

Exact Diagonalization: Introduction

Condensed Matter Summer School 9/5 - 9/9

Matrix

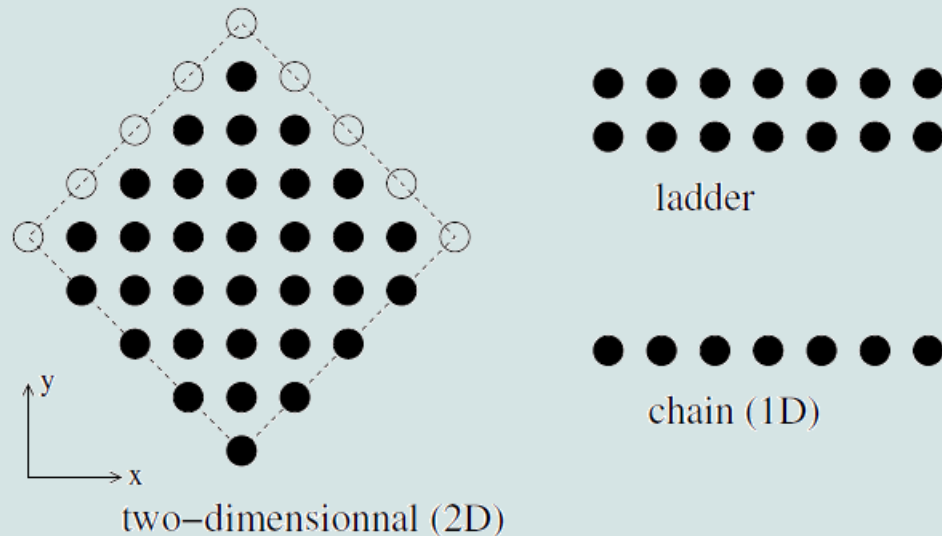
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Ingredients

- ▶ Hilbert space
 - ▶ Basis representation, Lookup techniques
 - ▶ Symmetries
- ▶ Hamiltonian Matrix
 - ▶ Sparse Matrix representation (memory/disk)
 - ▶ Matrix recalculation on the fly (matrix-free)
- ▶ Linear Algebra : Eigensolver / Time propagation
 - ▶ LAPACK full diagonalization
 - ▶ Lanczos type diagonalization (needs only operations)
 - ▶ More exotic eigensolver techniques, imaginary-time propagation,
- ▶ Observables
 - ▶ Static quantities (multipoint correlation functions, correlation density matrices,...)
 - ▶ Dynamic observables (spectral functions, density of states,...)
 - ▶ Real-time evolution

Lattice model

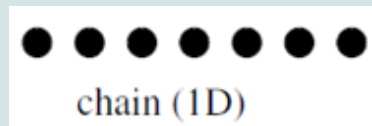
Direct diagonalization of Hamiltonian matrix on finite clusters



- ▶ Hilbert space, Symmetries
- ▶ Hamiltonian matrix
- ▶ Linear Algebra
- ▶ Observables

Representation of Many-Body States

- ▶ Mapping to (binary) integers:
 - ▶ Spin-1/2 Heisenberg:



$$|j, m\rangle = |\uparrow\rangle \otimes \dots \otimes |\downarrow\rangle$$

- ▶ For example

$$|\uparrow\downarrow\uparrow\rangle = |101\rangle = |5\rangle$$

- ▶ Hubbard:

- ▶ For example

$$\begin{array}{c} \uparrow \\ \circ \end{array} \quad \begin{array}{c} \downarrow \\ \circ \end{array} \quad |01\rangle_{\downarrow} |10\rangle_{\uparrow} = 6$$

Symmetries in ED I

- ▶ The inclusion of symmetries in an ED code has two major advantages:
 - ▶ Quantum number resolved energies and states.
 - ▶ Reduction of the Hilbert space to be diagonalized.
- ▶ Given group G with generators $\{g_p\}$
 - ▶ $[H, g_p] = 0 \rightarrow H$ block diagonal (Hilbert space can be divided)

Symmetries in ED II

- ▶ Consider the simplest parent Hamiltonian, *the Heisenberg model* on a chain.

$$H = J \sum_i \mathbf{s}_i \cdot \mathbf{s}_{i+1}$$

- ▶ The dimension of Hilber space is 2^N
 - ▶ For exmaple $N = 20 \rightarrow 2^{20} = 1048576 \cong 10^6$
 - ▶ To store a *real number (double precision)* in computer we need 8 Bytes.
 - ▶ To store the matrix element of the above Hamiltonian we need:
 - ▶ Number of bytes = $8 \cdot (10^6)^2 \cong 8000\text{GB}$
 - ▶ Considering the S_z symmetry
 - ▶ For example, the most amount of basis is total $S_z = 0$
 - ▶ 10 up-spin, 10 down-spin
- $$C_{10}^{20} = \frac{20!}{10!10!} = 184756 \cong 10^5$$
- ▶ Number of bytes = $8 (10^5)^2 \cong 80\text{GB}$

Symmetries in ED III

- ▶ Continuous:
 - ▶ $U(1)$ related symmetries
 - ▶ Conservation of particle number(s)
 - ▶ Conservation of total S_z
 - ▶ $SU(2)$ symmetry
 - ▶ Difficult to implement together with spatial symmetries
 - ▶ For $S_z = 0$ an operation called “spin inversion (Z_2)” splits the Hilbert space in even and odd spin sectors.
- ▶ Spatial symmetry groups:
 - ▶ Translation symmetry (abelian symmetry, therefore 1D irreps)
 - ▶ Pointgroup symmetries (in general non-abelian)

Sparse Matrix

$$H = \begin{pmatrix} 0 & & & & & & & & \\ & 0 & -t & & & & & & \\ & -t & 0 & & & & & & \\ & & & 0 & -t & & & & \\ & & & -t & 0 & & & & \\ & & & & & 0 & & & \\ & & & & U & -t & -t & & \\ & & & -t & 0 & & & -t & \\ & & & -t & & 0 & & -t & \\ & & & & -t & -t & U & & \\ & & & & & & & 0 & \\ & & & & & & U & -t & \\ & & & & & & -t & U & \\ & & & & & & & & U & -t \\ & & & & & & & & -t & U \\ & & & & & & & & & & 2U \end{pmatrix}$$

Hamiltonian Matrix Storage

- ▶ Different possibilities exist:
 - ▶ Store Hamiltonian matrix elements in memory in a sparse matrix format
 - ▶ Fast matrix vector multiplies, but obviously limited by available memory.
 - ▶ Store Hamiltonian matrix elements on disk in a sparse matrix format.
 - ▶ In principle possible due to the vast disk space available, but I/O speed is much slower than main memory access times. Difficult to parallelize.
 - ▶ Recalculate the Hamiltonian matrix elements in each iterations “on the fly”.
 - ▶ Needed for the cutting edge simulations, where the whole memory is used by the Lanczos vectors. Can be parallelized on most architectures.

Diagonalization routine

- ▶ If H is dense or system small enough
 - ▶ Use
 - ▶ Jacobi
 - ▶ Householder
 - ▶ LAPACK
 - ▶ ...
 - ▶ (all these apply orthogonal transformations to H until tridiagonal form, then quickly diagonalize)
- ▶ If H is sparse
 - ▶ Use
 - ▶ ARPACK
 - ▶ IETL/ALPS
 - ▶ DiagHam
 - ▶ (these are iterative solvers based on variants of Lanczos algorithm which preserves the sparseness of H)

LAPACK

- ▶ Complete diagonalization routine written in F77 (Fortran)
- ▶ Support all kind of data structure
 - ▶ Real, Real (double precision), complex, complex (double precision)
- ▶ Support most of matrix formation
 - ▶ Symmetric, Hermitian, Tri-diagonal, ...
- ▶ Useful for:
 - ▶ Simple problems, testing
 - ▶ Matrix H dense
 - ▶ Many eigenstates required
- ▶ But
 - ▶ H must be stored
 - ▶ Entire matrix must be diagonalized

Linear Algebra I

- ▶ Lanczos algorithm:
 - ▶ C. Lanczos, J. Res. Natl. Bur. Stand. 45, 255 (1950).
 - ▶ Iterative “Krylov” space method which brings matrices into **tridiagonal** form.
 - ▶ Method of choice in many large-scale ED programs.
 - ▶ Works well for sparse, short-range H
 - ▶ Very rapid convergence but numerically unstable (can be controlled by a suitable techniques).
 - ▶ Easy to implement and memory efficient – only 3 vectors needed at once
 - ▶ If the ground state (energy) or some low energy spectrum is interesting, the Lanczos method may help us. Using Lanczos and considering the $S_z = 0$ symmetry of the model above, the required memory to get the ground state (energy) and few low lying states is:
 - ▶ Number of bytes $\cong 30 \times 184756 = 5542680 \text{ Bytes} \cong 55 \text{ MB}$
 - ▶ The required memory is available as RAM in any personal computer.

Linear Algebra II

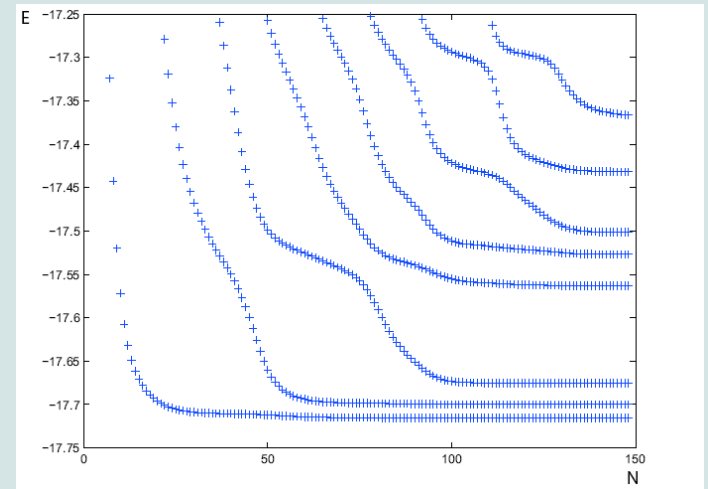
- ▶ “modified Lanczos” algorithm:
 - ▶ E. Gagliano, et al. Phys. Rev. B 34, 1677 (1986).
 - ▶ actually more like a Power-method, therefore rather slow convergence.
 - ▶ needs only two vectors in memory
 - ▶ At each step the approximate groundstate wavefunction is available
 - ▶ Difficult to get excited states
- ▶ Jacobi-Davidson Algorithm:
 - ▶ E. R. Davidson, Comput. Phys. 7, 519 (1993).
 - ▶ subspace expanded by diagonal approximation to inverse iteration
 - ▶ varying number of vectors in memory
 - ▶ higher-order convergence than Lanczos (usually)
 - ▶ Rapid convergence, especially for diagonally dominant matrices (Hubbard model)
 - ▶ Often used in DMRG programs as well.

Lanczos

$$\begin{aligned}
 b_0 &= 0 & |\phi'\rangle &= H|\phi_n\rangle - b_n|\phi_{n-1}\rangle & \blacktriangleright & \text{At each step } n, \text{ find the lowest eigenvalue of that matrix} \\
 a_n &= \langle \phi_n | \phi' \rangle \\
 |\phi''\rangle &= |\phi'\rangle - a_n |\phi_n\rangle & \blacktriangleright & \text{Stop when the lowest eigenvalue } E_0 \text{ has converged } (\Delta E_0/E_0 < 10^{-12}) \\
 b_{n+1} &= \|\phi''\| = \sqrt{\langle \phi'' | \phi'' \rangle} \\
 |\phi_{n+1}\rangle &= |\phi''\rangle / b_{n+1}
 \end{aligned}$$

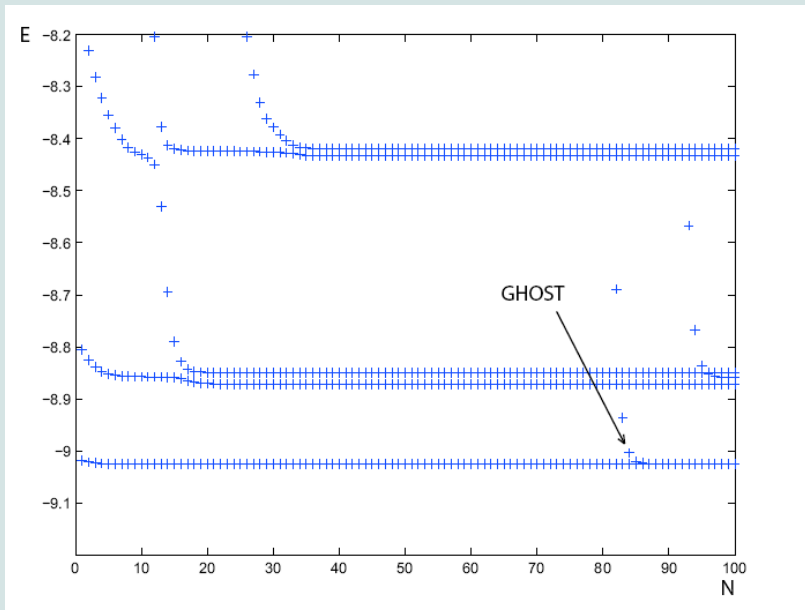
$$H^N |\phi_0\rangle = a_N |\phi_N\rangle, \quad |N\rangle = \frac{|\phi_N\rangle}{\sqrt{\langle \phi_N | \phi_N \rangle}}, \quad \text{Krylov space}$$

$$T = \begin{matrix} \langle 1 | \\ \langle 2 | \\ \langle 3 | \\ \langle \dots | \\ \langle N-1 | \\ \langle N | \end{matrix} \begin{pmatrix} a_1 & b_2 & & & & \\ b_2 & a_2 & b_3 & & & \\ & b_3 & a_3 & b_4 & & \\ & & \ddots & \ddots & \ddots & \\ & & & b_{N-1} & a_{N-1} & b_N \\ & & & & b_N & a_N \end{pmatrix}$$



Roundoff errors and ghosts

- ▶ In exact arithmetic the vectors $\{\phi_i\}$ are orthogonal and the Lanczos iterations stop after at most $N - 1$ steps. The eigenvalues of T are then the exact eigenvalues of H .
- ▶ Roundoff errors in finite precision cause a loss of orthogonality so that the matrix T contains extra spurious eigenvalues, called “ghosts”.



Roundoff errors and ghosts

- ▶ There are two ways to deal with that:
 - ▶ Reorthogonalization of the vectors after every step. This requires storing all of the vectors $\{\phi_i\}$ and is memory intensive. (you don't want to do it)
 - ▶ Control of the effects of roundoff.
- ▶ We will discuss the second solution as it is faster and needs less memory.
- ▶ The main effect of roundoff errors is that the matrix T contains extra spurious eigenvalues, called “ghosts”. These ghosts are not real eigenvalues of H .
- ▶ However they converge towards real eigenvalues of H over time and increase their multiplicities.

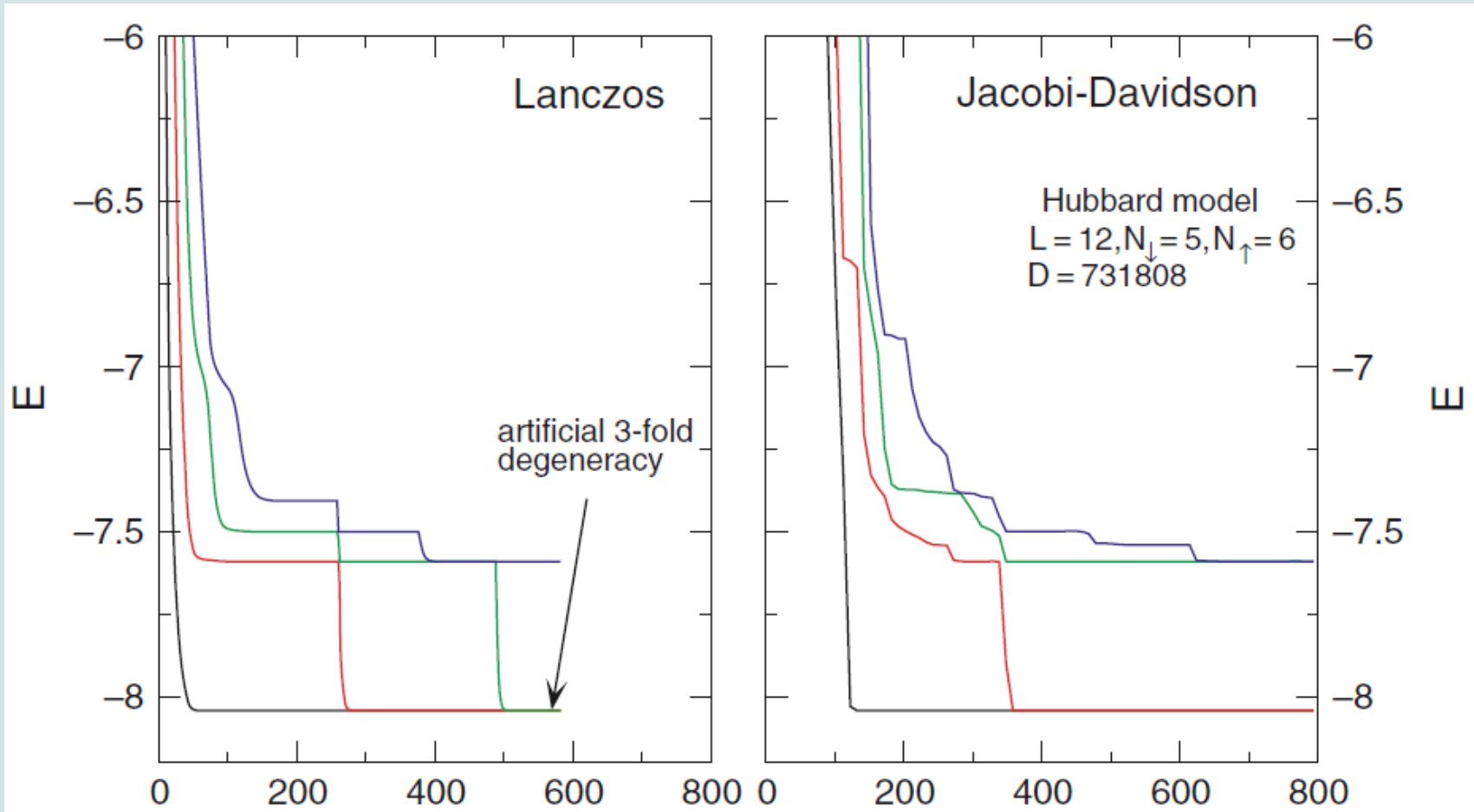
Roundoff errors and ghosts

- ▶ A simple criterion distinguishes ghosts from real eigenvalues. Ghosts are caused by roundoff errors. Thus they do not depend on the starting vector ϕ_i . As a consequence these ghosts are also eigenvalues of the matrix T' , which can be obtained from T by deleting the first row and column:

$$T' = \begin{pmatrix} \cancel{a_1} & \cancel{b_2} & & & & \\ \cancel{b_2} & a_2 & b_3 & & & \\ & b_3 & a_3 & b_4 & & \\ & & \ddots & \ddots & \ddots & \\ & & & b_{N-1} & a_{N-1} & b_N \\ & & & & b_N & a_N \end{pmatrix}$$

- ▶ Criterion to distinguish ghosts from real eigenvalues:
 - ▶ All multiple eigenvalues are real, but their multiplicities might be too large.
 - ▶ All single eigenvalues of T which are not eigenvalues of T' are also real.
-

Degeneracy



Green's function–Continued fraction

- ▶ Zero temperature Green's function

$$G_{ij}(t) = -i \langle \Omega | c_i(t) c_j^\dagger(0) | \Omega \rangle$$

$$G_{ij}(\omega) = \left\langle \Omega \left| c_i \frac{1}{\omega - H + E_0} c_j^\dagger \right| \Omega \right\rangle$$

- ▶ Consider the diagonal element

$$|\phi_i\rangle = c_i^\dagger |\Omega\rangle \Rightarrow G_{ii} = \left\langle \phi_i \left| \frac{1}{\omega - H + E_0} \right| \phi_i \right\rangle$$

- ▶ Use the expansion

$$\frac{1}{z - H} = \frac{1}{z} + \frac{1}{z^2} H + \frac{1}{z^3} H^2 + \dots$$

Green's function–Continued fraction

- ▶ Truncated expansion evaluated exactly in Krylov subspace generated by $|\phi_i\rangle$
- ▶ We perform a Lanczos procedure on $|\phi_i\rangle$, then G_{ii} is given by Jacobi continued fraction

$$G_{ii} = \frac{\langle \phi_i | \phi_i \rangle}{\omega - a_0 - \frac{b_1^2}{\omega - a_1 - \frac{b_2^2}{\omega - a_2 - \cdots}}}$$

Green's function–Continued fraction

- ▶ Non-diagonal element

$$G_{ij}^+(\omega) = \left\langle \Omega \left| (c_i + c_j) \frac{1}{\omega - H + E_0} (c_i + c_j)^+ \right| \Omega \right\rangle$$

- ▶ Define $|\phi_k\rangle = (c_i + c_j)^+ |\Omega\rangle$, we obtain

$$G_{ij}^+ = \frac{\langle \phi_k | \phi_k \rangle}{\omega - a_0 - \frac{b_1^2}{\omega - a_1 - \frac{b_2^2}{\omega - a_2 - \dots}}}$$

- ▶ Since $G_{ij} = G_{ji}$, then

$$G_{ij} = \frac{1}{2} [G_{ij}^+ - G_{ii} - G_{jj}]$$

Green's function–Lehmann representation

- ▶ Lehmann representation

$$\begin{aligned} G_{ij}(\omega) &= \sum_m \langle \Omega | c_i | m \rangle \frac{1}{\omega - E_m + E_0} \langle m | c_j^\dagger | \Omega \rangle + \sum_n \langle \Omega | c_j | n \rangle \frac{1}{\omega + E_n - E_0} \langle n | c_i^\dagger | \Omega \rangle \\ &= \sum_m \frac{|\langle \Omega | c_i | m \rangle|^2}{\omega - E_m + E_0} + \sum_n \frac{|\langle n | c_i^\dagger | \Omega \rangle|^2}{\omega + E_n - E_0} \end{aligned}$$

$$|\phi_i\rangle = c_i^\dagger |\Omega\rangle, \quad i = 1 \dots L$$

- ▶ Extended Krylov space form a band Hamiltonian ($2L+1$ diagonals)

$$H^N |\phi_i\rangle = a_i^{(N)} |\phi_i^N\rangle, \quad i = 1 \dots L$$

Green's function–Recursion

- ▶ Green's function

$$G_{ij}(\omega) = \left\langle \Omega \left| c_i \frac{1}{\omega - H + E_0} c_j^\dagger \right| \Omega \right\rangle = \left\langle \phi_1 \left| \frac{1}{\omega - H + E_0} \right| \phi_1 \right\rangle$$

$$|\phi_i\rangle = c_i^\dagger |\Omega\rangle, \quad i = 1 \dots L$$

- ▶ **$HQ = QT$**

$$\mathbf{T} = \begin{pmatrix} a_1 & b_{21} & b_{32} & \dots & & \\ b_{21} & a_2 & b_{31} & b_{42} & & \\ b_{32} & b_{31} & a_3 & b_{41} & & \\ \vdots & b_{42} & \ddots & \ddots & \ddots & \\ & & & b_{N-1,1} & a_{N-1} & b_{N,1} \\ & & & & b_{N,1} & a_N \end{pmatrix} \quad \mathbf{Q} = [\boldsymbol{\varphi}_1 \quad \boldsymbol{\varphi}_2 \quad \boldsymbol{\varphi}_3 \quad \dots]$$

- ▶ where $\boldsymbol{\varphi}_i$ is calculated by band Lanczos method

Green's function–Recursion

► Recursion

$$\begin{aligned} G_{ij}(\omega) &= \boldsymbol{\varphi}_1 \frac{1}{\omega - \mathbf{H} + E_0} \boldsymbol{\varphi}_1 \\ &= \boldsymbol{\varphi}_1 \mathbf{Q} \mathbf{Q}^+ \frac{1}{\omega - \mathbf{H} + E_0} \mathbf{Q} \mathbf{Q}^+ \boldsymbol{\varphi}_1 \\ &= \boldsymbol{\varphi}_1 \mathbf{Q} \frac{1}{\omega - \mathbf{T} + E_0} \mathbf{Q}^+ \boldsymbol{\varphi}_1 \\ &= \left[\frac{1}{\omega - \mathbf{T} + E_0} \right]_{11} \end{aligned}$$

► where

$$\mathbf{Q}^+ \boldsymbol{\varphi}_1 = \begin{bmatrix} \boldsymbol{\varphi}_1 \\ \boldsymbol{\varphi}_2 \\ \boldsymbol{\varphi}_3 \\ \vdots \end{bmatrix} \quad \boldsymbol{\varphi}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

Finite Temperature Lanczos Methods

- ▶ These methods (FTLM, LTLM) combine the Lanczos method and random sampling to go to larger systems.
- ▶ For high to moderate temperatures these methods can obtain results basically in the thermodynamic limit.
- ▶ The Low Temperature Lanczos Method can also go to low temperatures to get correct results on a given sample. Finite size effects however persist.
- ▶ Like the $T = 0$ ED method it is most useful, where QMC or T -DMRG etc fail, such as frustrated and fermionic models.

FTLM

- ▶ In the canonical ensemble, the thermal average of a physical quantity A at finite $-T$ in orthonormal basis $|n\rangle$ is defined as

$$\langle A \rangle = \frac{1}{Z} \sum_n \langle n | A e^{-\beta H} | n \rangle$$

$$Z = \sum_n \langle n | e^{-\beta H} | n \rangle$$

- ▶ High temperature expansion gives

$$\langle A \rangle = \frac{1}{Z} \sum_n \sum_{k=0}^{\infty} \frac{(-\beta)^k}{k!} \langle n | H^k A | n \rangle$$

$$Z = \sum_n \sum_{k=0}^{\infty} \frac{(-\beta)^k}{k!} \langle n | H^k | n \rangle$$

- ▶ From two successive Lanczos procedures for two operators

$$\langle n | H^k B H^l A | n \rangle = \sum_{i=0}^M \sum_{j=0}^M \langle n | \psi_i \rangle \langle \psi_i | B | \tilde{\psi}_j \rangle \langle \tilde{\psi}_j | A | n \rangle (\varepsilon_i)^k (\tilde{\varepsilon}_j)^l$$

FTLM

- ▶ We obtain ($k < M$)

$$\langle n | H^k A | n \rangle = \sum_{i=0}^M \langle n | \psi_i^{(n)} \rangle \langle \psi_i^{(n)} | A | n \rangle (\varepsilon_i^{(n)})^k$$

- ▶ The superscript (n) means that $|\psi_i^{(n)}\rangle$ and $\varepsilon_i^{(n)}$ are obtained in Lanczos algorithm with the initial state $|\psi_i^{(n)}\rangle$
- ▶ Extend the results to $k > M$, the approximate expressions are

$$\begin{aligned} \langle A \rangle &\approx \frac{1}{Z} \sum_n \sum_{k=0}^{\infty} \frac{(-\beta)^k}{k!} \sum_{i=0}^M \langle n | \psi_i^{(n)} \rangle \langle \psi_i^{(n)} | A | n \rangle (\varepsilon_i^{(n)})^k = \frac{1}{Z} \sum_n \sum_{i=0}^M e^{-\beta \varepsilon_i^{(n)}} \langle n | \psi_i^{(n)} \rangle \langle \psi_i^{(n)} | A | n \rangle \\ Z &\approx \sum_n \sum_{k=0}^{\infty} \frac{(-\beta)^k}{k!} \sum_{i=0}^M \langle n | \psi_i^{(n)} \rangle \langle \psi_i^{(n)} | n \rangle (\varepsilon_i^{(n)})^k = \frac{1}{Z} \sum_n \sum_{i=0}^M e^{-\beta \varepsilon_i^{(n)}} \langle n | \psi_i^{(n)} \rangle \langle \psi_i^{(n)} | n \rangle \end{aligned}$$

- ▶ It is very expensive to calculate all n

FTLM

- ▶ Stochastic sampling of Krylov space: Replace the summation of the full Hilbert space by a sum over random number of states

$$\langle A \rangle \approx \frac{1}{Z} \sum_s \sum_r \sum_i^M e^{-\beta \varepsilon_i^{(r)}} \langle r | \psi_i^{(r)} \rangle \langle \psi_i^{(r)} | A | r \rangle$$

$$Z = \sum_s \sum_r \sum_i^M e^{-\beta \varepsilon_i^{(r)}} \left| \langle r | \psi_i^{(r)} \rangle \right|^2$$

- ▶ Σ_s over symmetry sector of dimension N_s
- ▶ Σ_r over R random starting vectors $|r\rangle$

LTLM

- ▶ Since FTLM is related to high- T expansion, i.e., $T \rightarrow \infty$ limit correct, the limit of $T \rightarrow 0$ does not lead to the ground state expectation value.

$$\langle A \rangle \approx \frac{1}{Z} \sum_s \sum_r \sum_i^M \sum_j^M e^{-\beta \varepsilon_i^{(r)}/2} e^{-\beta \varepsilon_j^{(r)}/2} \langle r | \psi_j^{(r)} \rangle \langle \psi_j^{(r)} | A | \psi_i^{(r)} \rangle \langle \psi_i^{(r)} | r \rangle$$

$$Z = \sum_s \sum_r \sum_i^M e^{-\beta \varepsilon_i^{(r)}} \left| \langle r | \psi_i^{(r)} \rangle \right|^2$$

- ▶ One more Lanczos procedure for each sampling
- ▶ In the limit of $T \rightarrow 0$

$$\langle A \rangle \approx \langle \psi_0 | A | \psi_0 \rangle$$

Strongly Correlated Systems

- ▶ Hubbard model
- ▶ Heisenberg model
- ▶ High- T_c superconductor
- ▶ Quantum Phase Transitions

Applications

- ⊗ **Quantum Magnets**: nature of novel phases, critical points in 1D, dynamical correlation functions in 1D & 2D
- ⊗ **Fermionic models (Hubbard/ t - J)**: gaps, pairing properties, correlation exponents, etc
- ⊗ **Fractional Quantum Hall states**: energy gaps, overlap with model states, entanglement spectra
- ⊗ **Quantum dimer models or other constrained models (anyon chain..)**
- ⊗ **Full Configuration Interaction in Quantum Chemistry**

Largest system reached

- ▶ $S = 1/2$ spin models
 - ▶ square lattice: $N = 40$
 - ▶ triangular lattice: $N = 39$
 - ▶ maximum dimension of basis: 1.5 billion
- ▶ t - J models
 - ▶ checkerboard lattice with 2 holes: $N = 32$
 - ▶ square lattice with 2 holes: $N = 32$
 - ▶ maximum dimension of basis: 2.8 billion
- ▶ Hubbard models
 - ▶ square lattice at half filling: $N = 20$
 - ▶ quantum dot structure: $N = 20$
 - ▶ maximum dimension of basis: 3 billion
- ▶ Holstein models
 - ▶ chain with $N = 14$ + phonon pseudo-sites
 - ▶ maximum dimension of basis: 30 billion

Exact diagonalization is expensive!

- ▶ Why use ED then?
 - ▶ Robust, unbiased and completely versatile—almost anything can be calculated!
 - ▶ There are models which are not easy to access via other models (e.g. frustrated magnets)
 - ▶ Error is at least as low as 10^{-14} – numerical precision
 - ▶ Exploiting symmetries reduces computational effort and gives physical information about eigenstates (good quantum numbers)

The best we can do: Low-lying states

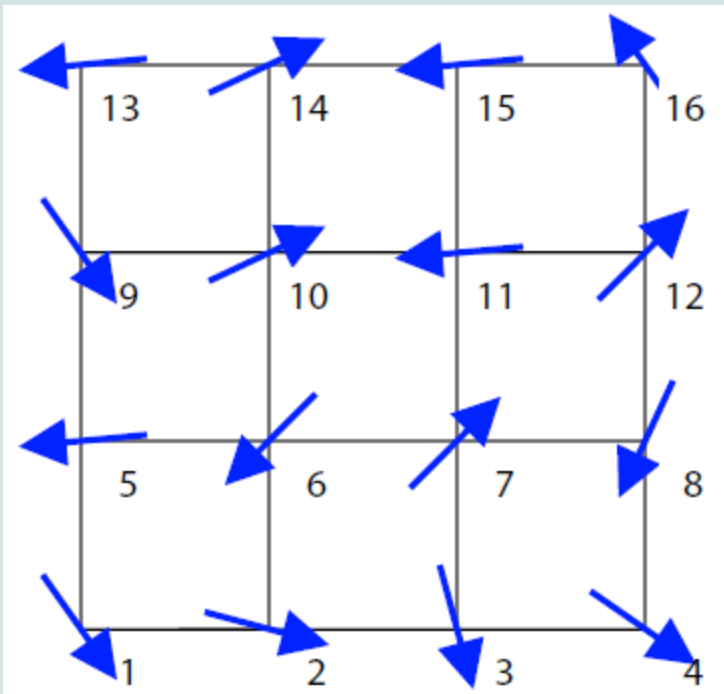
- ▶ Low-lying states sufficient for low-temperature physics;
- ▶ Low-lying states important for quantum phase transitions, e.g. magnetization jumps.
- ▶ Several methods available to obtain extreme eigenvalues: projection, Lanczos, DMRG.

Exact Diagonalization: Application

Spin-1/2

Heisenberg model

$$\mathcal{H} = -J \sum_{\langle ij \rangle \sigma} \mathbf{S}_i \cdot \mathbf{S}_j + h \sum_i \mathbf{B} \cdot \mathbf{S}_i$$



Spin-1/2 system

- ▶ Define ladder operators:

$$S_i^+ = S_i^x + iS_i^y$$

$$S_i^- = S_i^x - iS_i^y$$

- ▶ The basis of Hilbert space

$$|j, m\rangle = |\uparrow\rangle \otimes \dots \otimes |\downarrow\rangle$$

- ▶ The Hamiltonian becomes

$$H = J \sum_{ij} S_i^z S_j^z + [S_i^x S_j^x + S_i^y S_j^y] = J \sum_{ij} S_i^z S_j^z + \frac{1}{2} [S_i^+ S_j^- + S_i^- S_j^+]$$

Three spin-1/2 Heisenberg model

- ▶ Non-periodic boundary condition

$$H = JS_1^z S_2^z + J \frac{1}{2} [S_1^+ S_2^- + S_1^- S_2^+] + JS_2^z S_3^z + J \frac{1}{2} [S_2^+ S_3^- + S_2^- S_3^+]$$

- ▶ The size of the system is $2^3 = 8$
- ▶ The trial basis states are

$$|\downarrow\downarrow\downarrow\rangle = |000\rangle = |0\rangle$$

$$|\downarrow\downarrow\uparrow\rangle = |001\rangle = |1\rangle$$

$$|\downarrow\uparrow\downarrow\rangle = |010\rangle = |2\rangle$$

$$|\downarrow\uparrow\uparrow\rangle = |011\rangle = |3\rangle$$

$$|\uparrow\downarrow\downarrow\rangle = |100\rangle = |4\rangle$$

$$|\uparrow\downarrow\uparrow\rangle = |101\rangle = |5\rangle$$

$$|\uparrow\uparrow\downarrow\rangle = |110\rangle = |6\rangle$$

$$|\uparrow\uparrow\uparrow\rangle = |111\rangle = |7\rangle$$

The Hamiltonian matrix

- ▶ The Hamiltonian in the given basis is

$$H = \begin{pmatrix} \langle \downarrow \downarrow \downarrow | & \frac{J}{2} & & & & & \\ \langle \downarrow \downarrow \uparrow | & 0 & \frac{J}{2} & & & & \\ \langle \downarrow \uparrow \downarrow | & \frac{J}{2} & -\frac{J}{2} & 0 & 0 & \frac{J}{2} & \\ \langle \downarrow \uparrow \uparrow | & & 0 & 0 & 0 & \frac{J}{2} & \\ \langle \uparrow \downarrow \downarrow | & & 0 & 0 & 0 & 0 & \\ \langle \uparrow \downarrow \uparrow | & & \frac{J}{2} & \frac{J}{2} & 0 & -\frac{J}{2} & \frac{J}{2} \\ \langle \uparrow \uparrow \downarrow | & & & & & \frac{J}{2} & 0 \\ \langle \uparrow \uparrow \uparrow | & & & & & & \frac{J}{2} \end{pmatrix}$$

Magnetization

- ▶ The magnetization is

$$M_z = \sum_i S_i^z$$

- ▶ The size of the largest block is

$$M_z = 1/2 \text{ or } -1/2$$

$$C_1^3 = 3$$

- ▶ The new basis states become

$$\begin{aligned} |\downarrow\downarrow\downarrow\rangle &= |M_z = -\frac{3}{2}\rangle = |0\rangle & |\downarrow\uparrow\uparrow\rangle &= |M_z = \frac{1}{2}\rangle = |3\rangle \\ |\downarrow\downarrow\uparrow\rangle &= |M_z = -\frac{1}{2}\rangle = |1\rangle & |\uparrow\downarrow\uparrow\rangle &= |M_z = \frac{1}{2}\rangle = |5\rangle \\ |\downarrow\uparrow\downarrow\rangle &= |M_z = -\frac{1}{2}\rangle = |2\rangle & |\uparrow\uparrow\downarrow\rangle &= |M_z = \frac{1}{2}\rangle = |6\rangle \\ |\uparrow\downarrow\downarrow\rangle &= |M_z = -\frac{1}{2}\rangle = |4\rangle & |\uparrow\uparrow\uparrow\rangle &= |M_z = \frac{3}{2}\rangle = |7\rangle \end{aligned}$$

Magnetization

- The Hamiltonian becomes

$$H = \begin{pmatrix} \left\langle -\frac{3}{2} \right| & \left\langle -\frac{1}{2} \right| & \left\langle -\frac{1}{2} \right| & \left\langle -\frac{1}{2} \right| & \left\langle \frac{1}{2} \right| & \left\langle \frac{1}{2} \right| & \left\langle \frac{1}{2} \right| & \left\langle \frac{3}{2} \right| \\ \frac{J}{2} & 0 & \frac{J}{2} & \frac{J}{2} & 0 & \frac{J}{2} & 0 & \frac{J}{2} \\ & \frac{J}{2} & -\frac{J}{2} & \frac{J}{2} & \frac{J}{2} & -\frac{J}{2} & \frac{J}{2} & \\ & & \frac{J}{2} & 0 & \frac{J}{2} & 0 & 0 & \\ & & & \frac{J}{2} & -\frac{J}{2} & \frac{J}{2} & 0 & \\ & & & & \frac{J}{2} & 0 & \frac{J}{2} & \\ & & & & & \frac{J}{2} & 0 & \end{pmatrix}$$

How does the computer work

- ▶ Declare a vector $A[m]$ to store basis states

- ▶ For example:

$$A[1] = 0 \quad A[2] = 1 \quad A[3] = 2 \quad A[4] = 3$$

$$A[5] = 4 \quad A[6] = 5 \quad A[7] = 6 \quad A[8] = 7$$

- ▶ Next, act H on the basis states

- ▶ For example:

$$H|A[3]\rangle = H|2\rangle = H|\downarrow\uparrow\downarrow\rangle = \frac{J}{2}|\downarrow\downarrow\uparrow\rangle + \frac{J}{2}|\uparrow\downarrow\downarrow\rangle = \frac{J}{2}|1\rangle + \frac{J}{2}|4\rangle$$

- ▶ Then, we need to find out the corresponding index with respect to the basis

$$|1\rangle = |A[2]\rangle, \quad j_1 = 2$$

$$|4\rangle = |A[5]\rangle, \quad j_2 = 5$$

How does the computer work

- ▶ The most efficient way: declare a vector $B[m]$, $m = 0$ to 7

- ▶ Therefore we have

$$B[0] = 1 \quad B[1] = 2 \quad B[2] = 3 \quad B[3] = 4$$

$$B[4] = 5 \quad B[5] = 6 \quad B[6] = 7 \quad B[7] = 8$$

- ▶ and

$$B[A[2]] = 2 = j_1$$

$$B[A[4]] = 4 = j_2$$

- ▶ The matrix elements are

$$H(i, j_1) = H(3, 2) = \frac{J}{2}$$

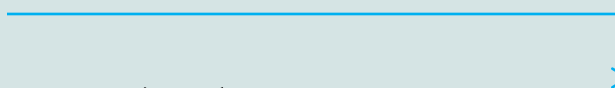
$$H(i, j_2) = H(3, 4) = \frac{J}{2}$$

Two-Table method

- ▶ Divide lattice to two sublattice
- ▶ For example: site-1,2 and site-3

$$i = 2^2 i_3 + i_{1,2}$$

$$j = j_3 + j_{1,2}$$



$ \downarrow\rangle$	0	1	$ \downarrow\downarrow\rangle$	0	0	1
$ \downarrow\rangle$	0	1	$ \downarrow\uparrow\rangle$	1	1	2
$ \downarrow\rangle$	0	1	$ \uparrow\downarrow\rangle$	2	2	3
$ \downarrow\rangle$	0	1	$ \uparrow\uparrow\rangle$	3	3	4
$ \uparrow\rangle$	1	5	$ \downarrow\downarrow\rangle$	0	0	5
$ \uparrow\rangle$	1	5	$ \downarrow\uparrow\rangle$	1	1	6
$ \uparrow\rangle$	1	5	$ \uparrow\downarrow\rangle$	2	2	7
$ \uparrow\rangle$	1	5	$ \uparrow\uparrow\rangle$	3	3	8

Two spin-3/2 Heisenberg model

$$H = JS_1^z S_2^z + J \frac{1}{2} [S_1^+ S_2^- + S_1^- S_2^+]$$

- ▶ The size of the system is $4^2 = 16$

$$\left| -\frac{3}{2} - \frac{3}{2} \right\rangle \quad \left| \frac{1}{2} - \frac{3}{2} \right\rangle$$

$$\left| -\frac{3}{2} - \frac{1}{2} \right\rangle \quad \left| \frac{1}{2} - \frac{1}{2} \right\rangle$$

$$\left| -\frac{3}{2} \frac{1}{2} \right\rangle \quad \left| \frac{1}{2} \frac{1}{2} \right\rangle$$

$$\left| -\frac{3}{2} \frac{3}{2} \right\rangle \quad \left| \frac{1}{2} \frac{3}{2} \right\rangle$$

$$\left| -\frac{1}{2} - \frac{3}{2} \right\rangle \quad \left| \frac{3}{2} - \frac{3}{2} \right\rangle$$

$$\left| -\frac{1}{2} - \frac{1}{2} \right\rangle \quad \left| \frac{3}{2} - \frac{1}{2} \right\rangle$$

$$\left| -\frac{1}{2} \frac{1}{2} \right\rangle \quad \left| \frac{3}{2} \frac{1}{2} \right\rangle$$

$$\left| -\frac{1}{2} \frac{3}{2} \right\rangle \quad \left| \frac{3}{2} \frac{3}{2} \right\rangle$$

Two-bit per-site

$$|00\rangle = \left| -\frac{3}{2} \right\rangle$$

$$|01\rangle = \left| -\frac{3}{2} \right\rangle$$

$$|10\rangle = \left| \frac{1}{2} \right\rangle$$

$$|11\rangle = \left| \frac{3}{2} \right\rangle$$

$$\left| -\frac{3}{2} - \frac{3}{2} \right\rangle = |00 \ 00\rangle \quad \left| \frac{1}{2} - \frac{3}{2} \right\rangle = |10 \ 00\rangle$$

$$\left| -\frac{3}{2} - \frac{1}{2} \right\rangle = |00 \ 01\rangle \quad \left| \frac{1}{2} - \frac{1}{2} \right\rangle = |10 \ 01\rangle$$

$$\left| -\frac{3}{2} \frac{1}{2} \right\rangle = |00 \ 10\rangle \quad \left| \frac{1}{2} \frac{1}{2} \right\rangle = |10 \ 10\rangle$$

$$\left| -\frac{3}{2} \frac{3}{2} \right\rangle = |00 \ 11\rangle \quad \left| \frac{1}{2} \frac{3}{2} \right\rangle = |10 \ 11\rangle$$

$$\left| -\frac{1}{2} - \frac{3}{2} \right\rangle = |01 \ 00\rangle \quad \left| \frac{3}{2} - \frac{3}{2} \right\rangle = |11 \ 00\rangle$$

$$\left| -\frac{1}{2} - \frac{1}{2} \right\rangle = |01 \ 01\rangle \quad \left| \frac{3}{2} - \frac{1}{2} \right\rangle = |11 \ 01\rangle$$

$$\left| -\frac{1}{2} \frac{1}{2} \right\rangle = |01 \ 10\rangle \quad \left| \frac{3}{2} \frac{1}{2} \right\rangle = |11 \ 10\rangle$$

$$\left| -\frac{1}{2} \frac{3}{2} \right\rangle = |01 \ 11\rangle \quad \left| \frac{3}{2} \frac{3}{2} \right\rangle = |11 \ 11\rangle$$

Magnetization $M_z = -3, -2, -1, 0, 1, 2, 3$

$$|-\frac{3}{2} - \frac{3}{2}\rangle = |00 \ 00\rangle$$

$$|-\frac{3}{2} - \frac{1}{2}\rangle = |00 \ 01\rangle \quad |-\frac{1}{2} - \frac{3}{2}\rangle = |01 \ 00\rangle$$

$$|-\frac{3}{2} \frac{1}{2}\rangle = |00 \ 10\rangle \quad |-\frac{1}{2} - \frac{1}{2}\rangle = |01 \ 01\rangle \quad |\frac{1}{2} - \frac{3}{2}\rangle = |10 \ 00\rangle$$

$$|-\frac{3}{2} \frac{3}{2}\rangle = |00 \ 11\rangle \quad |-\frac{1}{2} \frac{1}{2}\rangle = |01 \ 10\rangle \quad |\frac{1}{2} - \frac{1}{2}\rangle = |10 \ 01\rangle \quad |\frac{3}{2} - \frac{3}{2}\rangle = |11 \ 00\rangle$$

$$|-\frac{1}{2} \frac{3}{2}\rangle = |01 \ 11\rangle \quad |\frac{1}{2} \frac{1}{2}\rangle = |10 \ 10\rangle \quad |\frac{3}{2} - \frac{1}{2}\rangle = |11 \ 01\rangle$$

$$|\frac{1}{2} \frac{3}{2}\rangle = |10 \ 11\rangle \quad |\frac{3}{2} \frac{1}{2}\rangle = |11 \ 10\rangle$$

$$|\frac{3}{2} \frac{3}{2}\rangle = |11 \ 11\rangle$$

The Basis and lookup tables

$$\begin{array}{cccccccc} A[1] = 0 & A[2] = 1 & A[3] = 4 & A[4] = 2 & A[5] = 5 & A[6] = 8 & A[7] = 3 & A[8] = 6 \\ A[9] = 9 & A[10] = 12 & A[11] = 7 & A[12] = 10 & A[13] = 13 & A[14] = 11 & A[15] = 14 & A[16] = 15 \end{array}$$

$$\begin{array}{cccccccc} B[0] = 1 & B[1] = 2 & B[2] = 4 & B[3] = 7 & B[4] = 3 & B[5] = 5 & B[6] = 8 & B[7] = 11 \\ B[8] = 6 & B[9] = 9 & B[10] = 12 & B[11] = 14 & B[12] = 10 & B[13] = 13 & B[14] = 15 & B[15] = 16 \end{array}$$

Hamiltonian matrix elements

$$H = \begin{pmatrix} \langle 0 | & \frac{9}{4}J & & & & & & & & & & & & & & \\ \langle 1 | & & \frac{3}{4}J & \frac{3}{2}J & & & & & & & & & & & & \\ \langle 4 | & & \frac{3}{2}J & \frac{3}{4}J & & & & & & & & & & & & \\ \langle 2 | & & & & -\frac{3}{4}J & \sqrt{3}J & 0 & & & & & & & & & \\ \langle 5 | & & & & \sqrt{3}J & -\frac{3}{4}J & \sqrt{3}J & & & & & & & & & \\ \langle 8 | & & & & 0 & \sqrt{3}J & -\frac{3}{4}J & & & & & & & & & \\ \langle 3 | & & & & & & & -\frac{9}{4}J & \frac{3}{2}J & 0 & 0 & & & & & \\ \langle 6 | & & & & & & & \frac{3}{2}J & -\frac{1}{4}J & 2J & 0 & & & & & \\ \langle 9 | & & & & & & & 0 & 2J & -\frac{1}{4}J & \frac{3}{2}J & & & & & \\ \langle 12 | & & & & & & & 0 & 0 & \frac{3}{2}J & -\frac{9}{4}J & & & & & \\ \langle 7 | & & & & & & & & & & & -\frac{3}{4}J & \sqrt{3}J & 0 & & \\ \langle 10 | & & & & & & & & & & & \sqrt{3}J & \frac{1}{4}J & \sqrt{3}J & & \\ \langle 13 | & & & & & & & & & & & 0 & \sqrt{3}J & -\frac{3}{4}J & & \\ \langle 11 | & & & & & & & & & & & & & & \frac{3}{4}J & \frac{3}{2}J \\ \langle 14 | & & & & & & & & & & & & & & \frac{3}{2}J & \frac{3}{4}J \\ \langle 15 | & & & & & & & & & & & & & & & \frac{9}{4}J \end{pmatrix}$$

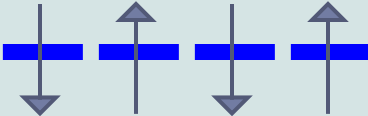
Exact Diagonalization: Application

Hubbard Model

Hubbard Model

$$\mathcal{H} = -t \sum_{\langle ij \rangle \sigma} c_{i\sigma}^+ c_{j\sigma} + \sum_i U \hat{n}_{i\uparrow} \hat{n}_{i\downarrow}$$

UHO $\varepsilon + U$ 

LHO ε 

1 2 3 4

Atomic Ground State ($t = 0$)

- Four possible occupation of a single site



$$H = \begin{matrix} \langle 0 | \\ \langle \uparrow | \\ \langle \downarrow | \\ \langle 2 | \end{matrix} \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & U \end{pmatrix}$$

A. Non-interacting electrons

$$H_0 = \sum_{ij} t_{ij} c_i^\dagger c_j$$

- ▶ Assume ϕ_j to be Wannier wave functions (Fourier transform of Bloch orbitals)

$$t_{ij} = \left\langle \phi_i \left| -\frac{\hbar^2 \nabla^2}{2m} + \hat{v} \right| \phi_j \right\rangle$$

- ▶ Assume t_{ij} is non-zero only when i, j are nearest neighbors

$$H_0 = -t \sum_{ij} (c_i^\dagger c_j + c_j^\dagger c_i)$$

- ▶ Diagonalize the Hamiltonian by a Fourier transformation

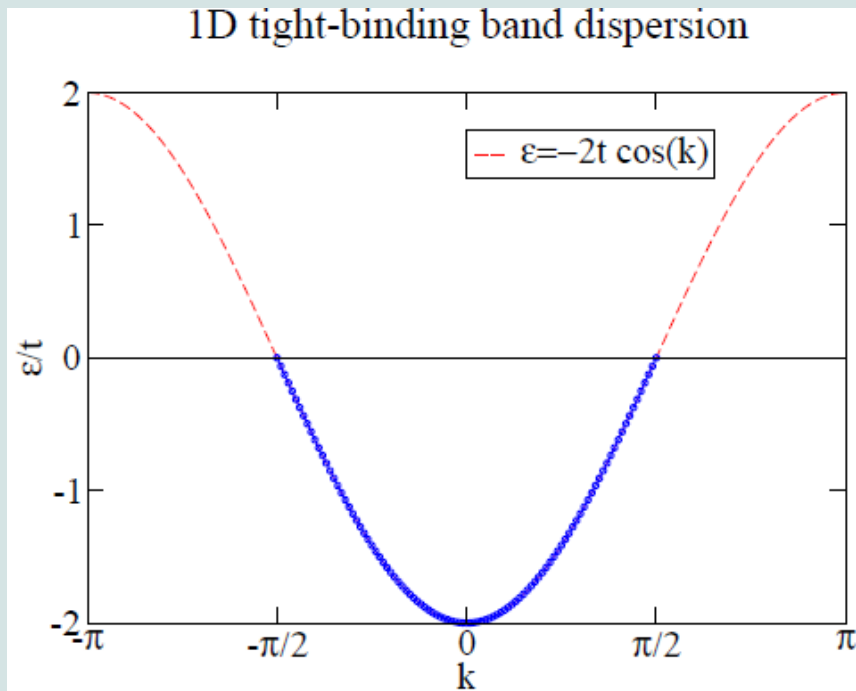
$$c_k^\dagger = \frac{1}{\sqrt{L}} \sum_j e^{ikj} c_j^\dagger$$

A. Non-interacting electrons

- ▶ Then the Hamiltonian in ϕ_j basis becomes

$$H_0 = \sum_k \varepsilon_k c_k^\dagger c_k$$

where $\varepsilon_k = -2t \cos(k)$



$$|GS\rangle = \prod_{k < k_F} c_{k\uparrow}^\dagger c_{k\downarrow}^\dagger |0\rangle$$

A. Non-interacting electrons

- ▶ Particle-hole excitation

$$H_0|\psi_n\rangle = H_0 c_{k+q\uparrow}^+ c_{k\uparrow} |GS\rangle = \varepsilon_{kq} |\psi_n\rangle$$

where $\varepsilon_k(q) = \varepsilon_{k+q} - \varepsilon_k$

- ▶ As the energy of the particle-hole excitation $\varepsilon_k(q)$ can be made arbitrarily small, $|GS\rangle$ characterizes a metal.

B. Electron-electron interaction

► Two-site case









$$H = H_t + H_U = -t(c_{0\uparrow}^+ c_{1\uparrow} + c_{1\uparrow}^+ c_{0\uparrow} + c_{0\downarrow}^+ c_{1\downarrow} + c_{1\downarrow}^+ c_{0\downarrow}) + U(n_{0\uparrow} n_{0\downarrow} + n_{1\uparrow} n_{1\downarrow})$$

$$H = \begin{pmatrix} \langle 00 | & & & & & & & & & & & & & & & & \\ \langle 0\uparrow | & & -t & & & & & & & & & & & & & & \\ \langle 0\downarrow | & & & & -t & & & & & & & & & & & & \\ \langle 02 | & & U & & -t & & & -t & & & & & & & & & \\ \langle \uparrow 0 | & -t & & & & & & & & & & & & & & & \\ \langle \uparrow\uparrow | & & & & & & & & & & & & & & & & \\ \langle \uparrow\downarrow | & & -t & & & & & -t & & & & & & & & & \\ \langle \uparrow 2 | & & & & U & & & & & -t & & & & & & & \\ \langle \downarrow 0 | & -t & & & & & & & & & & & & & & & \\ \langle \downarrow\uparrow | & & -t & & & & & -t & & & & & & & & & \\ \langle \downarrow\downarrow | & & & & & & & & & & & & & & & & \\ \langle \downarrow 2 | & & & & & & U & & & -t & & & & & & & \\ \langle 20 | & & & -t & & -t & & U & & & & & & & & & \\ \langle 2\uparrow | & & & & -t & & & & U & & & & & & & & \\ \langle 2\downarrow | & & & & & & -t & & & U & & & & & & & \\ \langle 22 | & & & & & & & & & & 2U & & & & & & \end{pmatrix}$$

B. Electron-electron interaction

- ▶ Organizing the Hilbert space
 - ▶ Operators commute with Hamiltonian
 - ▶ Number operator $N = \sum n_{i\sigma}$
 - ▶ Total $S_z = \frac{1}{2} \sum (n_{i\uparrow} - n_{i\downarrow})$
 - ▶ The structure of the Hamiltonian will be block diagonal where each block corresponds to a fixed value of N and S_z
 - ▶ For example: $N=2$, the size of this block is $C_2^4 = 6$, which is also the largest block of the Hamiltonian matrix
 - ▶ *Case1: $S_z = 0$*
 - ▶ The size is 4
 - ▶ *Case2: $S_z = -1$*
 - ▶ The size is 1
 - ▶ *Case3: $S_z = 1$*
 - ▶ The size is 1

$$N=2 \text{ and } S_z=0$$

		$ 01\rangle_{\downarrow} 01\rangle_{\uparrow}$	5
		$ 01\rangle_{\downarrow} 10\rangle_{\uparrow}$	6
		$ 10\rangle_{\downarrow} 01\rangle_{\uparrow}$	9
		$ 10\rangle_{\downarrow} 10\rangle_{\uparrow}$	10
Site 1	Site 0	→ Binary	→ Decimal

► Declare a matrix $A(1:4)$

$A(1)$	→	5	} Basis States
$A(2)$	→	6	
$A(3)$	→	9	
$A(4)$	→	10	

$N=2$ and $S_z=0$: Diagonal elements

- ▶ Basis states

$$|\varphi_1\rangle = |A(1)\rangle = |5\rangle = |01\rangle_{\downarrow} |01\rangle_{\uparrow}$$

- ▶ Acting on-site Coulomb interaction H_U on the basis states

$$H_U = \begin{matrix} \downarrow\langle 01|_{\uparrow}\langle 01| \\ \downarrow\langle 01|_{\uparrow}\langle 10| \\ \downarrow\langle 10|_{\uparrow}\langle 01| \\ \downarrow\langle 10|_{\uparrow}\langle 10| \end{matrix} \begin{pmatrix} U & & & \\ & 0 & & \\ & & 0 & \\ & & & U \end{pmatrix}$$

$N=2$ and $S_z=0$: Off-Diagonal elements

- ▶ Acting hopping terms H_t on the basis states

$$|\varphi_1\rangle = |A(1)\rangle = |5\rangle = |01\rangle_{\downarrow} |01\rangle_{\uparrow} = c_{0\downarrow}^+ c_{0\uparrow}^+ |0\rangle$$

$$\begin{aligned} H_t |\varphi_1\rangle &= -t(c_{0\uparrow}^+ c_{1\uparrow} + c_{1\uparrow}^+ c_{0\uparrow} + c_{0\downarrow}^+ c_{1\downarrow} + c_{1\downarrow}^+ c_{0\downarrow}) c_{0\downarrow}^+ c_{0\uparrow}^+ |0\rangle \\ &= -t c_{1\uparrow}^+ c_{0\uparrow} c_{0\downarrow}^+ c_{0\uparrow}^+ |0\rangle - t c_{1\downarrow}^+ c_{0\downarrow} c_{0\downarrow}^+ c_{0\uparrow}^+ |0\rangle \\ &= +t c_{1\uparrow}^+ c_{0\downarrow}^+ c_{0\uparrow} c_{0\uparrow}^+ |0\rangle - t c_{1\downarrow}^+ c_{0\uparrow}^+ c_{0\downarrow} c_{0\downarrow}^+ |0\rangle \\ &= -t c_{0\downarrow}^+ c_{1\uparrow}^+ c_{0\uparrow} c_{0\uparrow}^+ |0\rangle - t c_{1\downarrow}^+ c_{0\uparrow}^+ c_{0\downarrow} c_{0\downarrow}^+ |0\rangle \\ &= -t |\varphi_2\rangle - t |\varphi_3\rangle \end{aligned}$$

- ▶ How does the computer work to come out the same results?

$N=2$ and $S_z=0$: Off-Diagonal elements

- ▶ Acting hopping terms H_t on the basis states

$$|\phi\rangle = |\phi\rangle_{\downarrow} |\phi\rangle_{\uparrow} = [c_{r_1\downarrow}^+ \cdots c_{r_N\downarrow}^+] |0\rangle_{\downarrow} [c_{r_1\uparrow}^+ \cdots c_{r_N\uparrow}^+] |0\rangle_{\uparrow}$$

$$\begin{aligned} c_{r_i\uparrow}^+ c_{r_j\uparrow} |\phi\rangle &= c_{r_i\uparrow}^+ c_{r_j\uparrow} |\phi\rangle_{\downarrow} |\phi\rangle_{\uparrow} \\ &= c_{r_i\uparrow}^+ c_{r_j\uparrow} [c_{r_1\downarrow}^+ \cdots c_{r_N\downarrow}^+] |0\rangle_{\downarrow} [c_{r_1\uparrow}^+ \cdots c_{r_N\uparrow}^+] |0\rangle_{\uparrow} \\ &= [c_{r_1\downarrow}^+ \cdots c_{r_N\downarrow}^+] |0\rangle_{\downarrow} (-1)^N (-1)^N c_{r_i\uparrow}^+ c_{r_j\uparrow} [c_{r_1\uparrow}^+ \cdots c_{r_N\uparrow}^+] |0\rangle_{\uparrow} \\ &= |\phi\rangle_{\downarrow} c_{r_i\uparrow}^+ c_{r_j\uparrow} |\phi\rangle_{\uparrow} \end{aligned}$$

- ▶ Count the number of electrons between site i and site j

$$c_{1\uparrow}^+ c_{0\uparrow} |\phi_1\rangle = c_{1\uparrow}^+ c_{0\uparrow} c_{0\downarrow}^+ c_{0\uparrow}^+ |0\rangle = (-1)^0 |\phi_2\rangle$$

$$c_{1\uparrow}^+ c_{0\uparrow} |\phi_1\rangle = c_{1\downarrow}^+ c_{0\downarrow} c_{0\downarrow}^+ c_{0\uparrow}^+ |0\rangle = (-1)^0 |\phi_3\rangle$$

$N=2$ and $S_z=0$: Off-Diagonal elements

- ▶ Similar to other basis states

$$H_t|\phi_1\rangle = -t|\phi_2\rangle - t|\phi_3\rangle$$

$$H_t|\phi_2\rangle = -t|\phi_1\rangle - t|\phi_4\rangle$$

$$H_t|\phi_3\rangle = -t|\phi_1\rangle - t|\phi_4\rangle$$

$$H_t|\phi_4\rangle = -t|\phi_2\rangle - t|\phi_3\rangle$$

- ▶ Therefore in these basis states, the Hamiltonian has the following matrix representation.

$$H_t = \begin{matrix} \downarrow\langle 01|_{\uparrow}\langle 01| \\ \downarrow\langle 01|_{\uparrow}\langle 10| \\ \downarrow\langle 10|_{\uparrow}\langle 01| \\ \downarrow\langle 10|_{\uparrow}\langle 10| \end{matrix} \begin{pmatrix} & -t & -t & \\ -t & & & -t \\ -t & & & -t \\ & -t & -t & \end{pmatrix}$$

$N=2$ and $S_z = -1, 1$

- ▶ For $S_z = 1$, the basis state is

$$\begin{array}{|c|} \hline \uparrow \\ \hline \end{array} \begin{array}{|c|} \hline \uparrow \\ \hline \end{array} \rightarrow |00\rangle_{\downarrow} |11\rangle_{\uparrow}$$

$$|\phi\rangle = |00\rangle_{\downarrow} |\uparrow\uparrow\rangle_{\uparrow}$$

$$H_t |\phi\rangle = 0 \quad \text{and} \quad H_U |\phi\rangle = 0$$

- ▶ For $S_z = -1$, the basis state is

$$\begin{array}{|c|} \hline \downarrow \\ \hline \end{array} \begin{array}{|c|} \hline \downarrow \\ \hline \end{array} \rightarrow |11\rangle_{\downarrow} |00\rangle_{\uparrow}$$

$$|\phi\rangle = |\downarrow\downarrow\rangle_{\downarrow} |00\rangle_{\uparrow}$$

$$H_t |\phi\rangle = 0 \quad \text{and} \quad H_U |\phi\rangle = 0$$

$N = 0$ Block Diagonal Structure

$$H_t = \begin{matrix} \downarrow \langle 00 |_{\uparrow} \langle 11 | \\ \downarrow \langle 01 |_{\uparrow} \langle 01 | \\ \downarrow \langle 01 |_{\uparrow} \langle 10 | \\ \downarrow \langle 10 |_{\uparrow} \langle 01 | \\ \downarrow \langle 10 |_{\uparrow} \langle 10 | \\ \downarrow \langle 11 |_{\uparrow} \langle 00 | \end{matrix} \begin{pmatrix} 0 & & & & \\ & U & -t & -t & \\ & -t & & & -t \\ & -t & & & -t \\ & & -t & -t & U \\ & & & & & 0 \end{pmatrix}$$

The Block Diagonal Structure

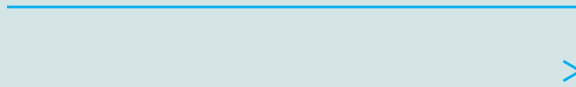
$$H = \begin{pmatrix} \langle 00 | & 0 & & & & & & & & & & & & & \\ \langle 0\uparrow | & & 0 & -t & & & & & & & & & & & \\ \langle \uparrow 0 | & & -t & 0 & & & & & & & & & & & \\ \langle 0\downarrow | & & & & 0 & -t & & & & & & & & & \\ \langle \downarrow 0 | & & & & -t & 0 & & & & & & & & & \\ \langle \uparrow\uparrow | & & & & & & 0 & & & & & & & & \\ \langle 02 | & & & & & & & U & -t & -t & & & & & \\ \langle \uparrow\downarrow | & & & & & & & -t & 0 & & -t & & & & \\ \langle \downarrow\uparrow | & & & & & & & -t & & 0 & -t & & & & \\ \langle 20 | & & & & & & & & -t & -t & U & & & & \\ \langle \downarrow\downarrow | & & & & & & & & & & & 0 & & & \\ \langle \uparrow 2 | & & & & & & & & & & & & U & -t & \\ \langle 2\uparrow | & & & & & & & & & & & & -t & U & \\ \langle \downarrow 2 | & & & & & & & & & & & & & & U & -t \\ \langle 2\downarrow | & & & & & & & & & & & & & -t & U & \\ \langle 22 | & & & & & & & & & & & & & & & 2U \end{pmatrix}$$

Two-Table method

- ▶ For example: Up configuration and Down configuration

$$i = 2^2 i_1 + i_2$$

$$j = j_1 + j_2$$



$ 00\rangle$	0	0	$ 11\rangle$	3	1	1
$ 01\rangle$	1	1	$ 01\rangle$	1	1	2
$ 01\rangle$	1	1	$ 10\rangle$	2	2	3
$ 10\rangle$	2	3	$ 01\rangle$	1	1	4
$ 10\rangle$	2	3	$ 10\rangle$	2	2	5
$ 11\rangle$	3	5	$ 00\rangle$	0	1	6

How to implement

- ▶ Use subroutine (in Fortran 77 or c), module (in Fortran 90/95/2003), or object (in c++ or python) to structure your code
- ▶ Write an object (here it means subroutine, module, or object in different language) to store the symmetrized basis states and lookup table
- ▶ Write an object to calculate the matrix elements and assign it to a matrix
- ▶ Call the diagonalization routine using LAPACK/ARPACK/...
- ▶ Write an object to calculate the observable quantities

Main piece of wisdom

- ▶ DO NOT start writing your own code from the scratch (unless really forced to)