

## Exercise 5.1: Lanczos algorithm - ground state and dynamical properties

The file `lanczos.py` contains an implementation of the lanczos algorithm and functions to generate a sparse matrix representing the (by now familiar) Hamiltonian of the transverse field Ising model, where we choose ferromagnetic  $J = 1$  for now. For simplicity, we use the full basis and do not exploit symmetries to block-diagonalize  $H$ .

- a) Generate the Hamiltonian for  $L = 14$  and  $g = 1.5$  using the functions provided in `lanczos.py`. Moreover, generate a random state (in the full basis), which we can use as a starting vector for the Lanczos iteration.
- b) Call the function `lanczos()` with parameters `N=200`, `stabilize=False`. The function returns the tridiagonal matrix  $T$  and orthonormal basis of the Krylov space generated during the Lanczos iteration. Determine the 10 smallest eigenvalues of  $T$  using `np.linalg.eigvalsh`. Do you find a ground state degeneracy? Do you expect a degeneracy for these parameters?

Call `lanczos()` again with `stabilize=True`. What does this option do? Do you get the expected degeneracy now? Can you explain this? Confirm the results by comparing with the (quasi-exact) energies returned by `scipy.sparse.linalg.eigsh`.

- c) Find the ground state  $|u_0\rangle$  of  $T$  and use it to find the ground state  $|\psi_0\rangle$  of  $H$  in the full basis. Check that the state you obtain is normalized, has the correct energy  $E_0 = \langle \psi_0 | H | \psi_0 \rangle$ . Calculate the variance  $\langle \psi_0 | H^2 | \psi_0 \rangle - \langle \psi_0 | H | \psi_0 \rangle^2$  to see if  $|\psi_0\rangle$  is an eigenstate of  $H$ .

*Hint:* Let  $V$  denote the matrix containing the vectors returned by `lanczos()` as columns, then  $V^\dagger H V = T$ . Hence, if  $|u\rangle$  is an eigenvector of  $T$ ,  $|\psi_0\rangle := V |u_0\rangle$  is an (approximate) eigenvector of  $H$ .

To determine dynamical properties, we want to calculate

$$I(\hat{O}, z) = -\frac{1}{\pi} \text{Im} \underbrace{\langle \psi_0 | \hat{O}^\dagger \frac{1}{z - H} \hat{O} | \psi_0 \rangle}_{\equiv x_0} \quad \text{with } z = \omega + E_0 + i\epsilon \quad (1)$$

Recall from class that this can be done by starting a Lanczos iteration from  $|\phi_0\rangle = \hat{O} |\psi_0\rangle$ , which gives us the tridiagonal matrix  $T$  such that

$$z - T = \begin{pmatrix} z - \alpha_0 & -\beta_1 & & \\ -\beta_1 & z - \alpha_1 & -\beta_2 & \\ & -\beta_2 & z - \alpha_1 & \ddots \\ & & \ddots & \ddots \end{pmatrix} \quad (2)$$

Using Cramer's Rule, we can evaluate  $x_0 = \frac{\det A}{\det(z-T)}$ , where  $A$  is  $z - T$  with the first column replaced by the unit vector  $(1, 0, \dots)^T$ . This leads to the continued fraction

$$\begin{aligned} x_0 &= \frac{1}{z - \alpha_0 - \beta_1^2 \frac{\det D_2}{\det D_1}}, \\ \frac{\det D_2}{\det D_1} &= \frac{1}{z - \alpha_1 - \beta_2^2 \frac{\det D_3}{\det D_2}}, \\ \frac{\det D_3}{\det D_2} &= \frac{1}{z - \alpha_2 - \beta_3^2 \frac{\det D_4}{\det D_3}}, \dots \end{aligned} \quad (3)$$

where  $D_n$  is the matrix  $z - T$  with the first  $n$  rows and columns omitted. For a  $N \times N$  matrix  $T$ , the last ratio evaluates as

$$\frac{\det D_{N-1}}{\det D_{N-2}} = \frac{1}{z - \alpha_{N-2} - \frac{\beta_{N-1}^2}{z - \alpha_{N-1}}} \quad (4)$$

- d) Construct the state  $|\phi_0\rangle = S_0^+ |\psi_0\rangle$  and get the tridiagonal matrix  $T$  of a Lanczos iteration starting from this initial state. Here,  $S_j^+ = \frac{1}{2}(\sigma_j^x + i\sigma_j^y)$  labels the spin-raising operator on site  $j$ .

*Hint:* To construct  $S_0^+$ , take a look at `lanczos.py` again.

- e) Write a function which (given  $z$  and  $T$ ) evaluates the continued fractions of eq. (3) to calculate  $I$ .

*Hint:* You can extract  $\alpha$  and  $\beta$  from  $T$  using `np.diag`. Take care that we labeled  $\alpha_n$  starting from  $n = 0$ , while for  $\beta_n$  we start from  $n = 1$ .

- f) Plot  $I(S_0^+, \omega)$  versus  $\omega$ . Choose  $z = \omega + E_0 + i\epsilon$  for  $\omega \in [-1, 10]$  and  $\epsilon$  in the order of  $0.001 \lesssim \epsilon \lesssim 0.1$ . What is the influence of  $\epsilon$ ?

*Hint:* There's a good chance that the function calculating  $I$  from e) works with  $z$  being a numpy array with different  $\omega$  values; this will lead to a faster evaluation than calling it for each  $\omega$  separately.

- g) To get more physical insight, we can choose a momentum-dependent operator for  $\hat{O}$ . Calculate  $I(S_k^+, \omega)$  for the  $k$  values compatible with the chosen  $L = 14$ , where  $S_k^+$  is defined as

$$S_k^+ = \frac{1}{\sqrt{L}} \sum_{j=0}^{L-1} e^{ijk} S_j^+. \quad (5)$$

Create a 2D colorplot of  $I(S_k^+, \omega)$  with  $k$  on the  $x$ -axis and  $\omega$  on the  $y$ -axis.

*Hint:* To create the colorplot, you can use the function provided in `lanczos.py`.

- h) Regenerate similar colorplots for other values of  $g$  in both phases and at the critical point, both for ferromagnetic  $J = 1$  and antiferromagnetic  $J = -1$ .

Bonus Use for  $\hat{O}$  the fermionic creation ( $c_k^\dagger$ ) and annihilation ( $c_k$ ) from the Jordan-Wigner transformation, defined by

$$c_k^\dagger = \frac{1}{\sqrt{L}} \sum_j e^{-ij k} c_j^\dagger, \quad c_j^\dagger = \sigma^z \otimes \dots \otimes \sigma^z \otimes S_j^- \otimes \mathbb{1} \otimes \dots \otimes \mathbb{1}. \quad (6)$$