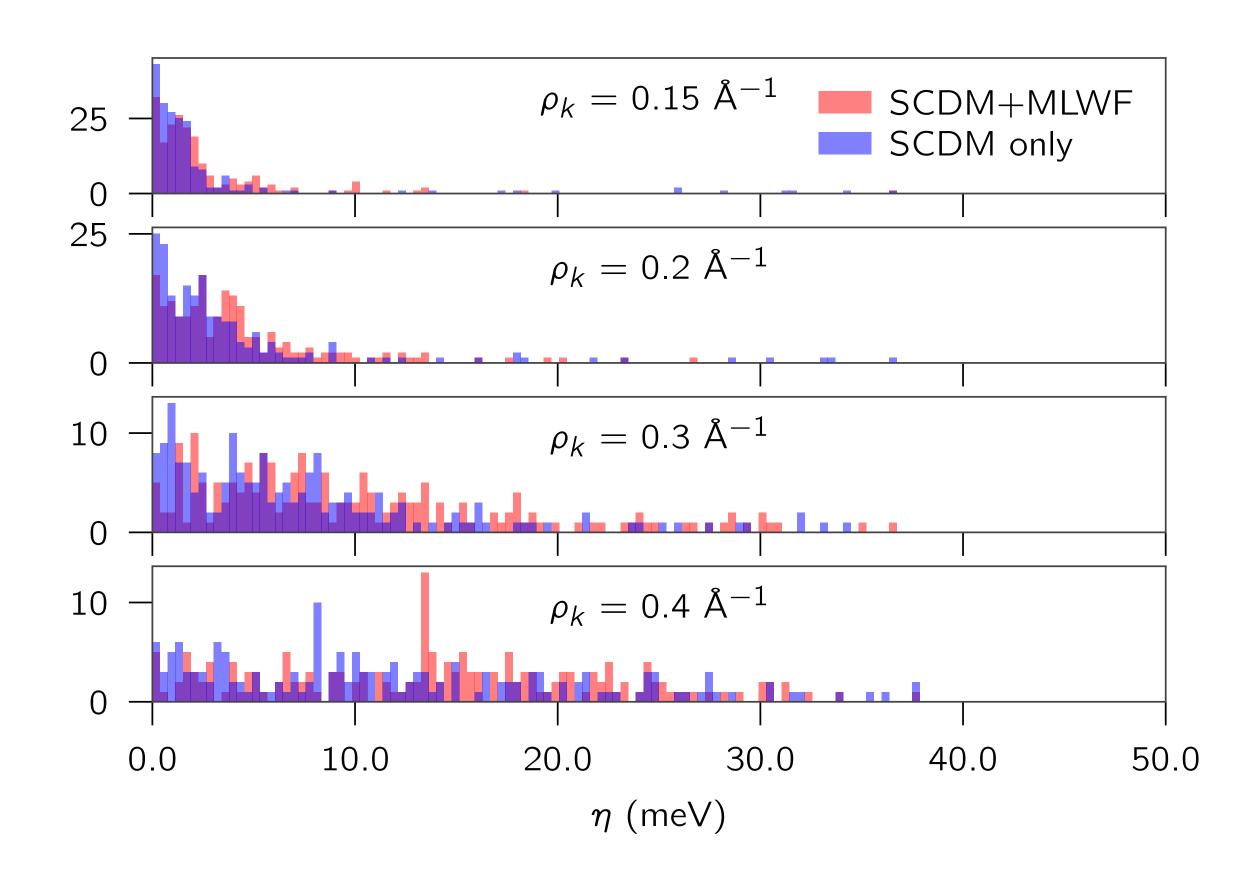


Entangled bands: (subset of the) results



Comparison of SCDM + MLWF with SCDM only (200 materials with entangled bands)

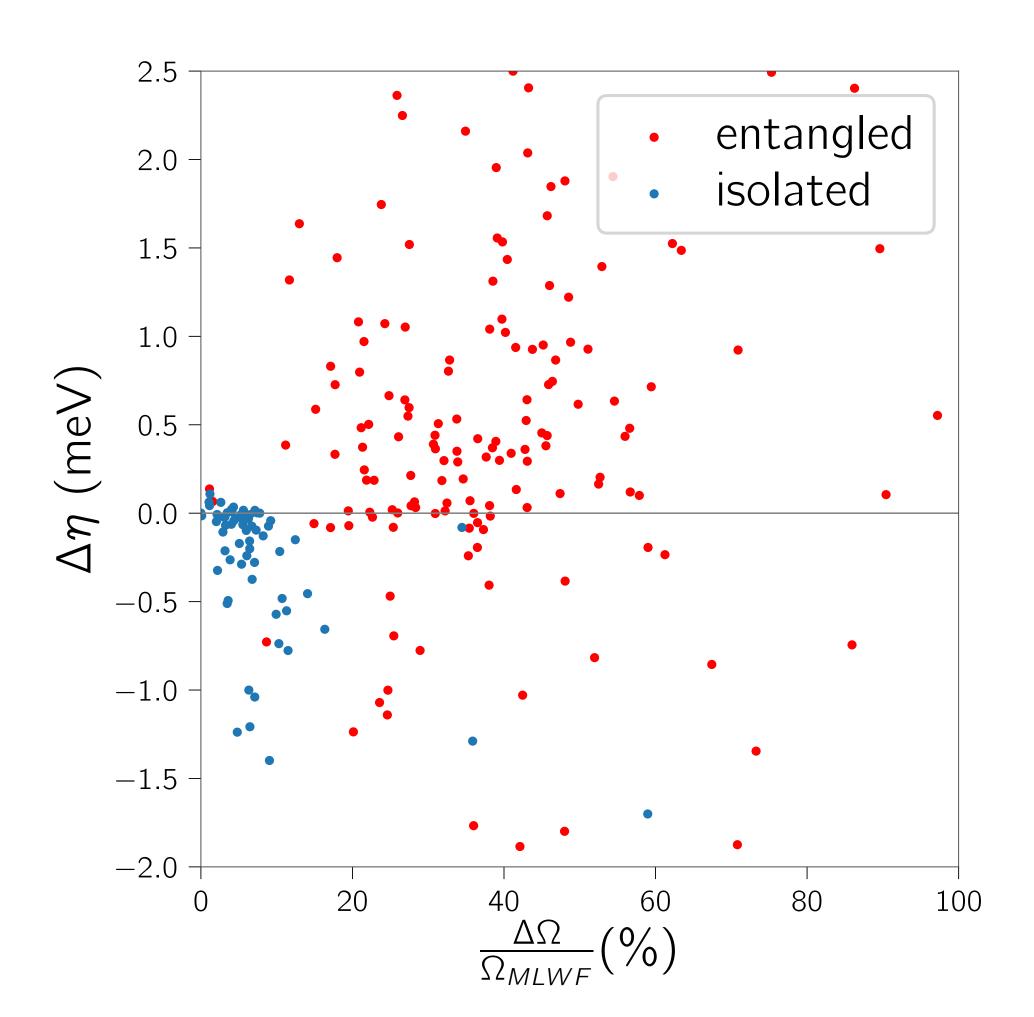




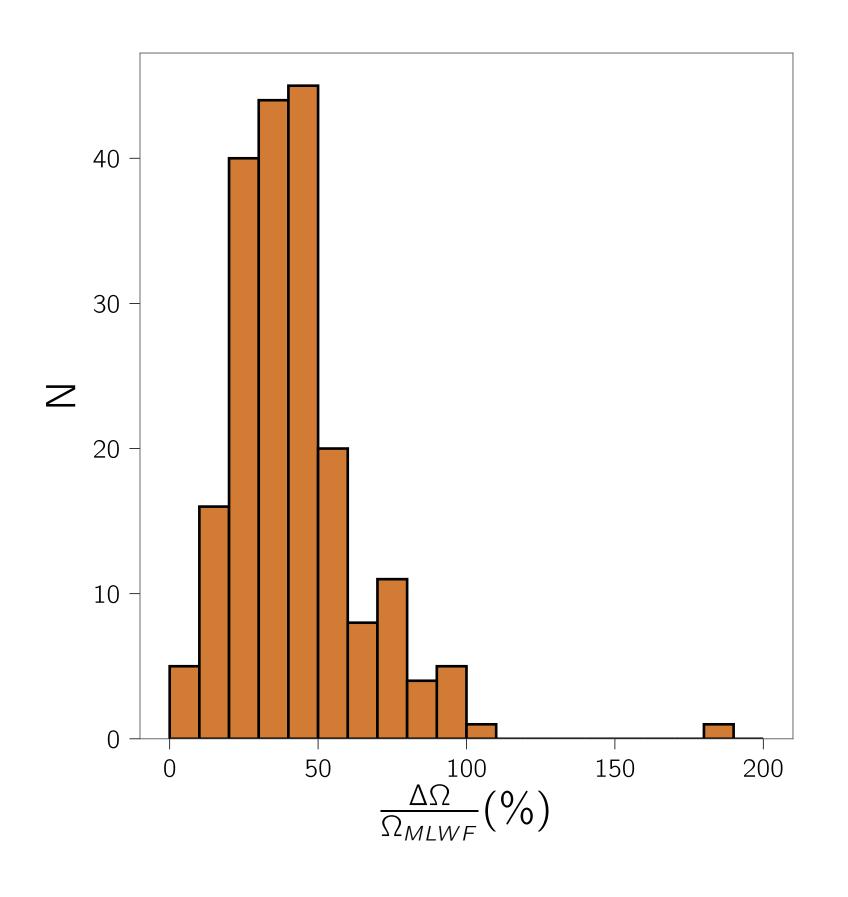
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Comparison of SCDM + MLWF with SCDM only (200 materials with entangled bands)

 In many cases, running a MLWF in the entangled case worsens the bands distance (even if only by a little)



- Spread reduction is relatively large, often
 50% or more
- SCDM with entangled bands is not maximally localised



Conclusions (part 1)

- SCDM method now implemented in Quantum ESPRESSO (since QE 6.3)
- We have **defined an algorithm** for the choice of "free" parameters in SCDM (N, μ , σ)
- We have implemented automatic workflows within AiiDA for fully automatic
 Wannierisation
- We have validated the SCDM method and our algorithm on a set of 200 crystalline materials (~80 insulators) covering the chemical space

Acknowledgements



THEORY AND SIMULATION

OF MATERIALS







Valerio Vitale



Antimo Marrazzo



Nicola Marzari



Jonathan R. Yates



Arash A. Mostofi

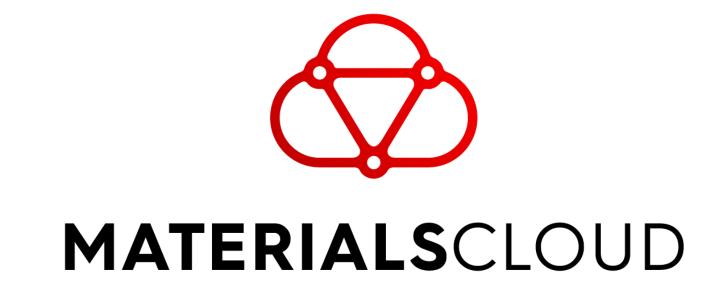
Open Science Platform for Materials Science: AiiDA and the Materials Cloud

Giovanni Pizzi (EPFL)



http://www.aiida.net

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



http://www.materialscloud.org









Our two core infrastructures



AiiDA as the "operating system" to manage, automate and store simulations and their results

and



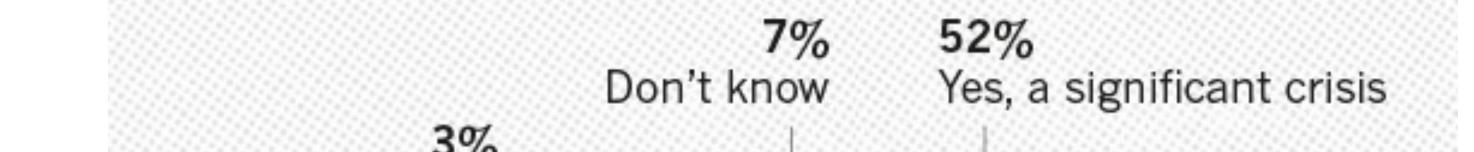
Materials Cloud as the open-science dissemination portal and cloud simulation platform

Reproducibility: a cornerstone of the scientific method

IS THERE A REPRODUCIBILITY CRISIS?



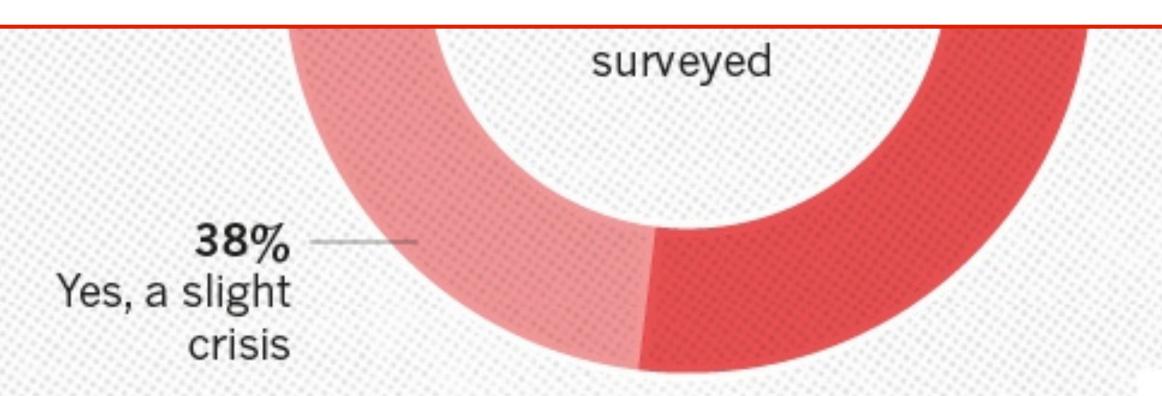
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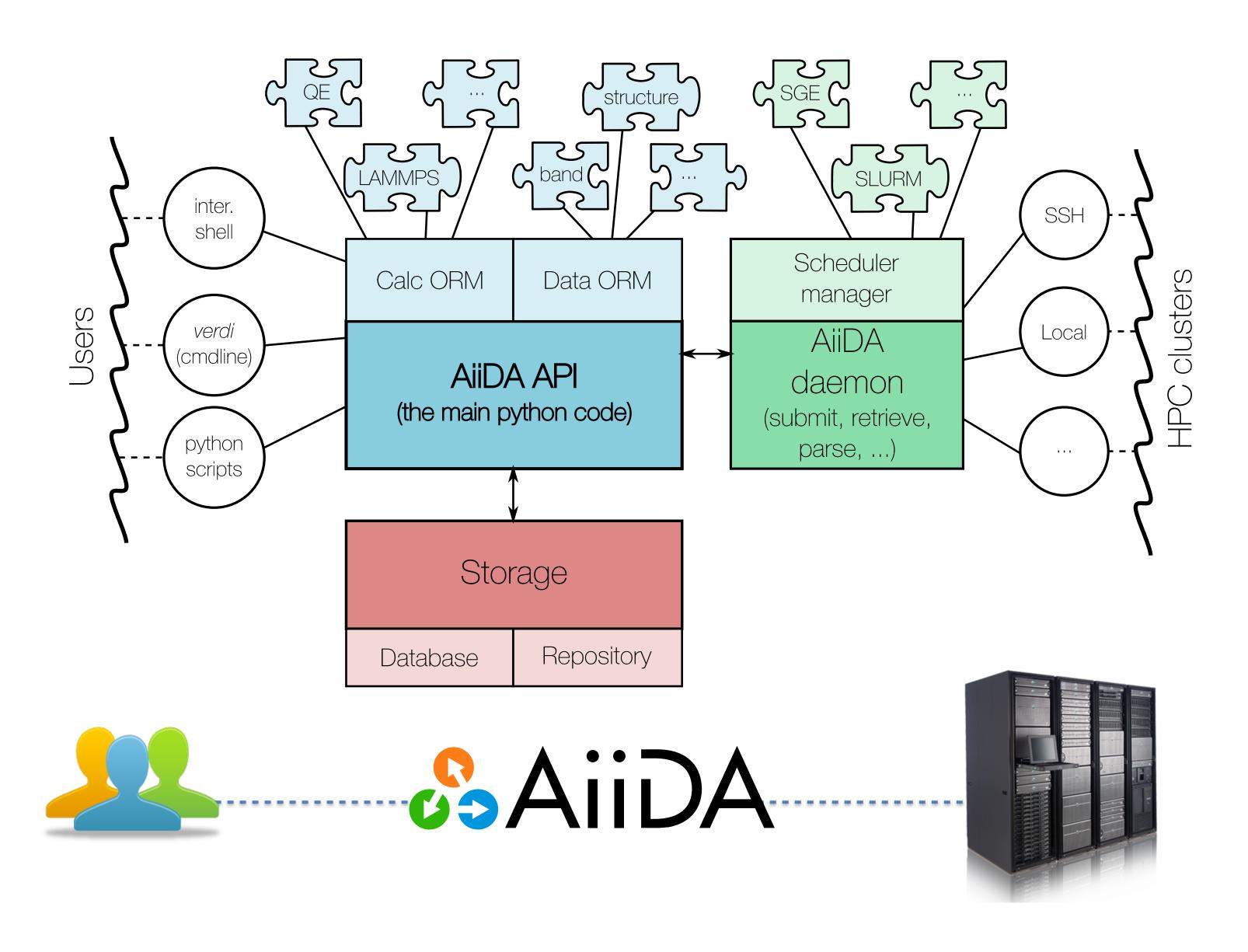
No excuses in computational science

We can and must be fully reproducible

CHALLENGE #2: make open-science easier



AiiDA



Main features

- Python (2.7 & 3.6) infrastructure
- SQL database backend, access via a Python ORM
- Local connection to clusters, or via ssh using a python API
- Interface to various job schedulers (SGE, Torque, LSF, PBS Pro, SLURM, ...)
- Event-based daemon with remote management and workflow execution manager
- REST APIs using Flask to expose one own's data
- Plugin management system and extended code support
- Easy sharing of the results with other users in the community

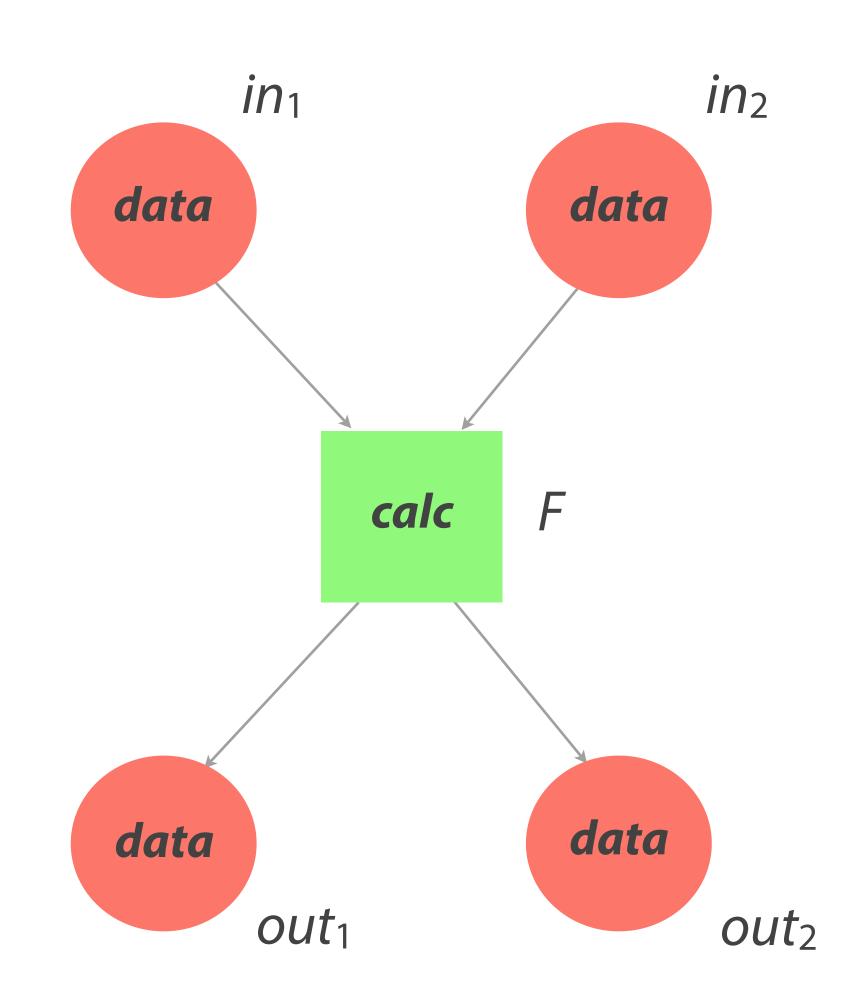
Storage and provenance



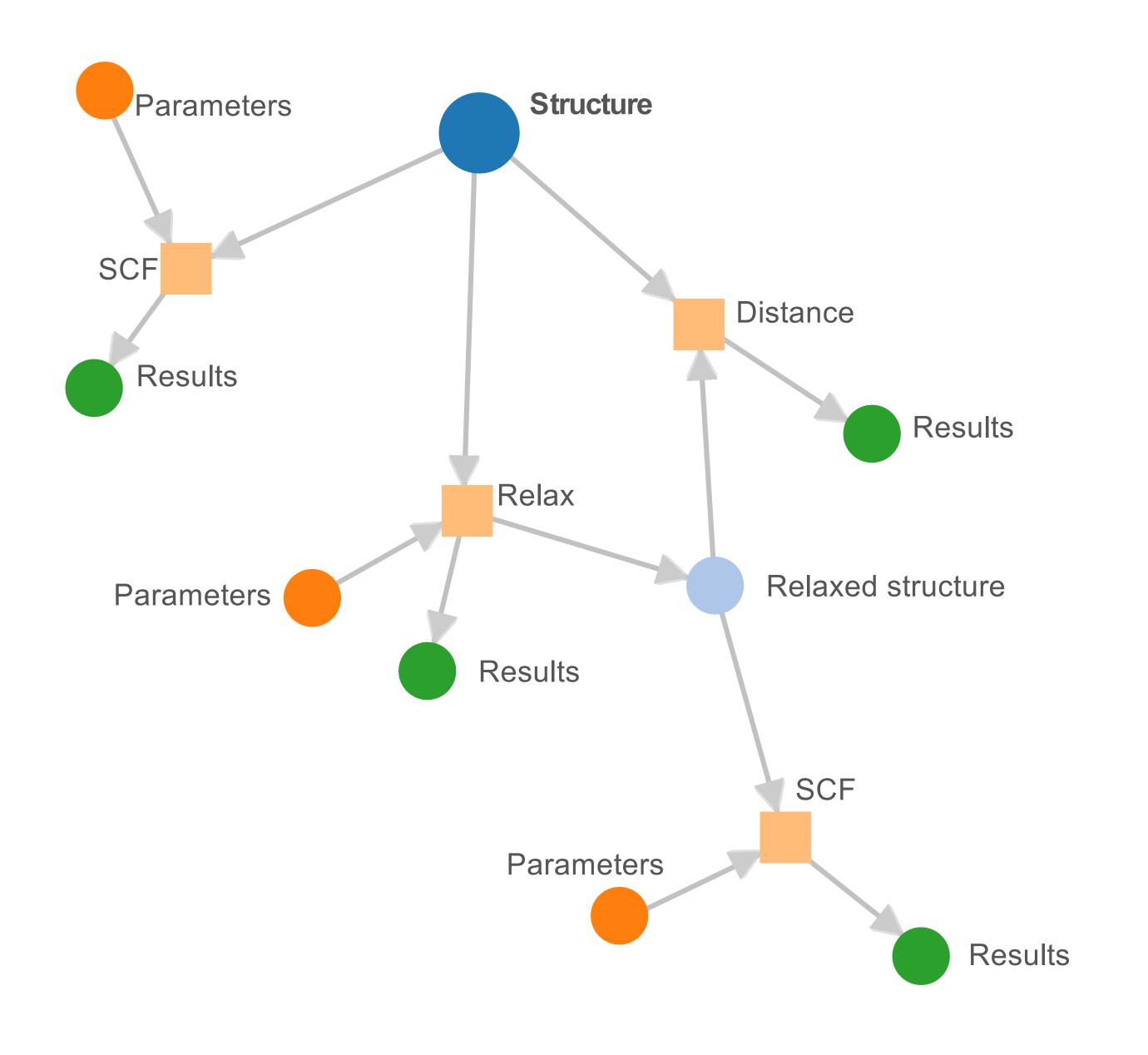
- Calculated properties: result of complex, connected calculations
- How do we store simulations preserving the connected structure?
- Inspiration from the open provenance model
- Any calculation: a function, converting inputs to outputs:

$$out_1$$
, $out_2 = F(in_1, in_2)$

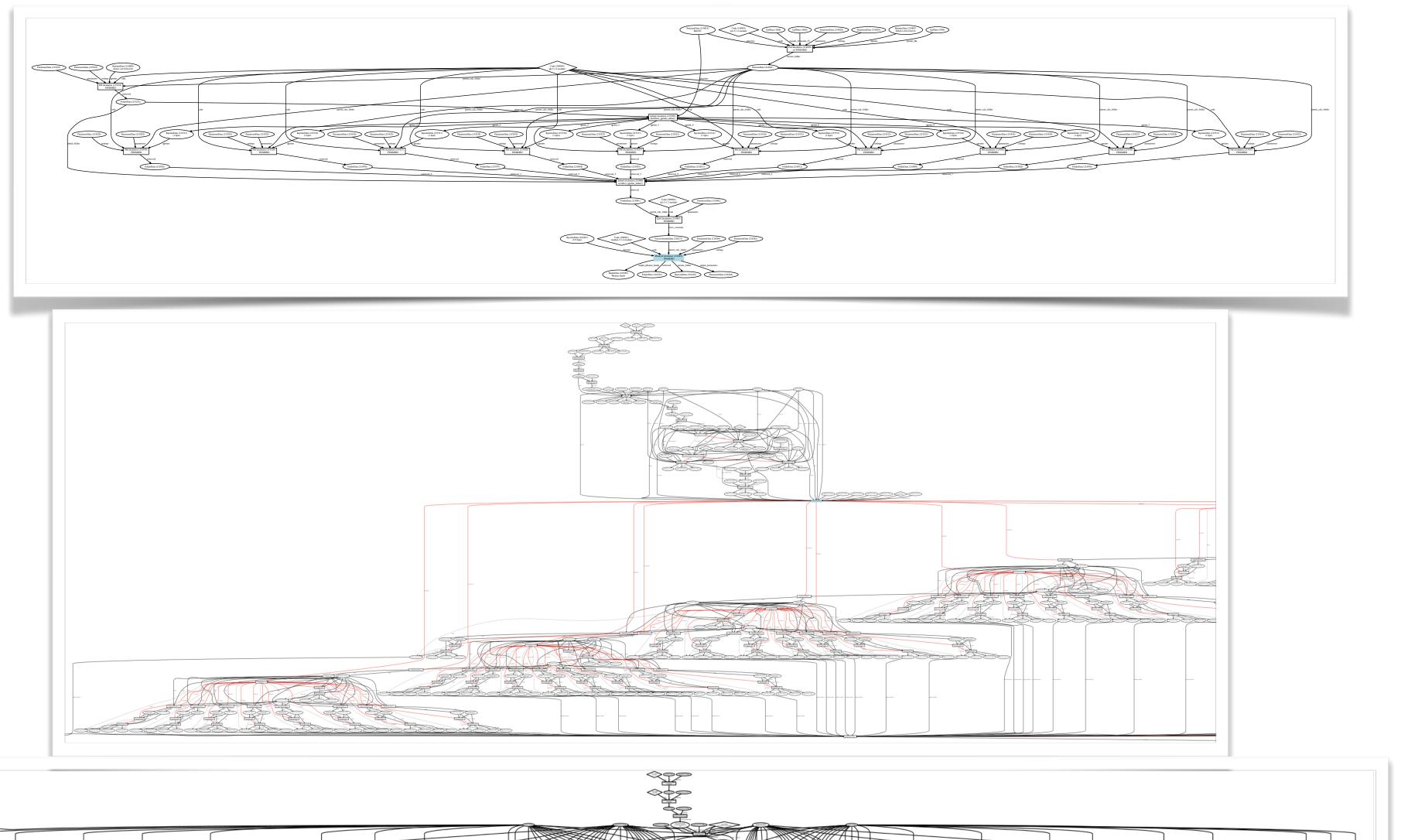
- Each object is a node in a graph, connected by directional labeled links
- Output nodes can be used as inputs



Data provenance: directed acyclic graphs



"Simple" graphs of workflows for a single material



Phonon dispersion

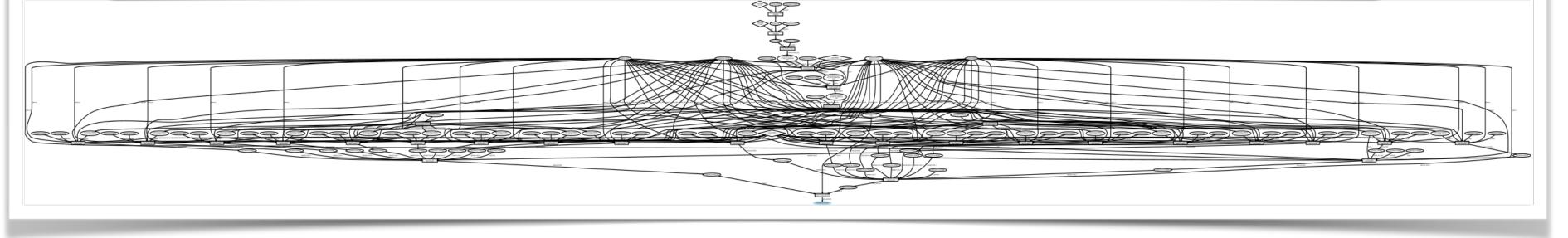
Thermal transport, electronic mobility, ...

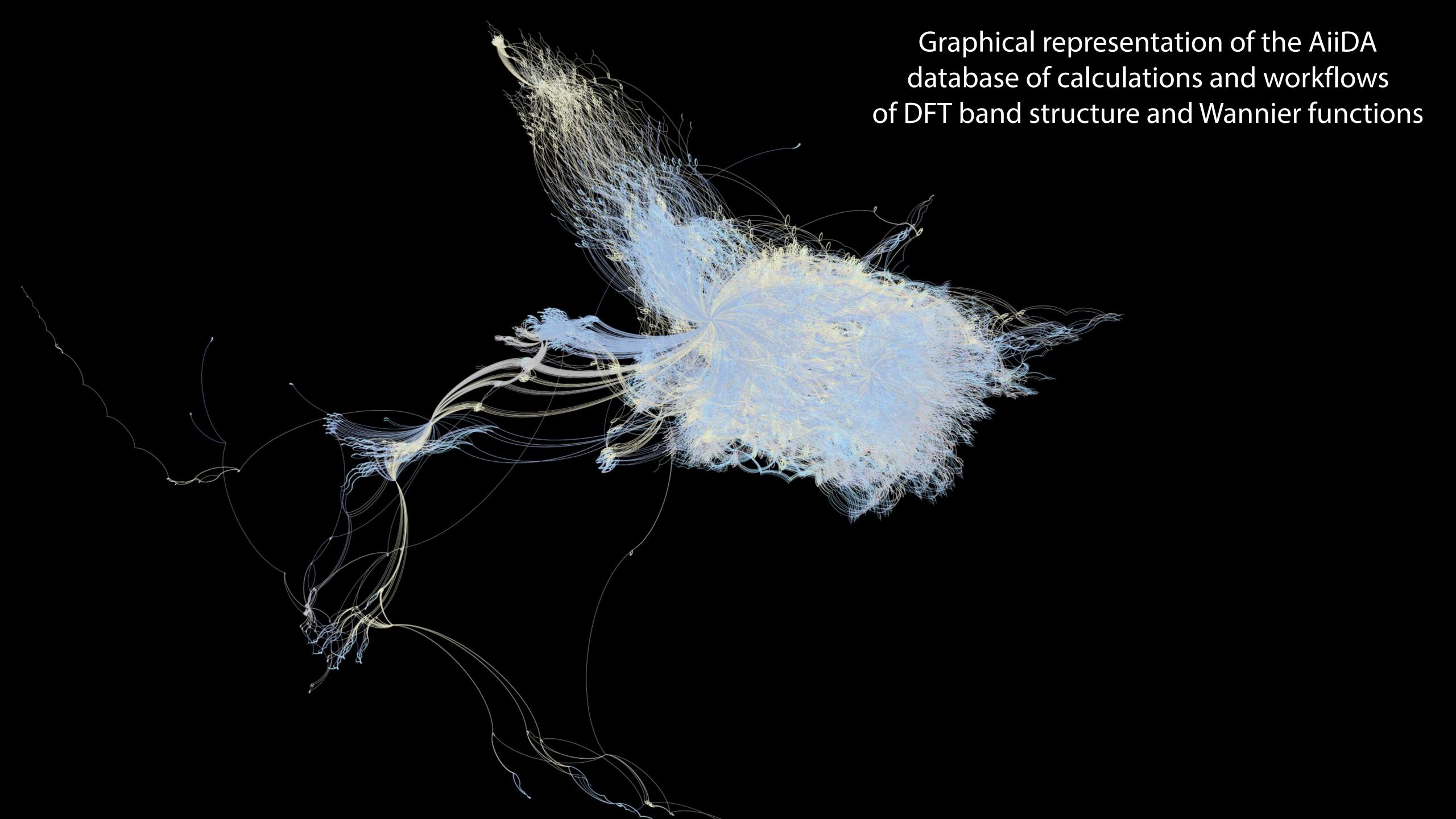
Molecular dynamics of Li ions in a solid-state electrolyte

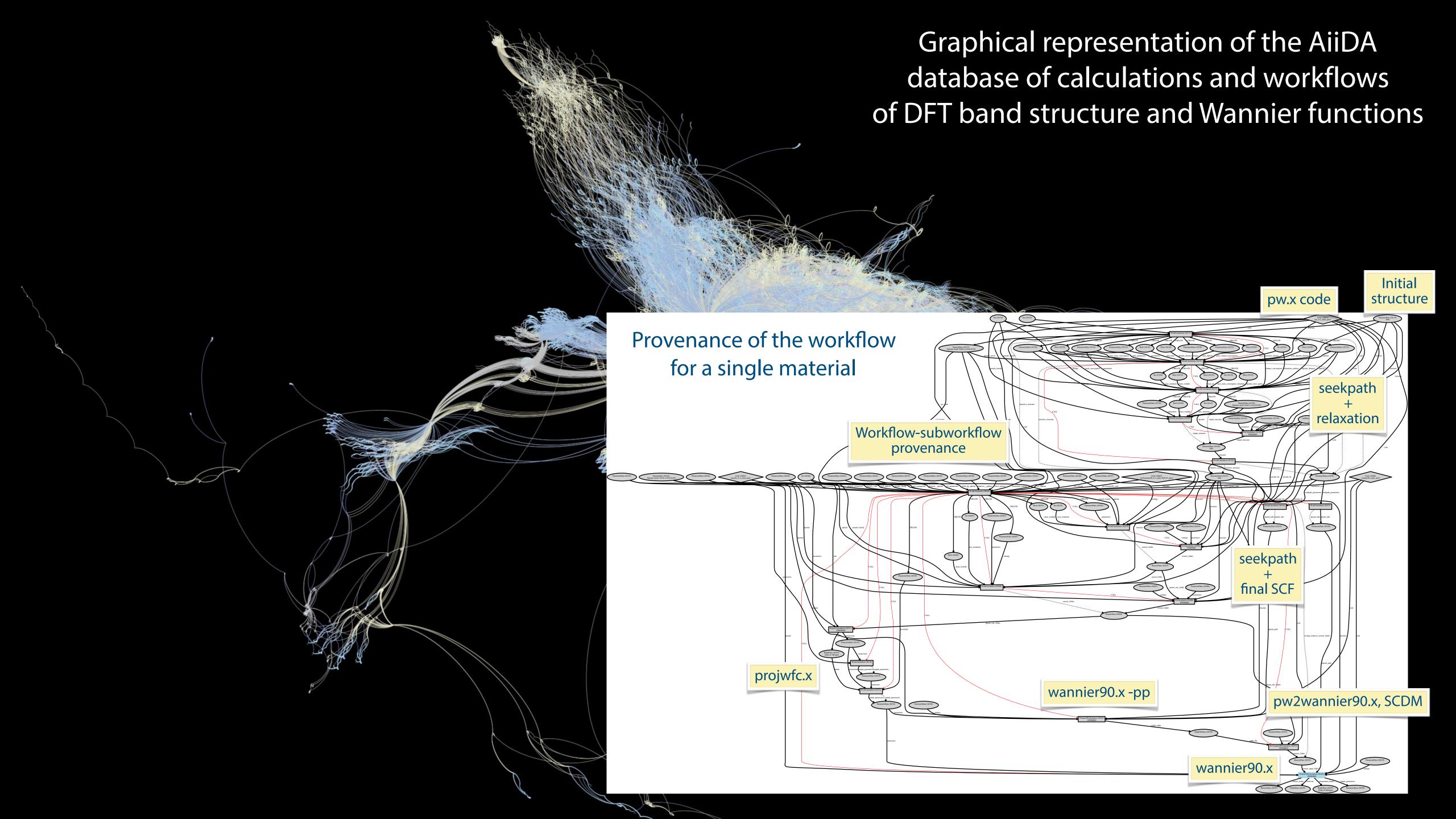
Novel, safe and efficient Li-batteries

Elastic constants

Response of materials to stresses and deformations



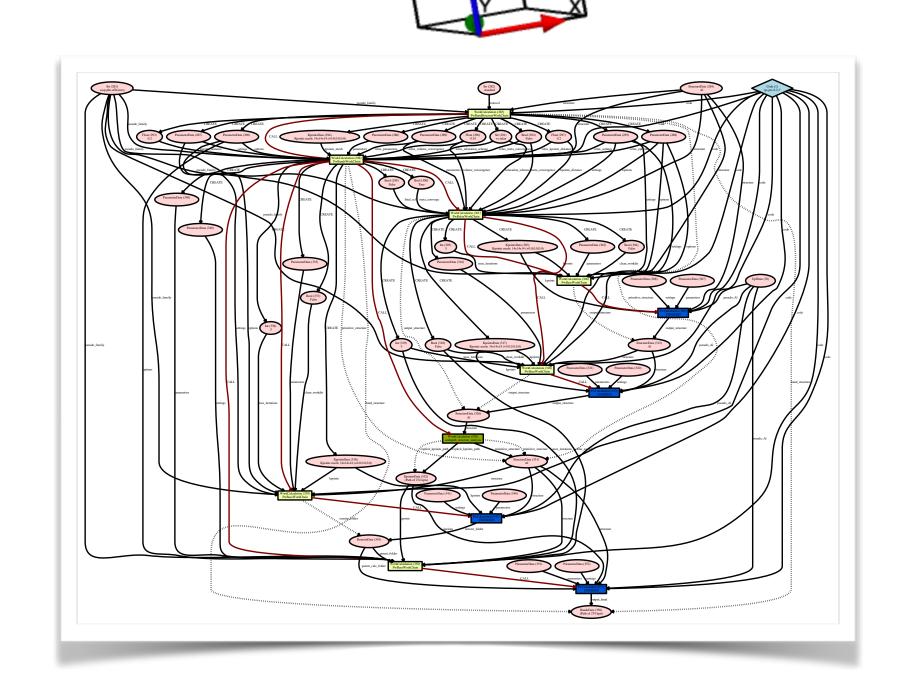


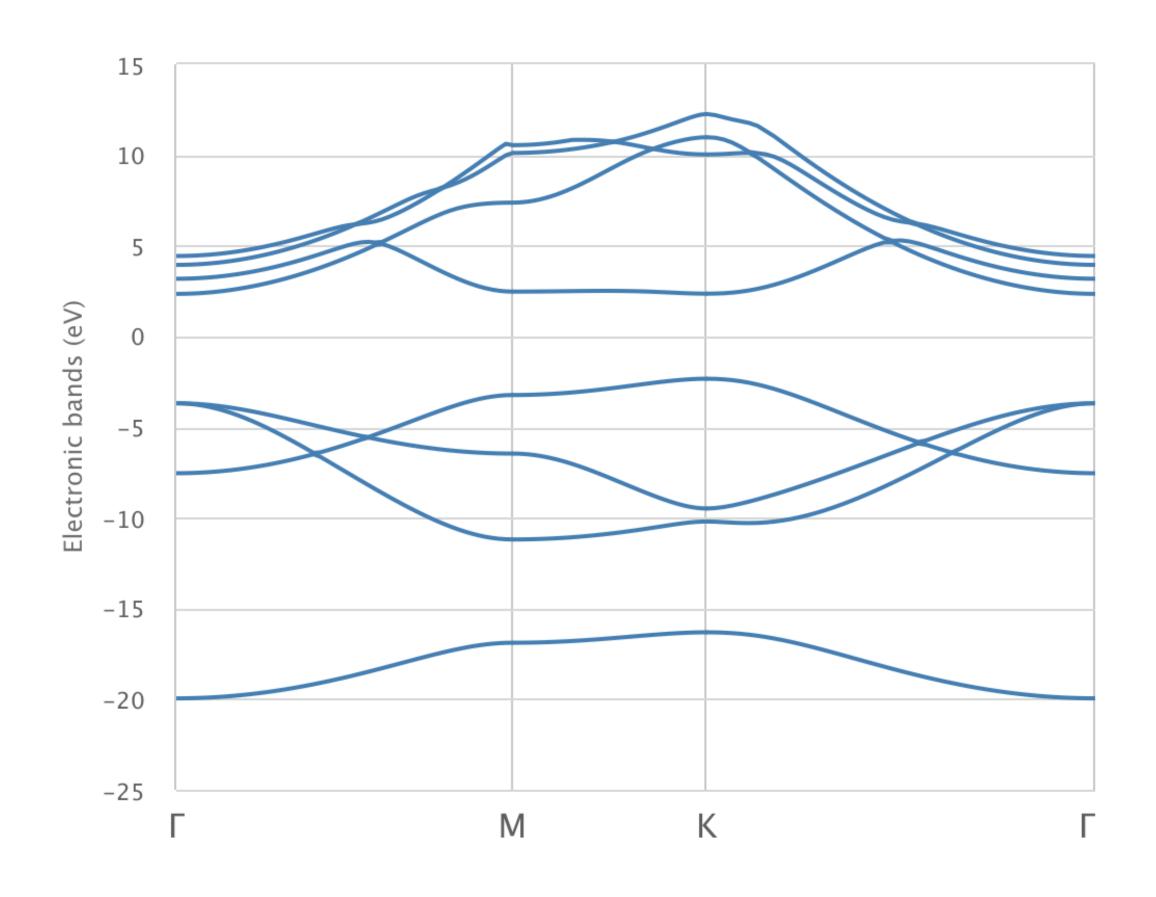


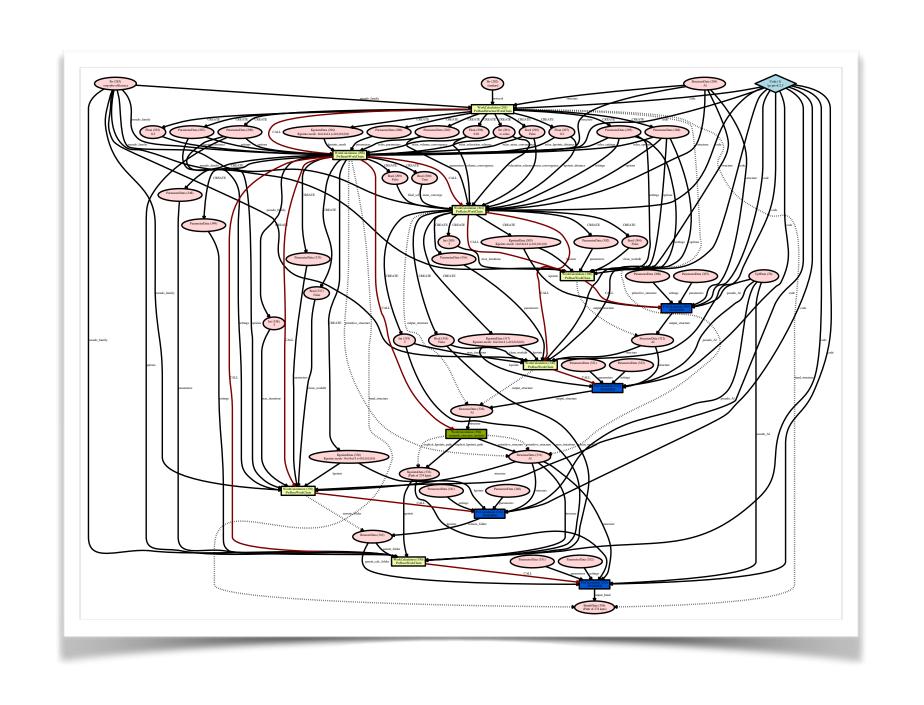
• Given a material, we often need to compute advanced quantities

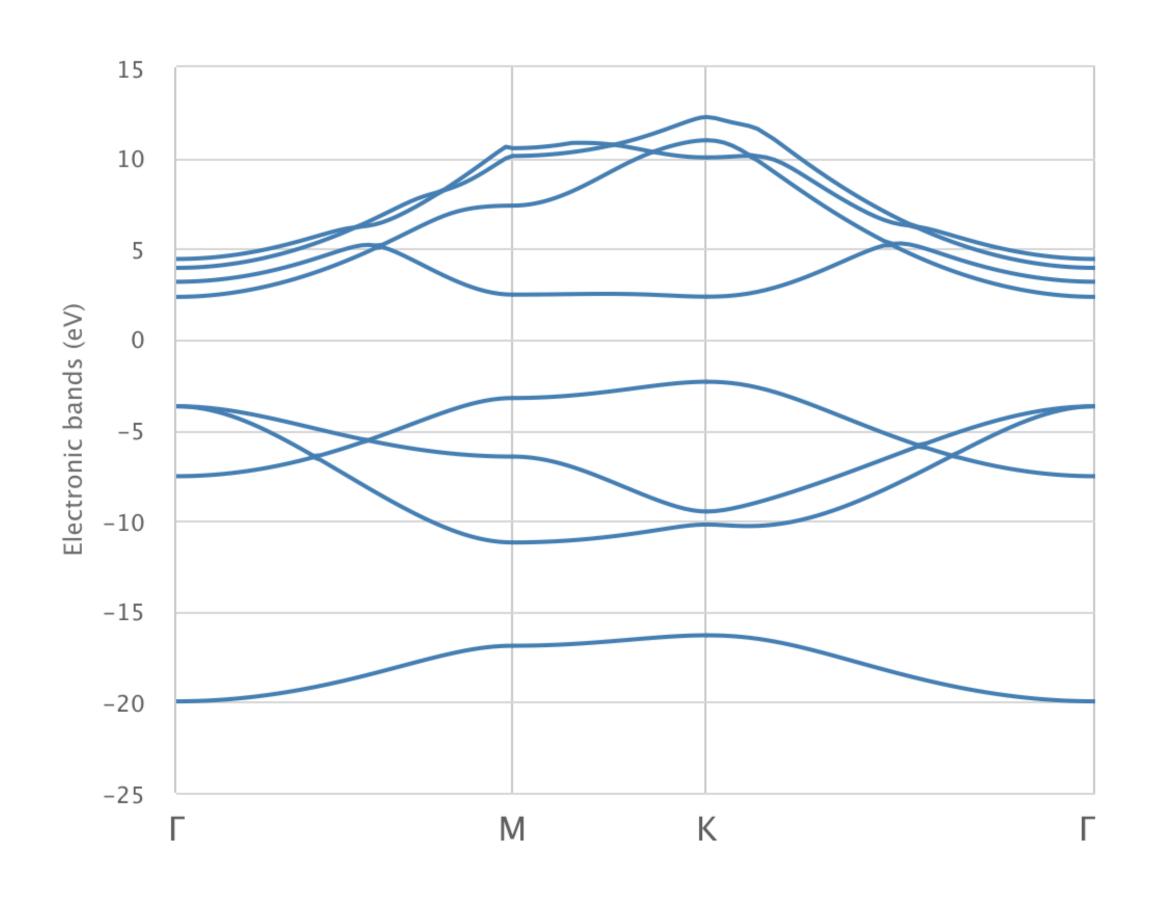
• These are often non-trivial and result from a

complex workflow









- The AiiDA provenance graph allows to know how the structure was computed and to reproduce that single specific calculation: log of "what happened in the past"
- We need also an easy way to re-run the same calculation again with different parameters or for a different material: turn-key workflows

```
class PwBandsWorkChain(WorkChain):
   @classmethod
   def define(cls, spec):
        spec.input('code',
                   valid_type=Code)
        spec.input('structure',
                   valid_type=StructureData)
        spec.input('pseudo_family',
                   valid_type=Str)
        spec.outline(
            cls.setup,
            cls.validate_inputs,
            if_(cls.should_do_relax)(
                cls.run_relax,
            cls.run_seekpath,
            cls.run_scf,
            cls.run_bands,
            cls.results,
```

- "Operating system" for all calculations
- Automatic provenance tracking in the DB
- Control provenance granularity store level of detail relevant to the workflows
- Progress checkpointing restart from arbitrary step, retry on failure, allows to shut down daemon and continue later
- Easy debugging, self-documenting

```
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       spec.outline(
           cls.setup,
           cls.validate_inputs,
           if_(cls.should_do_relax)(
               cls.ru
                     Turn-key solution:
           cls.run_s
                     PwBandStructureWorkChain run(
           cls.run_s
                          code=Code get_from_string(
           cls.run_b
                               'qe-pw-6.2.1@localhost'),
           cls.resul
                          structure=StructureData(
                              ase=ase.build.bulk('Al')),
                           pseudo_family=Str('sssp-pbe-efficiency'))
```

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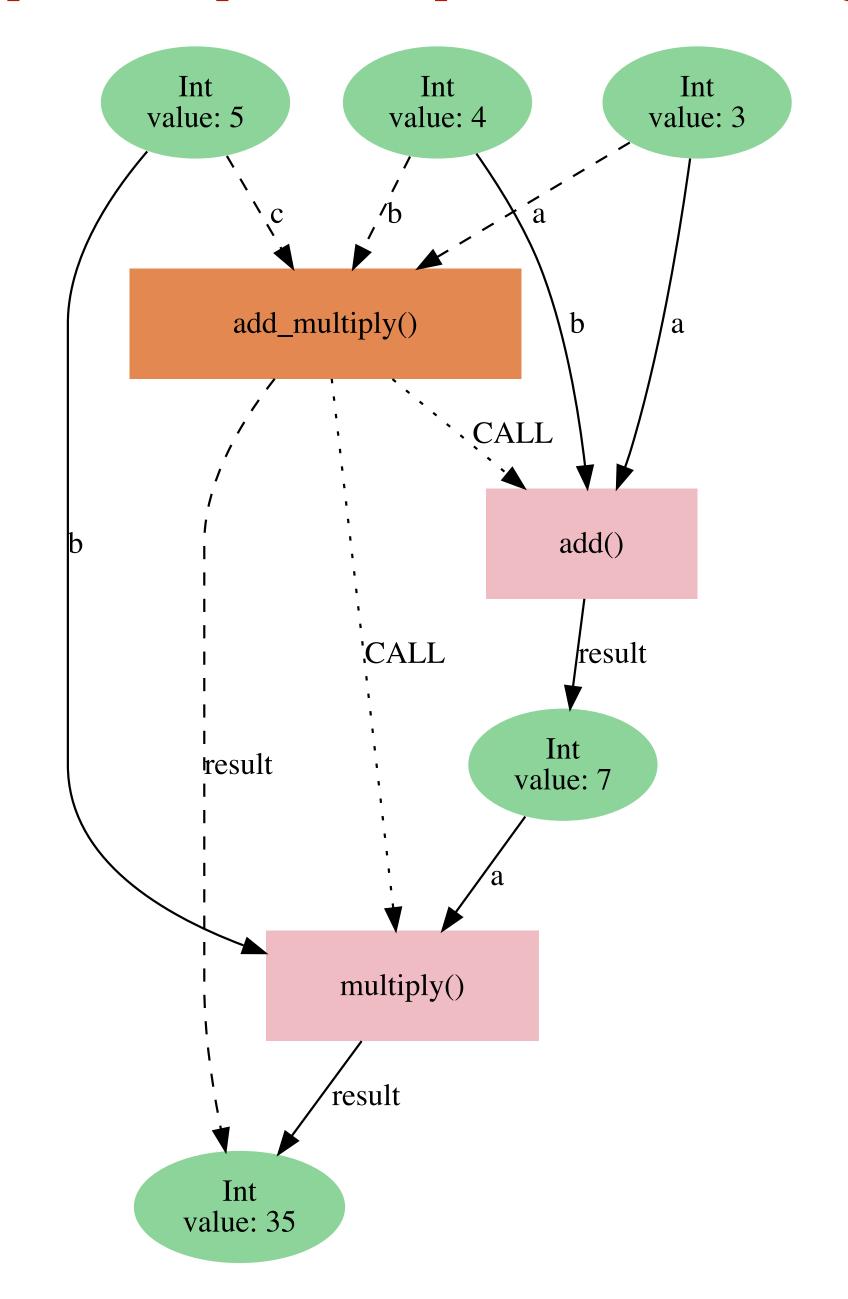
ugging, self-documenting

Simple provenance tracking also for pre-/post-processing

```
def add(a, b):
    return a+b
def multiply(a, b):
    return a*b
def add_multiply(a, b, c):
    return multiply(add(a, b), c)
final_value = add_multiply(
    a=3,
    b=4
    c=5)
final_value: (3+4)*5=35
```

Simple provenance tracking also for pre-/post-processing

```
from aiida.engine import calcfunction, workfunction
@calcfunction
def add(a, b):
    return a+b
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def multiply(a, b):
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@workfunction
def add_multiply(a, b, c):
    return multiply(add(a, b), c)
final_value = add_multiply(
    a=Int(3),
    b=Int(4)
    C=Int(5)
final value:
<Int: uuid: e25b98c2-db06-4451-86dc-4e1b179d00cc (pk: 1547) value: 35>
```

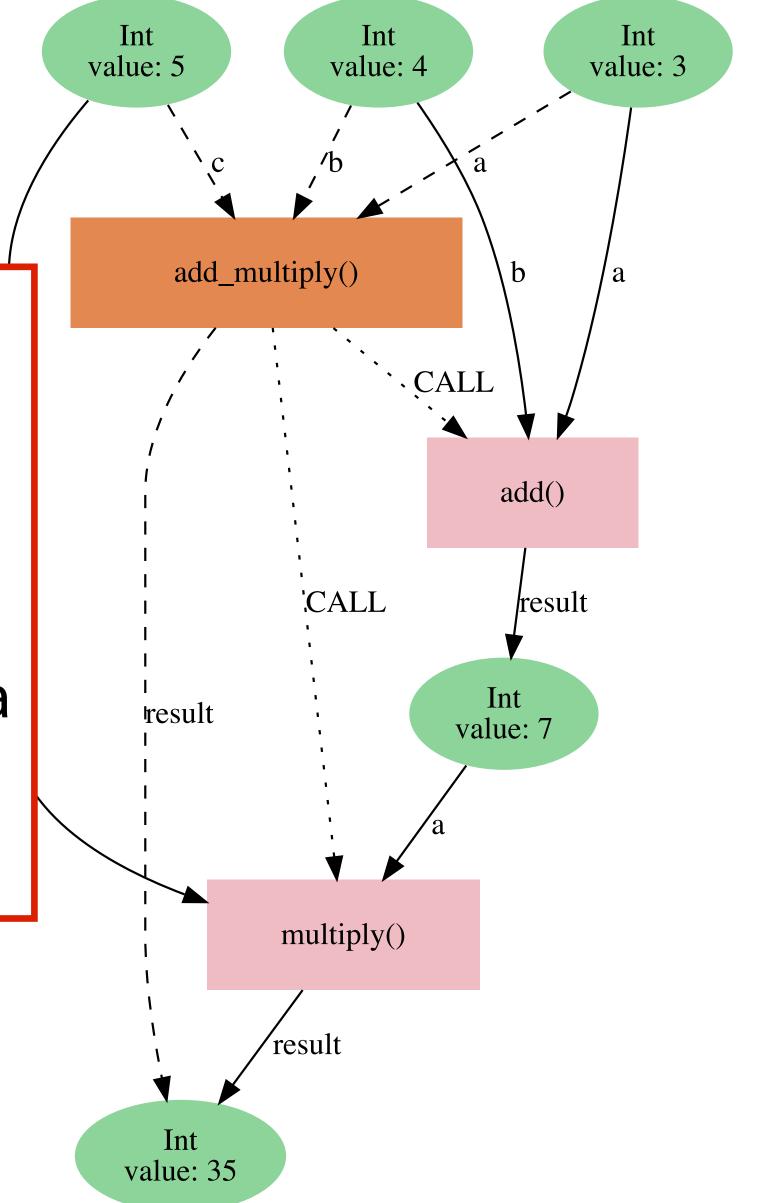


Simple provenance tracking also for pre-/post-processing

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from aiida engine import calcfunction, workfunction
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def add(a, b):
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    return multiply(add(a, b),
final_value = add_multiply(
    a=Int(3),
    b=Int(4)
```

AUTOMATIC PROVENANCE TRACKING

By very minimal modifications, provenance is automatically tracked also for simple, local data processing python functions



final_value:

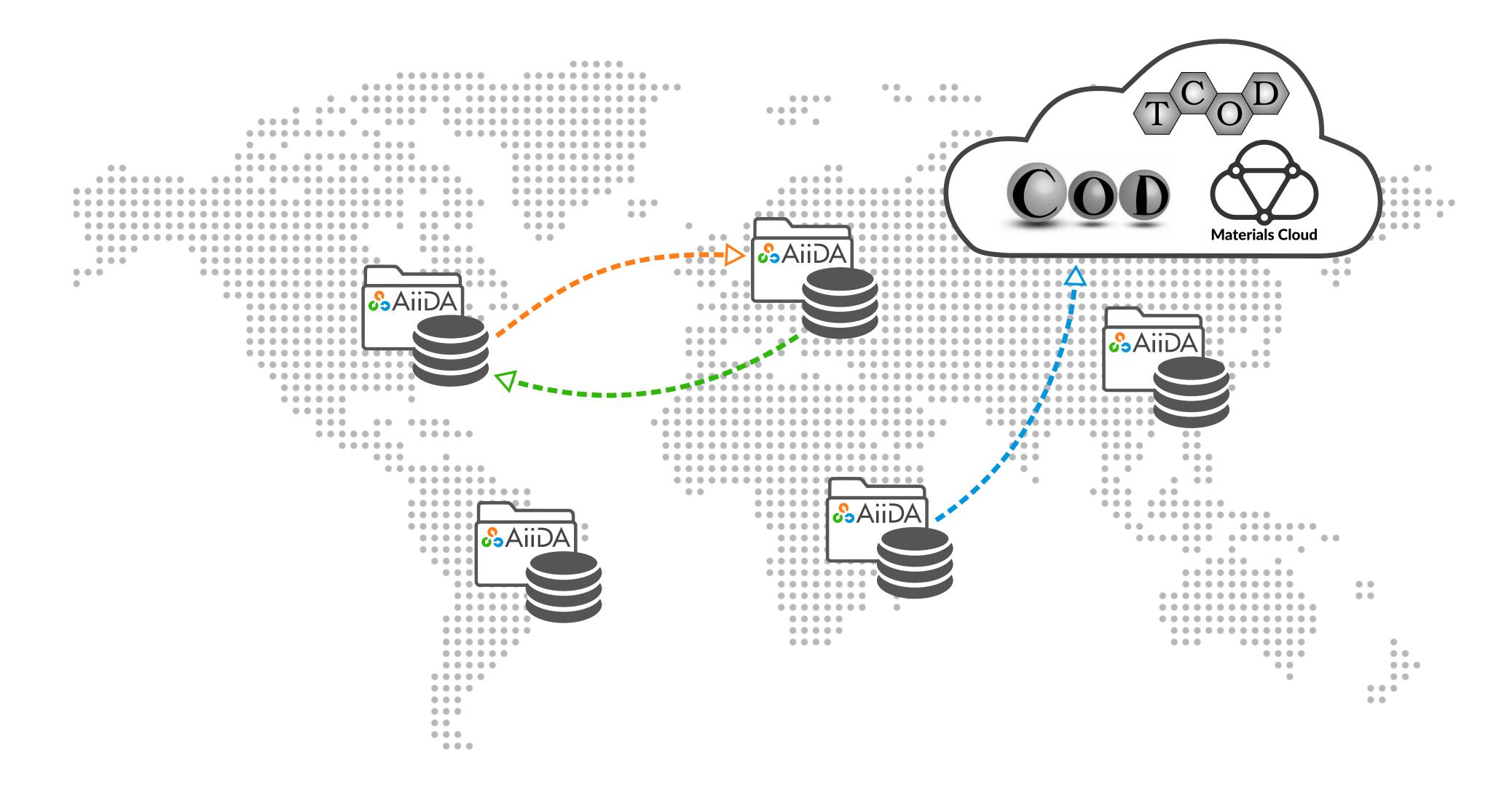
C=Int(5)

<Int: uuid: e25b98c2-db06-4451-86dc-4e1b179d00cc (pk: 1547) value: 35>

Sharing in AiiDA: data and graphs

- Private AiiDA instances
- UUIDs to uniquely identify nodes

Data can be shared to other AiiDA repositories
 or to online repositories



Sharing in AiiDA: *not only data* Codes, plugins and workflows













Calculation

Data

Parsers

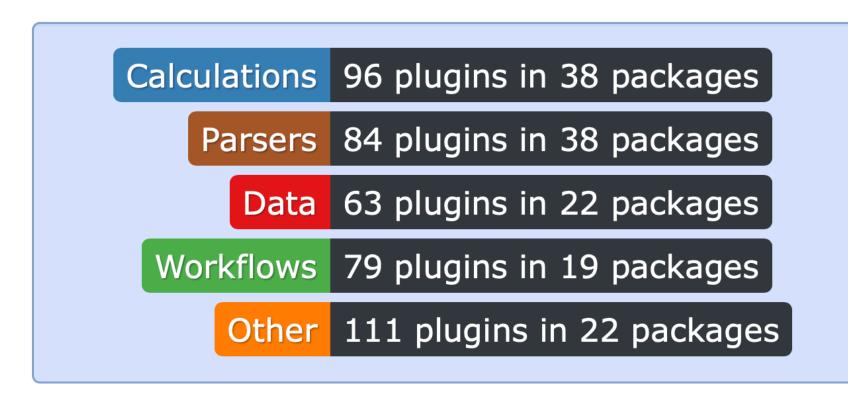
Transport and scheduler

Workflows

Importers & exporters

AiiDA plugin registry

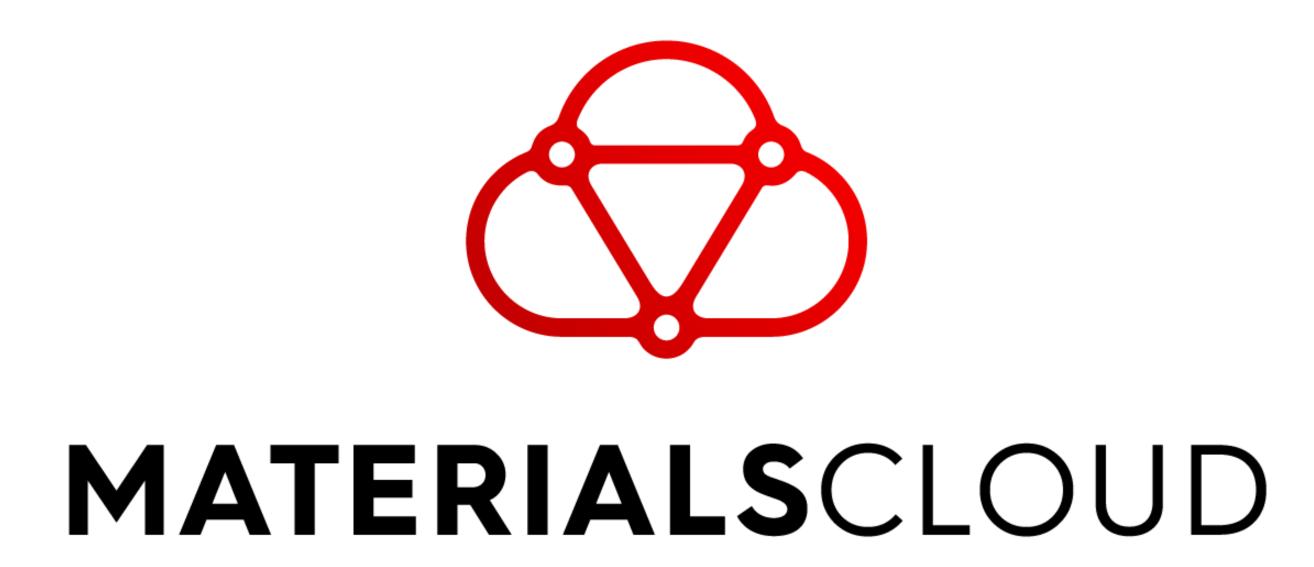
Registered plugin packages: 48



- Plugins are collected in the AiiDA plugin registry
- Over 90 different codes currently supported, with almost 80 workflows
- Many are community-contributed

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

OPEN SCIENCE PLATFORM:



https://www.materialscloud.org

Materials Cloud

- •AiiDA is the 'engine', like Git used in production since 2015
- •Materials Cloud is the dissemination platform (like GitHub) and more (cloud computing and data generation platform) online since Dec 2017

ARCHIVE

- Long-term storage
- Metadata protocols
- DOIs

•• EXPLORE

- AiiDA graph browser
- Raw data + provenance (inputs, outputs)

LEARN

- Educational materials
- Videos, courses, tutorials

% WORK

- Simulation tools and services
- AiiDA lab in the cloud or on premises

V DISCOVEF

- Curated datasets
- Interactive interfaces and visualizations

Open data sharing: Archive, Discover, Explore

materialscloud:2017.0008







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

Authors: 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
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DOI 10.24435/materialscloud:2017.0008/v2 (/ersion v2, submitted on 21 March 2018)

How to cite this entry

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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 from 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, etc.) together with the provenance of all data and calculations as stored by AiiDA.

Direct links to Discover & Explore

assigned

to Discover & Materials Cloud sections using this data

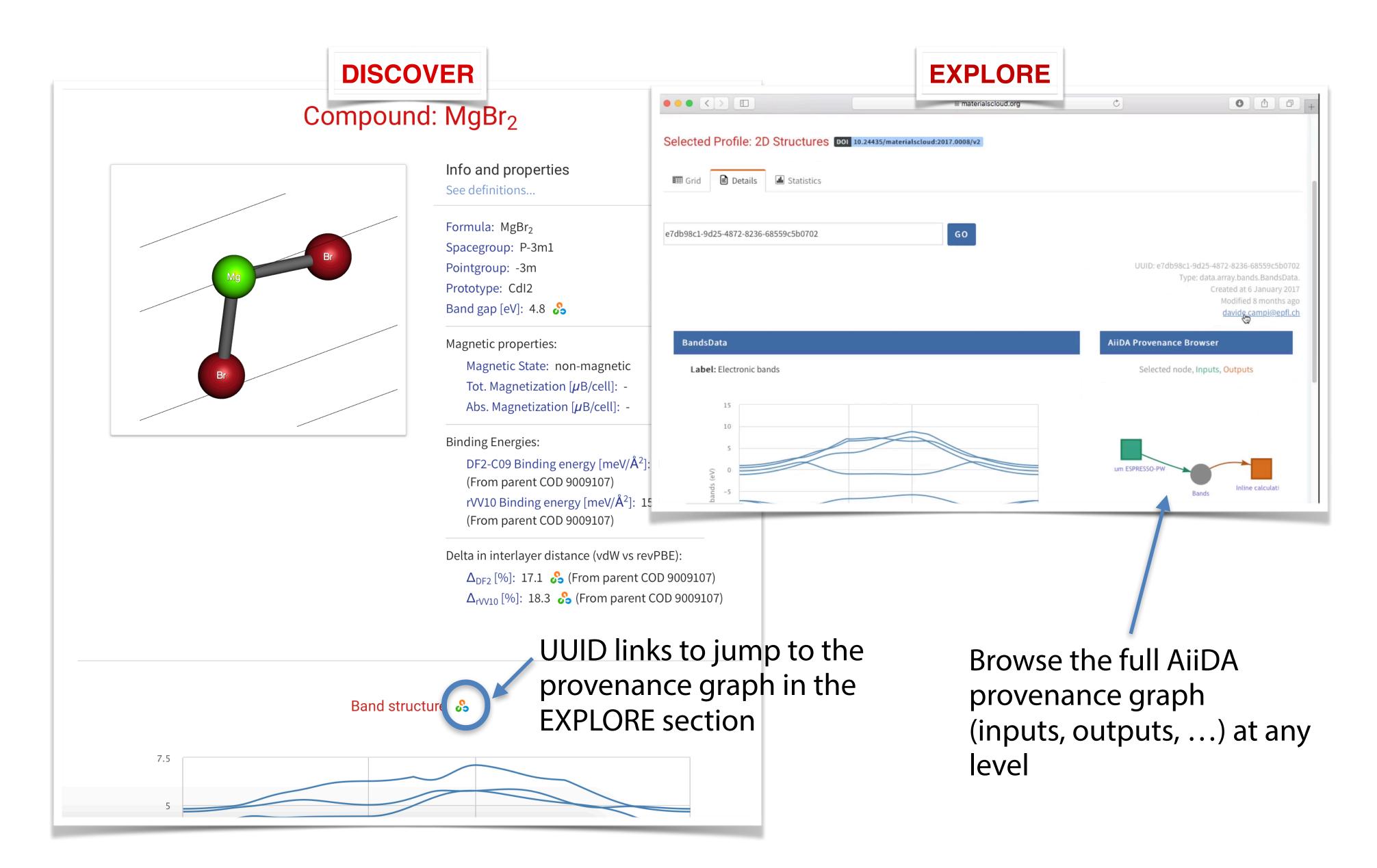
- elect 2d materials via interactive periodic table and view their properties (with links to provenance)
- splore interface providing access to the full database

FAIRsharing.org re3data.org

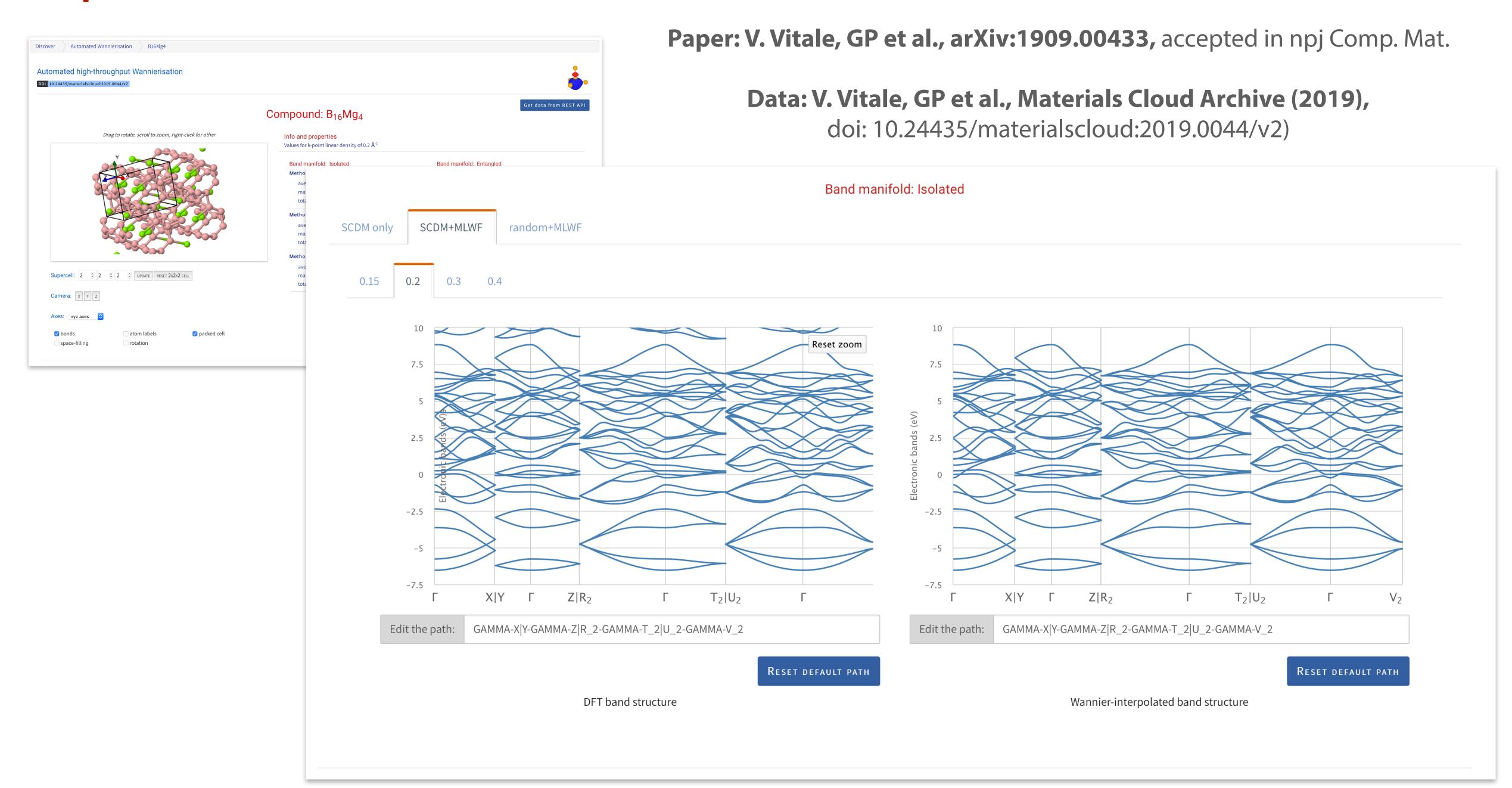
+

Recommended data repository by Nature's journal Scientific Data

Open data sharing: Archive, Discover, Explore



Explore+Discover section for Automatic Wannier functions will be online soon



Data Management Plans (DMPs) and FAIR principles

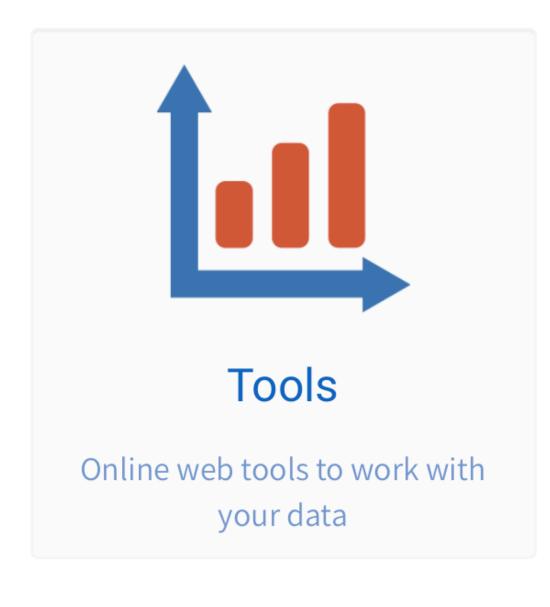
- Combination of AiiDA + Materials Cloud (Discover, Explore, Archive):
 FAIR-compliant sharing
- Findable: DOIs with standardized metadata
- Accessible: web interface to browse data, calculations and provenance, curated data in Discover section
- Interoperable: data linked via the AiiDA directed graph; data structures reusable between different codes
- Reusable: downloadable data, encourage open (CC) licences, reproduce in the AiiDA Lab thanks to full provenance

 We provide DMP templates for researchers using Materials Cloud (and we are coordinating with EMMC for a EU H2020 template)



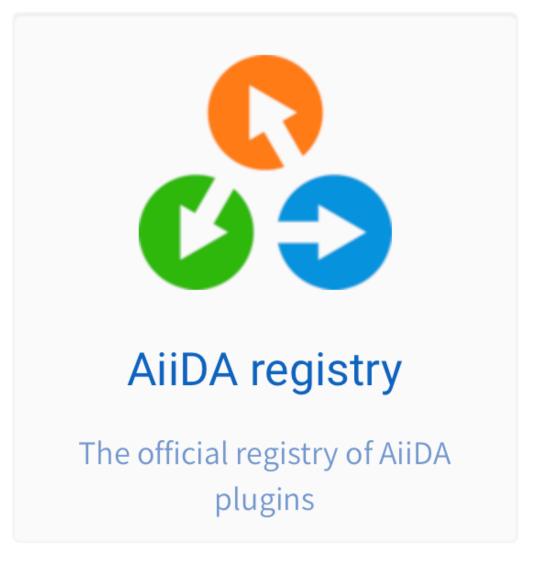
Work with your data in the cloud

You can access a number of services related to AiiDA and computational materials science that require almost no setup. They are either cloud online services or downloadable directly to your machine.



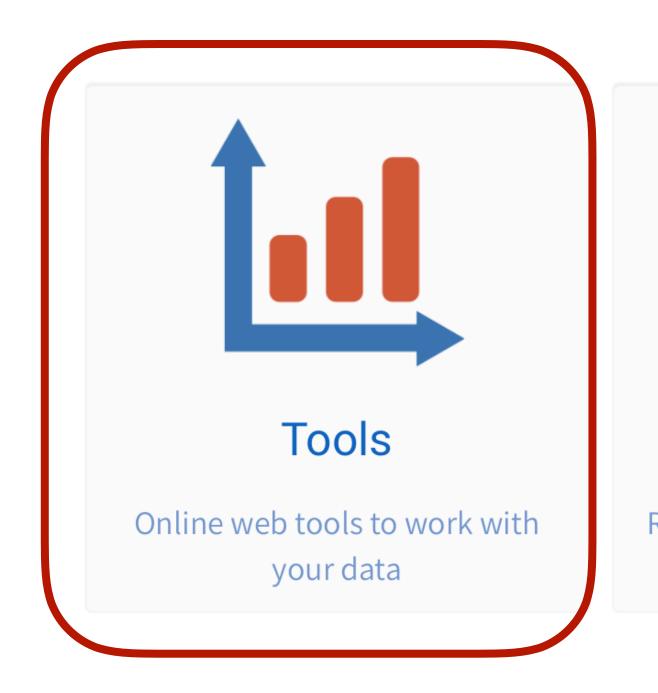




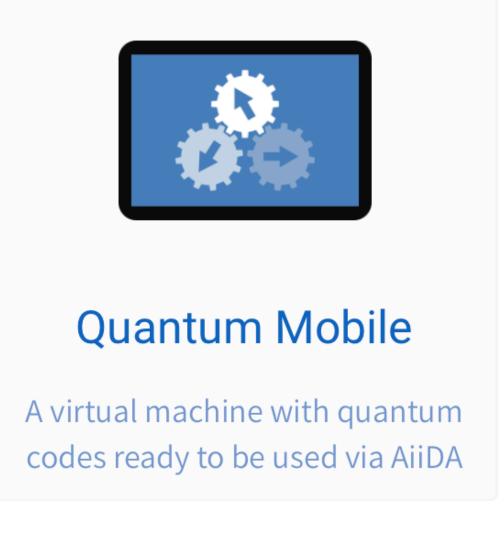


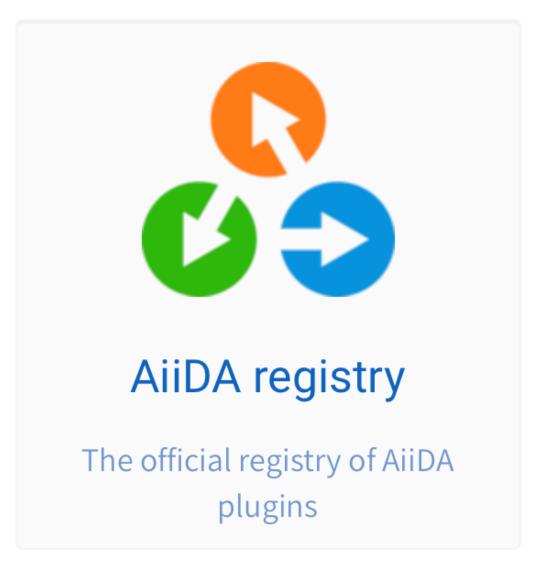
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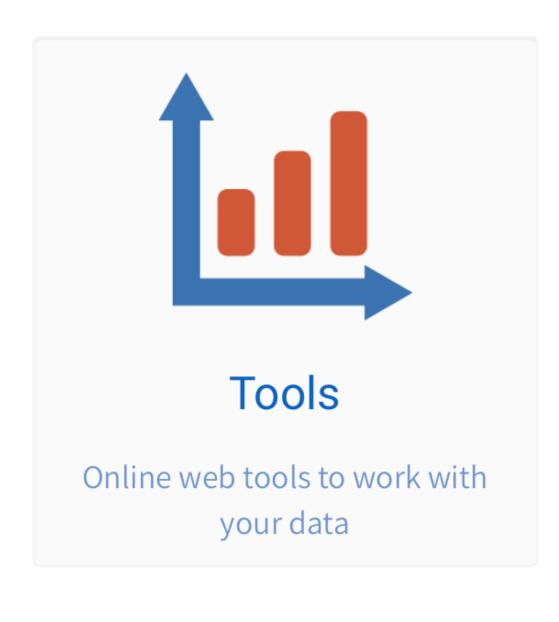




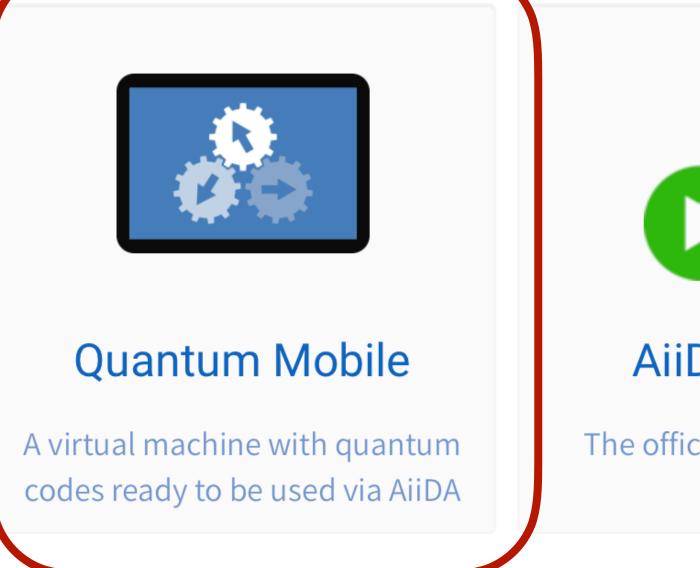
For instance: cloud/web version of **seekpath**

Work with your data in the cloud

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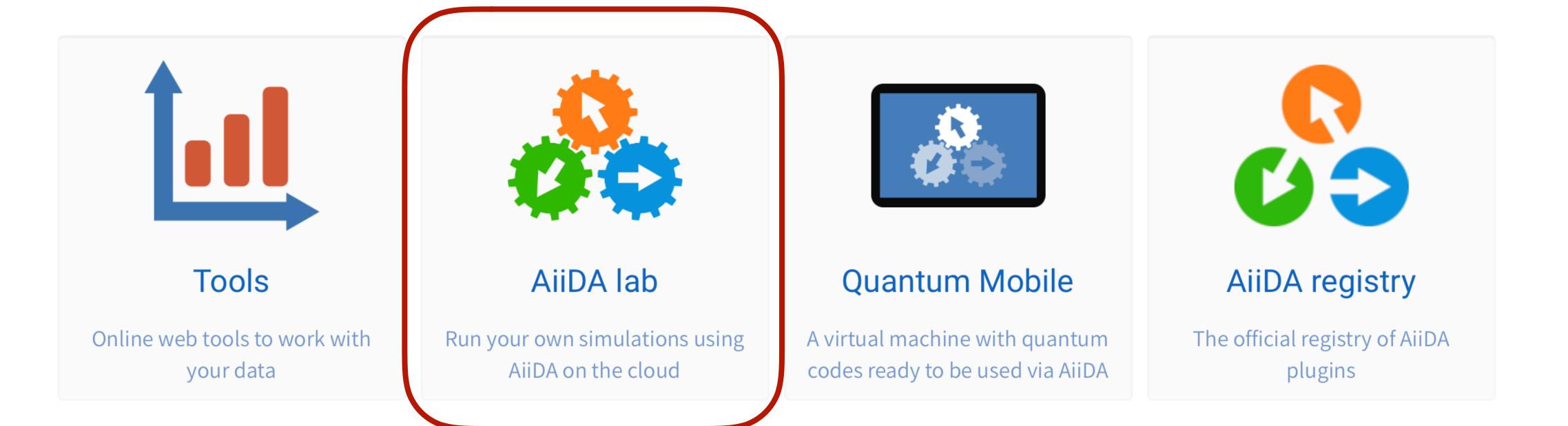


You are using it for this tutorial

Feel free to use it also for other events and tutorials

Work with your data in the cloud

You can access a number of services related to AiiDA and computational materials science that require almost no setup. They are either cloud online services or downloadable directly to your machine.



AiiDA on the cloud using a jupyter interface

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AiiDA core and Materials Cloud authors

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All AiiDA core contributors
All contributors for the 45+ plugins

The CSCS support teams









Moreover:

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

Swissuniversities P-5 "Materials Cloud"

Scaling the web platform, extending to more disciplines

H2020 Marketplace (providing data and simulation services in a EU Markeplace platform also for industry)

H2020 Intersect (develop AiiDA workflows to compute transport properties of materials)

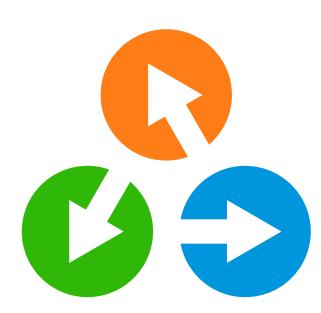
Conclusions (part 2 - AiiDA and Materials Cloud)

- AiiDA: reproducibility and automation engine
 - •turn-key workflows and automation with full provenance tracking as a graph
 - Share data, plugins and workflows
- Materials Cloud: FAIR dissemination portal
 - Archive: Findable long-term storage + Discover (curated data) + Explore (raw AiiDA data)
 - Quantum Mobile as a tool for tutorials and schools

Conclusions

- Automated Wannier functions: all data and provenance on Materials Cloud Archive (Valerio Vitale, GP *et al*, Materials Cloud Archive (2019), doi: 10.24435/materialscloud:2019.0044/v2)
- Archive entry contains a reproducible Virtual Machine to run the same (or new) simulations
- Workflows now updated to most recent version of AiiDA, you will try in the tutorial
- Discover+Explore sections for the automatic Wannier functions project: soon online

Contacts



Website: http://www.aiida.net

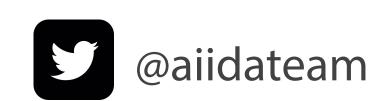
Docs: http://aiida-core.readthedocs.io

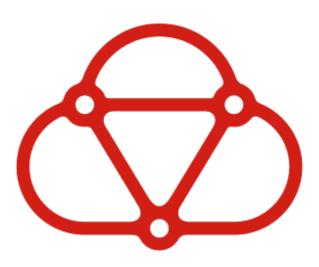
Git repo: https://github.com/aiidateam/aiida_core/

Plugin registry: http://aiidateam.github.io/aiida-registry



https://www.facebook.com/aiidateam





Materials Cloud: http://www.materialscloud.org

- AiiDA Lab: http://aiidalab.materialscloud.org

- Archive: http://archive.materialscloud.org

Quantum Mobile: http://www.materialscloud.org/work/

quantum-mobile