

# Accurate structural optimizations in solids with auxiliary-field quantum Monte Carlo

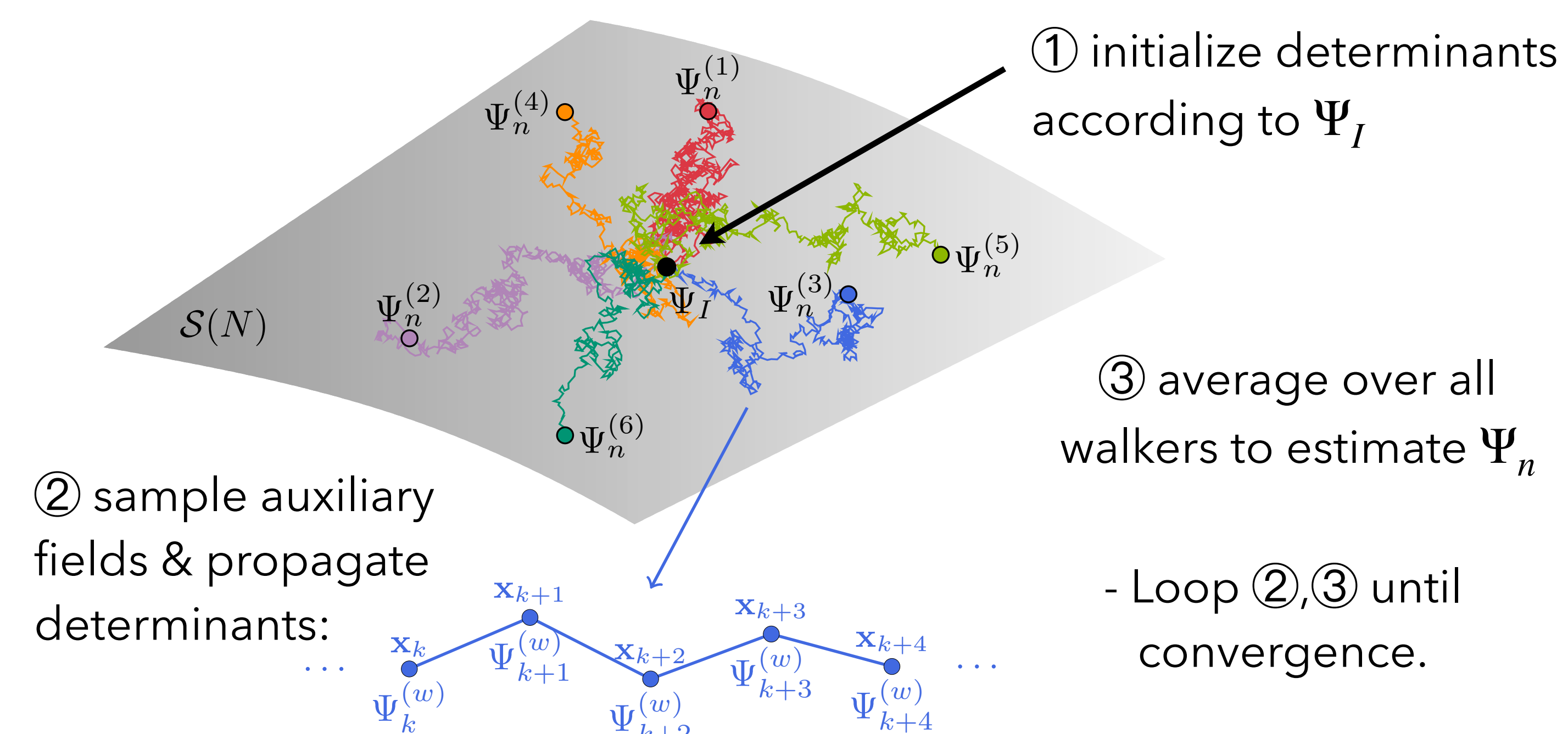
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## Abstract

Structural optimization by accurate, non-perturbative methods has been an outstanding challenge in many-body electronic structure computations. We present direct computation of forces and stresses in solids by plane-wave AFQMC. With them, we perform full structural optimizations in several solids. Additionally, we propose a general optimization algorithm, FSSD×SET, for gradients which have intrinsic stochastic noise. This algorithm is found to outperform standard optimization methods and several machine learning algorithms in efficiency and robustness.

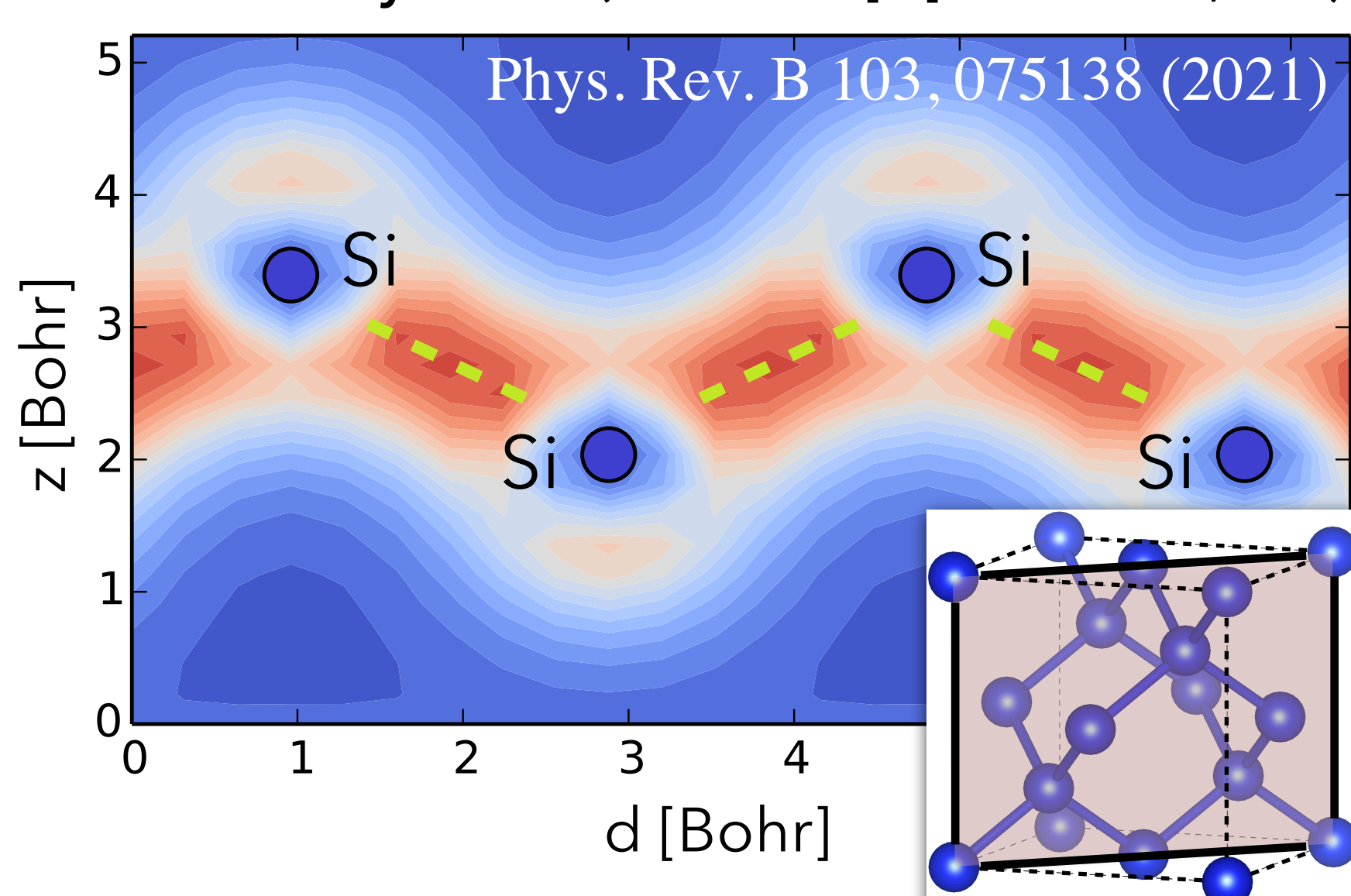
## Auxiliary-field QMC

AFQMC [1-3] approaches the ground state of a system with imaginary time propagation  $e^{-\beta H} |\Psi_I\rangle \rightarrow |\Psi_{GS}\rangle$ . With the introduction of auxiliary fields, two-body propagations transform into integrals of one-body propagations, and can be sampled with a random walk.

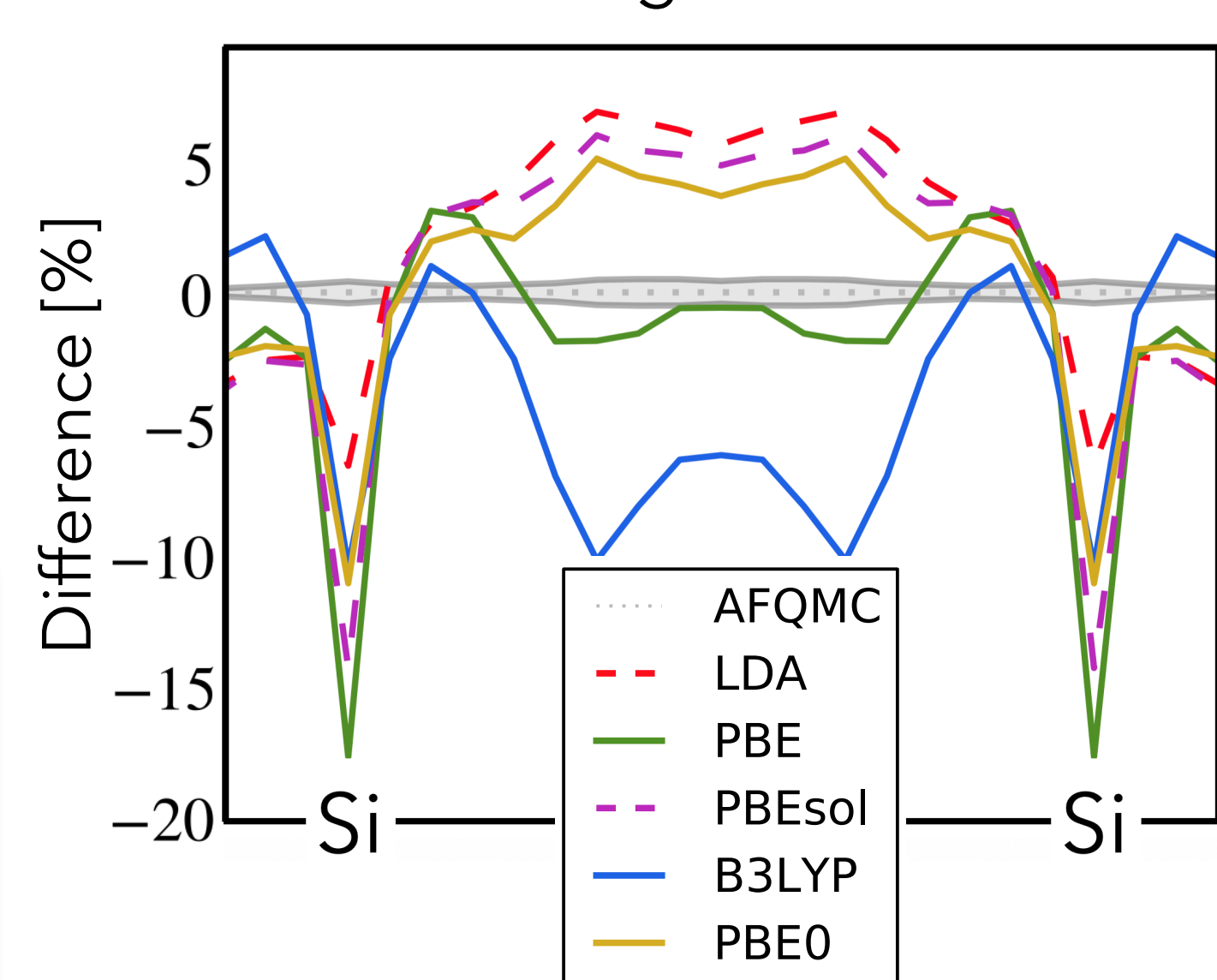


AFQMC has demonstrated excellent accuracy in lattice models, molecules, and solids.

Density of Si (see Ref. [4] for NaCl, Cu)



Benchmarking DFT functionals



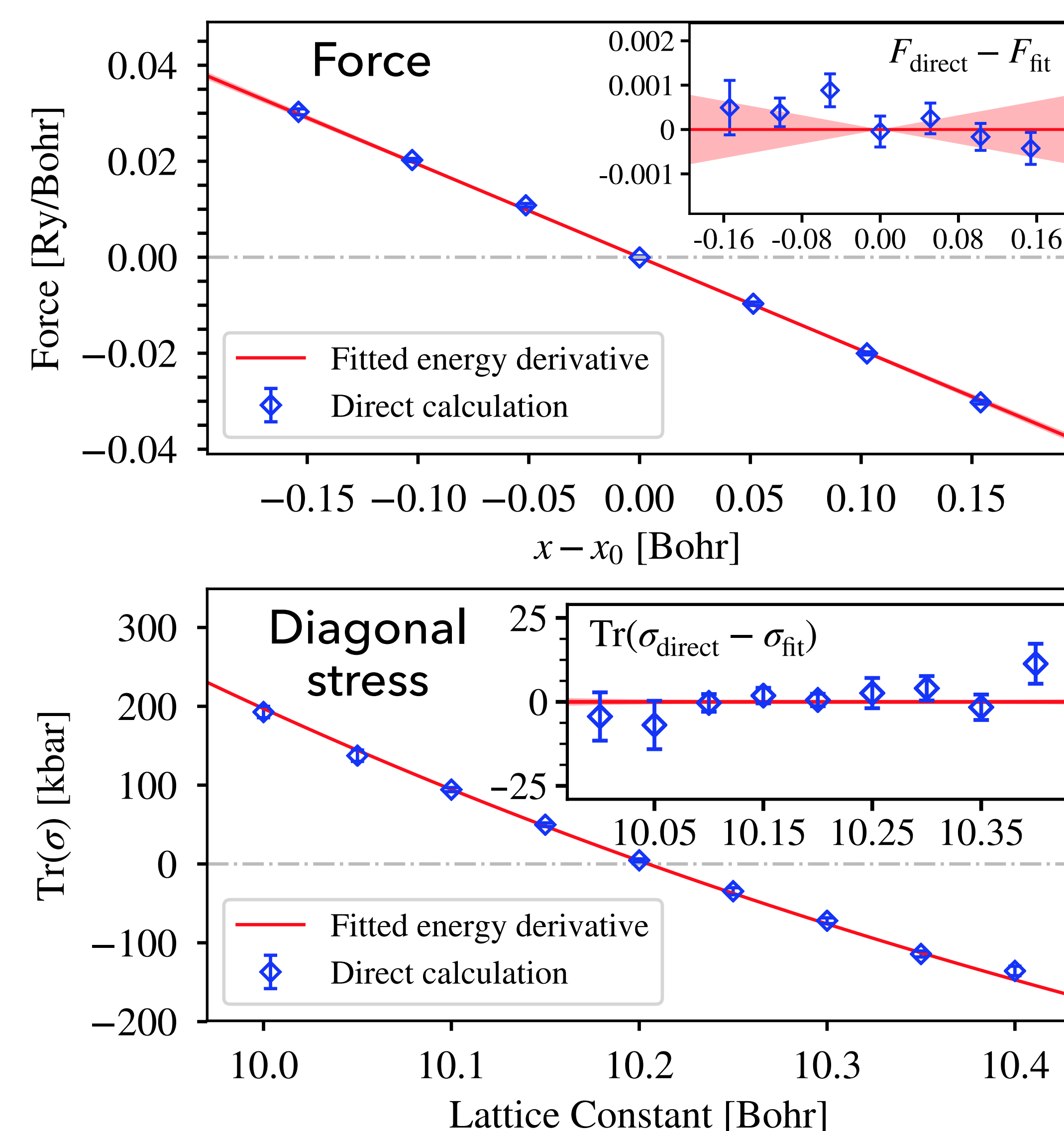
## Forces and Stresses

Computation of forces is a kind of “holy grail” in QMC. We present forces and stresses computed directly from BP + Hellmann-Feynman in plane-wave AFQMC [5].

Here we compare:

- ① directly computed force/stress
- ② fitting AFQMC equation-of-state and obtaining its derivative

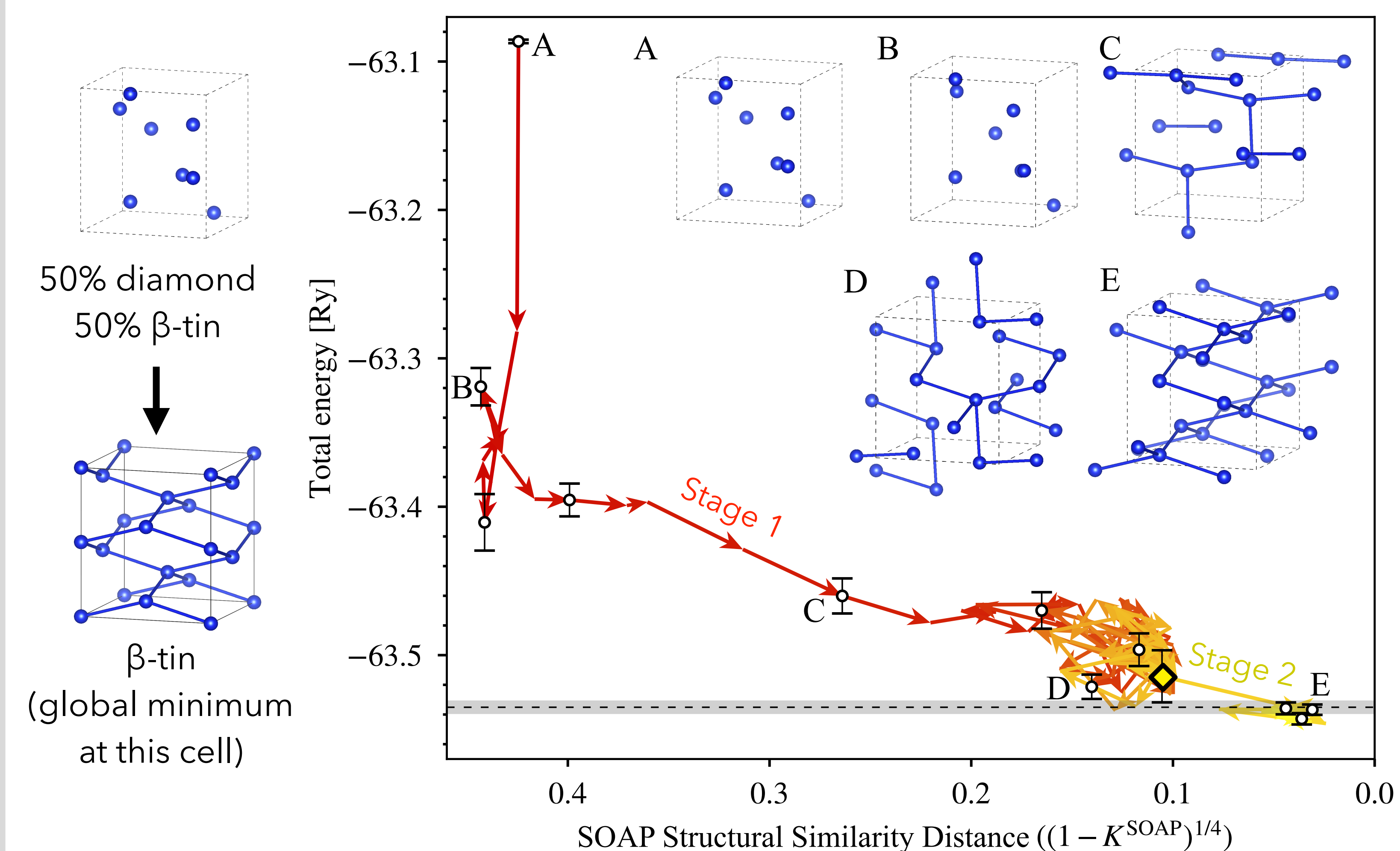
Excellent agreement →



## Structural Optimization with AFQMC

With AFQMC forces/stresses and our new optimization algorithm, we perform two direct AFQMC structural optimizations [5].

### ① Atom-geometry optimization: Si (diamond → $\beta$ -tin)



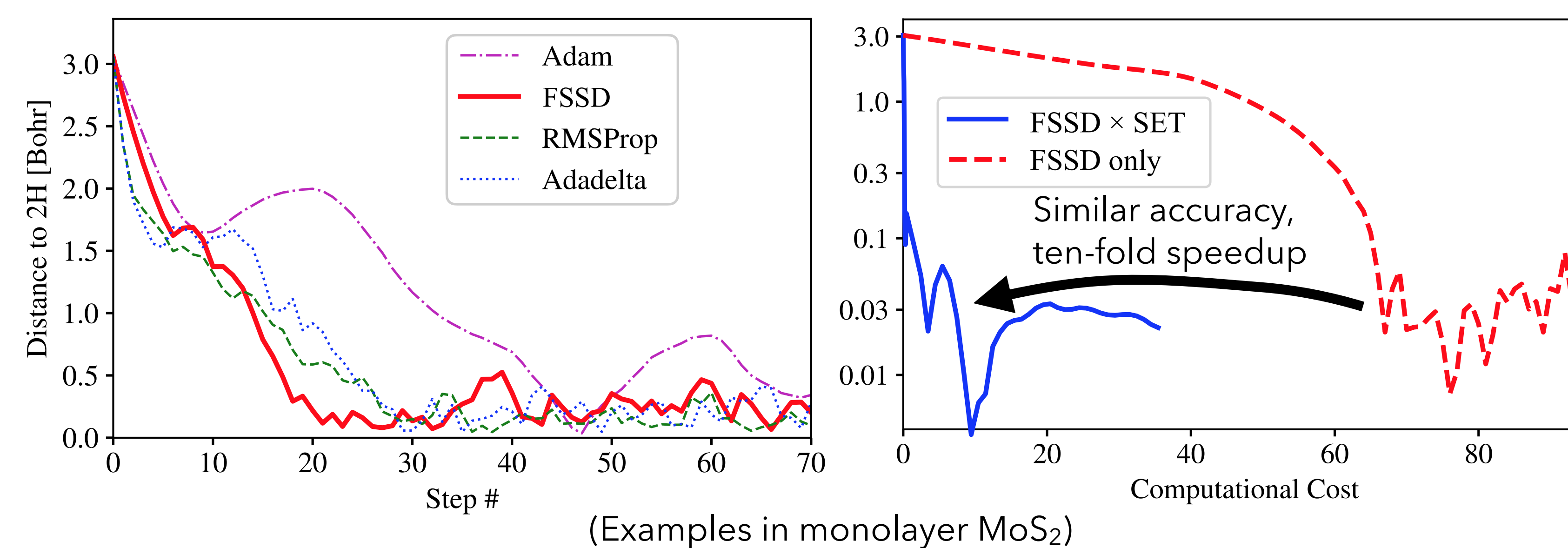
## FSSD×SET

AFQMC forces/stresses contain statistical uncertainty. What’s the best way to perform structural optimizations in the presence of noisy forces?

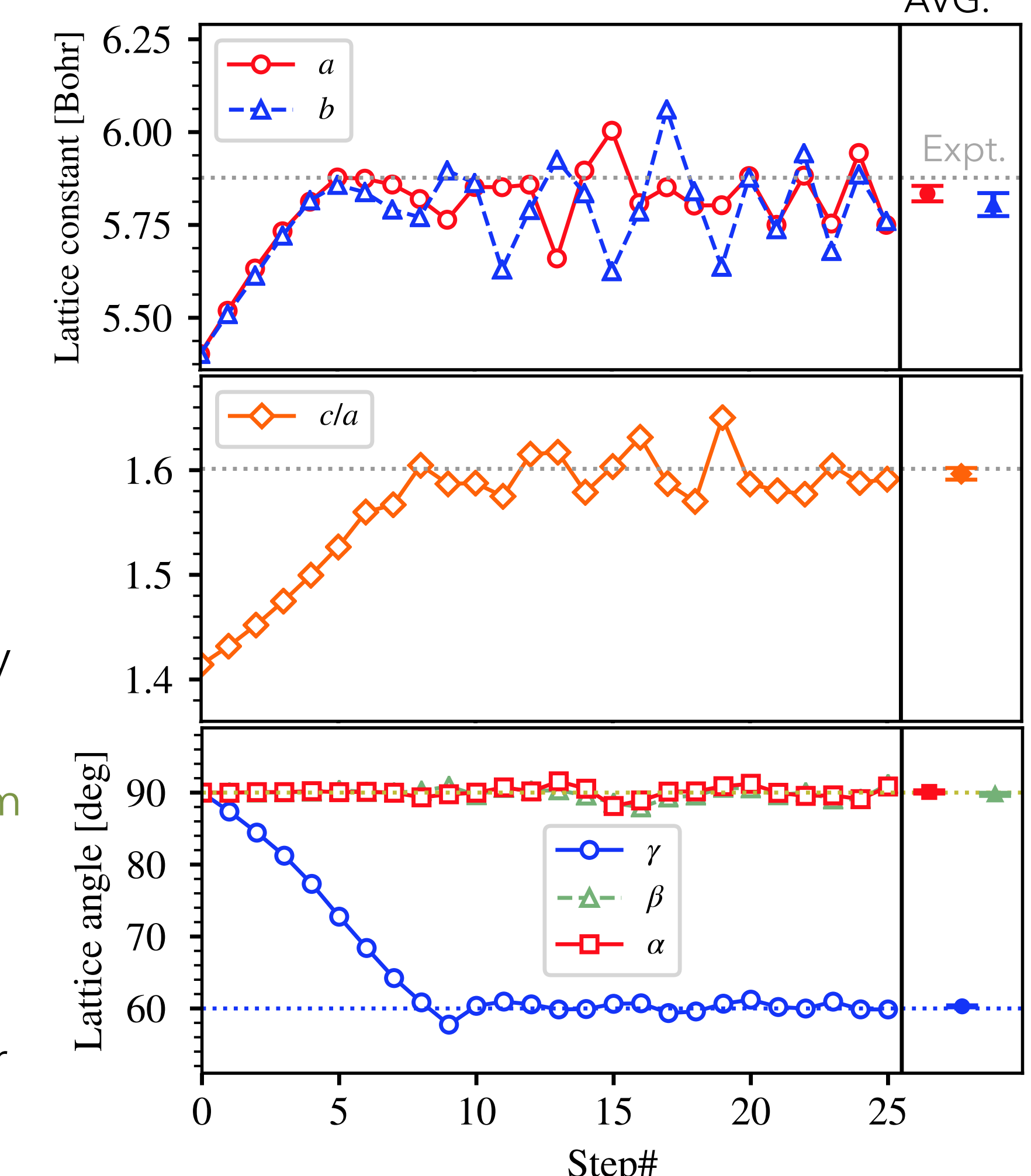
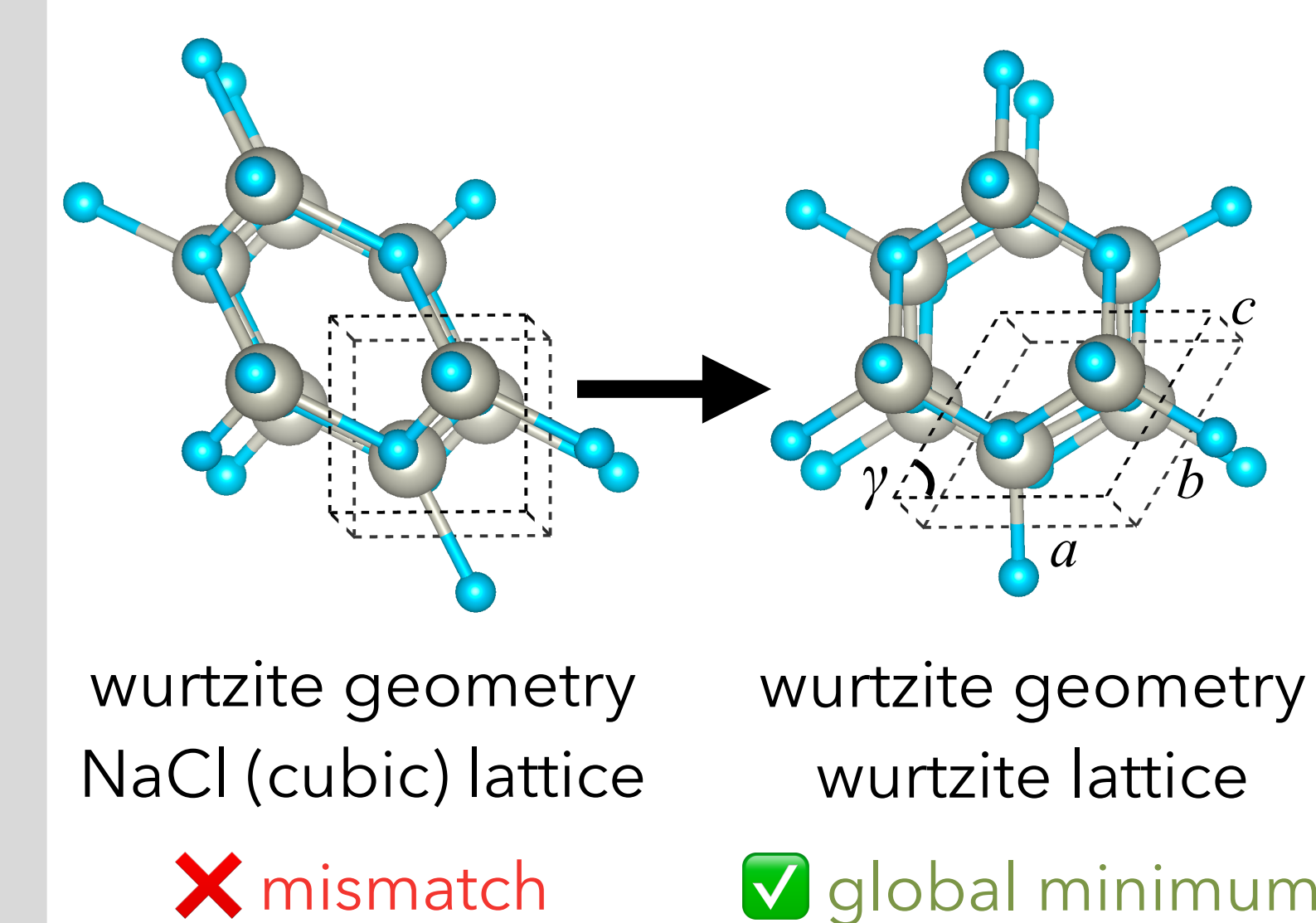
We propose a new algorithm called FSSD×SET ([arXiv:2204.12074](https://arxiv.org/abs/2204.12074) [6]):

- **Update rule** “Fixed Step-size Steepest Descent” (FSSD),
- **Scheduling workflow** “Staged error targeting” (SET).

FSSD outperforms several common machine learning algorithms, while SET provides an additional boost.



### ② Lattice optimization: AlN



\*AVG. = average of lattice parameters after convergence (required by SET)

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6. S. Chen and S. Zhang, [arXiv:2204.12074](https://arxiv.org/abs/2204.12074) (2022).