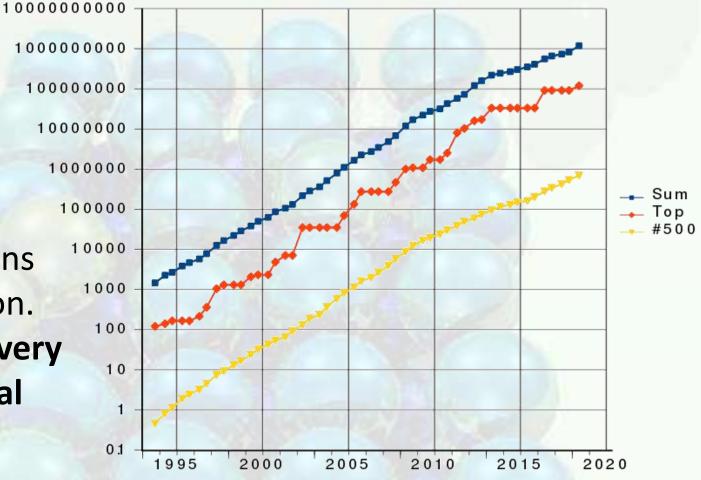
THE GREAT ACCELERATION IN THE DESIGN AND DISCOVERY OF NOVEL MATERIALS Nicola Marzari, EPFL & PSI

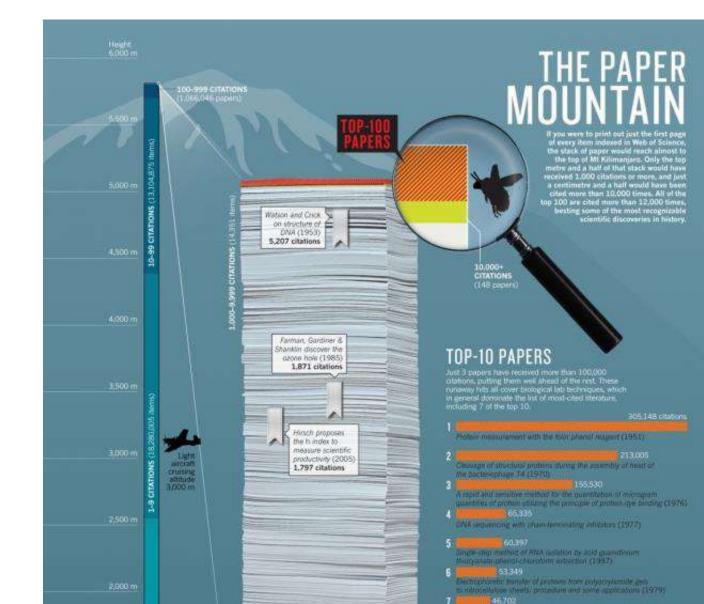
THE RISE OF COMPUTATIONAL SCIENCE

A calculation that took one year in 1992 takes one second in 2021 (33-million-fold increase).

And this is just with bits: neurons are in, and qubits on the horizon. 21st-century science and discovery will be driven by computational science.



IMPACT OF COMPUTATIONAL QUANTUM MECHANICS



THE TOP 100 PAPERS: 12 papers on densityfunctional theory in the top-100 most cited papers in the entire scientificmedical-engineering literature, ever.

NATURE, OCT 2014

MOST CITED PAPERS IN THE HISTORY OF APS

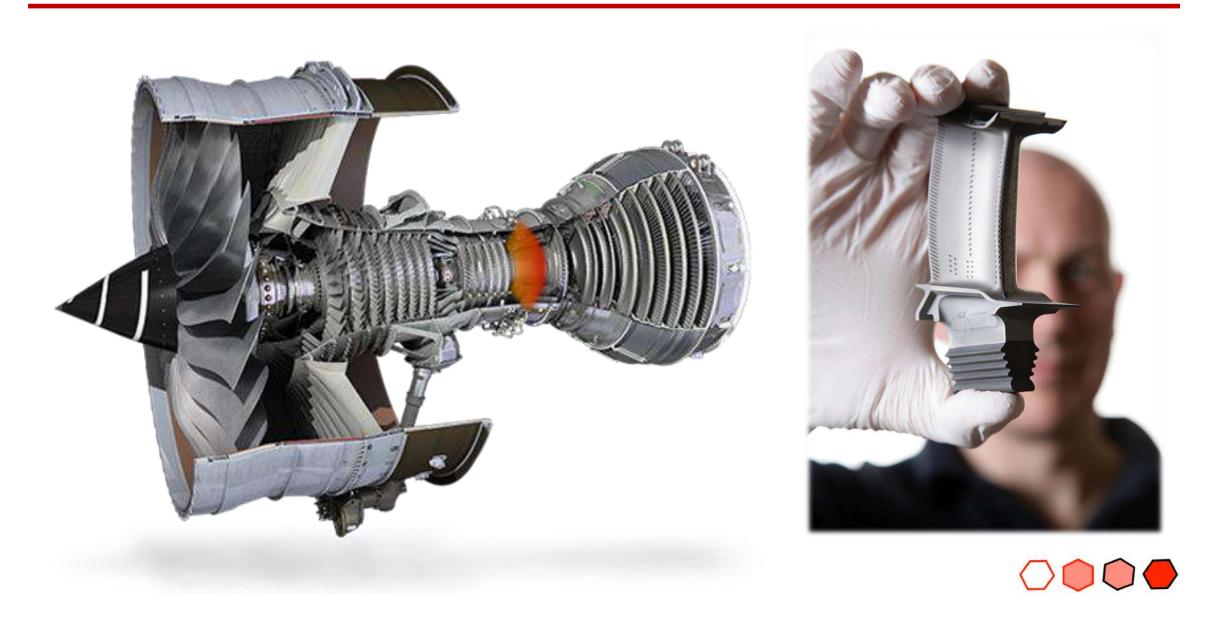
	Journal	# cites	Title	Author(s)
1	PRL (1996)	78085	Generalized Gradient Approximation Made Simple	Perdew, Burke, Ernzerhof
2	PRB (1988)	67303	Development of the Colle-Salvetti Correlation-Energy	Lee, Yang, Parr
3	PRB (1996)	41683	Efficient Iterative Schemes for Ab Initio Total-Energy	Kresse and Furthmuller
4	PR (1965)	36841	Self-Consistent Equations Including Exchange and Correlation	Kohn and Sham
5	PRA (1988)	36659	Density-Functional Exchange-Energy Approximation	Becke
6	PRB (1976)	31865	Special Points for Brillouin-Zone Integrations	Monkhorst and Pack
7	PRB (1999)	30940	From Ultrasoft Pseudopotentials to the Projector Augmented	Kresse and Joubert
8	PRB (1994)	30801	Projector Augmented-Wave Method	Blochl
9	PR (1964)	30563	Inhomogeneous Electron Gas	Hohenberg and Kohn
10	PRB (1993)	19903	Ab initio Molecular Dynamics for Liquid Metals	Kresse and Hafner
11	PRB (1992)	17286	Accurate and Simple Analytic Representation of the Electron	Perdew and Wang
12	PRB (1990)	15618	Soft Self-Consistent Pseudopotentials in a Generalized	Vanderbilt
13	PRB (1992)	15142	Atoms, Molecules, Solids, and Surfaces - Applications of the	Perdew, Chevary,
14	PRB (1981)	14673	Self-Interaction Correction to Density-Functional Approx	Perdew and Zunger
15	PRB (1986)	13907	Density-Functional Approx. for the Correlation-Energy	Perdew
16	RMP (2009)	13513	The Electronic Properties of Graphene	Castro Neto et al.
17	PR (1934)	12353	Note on an Approximation Treatment for Many-Electron Systems	Moller and Plesset
18	PRB (1972)	11840	Optical Constants on Noble Metals	Johnson and Christy
19	PRB (1991)	11580	Efficient Pseudopotentials for Plane-Wave Calculations	Troullier and Martins
20	PRL (1980)	10784	Ground-State of the Electron-Gas by a Stochastic Method	Ceperley and Alder

Marzari (11 Apr 2019)

THE RISE OF MATERIALS SCIENCE

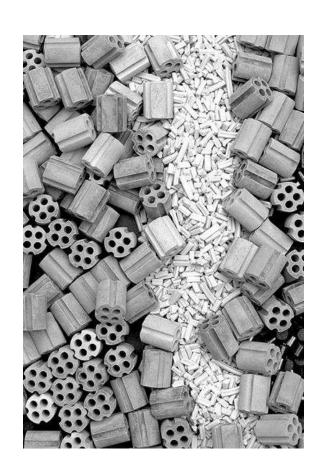


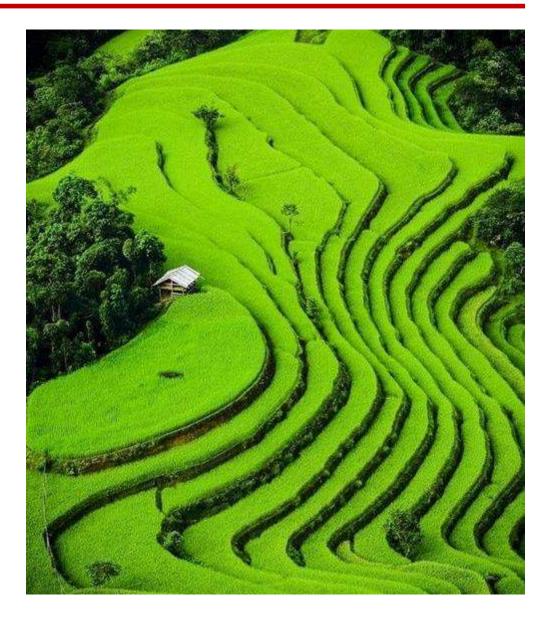
IF WE FLY AGAIN...



THE MOST IMPORTANT MATERIAL EVER?







MATERIALS ARE KEY TO SOCIETAL WELL BEING

We need novel materials for:

- Energy harvesting, conversion, storage, efficiency
- Environmental protection and reparation
- High-tech and high-value industries
- Information and communication technologies
- Health care and biomedical engineering
- Pharmaceuticals (crystallization, stability, polytypes)
- Monitoring, provenance, and safety of foods
- Fundamental science (graphene and 2D materials, topological insulators, entangled spins for quantum computing, high-T_c)
- Experimental science (detectors, sensors, magnets)

3 Technologies That Could Create Trillion-Dollar Markets Over the Next Decade

By Greg Satell Updated April 21, 2019 9:00 a.m. ET



Yet today, we're in the midst of a materials revolution. Powerful simulation techniques, combined with increased computing power and machine learning, are enabling researchers to automate much of the discovery process, vastly accelerating the development of new materials

BARRON'S (April 2019)



MATERIALS MODELLING

The frontiers and the challenges

Materials simulations have become a dominant force in the world of science and technology. The intellectual challenges lying ahead to sustain such a paradigm shift are discussed.

Nicola Marzari

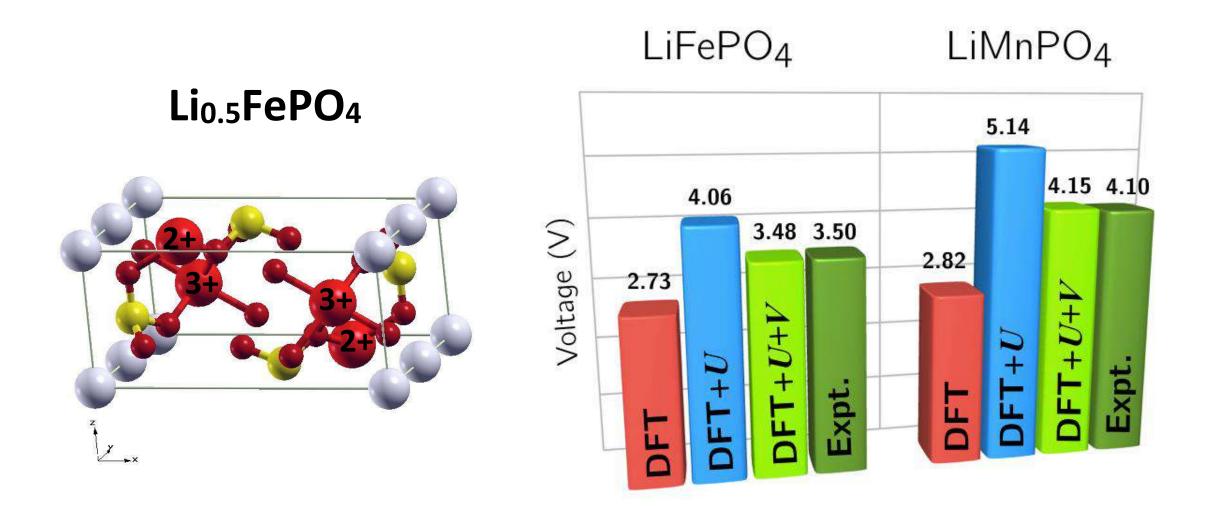
1) PREDICTIVE ACCURACY

2) REALISTIC COMPLEXITY

3) MATERIALS' INFORMATICS

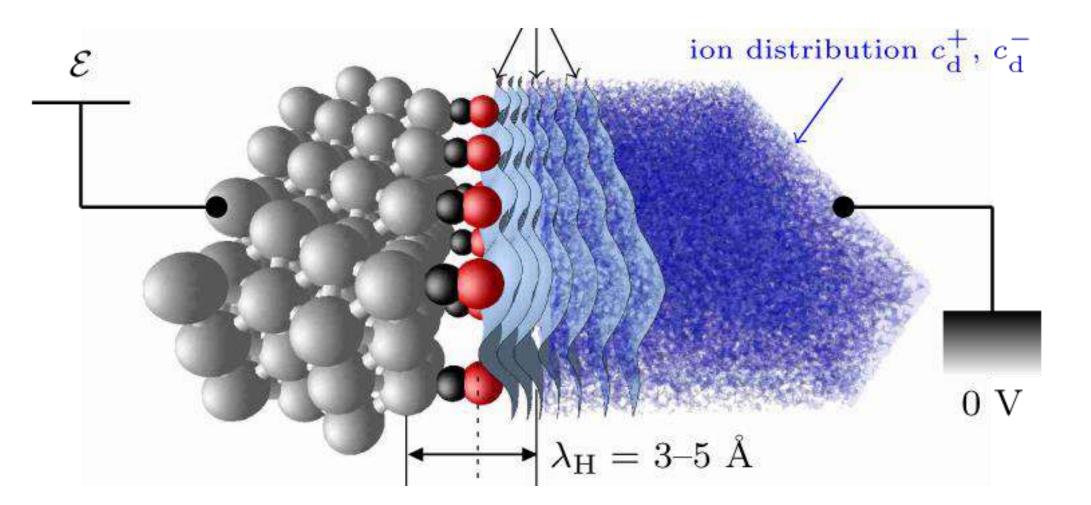
NATURE MATERIALS | VOL 15 | APRIL 2016 | www.nature.com/naturematerials

HUBBARD FUNCTIONALS (DFT+U+V)



M. Cococcioni and N. Marzari, Phys. Rev. Materials 3, 033801 (2019).

COMPLEXITY: MULTI-SCALE MODELING FOR ELECTROCHEMISTRY



O. Andreussi *et al.,* J. Chem. Phys. 136, 064102 (2012) F. Nattino *et al.,* J. Chem. Phys. 150, 041722 (2019)

COMPLEXITY: PREDICTING THE COLOUR OF A MATERIAL













G. Prandini, G.M. Rignanese, and N. Marzari, npj Computational Materials 5, 129 (2019)



COMPLEXITY: MULTI-PHYSICS MODELING OF TRANSPORT

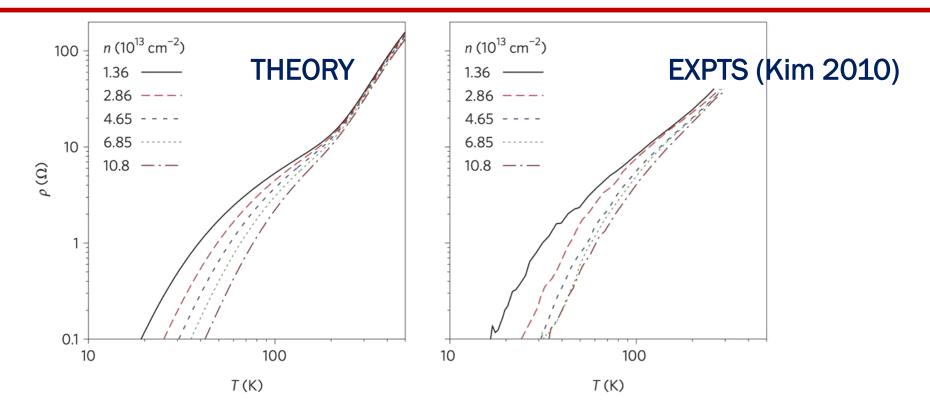
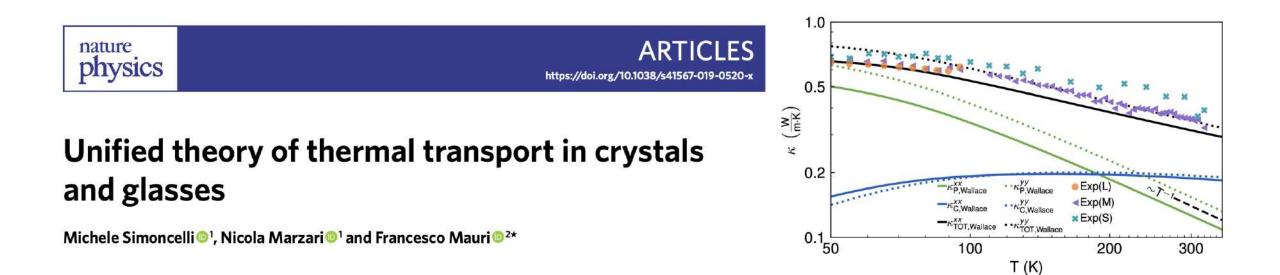


Figure 1 Electrical resistivity of graphene as a function of temperature and doping (ρ , electrical resistivity; *T*, temperature; *n*, carrier density). Left panel: first-principles results obtained using a combination of density-functional perturbation theory, many-body perturbation theory and Wannier interpolations to solve the Boltzmann transport equation. Right panel: experimental data. Adapted from ref. 4, American Chemical Society.

C.-H. Park *et al.,* Nano Letters (2014) T. Y. Kim, C.-H. Park, and N. Marzari, Nano Letters (2016)

THERMOELECTRICS AND THERMAL BARRIER COATINGS



$$\begin{split} \kappa^{\alpha\beta} &= \kappa_{\rm P}^{\alpha\beta} + \frac{\hbar^2}{k_B T^2} \frac{1}{\mathcal{V}N_{\rm c}} \sum_{\boldsymbol{q}} \sum_{s \neq s'} \frac{\omega(\boldsymbol{q})_s + \omega(\boldsymbol{q})_{s'}}{2} V^{\alpha}(\boldsymbol{q})_{s,s'} V^{\beta}(\boldsymbol{q})_{s',s} \times \\ & \times \frac{\omega(\boldsymbol{q})_s \bar{N}^T(\boldsymbol{q})_s [\bar{N}^T(\boldsymbol{q})_s + 1] + \omega(\boldsymbol{q})_{s'} \bar{N}^T(\boldsymbol{q})_{s'} [\bar{N}^T(\boldsymbol{q})_{s'} + 1]}{4[\omega(\boldsymbol{q})_{s'} - \omega(\boldsymbol{q})_s]^2 + [\Gamma(\boldsymbol{q})_s + \Gamma(\boldsymbol{q})_{s'}]^2} [\Gamma(\boldsymbol{q})_s + \Gamma(\boldsymbol{q})_{s'}] \end{split}$$

RELIABLY, REPRODUCIBLY, HIGH-THROUGHPUT

VOLUME 88, NUMBER 25 PHYSICAL REVIEW LETTERS

24 JUNE 2002

Combined Electronic Structure and Evolutionary Search Approach to Materials Design

G. H. Jóhannesson, T. Bligaard, A. V. Ruban, H. L. Skriver, K. W. Jacobsen, and J. K. Nørskov Center for Atomic-Scale Materials Physics, Department of Physics, Technical University of Denmark, DK-2800, Lyngby, Denmark (Received 20 February 2002; published 10 June 2002)

We show that density functional theory calculations have reached an accuracy and speed making it possible to use them in conjunction with an evolutionary algorithm to search for materials with specific properties. The approach is illustrated by finding the most stable four component alloys out of the 192016 possible fcc and bcc alloys that can be constructed out of 32 different metals. A number of well known and new "super alloys" are identified in this way.

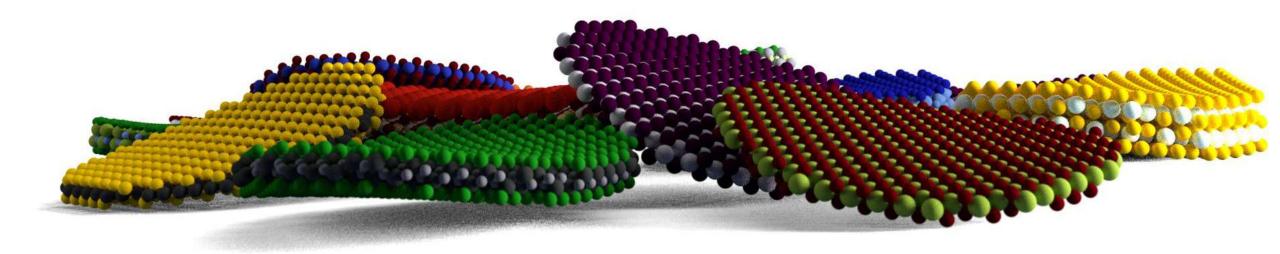
DOI: 10.1103/PhysRevLett.88.255506

PACS numbers: 81.05.Bx, 61.66.Dk, 71.15.Mb

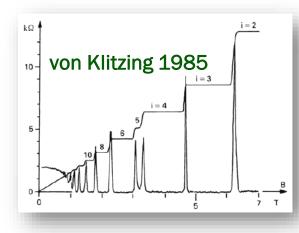


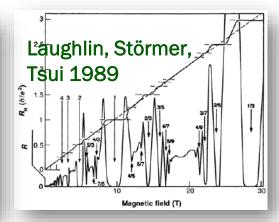


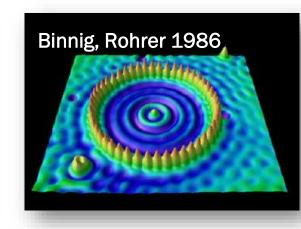
EXAMPLE: COMPUTATIONAL EXFOLIATION OF ALL KNOWN INORGANIC MATERIALS

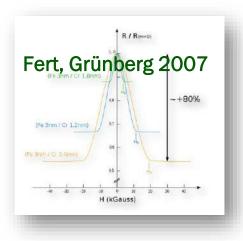


PHYSICS AND CHEMISTRY IN LOW DIMENSIONS





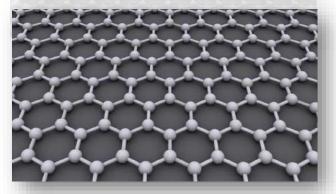


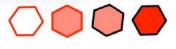




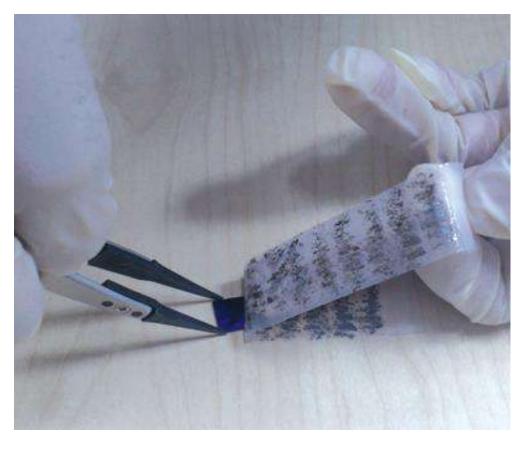




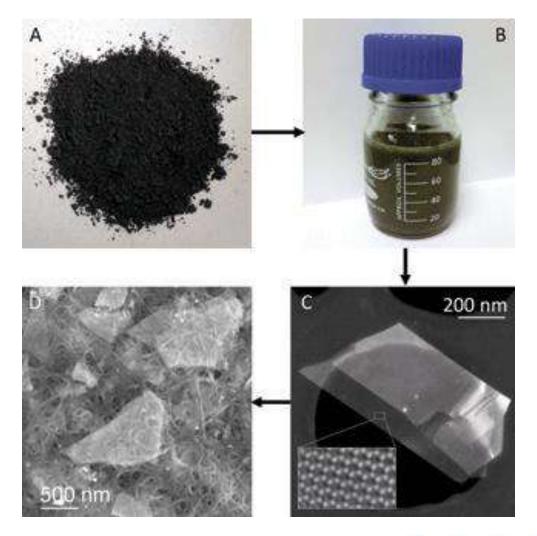


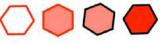


HOW DO WE PRODUCE 2D MATERIALS?

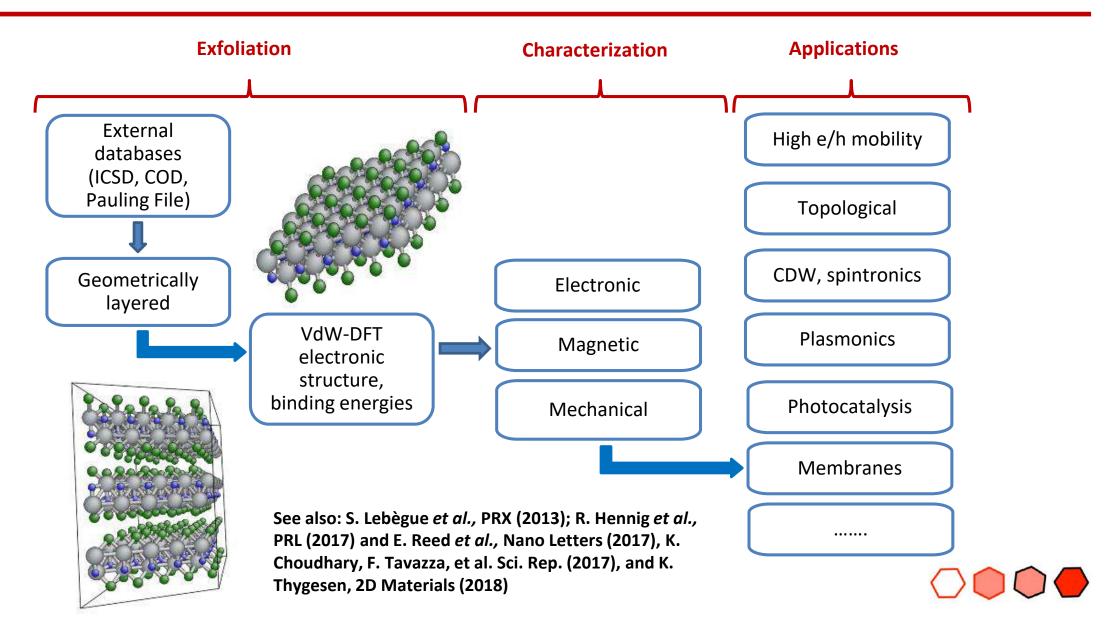


Mechanical (e.g. Geim/Novoselov, fig. from Nature/NUS) or liquid exfoliation (e.g. Nicolosi/Coleman, fig. from Science), electrochemical intercalation. Also, bottomup: CVD and wet chemical synthesis.

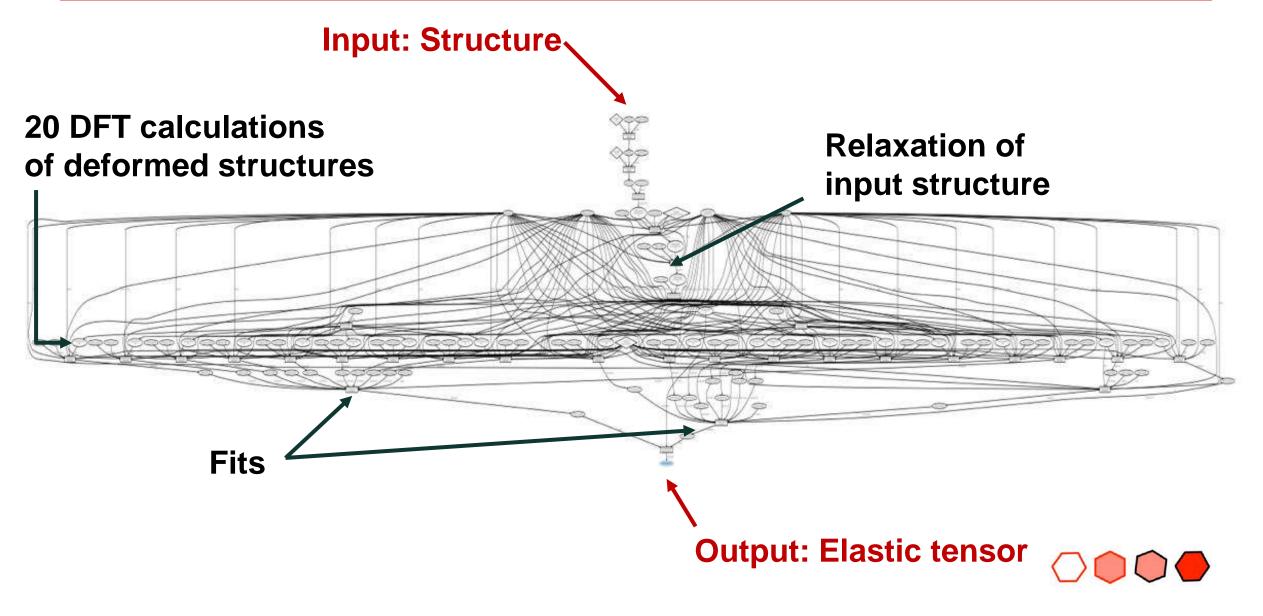




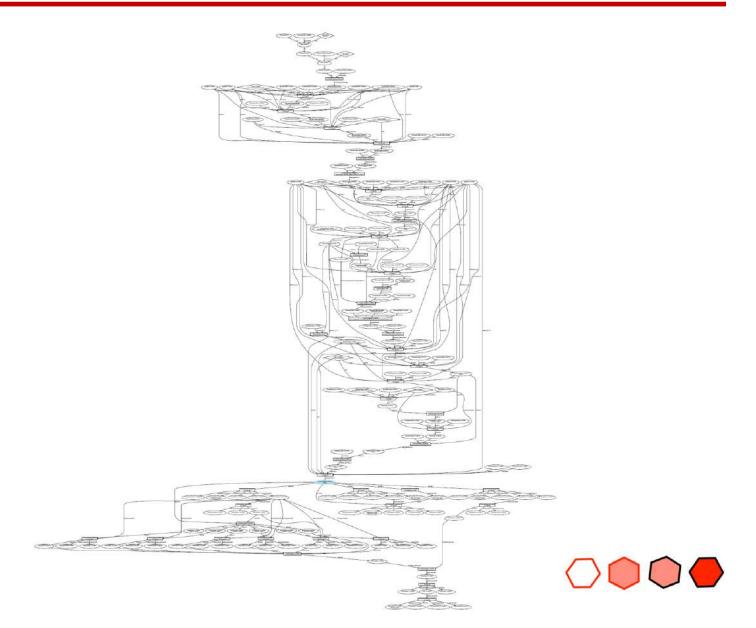
HIGH-THROUGHPUT COMPUTATIONAL EXFOLIATION



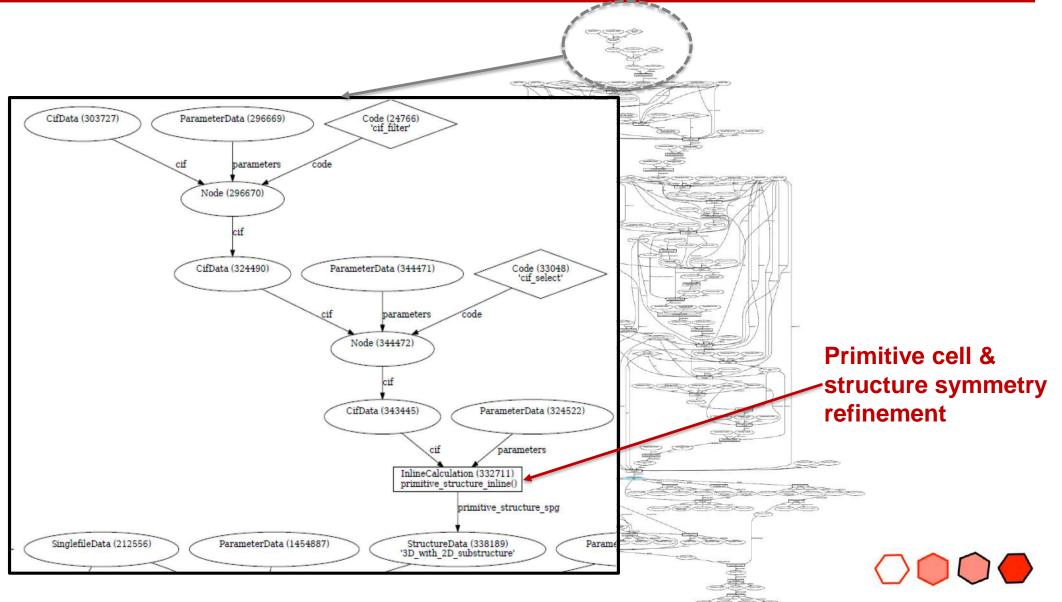
AUTOMATIC WORKFLOWS: FROM STRUCTURE TO PROPERTY



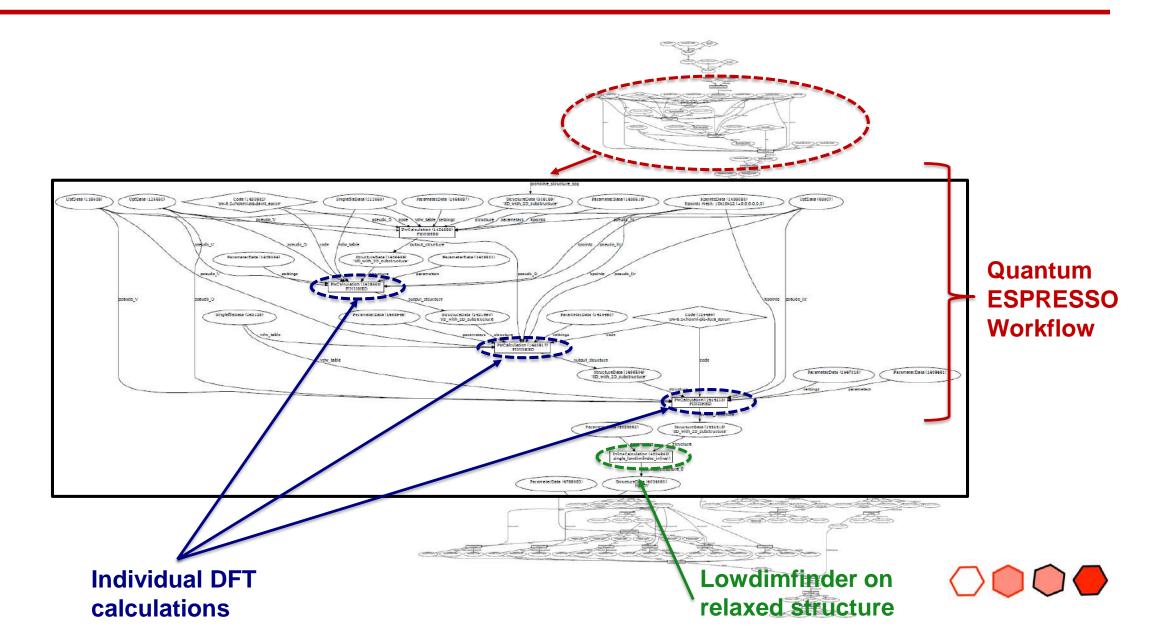
LET'S START FROM A MATERIAL (VOBr₂)



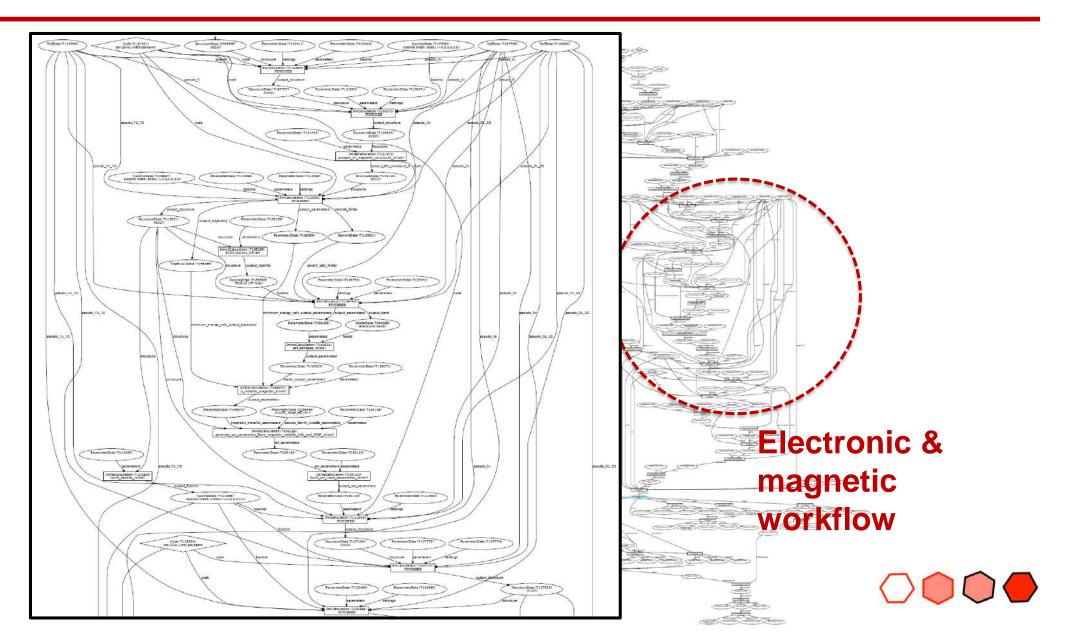
FROM DATABASE ENTRY TO A WORKING STRUCTURE



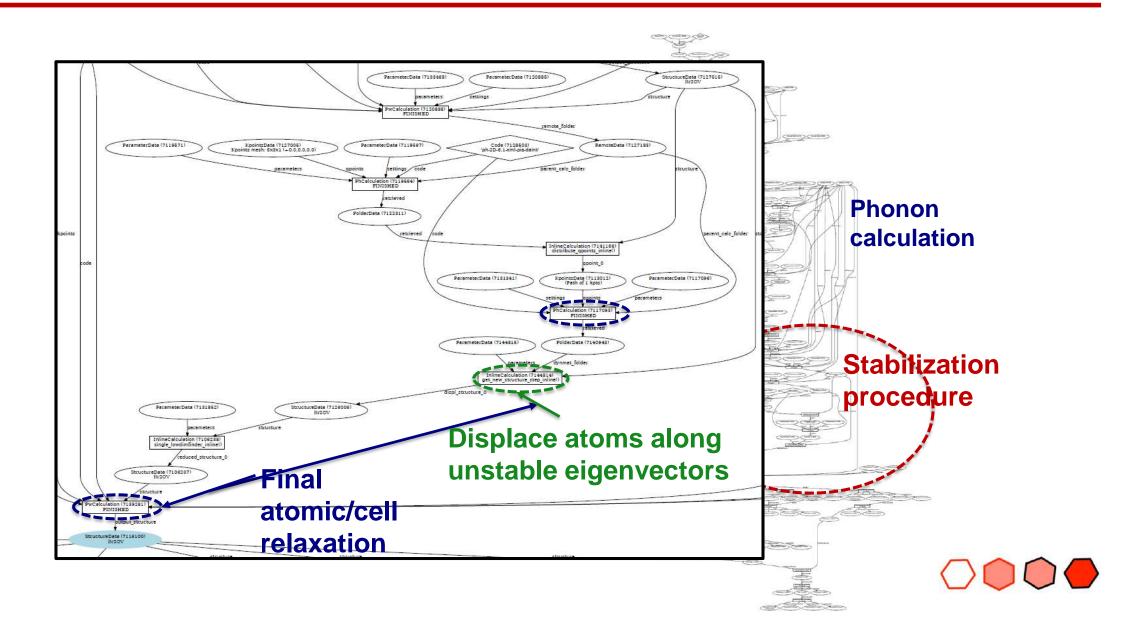
3D RELAXATION



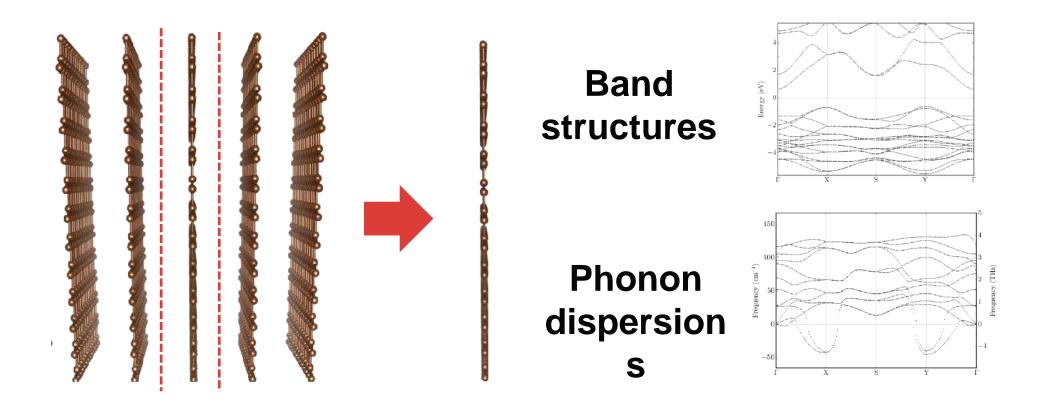
MAGNETIC SCREENING OF THE 2D MONOLAYER

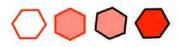


REMOVING MECHANICAL INSTABILITIES



ALL AUTOMATED...





FINALLY...

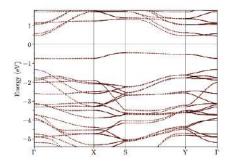
215 VBr₂O (Pmm2)

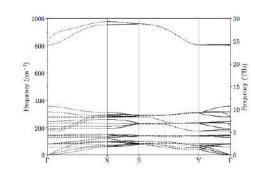
Info and properties (for more details and definitions see page 2)

Formula VBr₂O Spacegroup Pmm2 Prototype VCl_2O (Pmm2) Parent 3D VBr₂O Source DB ICSD DB ID 24381

DF2-C09 Binding energy [meV/Å²] 14.4rVV10 Binding energy [meV/Å²] 21.6Band gap [eV] 0.9 Magnetic State AFM Tot. Magnetization $[\mu_B/\text{cell}]$ 0.0 Abs. Magnetization $[\mu_B/\text{cell}]$ 2.54

Band structure and phonon dispersions





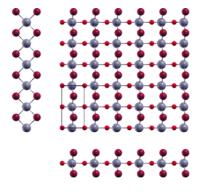
Band structure: energy bands of VBr₂O (66 electrons) in Phonon dispersions: phonon frequencies of VBr₂O (8 a window around the chemical potential and along a highsymmetry path. The number of bands included in the calculation is 80.

atoms/cell) along a high-symmetry path.

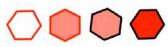
Crystal structure

Structural parameters: cell (top) and atomic positions (bottom) of VBr₂O in cartesian coordinates.

		x [Å]	y [Å]	<i>z</i> [Å]	
$egin{array}{c} a_1\ a_2\ a_3 \end{array}$	3.80622044 0.00000000 0.00000000		0.00000000 7.17029927 0.00000000	$\begin{array}{c} 0.00000000\\ 0.00000000\\ 19.47346306\end{array}$	
		x [Å]	y [Å]	z [Å]	
٠	Br	2.00107500	5.37772439	-1.78545446	
٠	\mathbf{Br}	2.00107500	1.79257489	-1.78545446	
9	V_1	1.70214333	3.58514964	0.00000000	
	V_2	1.70214341	0.00000000	0.00000000	
۲	Br	2.00107500	5.37772439	1.78545446	
٠	Br	2.00107500	1.79257489	1.78545446	
•	O_1	0.06788642	3.58514964	0.00000000	
•	O_2	0.06788661	0.00000000	0.00000000	



Orthographic projections: different views of VBr₂O from the x axis (left), the y axis (bottom) and the z axis (center).

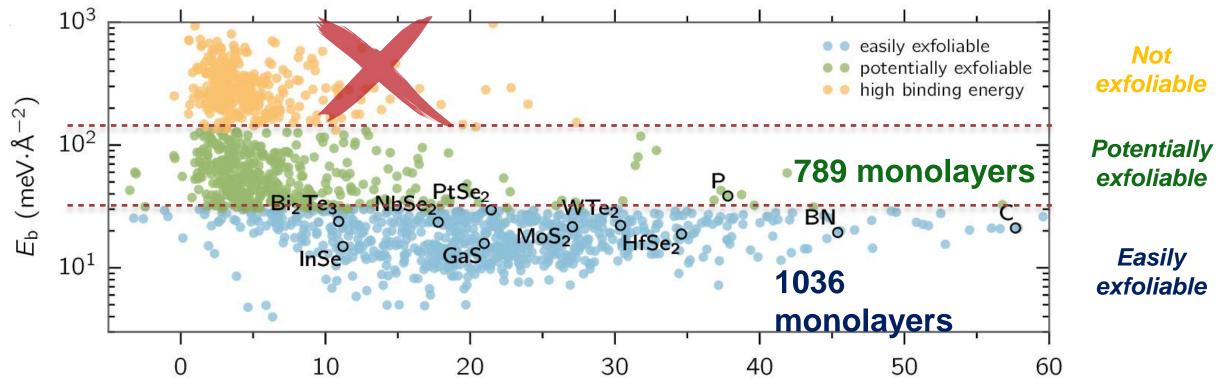


	Unique to COD	Unique to ICSD	Common to both	Total
Entries analyzed	307616	172370		479986*
CIF inputs	99212	87070		186282*
Unique 3D structures	60354	34548	13521	108423
Layered 3D structures	1180	3257	1182	5619

*At this level unicity is not tested



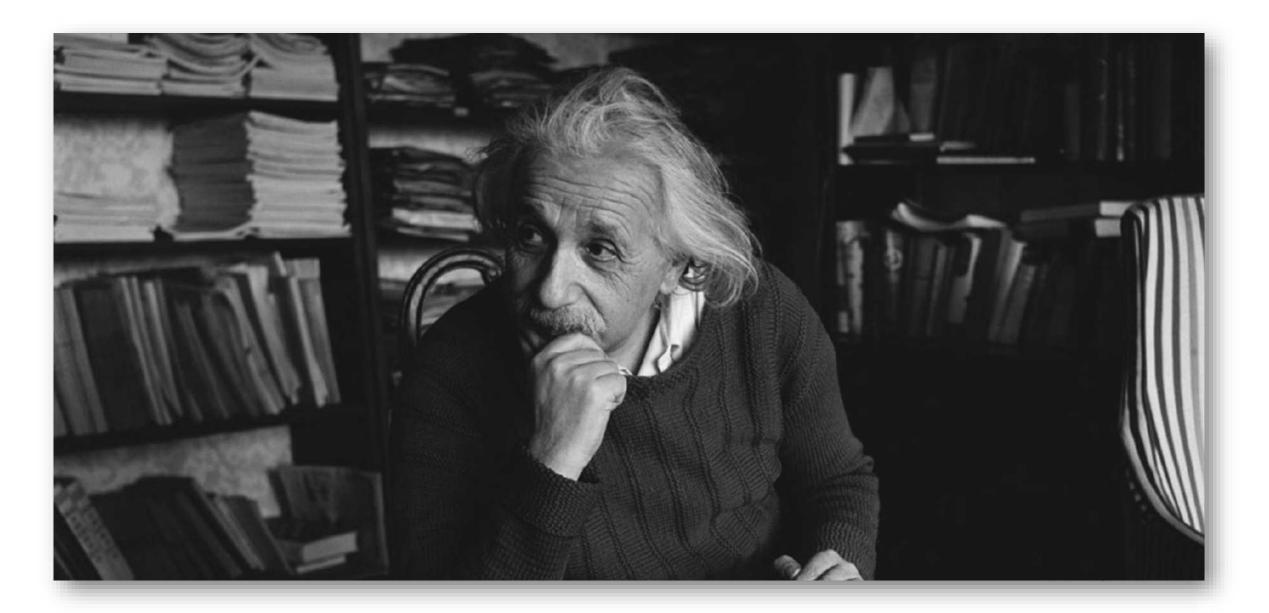
HOW MANY CANDIDATES? QUANTUM SCREENING



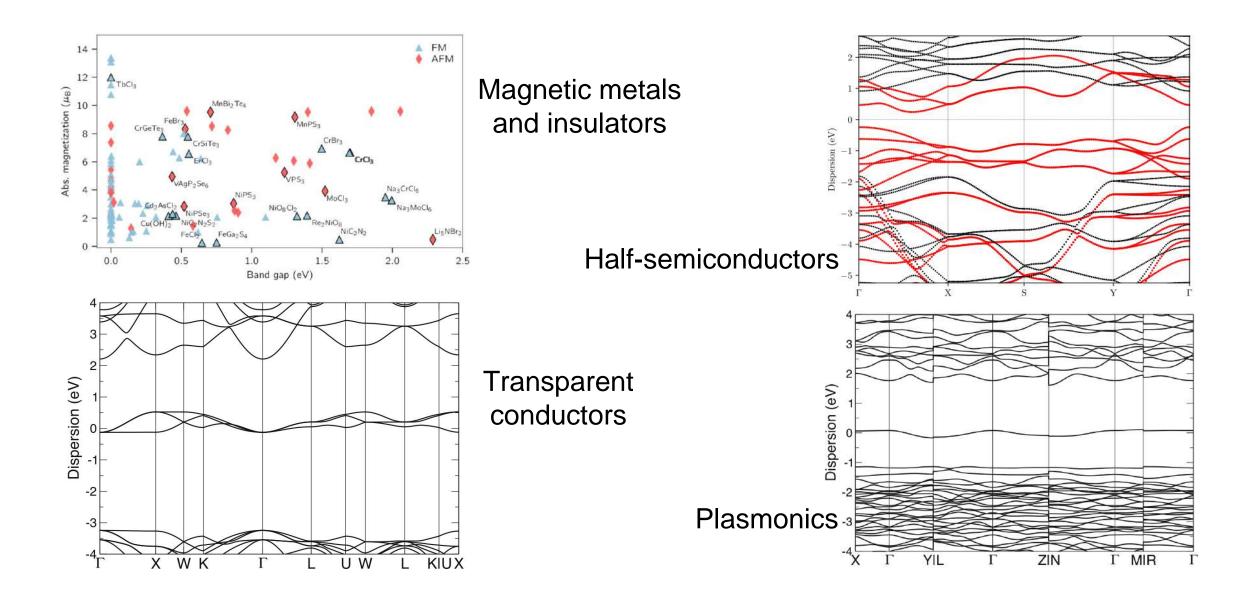
Difference in interlayer distance when computed with/without vdW functionals (%)

- $E_b < 30 \text{ meV/Å}^2$ (DF2-C09) or $E_b < 35 \text{ meV/Å}^2$ (rVV10) \rightarrow 2D, easily exfoliable
- In-between \rightarrow 2D, potentially exfoliable
- $E_b > 130 \text{ meV/Å}^2 \rightarrow \text{not 2D}$ (discarded)

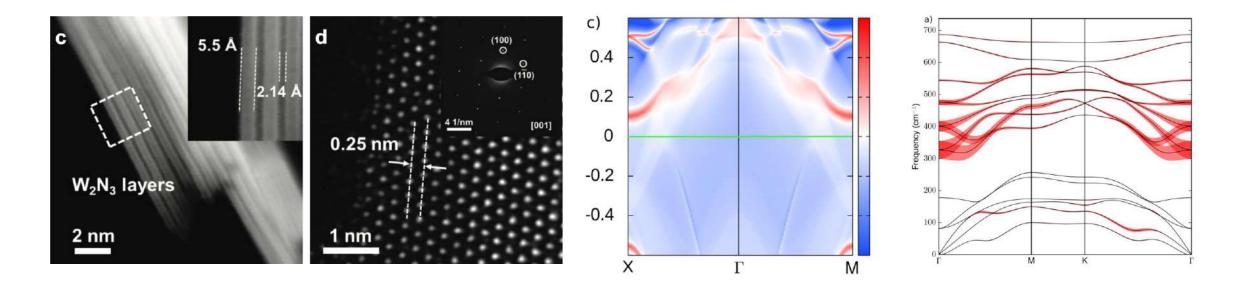
WHAT TO DO NEXT?



FROM ELECTRONICS...



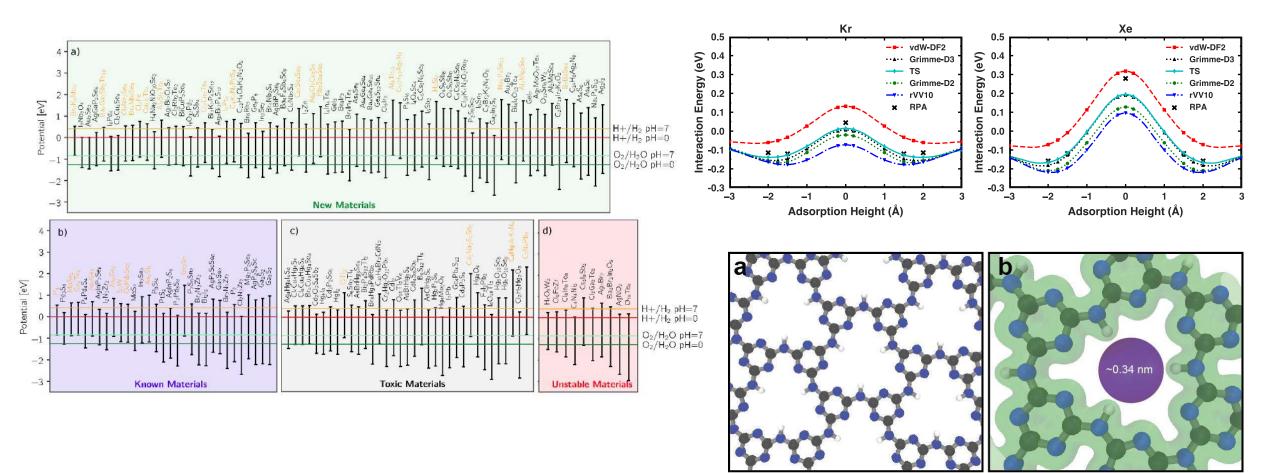
... TO THE LARGEST SUPERCONDUCTING Tc IN 2D...



H. Jun *et al.,* Advanced Materials 31, 1902709 (2019)

D. Campi, S. Kumari, and N. Marzari, Nano Letters 21, 3435 (2021)

...TO MATERIALS: PHOTOCATALYSIS, MEMBRANES

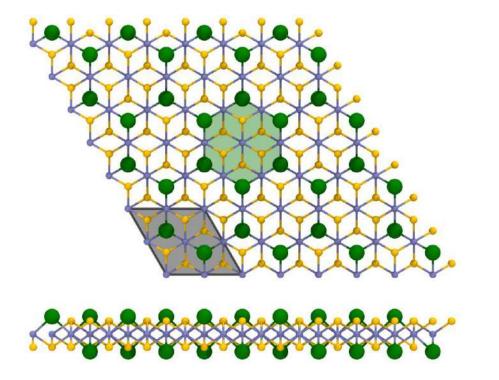


Science Advances (2019), and under review (2021)

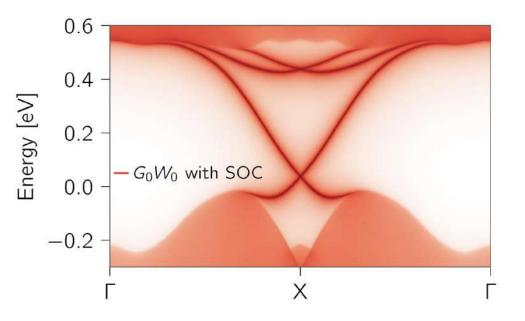
THE DISCOVERY OF JACUTINGAITE



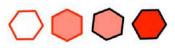
THE DISCOVERY OF JACUTINGAITE



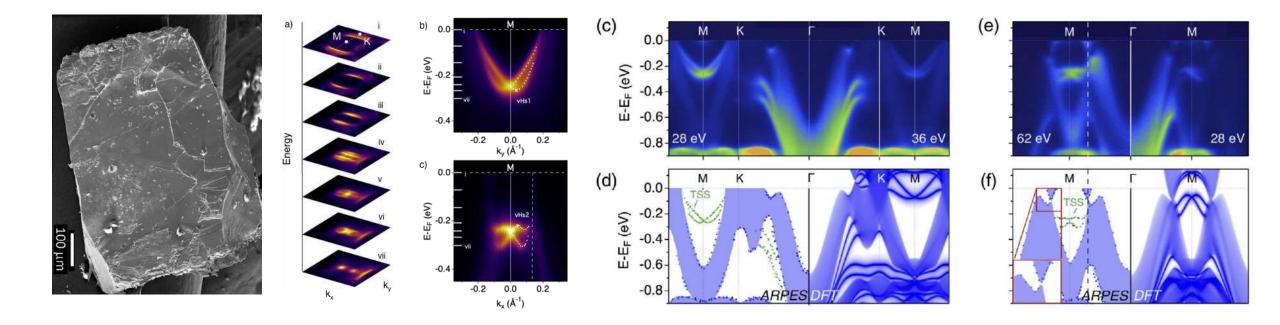
Classified as potentially exfoliable (binding energy of 60 meV Å⁻²)



A. Marrazzo *et al.,* Phys. Rev. Lett. 120, 117701 (2018)



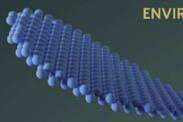
ROOM-TEMPERATURE KANE-MELE QSHI



A. Marrazzo *et al.,* Phys. Rev. Lett. 120, 117701 (2018) I. Cucchi, *et al.,* Phys. Rev. Lett. 124, 106402 (2020) A. Marrazzo, N. Marzari, and M. Gibertini, Phys. Rev. Res. 2, 012063(R) (2020)

nature MARCH 2018 VOL 13 NO 3 www.nature.com/naturenanotechnology nanotechnology

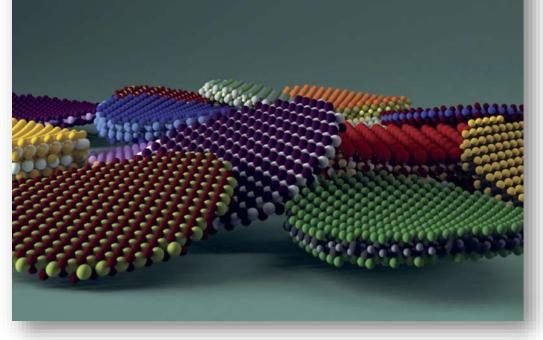
Computational quest for 2D materials



ENVIRONMENTAL NANOTECHNOLOGY Interacting with the community

> ACHROMATIC METALENSES Visible images

> > NEUROMODULATION Wireless excitement



THERE IS PLENTY OF ROOM AT THE

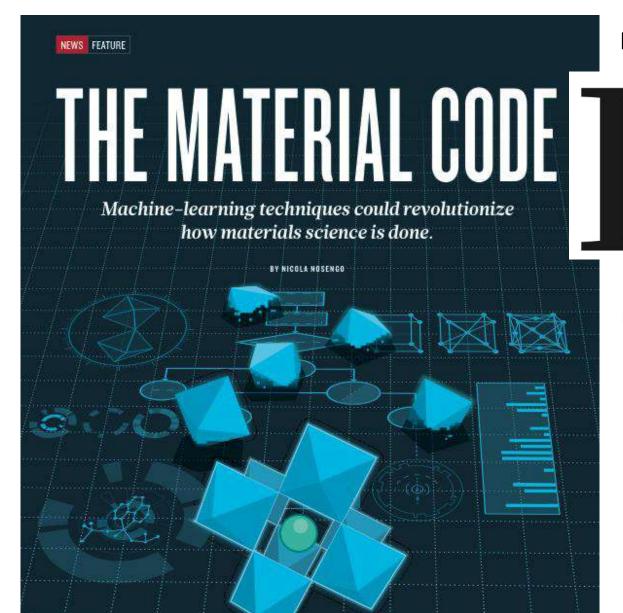
- High electron/hole mobility devices
- Topological insulators, quantum computing
- Ferromagnetic/spintronics in 2D
- Charge-density waves and superconductors
- Plasmonics, transparent conductors

3D layered parents:

- Solid-state ionic conductors
- Hydrogen or oxygen evolution catalysts
- Membranes for filtration/separation
- Piezo, ferro, and thermoelectrics

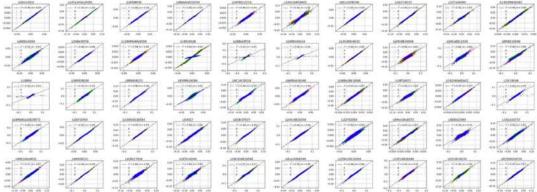
N. Mounet, M. Gibertini, P. Schwaller, D. Campi, A. Merkys, A. Marrazzo, T. Sohier, I. E. Castelli, A. Cepellotti, G. Pizzi and N. Marzari, Nature Nanotechnology 13, 246 (2018)

MACHINE LEARNING AS THE GREAT ACCELERATOR

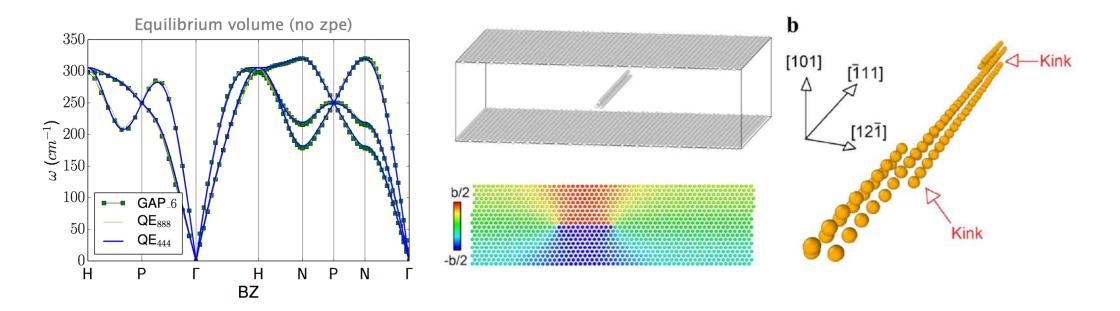


Nature, May 2016

t's a strong contender for the geekiest video ever made: a close-up of a smartphone with line upon line of numbers and symbols scrolling down the screen. But when visitors stop by Nicola Marzari's office, which overlooks Lake Geneva, he can hardly wait to show it off. "It's from 2010," he says, "and this is my cellphone calculating the electronic structure of silicon in real time!"



MACHINE-LEARNING QUANTUM MECHANICS



Dislocation glide by kink-pair nucleation and propagation

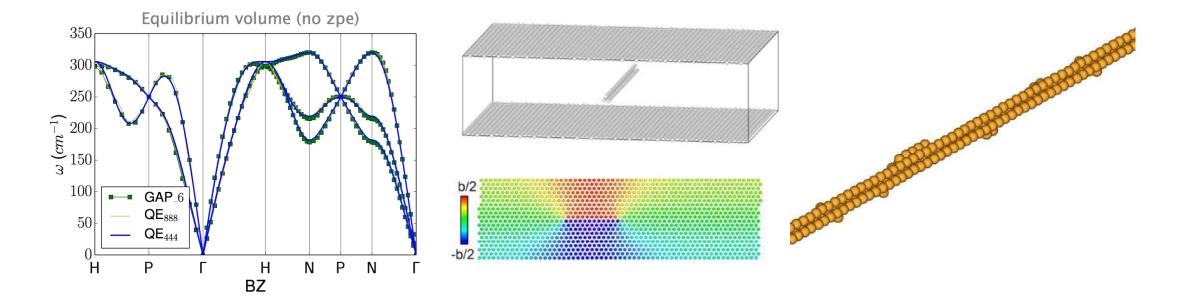
D. Dragoni, T. Duff, G. Csányi, and N. Marzari, Phys. Rev. Materials 2, 013808 (2018)

F. Maresca, D. Dragoni, G. Csányi, N. Marzari, and W. A. Curtin, npj Comput Mater 4, 69 (2018)



MACHINE-LEARNING QUANTUM MECHANICS

Dislocation glide by kink-pair nucleation and propagation

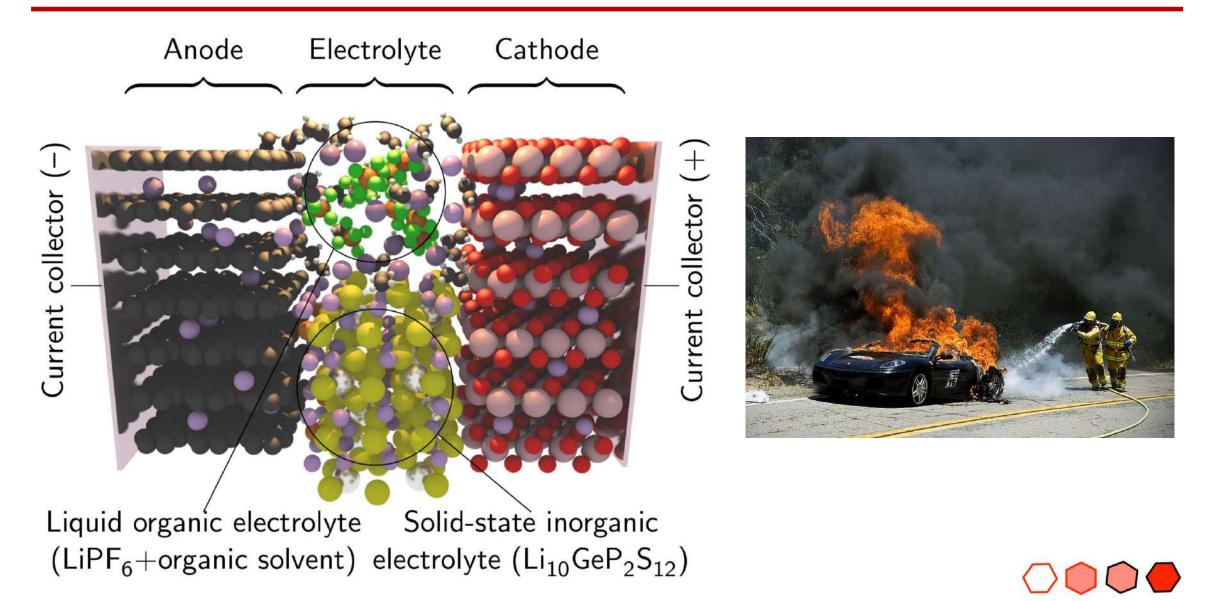


D. Dragoni, T. Duff, G. Csányi, and N. Marzari, Phys. Rev. Materials 2, 013808 (2018)

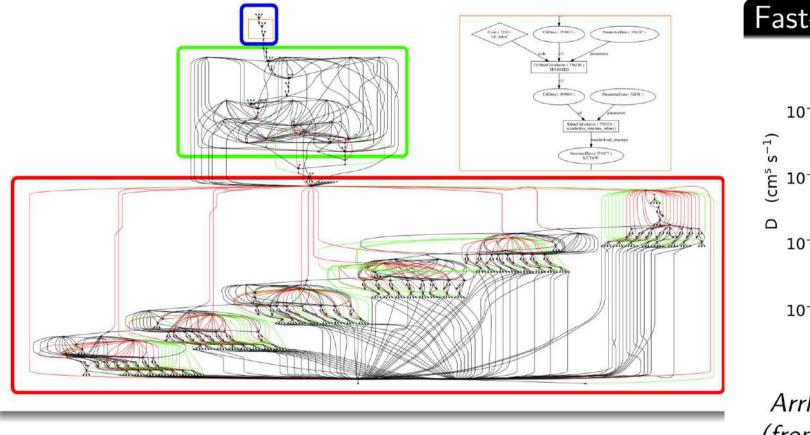
F. Maresca, D. Dragoni, G. Csányi, N. Marzari, and W. A. Curtin, npj Comput Mater 4, 69 (2018)



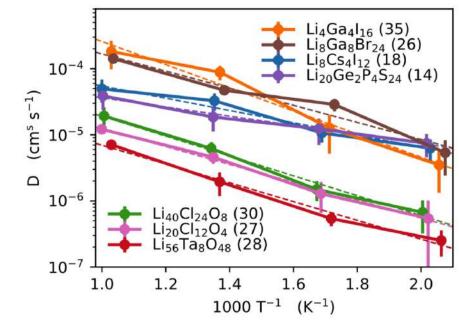
ALL SOLID-STATE BATTERIES



AUTOMATED SCREENING



Fast-ionic conductors



Arrhenius behavior of tracer diffusion (from MSD) for best Li-ion conductors

L. Kahle, A. Marcolongo, and N. Marzari, Energy & Environmental Science 13, 928 (2020)

SCIENCE IN THE CLOUD: TOWARDS A DIGITAL INFRASTRUCTURE

OPEN SCIENCE TECHNOLOGY STACK

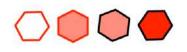
1. Widely used, **open-source community codes**

2. An **operating system** for high-throughput computational science, data proversional discrete computational science, data proversional discrete computational science and the second discrete computation of the second discrete

3. A **work environment** for non specialist where to run simulations

4. A dissemination platform for raw and CLOUD CLOUD CLOUD





Automation Sharing

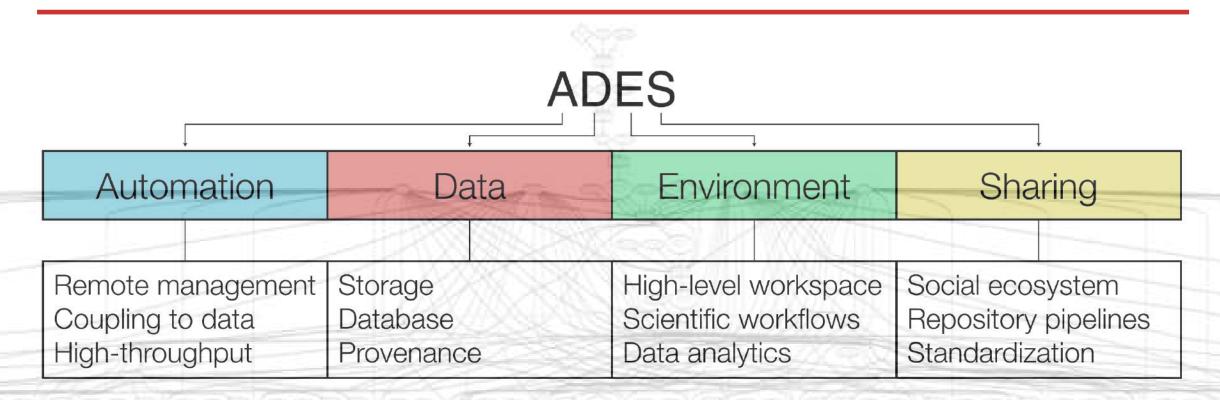
http://www.aiida.net

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

Data

Environment

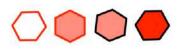
ADES MODEL FOR COMPUTATIONAL SCIENCE



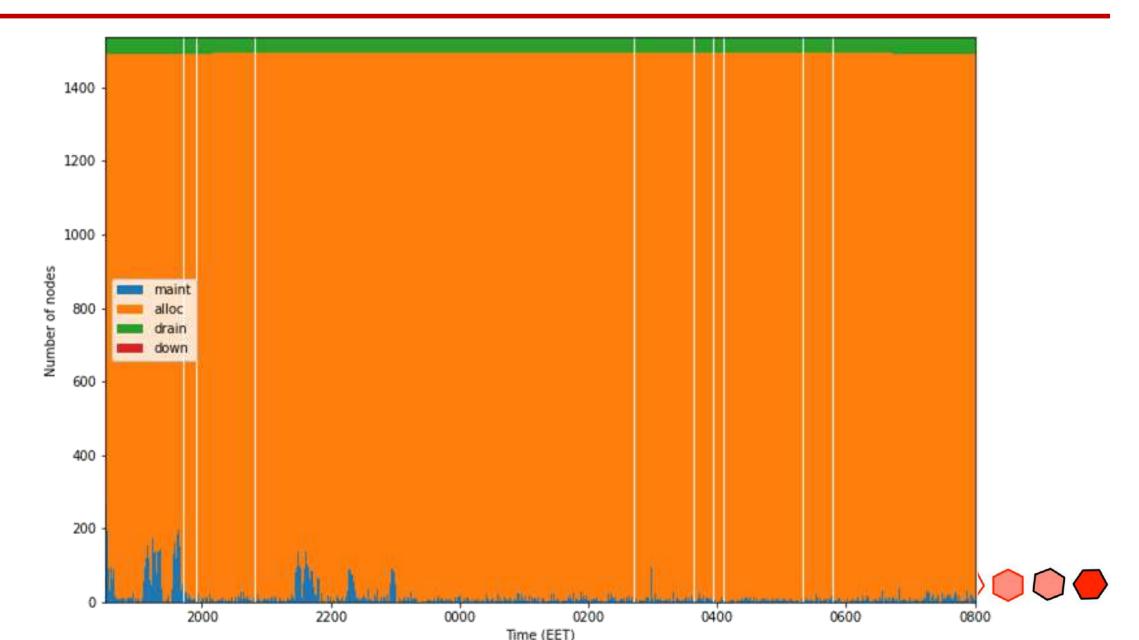
Low-level pillars

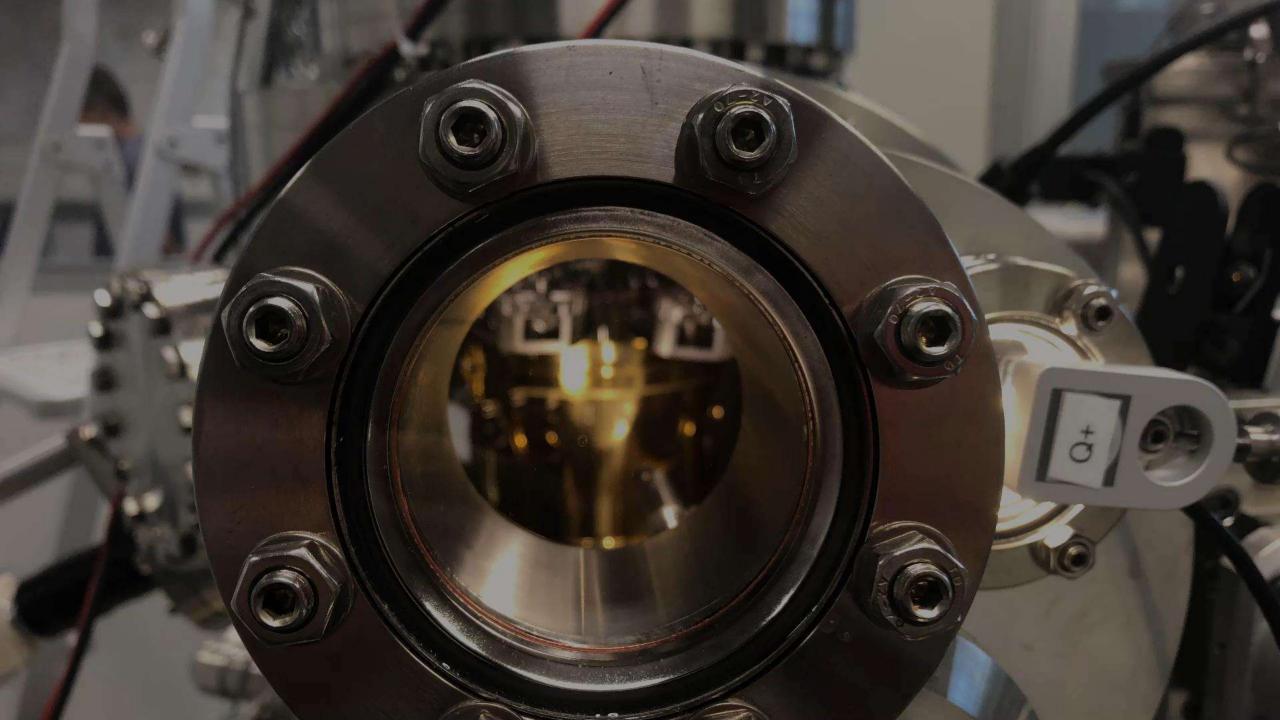
User-level pillars

S.P. Huber et al., Nature Scientific Data (2020) G. Pizzi et al., Comp. Mat. Sci. 111, 218 (2016)



LUMI-C HERO RUN





READY TO GO IN THE QUANTUM MOBILE

20.11.2a

Q. Search the docs ...

Quantum Mobile

Releases

USERS

Launching Quantum Mobile

Using Quantum Mobile

VirtualBox FAQ

Troubleshooting

DEVELOPERS

Customise Quantum Mobile

Build a Desktop VM

Build a Cloud VM

Build a Docker container

Create a new ansible role

MAINTAINERS

Developing Quantum Mobile Preparing releases

Theme by the Executable Book Project

Quantum Mobile

What is Quantum Mobile

Quantum Mobile is a Virtual Machine for computational materials science.

Quantum Mobile provides a uniform environment for quantum mechanical materials simulations. Simulation codes are set up and ready to be used either directly or through the AiiDA python framework for automated workflows and provenance tracking.

Open source throughout

Based on Ubuntu Linux

Pre-built images

4

Available for Linux, MacOS or Windows computers, using VirtualBox. Or deploy on cloud services like OpenStack or Amazon Elastic Compute Cloud using ansible.

Simulation codes pre-installed

Abinit, BigDFT, CP2K, Fleur, Quantum ESPRESSO, Siesta, Wannier90, Yambo, together with AiiDA, JupyterLab, and the AiiDAlab Jupyter environment.

Tools pre-installed

WANNIER 90

atomistic (xcrysden, jmol, cif2cell, ase, pymatgen, seekpath, spglib, pycifrw), visualization (grace, gnuplot, matplotlib, bokeh, jupyter), simulation environment (slurm, OpenMPI, FFT/BLAS/LAPACK, gcc, gfortran, singularity). Modular setup

with individually tested ansible roles. Build your own flavour tailored to your use case.



PostareSOL

L Contents

53

0

What is Quantum Mobile

Quantum Mobile Flavours Testimonials Acknowledgements

COMMON WORKFLOWS



ARTICLE OPEN

Check for updates

H₂ dissociation curve

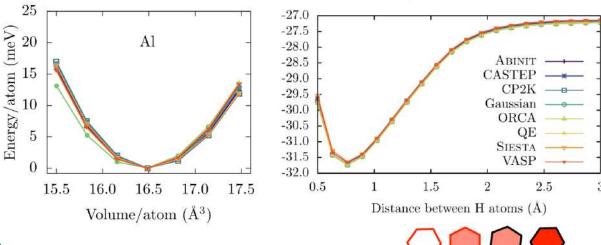
Common workflows for computing material properties using different quantum engines

Sebastiaan P. Huber ¹[∞], Emanuele Bosoni ², Marnik Bercx¹, Jens Bröder ^{3,4}, Augustin Degomme ⁵, Vladimir Dikan², Kristjan Eimre ⁶, Espen Flage-Larsen ^{7,8}, Alberto Garcia ², Luigi Genovese ⁵, Dominik Gresch⁹, Conrad Johnston ¹⁰, Guido Petretto ¹¹, Samuel Poncé¹, Gian-Marco Rignanese ¹¹, Christopher J. Sewell¹, Berend Smit ¹², Vasily Tseplyaev^{3,4}, Martin Uhrin ¹⁴, Daniel Wortmann ¹³, Aliaksandr V. Yakutovich ^{1,12}, Austin Zadoks¹, Pezhman Zarabadi-Poor ^{13,14}, Bonan Zhu ^{14,15}, Nicola Marzari ¹ and Giovanni Pizzi ¹²

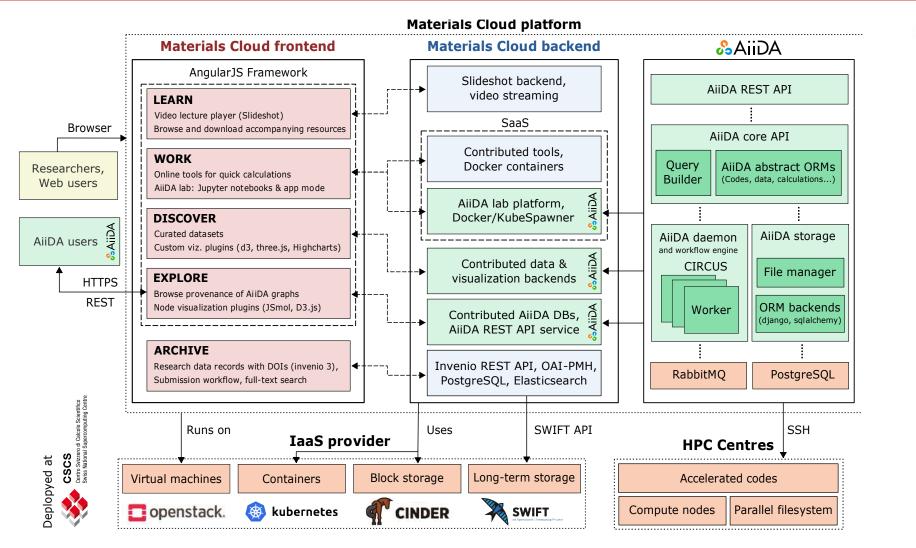
\$ aiida-common-workflows launch eos siesta --structure=Al --protocol=precise



S. Huber et al., npj Computational Materials 7, 136 (2021)



MATERIALS CLOUD



L. Talirz et al., Scientific Data 7, 299

(2020)

Indexed by

SCIENTIFIC DATA













MATERIALS CLOUD - DISCOVER

Discover curated data sets

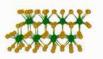
Add DISCOVER section

This section will contain a curated set of results including structures and their properties as generated by NCCR members.



Standard solid-state pseudopotentials (SSSP) DOI 10.24435/materialscloud:2018.0001/v3

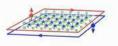
Authors: Gianluca Prandini, Antimo Marrazzo, Ivano E. Castelli, Nicolas Mounet & Nicola Marzari Description: A Standard Solid State Pseudopotentials (SSSP) library optimized for precision and efficiency.



2D structures and layered materials DOI 10.24435/materialscloud:2017.0008/v2

Authors: Nicolas Mounet, Marco Gibertini, Philippe Schwaller, Davide Campi, Andrius Merkys, Antimo Marrazzo, Thibault Sohier, Ivano E. Castelli, Andrea Cepellotti, Giovanni Pizzi & Nicola Marzari Description: Results from screening known 3D crystal structures finding those that can be computationally exfoliated, producing 2D materials candidates. If you use this work please cite N. Mounet et al, Nat. Nanotech., doi:10.1038/s41565-017-0035-5 (2018).

2D topological insulators



Authors: Antimo Marrazzo, Marco Gibertini, Davide Campi, Nicolas Mounet & Nicola Marzari Description: Results from screening exfoliable materials for 2D topological insulators (Quantum Spin Hall Insulators).

MATERIALS CLOUD - ARCHIVE



Asymmetric azide-alkyne Huisgen cycloaddition on chiral metal surfaces

DOI 10.24435/materialscloud:tx-8g

Samuel Stolz, Michael Bauer, Carlo A. Pignedoli, Nils Krane, Max Bommert, Elia Turco, Nicolo Bassi, Amogh Kinikar, Néstor Merino-Diez, Roland Hany, Harald Brune, Oliver Gröning, Roland Widmer

The record contains the data supporting our recent findings on asymmetric azide-alkyne Huisgen cycloaddition on chiral metal surfaces: Achieving fundamental understanding of enantioselective heterogeneous synthesis is maned by the permanent presence of multitudinous arrangements of catalytically active sites in real catalysts. We address this issue by using structurally comparatively simple, well-defined, and chiral intermetallic PdGa(111) surfaces as catalytic substrates. We demonstrate the impact of chirality fransfer and ensemble effect for the thermally activated azide-alkyne Huisgen cycloaddition between 3-(4-azidopheny/ipropionic acid and 9-ethyn/phenanthrene on these threefold symmetric intermetallic surfaces under ultrahigh vacuum conditions. Specifically, we encounter a dominating ensemble effect for this reaction as on the Pd3-terminated PdGa(111) surfaces no stable heterocoupled structures are created, while on the Pd1-terminated PdGa(111) surfaces, the cycloaddition ...

Latest version: v1 Publication date: Mar 02, 2021

Reversible dehalogenation in on-surface aryl-aryl coupling

DOI 10.24435/materialscloud:71-t1

Samuel Stolz, Marco Di Giovannantonio, José I. Urgel, Qiang Sun, Amogh Kinikar, Gabriela Borin Barin, Max Bommert, Roman Fasel, Roland Widmer

The record contains the data to support the findings of our recent work on reversibility of the dehalogenation process in on-surface aryl-aryl coupling. In the emerging field of on-surface synthesis, dehalogenative aryl-aryl coupling is unarguably the most prominent tool for the fabrication of covalently bonded carbon-based nanomaterials. Despite its importance, the reaction kinetics are still poorly understood. Here we present a comprehensive temperature-programmed x-ray photoelectron spectroscopy investigation of reaction kinetics and energetics in the prototypical on-surface dehalogenative polymerization of 4,4''-dibromo-p-terphenyl into poly(para-phenylene) on two coinage metal surfaces, Cu(111) and Au(111). We find clear evidence for reversible dehalogenation on Au(111), which is inhibited on Cu(111) oving to the formation of organometallic intermediates. The incorporation of reversible dehalogenation in the reaction rate equations leads to excellent agreement with ...

Latest version: v1 Publication date: Mar 02, 2021

Extensive benchmarking of DFT+U calculations for predicting band gaps

DOI 10.24435/materialscloud;jx-fp

Nicole Kirchner-Hall, Wayne Zhao, Yihuang Xiong, Iurii Timrov, Ismaila Dabo

Accurate computational predictions of band gaps are of practical importance to the modeling and development of semiconductor technologies, such as (opto)electrohic devices and photoelectrochemical ceils. Among available electronic-structure methods, density-functional theory (DFT) with the Hubbard U correction (DFT+U) applied to band edge states is a computationally tractable approach to improve the accuracy of band gap predictions based on (semi)local functionals. At variance with DFT approximations, which are not intended to describe optical band objective-state properties, DFT+U can be interpreted as approximations, which are not intended to describe optical band gaps and other excited-state properties, DFT+U can be interpreted as approximations, which are not intended when U is determined by intercent of the total energy with respect to electronic occupations in the Hubbard manifold (thus removing self-interaction errors in this subspace), thereby providing a (heurity) justication or sing DFT+U or predict band gaps. However, it is ...

Latest version: v1 Publication date: Mar 02, 2021

Building a consistent and reproducible database for adsorption evaluation in Covalent-Organic Frameworks 🕁 🖧

DOI 10.24435/materialscloud:5q-jt

Daniele Ongari, Aliaksandr V. Yakutovich, Leopold Talirz, Berend Smit

We present a workflow that traces the path from the bulk structure of a crystalline material to assessing its performance in carbon capture from coal's postcombustion flue gases. This workflow is applied to a database of 324 covalent-organic frameworks (COFs) reported in the literature, to characterize their CO2 adsorption properties using the following steps: (1) optimization of the crystal structure of a crystalline material to assessing its performance in carbon capture from coal's postcombustion flue gases. This workflow is applied to a database of 324 covalent-organic frameworks (COFs) reportation to the interature, to characterize their CO2 adsorption properties using the following steps: (1) optimization of the crystal structure (atomic positions and unit cell) using density functional theory, (2) fitting atomic point charges based on the electron density of the structures before and after optimization, (4) computing atomic topication at the interaction properties using grand canonical Monte Carlo simulations with an empirical interaction potential, and finally, (5) assessing the CO2 parabitic energy via process modeling. The full workflow has been encoded in the Automated Interactive Infrastructure and Database for Computational Science (AiiDA). Both the workflow and the ...

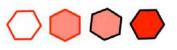
Latest version: v8 Publication date: Feb 24, 2021

LET'S BROADEN THE HORIZON



THE BEZOS MANDATE: EXTERNALIZABLE SERVICE INTERFACES

- 1) All teams will henceforth expose their data and functionality through service interfaces.
- 2) Teams must communicate with each other through these interfaces.
- 3) There will be no other form of interprocess communication allowed.
- 4) All service interfaces, without exception, must be designed from the ground up to be externalizable.



IN ACTION: OPTIMADE UNIVERSAL REST API



About us

Designing new materials suitable for specific applications is a long, complex, and costly process. Researchers think of new ideas based on intuition and experience. Their synthesis and evaluation require a tremendous amount of trial and error. In the last few years, there has been a major game change in materials design. Thanks to the exponential growth of computer power and the development of robust first-principles electronic structure codes, it has become possible to perform large sets of calculations automatically. This is the burgeoning area of high-throughput ab initio computation. Such calculations have been used to create large databases containing the calculated properties of existing and hypothetical materials, many of which have appeared online:

- the AFLOW distributed materials property repository
- the Harvard Clean Energy Project Database
- the Materials Cloud
- . the Materials Project
- the NoMaD (Novel Materials Discovery) Repository
- · the Open Quantum Materials Database
- · the Computational Materials Repository
- · the Data Catalyst Genome
- the Materials Platform for Data Science
- the Joint Automated Repository for Various Integrated Simulations

• ...

The Open Databases Integration for Materials Design (OPTIMADE) consortium aims to make materials databases interoperational by developing a common REST API.

Help improve the application: # Report a bug * Suggest a feature/change

This is a friendly client to search through databases and other implementations exposing an OPTIMADE RESTful API. To get more information about the OPTIMADE API, please see the official web page. All providers are retrieved from the OPTIMADE consortium's list of providers.

Note: The structure property assemblies is currently not supported. Follow the issue on GitHub to learn more.

► FAQ

Client version: 2021.2.23.1 Source code: GitHub

+ Log

Query a provider's database

Selec	t a provider	~															
No provider chosen 🗸 🗸					~												
< C Showing 0 of 0 results > >																	
pply	filters																
Basic Raw																	
Chemistry Chemical Formula e.g. [FEOI2 Na Elements						<u>∆</u> Hid	le Peric	idic Tal	ble								
		Stru	cture	es can	include	any c	hose	en ele	ments	(inste	ad of all	0					
		н															He
Li Be										в	С	N	0	F	Ne		

IN ACTION: OPTIMADE UNIVERSAL RESTAPI

OPTIMADE Open Databases Integration for Materials Design	Apply filters Basic Raw Chemistry Chemical Formula Elements Structures can include any chosen	Results		0 0.9288990900 0.1181459098 0.9953740099 0 0.3026540000 0.4782289090 0.381559090 0 0.4782280000 0.3816550000 0 0.1973360000 0.371919000 0 0.9237720000 0.3244359900 0.9935260000 0 0.9035260000 0.3244359900 0.324436000 0 0.9035260000 0.324136000 0.324436000 0 0.5953740000 0.3200990000 0.1181450000 0 0.5953740000 0.320090000 0.1197366000 0 0.8318550000 0.9782280000 0.8026640000 K_POINTS automatic 4 4 0
Currently valid OPTIMADE API version: 92.9.1 Client version: 2021,3.29 Source code: 00450 Help improve the application: IF Reports burg This is a friendly client to search through database and other im; get more information about the OPTIMADE API, please see the o OPTIMADE consortiant's fist of provides.	H Li Be Na Mg K Ca Sc TI V Cr Mn Fe		 x Sort Quantum ESPF About the Quantum ESPRESSO in 	CELL_PARAMETERS angstrom -5.9308880000 5.9308880000 5.9308880000 -5.9308880000 5.9308880000 5.9308880000 5.9308880000 5.9308880000
Note: The structure property assessibilities is currently not support	Rb Sr Y Zr Nb Mo Tc Ru Cs Ba · Ht Ta W Re Os Fr Ra # Rt Db Sg Bh Hs		Instructions Acknowledgements The crystal structure has been su Adapt the parameters below and	Drag to rotate, scroll to zoom, right-click for other
Select a provider S database Select a provider S database Select a provider Apply filters Basic Raw Chemistry Chemistry Chemistry Estimate Select a control	La Ce Pr Nd Pn # Ac Th Pa U Np Number of Elements		Select here the pseudopotential lib Select here the magnetism/smean Select here the k-points distance	
Structures can include any chosen e H LL Be Na Mg K Ca Sc Ti V Cr Mn Fe	Dimensionality Molecule Wire Planar Bu Number of Sites Provider specific Provider ID NBI Will take precedence	Crystallographic Information File v1.0 (.cif) V Use in QE Input Generator	(and smearing (eV) in case of frac	
Rb Sr V Zr Nb Mo Ta Ru Cs Ba * HI Ta W Fa Os Fr Ra # RI Db Sg Bh Ha	Q, Search Results Ascending id Co8Ge12LH2O48 (id=mp-1013807)	ructure details Sites		Supercell: 2 0 2 0 2 0 LPDARE RESET 2x2x2 CEL

 $\bigcirc \cup \cup$

QUANTUM-AS-A-SERVICE – AiiDAlab deployment

🖉 🖉 🥥 🧧 .start-0 - Jupyter Notebook 🛛 🗙 🖉 aiida	@b8f2bf3a041a: ~ × +											0
$\langle \cdot \rangle \rightarrow C \land = ab63397873ac.eu.ngrok.io/apps//$								Ⅲ ☆		Tp (5 *	-
	💭 Jupyter					Edit App	Logout				5052 VBA	
	Executing 'aiidalab' in DI	EVELOP mode .	Tasks	App Store	? Help							
	title) metadatā.json file is not pr	resent		ESPRESSO	S Manage App	Modified			k			
A. V. Yakutovich et al., Comp. Mat. Sci. 188, 110165 (2021)	OPTIMADE Client		OPTIMAI Open Datat for Material	bases Integration	Manage App	IUpdate available						

QUANTUM-AS-A-SERVICE – AiiDAlab deployment

🖉 🖉 🖉 .start-0 - Jupyter Notebook 🗙 📐 aiida@	@b8f2bf3a041a:~ × +												0
← → C △	apps/home/start.ipynb							Ⅲ ☆		Tp 🔘	🚺 Se	* 0	÷
	📁 jupyter					Edit App	Logout						
	Executing 'aiidalab' in DI Image: File Manager	EVELOP mode.	Tasks	App Store	? Help								
	✓ (title) metadata.json file is not pr	esent	Gurntum	ESPRESSO	S Modif	fied			×				
	✓ OPTIMADE Client		OPTIMA Open Datat for Material	DE bases Integration is Design	O Upda Manage App URL	te available							
	- AliDAlah Rasa Widgata					2							

CONCLUSIONS

1. Materials enable the technologies that power our economy, our lives and our society

2. We can discover novel materials with a speed that mirrors ICT technologies, rather than any physical infrastructure

3. We can redistribute simulations tools, data, and services at will and to the entire world

NATURE MATERIALS | VOL 20 | JUNE 2021 | 736-749 | www.nature.com/naturematerials

REVIEW ARTICLE | INSIGHT

nature materials

Check for updates

Electronic-structure methods for materials design

Nicola Marzari¹¹², Andrea Ferretti² and Chris Wolverton³

The accuracy and efficiency of electronic-structure methods to understand, predict and design the properties of materials has driven a new paradigm in research. Simulations can greatly accelerate the identification, characterization and optimization of materials, with this acceleration driven by continuous progress in theory, algorithms and hardware, and by adaptation of concepts and tools from computer science. Nevertheless, the capability to identify and characterize materials relies on the predictive accuracy of the underlying physical descriptions, and on the ability to capture the complexity of realistic systems. We provide here an overview of electronic-structure methods, of their application to the prediction of materials properties, and of the different strategies employed towards the broader goals of materials design and discovery.

ACKNOWLEDGEMENTS





Francisco F.

Ramirez

(EPFL)

Adorf

(EPFL)



Aliaksandr

Yakutovich

(EPFL)

Flaviano dos Santos (EPFL)

Andersen (EPFL)



Chris

Sewell

(EPFL)

Valeria

Granata (EPFL)



Smit

(EPFL)

Elsa

Passaro

(EPFL)

Thomas Schulthess (ETHZ,CSCS)

Marnik

Bercx

(EPFL)

http://www.aiida.net http://www.materialscloud.org

Contributors for the 40+ plugins: Quantum ESPRESSO, Wannier90, CP2K, FLEUR, YAMBO, SIESTA, VASP, CASTEP, CRYSTAL, ...

Leopold

Talirz

(EPFL)

Contributors to aiida-core and former AiiDA team members — Oscar Arbelaez, Michael Atambo, Valentin Bersier, Marco Borelli, Jocelyn Boullier, Jens Bröder, Ivano E. Castelli, Andrea Cepellotti, Keija Cui, Vladimir Dikan, Marco Dorigo, Y.-W. Fang, Fernando Gargiulo, Marco Gibertini, Davide Grassano, Dominik Gresch, Conrad Johnston, Rico Häuselmann, Daniel Hollas, Eric Hontz, Jianxing Huang, Christoph Koch, Espen Flage-Larsen, Ian Lee, Daniel Marchand, Antimo Marrazzo, Andrius Merkys, Simon Pintarelli, Nicolas Mounet, Tiziano Müller, Gianluca Prandini, Philip Rüßmann, Riccardo Sabatini, Ole Schütt, Phillippe Schwaller, Andreas Stamminger, Atsushi Togo, Daniele Tomerini, Nicola Varini, Martin Uhrin, Jason Yu, Austin Zadoks, Bonan Zhu, Mario Zic, Spyros Zoupanos







(EPFL)





VandeVondele

(ETHZ,CSCS)

ACKNOWLEDGEMENTS



http://nccr-marvel.ch

Swiss National Centre for Computational Design and Discovery of Novel Materials (2014-18, 2018-22, 2022-26)

http://max-centre.eu

DRIVING THE EXASCALE TRANSITION

H2020 Centre of Excellence MaX: Materials Design at the Exascale (2015-18, 2018-21)



https://www.big-map.eu

H2020 Battery Interface Genome – Materials Acceleration Platform (Battery 2030+) (2020-23)

Related projects: H2020 Nanoscience Foundries and Fine Analysis H2020 European Materials Modelling Council H2020 Marketplace EPEL H2020 Intersect H2020 DOME 4.0 H2020 OpenModel H2020 NEP SOLVAY H2020 EPFL Fellows H2020 EPFL Innovators H2020 Marie Curie MarketPlace FL Open Science HEMONT PASC PRACE ENSINE SWISS NATIONAL SCIENCE FOUNDATION IBM Constellium SAMSUNG Innosuisse **---**Solvay for Advanced Scientific Computin Samsung Richemont Varinor PRACE * EIMINC #



"Things were done very differently on the farm when I was your age, Kenny."