

Quantum Alchemy: Treating Chemical Space with Perturbations

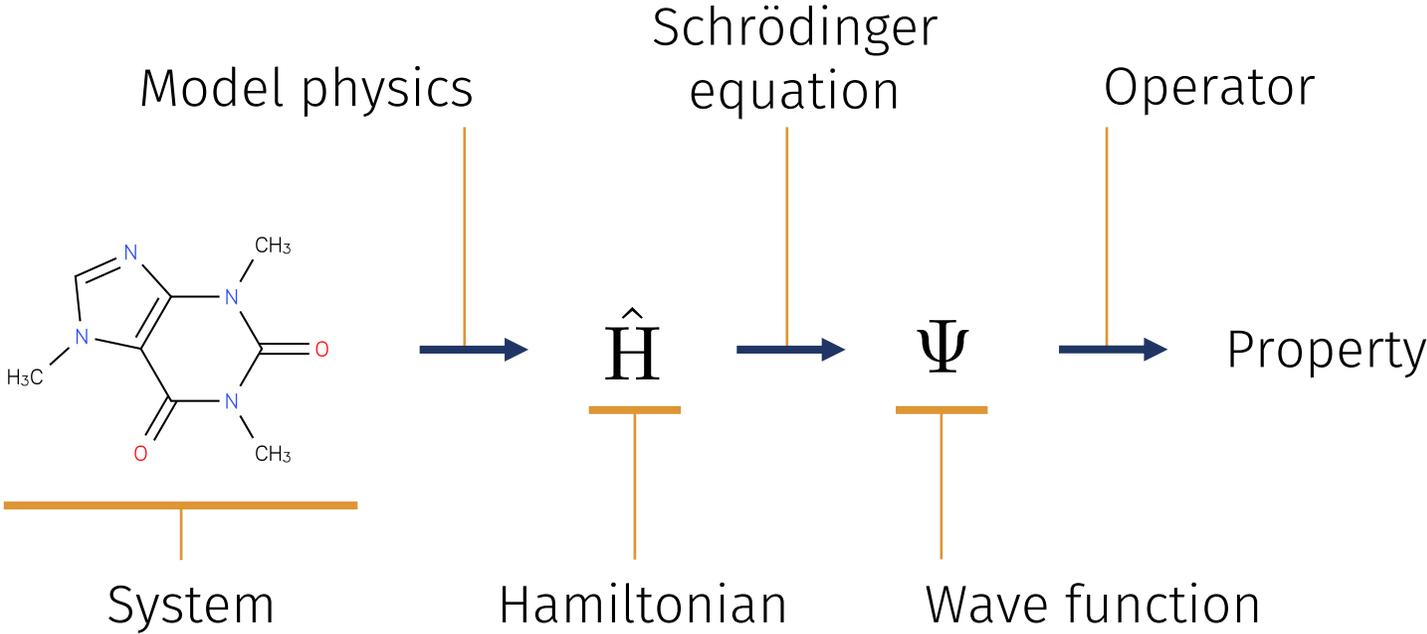
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 nablachem.org/talks

 [ferchault](https://github.com/ferchault)

 [@ferchault](https://twitter.com/ferchault)



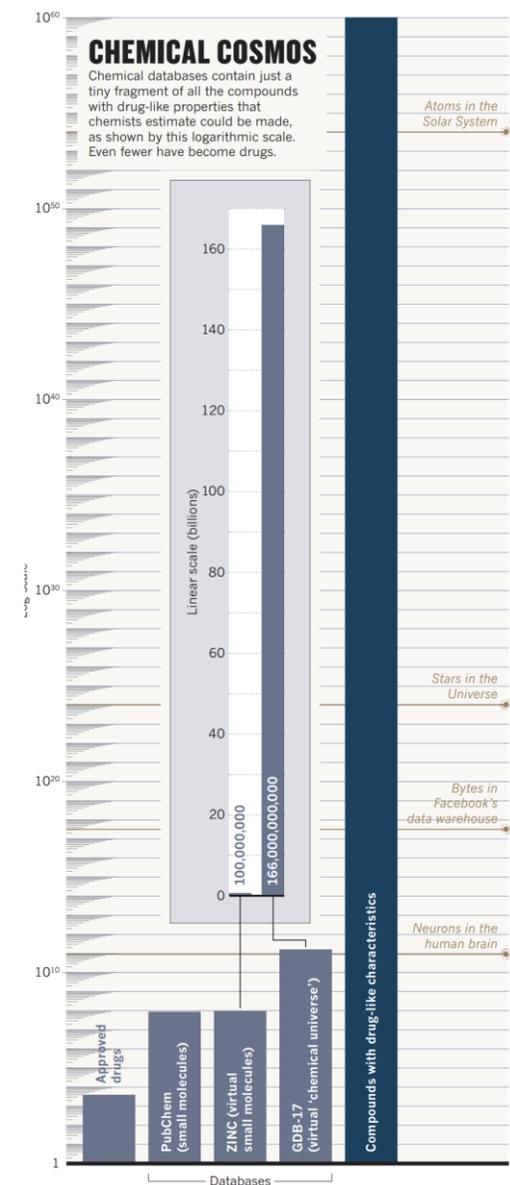
Scaling of Chemical Space

Commercial databases

- 164 million molecules
- 15k added daily

Scale

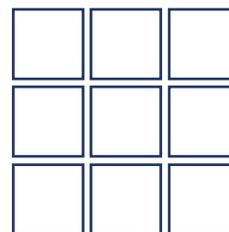
- One person: 1 million compounds/second
- 10 billion people on earth
- 10^{26} universe ages to go through



Face centered cubic and 70 elements only

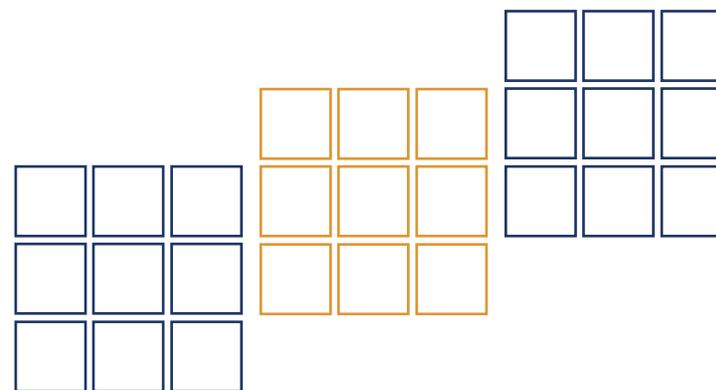
9 primitive cells

- Binary 10^7
- Ternary: 10^{13}
- Quaternary: 10^{15}

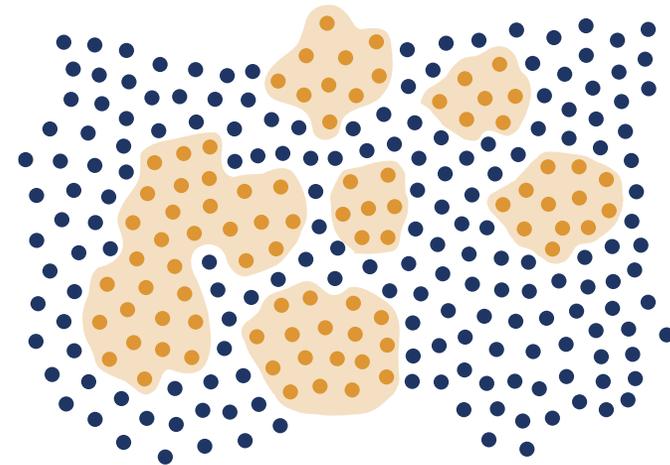
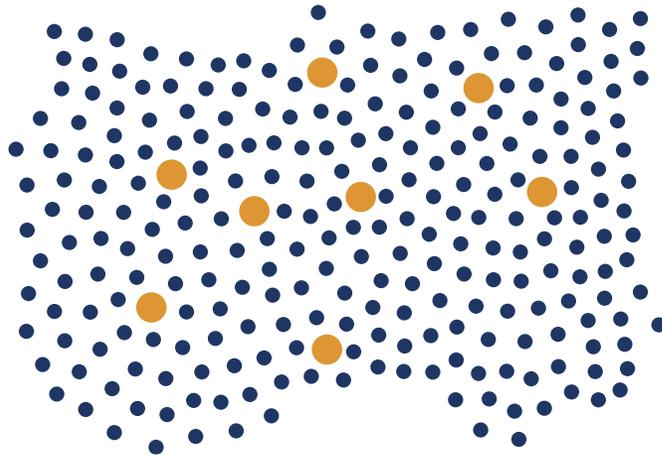


27 primitive cells

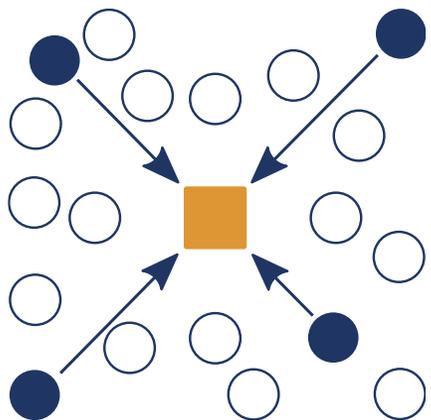
- Binary: $\sim 10^{17}$
- Ternary: $\sim 10^{29}$
- Quaternary: $\sim 10^{36}$



Speed does not matter:
even enumeration is impossible.



Machine Learning



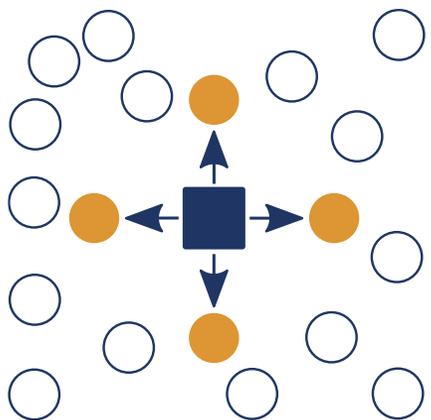
Foundations | Statistical modelling

Accuracy | Systematically improvable through data and training

Specialty | Universal, scale-bridging, data-driven approach

Limitation | Requires training data, no black box

Quantum Alchemy



Foundations | Perturbation theory

Accuracy | Systematically improvable through higher orders terms

Specialty | Combinatorial scaling with chemical diversity

Limitation | Finite range in chemical space

Idea

Treat system changes perturbatively^[1,2,3]

Build a Taylor/Padé approximant^[4]: often 100.000 times faster

Steps

Choose system



Alter system, calculate property response functions



Predict many modified systems



1 | L. L. Foldy, *Phys Rev.* 1951. 2 | E. B. Wilson, *J. Chem. Phys.* 1962.

3 | GFvR, O. A. von Lilienfeld, *Phys. Rev. Res.*, 2020. 4 | GFvR, *J. Chem. Phys.*, 2021.

Also: Ayers, Cárdenas, Keith, Geerlings, Politzer, ...

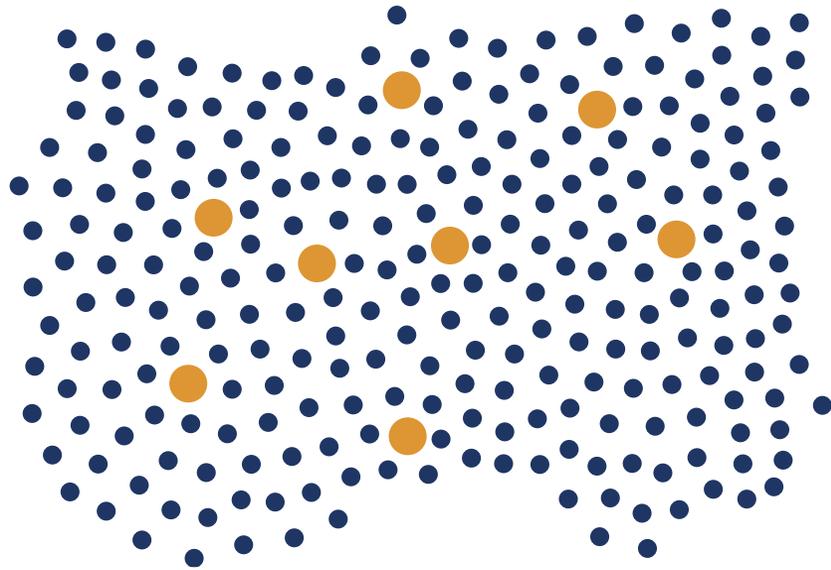
Few highly accurate calculations
instead of many intermediate ones

$$\hat{H} = \hat{H}(\underbrace{Z_i}_{4N}, \underbrace{\mathbf{R}_i}_{1D, \text{ close to } \sum_i Z_i}, \underbrace{N_e}_{1D}, \sigma)$$

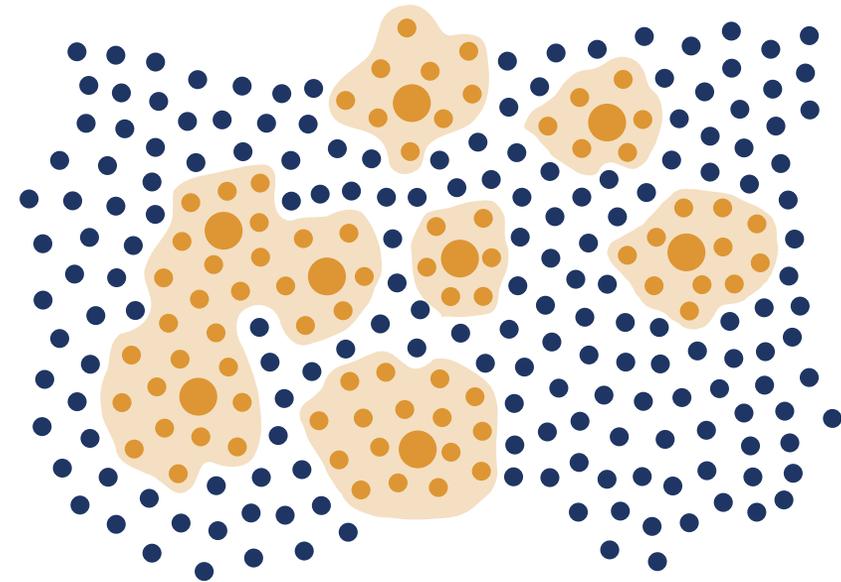


Joseph Wright, 1771

Without Perturbation



With Perturbation



Systems

- Any
- Known
- Approximated

Two systems: Interpolate between molecular isoelectronic Hamiltonians

$$\hat{H}(\lambda) \equiv \lambda \hat{H}_t + (1 - \lambda) \hat{H}_r$$

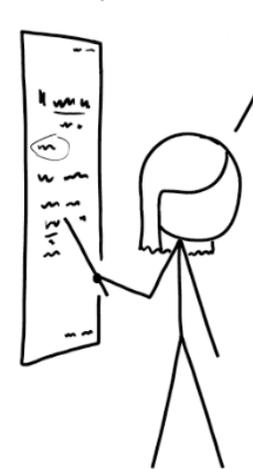
Taylor expansion around reference molecule

$$E_t = E_r + \Delta E^{\text{NN}} + \int_{\Omega} d\mathbf{r} \sum_{n=0}^{\infty} \frac{1}{(n+1)!} \Delta v \frac{\partial^n \rho_{\lambda}(\mathbf{r})}{\partial \lambda^n} \Big|_{\lambda=0}$$

(Alchemical Perturbation Density Functional Theory, APDFT)

- Gives consistent energies, densities, forces, ...
- Uses the same derivatives for all predictions

AT THIS POINT, YOU'RE PROBABLY THINKING, "I LOVE THIS EQUATION AND WISH IT WOULD NEVER END!"
WELL, GOOD NEWS!



Differentiable / Analytic + Converge quickly

- ✓ Total Energy ^[1,2]
- ✓ Dipole moments ^[2]
- ✓ Deprotonation energies ^[3]
- ✓ Photoelectron circular dichroism ^[5]
- ✓ Electron density ^[1,2]
- ✓ Non-covalent interactions ^[1]
- ✓ Ionisation Energy ^[4]
- ✓ Orbital eigenvalues ^[2]
- ✓ Binding energies ^[1,2]
- ✓ Electron Affinity ^[4]

In progress

NMR spectra, QMC energies

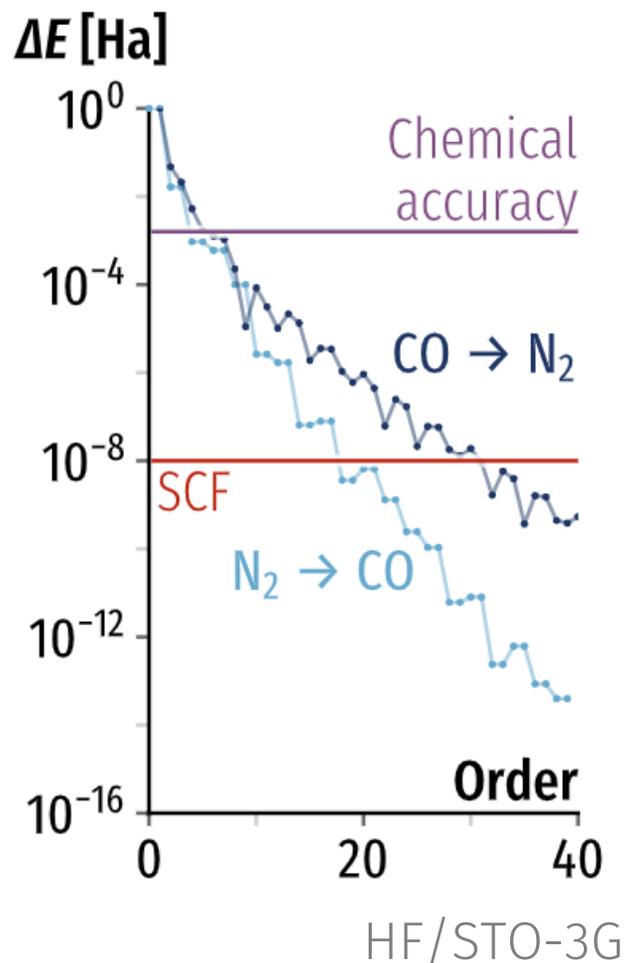
Tested in: Gaussian, Psi4, PySCF, PySCF-AD, dqc, ORCA, MRCC, cp2k, CPMD

Tested with: HF, KS-DFT, CCSD



NablaChem/nablachem
`pip install nablachem`

1 | GFvR, O. A. von Lilienfeld, *Phys. Rev. Res.*, 2020. **2** | GFvR, *J. Chem. Phys.*, 2021. **3** | GFvR, O. A. von Lilienfeld, *Phys. Chem. Chem. Phys.*, 2020. **4** | E Eikey, A Maldonado, C Griego, GFvR, J Keith, *J. Chem. Phys.*, 2022. **5** | GFvR, A Artemyev, B Lagutin, P Demekhin, *J. Chem. Phys.*, 2024.

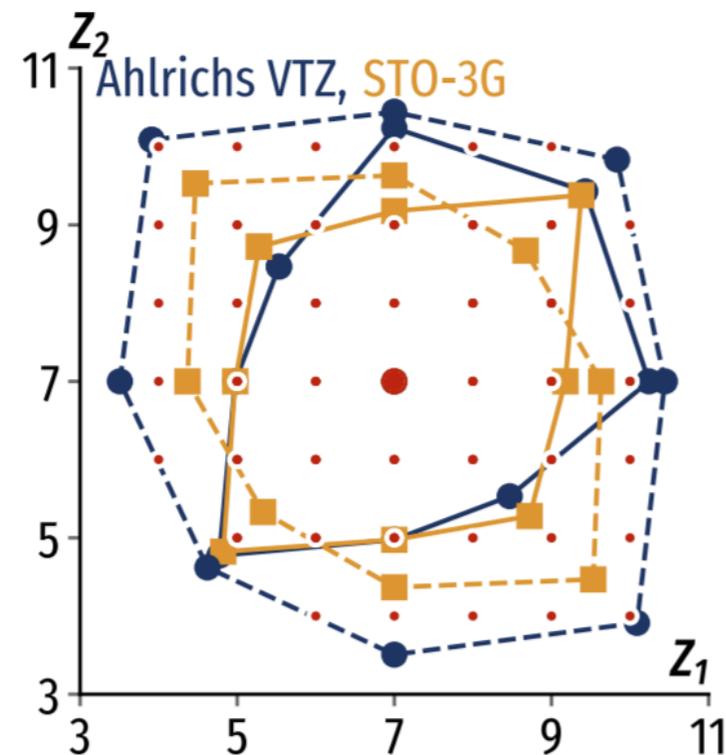


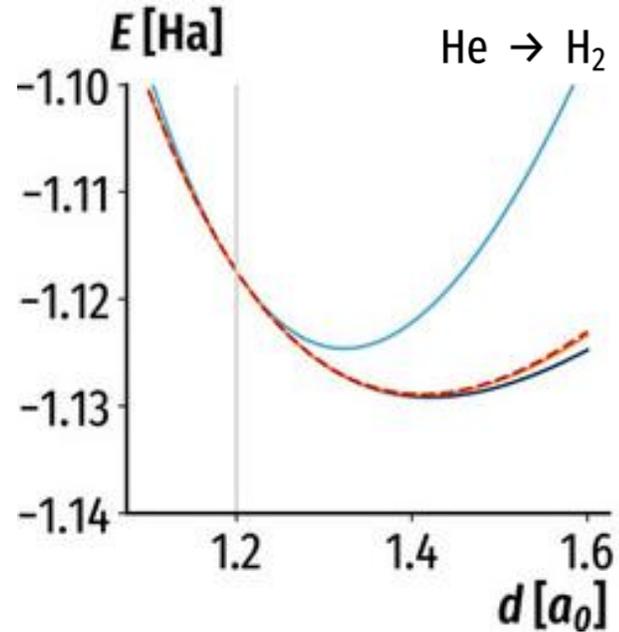
Taylor expansion

- First terms accurate enough
 - Truncate early
- Converges to the right value
- Large convergence radius
- Scales with chemical space

Arbitrary precision Hartree-Fock

 NablaChem/APHF



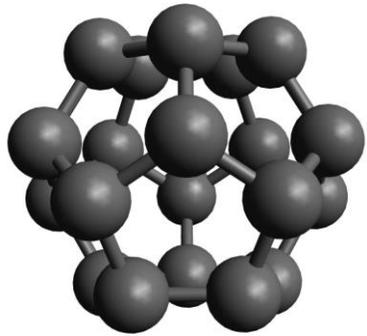


Taylor expansion

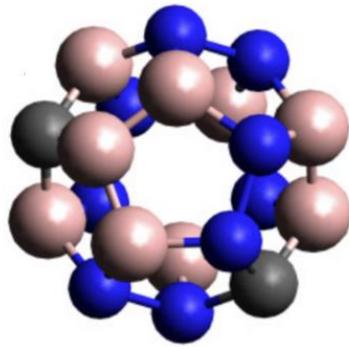
- Large changes still converge (more slowly)
- Geometric response can be recovered

Scaling with chemical space

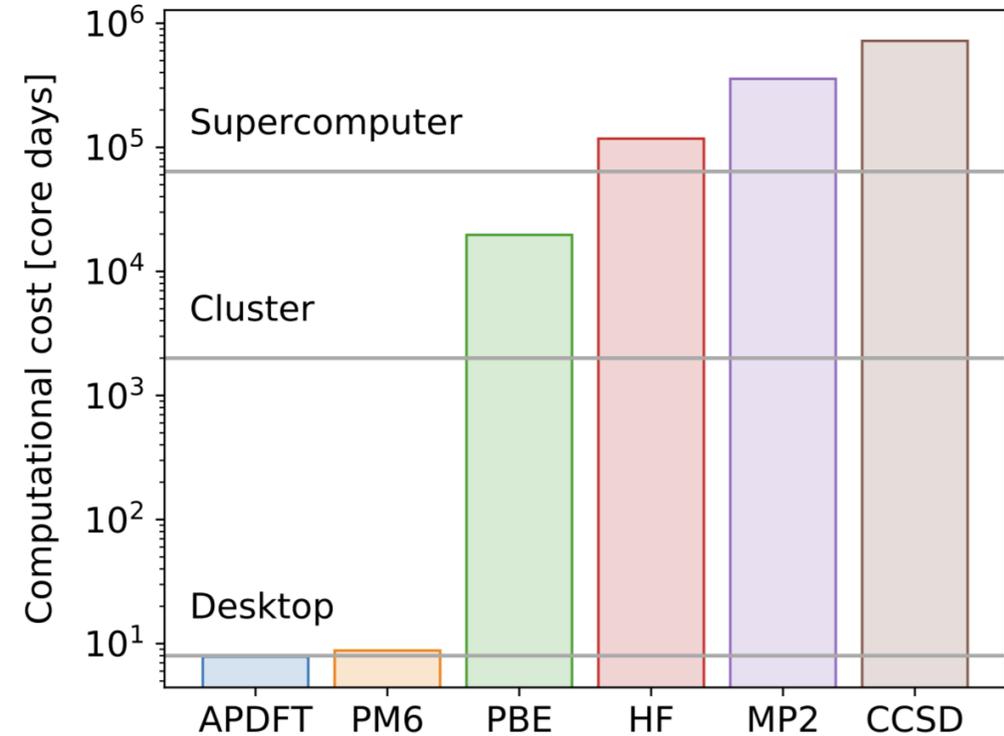
- 1 derivative for second order
- 5 derivatives for third order



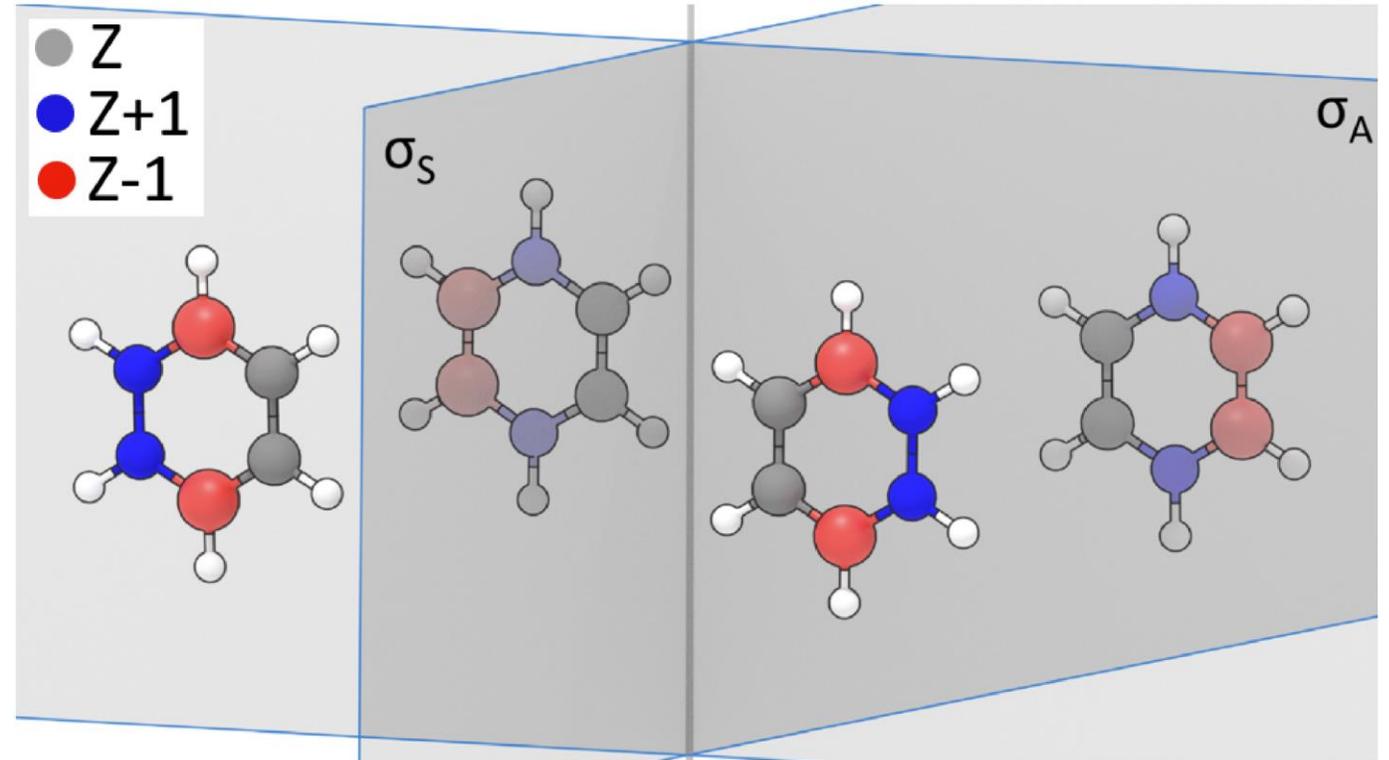
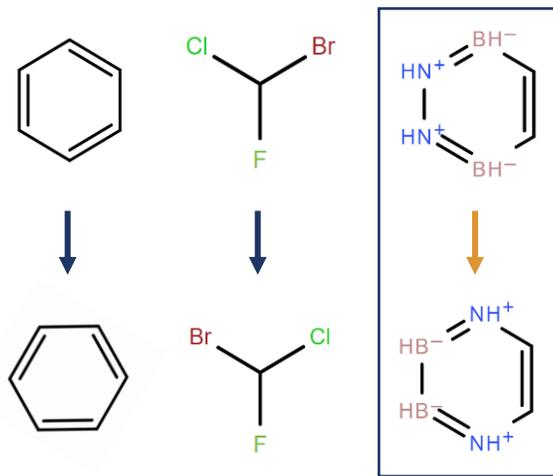
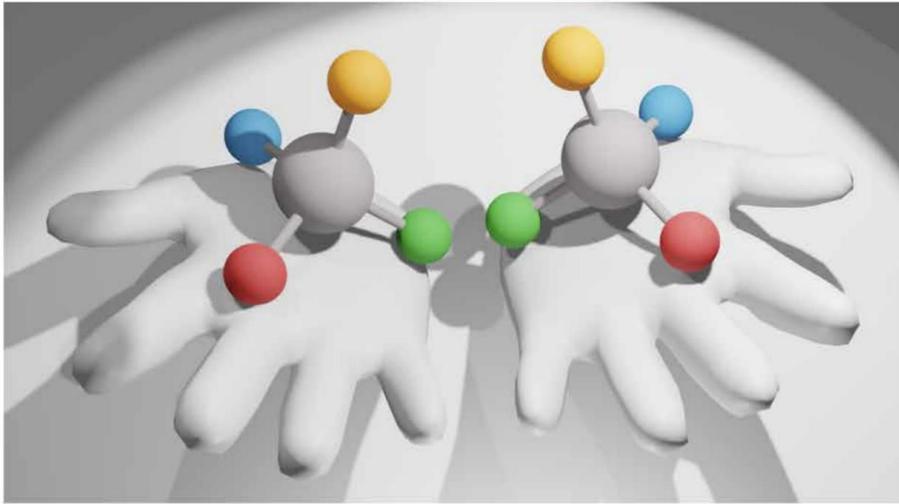
C_{20}



$3.1 \cdot 10^6$
targets

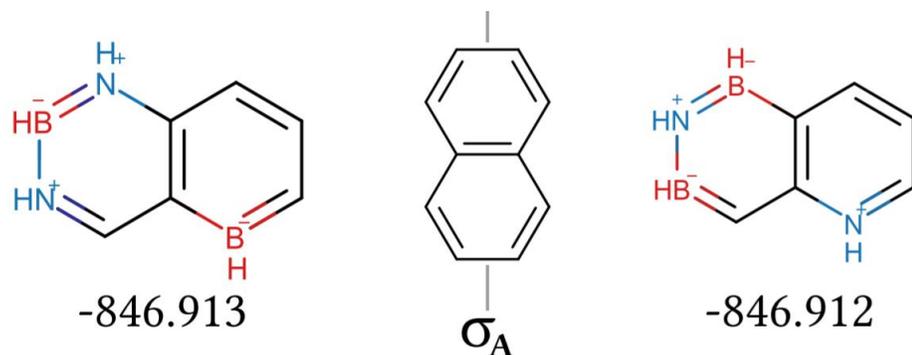


QA: 80.000x faster



Fundamentally new symmetry

Electronic energy only



Bond energy rules

Consecutive Elements

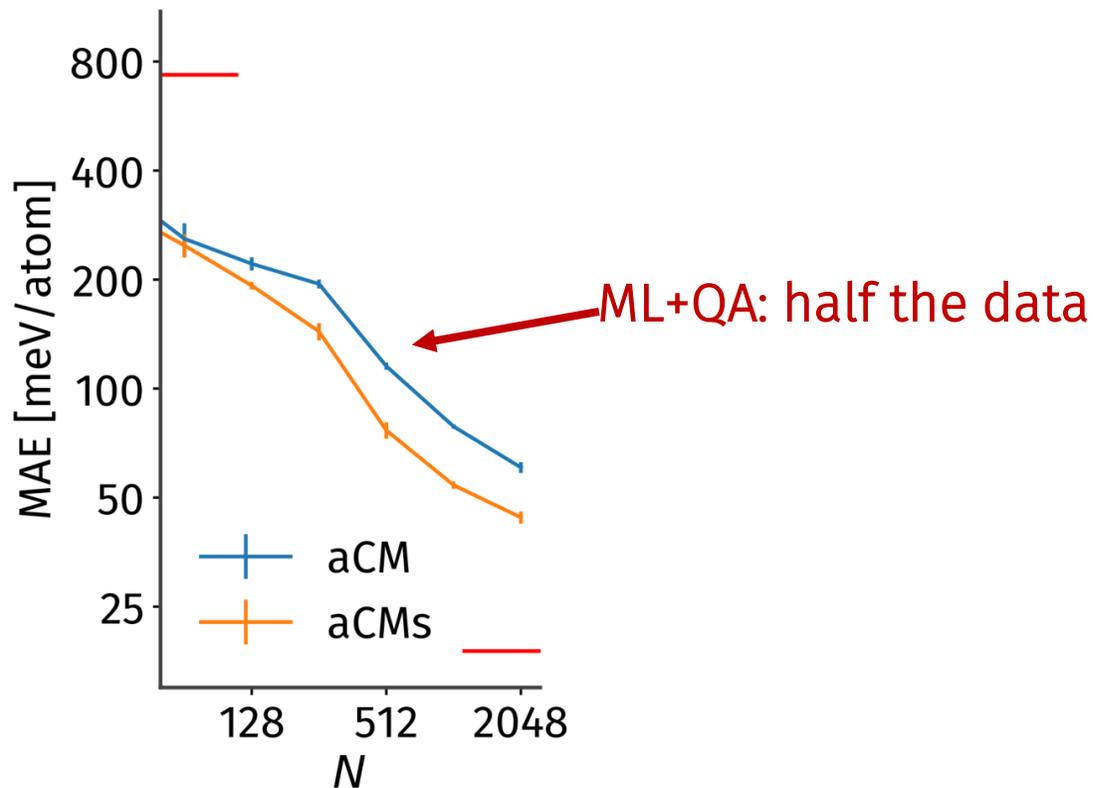
Q R S

B C N

$$E_{QR} \simeq E_{SR} + 0.5(E_{QQ} - E_{SS})$$

+ more rules for angles

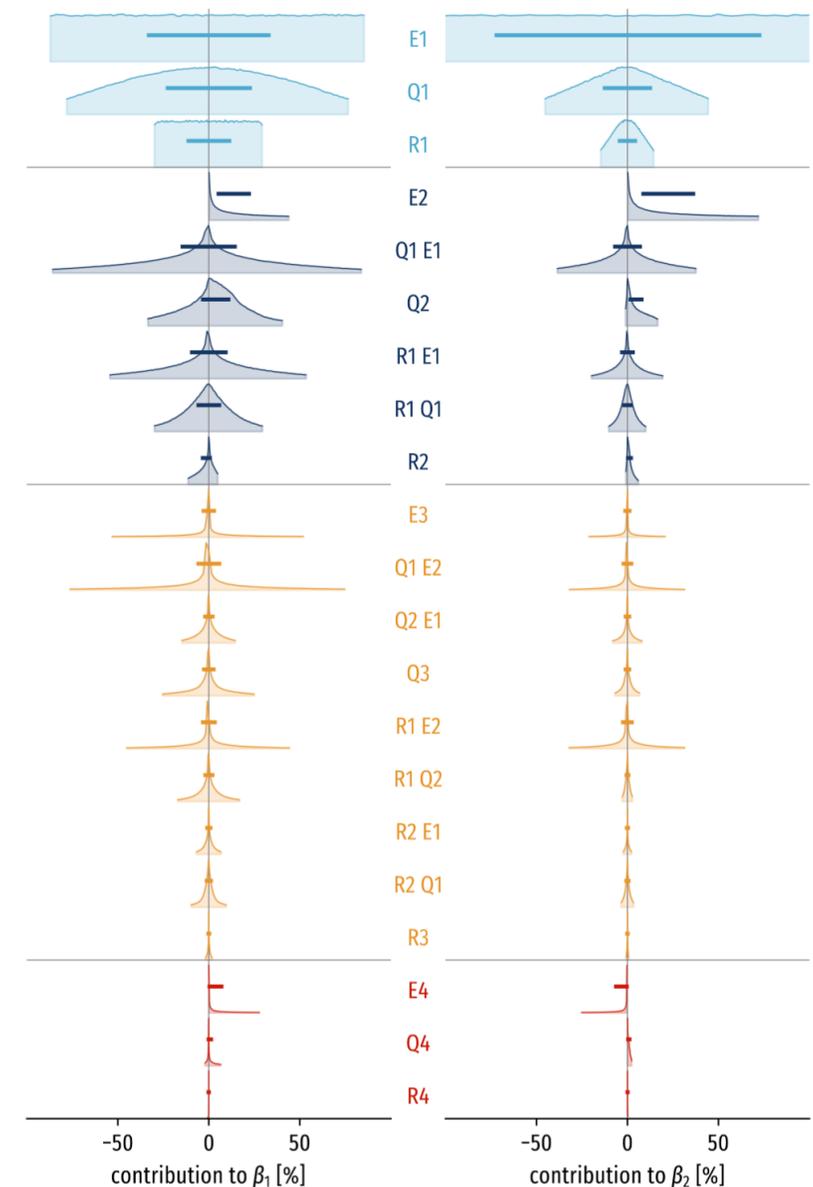
Speed up machine learning



Angular emission

- Expensive to calculate
- Highly coupled degrees of freedom: multidimensional expansion

$$\frac{d\sigma^\pm}{d\Omega} = \frac{\sigma}{4\pi} \left[1 \pm \underbrace{\beta_1 P_1(\cos \theta)}_{\text{dichroic parameter}} - \frac{1}{2} \underbrace{\beta_2 P_2(\cos \theta)}_{\text{anisotropy parameter}} \right]$$



Definition Intrinsic Dimension

Minimal number of degrees of freedom to describe a property.

≠ intrinsic dimension of a point cloud!

$$f(x, y) = x + y$$

Example: Dimers

- Energy: 3 dimensions
- Net charge: 1 dimension

Example: Atom

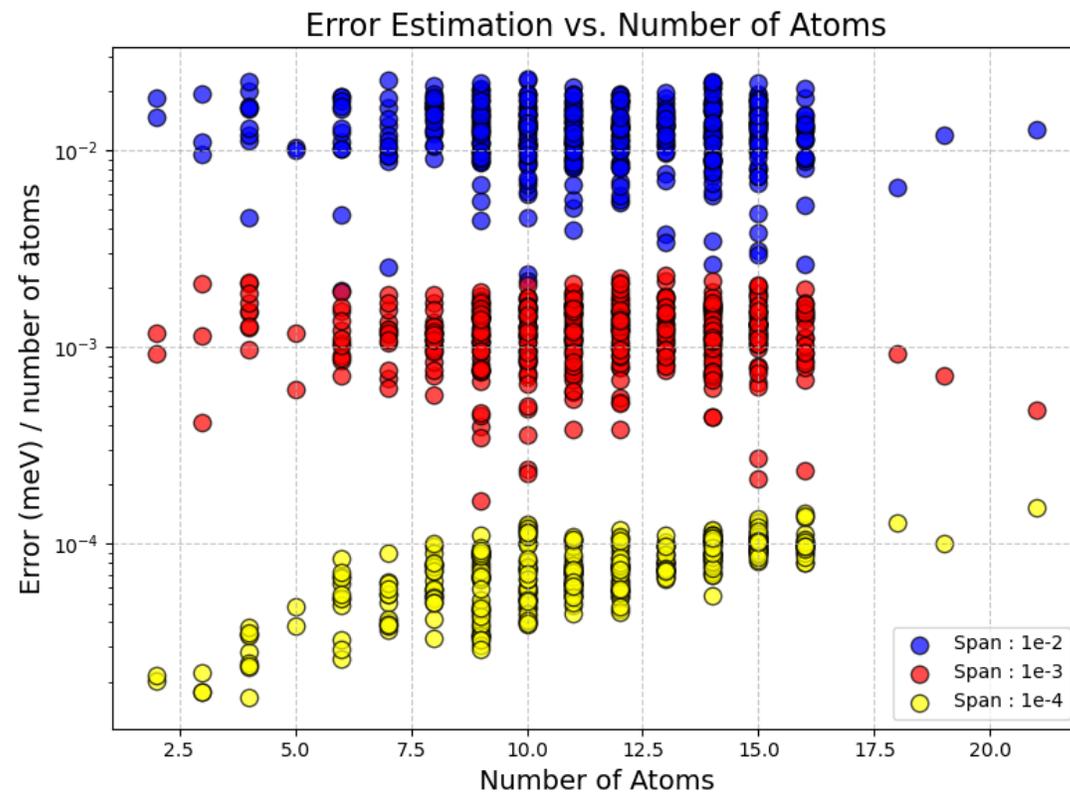
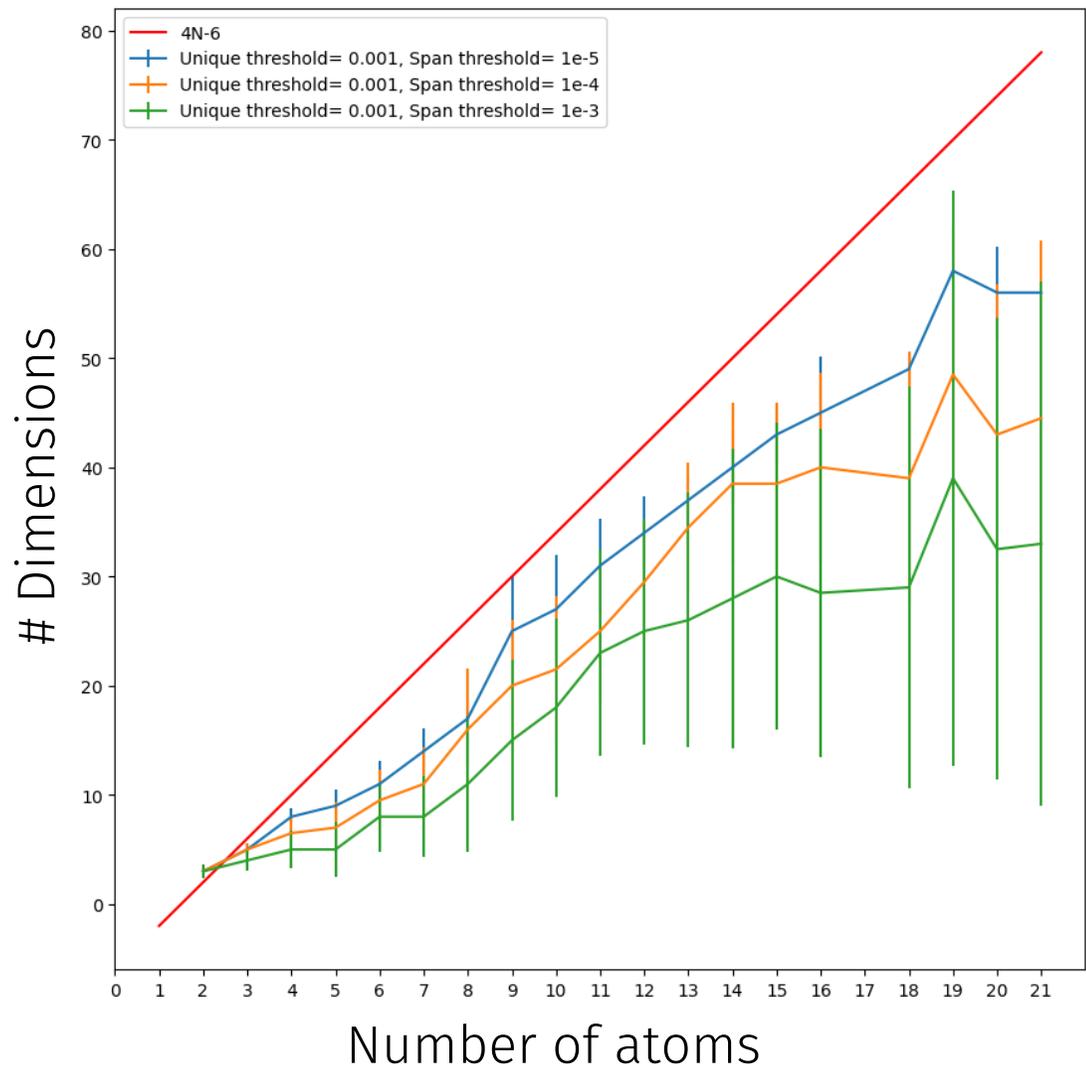
- 1 dimension
- 1 dimension

$$E(\mathbf{R}_I, \mathbf{Z}_I)$$

Distance Net charge Z_1+Z_2
Asymmetry Z_1-Z_2



Ali Banjafar



Both energy and density derivatives are hard

- Finite differences expensive, numerical instabilities
- Hellmann-Feynman finite order, basis set inaccurate
- Coupled-perturbed finite order, tedious
- Automatic differentiation niche: DiffiQult, quax, dqc, ...

 ferchault/APDFT

 NablaChem/APHF

 aspuru-guzik-group/DiffiQult

 CCQC/Quax

 diffqc/dqc **Alchemy!**

 fishjojo/pyscfad

Gaussian basis sets not overly cooperative

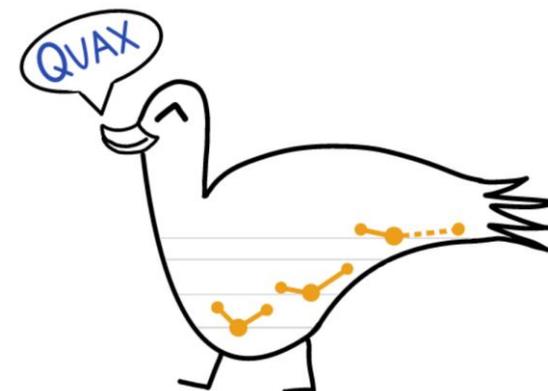
- Elements are discrete, derivatives are not
- Density converges more slowly with basis set quality than energy
- Unless complete basis set limit: Pulay terms

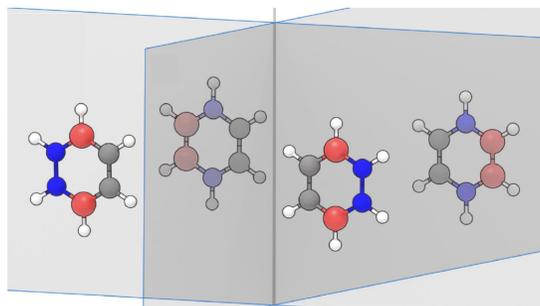
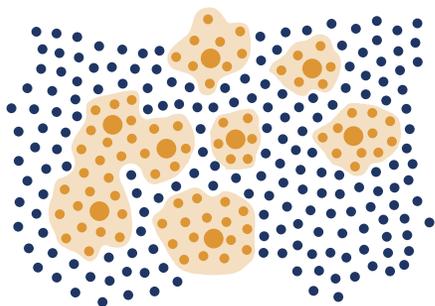
Pseudopotentials

- Discrete Elements, parameters not differentiable

Convergence

- Finite radius
- Not all systems are made equally





Efficient | Re-use knowledge, no one-by-one

Symmetries | Reducing (“folding”) search space

Physics-driven | Predictive power, hard constraint

Differentiable Chemistry | Arbitrary derivatives of many properties

Closed-Form | Explainable and shows structure

Fast Design | Can assess whole regions at once

Thanks

Anton Artemyev

Ali Banjafar

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Philipp Demekhin

Giorgio Domenichini

Emily Eikey

Chasz Griego

Nicolas Grimblat

John Keith

Simon Krug

Boris Lagutin

Anatole von Lilienfeld

Alex Maldonado

Michael Sahre

Relate to cost

- Expensive terms also irrelevant
- For molecules, subset only
- Allows for efficient stencil design

Estimated speedup

$$\frac{3^N}{N^2 + 12N - 1}$$

