

Electronic structure software development: something old & something new

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Overview

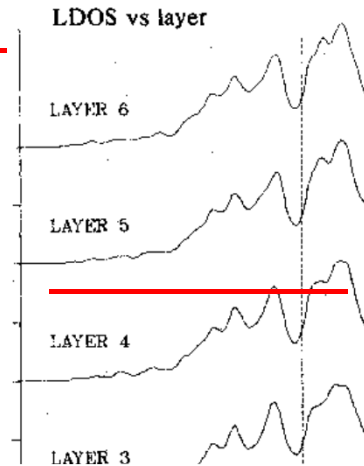
1. HSE functional implementation in SeqQuest
2. Some thoughts on scientific software development
3. Localized & randomized linear-scaling electronic structure

History of SeqQuest

I. INTRODUCTION

It has been shown that calculation of the surface structure of d -band materials is within present day computational techniques therefore timely to begin a study of effects of surface electronic properties across the periodic table. Among the 3d metals self-consistent surface-electronic-structure calculations performed only for Cu,¹ and Ni.² (For the transition metals Nb,³ Mo,⁴ Pd,⁵ and Ag.⁶) We present here the results of a self-consistent calculation of the electronic properties of a Ti(0001) film.

The method is based on a linear combination of atomic orbitals approach, which permits



[Phys. Rev. B **20**, 1433 (1979)]

Mix of physics & chemistry methods:

- spd Gaussian atomic orbitals (DZP)
- ECP pseudopotentials (large core)
- local Δ -potentials on a uniform grid
- semilocal DFT (no Fock exchange)

A systematic first-principles pseudopotential plane-wave and linear combination of atomic orbitals basis-set convergence study has been performed for a 64-atom amorphous "tetrahedral" carbon network (Drabold *et al.*). Converged results of the two methods are consistent with each other and indicate that when large, site-dependent bond-angle and bond-length distortions are present, variational freedom within a basis-set representation is crucial for an accurate representation of atomic forces. Minimal basis-set/non-self-consistent methods are not adequate for these systems, but are more applicable to systems with minimal site-dependent, electron-density distortions.

I. INTRODUCTION

The rich variety of electronic, optical, and structural properties exhibited by "diamondlike" or tetrahedral amorphous carbon (a -tC) materials¹ is intimately related to the richness in bonding displayed by carbon. Qualitatively, the ratio of sp^3 - to sp^2 -like atoms and the connectivity are the main factors in determining the properties of an individual material. Neutron scattering² is often used to quantify deviations from ideal tetrahedral and graphitic bonding through differences in the radial distribution function compared with that of diamond and

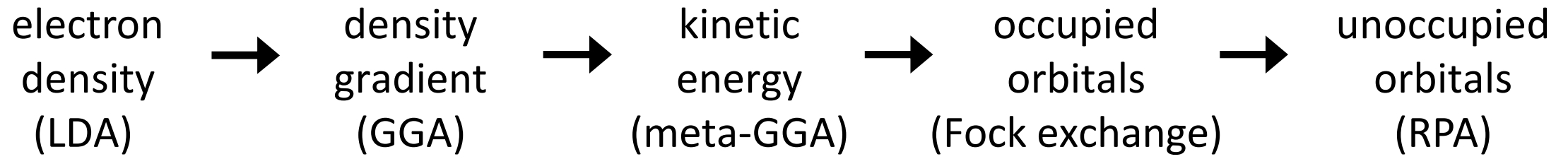
TBMD (Ref. 3) at the same density. Wang and Ho³ point out that the TBMD model gives a better description than LBDF (Ref. 5) of the relative energetics of the graphite and diamond phases, although the energy differences are small and may not be of differentiating significance for topological models of a -tC. The LBDF (Ref. 5) predicts the diamond structure to be lower in energy than the graphite structure in contrast with TBMD,³ but consistent with fully self-consistent density-functional calculations. Drabold, Stumm, and Fedders argue the TBMD is not universally transferable, whereas LBDF,⁴ based on the Harris version of density-functional theory,⁷ has added

[Phys. Rev. B **52**, 9254 (1995)]

- <http://dft.sandia.gov/Quest> & <http://nanoHUB.org>
- **primary uses:** surface science & point-defect chemistry
- 100k lines of Fortran 77, 4k lines for main control flow
- MPI parallelization w/ few-node scalability

Heyd-Scuseria-Ernzerhof density functional

Perdew's "Jacob's ladder" of universal density functionals [Perdew & Schmidt, AIP Conf. Proc. **577**, 1 (2001)]:



↓
exchange screening
($\approx 1/\epsilon$ -fraction hybrids)

↙
d-shells in solid state
(short-ranged hybrids)

↘
molecules in vacuum
(long-ranged hybrids)

PBE/HSE exchange-hole model:

[Heyd, Scuseria, & Ernzerhof, J. Chem. Phys. **118**, 8207 (2003)]

$$E_x = \int d\mathbf{r} \rho(\mathbf{r}, \mathbf{r}) \epsilon_x(\mathbf{r}, \dots)$$

$$= \int d\mathbf{r} d\mathbf{r}' \rho(\mathbf{r}, \mathbf{r})^2 J(\mathbf{r}, |\mathbf{r} - \mathbf{r}'|, \dots) V(|\mathbf{r} - \mathbf{r}'|)$$

$$= -\frac{1}{2} \int d\mathbf{r} d\mathbf{r}' \rho(\mathbf{r}, \mathbf{r}')^2 V(|\mathbf{r} - \mathbf{r}'|)$$

Fock exchange: physics vs. chemistry

implicit ← $h_x(\mathbf{r}, \mathbf{r}') = -\rho(\mathbf{r}, \mathbf{r}')V(|\mathbf{r} - \mathbf{r}'|)$ → explicit

$$\psi_{\text{out}}(\mathbf{r}) := \int d\mathbf{r}' h_x(\mathbf{r}, \mathbf{r}') \psi_{\text{in}}(\mathbf{r}')$$

$$H_{ij} := \int d\mathbf{r} d\mathbf{r}' \phi_i(\mathbf{r}) h_x(\mathbf{r}, \mathbf{r}') \phi_j(\mathbf{r}')$$

$$\rho(\mathbf{r}, \mathbf{r}') = \sum_{i \in \text{occ}} \psi_i(\mathbf{r}) \psi_j(\mathbf{r}')$$

$$\rho(\mathbf{r}, \mathbf{r}') = \sum_{i, j \in \text{basis}} \phi_i(\mathbf{r}) P_{ij} \phi_j(\mathbf{r}')$$

$$v_i(\mathbf{r}) = \int d\mathbf{r}' V(|\mathbf{r} - \mathbf{r}'|) \psi_i(\mathbf{r}') \psi_{\text{in}}(\mathbf{r}')$$

$$V_{ijkl} = \int d\mathbf{r} d\mathbf{r}' \phi_i(\mathbf{r}) \phi_j(\mathbf{r}) V(|\mathbf{r} - \mathbf{r}'|) \phi_k(\mathbf{r}') \phi_l(\mathbf{r}')$$

$$\phi_{\text{out}}(\mathbf{r}) = - \sum_{i \in \text{occ}} \phi_i(\mathbf{r}) v_i(\mathbf{r})$$

$$H_{ij} = - \sum_{k, l \in \text{basis}} V_{ikjl} P_{kl}$$

cost (per vector) $\propto N_{\text{occ}} N_{\text{grid}} \log(N_{\text{grid}})$

cost $\propto N_{\text{basis}} (n_{\text{overlap}})^2 n_{\text{kernel}}$

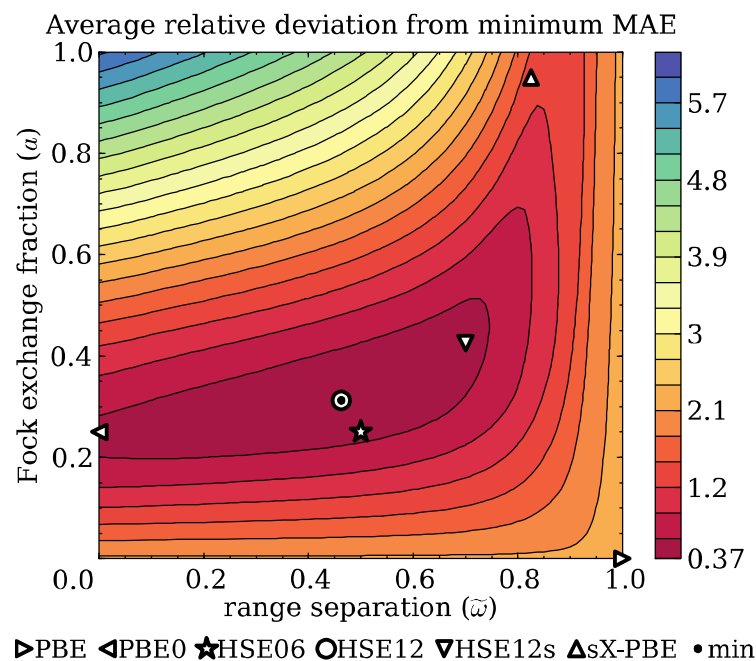
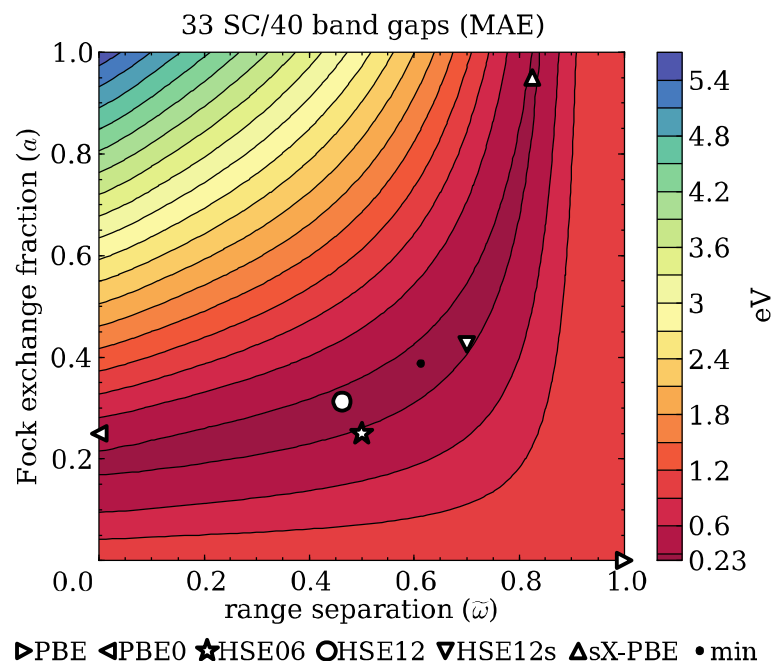
VASP, Quantum ESPRESSO, CASTEP, ...

GAUSSIAN, CRYSTAL, PySCF (solid state)

Recalibrating HSE

Simultaneously optimize HSE fraction (0.25) & range (9.1 bohr) to:

understand physical significance, reduce range/cost, & improve accuracy



tests: SC/40, G3/99, BH42/04, & T-96R

“HSE12”: (0.313, 10.2 bohr)

“HSE12s”: (0.425, 4.6 bohr)

min-range: (1.0, 2.5 bohr)
(narrower than atomic orbitals)

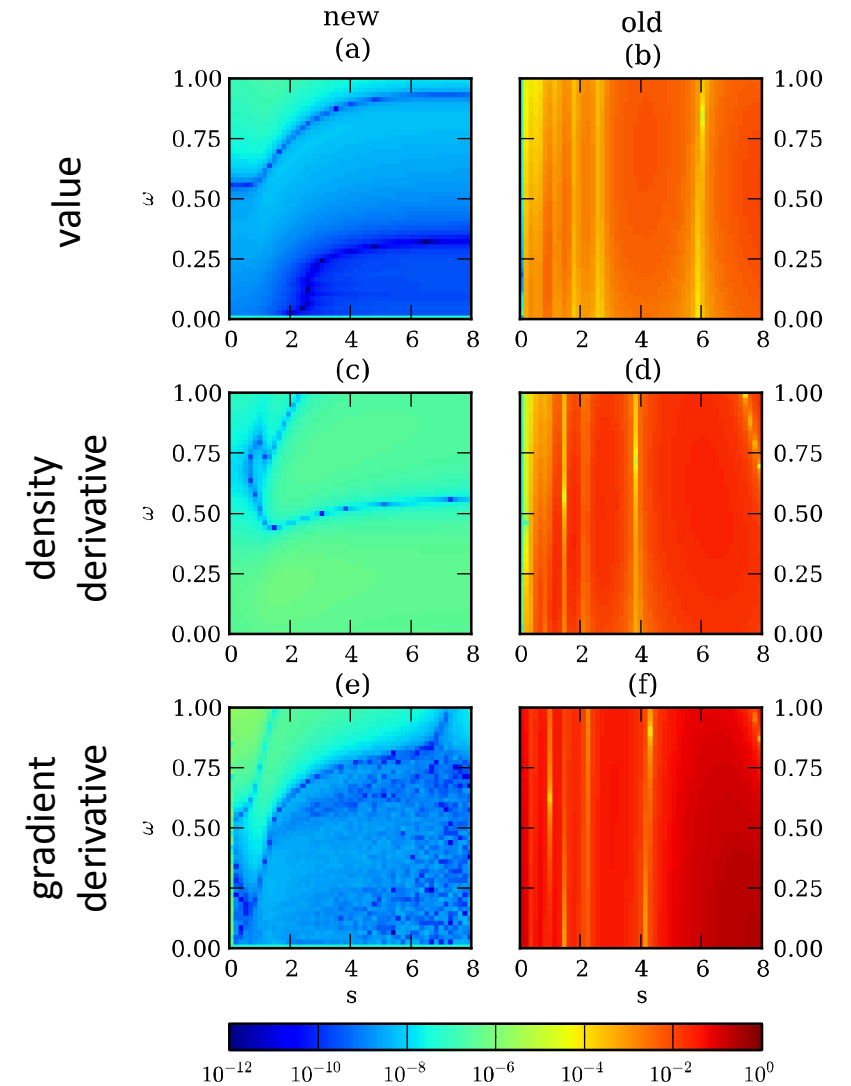
[Moussa, Shultz, & Chelikowsky, J. Chem. Phys. **136**, 204117 (2012)]

*Truhlar’s “Minnesota” density functionals constructed from a similar methodology

Refactoring HSE

$$\epsilon_x(\mathbf{r}, \dots) = \rho(\mathbf{r}, \mathbf{r}) \int d\mathbf{r}' J(\mathbf{r}, |\mathbf{r} - \mathbf{r}'|, \dots) V(|\mathbf{r} - \mathbf{r}'|)$$

- numerical & regularization problems in standard implementation
- replaced integration quadrature with semi-analytic evaluation
- refactored twice (we could afford to make this into a reliable black box)



[Moussa & Schultz, arXiv:1210.8233 (2012)]

Verifying HSE

- Coulomb integrals & derivatives: **Libint 2.3.1** w/ added C wrapper
- **no pre-computed** integrals (“direct”) or screening (envelop tests)
- SeqQuest basis sets & ECPs **interoperable*** w/ chemistry codes
- G2/97 test set for numerical validation:

	PBE	HSE
experiment – all-electron:	0.20	0.05
all-electron – DZP:	0.07	0.07
DZP – DZP+ECP:	0.23	0.18
GAUSSIAN – SeqQuest (DZP+ECP):	0.02	0.02

RMS errors (eV/atom)

Screening HSE

bulk Al, DZP+ECP, 2.5 bohr exchange length

LOOP atomic shell I (central cell):

LOOP atomic shell J :

screen (I,J) pairs

LOOP atomic shell K :

screen (I,J,K) triplets

LOOP atomic shell L :

screen (I,J,K,L) quartets

read 1RDM blocks

screen w/ 1RDM blocks

Fock exchange on (I,J,K,L)

write Hamiltonian blocks

$\epsilon = 10^{-7}$ ($E_x = 122.371$)

1×10^4 pairs

3×10^3 unscreened pairs

7×10^6 triplets

3×10^5 unscreened triplets

6×10^8 quartets

1×10^7 unscreened quartets

2×10^7 pre-screened quartets

$\epsilon = 10^{-8}$ ($E_x = -0.309$)

1×10^4 pairs

4×10^3 unscreened pairs

8×10^6 triplets

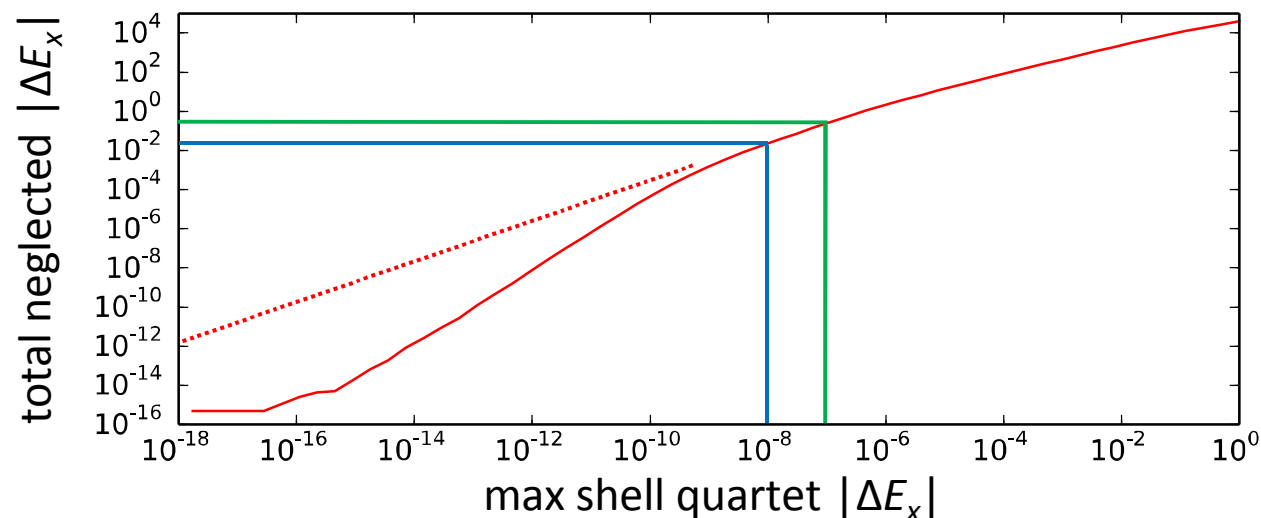
6×10^5 unscreened triplets

1×10^9 quartets

3×10^7 unscreened quartets

4×10^7 pre-screened quartets

ill-conditioned summation →



Work in progress

- **Precompute/accelerate** screening to mitigate cost
- Fully **verify** periodic HSE implementation
- Compatible MPI **parallelization**
- Coulomb tensor decompositions?

I am a “software scientist”

- Some custom software for almost every paper
- Firm believer in **open/reproducible** computing
- Attempts at **pedagogical** software w/ methods papers
- Monolithic methods/codes should exist in a **primordial sea** of software fragments that explore/communicate new ideas

Pros & cons of emphasizing software

- + accelerate development cycle
- + maximize hardware performance
- + maximize algorithm performance
- + enable quantitative comparisons

- attachment to mature software
- difficult to follow HPC evolution
- methodological ossification
- can't fix unhealthy competition

Will this *help* or *hinder* the adoption of new, asymptotically better ideas?

Linear scaling electronic structure methods

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Methods exhibiting linear scaling with respect to the size of the system, the so-called $O(N)$ methods, are an essential tool for the calculation of the electronic structure of large systems containing many atoms. They are based on algorithms that take advantage of the decay properties of the density matrix. In this article the physical decay properties of the density matrix will first be studied for both metals and insulators. Several strategies for constructing $O(N)$ algorithms will then be presented and critically examined. Some issues that are relevant only for self-consistent $O(N)$ methods, such as the calculation of the Hartree potential and mixing issues, will also be discussed. Finally some typical applications of $O(N)$ methods are briefly described. [S0034-6861(99)00104-X]

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← [Rev. Mod. Phys. **71**, 1085 (1999)]

post-HF methods:

[Saebo & Pulay, Annu. Rev. Phys. Chem. **44**, 213 (1993)]

stochastic methods:

[Silver & Roder, Int. J. Mod. Phys. C **5**, 735 (1994)]

[Baer, Neuhauser, & Rabani, Phys. Rev. Lett. **111**, 106402 (2013)]

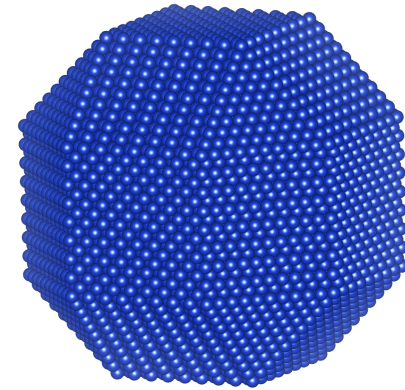
available linear-scaling DFT codes: ONETEP

“Random Phase Approximation: Models and Algorithms for Electron Correlation” [my 2014 BES early career proposal]

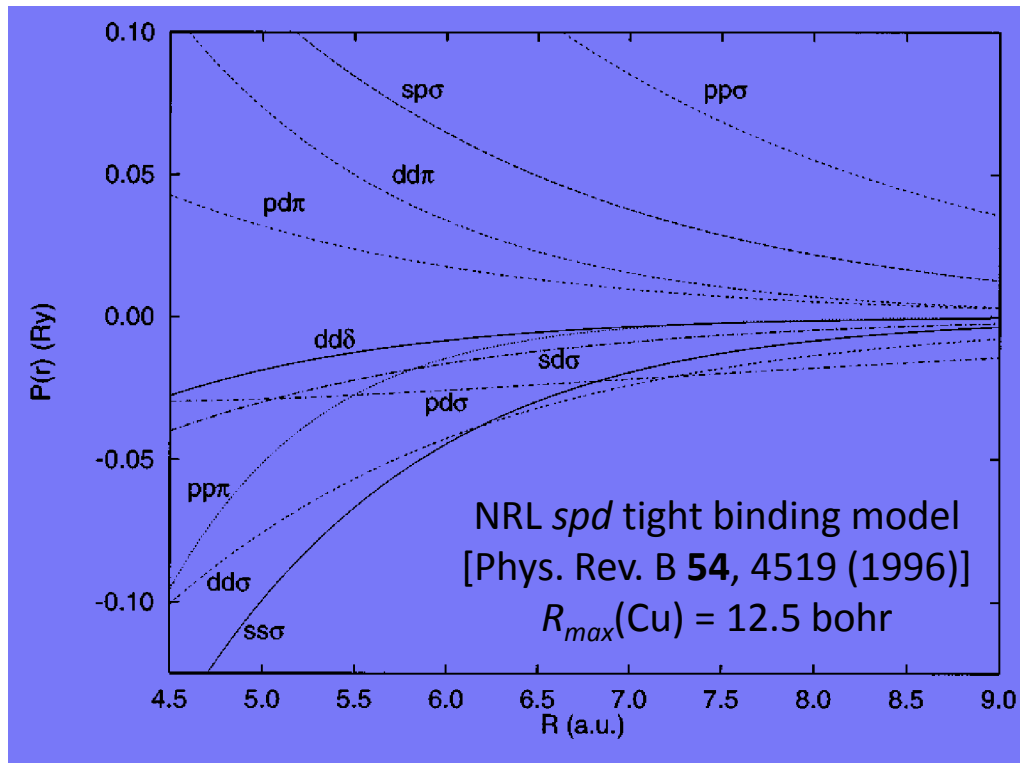
	Atomistic simulation method	Computer time per atom per simulation time step (seconds)	Maximum number of atoms simulated
(2012)	Embedded-atom MM [1]	0.000003	1,000,000
(2012)	ReaxFF MM [1]	0.0004	1,000,000
(2012)	Localized tight binding [2]	0.002	3,000
(2010)	Machine-learned MM [3]	0.01	300
	Target of this proposal	≈ 1 (aspirational)	–
(2014)	Localized DFT [4]	10	100,000
(2014)	Stochastic RPA [5]	100	1,000
(2013)	Localized MP2 [6]	2,000 (1D)	2,000
(2013)	Stochastic DFT [7]	3,000	1,000
(2013)	Localized CCSD [8]	3,000 (1D)	500

Table 1: Comparison of cost between linear-scaling atomistic simulation methods.

“Realistic” test problems

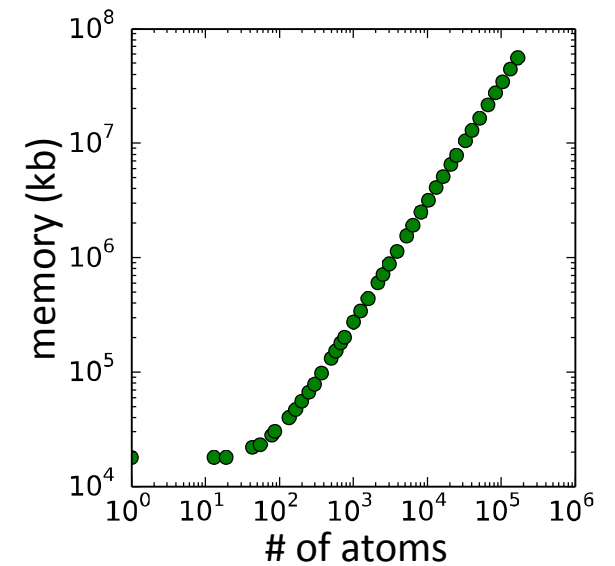
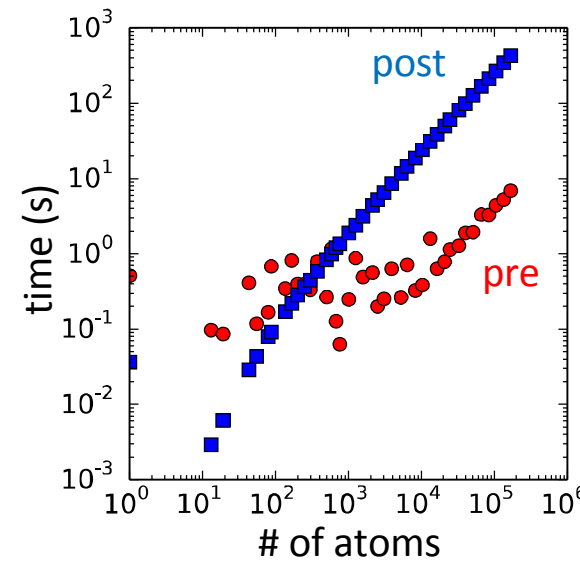


rhombicuboctahedral
copper clusters



target temperatures: 0.03 eV (ambient)
3.0 eV (warm dense matter)

target force accuracy: 0.01 eV/Å



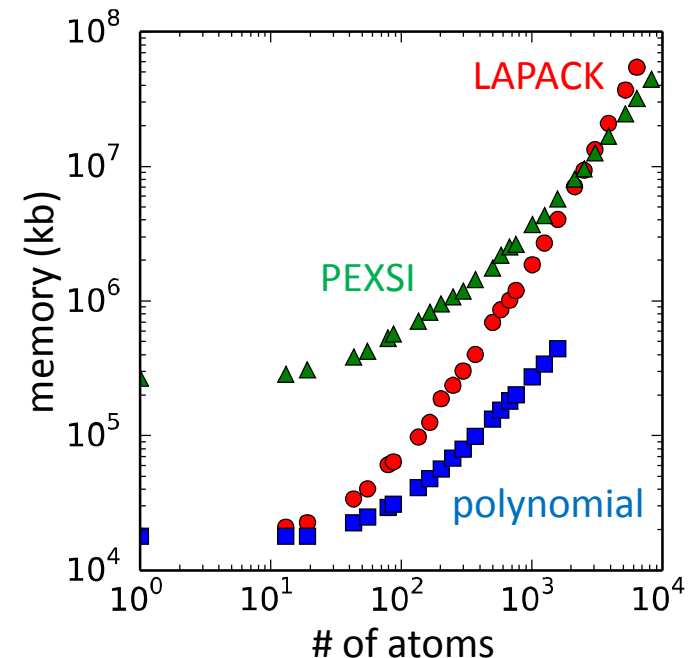
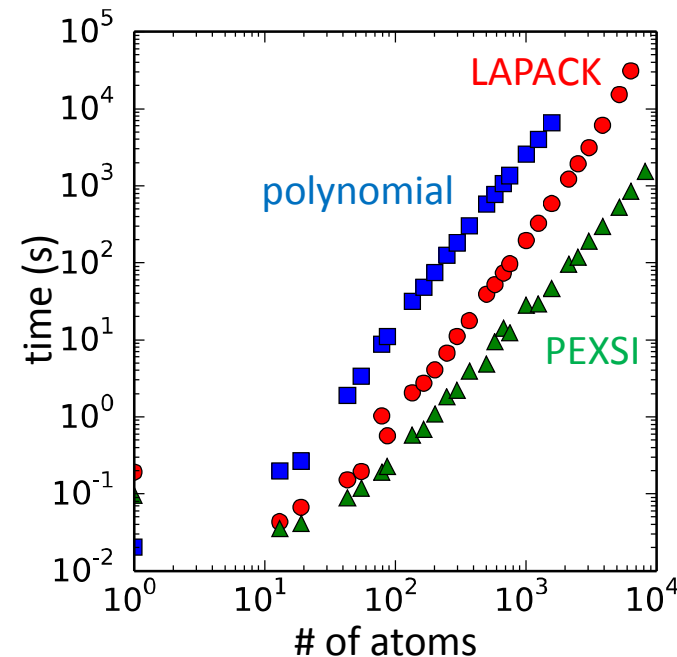
pre/post-processing benchmarks

Basic numerical ingredients

- NRL tight-binding model
- pseudorandom number generation
- block-sparse matrix operations
- geometric localization / partitioning
- polynomial Fermi-Dirac approximation
- rational Fermi-Dirac approximation
- iterative linear solvers (CG & CR)
- PEXSI solver interface

3k lines of C w/ OpenMP & MPI (PEXSI only)

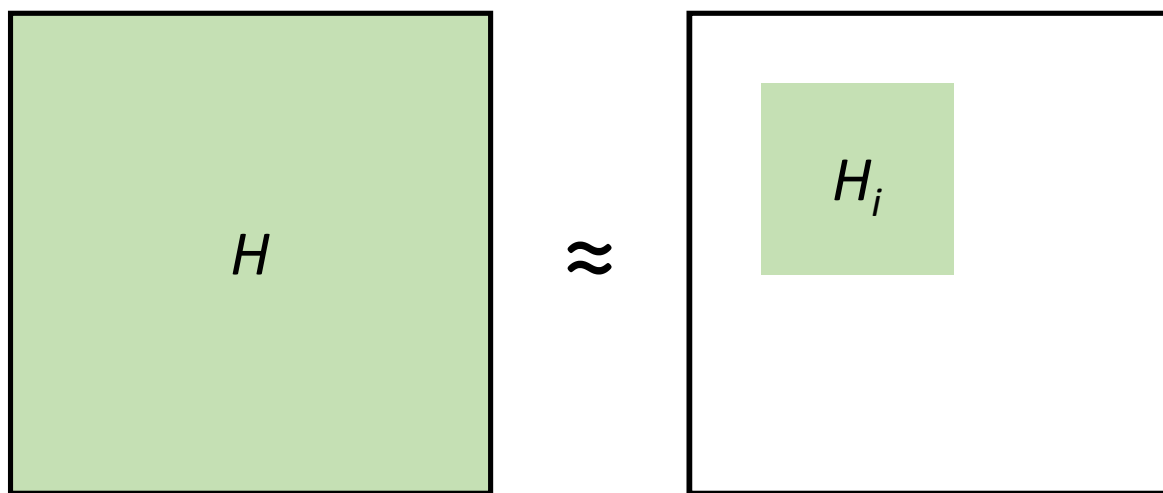
input: Fermi energy & *.xyz
output: total energy & forces



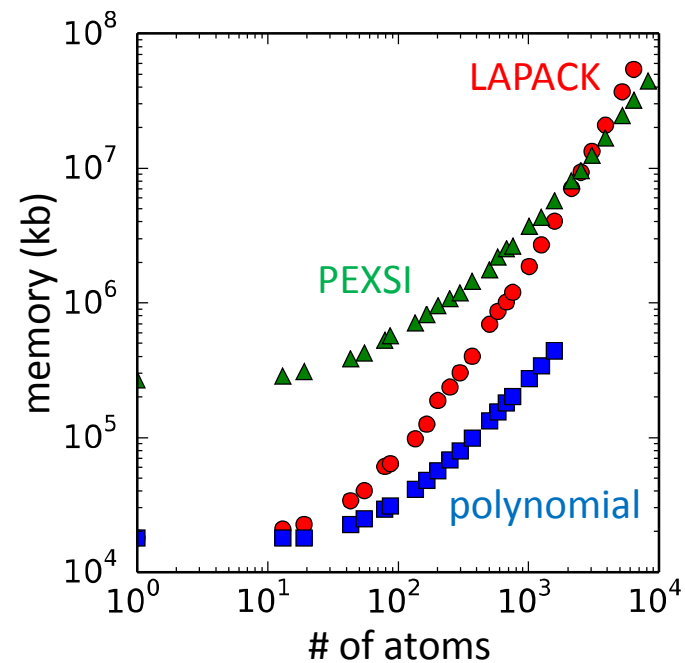
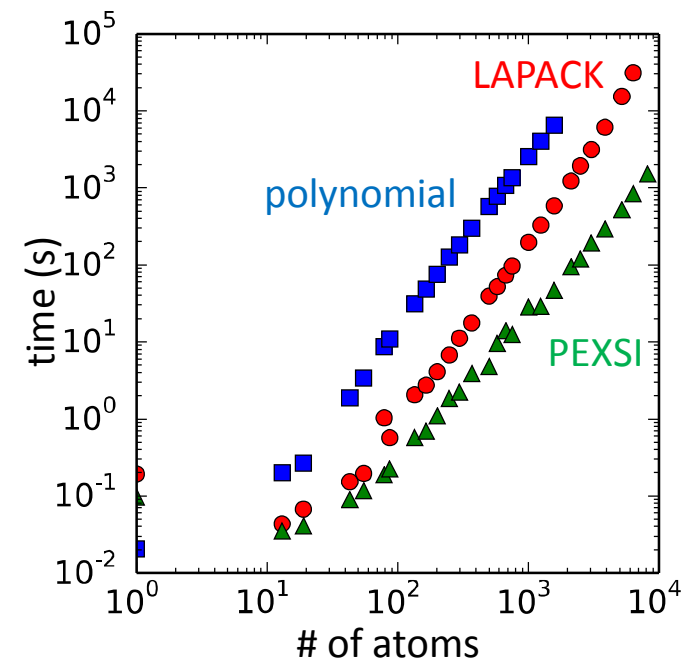
6-pole rational *or* degree-16 polynomial @ $T = 3.0$ eV

Localization *a priori*

$$f(H)e_i \approx f(H_i)e_i$$



error depends on T , gap, & sparsity



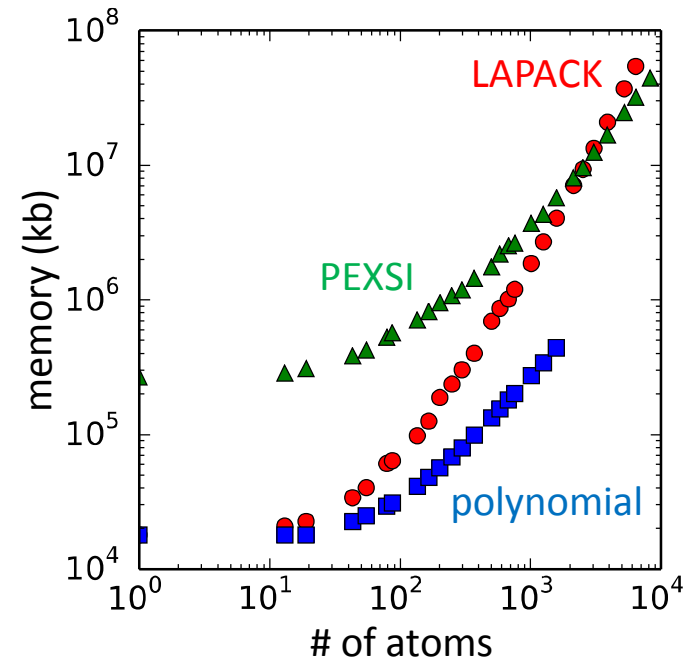
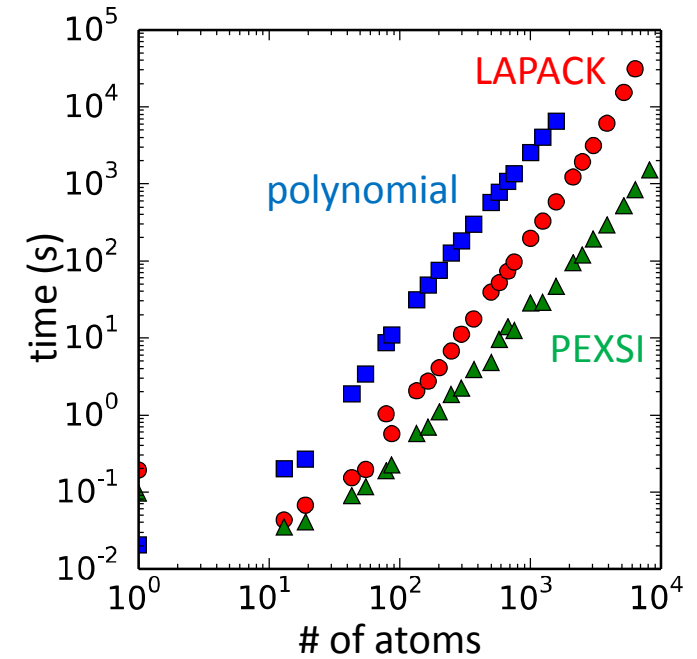
Randomization

$$\text{tr}[f(H)] \approx \frac{1}{s} \sum_{i=1}^s x_i^T f(H) x_i$$

stochastic matrix trace estimation

[Avron & Toledo, J. ACM **58**, 8 (2011)]

error $\propto s^{-1/2}$



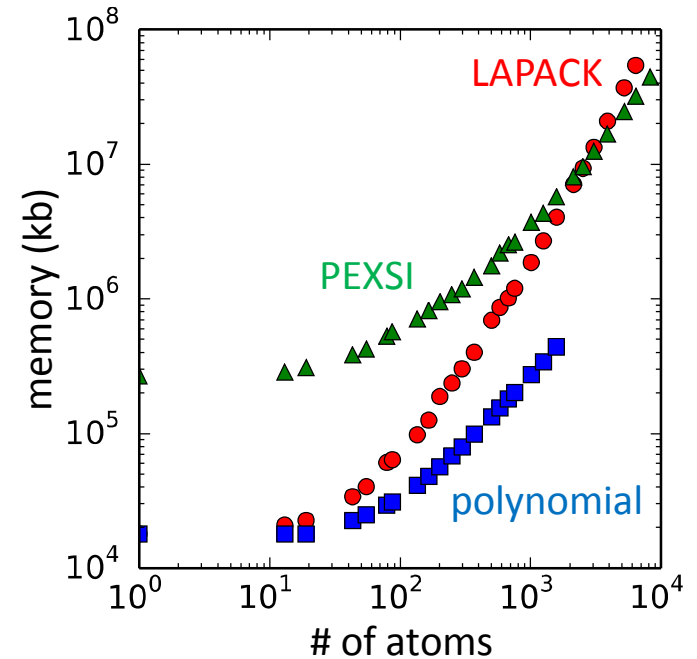
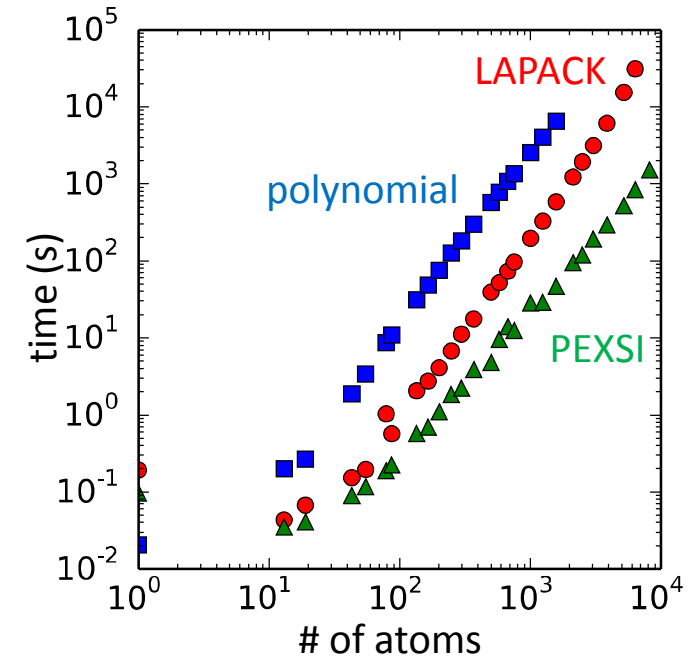
Localization *a posteriori*

$$\min_P \sum_{i=1}^s \|[P - f(H)]x_i\|_2^2$$

fixed-sparsity analog of
compressed sensing

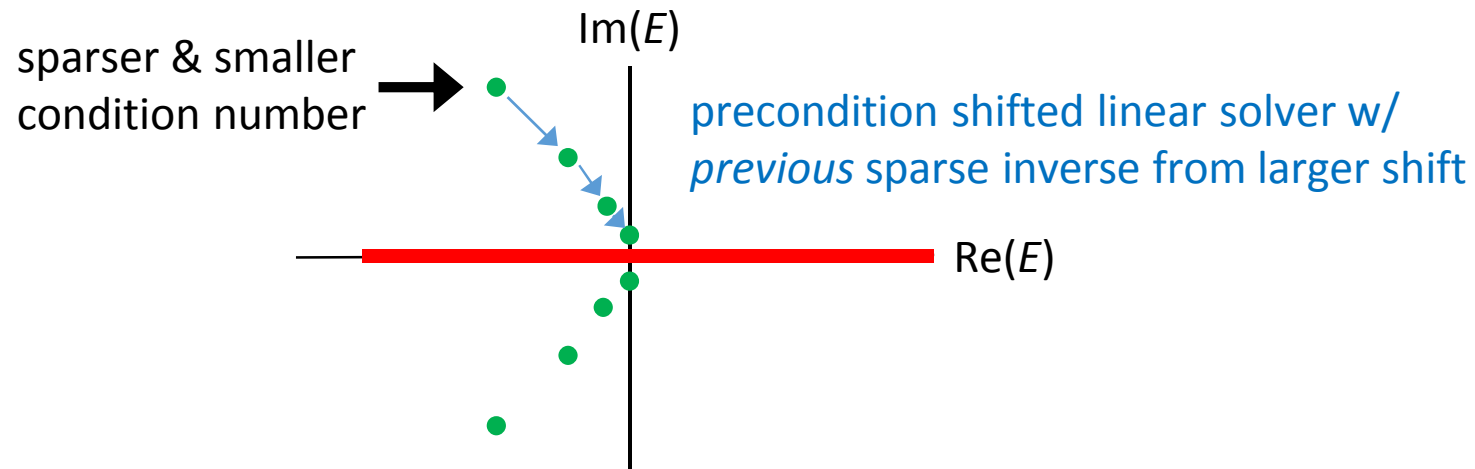
[Candes, Romberg, & Tao, Comm. Pure Appl. Math. **59**, 1207 (2006)]

error \propto [localization error] * $s^{-1/2}$



Rational approximation & preconditioning

$$f(H)x_i = \sum_j c_j (H - z_j S)^{-1} x_i$$



@ 2018 APS March meeting (3PM, March 7th)

Does this have a future?

- Yes, if **moderate** numerical accuracy suffices
- Efficient **$\log(1/\epsilon)$ multiscale** methods are *long-term* research
- **Holistic design**: electron correlation model from fast ingredients
- Compatible w/ fast MP2 & RPA methods
[Moussa, J. Chem. Phys. **140**, 014107 (2014)]

Future of electronic structure?

- Method/software development has always been technically difficult & will only get more difficult as sophistication increases.
- Electronic structure (e.g.@Sandia: QMCPACK, SeqQuest, Socorro, PyQuante, MPQC, LMTO-ASA, MECCA) has struggled to achieve as much practical impact as classical MD (e.g.@Sandia: LAMMPS).

My research plan:

1. linear-scaling solver
2. RPA correlation model
- 3a. semi-empirical basis (stopgap)
- 3b. all-electron basis

“Big-science” electronic structure:

- Are there enough people?
- Is there enough money?
- Is there a design that *can* succeed?
- Is there enough consensus?

At Sandia:

- No.
- No.
- No.
- No.