

# Quantum Monte Carlo methods

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# The Theory of Everything

$$i\hbar \frac{\partial \Psi}{\partial t} = \mathcal{H} \Psi$$

$$\mathcal{H} = - \sum_j^N \frac{\hbar^2}{2m} \nabla_j^2 - \sum_\alpha^M \frac{\hbar^2}{2M_\alpha} \nabla_\alpha^2 - \sum_j^N \sum_\alpha^M \frac{Z_\alpha e^2}{|r_j - R_\alpha|} \\ + \sum_{j < k}^N \frac{e^2}{|r_j - r_k|} + \sum_{\alpha < \beta}^M \frac{Z_\alpha Z_\beta e^2}{|R_\alpha - R_\beta|}$$

- |          |           |              |                  |
|----------|-----------|--------------|------------------|
| * Air    | * Steel   | * Paper      | * Vitamins       |
| * Water  | * Plastic | * Dynamite   | * Ham Sandwiches |
| * Fire   | * Glass   | * Antifreeze | * Ebola Virus    |
| * Rocks  | * Wood    | * Glue       | * Economists     |
| * Cement | * Asphalt | * Dyes       | * ...            |

Robert Laughlin: Nobel Prize Talk, viewgraph #2

# Electronic structure and properties of materials

## Hamiltonian of interacting electrons and ions

$$H = -\frac{1}{2} \sum_i \nabla_i^2 - \sum_{i,I} \frac{Z_I}{r_{iI}} + \sum_{i<j} \frac{1}{r_{ij}} + E_{ion-ion}$$

$$H \psi(\mathbf{r}_1, \mathbf{r}_2, \dots) = E \psi(\mathbf{r}_1, \mathbf{r}_2, \dots)$$

- ground states
- excited states, optical properties
- responses to external fields
- $T > 0$ , etc ...

# Traditional electronic structure methods

## Density Functional Theory :

-based on one-particle density

$$E_{tot} = \int F[\rho(\mathbf{r})] d\mathbf{r}$$

- exact functional  $F$  is unknown
- various approximations for  $F$  :  
LDA (local density approx.)  
GGA (generalized grad. approx.)

**Problem: efficient but inaccurate**  
(need accuracy 0.1 eV or higher)

**Hartree-Fock :** wavefunction as Slater determinant (antisymmetry) of one-particle orbitals

$$\Psi_{HF}(\mathbf{r}_1, \mathbf{r}_2, \dots) = \text{Det}[\{\phi_i(\mathbf{r}_j)\}]$$

**Post-Hartree-Fock: expansion in excitations**

$$\Psi_{corr}(\mathbf{r}_1, \mathbf{r}_2, \dots) = \sum_n d_n \text{Det}_n[\{\phi_i(\mathbf{r}_j)\}]$$

**More advanced: CC, MBPT etc**

**Problem: accurate but inefficient**

Alternatives ?

Quantum Monte Carlo (QMC)

# Accuracies which we need

- cohesion, optical excitations, barriers :  $\sim 0.1 - 0.01\text{eV} \sim 1000 - 100\text{K}$
- magnetism, superconductivity, spintronics :  $\sim 0.001\text{ eV} \sim 10\text{K}$
- QED (important) :  $\sim 0.000001\text{ eV}$
- recent calculations of sixth order QED corrections for He atom:  
12 digit accuracy

Nature employs energy, length, etc scales as a composer employs various orchestral instruments

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# Traditional electronic structure methods and quantum Monte Carlo

## Post-HF methods:

- convergence in one-particle basis sets is slow, inefficient description of many-body effects
- need to explicitly evaluate integrals restricts functional forms which can be used

## DFT approaches:

- difficult systematical improvement (the fundamental proof is not constructive)

## QMC:

- use stochastic methods to map the many-body problem onto a sampling/simulation problem
- focus on many-body effects and efficiency of their description
- many ideas applicable to other systems/models: lattices, etc.
- in many ways, complementary to traditional approaches

# Variational Monte Carlo (VMC): Slater-Jastrow wavefunction

"Zero order" wave function: Slater determinant(s)

$$\psi_0(\mathbf{R}) = \sum_n d_n \text{Det}_n[\phi_\alpha] \text{Det}_n[\phi_\beta]$$

one-particle orbitals from - Hartree-Fock, MCSCF

- Density Functional Theory (DFT)

- ...

Correlate the Slater wave function explicitly: e-e, e-e-l terms, ...

$$\psi_T(\mathbf{R}) = \sum_n d_n \text{Det}_n[\phi_\alpha] \text{Det}_n[\phi_\beta] \exp(U_{corr})$$

$$U_{corr} = \sum_{i,j} \sum_I \sum_{k,l,m} c_{klm} a_k(r_{ij}) b_l(r_{iI}) b_k(r_{jI})$$

*electrons*

*ions*

*correlation basis expansion*

Other functional forms: pairing BCS wavefunctions, pfaffians, ...



# Variational Monte Carlo (VMC): stochastic methods for multi-D integrals

- evaluate the expectation values by sampling  $3N$  electron coordinates
- no other choice! For example, variational energy is given by

$$E_{VMC} = \frac{\int \psi_T^2 [H \psi_T / \psi_T] d\mathbf{R}}{\int \psi_T^2 d\mathbf{R}} = \lim_{M \rightarrow \infty} \frac{1}{M} \sum_{sample} \frac{H \psi(\mathbf{R}_{sample})}{\psi(\mathbf{R}_{sample})}$$

where samples are distributed as  $\psi_T^2(\mathbf{R})$  and  $\mathbf{R} = (r_1, r_2, \dots, r_N)$

- optimize any variational parameters  $\{c_{klm}\}$ ,  
eg, minimize the energy or local energy variance (Umrigar et al, '88)

$$\sigma_{VMC}^2 \approx \frac{1}{M} \sum_{sample} \left[ \frac{H \psi(\mathbf{R}_{sample})}{\psi(\mathbf{R}_{sample})} - E_{VMC} \right]^2 = \min$$

# Beyond VMC

**VMC** is straightforward, fast, efficient, but not accurate enough:  
**variational bias**

**Eliminate the VMC bias by** projecting out the (ground) state in imaginary time

$$\psi(\mathbf{R}, t) = \exp(-tH) \psi_T(\mathbf{R})$$

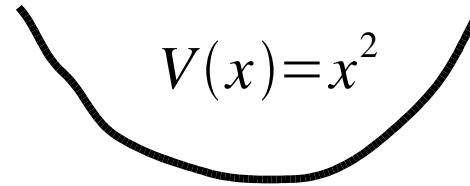
$$-\partial_t \psi(\mathbf{R}, t) = H \psi(\mathbf{R}, t)$$

$$\psi(\mathbf{R}, t + \tau) = \int G(\mathbf{R}, \mathbf{R}', \tau) \psi(\mathbf{R}', t) d\mathbf{R}'$$

Wave function can be sampled and the equation solved by interpreting the Green's functions as a transition probability density: simulate by a stochastic process

# Toy model: 1D harmonic oscillator

$$H = T + V(x)$$



**Propagator**

$$G(x, x', \tau)$$



$$C e^{-(x-x')^2/4\tau} \cdot e^{-(V(x)-E_T)\tau}$$

diffusion

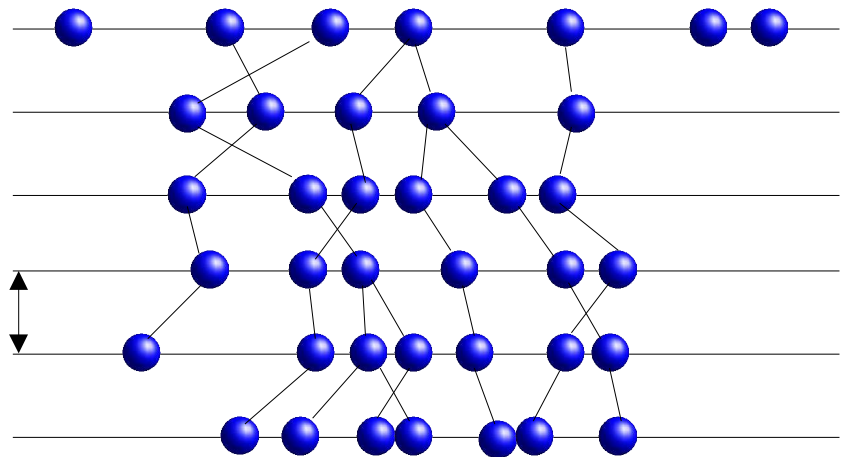
renorm

$$\psi_{init}(x)$$

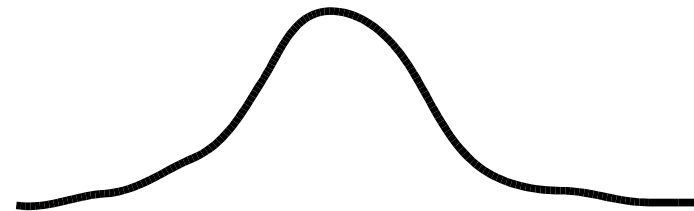


$t \downarrow$

$\tau$



$$\psi_{ground}(x)$$



# Fermion sign problem

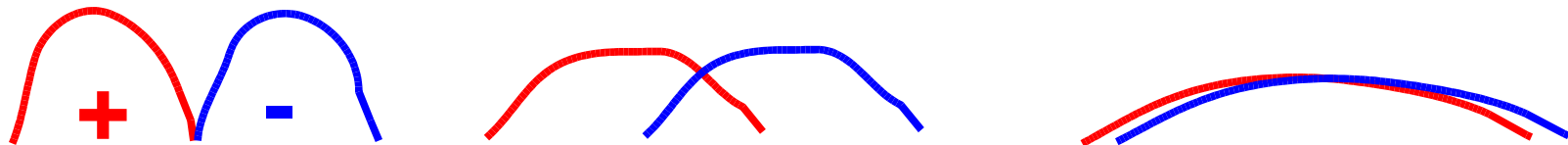
Naïve approach for fermionic wave functions: decompose to + and -

$$\psi_T(\mathbf{R}) = \psi_T^+(\mathbf{R}) - \psi_T^-(\mathbf{R})$$

$$-\partial_t \psi^+(\mathbf{R}, t) = H \psi^+(\mathbf{R}, t)$$

$$-\partial_t \psi^-(\mathbf{R}, t) = H \psi^-(\mathbf{R}, t)$$

Unfortunately, + and - components converge independently to the lowest energy solution (which is bosonic) because Schr. eq. is linear!



$$\lim_{t \rightarrow \infty} \psi^+(\mathbf{R}, t) - \lim_{t \rightarrow \infty} \psi^-(\mathbf{R}, t) \propto \exp[-(E_{Fermi} - E_{Boson})t]$$

Fermion "signal" decays exponentially quickly into a bosonic "noise"

# Importance sampling and fixed-node diffusion Monte Carlo (FNDMC)

$$f(\mathbf{R}, t+\tau) = \int G^*(\mathbf{R}, \mathbf{R}', \tau) f(\mathbf{R}', t) d\mathbf{R}'$$

$$f(\mathbf{R}, t) = \psi_T(\mathbf{R}) \phi(\mathbf{R}, t),$$

$$\psi_T = \psi_{HF} e^{U_{corr}} = \det\{\phi_\alpha\} \det\{\phi_\beta\} e^{U_{corr}}$$

$$f(\mathbf{R}, t \rightarrow \infty) \propto \psi_T(\mathbf{R}) \phi_{ground}(\mathbf{R})$$

$$G^*(\mathbf{R}, \mathbf{R}', \tau) = \frac{\langle \mathbf{R} | \exp(-\tau H) | \mathbf{R}' \rangle}{\psi_T(\mathbf{R}') \psi_T^{-1}(\mathbf{R})}$$

**Fermion node: (3N-1)-dimen. hypersurface defined as**  $\phi(\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_N) = 0$

**Fixed-node approximation:**  $f(\mathbf{R}, t) > 0$

- antisymmetry (nonlocal) replaced by a boundary (local)
- exact node implies recovering exact energy (in polynomial time)

**Accuracy quite high: energy differences within a few % of experiment**

# Fermion node toy model: excited state of harmonic oscillator

$$H = T + V(x)$$

**Propagator**

$$G(x, x', \tau)$$

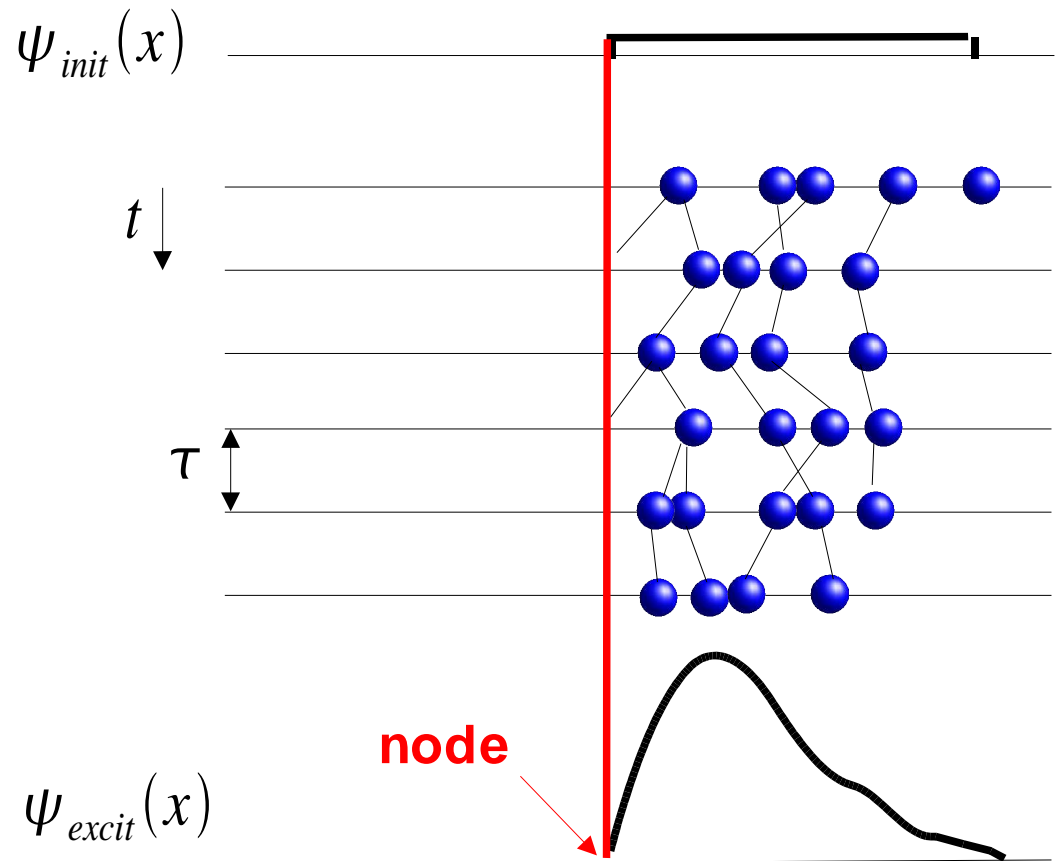
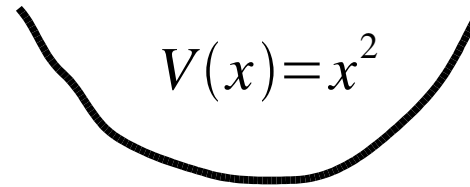


$$C e^{-(x-x')^2/4\tau} \cdot e^{-(V(x)-E_T)\tau}$$

↖  
**diffusion**

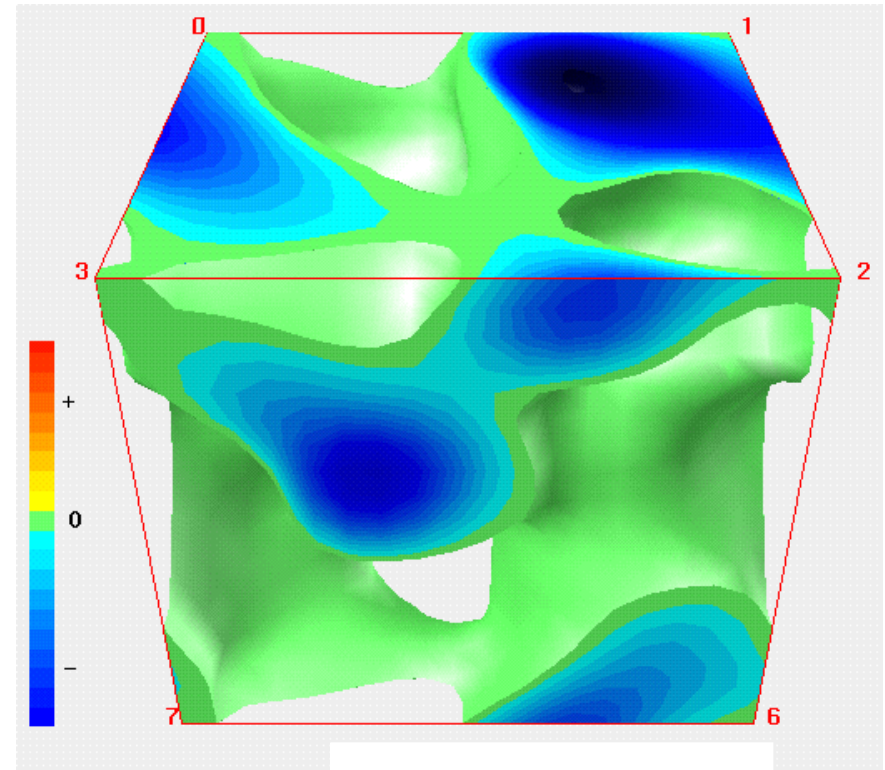
↖  
**renorm**

**+ boundary condition  
(evaluate trial function)**



# Fixed-node approximation

- bosonization of the fermionic problem
- important shift (and a leap of faith):  
**antisymmetry** (nonlocal)  $\rightarrow$  **boundary condition** (local)
- fermion node is  $(3N-1)$ -dim hyper-surface:
- easy to enforce (check the sign of the determinant)
- difficult to parametrize with arbitrary accuracy



Green surface: 3D cut of 59-dimensional fermion node hypersurface

# The key paradigm shift in QMC

- the many-body (Schrodinger) differential equation mapped onto equivalent stochastic process which is easy to simulate

density of walkers in  $3N$ -dim space  $\Leftrightarrow$  value of the wavefunction

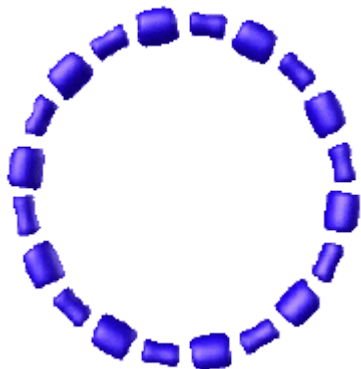


# The key directions of (my) QMC research

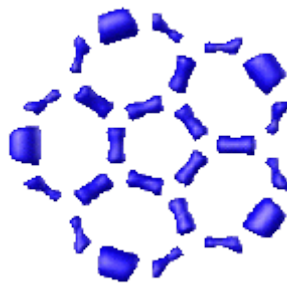
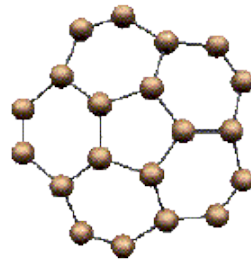
- applications of QMC to challenging materials and to materials design: **strongly correlated systems, competing many-body effects, new materials**
- development of fundamental understanding of quantum many-body systems and methods capable of solving such challenging problems: **fermion nodes, treatment of magnetic effects, etc**
- coupling of QMC with other approaches for multi-scale application: **coupling of QMC with molecular dynamics,  $T > 0$ , etc**

# Example of application: which is the lowest energy isomer???

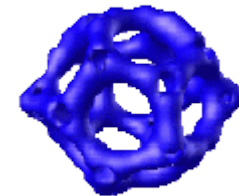
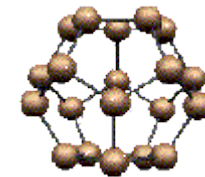
C20 challenge: ring



bowl

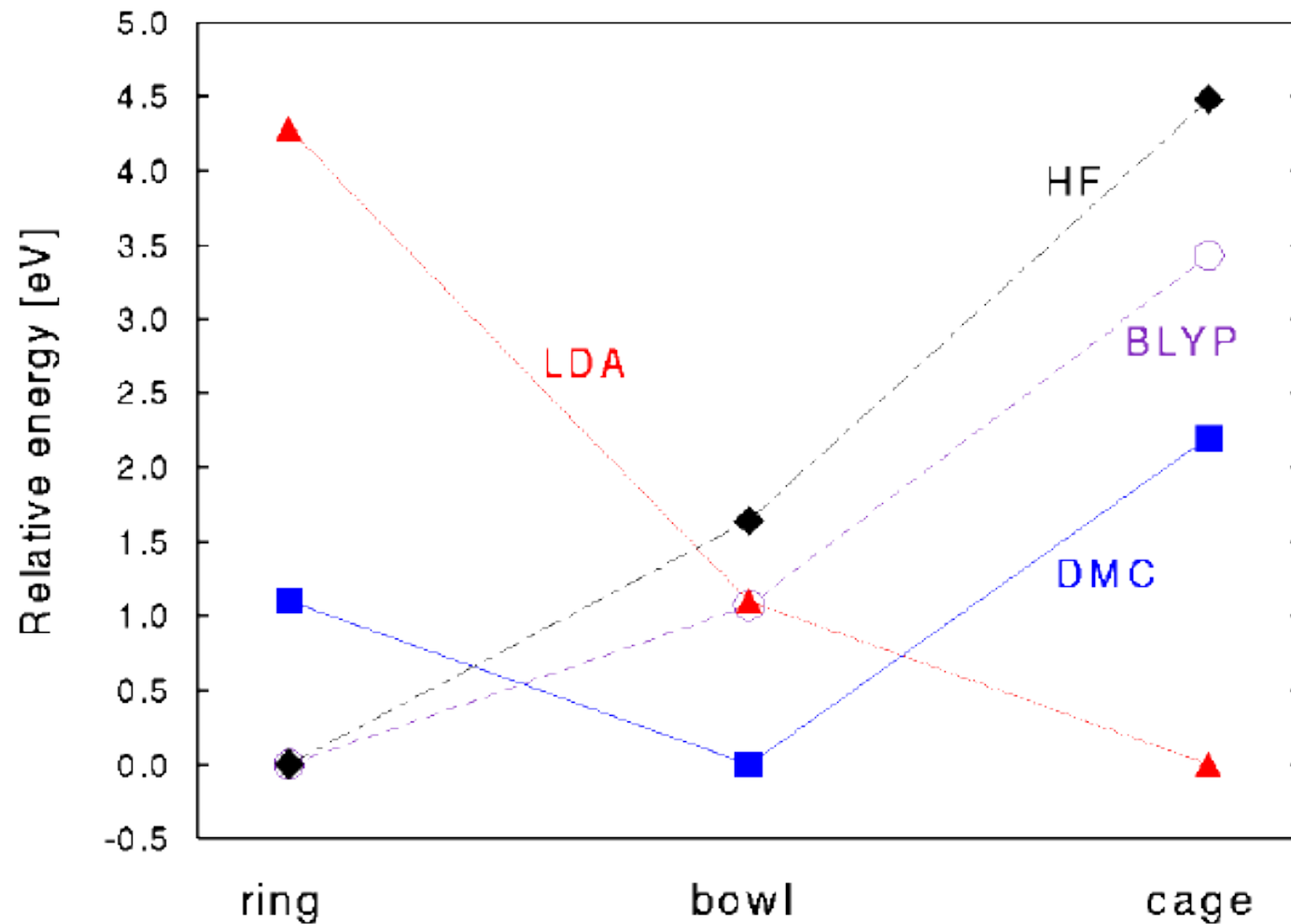


cage



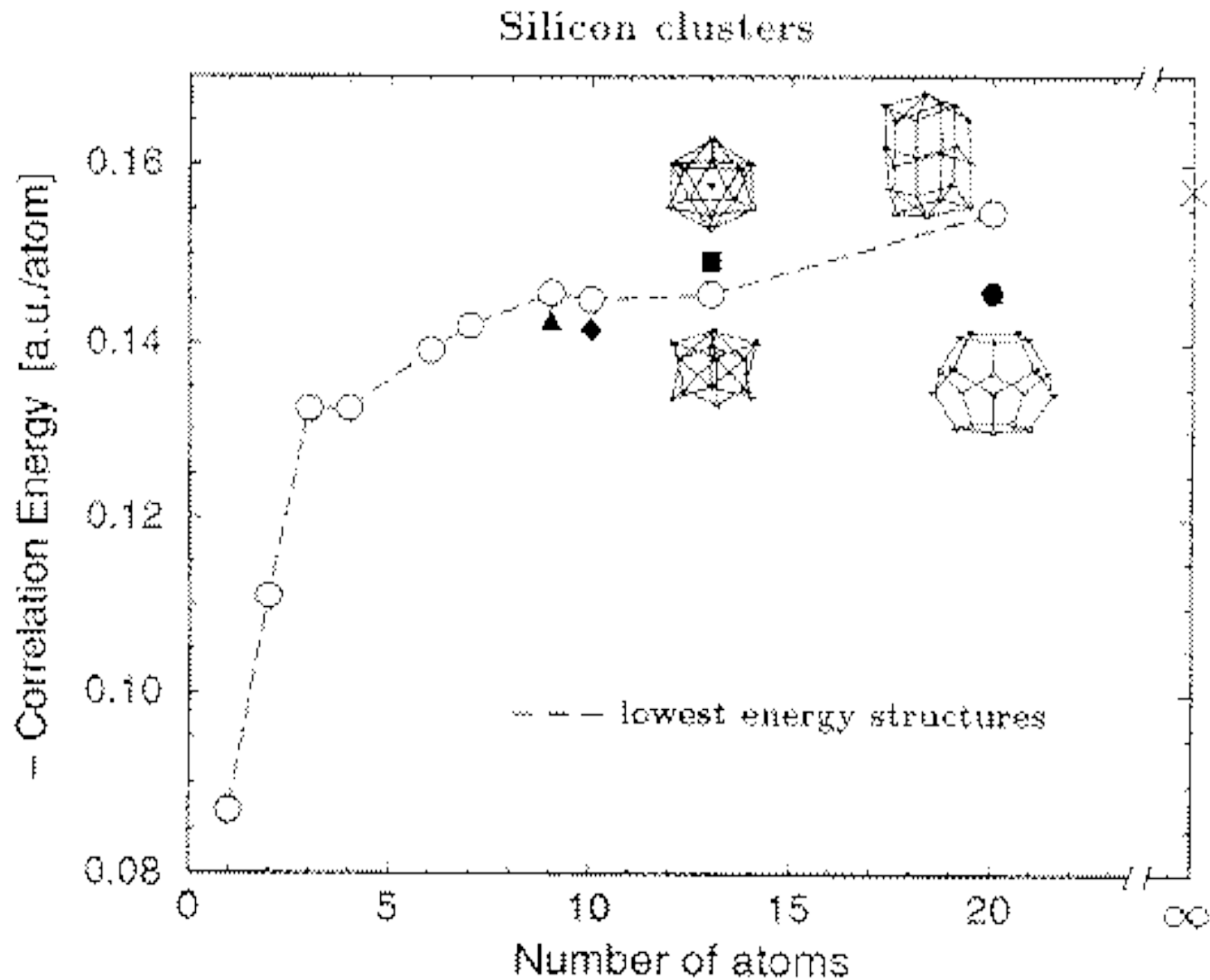
J.C. Grossman, L. Mitas, K. Raghavachari, Phys. Rev. Lett. 75, 3870 (1995)

# QMC was the first method to predict the energy order of isomers

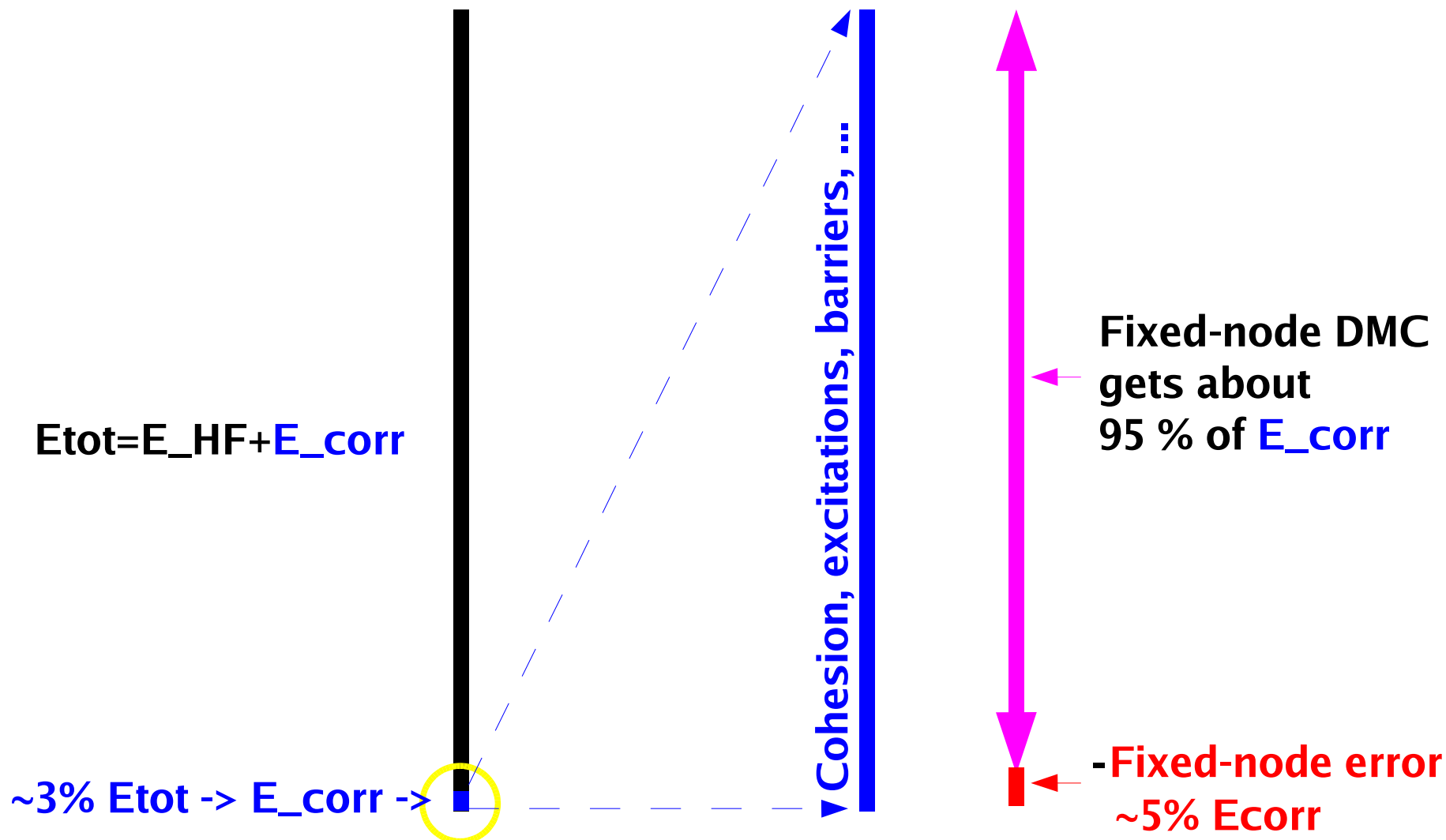


J.C. Grossman, L. Mitas, K. Raghavachari, Phys. Rev. Lett. 75, 3870 (1995)

# Calculate correlation energy: from a single atom to molecule to clusters all the way to the solid



# Performance of the fixed-node DMC method using a few-determinant trial functions



# Solving the stationary Schrödinger equation: key ingredients

The many-body problem splits into two parts

←  
- fermion nodes, ie, boundary conditions

- often one or few determinants are very reasonable

- "non-dynamical" correlation, multi-reference wavefunctions

- can be obtained within mean-field/basis set methods

→  
- "bosonic" correlations within the fermion node domains

- e-e cusps, multiple collision points, etc

- "dynamical" correlation, very difficult for other methods

- easy within QMC

The combined strategy "corners" the error in the treatment of many-body effects into the last few percent of correlation energy. The methods are complementary: together, they produce

# Summary on application and performance of QMC methods

- fixed-node DMC typically recovers about 95% of the valence correlation energy for up to hundreds of electrons
- energy differences agree with experiments within a few percent
- method scales like a  $N^3$  where  $N$  is the number of valence electrons (core electrons eliminated by pseudopotentials)
- applied to a number of systems, eg, electron gas and quantum liquids, atoms, molecules, solids etc; often the results became benchmarks for other methods
- about two orders of magnitude slower than mean-field methods but very efficient (perfectly scalable) on parallel architectures

# Computational advantages

- scaling in the number of particles:  $N^a$ ,  $a = 1 \sim 3$
- **sampling walkers independent** -> natural parallelism and scalability, slow networks OK, heterogeneous clusters, ...
- **cycle intensive, less memory intensive**
- **robust**
- **enable to focus in interesting physics**



# Impact of QMC methods

## Breakthroughs and benchmark calculations:

- homogeneous electron gas in '80 (Ceperley & Alder)
- quantum liquids and solids
- barrier of  $\text{H}+\text{H}_2 \rightarrow \text{H}_2+\text{H}$  with **0.001 eV** accuracy (J. B. Anderson)
- calculations of solids, clusters, etc (last 15 years)

## Relative accuracy for s-p systems

- energy differences typically within 1-3% of experiment

## Sizes of systems and timing

- 100-200 valence electron systems becoming routine;  
(1000 or more doable at the current level of development)
- typical run: 32 processors for a day
- about 100 times slower than mean-field approaches

# Current limitations of QMC methods

**Fundamentally, fixed-node approximation is still present:**

- sometimes the physical effect of interest is in the last few percent of the correlation

**Quantities beyond energies:**

- we need better methods for getting correlation functions, forces, responses to external fields, etc
- small energy scales (ie, magnetism) is burried in the error bars

**Need inputs (orbitals) from other approaches:**

- often that is the most time consuming part :-)
- many of currently available DFT/HF codes are not built for the purpose to provide the best possible inputs to QMC

# One- and two-particle quantities vs many-particle wave functions

**Traditional approaches are based on one- or two-particle quantities:**

- one-particle density (matrix), orbitals
- one-particle Green's function
- pair-correlation functions, etc

since these contain just the right amount of information necessary to describe the physics.

**We think and "understand" in terms of reduced quantities.**

**QMC, however, is based on wave functions: ... despite that**

- in general, wave functions are intractable and contain too much of information most of which is irrelevant
- limited to small number of particles, otherwise intractable
- difficult to understand the physics

**QMC shows that maybe the traditional reductionism might have reached its limits ...**

# QMC and wavefunctions: only an accurate method or a paradigm shift ?

- in many cases it is more efficient to carry out the many-particle calculations rather than to (re)build a better mean-field
- let the machine to worry about reducing the unnecessary information
- combination of analytical insights and stochastic techniques a key for getting high accuracy solutions of Schr. eq.
- in working with wave functions one is closest to the many-body physics and understanding (perhaps even the most efficient way how to capture the correlation effects)
- history of high accuracy benchmark/reference calculations