

# Complex Systems Exercises

## Exercise 1.0.1 (Ising Model):

Consider a 1-dimensional Ising Model with nearest-neighbour ferromagnetic interaction in an external uniform field with energy function given by:

$$\mathcal{H}(\boldsymbol{\sigma}) = -J \sum_{x=1}^N \sigma_x \sigma_{x+1} - B \sum_{x=1}^N \sigma_x \quad J > 0$$

where periodic boundary conditions are used, i.e.  $\sigma_{N+1} \equiv \sigma_1$ . Define  $K \equiv \beta J$  and  $h \equiv \beta B$ .

**Part A.** Using the transfer matrix  $T(\sigma, \sigma') = \exp(K\sigma\sigma' + h(\sigma + \sigma')/2)$  and its spectral decomposition, determine:

1. The **partition function**  $Z(K, h)$
2. The **free energy** per node in the thermodynamic limit and its plot for  $h = 0$  versus  $1/K$
3. The **entropy** per node in the thermodynamic limit and its plot for  $h = 0$  versus  $1/K$
4. The **mean energy** per node in the thermodynamic limit and its plot for  $h = 0$  versus  $1/K$
5. The **specific heat** per node in the thermodynamic limit and its plot for  $h = 0$  versus  $1/K$
6. The **average magnetization** at  $x$ ,  $\langle \sigma_x \rangle$ , in the thermodynamic limit and its plot for  $h = 0, 0.1, 0.2, 0.5, 1$  versus  $1/K$  and for  $K = 1$  versus  $h$  in the range  $(-5, 5)$
7. The **two-point correlation function**  $\langle \sigma_x \sigma_{x+y} \rangle$  in the thermodynamic limit and its plot for  $h = 0$  and  $K = 1$  versus  $y$ .

**Part B.** Consider the same model with **open** boundary conditions (node 1 is linked only to node 2, and node  $N$  only to node  $N - 1$ ):

$$\mathcal{H}(\boldsymbol{\sigma}) = -J \sum_{x=1}^{N-1} \sigma_x \sigma_{x+1} - B \sum_{x=1}^N \sigma_x$$

Show that the partition function for this case can be formally written as:

$$Z(K, h) = \mathbf{v}^T \mathbf{T}^N \mathbf{v} \equiv \sum_{\substack{\sigma_1 = \pm 1 \\ \sigma_N = \pm 1}} v(\sigma_1) \mathbf{T}^N(\sigma_1, \sigma_N) v(\sigma_N)$$

where  $v(\sigma) = e^{h\sigma/2}$ . Show that the free energy per node in the thermodynamic limit is the same as above.

**Part C.** Same as in part *B* with fixed boundary conditions  $\sigma_1 = 1 = \sigma_N$ , and  $v(\sigma) = e^{h/2}$  for both  $\sigma = \pm 1$ .

**Part D.** How would you try to solve the Ising model in 1-dimension with nearest neighbour and next-to-nearest neighbour interaction and periodic boundary condition ( $\sigma_{N+1} = \sigma_1$  and  $\sigma_{N+2} = \sigma_2$ ):

$$\mathcal{H}(\boldsymbol{\sigma}) = - \sum_{x=1}^N (J_1 \sigma_x \sigma_{x+1} + J_2 \sigma_x \sigma_{x+2}) - B \sum_{x=1}^N \sigma_x$$

## 1.1 Solution

### 1.1.1 Part A

Consider a  $d = 1$  system with  $N$  spins. The periodic boundary conditions are  $\sigma_{N+1} = \sigma_1$ .

1. The **partition function** is given by:

$$\begin{aligned} Z(K, h) &= \sum_{\{\boldsymbol{\sigma}\}} e^{-\beta \mathcal{H}(\boldsymbol{\sigma})} = \sum_{\{\boldsymbol{\sigma}\}} \exp \left( K \sum_{x=1}^N \sigma_x \sigma_{x+1} + h \sum_{x=1}^N \sigma_x \right) = \\ &= \sum_{\{\boldsymbol{\sigma}\}} \prod_{x=1}^N \exp \left( \underbrace{K \sigma_x \sigma_{x+1} + h \frac{\sigma_x + \sigma_{x+1}}{2}}_{T_{\sigma_x, \sigma_{x+1}}} \right) = \\ &= \sum_{\sigma_1=\pm 1} \cdots \sum_{\sigma_N=\pm 1} T_{\sigma_1 \sigma_2} T_{\sigma_2 \sigma_3} \cdots T_{\sigma_{N-1} \sigma_N} T_{\sigma_N \sigma_1} = \\ &= \sum_{\sigma_1=\pm 1} \underbrace{(T \cdots T)}_{N \text{ times}}_{\sigma_1 \sigma_1} = \sum_{\sigma_1=\pm 1} (T^N)_{\sigma_1 \sigma_1} = \text{Tr } T^N \end{aligned}$$

where  $T$  is a  $2 \times 2$  matrix given by:

$$T = \begin{bmatrix} \sigma' = +1 & \sigma' = -1 \\ e^{K+h} & e^{-K} \\ e^{-K} & e^{K-h} \end{bmatrix} \begin{matrix} \sigma = +1 \\ \sigma = -1 \end{matrix} \quad (1.1)$$

As the trace is basis-independent, we can compute it in the basis that diagonalizes  $T$ . Let  $\lambda_1$  and  $\lambda_2$  be the eigenvalues of  $T$ , with  $\lambda_1 < \lambda_2$ . By solving:

$$\det(T - \lambda \mathbb{I}) \stackrel{!}{=} 0$$

we find:

$$\lambda_{1,2} = e^K \cosh h \mp \sqrt{e^{2K} \sinh^2 h + e^{-2K}} \quad (1.2)$$

When diagonalized,  $T = \text{diag}(\lambda_1, \lambda_2)$ , and so:

$$Z(K, h) = \text{Tr } T^N = (\lambda_1^N + \lambda_2^N) \quad (1.3)$$

2. The **free energy** per node  $f(K, h)$  is defined by the relation:

$$Z = e^{-\beta N f(K, h)} \stackrel{(1.3)}{=} (\lambda_1^N + \lambda_2^N)$$

Taking the ln of both sides and dividing by  $N$  leads to:

$$\frac{\ln Z}{N} = -\beta f(K, h) = \frac{1}{N} \ln(\lambda_1^N + \lambda_2^N)$$

In the thermodynamic limit  $N \rightarrow +\infty$  only the greatest eigenvalue ( $\lambda_1$ ) will dominate:

$$\begin{aligned} &= \frac{1}{N} \ln \left( \lambda_2^N \left[ 1 + \left( \frac{\lambda_1}{\lambda_2} \right)^N \right] \right) = \\ &\xrightarrow{N \rightarrow +\infty} \frac{1}{N} N \ln \lambda_2 = \ln \lambda_2 \end{aligned}$$

Thus:

$$f(K, h) = -\frac{1}{\beta} \ln \lambda_2 = -\frac{1}{\beta} \ln [e^K \cosh h + \sqrt{e^{2K} \sinh^2 h + e^{-2K}}] \quad (1.4)$$

When  $h = 0$ :

$$\begin{aligned} f(K, h = 0) &= -\frac{1}{\beta} \ln [e^K \cdot 1 + \sqrt{e^{2K} \cdot 0 + e^{-2K}}] = \\ &= -\frac{J}{\beta J} \ln 2 \frac{[e^K + e^{-K}]}{2} = -\frac{J}{K} \ln(2 \cosh K) \end{aligned} \quad (1.5)$$

where  $K = \beta J = J/(k_B T)$ , and so  $1/K \propto T$ .

As a function of  $T$  (or  $1/K$ ), we have that:

$$f(K, 0) \xrightarrow{T \rightarrow 0^+} -J$$

And for large  $T$ :

$$f(K, 0) \underset{T \gg 1}{\sim} -JT \log 2$$

So, if  $J = k_B = 1$ ,  $f(K, 0)$  stationarizes at  $-1$  for  $T \rightarrow 0^+$ , and goes to  $-\infty$  linearly (with a  $\log 2$  factor) as  $T \rightarrow +\infty$ .

A plot of  $f(1/K)$  is shown in fig. 1.1a.

3. The **entropy** per node is obtained by differentiating the free energy per node:

$$s \equiv -\frac{\partial f}{\partial T} = -\frac{\partial f}{\partial \beta} \frac{\partial \beta}{\partial T} = \frac{1}{k_B T^2} \frac{\partial f}{\partial \beta} \quad (1.6)$$

Using (1.4) we get:

$$\begin{aligned} \frac{\partial f}{\partial \beta} &= +\frac{1}{\beta^2} \ln [e^K \cosh h + \sqrt{e^{2K} \sinh^2 h + e^{-2K}}]. \\ &\left( e^K J \cosh h + e^K B \sinh h + \frac{1}{2\sqrt{e^{2K} \sinh^2 h + e^{-2K}}} \right). \end{aligned}$$

$$\cdot \left[ 2e^{2K} J \sinh^2 h + 2e^{2K} B \cosh h \sinh h - 2Je^{-2K} \right]$$

Taking  $h = 0$ :

$$\begin{aligned} \frac{\partial f}{\partial \beta} \Big|_{h=0} &= \frac{1}{\beta^2} \ln 2 \frac{[e^K + e^{-K}]}{2} - \frac{J\beta}{\beta \cdot \beta} \frac{e^K - e^{-K}}{e^K + e^{-K}} = \\ &= \frac{1}{\beta^2} \ln[2 \cosh(K)] - \frac{K}{\beta^2} \tanh(K) \end{aligned}$$

The result is the same we would have obtained by directly differentiating (1.5), since:

$$\frac{\partial f}{\partial \beta}(K, h) = \frac{\partial f}{\partial K} \frac{\partial K}{\partial \beta} + \underbrace{\frac{\partial f}{\partial h} \frac{\partial h}{\partial \beta}}_B$$

and  $h = 0 \Rightarrow B = 0$ , meaning that the rightmost term vanishes (assuming that  $\frac{\partial f}{\partial h}$  is well behaved).

Substituting back in (1.6) we get:

$$s = k_B \left[ \ln(2 \cosh K) - K \tanh K \right] \quad (1.7)$$

A plot of  $s(1/K)$  is shown in fig. 1.1b.

4. The **mean energy** is given by:

$$\begin{aligned} \langle \epsilon \rangle &= -\frac{\partial}{\partial \beta} \frac{\ln Z}{N} = -\frac{\partial}{\partial \beta} (\beta f(K, h)) = \\ &= \frac{1}{e^K \cosh h + \sqrt{e^{2K} \sinh^2 h + e^{-2K}}} \left( e^K J \cosh h + e^K B \sinh h + \right. \\ &\quad \left. + \frac{1}{2\sqrt{e^{2K} \sinh^2 h + e^{-2K}}} \cdot \left[ 2e^{2K} J \sinh^2 h + 2e^{2K} B \sinh h \cosh h - 2Je^{-2K} \right] \right) \end{aligned}$$

When  $h = 0$ :

$$\epsilon(K, h = 0) = -J \frac{e^K - e^{-K}}{e^K + e^{-K}} = -J \tanh K \quad (1.8)$$

which is plotted in fig. 1.1c.

Note that, alternatively, we could have used the free energy definition to determine  $\epsilon$ :

$$f = \epsilon - Ts$$

5. The **specific heat per node** is defined as:

$$c \equiv \frac{\partial \epsilon}{\partial T} = -\frac{\partial \epsilon}{\partial \beta} \frac{\partial \beta}{\partial T} = -k_B \beta^2 \frac{\partial \epsilon}{\partial \beta}$$

Since the expression is quite complicated, we use the argument we made when computing the entropy to take  $h = 0$  *before* computing the derivative:

$$c(h = 0) = -k_B \beta^2 \frac{\partial}{\partial \beta} (-J \tanh K) = \frac{k_B K^2}{\cosh^2 K}$$

A plot of  $c(1/K)$  is shown in fig. 1.1d.

6. The **magnetization** can be directly computed by differentiating the free energy:

$$\begin{aligned}
\langle \sigma_x \rangle &= -\beta \frac{\partial}{\partial h} f(K, h) = \frac{\partial}{\partial h} (-\beta f(K, h)) = \\
&\stackrel{(1.4)}{=} \frac{\partial}{\partial h} \ln \left( e^K \cosh h + \sqrt{e^{2K} \sinh^2 h + e^{-2K}} \right) = \\
&= \frac{1}{e^K \cosh h + \sqrt{e^{2K} \sinh^2 h + e^{-2K}}} \left( e^K \sinh h + \frac{2e^{2K} \sinh h \cosh h}{2\sqrt{e^{2K} \sinh^2 h + e^{-2K}}} \right) = \\
&= e^K \sinh h \left[ 1 + \frac{e^K \cosh h}{\sqrt{e^{2K} \sinh^2 h + e^{-2K}}} \right] \frac{1}{e^K \cosh h + \sqrt{e^{2K} \sinh^2 h + e^{-2K}}} = \\
&= e^K \sinh h \frac{\sqrt{e^{2K} \sinh^2 h + e^{-2K}} + e^K \cosh h}{\sqrt{e^{2K} \sinh^2 h + e^{-2K}}} \frac{1}{e^K \cosh h + \sqrt{e^{2K} \sinh^2 h + e^{-2K}}} = \\
&= \frac{e^K \sinh h}{\sqrt{e^{2K} \sinh^2 h + e^{-2K}}} \tag{1.9}
\end{aligned}$$

For  $h = 0$ :

$$\langle \sigma_x \rangle \equiv 0$$

Plots of  $\langle \sigma_x \rangle$  as function of  $h$  and  $K$  are shown in fig. 1.1e and 1.1f.

Alternatively, we can use the transfer matrix  $T$ . We start by writing explicitly the average:

$$\langle \sigma_x \rangle = \frac{1}{Z} \sum_{\{\sigma\}} \sigma_x e^{-\beta \mathcal{H}(\sigma)} = \frac{1}{Z} \sum_{\sigma_1=\pm 1} \cdots \sum_{\sigma_N=\pm 1} T_{\sigma_1 \sigma_2} \cdots T_{\sigma_{x-1} \sigma_x} \sigma_x T_{\sigma_x \sigma_{x+1}} \cdots T_{\sigma_N \sigma_1}$$

If we define:

$$T'_{\sigma_i \sigma_j} \equiv \sigma_i T_{\sigma_i \sigma_j} \tag{1.10}$$

We can still write the sum over all spin configurations as the trace of a matrix product:

$$\langle \sigma_x \rangle = \frac{1}{Z} \text{Tr}(T^{x-1} T' T^{N-x})$$

Then, using the cyclic property of the trace:

$$\text{Tr}(ABC) = \text{Tr}(CAB) = \text{Tr}(BCA)$$

we get:

$$\langle \sigma_x \rangle = \frac{1}{Z} \text{Tr}(T' T^{N-x} T^{x-1}) = \frac{1}{Z} \text{Tr}(T' T^{N-1}) \quad \forall x$$

As expected,  $\langle \sigma_x \rangle$  does not depend on  $x$ , since the system is translational invariant.

Explicitly,  $T'$  is given by:

$$T' = \begin{bmatrix} \sigma'=+1 & \sigma'=-1 \\ e^{K+h} & e^{-K} \\ -e^{-K} & -e^{K-h} \end{bmatrix} \begin{matrix} \sigma=+1 \\ \sigma=-1 \end{matrix}$$

Note that it can be written as:

$$\mathbb{T}' = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \mathbb{T} = \sigma_z \mathbb{T}$$

where  $\sigma_z$  is the third Pauli matrix (not to be confused with the  $z$ -th spin). Thus:

$$\langle \sigma_x \rangle = \frac{1}{Z} \text{Tr}(\sigma_z \mathbb{T}^N)$$

As before, since the trace is basis independent, this computation is easier in the basis that diagonalizes  $\mathbb{T}$ . Let  $|v_{1,2}\rangle$  be the two eigenvectors of  $\mathbb{T}$ , with eigenvalues  $\lambda_{1,2}$ . In the basis  $\{|v_{1,2}\rangle\}$ ,  $\mathbb{T}^N = \text{diag}(\lambda_1^N, \lambda_2^N)$ , while  $\sigma_z$  becomes:

$$V^{-1} \sigma_z V = \begin{pmatrix} \langle v_1 | \sigma_z | v_1 \rangle & \langle v_1 | \sigma_z | v_2 \rangle \\ \langle v_2 | \sigma_z | v_1 \rangle & \langle v_2 | \sigma_z | v_2 \rangle \end{pmatrix} \quad (1.11)$$

So, the argument of the trace in the  $\{|v_{\pm}\rangle\}$  basis is:

$$V^{-1} \sigma_z V \text{diag}(\lambda_1^N, \lambda_2^N) = \begin{pmatrix} \lambda_1^N \langle v_1 | \sigma_z | v_1 \rangle & \lambda_2^N \langle v_1 | \sigma_z | v_2 \rangle \\ \lambda_1^N \langle v_1 | \sigma_z | v_1 \rangle & \lambda_2^N \langle v_2 | \sigma_z | v_2 \rangle \end{pmatrix}$$

Thus:

$$\langle \sigma_z \rangle = \frac{1}{Z} \left[ \lambda_1^N \langle v_1 | \sigma_z | v_1 \rangle + \lambda_2^N \langle v_2 | \sigma_z | v_2 \rangle \right] \quad (1.12)$$

$Z(K, h)$  is given by (1.3), which in the thermodynamic limit becomes:

$$Z(K, h) = (\lambda_1^N + \lambda_2^N) = \lambda_2^N \left( 1 + \left( \frac{\lambda_1}{\lambda_2} \right)^N \right) \xrightarrow{N \rightarrow +\infty} \lambda_2^N \quad \lambda_1 < \lambda_2$$

Substituting back in (1.12) we get:

$$\langle \sigma_z \rangle = \frac{1}{\lambda_2^N} \left[ \lambda_1^N \langle v_1 | \sigma_z | v_1 \rangle + \lambda_2^N \langle v_2 | \sigma_z | v_2 \rangle \right] \xrightarrow{N \rightarrow +\infty} \langle v_2 | \sigma_z | v_2 \rangle \quad (1.13)$$

All that's left is to find the eigenvectors  $\{|v_{1,2}\rangle\}$  and compute the required matrix element.

Since  $\mathbb{T}$  is a  $2 \times 2$  matrix, we can write it as a *linear combination* of the Pauli Matrices  $\sigma_{x,y,z}$ , which, together with the identity  $\mathbb{1}$ , form a basis of  $\mathcal{M}_{2 \times 2}(\mathbb{C})$ . In other words, any  $2 \times 2$  matrix  $M$  can be written as:

$$M = a_0 \mathbb{1} + a_1 \sigma_x + a_2 \sigma_y + a_3 \sigma_z$$

with:

$$\sigma_x \equiv \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}; \quad \sigma_y \equiv \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix}; \quad \sigma_z \equiv \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

We can define a *vector of matrices*  $\boldsymbol{\sigma} \equiv (\sigma_x, \sigma_y, \sigma_z)^T$ , and write:

$$M = a_0 \mathbb{1} + \mathbf{a} \cdot \boldsymbol{\sigma} = a_0 \mathbb{1} + \underbrace{\|\mathbf{a}\|}_a (\hat{\mathbf{n}} \cdot \boldsymbol{\sigma})$$

where  $\hat{\mathbf{n}} = (n_x, n_y, n_z)^T$  is a unitary vector ( $\|\hat{\mathbf{n}}\| = 1$ ), and:

$$\hat{\mathbf{n}} \cdot \boldsymbol{\sigma} = \begin{pmatrix} n_z & n_x - in_y \\ n_x + in_y & -n_z \end{pmatrix}$$

This makes it simpler to find eigenvalues and eigenvectors of M. In fact, if  $\mathbf{v}$  is an eigenvector of  $\hat{\mathbf{n}} \cdot \boldsymbol{\sigma}$  with eigenvalue  $\lambda$ :

$$(\hat{\mathbf{n}} \cdot \boldsymbol{\sigma})\mathbf{v} = \lambda\mathbf{v}$$

then  $\mathbf{v}$  is an eigenvector also of M, but with eigenvalue  $a_0 + a\lambda$ :

$$M\mathbf{v} = [a_0\mathbb{1} + a(\hat{\mathbf{n}} \cdot \boldsymbol{\sigma})]\mathbf{v} = a_0\mathbf{v} + a\lambda\mathbf{v} = (a_0 + a\lambda)\mathbf{v}$$

The eigenvalues of  $\hat{\mathbf{n}} \cdot \boldsymbol{\sigma}$  are  $\pm 1$ :

$$\det(\hat{\mathbf{n}} \cdot \boldsymbol{\sigma} - \lambda\mathbb{1}) = \lambda^2 - \|\hat{\mathbf{n}}\|^2 \Rightarrow \lambda^2 = 1 \Rightarrow \lambda = \pm 1$$

If we parametrize the unit vector  $\hat{\mathbf{n}}$  in spherical coordinates:

$$n_x = \sin \theta \cos \varphi \tag{1.14}$$

$$n_y = \sin \theta \sin \varphi \tag{1.15}$$

$$n_z = \cos \theta$$

Then a pair of *orthonormal* eigenvectors of  $\hat{\mathbf{n}} \cdot \boldsymbol{\sigma}$  is given by:

$$|v_1\rangle = \begin{pmatrix} -\sin \frac{\theta}{2} \\ \cos \frac{\theta}{2} \end{pmatrix} \quad |v_2\rangle = \begin{pmatrix} \cos \frac{\theta}{2} \\ \sin \frac{\theta}{2} \end{pmatrix}$$

$|v_1\rangle$  corresponds to  $\lambda = -1$  and  $|v_2\rangle$  to  $\lambda_2 = +1$ .

So, let's write T in the Pauli basis, using the Hilbert-Schmidt inner product ( $\langle A, B \rangle = \text{Tr}(AB^*)$ ) to find the coefficients  $a_{0,1,2,3}$ :

$$T = \underbrace{\frac{\text{Tr}(T\mathbb{1})}{2}}_{a_0} \mathbb{1} + \underbrace{\frac{\text{Tr}(T\sigma_x)}{2}}_{a_1} \sigma_x + \underbrace{\frac{\text{Tr}(T\sigma_y)}{2}}_{a_2} \sigma_y + \underbrace{\frac{\text{Tr}(T\sigma_z)}{2}}_{a_3} \sigma_z$$

In our case:

$$a_0 = e^K \frac{e^h + e^{-h}}{2} = e^K \cosh h \tag{1.16}$$

$$a_1 = \frac{2e^{-K}}{2} = e^{-K}$$

$$a_2 = 0$$

$$a_3 = e^K \frac{e^h - e^{-h}}{2} = e^K \sinh h$$

Thus:

$$a = \|\mathbf{a}\|^2 = \sqrt{a_1^2 + a_2^2 + a_3^2} = \sqrt{e^{-2K} + e^{2K} \sinh^2 h}$$

$$\hat{\mathbf{n}} = \frac{\mathbf{a}}{a}$$

Since the eigenvectors of  $\mathbb{T}$  are the same of  $\hat{\mathbf{n}} \cdot \boldsymbol{\sigma}$  we can evaluate (1.13):

$$\langle v_2 | \sigma_z | v_2 \rangle = \cos^2 \frac{\theta}{2} - \sin^2 \frac{\theta}{2} = \cos \theta = n_z = \frac{a_3}{a} = \frac{e^K \sinh h}{\sqrt{e^{-2K} + e^{2K} \sinh^2 h}}$$

which coincides with the result we got in (1.9).

## 7. Two-point correlation

The same spectral method used to compute the magnetization can be used also for the two-point correlation  $\langle \sigma_x \sigma_{x+y} \rangle$ . As before, we start by explicitly writing the average:

$$\begin{aligned} \langle \sigma_x \sigma_{x+y} \rangle &= \frac{1}{Z} \sum_{\{\boldsymbol{\sigma}\}} \sigma_x \sigma_{x+y} e^{-\beta \mathcal{H}(\boldsymbol{\sigma})} = \\ &= \frac{1}{Z} \sum_{\sigma_1=\pm 1} \cdots \sum_{\sigma_N=\pm 1} \mathbb{T}_{\sigma_1 \sigma_2} \cdots \mathbb{T}_{\sigma_{x-1} \sigma_x} \sigma_x \mathbb{T}_{\sigma_x \sigma_{x+1}} \cdots \\ &\quad \cdots \mathbb{T}_{\sigma_{x+y-1} \sigma_{x+y}} \sigma_{x+y} \mathbb{T}_{\sigma_{x+y} \sigma_{x+y+1}} \cdots \mathbb{T}_{\sigma_N \sigma_1} = \\ &= \frac{1}{Z} \text{Tr}(\mathbb{T}^{x-1} \sigma_x \mathbb{T}^y \sigma_{x+y} \mathbb{T}^{N-x-y+1}) = \\ &\stackrel{(a)}{=} \frac{1}{Z} \text{Tr}(\sigma_z \mathbb{T}^y \sigma_z \mathbb{T}^{N-y}) \end{aligned}$$

where in (a) we used the cyclic property of the trace, and the fact that  $\sigma_n \mathbb{T}_{\sigma_n \sigma_{n+1}}$  is equivalent to  $\mathbb{T}' = \sigma_z \mathbb{T}$ , where  $\sigma_z$  is the third Pauli matrix.

In the continuum limit  $Z = \lambda_2^N$ . If we compute the trace in the basis  $|v_{1,2}\rangle$  that diagonalizes  $\mathbb{T}$ , we get:

$$\langle \sigma_x \sigma_{x+y} \rangle = \frac{1}{\lambda_2^N} \left[ \langle v_1 | \sigma_z \mathbb{T}^y \sigma_z | v_1 \rangle \lambda_1^{N-y} + \langle v_2 | \sigma_z \mathbb{T}^y \sigma_z | v_2 \rangle \lambda_2^{N-y} \right]$$

and since  $\lambda_1 < \lambda_2$ , when  $N \rightarrow +\infty$  the first term vanishes, leaving:

$$\langle \sigma_x \sigma_{x+y} \rangle = \frac{\langle v_2 | \sigma_z \mathbb{T}^y \sigma_z | v_2 \rangle}{\lambda_2^y}$$

Be careful not to mix different bases! The matrix product can be done in the canonical basis - but it's difficult since here  $\mathbb{T}$  has the form (1.1), thus making  $\mathbb{T}^y$  quite hard to compute. A better choice is to compute everything in the  $|v_{1,2}\rangle$  basis, where  $\mathbb{T} = \text{diag}(\lambda_1, \lambda_2)$ ,  $|v_1\rangle = (1, 0)^T$ ,  $|v_2\rangle = (0, 1)^T$  and  $\sigma_z$  is given by (1.11), i.e.:

$$\sigma_z = \begin{pmatrix} -\cos \theta & -\sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix}$$

An even better choice is to use completeness:

$$\langle v_2 | \sigma_z \mathbb{T}^y \sigma_z | v_2 \rangle = \sum_{i,j=1}^2 \langle v_2 | \sigma_z | v_i \rangle \langle v_i | \mathbb{T}^y | v_j \rangle \langle v_j | \sigma_z | v_2 \rangle$$

Since  $|v_{1,2}\rangle$  diagonalize  $\mathbb{T}$ , we have:

$$\langle v_1 | \mathbb{T}^y | v_1 \rangle = \lambda_1^y \quad \langle v_2 | \mathbb{T}^y | v_2 \rangle = \lambda_2^y \quad \langle v_1 | \mathbb{T}^y | v_2 \rangle = \langle v_2 | \mathbb{T}^y | v_1 \rangle = 0$$



Thus:

$$\langle v_2 | \sigma_z T^y \sigma_z | v_2 \rangle = \langle v_2 | \sigma_z | v_1 \rangle \lambda_1^y \langle v_1 | \sigma_z | v_2 \rangle + \langle v_2 | \sigma_z | v_2 \rangle \lambda_2^y \langle v_2 | \sigma_z | v_2 \rangle$$

Note that we already computed  $\langle v_2 | \sigma_z | v_2 \rangle = \cos \theta$ , and so we just need  $\langle v_2 | \sigma_z | v_1 \rangle$ , which is equal to  $\langle v_1 | \sigma_z | v_2 \rangle$  since  $\sigma_z$  is symmetric in the canonical basis, and symmetry is preserved in an orthonormal change of basis. We then find  $\langle v_1 | \sigma_z | v_2 \rangle = -\sin \theta$  and so:

$$\langle \sigma_x \sigma_{x+y} \rangle = \frac{\lambda_1^y \sin^2 \theta + \lambda_2^y \cos^2 \theta}{\lambda_2^y} = \cos^2 \theta + \left( \frac{\lambda_1}{\lambda_2} \right)^y \sin^2 \theta$$

$\lambda_{1,2}$  have been computed in (1.2), and from the parameterization of  $\hat{\mathbf{n}}$  (1.14) we have  $\cos^2 \theta = n_3^2 = (a_3/a)^2$  and  $\sin^2 \theta = n_x^2 + n_y^2 = (a_1/a)^2$ , with the values found in (1.16).

Since  $\lambda_1 < \lambda_2$ , when  $y \rightarrow +\infty$  the second term vanishes, and:

$$\langle \sigma_x \sigma_{x+y} \rangle \xrightarrow{y \rightarrow \infty} \cos^2 \theta = \cos \theta \cdot \cos \theta = \langle \sigma_x \rangle \langle \sigma_{x+y} \rangle$$

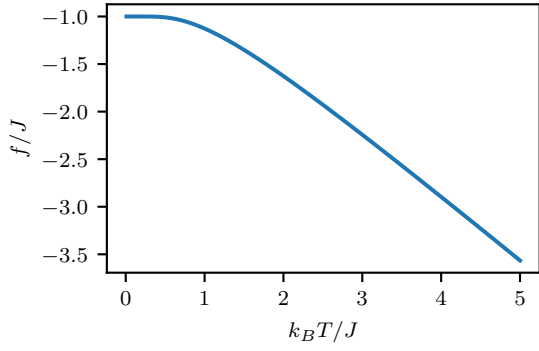
This means that two *spins* that are *infinitely* far apart are effectively *independent*.

When  $h = 0$ , the two-point correlation reduces to:

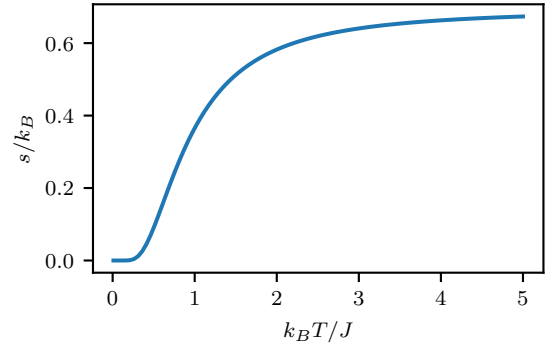
$$\langle \sigma_x \sigma_{x+y} \rangle = \left( \frac{e^K + e^{-K}}{e^K - e^{-K}} \right)^y = (\tanh K)^y$$

which coincides with the result already found in section 4.3.1 of the main notes, where we used *open* boundary conditions instead of periodic ones (in the thermodynamic limit they are effectively the same, as we will see in part B).

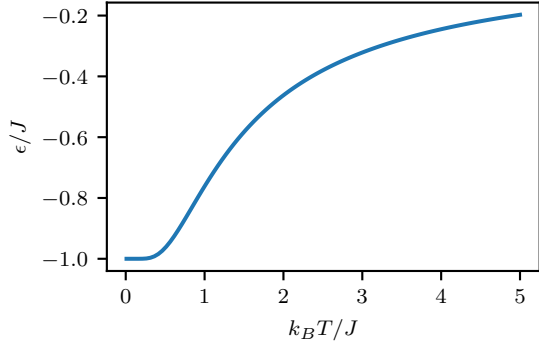
A plot of  $\langle \sigma_x \sigma_{x+y} \rangle$  as a function of  $y$  is shown in fig. 1.1g.



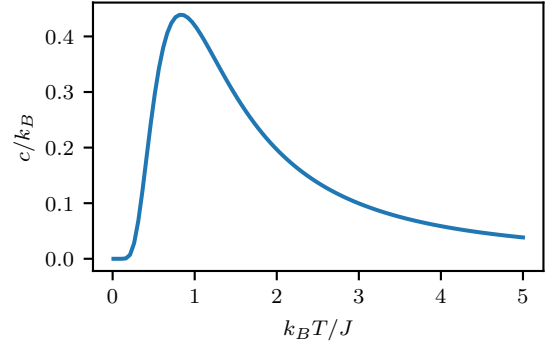
(a) Free energy  $f$  per node ( $h = 0$ ).



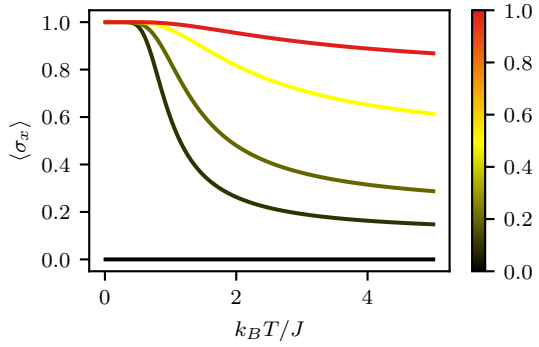
(b) Entropy  $s$  per node ( $h = 0$ ).



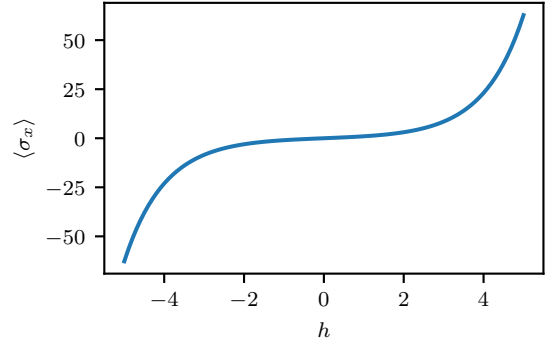
(c) Mean energy  $\epsilon$  per node ( $h = 0$ ).



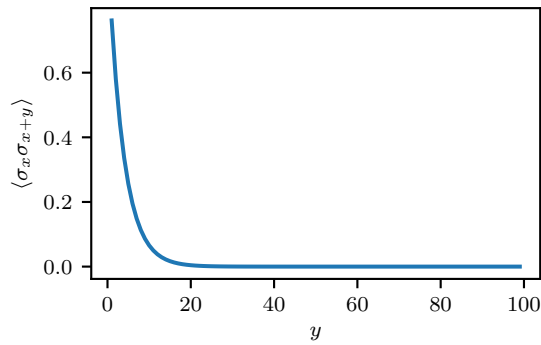
(d) Specific heat  $c$  per node ( $h = 0$ ).



(e) Average magnetization  $\langle \sigma_x \rangle$  for  $h = 0, 0.1, 0.2, 0.5, 1$ .



(f) Average magnetization  $\langle \sigma_x \rangle$  for  $K = 1$ .



(g) Two-point correlation function  $\langle \sigma_x \sigma_{x+y} \rangle$  for  $h = 0$  and  $K = 1$ .

**Figure (1.1)** – Plots of various quantities of interest. Note that  $K_B T/J = 1/K$ .

### 1.1.2 Part B

Let's consider the same model with **open boundary conditions**. The Hamiltonian is given by:

$$\mathcal{H}(\boldsymbol{\sigma}) = -J \sum_{x=1}^{N-1} \sigma_x \sigma_{x+1} - B \sum_{x=1}^N \sigma_x$$

We begin by computing the partition function  $Z$ :

$$Z = \sum_{\{\boldsymbol{\sigma}\}} e^{-\beta \mathcal{H}(\boldsymbol{\sigma})} = \sum_{\{\boldsymbol{\sigma}\}} \exp \left( K \sum_{x=1}^{N-1} \sigma_x \sigma_{x+1} + h \sum_{x=1}^N \sigma_x \right)$$

We rewrite the sum  $\sum_x \sigma_x$  in as follows:

$$\begin{aligned} \sum_{x=1}^N \sigma_x &= \frac{1}{2} \sum_{x=1}^N (\sigma_x + \sigma_{x+1}) + \sigma_1 + \sigma_N \\ &= \frac{1}{2} \left( \sum_{x=1}^{N-1} (\sigma_x + \sigma_{x+1}) + \sigma_1 + \sigma_N \right) \end{aligned} \quad (1.17)$$

Substituting back:

$$Z = \sum_{\{\boldsymbol{\sigma}\}} \prod_{x=1}^{N-1} \underbrace{\exp \left( K \sigma_x \sigma_{x+1} + h \frac{\sigma_x + \sigma_{x+1}}{2} \right)}_{T_{\sigma_x \sigma_{x+1}}} \exp \left( h \frac{\sigma_1}{2} \right) \exp \left( h \frac{\sigma_N}{2} \right)$$

We define the  $2 \times 2$  transfer matrix  $T$  as:

$$T_{\sigma \sigma'} = \exp \left( K \sigma \sigma' + h \frac{\sigma + \sigma'}{2} \right)$$

and the vector  $\boldsymbol{v} = (v(+1), v(-1))$  as:

$$v(\sigma) = \exp \left( h \frac{\sigma}{2} \right)$$

Leading to:

$$\begin{aligned} Z &= \sum_{\{\boldsymbol{\sigma}\}} \prod_{x=1}^{N-1} T_{\sigma_x \sigma_{x+1}} v(\sigma_1) v(\sigma_N) = \\ &= \sum_{\sigma_1=\pm 1} \sum_{\sigma_2=\pm 1} \cdots \sum_{\sigma_{N-1}=\pm 1} \sum_{\sigma_N=\pm 1} v(\sigma_1) T_{\sigma_1 \sigma_2} \cdots T_{\sigma_{N-1} \sigma_N} v(\sigma_N) = \\ &= \sum_{\sigma_1=\pm 1} \sum_{\sigma_N=\pm 1} v(\sigma_1) (T^{N-1})_{\sigma_1 \sigma_N} v(\sigma_N) = \boldsymbol{v}^T T^{N-1} \boldsymbol{v} = \langle v | T^{N-1} | v \rangle \end{aligned}$$

The scalar product can be computed in the basis  $|v_{1,2}\rangle$  where  $T = \text{diag}(\lambda_1, \lambda_2)$ . The change of basis can be done quickly by using completeness:

$$\langle v | T^{N-1} | v \rangle = \sum_{i,j=1}^2 \langle v | v_i \rangle \lambda_i^{N-1} \langle v_j | v \rangle = \langle v | v_1 \rangle^2 \lambda_1^{N-1} + \langle v | v_2 \rangle^2 \lambda_2^{N-1}$$

There is no necessity of computing  $\langle v|v_1\rangle$  or  $\langle v|v_2\rangle$ , as they won't be significant in the thermodynamic limit.

In fact, let's consider the free energy per node  $f$ :

$$\begin{aligned} \frac{\ln Z}{N} \equiv -\beta f &= \frac{1}{N} \ln \left[ \langle v|v_1\rangle^2 \lambda_1^{N-1} + \langle v|v_2\rangle^2 \lambda_2^{N-1} \right] = \\ &= \frac{1}{N} \ln \left[ \lambda_2^{N-1} \left( \langle v|v_1\rangle^2 \left( \frac{\lambda_1}{\lambda_2} \right)^{N-1} + \langle v|v_2\rangle^2 \right) \right] \end{aligned} \quad (1.18)$$

Since  $\lambda_1 < \lambda_2$ , when  $N \gg 1$ , the first term vanishes, leaving:

$$-\beta f = \frac{N-1}{N} \ln \lambda_2 + \frac{2}{N} \ln \langle v|v_2\rangle \xrightarrow{N \rightarrow +\infty} \ln \lambda_2$$

which is exactly the same result we found in (1.4). This also means that all the thermodynamic quantities we computed in part A have the same expression for the IM with o.b.c.

Intuitively, since in the thermodynamic limit the system is *infinite*, periodic and open boundary conditions are *effectively* the same.

### 1.1.3 Part C

We now consider the case with + boundary conditions:  $\sigma_1 = \sigma_N = +1$ . The partition function becomes:

$$Z = \sum_{\{\sigma\}} e^{-\beta \mathcal{H}(\sigma)} = \sum_{\sigma_2=\pm 1} \cdots \sum_{\sigma_{N-1}=\pm 1} \exp \left( K \sum_{x=1}^{N-1} \sigma_x \sigma_{x+1} + h \sum_{x=1}^N \sigma_x \right) \quad \sigma_1 = \sigma_N \equiv +1$$

We can repeat the argument we used in (1.17), leading to:

$$\begin{aligned} Z &= \sum_{\sigma_2=\pm 1} \cdots \sum_{\sigma_{N-1}=\pm 1} \exp \left( \frac{h}{2} \sigma_1 \right) T_{\sigma_1 \sigma_2} \cdots T_{\sigma_{N-1} \sigma_N} \exp \left( \frac{h}{2} \sigma_N \right) = \\ &= e^{h/2} (T^{N-1})_{1,1} e^{h/2} = \\ &= e^{h/2} \begin{pmatrix} 1 & 0 \end{pmatrix} T^{N-1} \begin{pmatrix} 1 \\ 0 \end{pmatrix} e^{h/2} = \begin{pmatrix} e^{h/2} & 0 \end{pmatrix} T^{N-1} \begin{pmatrix} e^{h/2} \\ 0 \end{pmatrix} = \\ &= \tilde{\mathbf{v}}^T T^{N-1} \tilde{\mathbf{v}} = \langle \tilde{v} | T^{N-1} | \tilde{v} \rangle \end{aligned}$$

where  $\tilde{\mathbf{v}} = (e^{h/2}, 0)^T$ . To compute the scalar product we use again *completeness*:

$$\langle \tilde{v} | T^{N-1} | \tilde{v} \rangle = \langle \tilde{v} | v_1 \rangle^2 \lambda_1^{N-1} + \langle \tilde{v} | v_2 \rangle^2 \lambda_2^{N-1}$$

In the thermodynamic limit, the term  $\lambda_2$  dominates, and everything else (including the prefactors) can be neglected. In fact, by repeating the same computation we did in (1.18), we obtain for the free energy:

$$-\beta f \xrightarrow{N \rightarrow +\infty} \ln \lambda_2$$

### 1.1.4 Part D

Consider the Ising Model with both nearest-neighbours and next-nearest-neighbours interactions in  $d = 1$ :

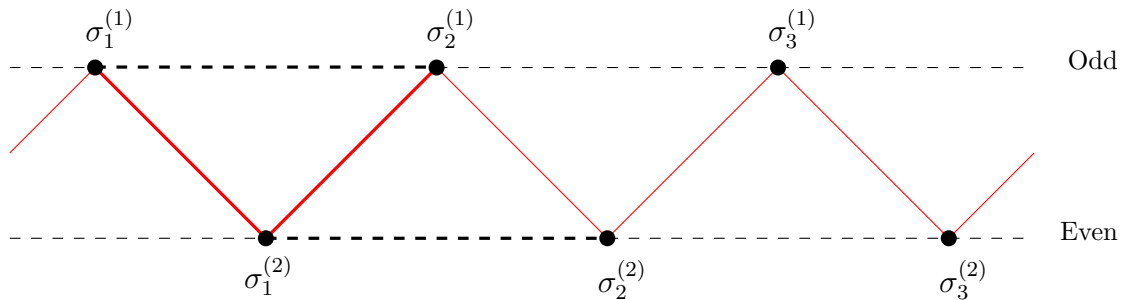
$$\mathcal{H}(\boldsymbol{\sigma}) = - \sum_{x=1}^N (J_1 \sigma_x \sigma_{x+1} + J_2 \sigma_x \sigma_{x+2}) - B \sum_{x=1}^N \sigma_x$$

To compute the partition function  $Z$ , we can still use the same logic from before, i.e. construct a transfer matrix  $T$ . However, we first need to rewrite the Hamiltonian as the product of terms  $T_{\sigma, \sigma'}$ , depending on only *two (consecutive) indices*. As of now, this is not possible - since we have 3 different indices:  $x$ ,  $x + 1$  and  $x + 2$ . There is no way to remove one of them if we want to account for both kind of interactions. So, the *trick* is to *add* a fourth index, and group them by 2, forming some kind of *multi-index* (or binary index).

We can do this by reasoning with **parity**. In fact, note that the nearest-neighbour interactions always involve spins with *different parity*, while the next-to-nearest-neighbour interactions only connect spins with the *same parity*. So, let's *group* the spins in two different *chains* depending on their parity. The first chain will contain all the *odd* spins  $\sigma_i^{(1)} \equiv \sigma_{2i+1}$ , and the second one all the *even* spins  $\sigma_i^{(2)} \equiv \sigma_{2i}$ . With this notation, the nearest neighbours interactions involve always spins of different chains (i.e. different parities), and the next-to-nearest-neighbours interaction always spins from the same chain.

We can then rewrite the Hamiltonian as follows (suppose, for simplicity, that  $N$  is even):

$$\mathcal{H}(\boldsymbol{\sigma}) = -J_1 \left[ \sum_{x=1}^{N/2} \sigma_x^{(1)} \sigma_x^{(2)} + \sigma_x^{(2)} \sigma_{x+1}^{(1)} \right] - J_2 \sum_{x=1}^{N/2} [\sigma_x^{(1)} \sigma_{x+1}^{(1)} + \sigma_x^{(2)} \sigma_{x+1}^{(2)}] - B \sum_{x=1}^{N/2} [\sigma_x^{(1)} + \sigma_x^{(2)}] \quad (1.19)$$



**Figure (1.2)** – Graphical representation of the IM model with both nearest-neighbour and next-to-nearest-neighbour interactions. Spins are represented as black dots, and ordered in two lines (chains) depending on their *parity*. The red continuous lines connect nearest-neighbours ( $J_1$  terms), while the black dashed lines join next-to-nearest-neighbours ( $J_2$  terms). The interactions described by the first term ( $x = 1$ ) of (1.19) are highlighted in bold.

The two *multi-indices* of the transfer matrix will be  $(\sigma_x^{(1)}, \sigma_x^{(2)})$  and  $(\sigma_{x+1}^{(1)}, \sigma_{x+1}^{(2)})$ , and so we need all terms to contain both of them:

$$\begin{aligned} \mathcal{H}(\boldsymbol{\sigma}) = & - \left[ \frac{J_1}{2} \sum_{x=1}^{N/2} [\sigma_x^{(1)} \sigma_x^{(2)} + 2\sigma_x^{(2)} \sigma_{x+1}^{(1)} + \sigma_{x+1}^{(1)} \sigma_{x+1}^{(2)}] + J_2 \sum_{x=1}^{N/2} [\sigma_x^{(1)} \sigma_{x+1}^{(1)} + \sigma_x^{(2)} \sigma_{x+1}^{(2)}] + \right. \\ & \left. + \frac{B}{2} \sum_{x=1}^{N/2} [\sigma_x^{(1)} + \sigma_x^{(2)} + \sigma_{x+1}^{(1)} + \sigma_{x+1}^{(2)}] \right] \end{aligned}$$

The partition function is given by:

$$Z = \sum_{\{\boldsymbol{\sigma}\}} e^{-\beta\mathcal{H}(\boldsymbol{\sigma})} = \sum_{\substack{\sigma_1^{(1)}=\pm 1 \\ \sigma_1^{(2)}=\pm 1}} \cdots \sum_{\substack{\sigma_{N/2}^{(1)}=\pm 1 \\ \sigma_{N/2}^{(2)}=\pm 1}} \prod_{i=1}^{N/2} \exp \left( \frac{K_1}{2} \left[ \sigma_x^{(1)}\sigma_x^{(2)} + 2\sigma_x^{(2)}\sigma_{x+1}^{(1)} + \sigma_{x+1}^{(1)}\sigma_{x+1}^{(2)} \right] \right. \\ \left. K_2 \left[ \sigma_x^{(1)}\sigma_{x+1}^{(1)} + \sigma_x^{(2)}\sigma_{x+1}^{(2)} \right] + \frac{B}{2} \left[ \sigma_x^{(1)} + \sigma_x^{(2)} + \sigma_{x+1}^{(1)} + \sigma_{x+1}^{(2)} \right] \right)$$

The exponential term is one entry of a  $4 \times 4$  transfer matrix  $\mathbb{T}$ :

$$\mathbb{T}_{(\sigma_x^{(1)}, \sigma_x^{(2)}), (\sigma_{x+1}^{(1)}, \sigma_{x+1}^{(2)})}$$

By mapping  $\sigma_x = \pm 1 \rightarrow \{0, 1\}$ , each “multi-index” is a binary number, defining a position in the matrix. For example, when  $\sigma_x^{(1)} = \sigma_x^{(2)} = \sigma_{x+1}^{(1)} = \sigma_{x+1}^{(2)} = +1$ , the matrix entry will be  $\mathbb{T}_{(1,1),(1,1)} \equiv \mathbb{T}_{4,4}$ . In this way, the sum of the product of exponentials can be interpreted as a *matrix product*, leading to:

$$Z = \text{Tr}(\mathbb{T}^{N/2})$$