

JARVIS

Computational Discovery of Two-Dimensional Materials, Evaluation of Force-Fields and Machine Learning

By

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Northwestern University, May 19, 2017

NIST Materials Genome Initiative
Gateway to Materials Genome Information

Acknowledgments



NIST



NIST



NIST/LBNL



NIST



NIST



NIST



NIST



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NIST



GMU



NIST



NIST



NIST



Google



NIST

Others : Richard Hennig (UF), Susan Sinnott (PSU), Tao Liang (PSU), Kiran Mathew (LBNL), Kristin Persson (LBNL)

Computational resources: CTCMS, RARITAN, XSEDE, NERSC

Software packages: VASP, LAMMPS, Pymatgen, ASE, Phonopy, Magpie, Matminer, Flask

Outline

1. JARVIS-FF

- Evaluation of elastic constants, convex hull
- Defects, surfaces, phonons

2. JARVIS-DFT:

- 2D or not 2D?
- Properties

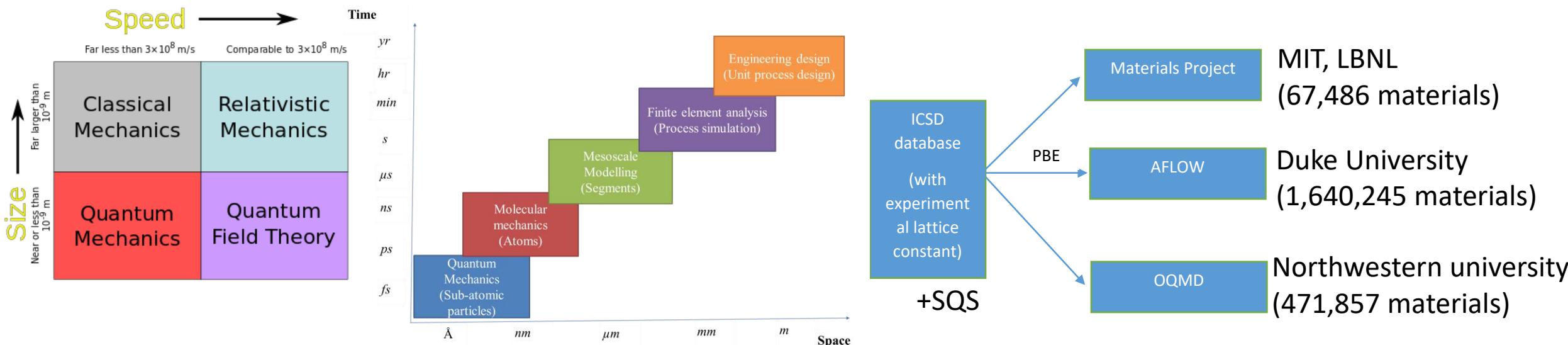
3. JARVIS-ML:

- Generalized Material-descriptors/features (challenge?)
- Neural-Network+classical physics based hybrid force-field

4. Conclusions

Few basics

- Materials Science is all about: Structure-Property-Performance relationship and minimization of free-energy
- Computationally: Structure = lattice constants (a, b, c), angles (alpha, beta, gamma) and basis vectors ([Si,0,0,0],[Si,0.5,0.0,0.5],....)
- To calculate property: classical physics (e.g. classical force-fields), quantum physics (e.g. density functional theory)
- MGI motivated current computational databases: Materials-project (MP), AFLOW, OQMD



Others: AIIDA, MaterialWeb, NREL-MatDB etc.

JARVIS-FF

1A Introduction to force-fields

1B Property evaluation: elastic constants, convex hull

1C Defects, surfaces, phonons

1A Intro Force-fields



- Coulomb potential $V(q_1(\vec{r}_1), q_2(\vec{r}_2)) = \frac{kq_1q_2}{|\vec{r}_2 - \vec{r}_1|}$ One parameter to fit/optimize
- Lennard-Jones potential $V_{LJ} = 4\epsilon \left[\left(\frac{\sigma}{r} \right)^{12} - \left(\frac{\sigma}{r} \right)^6 \right]$ Repulsive and attractive terms, Two parameters to fit
- Morse-potential $V(r) = D_e(e^{-2a(r-r_e)} - 2e^{-a(r-r_e)})$

These potentials lack angular information hence not able to capture elastic constants well

- Stilinger-Weber potential $V_{TOT} = \sum_{i,j}^N V_2(r_{ij}) + \sum_{i,j,k}^N V_3(r_{ij}, r_{ik}, \theta_{ijk})$ Uses angle but transferability problem
- EAM and MEAM potential $V_{TOT} = \sum_i^N F_i \left(\sum_j \rho(r_{ij}) \right) + \frac{1}{2} \sum_{i,j}^N V_2(r_{ij})$ Uses electron density, for metallic systems
- Bond-order/Tersoff/Brenner $V_{ij}(r_{ij}) = V_{repulsive}(r_{ij}) + b_{ijk}V_{attractive}(r_{ij})$ Uses bond information, for covalent systems

These potentials lack charge information

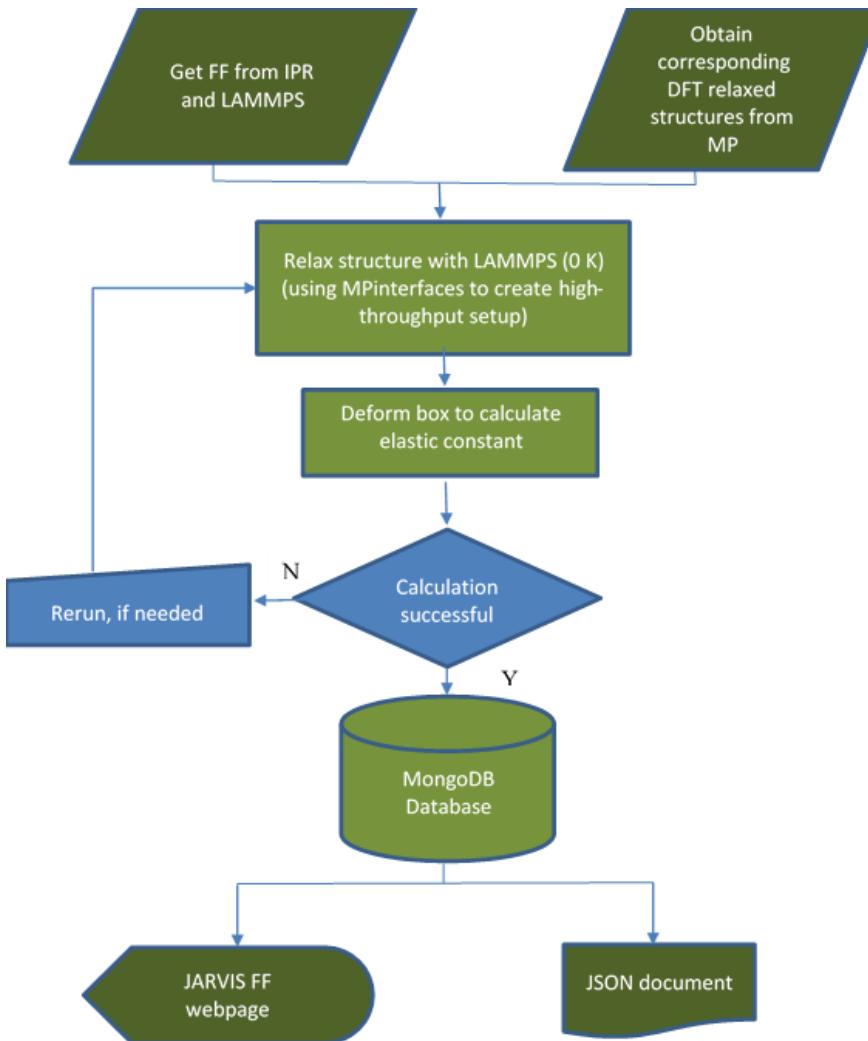
- Fixed charge potentials: Coulomb-Buckingham
- Other FFs: ReaxFF, COMB, AMBER, CHARMM, OPLS etc.

$$F_i = m_i a_i, \quad F_i = -\nabla_i V, \quad -\frac{dV}{dr_i} = m_i \frac{d^2 r_i}{dt^2}$$



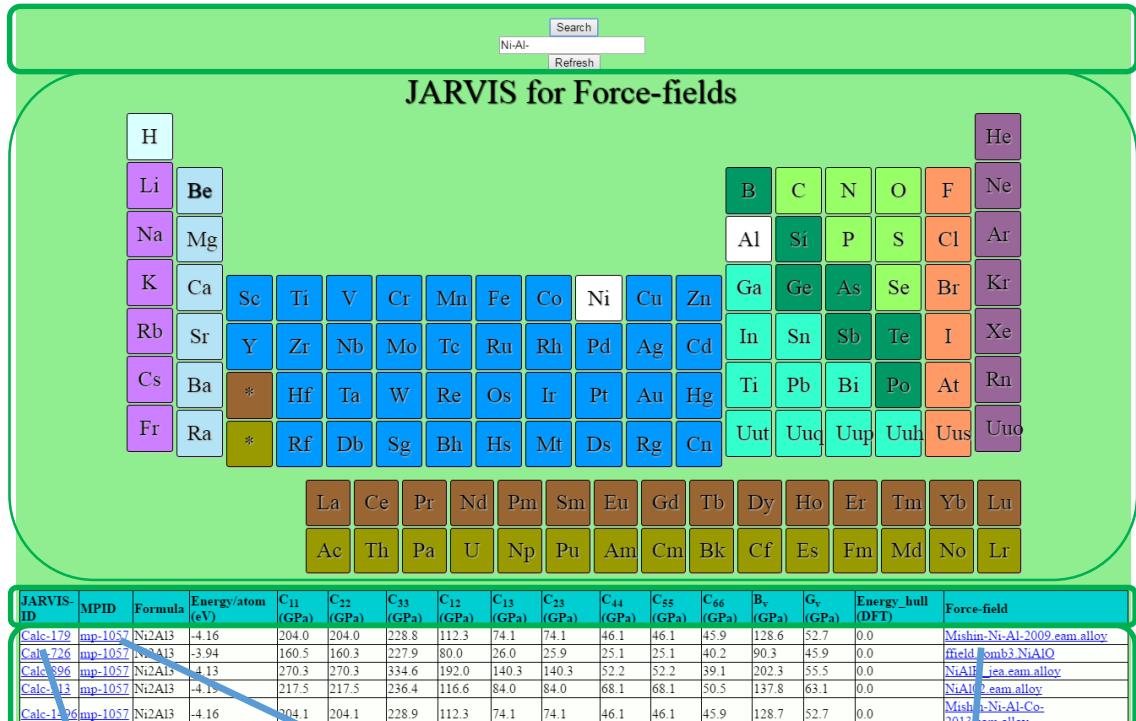
Berni Alder

JARVIS-FF-workflow



- Force-fields obtained from NIST's interatomic potential repository and LAMMPS-potential folder (e.g. Mishin-Ni-Al-2009.alloy)
- DFT relaxed structures obtained from materials-project for corresponding system (say Ni-Al system: Al, Ni, Ni₃Al, NiAl etc.)
- Convert the files from DFT format to LAMMPS format write corresponding input files from LAMMPS
- HT LAMMPS calculations

Public Webpage



a)

<http://www.ctcms.nist.gov/~knc6/periodic.html>
<http://www.ctcms.nist.gov/~knc6/JLAMMPS.html>

b)

c)

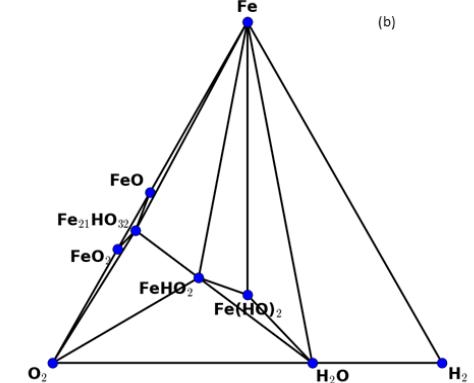
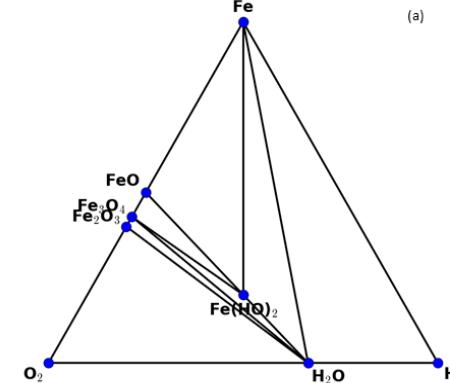
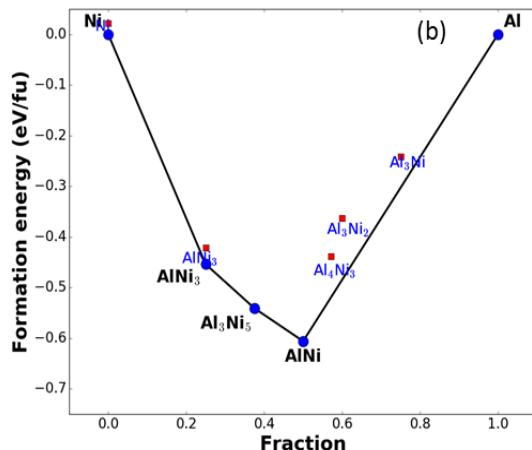
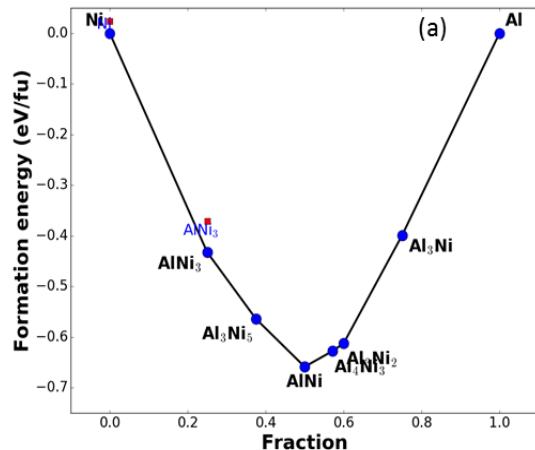
d)

JARVIS-ID	Formula	Spacegroup	Energy/atom (eV)	C ₁₁ (GPa)	C ₁₂ (GPa)	C ₁₃ (GPa)	C ₄₄ (GPa)	B _v (GPa)	G _v (GPa)	Force-field
JLMP-165	AlNi3	Pm-3m	-4.631478	238.2	166.4	166.4	130.2	190.3	92.5	Mishin-Ni-Al-2009.eam.alloy
JLMP-166	AlNi	Pm-3m	-4.510871	190.9	142.9	142.9	121.5	158.9	82.5	Mishin-Ni-Al-2009.eam.alloy
JLMP-167	Al ₃ Ni ₅	Cmmm	-4.582119875	188.1	178.5	179.7	86.7	201.0	95.0	Mishin-Ni-Al-2009.eam.alloy
JLMP-168	Al ₃ Ni	Pnma	-3.87520775	208.5	105.8	93.0	43.1	119.3	36.6	Mishin-Ni-Al-2009.eam.alloy
JLMP-169	Al ₄ Ni ₃	Ia-3d	-4.26581160714	187.6	117.8	117.8	77.4	141.1	60.4	Mishin-Ni-Al-2009.eam.alloy
JLMP-170	AlNi3	Fm-3m	-4.59878825	197.3	180.7	180.7	138.9	186.2	86.7	Mishin-Ni-Al-2009.eam.alloy
JLMP-171	Al ₃ Ni ₂	P-3m1	-4.1589292	204.0	112.3	74.1	46.1	128.6	52.7	Mishin-Ni-Al-2009.eam.alloy
JLMP-179	AlNi3	Pm-3m	-4.631478	238.2	166.4	166.4	130.2	190.3	92.5	Mishin-Ni-Al-Co-2013.eam.alloy
JLMP-180	AlNi	Pm-3m	-4.510871	190.8	142.9	142.9	121.5	158.9	82.5	Mishin-Ni-Al-Co-2013.eam.alloy
JLMP-181	Al ₃ Ni ₅	Cmmm	-4.582119875	188.1	178.5	179.8	86.7	201.1	95.0	Mishin-Ni-Al-Co-2013.eam.alloy
JLMP-182	Al ₃ Ni	Pnma	-3.87520775	208.5	105.8	93.0	43.1	119.3	36.6	Mishin-Ni-Al-Co-2013.eam.alloy
JLMP-183	Al ₄ Ni ₃	Ia-3d	-4.26581160714	187.6	117.8	117.8	77.4	141.1	60.4	Mishin-Ni-Al-Co-2013.eam.alloy
JLMP-184	AlNi3	Fm-3m	-4.59878825	197.3	180.7	180.7	138.9	186.2	86.7	Mishin-Ni-Al-Co-2013.eam.alloy
JLMP-185	Al ₃ Ni ₂	P-3m1	-4.1589292	204.1	112.3	74.1	46.1	128.7	52.7	Mishin-Ni-Al-Co-2013.eam.alloy
JLMP-209	AlNi3	Pm-3m	-4.631478	238.2	166.4	166.4	130.2	190.3	92.5	Mishin_updated-Ni-Al-Co-2013.eam.alloy
JLMP-210	AlNi	Pm-3m	-4.510871	190.8	142.9	142.9	121.5	158.9	82.5	Mishin_updated-Ni-Al-Co-2013.eam.alloy

Old version

Development version

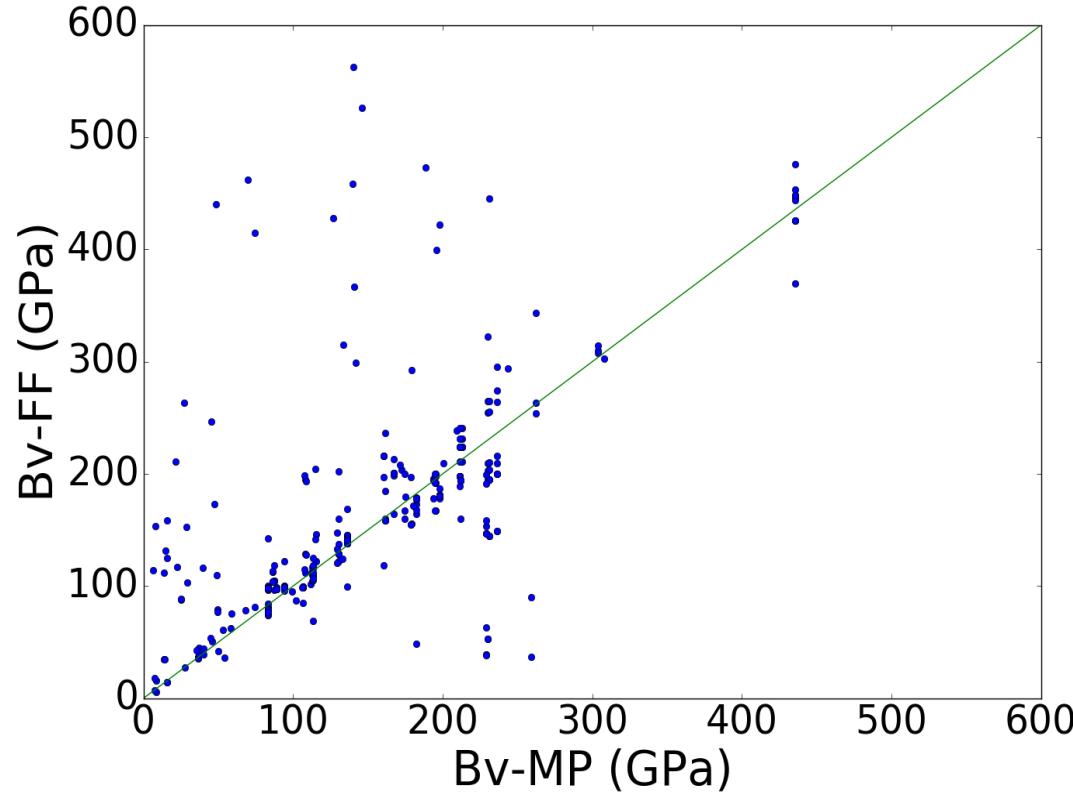
1B Convex-hull plot



- Comparison of DFT convex-hull plot with LAMMPS
- Some points shown unstable for FF, because they are not fit to it !

Choudhary et al., Nature:Scientific Data 4, 160125 (2017)
Choudhary et al., Computational Materials Science 113, 80 (2016)

1B Bulk modulus



- Comparison of DFT bulk modulus with LAMMPS, ~500 points in common
- Some FFs are not fit for elastic properties

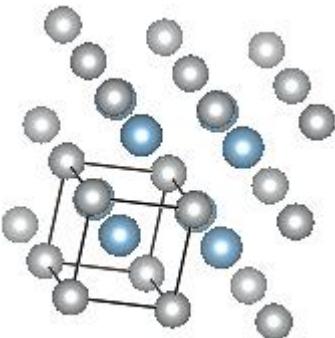
Choudhary et al, Nature:Scientific Data 4, 160125 (2017)

1C Vacancy formation energies

Vacancy-formation energy (eV)

Vacancy formation energies were calculated by deleting the symmetrically distinct atoms in the system. In the table, vacancy forming element, its multiplicity, and defect-formation energy are given. The reference element cohesive energies were calculated with the most stable structure for the element found on materials project database. The defect structures were generated with [Pydii](#) at 0 K. Defect structures were constructed with the fully-relaxed bulk system as input. For defect-structures energetics calculations, constant volume ensemble was used. We impose the defect structures to be at least 1.5 nm large in all directions.

Element	Mult.	Value
Al	1	2.027
Ni	1	2.845

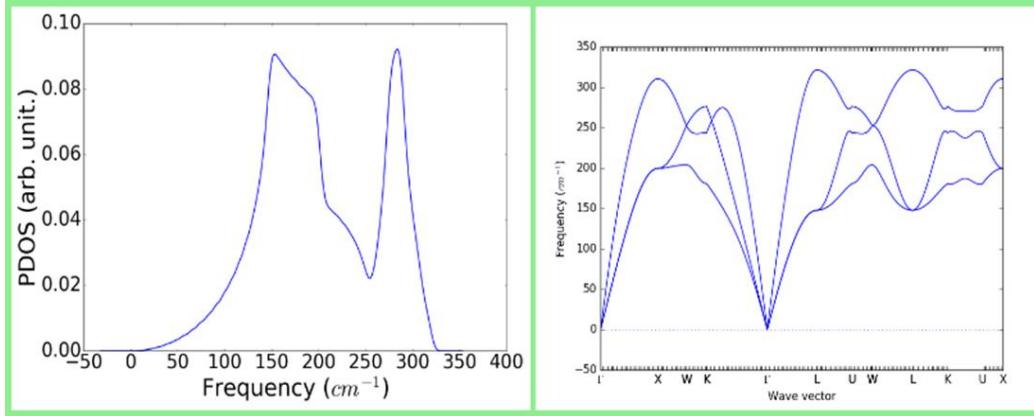


- Real materials are **VERY** defective !
- Vacancy structures based on unique Wyckoff positions
- Delete symmetrically distinct sites only

1C Phonons

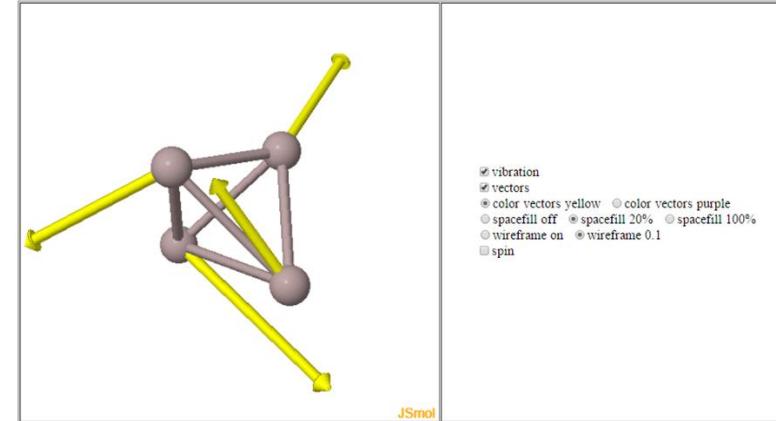
Phonon

Phonons were obtained by making an interface of JARVIS-FF with [Phonopy package](#) at 0 K. For deformed-structures, constant volume ensemble was used. The deformed structures were taken of at least 1.5 nm size in all directions. The band-indices for phonon bandstructure was obtained with [PyMatgen](#). The phonon representation were obtained with phonopy. The code for their Infrared (I) or Raman (R) activity is not directly implemented in Phonopy. The I, R designations were implemented in JARVIS and will be released in public domain soon.



Reference

mp-134



$$\Phi = \Phi_0 + \sum_{lk} \Phi_\alpha(lk) u_\alpha(lk) + \frac{1}{2} \sum_{ll'kk'} \Phi_{\alpha\beta}(lk, l'k') u_\alpha(lk) u_\beta(l'k') + \dots$$

$$\Phi_{\alpha\beta}(lk, l'k') = - \frac{\partial F_\beta(l'k')}{\partial u_\alpha(lk)} = \frac{\partial^2 \Phi}{\partial u_\alpha(lk) \partial u_\beta(l'k')}.$$

$$D_{\alpha\beta}^{kk'}(\mathbf{q}) = \frac{1}{\sqrt{M_k M_{k'}}} \sum_{l'} \Phi_{\alpha\beta}(0k, l'k') e^{i\mathbf{q}[\mathbf{r}(l'k') - \mathbf{r}(0k)]}$$

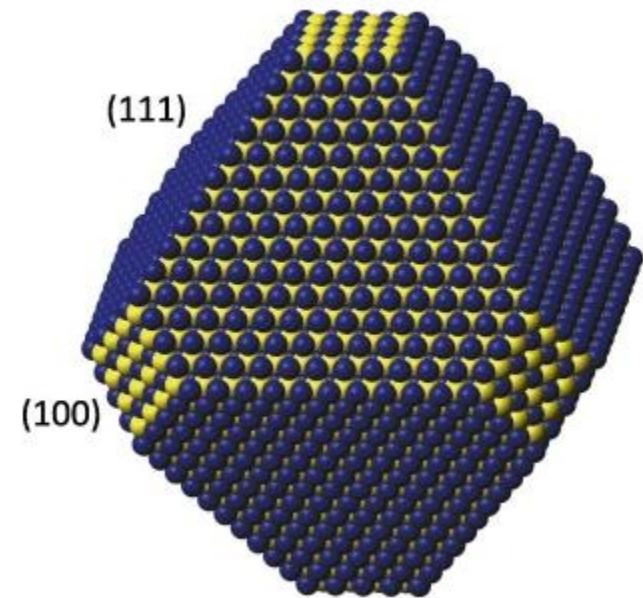
$$\sum_{\beta k'} D_{\alpha\beta}^{kk'}(\mathbf{q}) e_{\mathbf{q}j}^{\beta k'} = \omega_{\mathbf{q}j}^2 e_{\mathbf{q}j}^{\alpha k}$$

1C Surface energies

Surface energy (J/m²)

Surface energies were calculated for symmetrically distinct crystal surfaces obtained with [Pymatgen](#). In the table, (hkl) indices and surface energies are given. For surface-structure energetics, constant volume ensemble was used. We impose the slab thickness to be at least 2 nm and vacuum size of 2.5 nm. The maximum miller index is taken as 3.

Surface Value
(1 1 1) 0.875
(1 0 0) 0.949
(3 3 2) 0.952
(3 2 2) 0.977
(2 1 1) 0.999
(3 3 1) 1.006
(1 1 0) 1.013
(3 1 1) 1.019
(3 2 1) 1.033
(3 1 0) 1.042
(3 2 0) 1.05
(2 1 0) 1.053
(2 2 1) 1.058



Wulff construction: Used in predicting equilibrium shape of materials

Mathew et al., Computational Materials Science, 122 183 (2016)

Summary Part-1

- New website for FF calculation data
- Enable user to select FF before doing any calculations
- Can be complement to DFT where computational cost is too high, such as phonons, defects, grain boundaries etc.
- Genetic algorithm based evaluation, melting point, thermal expansion coefficient, thermal conductivity, Grüneisen parameter calculations on the way
- Outreach: thousands of website hits in few months, JARVIS-FF listed as a post-processing tools for LAMMPS

PART 2 JARVIS-DFT

2A Introduction to Density functional theory

2B 2D Materials

- Applications: Sub-micron electronics, Flexible and tunable electronics, photo-voltaics, energy-storage

2A Density Functional Theory

- Classical Newton's laws not applicable for electrons (**very fast, very tiny**);
- Schrödinger equation: mathematical equation that describes the evolution over time of a physical system in which quantum effects, such as **wave–particle duality**, are significant (such as electrons)
- Schrödinger equation of a **fictitious system** (the "Kohn–Sham system") of **non-interacting particles** (typically electrons) that generate the **same density** as any given system of interacting particles
- Uses **density** (scalar) vs **wavefunction** (vector) quantity

$$H\psi = E\psi \quad \left[-\frac{\hbar^2}{2m} \nabla^2 + V_{\text{eff}}(r) \right] \psi_i(r) = E_i(r) \psi_i(r) \quad V_{\text{eff}} = T + V_{Ne} + V_{ee} + V_{XC}$$

- **Different functionals:** LDA, GGA (PBE, PW91), HSE06 etc.
- **LDA:** developed for homogenous systems (Thomas-Fermi model), such as metals
- **GGA:** uses density gradient information on top of GGA
- **vdW-DF2:** uses exchange from GGA, correlation from LDA, quantum-Monte-Carlo based non-local corrections, **optB88** used here mainly

Second derivative wrt X

$$\frac{\partial^2 \psi}{\partial x^2} + \frac{8\pi^2 m}{h^2} (E - V) \psi = 0$$

Position Energy Potential Energy

The diagram shows the Schrödinger equation $\frac{\partial^2 \psi}{\partial x^2} + \frac{8\pi^2 m}{h^2} (E - V) \psi = 0$. Blue arrows point from the labels to the corresponding terms: 'Second derivative wrt X' points to $\frac{\partial^2 \psi}{\partial x^2}$; 'Schrodinger Wave-function' points to ψ ; 'Position Energy' points to E ; and 'Potential Energy' points to V .

Exchange-correlation



Walter Kohn

Hohenberg, Pierre; Walter Kohn (1964). "Inhomogeneous electron gas". *Physical Review*. **136** (3B): B864–B871

2B Two-dimensional materials

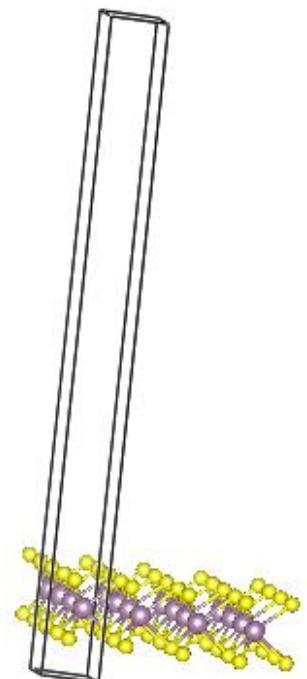
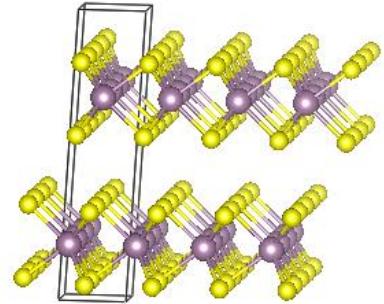
- Vander-Waal bonding in z-direction, covalent bonding in x, y-directions, e.g. Graphene, MoS₂

DIMENSIONALITY == PERIODICITY

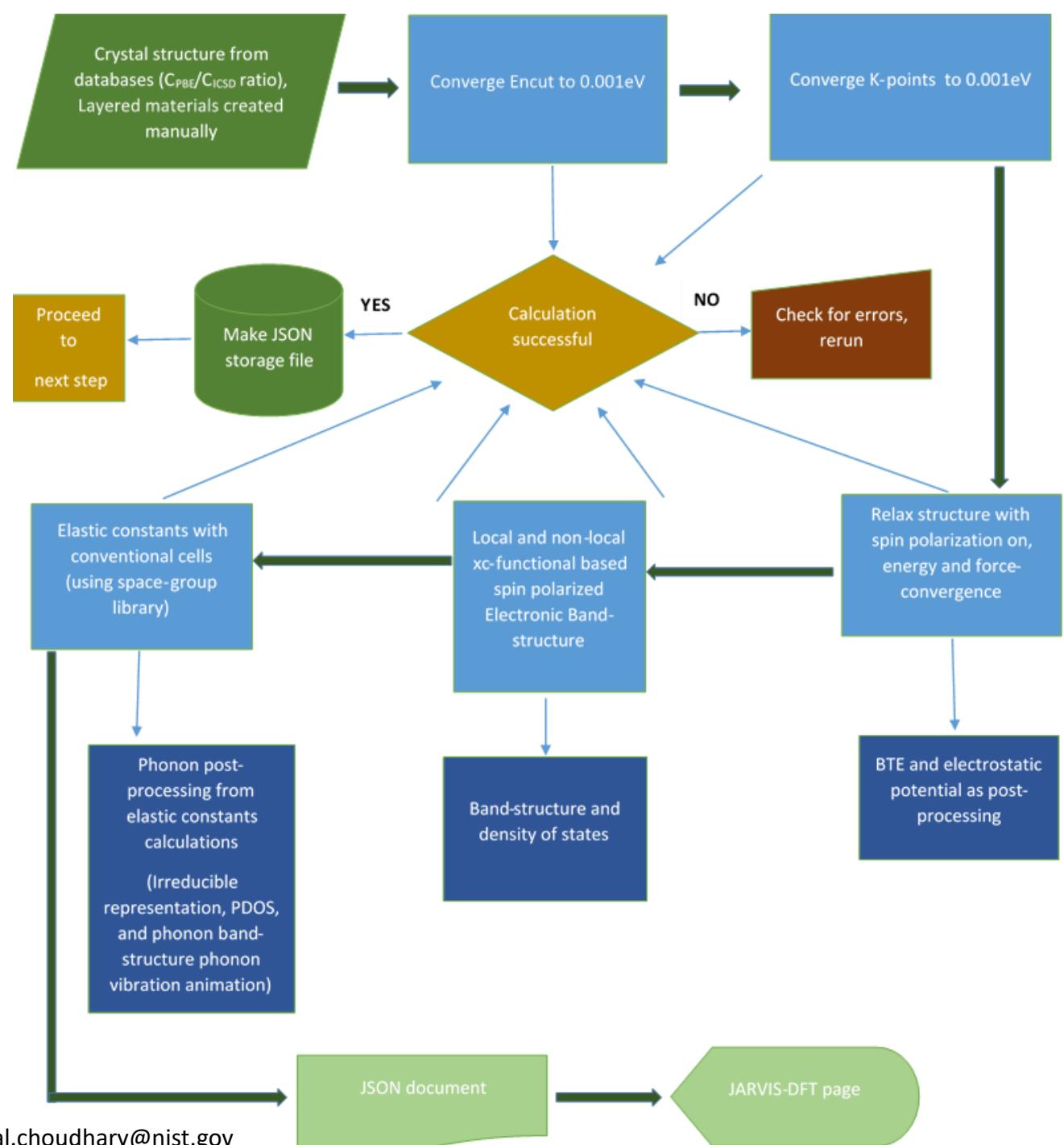
- Easily exfoliated, scotch-tape type, potential candidate for sub-nanometer technology, Si-replacement, flexible electronics, and other applications
- **For DFT calculation on 2D materials:** position of atoms (r) obtained from XRD and other experiments, ICSD database and other DFT databases
- DFT databases (**Materials Project, AFLOW, OQMD**) took structures from ICSD and used PBE functionals consistently for all structures, JARVIS-DFT took from them
- **Catch:** PBE functionals overestimate lattice constants (the r's in PBE-DFT are not correct)
- **JARVIS-DFT:** Using REST API at Materials-project got all the crystal structures with ICSD and PBE data and calculated relative error

$$\delta = \frac{|l_{PBE} - l_{ICSD}|}{l_{ICSD}}$$

- If the error is more than 5%, we predict them to be 2D materials; at least **1356 such materials** Found, recalculate with better optB88 functionals, tight DFT convergence



JARVIS-DFT workflow



- Written in python language
- Separate calculation for mono-layer and bulk materials
- Mono-layer created by adding additional vacuum (20 Angstrom) in z-direction
- c-lattice constant not-optimized for mono-layer (custom compilation of VASP)
- Example:

```
bulk@mp_48
1.0000000000000000
1.2325005105744926 -2.1347529556398626 0.0000000000000000
1.2325005105744926 2.1347529556398626 0.0000000000000000
0.0000000000000000 0.0000000000000000 6.6962686124617425
```

```
C
4
Direct
0.0000000000000000 0.0000000000000000 0.2500000000000000
0.0000000000000000 0.0000000000000000 0.7500000000000000
0.3333330000000032 0.6666669999999968 0.2500000000000000
0.6666669999999968 0.3333330000000032 0.7500000000000000
```

```
Surf-mp-48
1.0000000000000000
1.2320971008984494 -2.1340542301747005 0.0000000000000000
1.2320971008984494 2.1340542301747005 0.0000000000000000
0.0000000000000000 0.0000000000000000 30.803073000000013
```

```
C
2
Direct
0.0000000000000000 0.0000000000000000 0.0633300000000006
0.3333330000000032 0.6666669999999968 0.0633300000000006
```

Webpage

<http://www.ctcms.nist.gov/~knc6/JVASP.html>

- Click on elements, click search, click JARVIS-ID

JARVIS for DFT

Search
Mo-Te-
Refresh

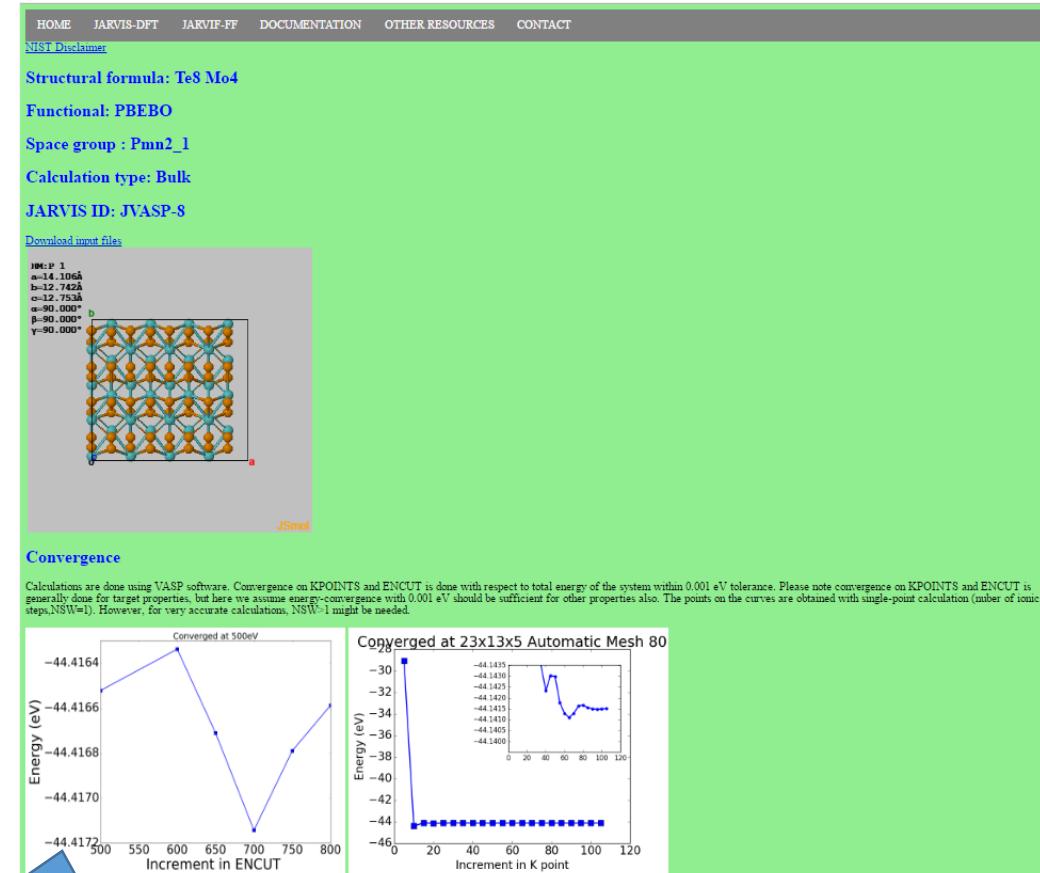
H	He
Li	Be
Na	Mg
K	Ca
Rb	Sc Ti V Cr Mn Fe Co Ni Cu Zn
Cs	Y Zr Nb Mo Tc Ru Rh Pd Ag Cd
Fr	Ba Hf Ta W Re Os Ir Pt Au Hg
	Rf Db Sg Bh Hs Mt Ds Rg Cn
	La Ce Pr Nd Pm Sm Eu Gd Tb Dy Ho Er Tm Yb Lu
	Ac Th Pa U Np Pu Am Cm Bk Cf Es Fm Md No Lr

JARVIS-ID | Formula | Functional | Calculation type | Bandgap (eV) & nature | HSE Bandgap | B_v (GPa) | G_v (GPa) | space group | Energy per atom (eV)

JVASP-109	Te8 Mo4	LDA	Bulk	0.01 na	63.511	32.607	P₂ 1 m	-6.86747197333	
JVASP-125	Te8 Mo4	PBE	Bulk	0.01 na	33.989	19.293	P₂ 1 m	-5.99087833083	
JVASP-127	Te8 Mo4	PBEBO	Bulk	0.01 na	49.011	25.467	P₂ 1 m	-3.69536699417	
JVASP-167	Te2 Mo1	PBE	IL	1.128 D	1.511 I	22.8	11.52	P-6m2	-6.00001043333
JVASP-168	Te2 Mo1	LDA	IL	1.18 I	na	25.978	13.527	P-6m2	-6.83157956667
JVASP-183	Te4 Mo2	PBEBO	IL	na	na	na	na	na	na
JVASP-186	Te4 Mo2	PBEBO	IL	0.01	na	26.133	10.84	P₂ 1 m	-3.608496505
JVASP-23	Te4 Mo2	PBE	IL	0.01	na	25.611	10.307	P₂ 1 m	-5.98622232667
JVASP-24	Te4 Mo2	LDA	IL	0.01	na	29.511	11.427	P₂ 1 m	-6.81637419167
JVASP-25	Te4 Mo2	PBE	IL	0.01	na	25.633	10.2	P₂ 1 m	-5.986301538
JVASP-26	Te4 Mo2	LDA	IL	0.01	na	29.544	11.267	P₂ 1 m	-6.816402333
JVASP-371	Te2 Mo1	PBEBO	IL	1.107 D	1.484 D	23.767	11.7	P-6m2	-3.61617248667
JVASP-316	Te8 Mo4	LDA	Bulk	0.01	na	63.256	32.433	Pmn2_1	-6.86714281083
JVASP-534	Te8 Mo4	PBE	Bulk	0.01	na	33.889	19.293	Pmn2_1	-5.99092602333
JVASP-535	Te4 Mo2	PBE	Bulk	0.879 I	na	31.133	22.68	P6 3/mmc	-6.005211005
JVASP-536	Te4 Mo2	LDA	Bulk	0.631 I	1.053 I	50.889	38.973	P6 3/mmc	-6.886117685
JVASP-537	Te4 Mo2	PBEBO	Bulk	0.754 I	na	43.444	29.847	P6 3/mmc	-3.707153305

Read WARNINGS!

- Kpoint, Energy cut-off and force convergence (N/A in other DFT databases, but very important!)
- Both bulk and single-layer materials available
- Makes ~5 times slower, but much reliable**

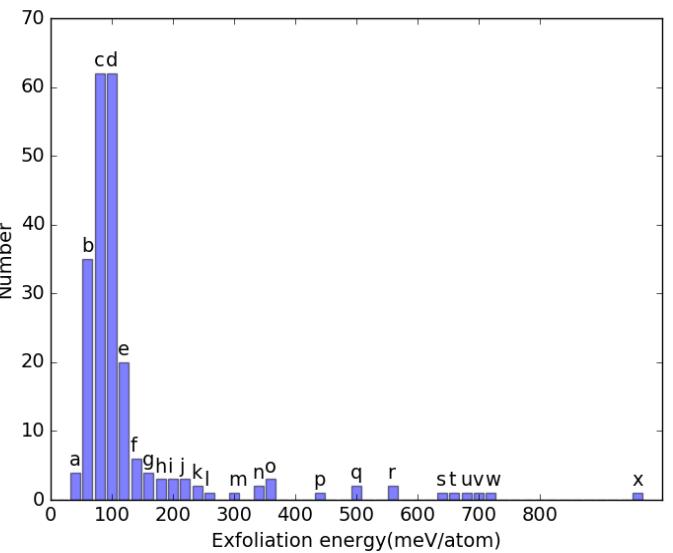


- >1280 hits in <6months
- Soon will be part of Citrine database

Exfoliation energy calculations

- Exfoliation energy:

$$E_f = \frac{E_{1L}}{N_{1L}} - \frac{E_{bulk}}{N_{bulk}}$$



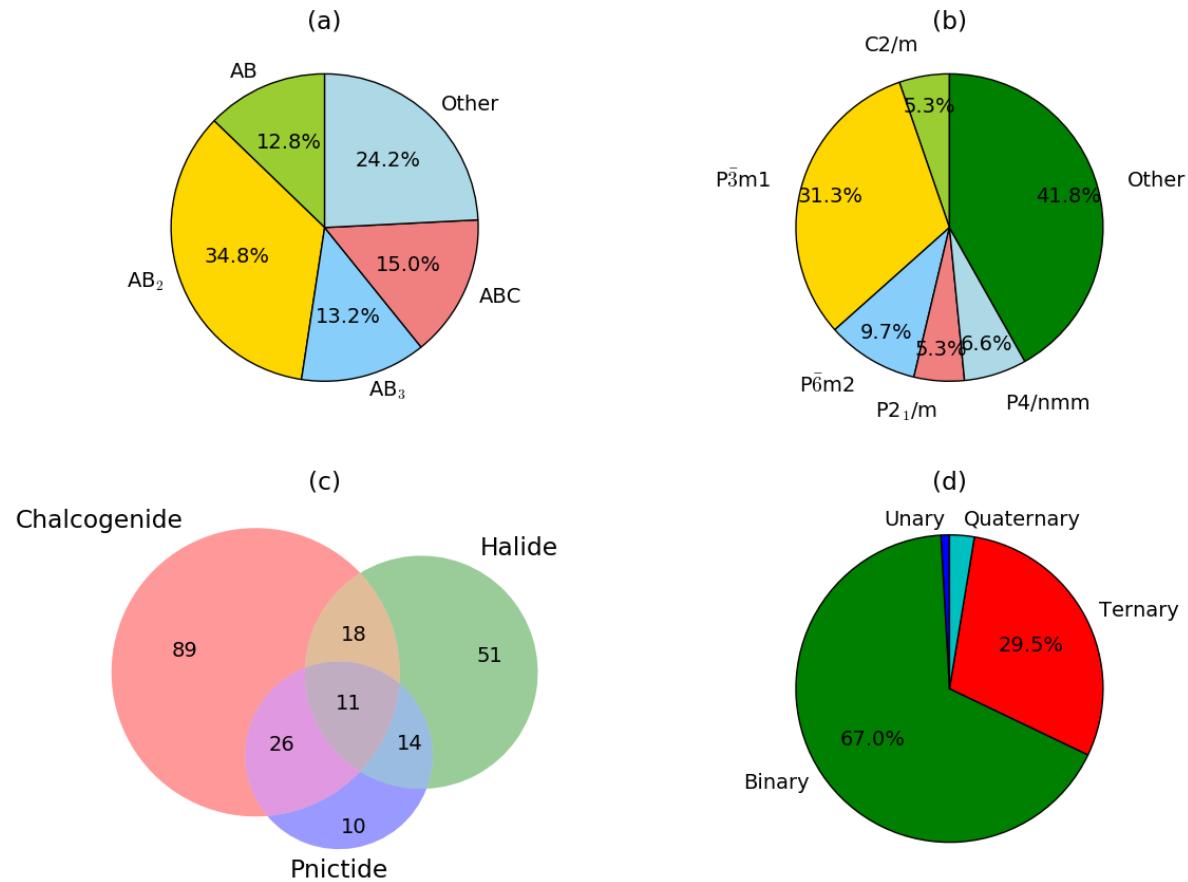
- 1012 bulk, 430 Single layer, 371 common, calculations still running (thousands of processors needed)

Energy range (meV)	Materials
a) 0-40	TiNCl, SiH ₄ , HfBrN, Mg(AlSe ₂) ₂
b) 40-60	GaSe, CrS ₂ , ZrS ₃ , NiO ₂ , GaS, ZrSe ₃ , NdTe ₃ , US ₃ , TiS ₃ , PrIO, DySI, Sc ₂ CCl ₂ , ThIN, TiBrN, InClO, LuSBr, SrHI, BiIO, BiBrO, KMnP, TiIN, Sc ₂ NCl ₂ , TiSbO ₃ , ZrCl, SmTe ₃ , PrTe ₃ , As ₂ O ₃ , Nb ₂ CS ₂ , RbMnAs, SiH, Bi ₂ TeI, ScCl, TbBr, Ge(BiTe2) ₂ , GaS
c) 60-80	WSe ₂ , WS ₂ , MoS ₂ , C, SnO ₂ , PtO ₂ , CdBr ₂ , ReSe ₂ , CrSe, MgCl ₂ , CoBr ₂ , ZrCl ₂ , MgBr ₂ , TcS ₂ , FeCl ₂ , MnCl ₂ , MnBr ₂ , InSe, CrBr ₃ , VCl ₃ , USe ₃ , IrCl ₃ , ScCl ₃ , RhCl ₃ , TaI ₂ O, DySBr, ErSeI, ErS ₂ , BiClO, OsCl ₂ O, CdCl ₂ , BN, Nb(SCl) ₂ , Bi ₂ Te ₂ S, ThBrN, HfCl ₄ , Bi ₂ Te ₂ Se, MgPSe ₃ , CdPS ₃ , ScPS ₄ , PPdS, TmAg(PSe ₃) ₂ , ScAg(PSe ₃) ₂ , ErAg(PSe ₃) ₂ , ScAg(PS ₃) ₂ , Nb ₃ Cl ₈ , Nb ₃ TeCl ₇ , InAg(PSe ₃) ₂ , Hf ₃ Te ₂ , SNCl, Sr ₃ Si ₂ , TiCl ₂ , HfFeCl ₆ , GaTe, CS ₂ , Nb(SeCl) ₂ , CrCl ₃ , BiI, TiBr ₂ , GaAg(PSe ₃) ₂ , CdPS ₃ , TiS ₂
d) 80-100	NbS ₂ , MoSe ₂ , NbSe ₂ , WTe ₂ , MoTe ₂ , VSe ₂ , ZrS ₂ , HfS ₂ , HfSe ₂ , MoS ₂ , PtO ₂ , PtS ₂ , SnS ₂ , SnSe ₂ , TiO ₂ , TiS ₂ , TiSe ₂ , ZrSe ₂ , TaS ₂ , SiTe ₂ , TaSe ₂ , VS ₂ , TaSe ₂ , MgI ₂ , SbI ₃ , PbI ₂ , GeI ₂ , SiS ₂ , MnI ₂ , CaI ₂ , RhBr ₃ , BiI ₃ , MoBr ₃ , RuBr ₃ , PCl ₃ , AuI, BPS ₄ , IrBr ₃ , Re(AgCl) ₃ , AlPS ₄ , AlSiTe ₃ , PPdSe, CrSiTe ₃ , Nb ₃ TeI ₇ , NdI ₂ , Al ₂ Te ₃ , S ₅ N ₆ , AlTeI ₇ , AlSeBr ₇ , CdI ₂ , PSe, Ta ₃ TeI ₇ , TmI ₂ , SbBr ₃ , P ₄ S ₅
e) 100-120	HfTe ₂ , PtSe ₂ , TiTe ₂ , WO ₂ , SnO, BCl ₃ , Te ₂ Br, Te ₂ I, PBr ₃ , TiI ₃ , BiTeCl, BiTeI, TiPt ₂ S ₃ , AlBr ₃ , BiSBr, CaN, Mn ₂ Bi, HgI ₂ , SrThBr ₆ , P
f) 120-140	BBR ₃ , AlI ₃ , TiTe ₃ Pt ₂ , SbSBr, TiPd ₂ Se ₃ , P ₂ Se ₅
g) 140-160	PdS ₂ , Te ₂ Pt, BI ₃ , Ta(ICI) ₂
h) 160-180	PdSe ₂ , NiTe ₂ , NbI ₅
i) 180-200	ZrS, PI ₃ , BaBrCl
j) 200-220	Te ₂ Pd, Te ₂ Ir, BiSe ₂
k) 220-240	SbSeI, BiSI
l) 240-260	Bi ₂ Te ₃
m) 260-300	VS ₂
n) 300-340	CaSn, KAuSe
o) 340-360	KAuS, RbAuS, RbAuSe
p) 360-440	Sc ₂ C
q) 440-500	GaN, TiSe ₂
r) 500-560	Sr ₂ H ₃ , Ca ₂ H ₃
s) 560-640	AlN
t) 640-660	TiTe ₂
u) 660-680	CrSe ₂
v) 680-700	ZrTe ₂
w) 700-720	CoAs ₂
x) 720-960	Ti ₂ O

Choudhary et al., (Submitted)

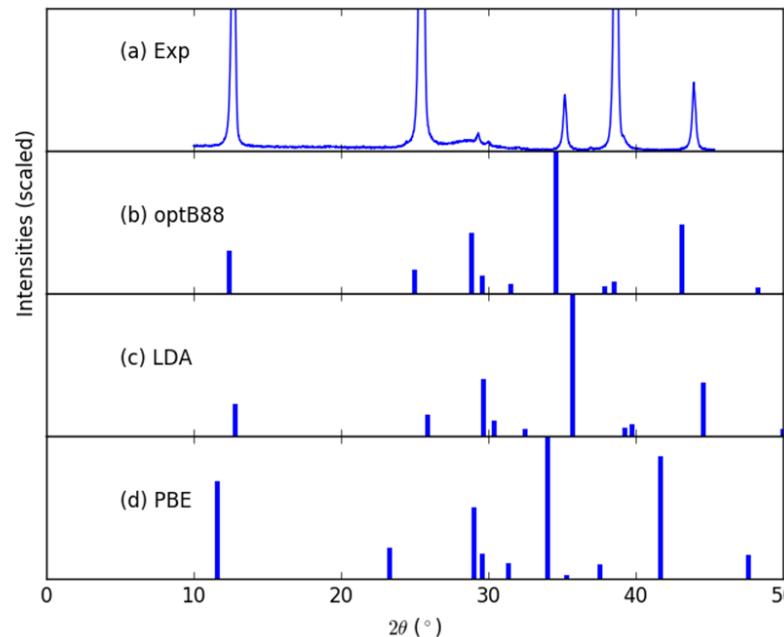
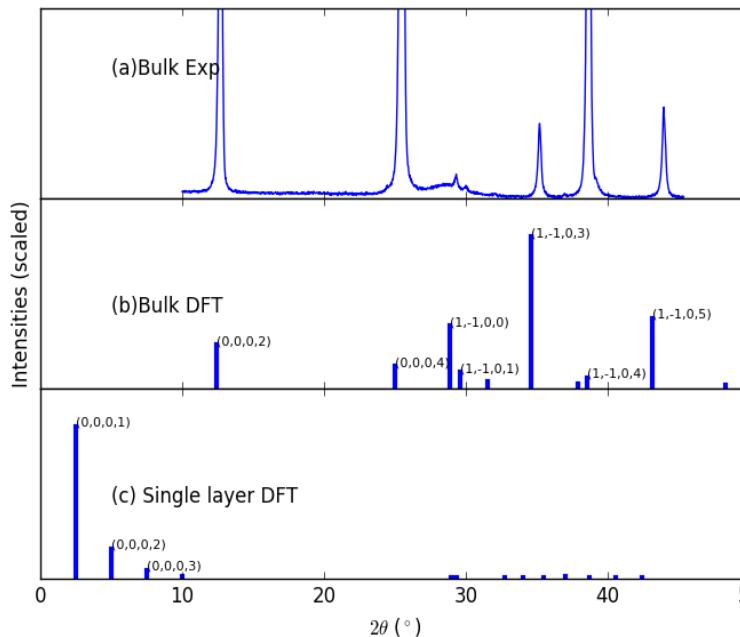
Email: kamal.choudhary@nist.gov

Classification of database



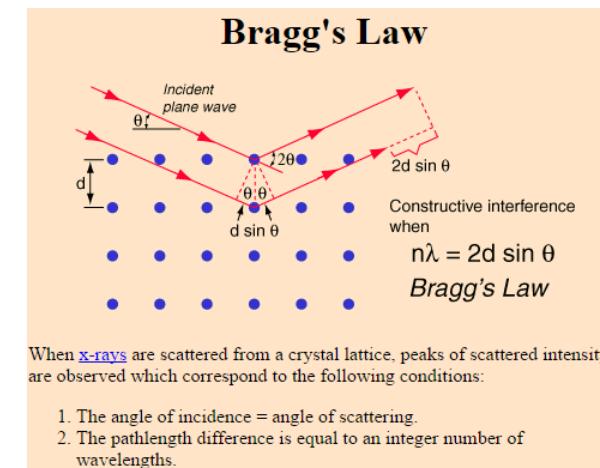
- 2D materials are not just AB₂
- More than 2H ($P\bar{6}m2$) and 1T ($P\bar{3}m1$) prototype
- Venn diagram
- Majority of 2D materials are binary

Computational X-Ray diffraction



- Experimental data from Irina Kalish ($\text{MoTe}_2\text{-}2\text{H}$)
- Computational XRD implemented in pymatgen
- Excellent agreement of optB88 and experiments
- PBE gives erroneous peaks
- Single-layer XRD also available

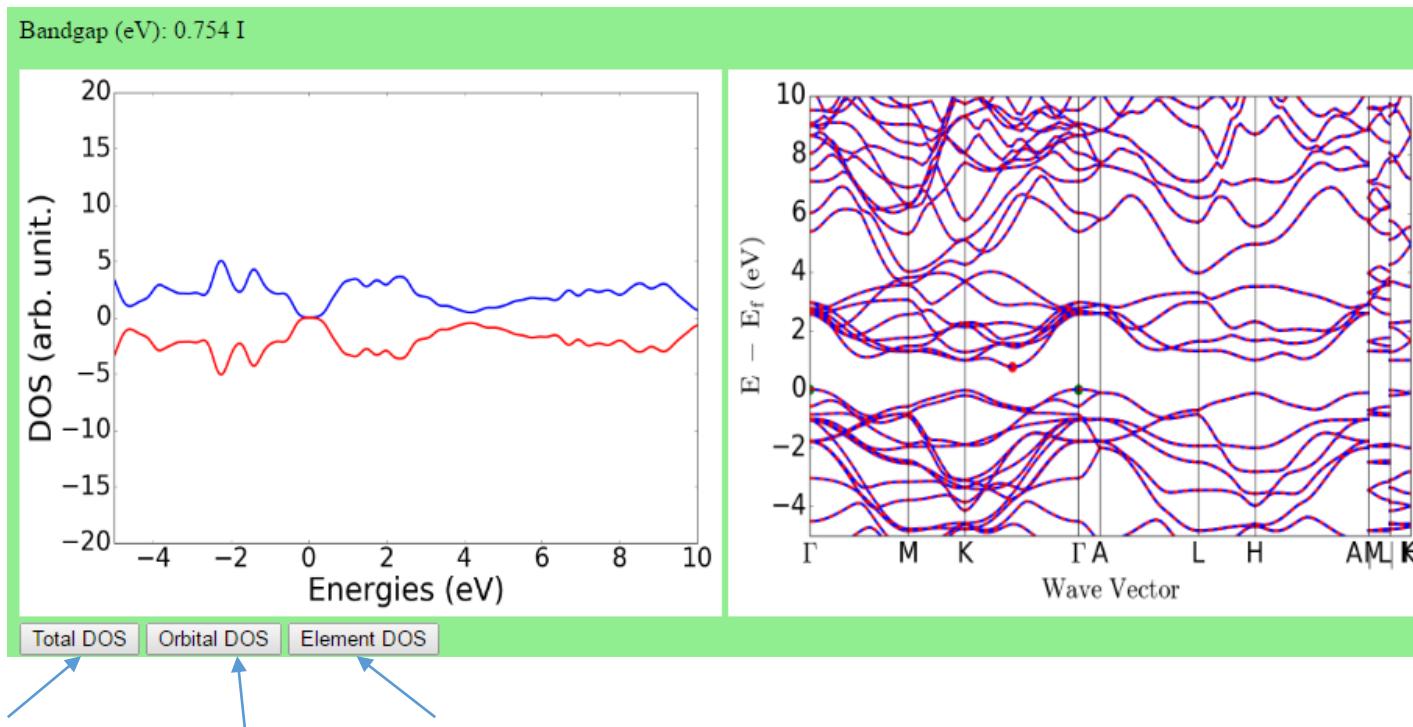
- Computational X-ray diffraction:



Choudhary et al., (Submitted)

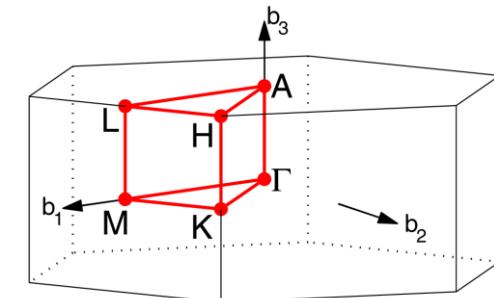
Computational band-structure

- Density of states and Band-structures:



- DOS can be used to approximately calculate magnetic property such as **Orbital magnetic moment**

- optB88 bandgaps
- Few LDA, PBE and **HSE06 bandgaps**

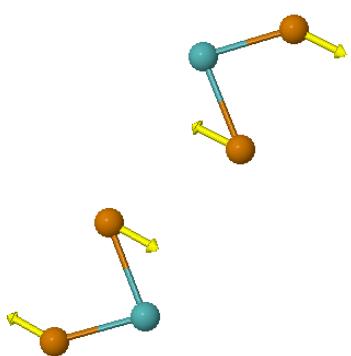


HEX path: Γ -M-K- Γ -A-L-H-A|L-M|K-H
[Setyawan & Curtarolo, DOI: 10.1016/j.commatsci.2010.05.010]

Brillouin zone for hexagonal system

Elastic constants, Γ -point phonons

Elastic Tensor C_{ij} GPa							Point group	
119.5 32.8 11.3 0.0 0.0 0.0							point_group_type: 6/mmm	
32.8 119.5 11.3 0.0 0.0 0.0							Visualize Phonons here	
11.3 11.3 41.2 0.0 0.0 0.0							Phonon mode (cm⁻¹) Representation	
0.0 0.0 0.0 43.3 0.0 0.0							-0.0465217481 None	
0.0 0.0 0.0 0.0 15.5 0.0							-0.0362848079 None	
0.0 0.0 0.0 0.0 0.0 15.5							0.0136711726 A2u I	
22.1183439196							22.1183439196 E2g R	
35.3915249465							35.3915249465 B2g	
112.721031921							112.721031921 E2u	
114.212804132							114.212804132 E1g R	
166.789595244							166.789595244 B1u	
170.535305026							170.535305026 A1g R	
226.002625615							226.002625615 E1u I	
226.037033636							226.037033636 E2g R	
275.946925877							275.946925877 A2u I	
279.963912554							279.963912554 B2g	

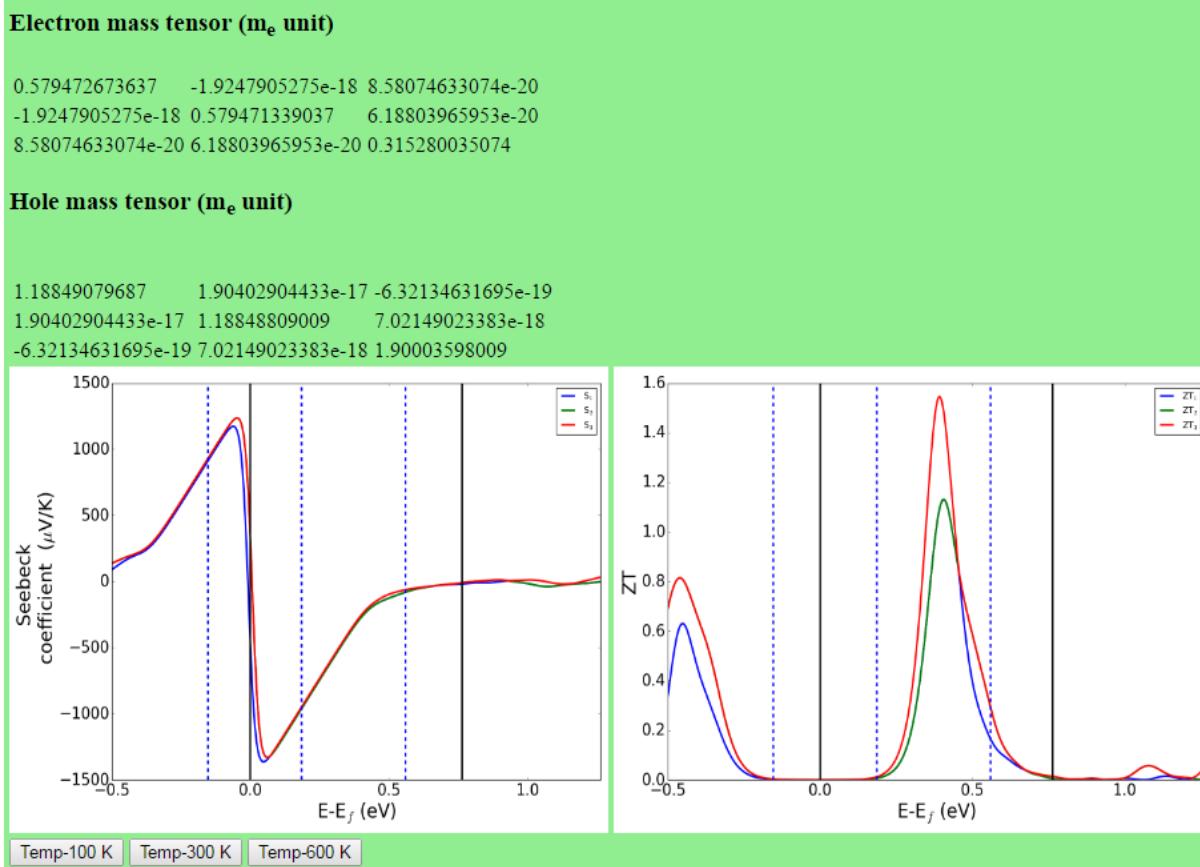


Read WARNINGS!

DFT Phonon mode 1x1x1 cell (cm ⁻¹)	DFT 1x1x1 Representation	Experimental modes	Experimental Representation	DFT Phonon modes for 2x2x2 cell	DFT 2 Represent
-0.13	Bu I			-0.58	Bu
-0.06	Au I			-0.04	Bu
-0.04	Bu I			0.06	Au
7.36	Bu I			6.08	Bu
25.80	Au I			24.65	Au
31.04	Bu I			26.09	Bu
76.66	Ag R			74.09	Ag
84.86	Ag R	80.56	Ag R	81.54	Ag
88.03	Bg R			87.63	Bg
90.32	Bg R			89.60	Bg
103.28	Bg R	96.54	Bg R	102.36	Bg
104.17	Bg R	108.32	Bg R	102.95	Bg
107.55	Au I			106.61	Au
108.54	Ag R			107.12	Ag
109.91	Au I			108.66	Au
112.03	Ag R	112.8	Ag R	111.10	Ag
114.06	Bu I			115.46	Bu
122.27	Bu I			121.19	Bu
124.29	Ag R			121.90	Ag
126.27	Ag R	129.2	Ag R	123.58	Ag
129.64	Bu I			126.25	Bu
134.19	Bu I			130.53	Bu
151.35	Ag R			150.19	Ag
156.27	Ag R	163.32	Ag R	153.05	Ag
175.28	Au I			176.42	Au
175.94	Au I			176.93	Au
184.24	Bg R			181.76	Bg
186.30	Bg R	191.64	Bg R	185.04	Bg
190.81	Bu I			189.22	Bu
190.93	Bu I			189.61	Bu
236.52	Ag R			234.90	Ag
239.60	Ag R	248.45	Ag R	239.42	Ag
252.88	Ag R	258.61	Ag R	250.04	Ag
253.04	Ag R	263.34	Ag R	252.26	Ag
264.26	Bu I			264.26	Bu
265.13	Bu I			264.73	Bu

- Experimental data from Ryan Beams, NIST
- <9% error, Beams et al., ACS Nano, 10 (10), 9626 (2016)

Thermoelectric properties



- BoltzTrap code in pymatgen
- Constant relaxation time approximation
- Only electronic contribution to thermal conductivity
- Perturbation in Fermi-distribution

$$f_0(\mathbf{r}, \mathbf{k}) = \frac{1}{1 + e^{\frac{E(\mathbf{k})}{k_B T}}} \quad f = f_0 + f_1.$$

For a stationary near-equilibrium distribution, the Boltzmann equation is

$$\frac{df}{dt} = 0 = \frac{\partial f}{\partial t} + \nabla_{\mathbf{r}} f \cdot \frac{\partial \mathbf{r}}{\partial t} + \nabla_{\mathbf{k}} f \cdot \frac{\partial \mathbf{k}}{\partial t} + \frac{\partial f}{\partial t} \Big|_{scat},$$

$$ZT = \frac{S^2 \sigma}{\kappa} T, \quad \kappa = \kappa_e + \kappa_L$$

- Doping dependent properties for:
Hall tensor, thermal conductivity,
electronic conductivity tensor, Seebeck tensor

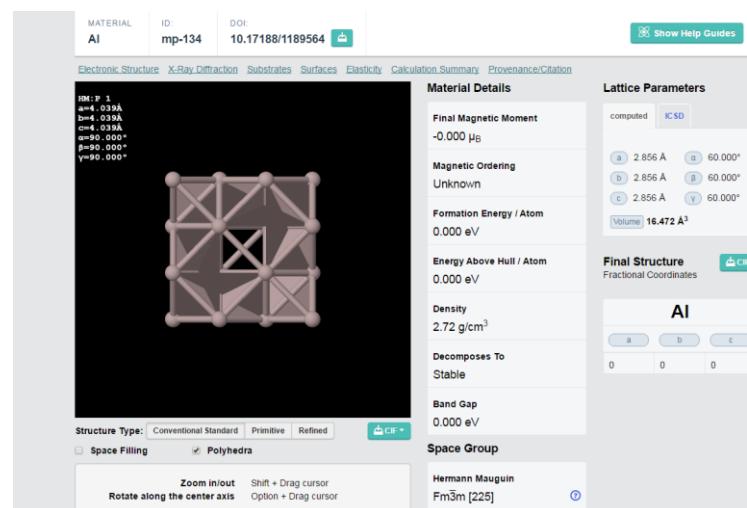
(UNDER CONSTRUCTION)

Read WARNINGS!

Providing References to other databases

Reference

mp-134
ICSD-ID: 150692
[AFLOW link](#)
[MP link](#)



Summary Part-2

- New publicly available database specially dedicated to 2D materials (both bulk and single layer)
- Simple criteria to identify 2D materials, Nanowires and Quantum dots?
- 2D-2L, piezoelectric properties, defects and interface properties on the way
- Largest dataset for optical materials, easy screening of materials, characterizing nature of optical transitions
- Many-body effects (GW,BSE), Nonlinear optical properties on the way
- JARVIS-DFT data will be available in Citrine database soon
- We welcome your experimental data for validation !

JARVIS-ML

- 3A Introduction
- 3B Implementation using current methods
- 3C New hybrid ML + FF

3A Intro Machine Learning

- Learning == Dimensionality reduction
- No one knows the Potential Energy Surface (PES) : Need of ML
- Similarly, for materials == linear combination of features
- Very simple examples for feature selection:
 - X: input data, say NaCl: [mean of atomic radii of constituents, electronegativity, atomic mass,...]
 - Y: target data: [heat of formation, bulk modulus, bandgap,.....]
 - Machine-Learning packages : Scikit-learn (Google), Tensor-flow (Google), Theano (Université de Montréal ,Canada)
 - Train a machine learning model with Neural network, linear regression, random-forest etc.
 - For new material say (KCl), convert in the format of X, then query the model

3B Example using JARVIS-DFT data

127.0.0.1:5000

Enter Material Composition
(Get predicted properties)

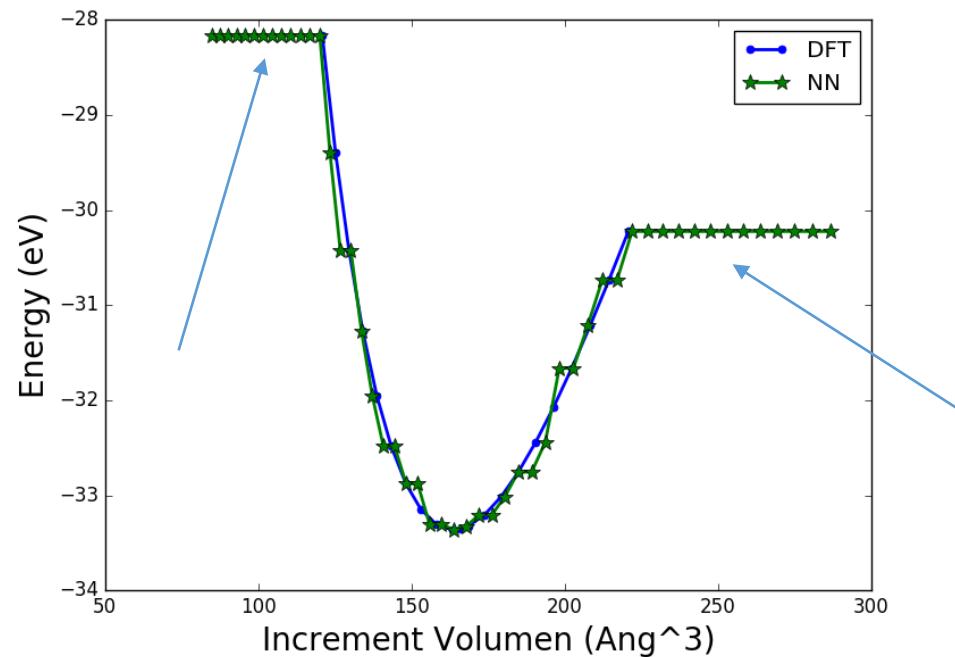
 Send

Bandgap (eV)	1.49800568727
Material	AlSHLiOZn
n	3.47446745993
C11 (GPa)	193.258274274

- Bandgap, refractive index elastic constant for example
- Currently running on my computer, testing and debugging may take several months

3C Artificial Neural network + FF

- Present trend: Thousands of DFT calculation, structure-property(energy) relationship, parametrize force-fields, limited to what we train—wrong extrapolation features! (E-V curve)
- Attempt to make new force-field: Machine-learning+FF-contribution (LJ, EAM, COMB and so on), Machine learning for interpolation, Physics for extrapolation
- ~500 DFT calculations for training database



For any atom i , we define the parameter

$$v_i^{(0)} = \frac{1}{\sigma^3} \sum_{j \neq i} e^{-r_{ij}^2/\sigma^2} f_c(r_{ij}),$$

where $f_c(r)$ is a truncation function defined by

$$f_c(r) = \begin{cases} \frac{(r - r_c)^4}{d^4 + (r - r_c)^4} & r \leq r_c \\ 0, & r \geq r_c. \end{cases}$$

*Notes Yuri Mishin

DEMO !

Conclusions and Future work

- New publicly available databases!
- Simply Google “**JARVIS-DFT**”, “**JARVIS-FF**” or browse <http://www.ctcms.nist.gov/~knc6/JARVIS.html>
- Identification and characterization of new/existing materials using classical and quantum methods
- Flexible python framework to accommodate other models as well: CALPHAD, FEM, PF; golden time for Multi-scale modelling
- Need of your **Experimental data** for validation!
- Soon JARVIS will be available on smart-phones with voice-recognition apps!
- JARVIS-Wiki for documentation, and Hands-on tutorials will be available soon



THANKS FOR YOUR ATTENTION !

JARVIS for YOU !
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