



Incremental Learning for 'more challenging' systems

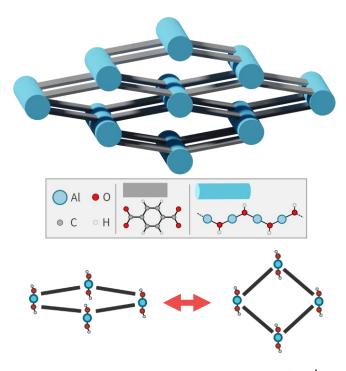
Sander Vandenhaute

advisor: Veronique Van Speybroeck





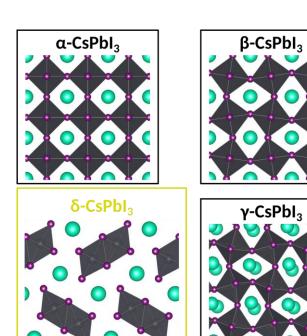
activated processes are everywhere



nanoporous materials

collective variable:

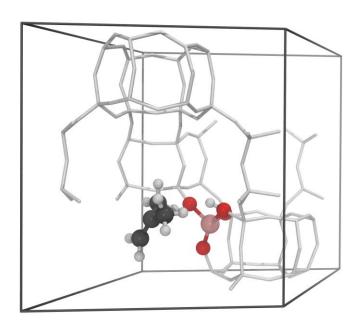
pore volume



perovskites

collective variable:

strain tensor



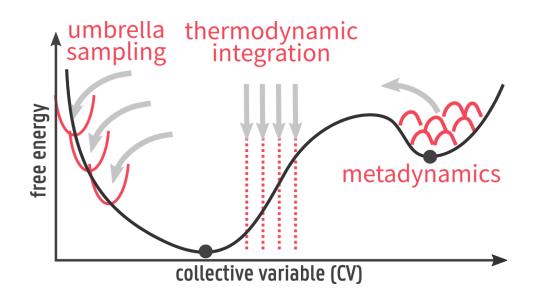
zeolite catalysis

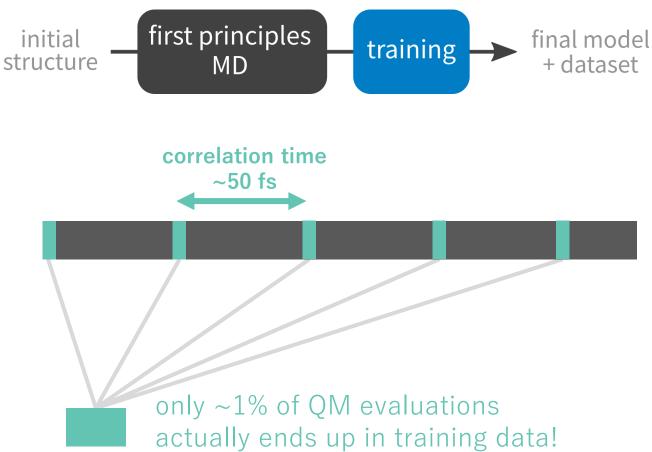
collective variable:

coordination numbers

the key challenge: data generation!

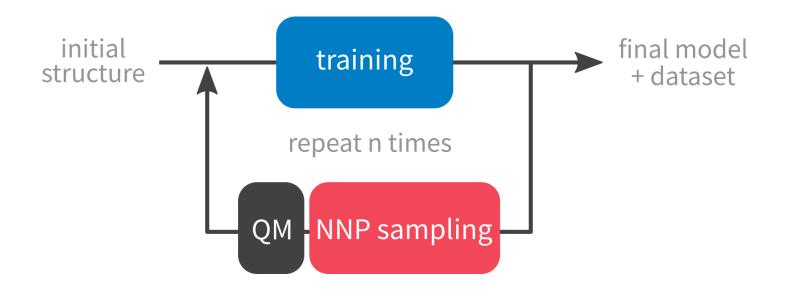
first principles MD is demonstrably inefficient



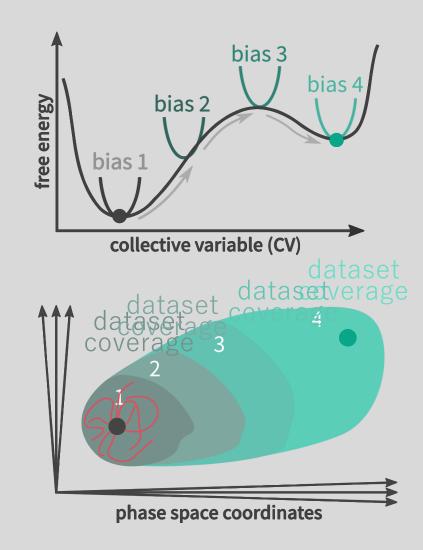


"incremental learning"

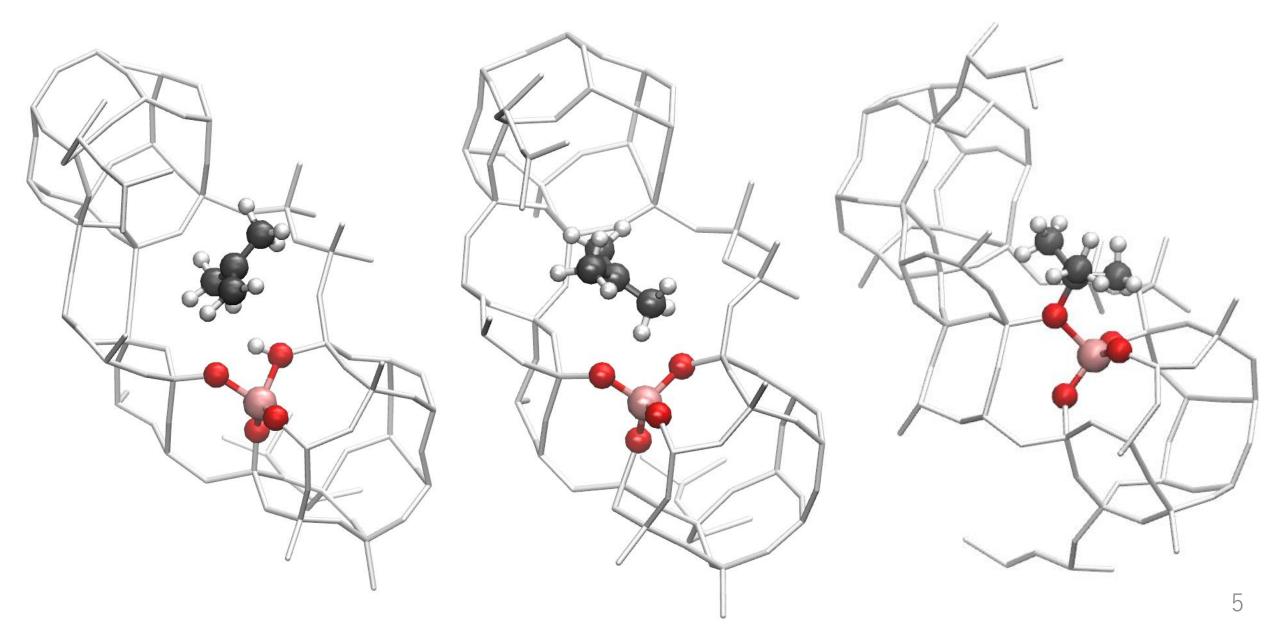
interleave training with short (multiple-walker) sampling



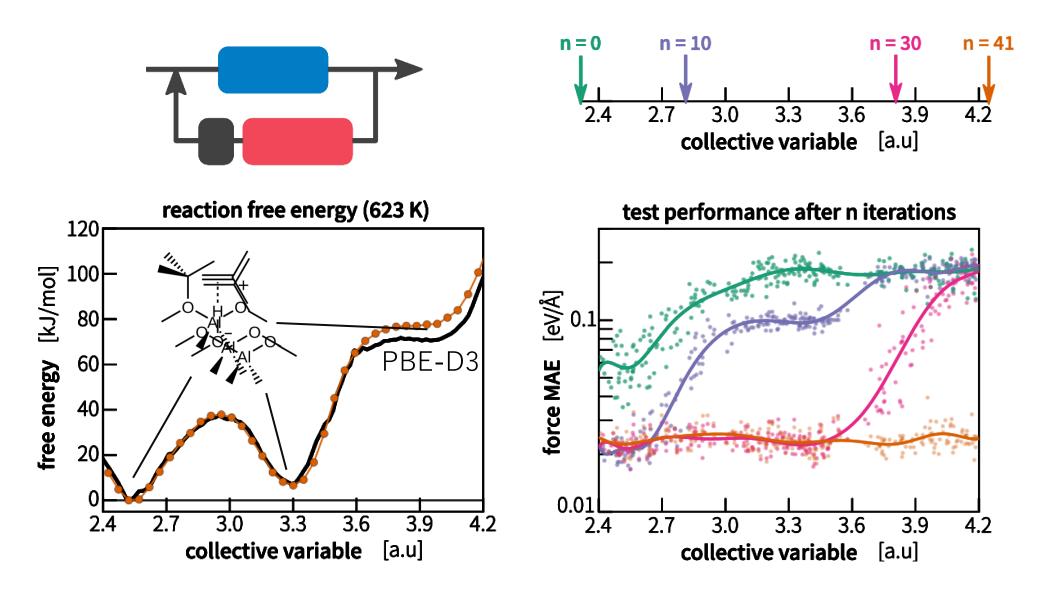




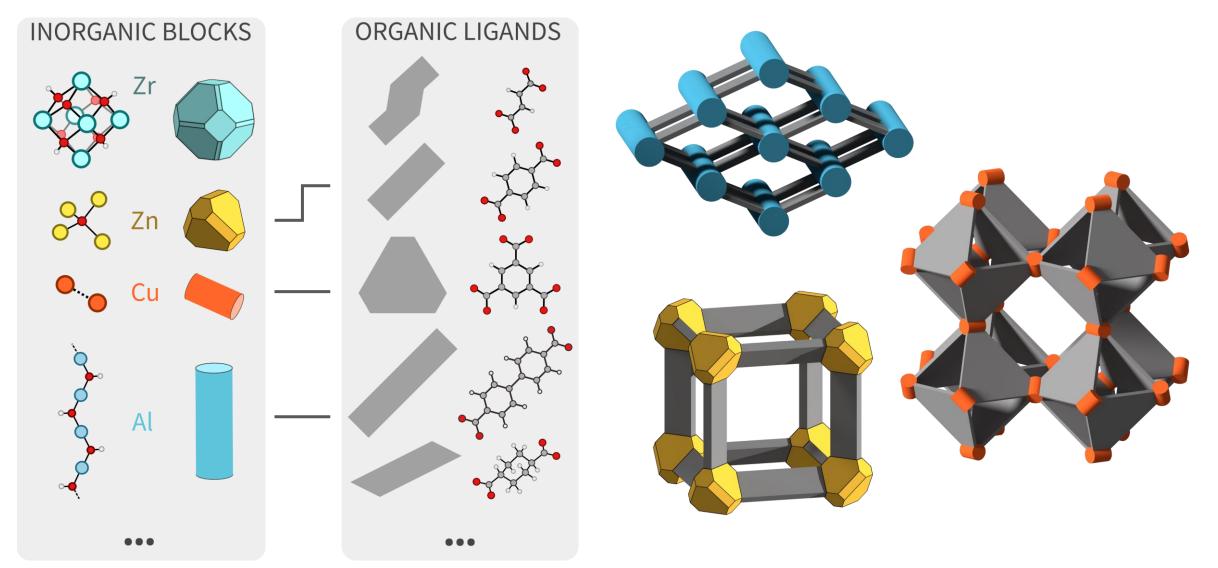
isobutene <> carbenium ion <> surface alkoxide



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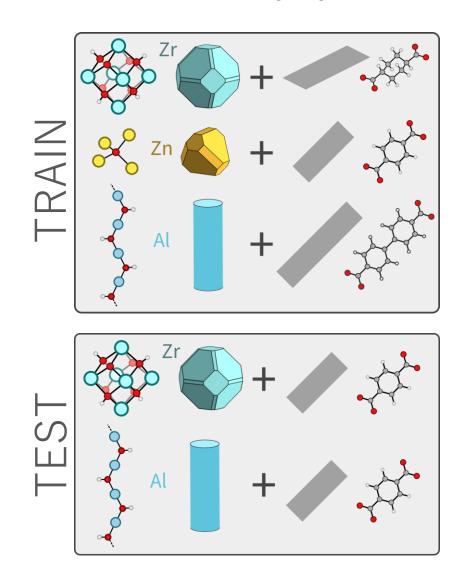


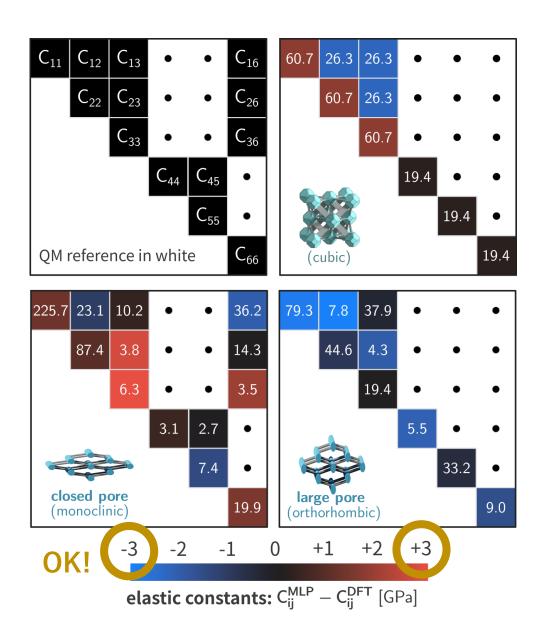
incremental learning for multiple materials?



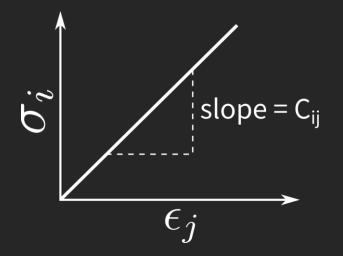
transferability towards 'unseen' combinations!

example: mechanical properties





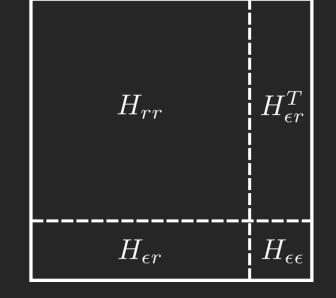
... exploit automatic differentiation!



$$C_{ij} \sim \left(\frac{\partial^2 U}{\partial \epsilon_i \partial \epsilon_j} \right)_{\substack{ ext{optimized positions}}}$$

$$C \sim H_{\epsilon\epsilon} - H_{\epsilon r} H_{rr}^{-1} H_{\epsilon r}^T$$

extended hessian:

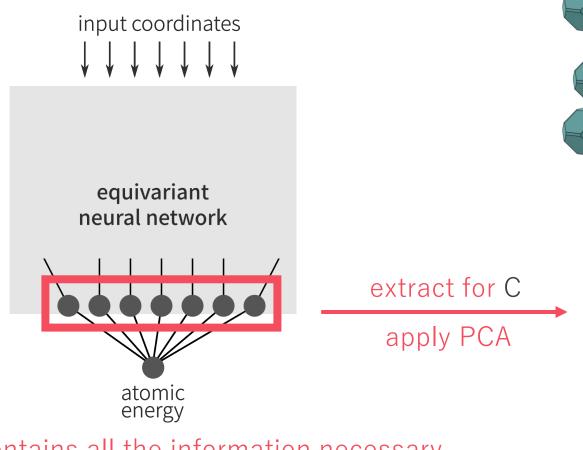




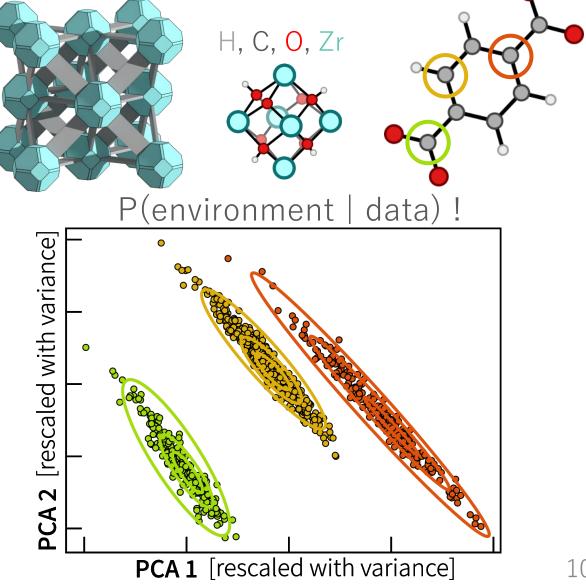
achieves double precision!

can we easily predict transferability?

analyze final output layer!



contains all the information necessary to predict the atomic energy, in a low-dimensional space! (d~10)



thank you!



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