

Tools for atomic-scale model construction Cluster expansions and force constants

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### **Tools for atomic-scale model construction**









### **Force constant expansions**

Taylor expansion of the potential energy

$$E = E_0 + \sum_i \Phi_i u_i + \frac{1}{2!} \sum_{ij} \Phi_{ij} u_i u_j + \frac{1}{3!} \sum_{ijk} \Phi_{ijk} u_i u_j u_k \dots$$

$$u_i \text{ Displacement from static equilibrium position}$$





### **Force constant expansions**

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$$u_i \text{ Displacement from static equilibrium position}$$

**Phonon dispersions** 

$$\Phi_{ij} \to D_{\alpha\beta}(\boldsymbol{q}) \to \omega_k(\boldsymbol{q})$$



Momentum CHALMERS

Frequency



### **Force constant extraction**

$$E = E_0 + \sum_{i} \Phi_i u_i + \frac{1}{2!} \sum_{ij} \Phi_{ij} u_i u_j + \frac{1}{3!} \sum_{ijk} \Phi_{ijk} u_i u_j u_k \dots$$

"Direct" approach: systematic enumeration

Harmonic approximation  $\rightarrow$  phonopy Third-order FCs  $\rightarrow$  phono3py, almaBTE  $\rightarrow$  Poor scaling with system size and order

$$\Phi_{ij} = \frac{\partial^2 E}{\partial u_i \partial u_j}$$
$$\approx -\frac{F_i(u_j) - F_i(0)}{u_j}$$





### **Force constant extraction**

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$$\approx -\frac{F_i(u_j) - F_i(0)}{u_j}$$

 $\Phi_{ij} = \Phi_{ij}(\boldsymbol{x})$ 

 $\min(\|Ax - f\|_2^2)$ 

"Regression" approach: fit to snapshots

→ TDEP & ALAMODE

Hellman et al. PRB 2011, Tadano et al. JPCM 2014

→ Compressive sensing Zhou, Ozolins *et al.* PRL 2014

### hiphive

Generalizes regression approach and makes it easily accessible



# Efficiency 2nd-order: Vacancy in BCC-Ta hiphive

### Convergence with training size





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### **Efficiency 3rd-order: Silicon**



Thermal conductivity of Silicon via Boltzmann transport

$$\kappa_l = \frac{1}{2V} \sum_{j\boldsymbol{q}} \lambda_{j\boldsymbol{q}}(T) v_{j\boldsymbol{q}} c_{j\boldsymbol{q}}(T)$$

### Requires third-order force constants





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Thermal conductivity of Silicon via Boltzmann transport





### **Efficiency 3rd-order: Silicon**

Thermal conductivity of Silicon via Boltzmann transport





# **Anharmonicity in clathrates**

Inorganic clathrate Ba<sub>8</sub>Ga<sub>16</sub>Ge<sub>30</sub>

• Anharmonic Ba (rattler) modes



Guest: Ba Host: Ga, Ge







# **Anharmonicity in clathrates**

Inorganic clathrate Ba<sub>8</sub>Ga<sub>16</sub>Ge<sub>30</sub>

- Anharmonic Ba (rattler) modes
- Fourth-order model (6000 parameters)
- Reproduces anharmonicity of Ba





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# **Anharmonicity in clathrates**

Inorganic clathrate Ba<sub>8</sub>Ga<sub>16</sub>Ge<sub>30</sub>

- Anharmonic Ba (rattler) modes
- Fourth-order model (6000 parameters)
- Reproduces anharmonicity of Ba
- Reproduces thermal conductivity





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Eriksson et al., Adv. Theory Simul. 2019, 1800184



## **Cluster Expansions**

Expansion of the total energy in site occupations



Need efficient means to extract effective cluster interactions (ECIs)  $J_{\alpha}$   $\rightarrow$  Again a linear problem



## **Cluster Expansions**

ice

Expansion of the total energy in site occupations

$$E(\boldsymbol{\sigma}) = E_0 + \sum_i J_i \sigma_i + \sum_{ij} J_{ij} \sigma_i \sigma_j + \sum_{ijk} J_{ijk} \sigma_i \sigma_j \sigma_k + \dots$$





# CE construction for a Ag-Pd alloy icet



- zerolet + singlet
- + 24 pairs
- + 20 triplets
- + 35 quadruplets
- = 81 parameters

ARDR: fast convergence and sparse solutionLASSO: more false-positivesRFE: requires more structures



# CE construction for a Ag-Pd alloy ice





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# **Ag-Pd alloy: CE sampling**

icet





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# **Ag-Pd alloy: CE sampling**

ice



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#### Clathrates

• Chemical order → thermoelectric performance Ångqvist *et al.* **2017** 









#### Clathrates

• Chemical order → thermoelectric performance Ångqvist *et al.* **2017** 





### Zeolites

- Aluminium distribution on framework
- Löwenstein's rule violated

Magnus Fant 2019







### Clathrates

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### Zeolites

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   Magnus Fant 2019



### Interface systems

- Surface segregation in PdAuCu (Pernilla Tanner 2019)
- Stability of WC-Co interfacial phases (Martin Gren 2019)







### icet features



### • Structure generation

- Enumeration
- SQS (Special Quasi-random Structures)
- Ground-states via MIP (Mixed Integer programming)
- Training via scikit-learn
  - LASSO, ARDR, RFE, ...
- Monte Carlo sampling
  - Canonical ensemble
  - Semi-Grand canonical ensemble
  - Variance-constrained semi-grand canonical ensemble
  - Simulated annealing
  - Wang-Landau sampling



### **Future work: Configurational + Vibrational**

 $\mathcal{Z} = \mathcal{Z}_{conf} \mathcal{Z}_{vib}$ 

Coupling between configurational and vibrational dofs → Very computational expensive



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### **Tutorials**

### icet tutorial Thursday 1pm

#### # setup

- cs = ClusterSpace(atoms, [6.0, 5.0], ['Ag', 'Pd'])
- sc = StructureContainer(cs)
- for structure in training\_structures: sc.add\_structure(structure)

#### # training

opt = Optimizer(sc.get\_fit\_data(), fit\_method='ardr')
opt.train()
ce = ClusterExpansion(cs, opt.parameters)

#### # sample

calc = ClusterExpansionCalculator(ce, atoms)
mc = CanonicalEnsemble(atoms, calc, temperature=600)
mc.run(1000)

### hiphive tutorial Thursday 3pm

#### # setuµ

- cs = ClusterSpace(atoms, [8.0, 6.0])
  sc = StructureContainer(cs)
  for structure in training structures:
  - sc.add structure(structure)

#### # training

opt = Optimizer(sc.get\_fit\_data(), fit\_method='rfe')
opt.train()
fcp = ForceConstantPotential(cs, opt.parameters)

#### t write to phono3py format

fcs = fcp.get\_force\_constants(supercell)
fcs.write\_to\_phonopy('fc2.hdf5')
fcs.write\_to\_phono3py('fc3.hdf5')



# The end



- Mattias Ångqvist
- William Muñoz
- Magnus Rahm
- Erik Fransson
- Joakim Brorsson
- Céline Durniak
- Piotr Rozyczko
- Thomas Rod
- Paul Erhart



- Fredrik Eriksson
- Erik Fransson
- Paul Erhart



https://materialsmodeling.org/software



Knut och Alice Wallenbergs Itiftelse



### **Extra**



## **Cluster decomposition**

### General procedure

- 1. Generate list of clusters
- 2. Obtain symmetry operations
- 3. Categorize clusters into orbits
- 4. Identify independent parameters
- 5. Apply sum rules





## **Structure decomposition**

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- 1. Generate structure e.g., by applying random displacements or superposing normal modes
- 2. Convert displacements u<sub>i</sub> into "cluster vectors"

   → each structure yields a matrix A<sub>i</sub> with
   N<sub>parameters</sub> columns and 3N<sub>atoms</sub> rows
- 3. Each structure comes with a target force vector





## **Parameter optimization**

- 1. Compile multiple structures into one (large) fit matrix A
- 2. Solve the linear problem Ap=f





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Ridge-regression/LASSO/ $||\boldsymbol{A}\boldsymbol{p} - \boldsymbol{f}||_2^2 + \alpha ||\boldsymbol{p}||_1 + \beta ||\boldsymbol{p}||_2^2$ 

Automatic relevance determination regression (ARDR) Recursive feature elimination (RFE)



**Cross-validation** 

