

$\leadsto$ Diagonalize T to find lowest eigenvalue E_0 and eigenvector $\vec{y}_0$.

not fully stored!

$\leadsto$ In order to find $|E_0\rangle = Q \vec{y}_0$, repeat Lanczos-iteration with one additional D-vector $|n\rangle$, initialized as $|n\rangle = y_0[0] |\psi_i\rangle$ and update in step j as $|n\rangle \rightarrow |n\rangle + y_0[j] |q_j\rangle$ at the end of the while-loop ($|q_j\rangle = \frac{q}{\beta[j]}$ at step j).
At the final value of j : $|n\rangle = Q \vec{y}_0 = |E_0\rangle$.

- Computational cost of Lanczos algorithm

$\hookrightarrow$ Runtime: $O(\dim(\mathcal{H}) \cdot \log(\dim \mathcal{H}) \cdot n)$

$\nearrow$
dim of Krylov space = $O(n)$

↳ Memory consumption: 2 real 32-bit floats 3 complex 64-bit doubles
($3 \times 2 \times 8$)

2-3 $\dim \mathcal{H}$ -vectors $\simeq (8-48) \cdot \dim \mathcal{H}$ Bytes

$\leadsto \dim \mathcal{H} = 2^N \simeq 10^9$ for $N=30$ sites doable on modern desktop computer/laptop with 8-64 GB RAM.

$\leadsto$ Runtime for one Lanczos iteration $\mathcal{O}(1-10 \text{ minutes})$ at $\dim \mathcal{H} = 10^9$.

- Numerical issues with Lanczos algorithm:

↳ $\beta[j]$ gets very small for $j \geq 50-100 \rightarrow$ loss of orthogonality between Krylov basis vectors $|q_i\rangle$. This problem is intrinsic to the algorithm as it comes hand in hand with the invariance of $K_n(H, |\varphi_i\rangle)$.

↳ Precision of eigenvalues typically higher than of eigenstates.

↳ Possible remedies: (i) Re-orthogonalize during iteration (costly as more vectors need to be stored)
(ii) Quick and dirty: Ignore loss of orthogonality and keep iterating to get better estimate for $|E_0\rangle$. Works in many cases (and can be verified by checking $\frac{\langle E_0 | H | E_0 \rangle}{\langle E_0 | E_0 \rangle} \stackrel{?}{\approx} E_0$)

However, a caveat is that spurious additional ground states (ghost states) may occur as an artefact.

↳ See, e.g. R.M. Noack and S.R. Manmana, arXiv: cond-mat/0510321

for a detailed discussion (and references) on the limitations of Lanczos algorithm and extensions, such as Arnoldi algorithm for non-Hermitian matrices.

B.3: Including Symmetries

- Basic idea: Symmetries manifest in operators that commute with the Hamiltonian.

→ Diagonalizing these operators (decomposing their representation on the Hilbert space into a direct sum of irreducible representations) leads to a block-diagonal form of the Hamiltonian, where each block corresponds to a fixed set of quantum numbers associated with the symmetry.

$$H = \begin{pmatrix} \begin{pmatrix} H^{(1)} \end{pmatrix} & 0 & - & 0 \\ 0 & \begin{pmatrix} H^{(2)} \end{pmatrix} & & \\ \vdots & & \ddots & \\ 0 & & & \end{pmatrix}$$

→ ED problem can be decomposed and only the blocks of interest need to be considered one by one.

- Key issues:- Basis states J respecting the symmetries need to be found and grouped into the aforementioned blocks.

- Since the action of operators is implemented more efficiently on unrestricted states I , an efficient mapping $I \leftrightarrow J$ needs to be constructed (lookup tables).

$\leadsto$ Schematic procedure when applying block $H^{(i)}$ to state:

(a) Loop on states J in block $H^{(i)}$ {

(b) Retrieve representative $\hat{I}(J)$ of this state in occupation number basis

(c) Implement action of H with bitwise operations

(d) Translate resulting I' back to J -basis ($\hat{J}(I')$).

[Optional:
(e) Store matrix-element $\langle \hat{J}[I'] | H^{(i)} | J \rangle$ to J
sparse matrix representation of $H^{(i)}$]

}

- Primary example: Total S_z conservation in spin systems (equivalent to particle number conservation in fermionic language)

$$\hookrightarrow S_z^{\text{tot}} = \sum_{j=1}^N S_z^{(j)} \quad \text{conservation for spin } \frac{1}{2} \text{ systems simply}$$

amounts to a fixed number of 1-bits in basis integers $I_6 = \sum_{j=1}^N b_j 2^{N-j}$

$\leadsto$ Size of block with $S_z^{\text{tot}} = (2n - N) \frac{\hbar}{2}$, $n = 0, \dots, N$
is $\dim(\mathcal{H}^{(n)}) = \binom{N}{n} < \dim(\mathcal{H}) = 2^N$.

Concrete example: For $N = 30$ sites ($\dim(\mathcal{H}) = 2^{30} = 1.07 \cdot 10^9$)
the bigger block (at $n = 15$) has dimension $\dim(\mathcal{H}^{(15)}) = \binom{30}{15} = 1.55 \cdot 10^8$
and at $n = 10$, $\dim(\mathcal{H}^{(10)}) = 3.0 \cdot 10^7 \leadsto$ block dimension
drastically reduced compared to $\dim(\mathcal{H})$

$\hookrightarrow$ To find all basis state in block n , start with $\sigma = (\underbrace{0, \dots, 0}_{N-n}, \underbrace{1, \dots, 1}_n)$ and generate all distinct permutations $P(\sigma)$. Order the associated integers $I_{P(\sigma)}$ from smallest (I_σ) to biggest ($I_{\tilde{\sigma}}$) with $\tilde{\sigma} = (\underbrace{1, 1, \dots, 1}_n, \underbrace{0, \dots, 0}_{N-n})$.

$\hookrightarrow$ Simplest lookup table $\hat{I}[j]$, with $j = 0, 1, 2, \dots, \binom{N}{n}-1$ is simply the ordered array of the basis integers $I_{P(\sigma)}$, i.e. $I[0] = I_\sigma, \dots, I[\binom{N}{n}-1] = I_{\tilde{\sigma}}$.

- Lookup table $\hat{j}[I]$ requires more thought: The naïve array of length 2^N would not only contain many "null" entries, but may also exceed memory resources (going to bigger N was the motivation for implementing symmetry).