

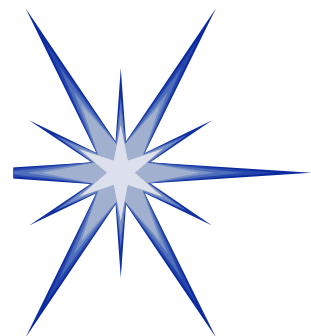
Stochastic Diagonalization

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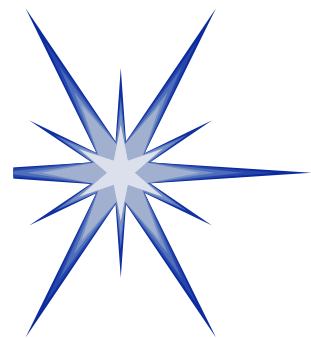
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Outline

- **Introduction**
 - Formulation of the problem
- **Minus-sign problem in Quantum Monte Carlo methods**
 - General aspects only
- **Stochastic diagonalization**
 - Theory
 - Comparison with other (QMC) methods
 - 2D tt' Hubbard model
 - Quantum Chemistry Problems
- **Application:**
 - Off-diagonal long-range order, superconductivity
- **Basis-set optimization**
- **Conclusions**



Formulation of the problem

- **Quantum statistical mechanics, (lattice) QED, (lattice) QCD:**

$$\langle X \rangle = \frac{\text{Tre}^{-\beta H} X}{\text{Tre}^{-\beta H}} = \frac{\sum_{\{\psi\}, \{\varphi\}} \langle \varphi | e^{-\beta H} | \psi \rangle \langle \psi | X | \varphi \rangle}{\sum_{\{\psi\}} \langle \varphi | e^{-\beta H} | \varphi \rangle}$$

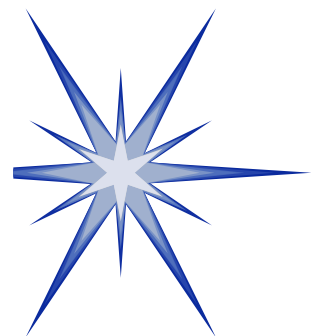
- **H : Hamiltonian**
- **X : Observable**
- **$\{\psi\}, \{\varphi\}$: Complete sets of states**
- **$\beta = 1/k_B T$: Inverse temperature**
- **Quantum dynamics:**

$$i\hbar \frac{\partial}{\partial t} |\Psi(t)\rangle = H |\Psi(t)\rangle \quad \Leftrightarrow \quad |\Psi(t)\rangle = e^{-itH/\hbar} |\Psi(0)\rangle$$

- **Ground-state ($T=0$) properties:**

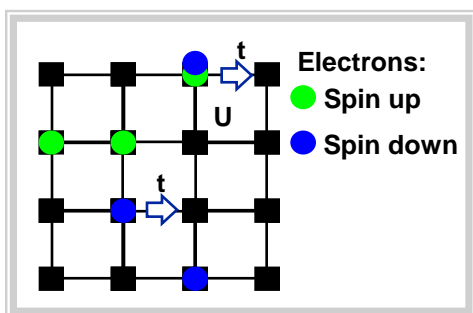
$$H |\Phi_1\rangle = E_1 |\Phi_1\rangle$$

- **E_1 is the smallest eigenvalue of H**

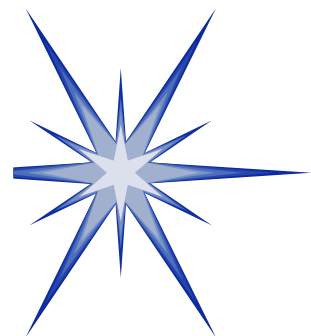


Formulation of the problem

- For numerical purposes, a state of the quantum system, i.e. the wave function, is represented by a vector of length M (the dimension of the Hilbert space) and the Hamiltonian H is a hermitian matrix of dimension $M \times M$.
- We know what we have to do: “Just” solve the Schrödinger equation $H|\Phi\rangle = E|\Phi\rangle$
 - A standard problem of linear algebra
 - Example: The Hubbard-Gutzwiller model (1963)



- Describes fermions (electrons) that hop from one lattice point to another and that feel each other via an on-site repulsion ($U > 0$).



Formulation of the problem

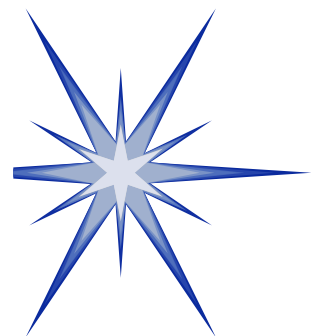
- Count the number of possible ways to distribute N electrons with spin up and N_{\downarrow} electrons with spin down over L lattice sites:

$$M = \binom{L}{N_{\uparrow}} \times \binom{L}{N_{\downarrow}}$$

- For $N = N_{\downarrow} = L/2$: $M \approx \frac{2^{2L+2}}{2\pi L}$
 $Memory \approx \frac{2^{2L-25}}{2\pi L}$ Giga bytes

L	$N_{\uparrow} = N_{\downarrow}$	M	Memory in Gb	Approx. cost
16	8	2×10^8	2	\$ 10^3
36	18	8×10^{19}	6×10^{11}	\$ 10^{14}
64	32	3×10^{36}	3×10^{28}	\$ 10^{20}

Very Large Matrices !



Formulation of the problem

- In general, there is no way to solve these monster eigenvalue problems using perturbative or conventional numerical (Eispack, Lanczos, Davidson, ...) methods.
- Key issue

How to solve (eigenvalue) problems involving extremely large matrices ?

- **General strategy: Trade memory for CPU time**
 - Quantum Monte Carlo (QMC) Methods
 - Stochastic Diagonalization



Quantum Monte Carlo Methods

- **Stochastic methods:**

- Use much less than M storage elements but use much more CPU time
- Example: A QMC method based on

$$\lim_{m \rightarrow \infty} (H - E_{shift})^m |\Phi_{initial}\rangle = |\Phi_1\rangle$$

- **Linear algebra: Power method**

$$\begin{aligned} (H - E_{shift})^m |\Phi_{initial}\rangle &= \sum_{i=1}^M (E_i - E_{shift})^m |\varphi_i\rangle \langle \varphi_i | \Phi_{initial}\rangle \\ \xrightarrow{m \rightarrow \infty} \max_i (E_i - E_{shift})^m |\varphi_i\rangle \langle \varphi_i | \Phi_{initial}\rangle &\propto |\Phi_1\rangle \end{aligned}$$

- for a “good” choice of E_{shift}

- **Physical quantities are related to matrix elements:**

$$\langle \Phi_i | (H - E)^m | \Phi_f \rangle = \sum_{\{\Phi_k\}} \langle \Phi_i | H - E | \Phi_1 \rangle \langle \Phi_1 | H - E | \Phi_2 \rangle \cdots \langle \Phi_{m-1} | H - E | \Phi_f \rangle$$

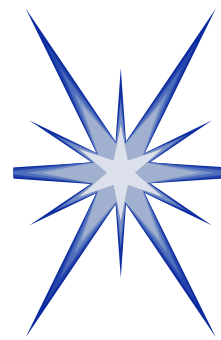
- Takes approximately mM operations
- Does not help much if we would like to sum all terms

Quantum Monte Carlo Simulation: General Strategy

- Importance sampling
 - Basic assumption (*hope*) : If the problem can be solved at all, it must be such that **not all** the states of the system are **equally important**. Using only a few of them should be sufficient to compute the physical properties with reasonable accuracy.
 - In practice a “few ” can still mean $10^4 - 10^8$ states
 - No a-priori justification in the case of quantum systems
 - OK for classical statistical problems
 - How to find these **important** states ?
 - Random search gives bad results



- Metropolis Monte Carlo method, molecular dynamics, Langevin dynamics, ...



Importance sampling

- **Importance sampling:**

- Search for *important* states uses

$$\rho = \rho(\{\Phi_k\}) = \langle \Phi_0 | F | \Phi_1 \rangle \langle \Phi_1 | F | \Phi_2 \rangle \cdots \langle \Phi_{m-1} | F | \Phi_m \rangle$$

as a “density” distribution for the states $\{\Phi_k\}$

- F depends on the application

- **Classical statistical mechanics**

$$F_{\text{Boltzmann}}(H) = e^{-\beta H} \quad ; \quad H \in \mathfrak{R}$$

- **Diffusion QMC (power method):**

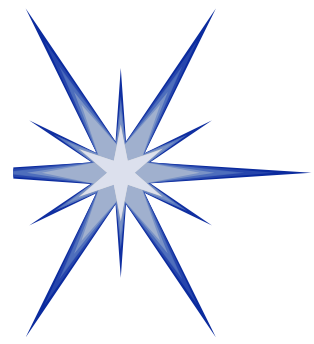
$$F_{\text{DQMC}}(H) = (1 - \tau H)^m$$

- **Green Function QMC (inverse iteration):**

$$F_{\text{GFMC}}(H) = [(E_{\text{shift}} + H)^{-1}]^m$$

- **Path Integral MC (Trotter-Suzuki product formulae)**

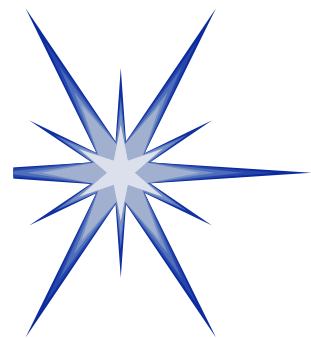
$$F_{\text{PI}}(H = A + B) = \left(e^{-\beta A/m} e^{-\beta B/m} \right)^m$$



Quantum Monte Carlo Methods: General Strategy

- All importance sampling methods require a probability distribution
 - Only positive functions qualify as probability distributions
- Is $\rho > 0$ for all possible states ?
 - In classical mechanics (Boltzmann factor): ALWAYS
 - In quantum mechanics: Almost **NEVER**
 - “Theorems”: If all the off-diagonal elements of the matrix representing the Hamiltonian are negative then $\rho > 0$
 - H. De Raedt and M . Frick, Phys. Rep. 231, 107-149 (1993)
 - The reverse is NOT true but it is save to expect it is.

Minus Sign Problem



Theorems on the positivity of $F(H)$

- **Classical statistical mechanics**

$$F_{\text{Boltzmann}}(H) = e^{-\beta H} > 0 \quad ; \quad H \in \mathfrak{R}$$

- **DQMC:** $F_{\text{DQMC}}(H) = (1 - \tau H)^m$

If $\langle \varphi | H | \psi \rangle \leq 0$ for all $|\varphi\rangle, |\psi\rangle$

then $\langle \varphi | (1 - \tau H) | \psi \rangle \geq 0$

- **GQMC:** $F_{\text{GFMC}}(H) = [(E_{\text{shift}} + H)^{-1}]^m$

If $\omega + H$ is positive definite and $\langle \varphi | H | \psi \rangle \leq 0$

for all $|\varphi\rangle \neq |\psi\rangle$ then $\langle \varphi | (\omega + H)^{-1} | \psi \rangle > 0$

- **Trotter-Suzuki:** $F_{\text{PI}}(H = A + B) = (e^{-\beta A/m} e^{-\beta B/m})^m$

If $\langle \varphi | X | \psi \rangle \leq 0$ for all $|\varphi\rangle \neq |\psi\rangle$

then $\langle \varphi | e^{-\tau X} | \psi \rangle \geq 0$ for all $\tau > 0$

- **Auxiliary-field QMC:**

$$F \propto \det(1 + e^{A_1} \dots e^{A_m}) = ?$$



Quantum Monte Carlo Minus Sign Problem

- **Why should negative contributions bother us at all ?**
 - **Importance sampling without (mathematical) justification**
 - **Trick: Transfer sign to measurement**
 - H. De Raedt and A. Lagendijk, Phys. Rev. Lett. 46, 77 (1981)

$$\langle A \rangle = \text{Tr} \rho A = \text{Tr} |\rho| \text{sign}(\rho) A = \frac{\langle \text{sign}(\rho) A \rangle_{|\rho|}}{\langle \text{sign}(\rho) \rangle_{|\rho|}}$$

- **Simulation methods suffer from statistical errors**
 - **This trick fails if $\langle \text{sign}(\rho) \rangle_{|\rho|} \rightarrow 0$ which is usually the case if**
 - the number of factors in the product formula increases, or
 - the system size increases, or
 - the temperature decreases, or
 - ...



Minus Sign Problem: Simple Example

- The minus sign problem can also appear when there are no **fermions** !
- **Example: A spin-1/2 model**

$$-\beta H = a\sigma^x - b\sigma^y = \begin{pmatrix} 0 & a+ib \\ a-ib & 0 \end{pmatrix}$$

Generic QMC approach: Rewrite the partition function using the Trotter-Suzuki formula:

$$Z = \lim_{m \rightarrow \infty} Z_m \quad ; \quad Z_m = \text{Tr} \left(e^{a\sigma^x / m} e^{-b\sigma^y / m} \right)^m$$

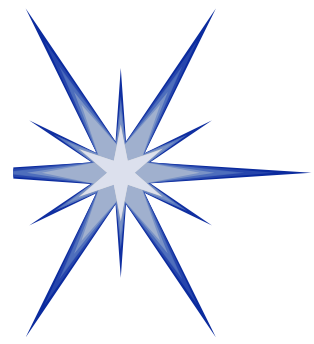
1) $Z_1 = \cosh a \cosh b \text{Tr}(1 + i\sigma^z \tanh a \tanh b)$



None of the contributions is REAL !

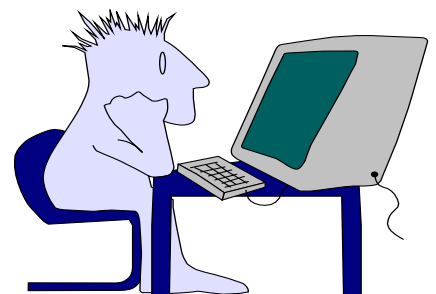
2) Change of representation : $-\beta H = a\sigma^x - b\sigma^z$

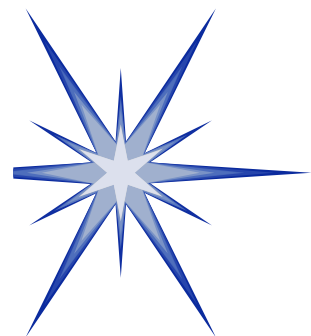
$$Z_m = \sum_{\{s_j = \pm 1\}} \prod_{j=0}^{m-1} \left[\delta_{s_j, s_{j+1}} + (1 - \delta_{s_j, s_{j+1}}) \tanh a \right] e^{-bs_j/m} \cosh \frac{a}{m}$$



Quantum Monte Carlo Methods: Summary

- Quantum Monte Carlo methods work well if there is no **minus sign** problem
- The minus sign problem is due to the use of a **product formula** in combination with an **importance sampling algorithm**
- Use of importance sampling seems unavoidable (just too many possibilities)
- Without using some product formula it seems very difficult to compute physical quantities without making (uncontrolled) approximations





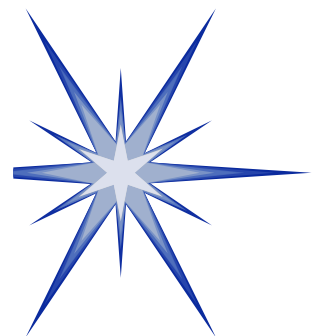
Stochastic Diagonalization

- **Method to compute the smallest eigenvalue and corresponding eigenvector of very (very) large matrices.**
 - **Basic idea: Assume that of all possible states of the quantum system, only a relatively small fraction is “important”**
 - The same assumption is at the heart of all Quantum Monte Carlo techniques
 - How to find these important states ?
 - HDR & W. von der Linden (1992):

$$|\Phi\rangle = \sum_{j=1}^M a_j |\phi_j\rangle = \sum_{j=1}^M a_{P_j} |\phi_{P_j}\rangle \quad ; \quad |a_{P_j}| \geq |a_{P(j+1)}|$$

$$|\Phi\rangle \approx |\tilde{\Phi}\rangle = \sum_{j=1}^{M_{imp}} a_{P_j} |\phi_{P_j}\rangle \quad ; \quad M_{imp} = \# \text{ Important states}$$

$$E \leq \tilde{E} = \frac{\langle \tilde{\Phi} | H | \tilde{\Phi} \rangle}{\langle \tilde{\Phi} | \tilde{\Phi} \rangle}$$



Stochastic Diagonalization

- **Lemma: If** $U^{-1} \begin{pmatrix} a & b \\ b^* & c \end{pmatrix} U \rightarrow \begin{pmatrix} a' & 0 \\ 0 & c' \end{pmatrix}$ **and if**

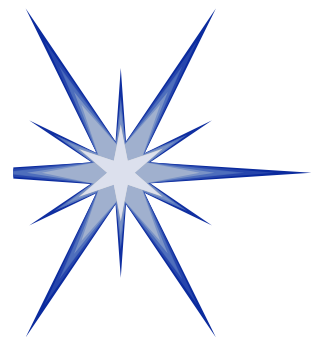
$$a \leq c \quad \text{and} \quad |b| \neq 0 \quad \text{then} \quad a' < a$$

- **Step 1: Modified Jacobi method**

$$U_{1,k}^{-1} \begin{pmatrix} a & b_1 & \dots & b_k & \dots & b_M \\ b_1^* & \bullet & \bullet & \bullet & \bullet & \bullet \\ \vdots & \bullet & \bullet & \bullet & \bullet & \bullet \\ b_k^* & \bullet & \bullet & \bullet & \bullet & \bullet \\ \vdots & \bullet & \bullet & \bullet & \bullet & \bullet \\ b_M^* & \bullet & \bullet & \bullet & \bullet & \bullet \end{pmatrix} U_{1,k} \rightarrow \begin{pmatrix} a' & b'_1 & \dots & 0 & \dots & b'_M \\ b_1'^* & \bullet & \bullet & \bullet & \bullet & \bullet \\ \vdots & \bullet & \bullet & \bullet & \bullet & \bullet \\ 0 & \bullet & \bullet & \bullet & \bullet & \bullet \\ \vdots & \bullet & \bullet & \bullet & \bullet & \bullet \\ b_M'^* & \bullet & \bullet & \bullet & \bullet & \bullet \end{pmatrix} \Rightarrow a' < a$$

- **Repeat until** $\rightarrow \begin{pmatrix} E_0 & 0 & \dots & 0 & \dots & 0 \\ 0 & \bullet & \bullet & \bullet & \bullet & \bullet \\ \vdots & \bullet & \bullet & \bullet & \bullet & \bullet \\ 0 & \bullet & \bullet & \bullet & \bullet & \bullet \\ \vdots & \bullet & \bullet & \bullet & \bullet & \bullet \\ 0 & \bullet & \bullet & \bullet & \bullet & \bullet \end{pmatrix}$

- **Step 2: Use the difference** $a - a' > 0$ **to set up an importance sampling scheme**

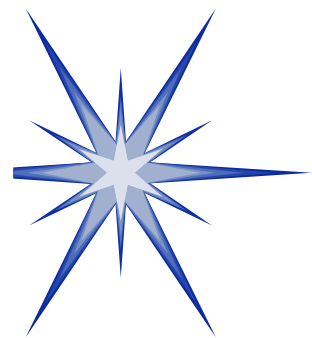


Stochastic Diagonalization

- Use plane rotations (2x2 matrices) to deflate the matrix and to simultaneously isolate the lowest eigenvalue

$$\begin{aligned} (E_0^{(0)}) &\rightarrow \begin{pmatrix} E_0^{(1)} & * \\ * & \bullet \end{pmatrix} \rightarrow \\ &\rightarrow \begin{pmatrix} E_0^{(2)} & * & * \\ * & \bullet & \bullet \\ * & \bullet & \bullet \end{pmatrix} \rightarrow \dots \rightarrow \begin{pmatrix} E_0 & 0 & \dots & 0 \\ 0 & \bullet & \bullet & \bullet \\ \vdots & \bullet & \bullet & \bullet \\ 0 & \bullet & \bullet & \bullet \end{pmatrix} \end{aligned}$$

- How to pick “candidates” ?
 - Use $\langle \varphi_{Pi} | H | \varphi_{Trial} \rangle \neq 0$
- Why does the (1,1) element converges to the ground state energy ?



Stochastic Diagonalization

- **Cauchy theorem:**

$$H = \begin{pmatrix} H^{(k)} & X^T \\ X & Y \end{pmatrix} ; \quad k = 1, \dots, M - 1$$

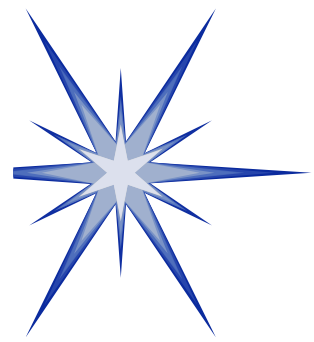
$E_j^{(k)}$: j -th eigenvalue of $H^{(k)}$

Then :

$$E_j \leq E_j^{(k)} \leq E_{M-k+j}$$

- **Poincaré separation theorem:**
Relates eigenvalues of a $(k-1) \times (k-1)$ matrix to the eigenvalues of a $k \times k$ matrix

$$E_1^{(k)} \leq E_1^{(k-1)} \leq E_2^{(k)} \leq \dots \leq E_{k-1}^{(k-1)} \leq E_k^{(k)}$$



SD: Algorithm

- **Initialize data structures**
- **Do**
- **If (Maximum of absolute value of off-diagonal elements of the first row < than threshold for rejecting plane rotations)**
then
 - **Generate a new trial state**
- **if (No important state has been found) then**
 - **Reduce the threshold(s)**
 - **else**
 - **Inflate the matrix**
 - **end if**
- **else**
 - **Annihilate the pair of off-diagonal elements with the largest absolute value by performing a plane rotation**
 - **end if**
- **end do**



SD : Practical aspects

- **Necessary condition to be useful in practice: The calculation of matrix elements of H should be (very) fast.**
 - **Each attempt to find a new important state requires the calculation of the matrix element**

$$\begin{aligned} H_{1,n+1}^{trial} &= ((U^{(1)} \dots U^{(p)})^T H U^{(p)} \dots U^{(1)})_{1,n+1} \\ &= \sum_{j=1}^n (U^{(1)} \dots U^{(p)})_{1,j}^T H_{j,n+1} \end{aligned}$$

- **Calculation of the product of plane rotation matrices $U^{(q)}$ is fast**
- **Calculating $H_{i,n+1}$ takes most of the CPU time**
 - **Can be done in parallel**



Computation of Physical Properties

- At $T=0$ the expectation value of a physical observable is given by

$$\begin{aligned}\langle A \rangle &= \langle \Phi | A | \Phi \rangle = \sum_{i,j=1}^M a_i a_j \langle \phi_i | A | \phi_j \rangle \\ &\approx \frac{\sum_{i,j=1}^{M_{imp}} a_{P_i} a_{P_j} \langle \phi_{P_i} | A | \phi_{P_j} \rangle}{\sum_{i=1}^{M_{imp}} a_{P_i}^2}\end{aligned}$$

- Takes of the order of $M_{imp} \times M_{imp}$ operations
 - For $M_{imp} \sim 10^5 - 10^6$ computing a physical quantity may take a substantial amount of CPU time



Example: Repulsive Hubbard Model

- **2D Hubbard model with nearest and next-nearest neighbor hopping**

$$\begin{aligned}
 H &= -t \sum_{\langle i,j \rangle} \sum_{\sigma=\uparrow,\downarrow} (c_{i,\sigma}^+ c_{j,\sigma} + c_{j,\sigma}^+ c_{i,\sigma}) + U \sum_i c_{i\uparrow}^+ c_{i\downarrow}^+ c_{i\downarrow} c_{i\uparrow} \\
 &\quad -t' \sum_{\langle\langle i,j \rangle\rangle} \sum_{\sigma=\uparrow,\downarrow} (c_{i,\sigma}^+ c_{j,\sigma} + c_{j,\sigma}^+ c_{i,\sigma}) \\
 &= \sum_k \sum_{\sigma=\uparrow,\downarrow} \varepsilon_k c_{k,\sigma}^+ c_{k,\sigma} + \frac{U}{L} \sum_{k,p,q} c_{k+q,\uparrow}^+ c_{p-q,\downarrow}^+ c_{p,\downarrow} c_{k,\uparrow}
 \end{aligned}$$

- **Ground state energy per site of 10 electrons on a 4x4 lattice filled as obtained by exact diagonalization (ED), stochastic diagonalization (SD), and projector quantum Monte Carlo (PQMC).**

$t' / t $	$U / t $	<i>ED</i>	<i>SD</i>	<i>PQMC</i>
0	-6	-2.458782	-2.4568	-2.460 ± 0.004
0	-4	-2.045849	-2.0453	-2.045 ± 0.002
0	-2	-1.731689	-1.7316	-1.731 ± 0.001
-0.22	2	-1.230034	-1.2300	-1.231 ± 0.001
-0.22	4	-1.126160	-1.1261	-1.125 ± 0.003
-0.22	6	-1.058717	-1.0581	-1.061 ± 0.005

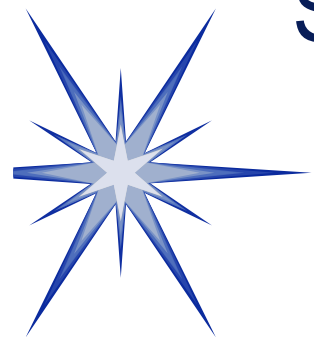


Example: Quantum Chemistry

- Energy and correlation energy (both in hartree) of the Mg atom as obtained by SD and various standard quantum chemistry methods. M_I denotes the number of the basis states (Slater determinants with D_{2h} symmetry) that has been used

<i>Method</i>	<i>E</i>	<i>E_{corr}</i>	<i>M_I</i>
<i>SCF</i>	-199.585212	0	1
<i>RASSCF</i>	-199.615701	-0.0305	4
<i>CISD</i>	-199.721386	-0.1362	2960
<i>CISDT</i>	-199.722039	-0.1368	102928
<i>CISDTQ</i>	-199.726256	-0.1410	1964232
<i>SD</i>	-199.724237	-0.1409	2000
<i>SD</i>	-199.726164	-0.1410	40000
<i>CI*</i>	-199.7263	-0.1411	2538603250

‡ J.Olsen et al., Chem. Phys. Lett. 169, 463 (1990)

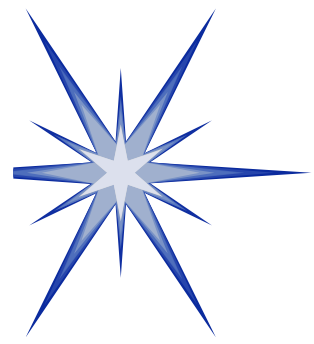


Stochastic Diagonalization versus Fixed-Node Diffusion Monte Carlo

- **Comparison between the ground state energy per site as obtained from a fixed-node approximation (FN), exact diagonalization (ED) and stochastic diagonalization (SD). The difference between the FN results of the second and third row stems from the choice of the trial state.**

Lattice	$U/ t $	FN*	ED	SD	M
2x2 $N_{\uparrow} = N_{\downarrow} = 2$	1	-3.3172	-3.3408	-3.3408 ^(a)	36
2x2x2 $N_{\uparrow} = N_{\downarrow} = 4$	10	-2.6597	-2.8652	-2.8634 ^(b)	4900
2x2x2 $N_{\uparrow} = N_{\downarrow} = 4$	10	-2.6382	-2.8652	-2.8634 ^(b)	4900

- ◆ **(A) : D.F.B. ten Haaf *et al.*, Phys. Rev. B (1995)**
 - (a) : Using < 100 s CPU-time on an IBM Thinkpad 755CD collecting 10 states, without using symmetry
 - (b) : Using < 1000 s CPU-time on an IBM Thinkpad 755CD collecting 600 states, without using symmetry



Stochastic Diagonalization: Main features

- Exact results (but requires much more effort than standard methods) for small systems because then we can put $M_{imp} = M$ from the start
- The “best” variational wave function that can be build from M_{imp} basis states
- **NO** minus sign problems
- “Quality” of the result depends on the choice of the representation (basis states)
 - Use real- and momentum space representation (very hard for QMC)
 - Checks: Compare with exact diagonalization (small systems) and Projector Quantum Monte Carlo



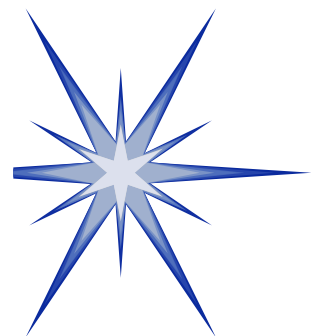
Application: Off-Diagonal Long-Range Order

- **Can a model system of (strongly) interacting electrons exhibit superconductivity ?**
 - No degrees of freedom other than those of the electrons
 - Dominant interaction between the electrons is repulsive
- **How would theoreticians know the system is in the superconducting state ?**
 - **Criteria:**
 - Infinite electrical conductivity
 - AND**
 - Off-Diagonal Long-Range order **OR**
 - Flux Quantization **OR**
 - Superfluid density



Off-Diagonal Long-Range Order (ODLRO)

- **Penrose & Onsager (1951) :**
 - Properties of the reduced one-particle density matrix can be used to “detect” Bose-Einstein condensation
- **Yang (1962) :**
 - Reduced two-particle density matrix can be used to “detect” superconductivity
 - “Detect” : ODLRO in the two-particle density matrix → superconductivity (zero resistance & Meissner effect)
 - Sewell (1990); Nieh, Su, Zhao (1995): ODLRO (+ ...) → Meissner effect

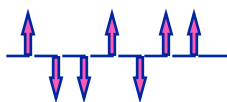


ODLRO

- **Ising Spin model**
- **Phase transition**
 - $T > T_c$: There is little correlation between spins on different sites
 - $T < T_c$: Strong correlation between spins on different sites

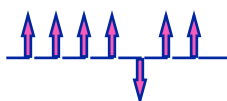
- **Quantitative measure of correlation:**

$$C(j) = \frac{1}{L} \sum_{i=1}^L \langle S_i S_{i+j} \rangle$$



$$T > T_c \Rightarrow C(j) \approx e^{-j/\lambda} \Rightarrow \sum_{j=1}^L C(j) = a$$

$$T < T_c \Rightarrow C(j) \approx 1 \Rightarrow \sum_{j=1}^L C(j) = a + bL$$



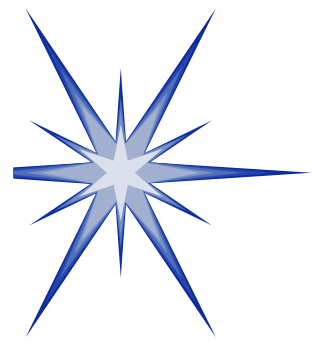
- **Fermions**
- **Normal ↔ super-conductor phase transition: Pairs of electrons**

- Yang: No long-range correlation in one-particle density matrices
- Correlation functions or reduced two-particle density matrix

$$C(j) = \frac{1}{L} \sum_{i=1}^L \langle c_{i,\uparrow}^+ c_{i+k,\downarrow}^+ c_{i+k+j,\downarrow} c_{i+j,\uparrow} \rangle$$

- k determines the “kind” of electron pairing
 - $k=0$ → on-site s-wave pairing

- **ODLRO** : $\sum_{j=1}^L C(j) = a + bL$

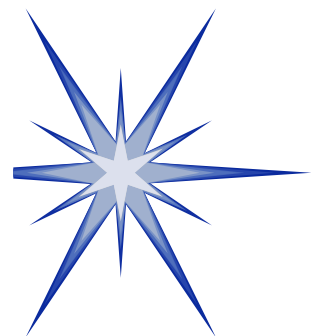


ODLRO: Full two-particle density matrix

- **Yang: Compute all entries of the reduced two-particle density matrix**

$$A(n, m) = \langle c_{i, \sigma}^+ c_{i', -\sigma}^+ c_{j', -\sigma'} c_{j, \sigma'} \rangle \quad ; n = (i, i', \sigma), m = (j, j', \sigma')$$

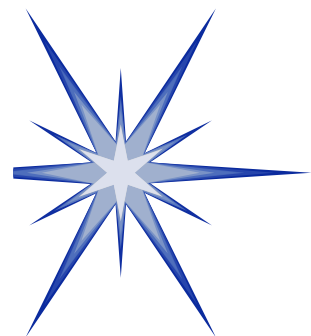
- **Compute $A(n, m)$: $2L^2 \times 2L^2$ matrix**
- **Find the largest eigenvalue λ_0 of A**
- **If $\lambda_0 = a + bL$ then there is ODLRO**
- **The eigenvector corresponding to λ_0 contains the information on the kind of pairing**
- **Analogy with magnetic moments:**
 - **Diagonalize $A(n, m) = \langle S_n S_m \rangle$ by Fourier transformation**
 - **$q=0 \Rightarrow$ ferromagnetic**
 - **$q=\pi \Rightarrow$ antiferromagnet**



BCS reduced Hamiltonian

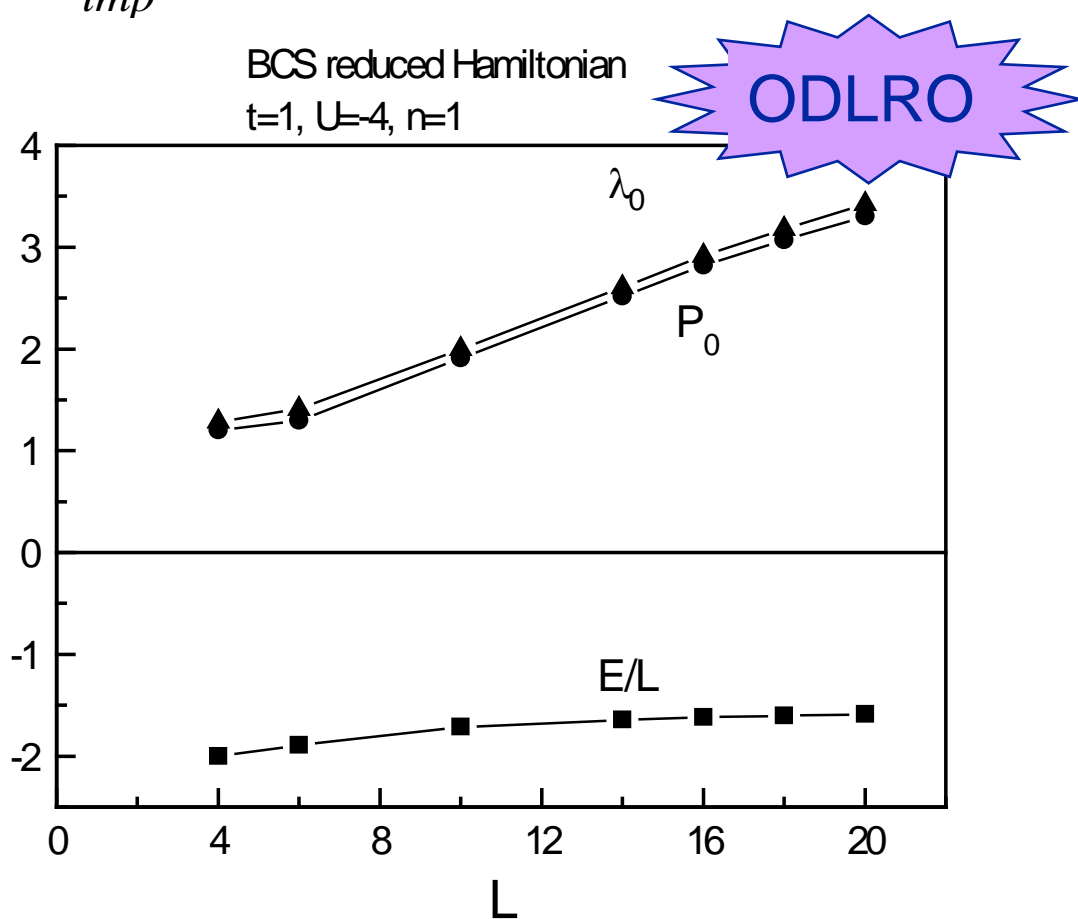
$$\begin{aligned} H_{BCS} &= -t \sum_{\langle i,j \rangle} \sum_{\sigma=\uparrow,\downarrow} (c_{i,\sigma}^+ c_{j,\sigma} + c_{j,\sigma}^+ c_{i,\sigma}) - \frac{|U|}{L} \sum_{i,j} c_{i\uparrow}^+ c_{i\downarrow}^+ c_{j\downarrow} c_{j\uparrow} \\ &= \sum_k \sum_{\sigma=\uparrow,\downarrow} \varepsilon_k c_{k,\sigma}^+ c_{k,\sigma} - \frac{|U|}{L} \sum_{k,p} c_{k,\uparrow}^+ c_{-k,\downarrow}^+ c_{-p,\downarrow} c_{p,\uparrow} \end{aligned}$$

- **For the BCS reduced Hamiltonian the BCS treatment yields the exact solution (in the thermodynamic limit)**
 - R.J. Bursill and C.J. Thompson (1993)
- **This system has to exhibit ODLRO**
 - the BCS wave function has ODLRO build in



BCS reduced Hamiltonian

SD: $M_{imp} \leq 100000$



On - site pairing correlation function

$$P_0 = \frac{1}{L} \sum_{i,k} \langle c_{i\uparrow}^+ c_{i\downarrow}^+ c_{k\downarrow} c_{k\uparrow} \rangle$$

$$P_0 \leq \lambda_0$$

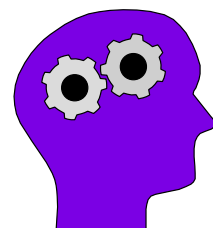
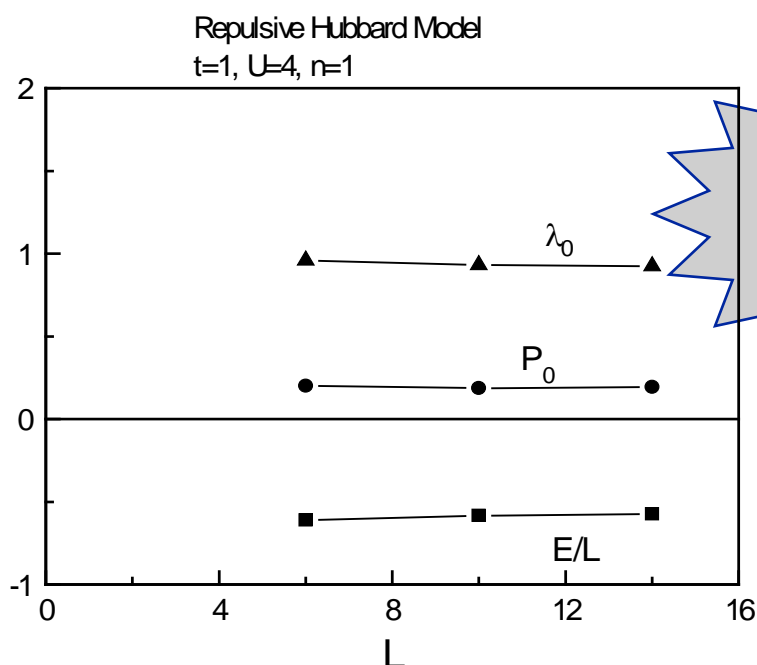


Repulsive Hubbard Model

$$H = -t \sum_{\langle i,j \rangle} \sum_{\sigma=\uparrow,\downarrow} (c_{i,\sigma}^+ c_{j,\sigma} + c_{j,\sigma}^+ c_{i,\sigma}) + |U| \sum_i c_{i\uparrow}^+ c_{i\downarrow}^+ c_{i\downarrow} c_{i\uparrow}$$
$$= \sum_k \sum_{\sigma=\uparrow,\downarrow} \varepsilon_k c_{k,\sigma}^+ c_{k,\sigma} + \frac{|U|}{L} \sum_{k,p,q} c_{k+q,\uparrow}^+ c_{p-q,\downarrow}^+ c_{p,\downarrow} c_{k,\uparrow}$$

- **BCS treatment: No ODLRO**
- **SD: $M_{imp} \leq 198000$**

$$M \leq 1.2 \times 10^6$$





Repulsive Hubbard Model with Correlated Hopping

$$\begin{aligned} H &= -t \sum_{\langle i,j \rangle \sigma = \uparrow, \downarrow} (c_{i,\sigma}^+ c_{j,\sigma} + c_{j,\sigma}^+ c_{i,\sigma}) + U \sum_i n_{i\uparrow} n_{i\downarrow} \\ &\quad + \Delta t \sum_{\langle i,j \rangle \sigma = \uparrow, \downarrow} (n_{i,-\sigma} + n_{j,-\sigma}) (c_{i,\sigma}^+ c_{j,\sigma} + c_{j,\sigma}^+ c_{i,\sigma}) \\ &= \sum_k \sum_{\sigma = \uparrow, \downarrow} \varepsilon_k c_{k,\sigma}^+ c_{k,\sigma} \\ &\quad + \frac{1}{L} \sum_{k,p,q} \left[U - \frac{\Delta t}{t} (\varepsilon_k + \varepsilon_{k+q}) \right] c_{k+q,\uparrow}^+ c_{p-q,\downarrow}^+ c_{p,\downarrow} c_{k,\uparrow} \end{aligned}$$

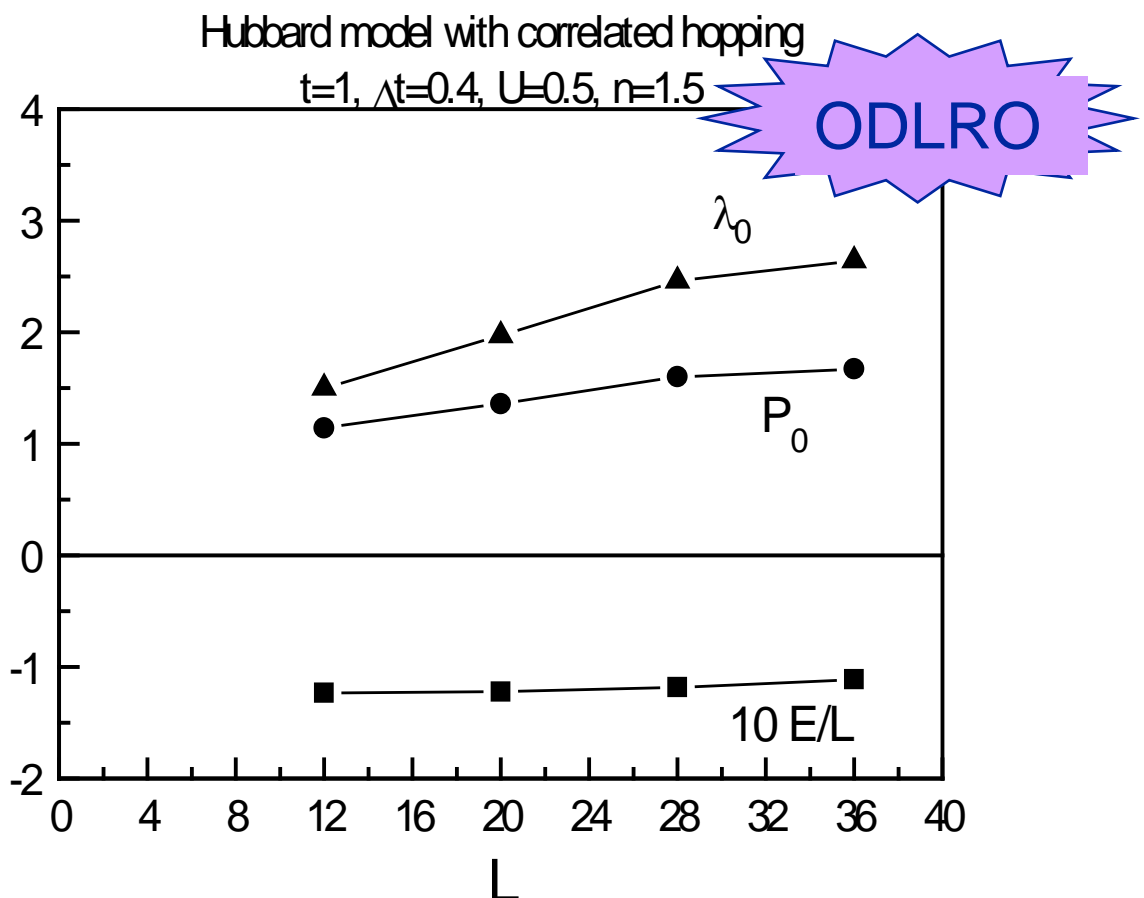
- Interaction may be “attractive” for some $(U, t, \Delta t, n)$
- Hirsch (1989): Basic model for (high- T_c) superconductivity (based on BCS treatment)
- Exact solution for $(U, t = \Delta t, n = 1)$: ODLRO for $U < -4/t$ but NO superconductivity
- Japaridze & Müller-Hartmann (1994): Continuum limit + Bosonization: ODLRO for $U \ll \Delta t \ll t, n_0 < n < n_1$



Repulsive Hubbard Model with Correlated Hopping

- SD: $M_{imp} \leq 192000$

$$M \leq 8.7 \times 10^{19}$$





Application: Summary

- **Off-Diagonal Long-Range Order (implies superconductivity ?) in models of interacting fermions.**
 - **The repulsive Hubbard model supplemented with correlated hopping may exhibit Off-Diagonal Long-Range Order, depending on the choice of $(U, t, \Delta t, n)$**
 - **Building the many-body states from single-particle states in either the real-space or k-space representation is far from optimal for this kind of problem**



Basis set optimization

- **SD can use “any” representation**
 - if $\langle \varphi | H | \psi \rangle$ can be calculated efficiently
 - why not use basis sets that are “more sophisticated” ?
- **An attempt motivated by our study of ODLRO**
 - real-space representation is adequate to describe e.g. local pairs
 - k-space is the representation of choice to describe e.g. extended structure
- **Can we let the algorithm find an optimum that interpolates between these two extremes ?**
 - basis set optimization (cfr. Quantum chemistry)



Optimization of the single-particle states

- **Example: Fermions on a lattice (atomic orbitals)**
- **Many-body basis states are build from single-particle states**

- **Real-space** : $|\varphi_i\rangle = c_i^+ |0\rangle$
- **Fourier-space** : $|\varphi_k\rangle = c_k^+ |0\rangle$
- **Many-body state (Slater determinant):**

$$|\phi_n\rangle = \sum_{i_j=1}^K a^{(n)}(i_1,1) \dots a^{(n)}(i_N,1) c_{i_1}^+ \dots c_{i_N}^+ |0\rangle$$

- **Matrix of coefficients:**

$$A^{(n,K)} = \begin{pmatrix} a^{(n)}(1,1) & a^{(n)}(1,2) & \dots & a^{(n)}(1,N) & \dots & a^{(n)}(1,K) \\ a^{(n)}(2,1) & a^{(n)}(2,2) & \dots & a^{(n)}(2,N) & \dots & a^{(n)}(2,K) \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ a^{(n)}(K,1) & a^{(n)}(K,2) & \dots & a^{(n)}(K,N) & \dots & a^{(n)}(K,K) \end{pmatrix}$$

- **e.g. “real space” :**

$$A^{(n,K)} = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{pmatrix}$$



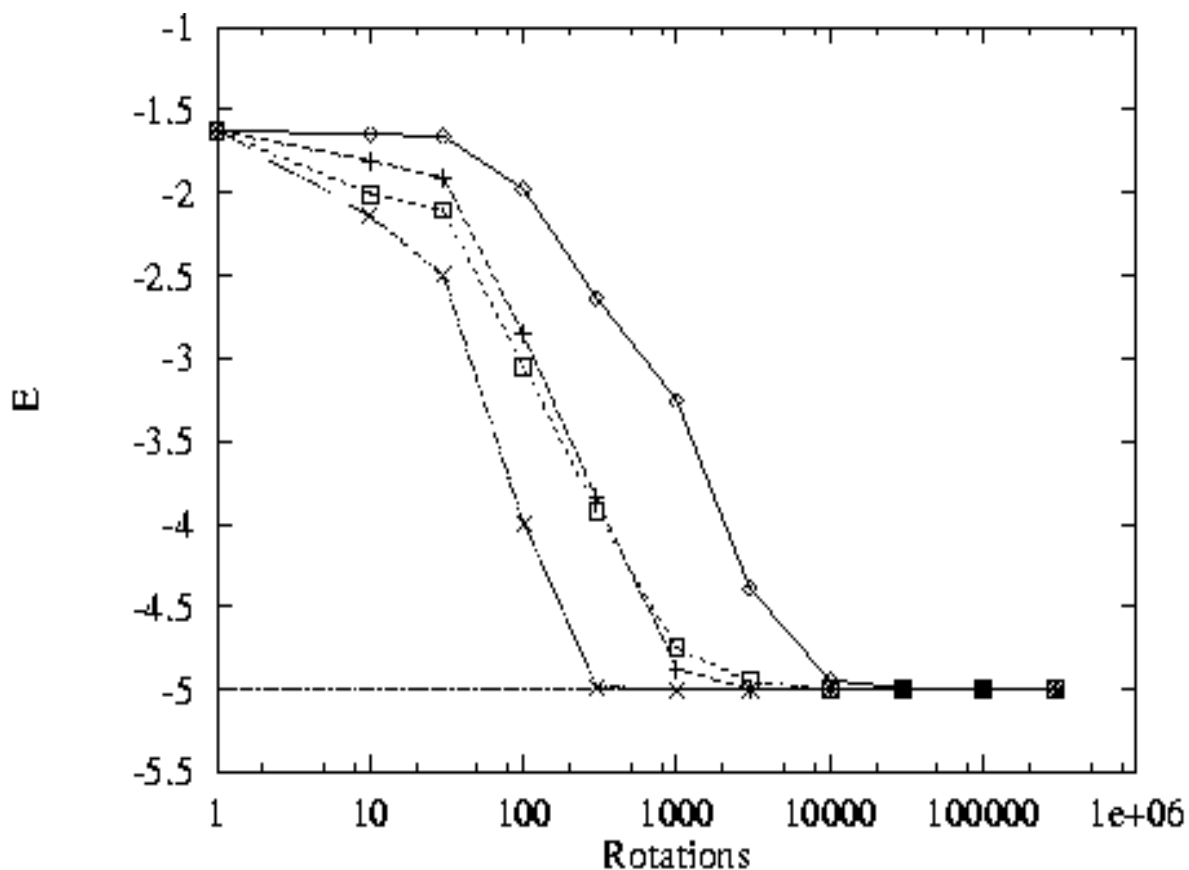
Optimization of the single particle states

- **Change the matrix $A^{(n,K)}$ using “simple” moves**
 - **keep single-particle states orthonormal**
 - Simplifies calculation of matrix elements of H
 - **Use plane-rotations:**
$$U = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$
 - **Update two rows (columns) p,q at a time**
 - **Mix columns:** $A^{(n,K)} \leftarrow A^{(n,K)} U^{(p,q)}$; $1 \leq p \leq N < q \leq K$
 - **Mix rows:** $A^{(n,K)} \leftarrow U^{(p,q)} A^{(n,K)}$; $p, q = 1, K$
 - **Preserves unitary character of $A^{(n,K)}$ and defines a “dynamics” for changing Slater determinants**



Basis-set optimization: Example

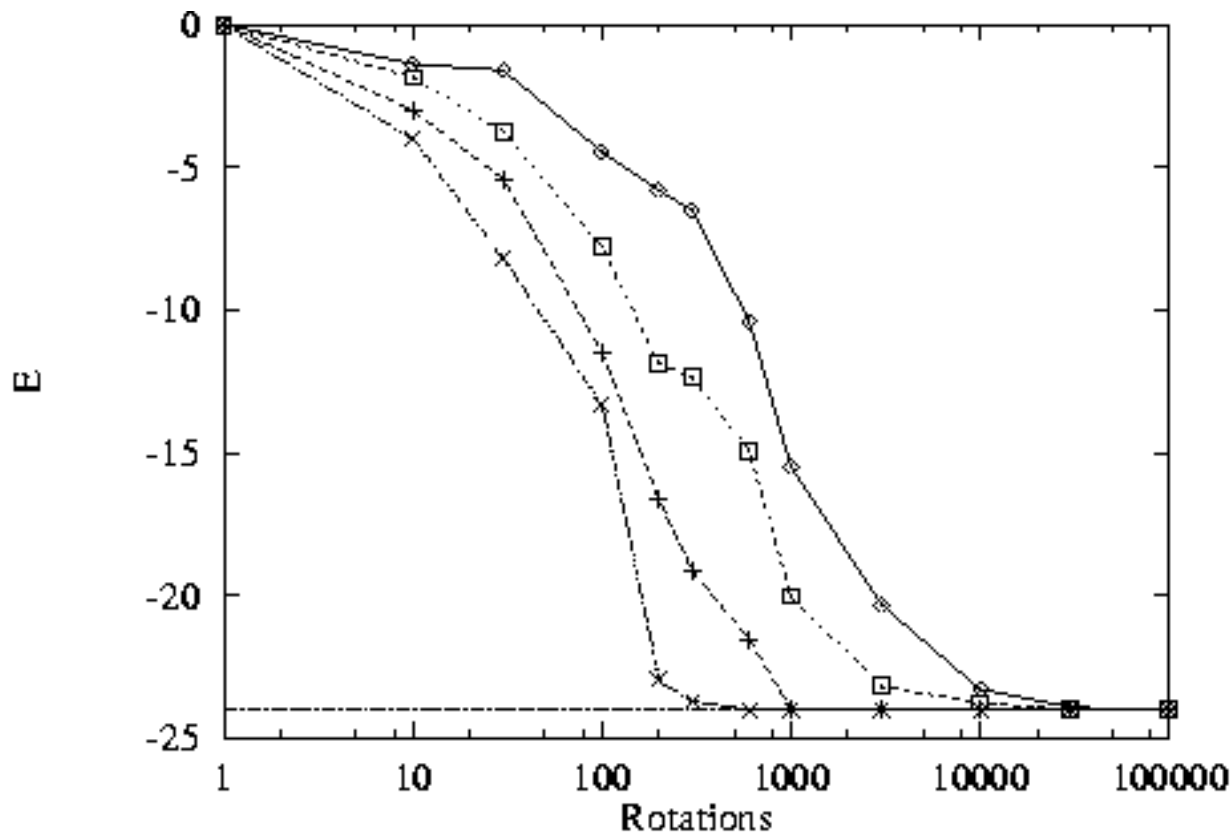
- The energy E as a function of the number of rotations as obtained by optimizing one Slater determinant for the case of a 4x4 Hubbard model ($N_\uparrow = N_\downarrow = 5, t = 0, U = -1$). The single-particle states used to construct the initial many-body wave function are $|\varphi_k\rangle = c_k^+ |0\rangle$. Various symbols correspond to different optimization strategies. The exact result is indicated by the dashed-dotted line.





Basis-set optimization: Example

- The energy E as a function of the number of rotations as obtained by optimizing one Slater determinant for the case of a 4x4 Hubbard model ($N_{\uparrow} = N_{\downarrow} = 5, t = 1, U = 0$). The single-particle states used to construct the initial many-body wave function are $|\varphi_i\rangle = c_i^+ |O\rangle$. Various symbols correspond to different optimization strategies. The exact result is indicated by the dashed-dotted line.

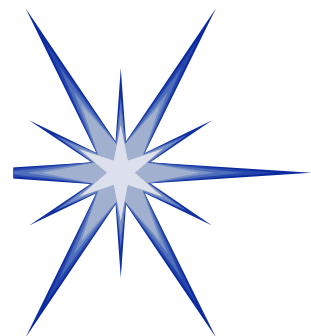




Basis-set optimization: Example

- **The ground state energy of the 4x4 Hubbard model ($N_{\uparrow} = N_{\downarrow} = 5, t = 1$) as obtained by various methods:**
 - E_1^{Exact} : Exact diagonalization
 - $E_1^{(k)}$: SD + basis-set optimization using $|\varphi_k\rangle = c_k^+|0\rangle$
 - $E_1^{(i)}$: SD + basis-set optimization using $|\varphi_i\rangle = c_i^+|0\rangle$
 - E_1^{SD} : SD
 - E_1^{HF} : Hartree Fock
- **In all SD calculations the limit to the number of important states was set to 200: $M_f = 200$.**
- **A greedy algorithm was used to optimize the single-particle states, using 10000 plane rotations for each many-body state added.**

U	E_1^{Exact}	$E_1^{(k)}$	$E_1^{(i)}$	E_1^{SD}	E_1^{HF}
6	-18.35837	-18.09710	-17.93031	-17.41384	-14.625
4	-19.58094	-19.50157	-19.42179	-19.24434	-17.750
2	-21.37695	-21.36576	-21.34991	-21.32807	-20.875
0	-27.12500	-27.12500	-27.68078	-27.64086	-27.125
-2	-27.70702	-27.69305	-27.68078	-27.64086	-27.125
-4	-32.73360	-32.51795	-32.45901	-32.05033	-30.250
-6	-39.34051	-38.19426	-38.27180	-36.85642	-33.375



Summary

- **General aspects of the sign problem in Quantum Monte Carlo methods**
- **Stochastic diagonalization method**
- **Application: Off-Diagonal Long-Range Order (← superconductivity) in models of interacting fermions.**
 - **The repulsive Hubbard model supplemented with correlated hopping may exhibit Off-Diagonal Long-Range Order, depending on the choice of $(U, t, \Delta t, n)$**
- **Basis-set optimization**