Classical statistical mechanics approach to multipartite entanglement

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Abstract
We characterize the multipartite entanglement of a system of $n$ qubits in terms of the distribution function of the bipartite purity over balanced bipartitions. We search for maximally multipartite entangled states, whose average purity is minimal, and recast this optimization problem into a problem of statistical mechanics, by introducing a cost function, a fictitious temperature and a partition function. By investigating the high-temperature expansion, we obtain the first three moments of the distribution. We find that the problem exhibits frustration.

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(Some figures in this article are in colour only in the electronic version)

1. Introduction

There is a profound diversity between quantum mechanical and classical correlations. Schrödinger [1, 2] coined the term 'entanglement' to describe the peculiar connection that can exist between quantum systems, that was first perceived by Einstein, Podolsky and Rosen [3] and has no analog in classical physics. Entanglement is a resource in quantum information science [4–6] and is at the origin of many unique quantum phenomena and applications, such as superdense coding [7], teleportation [8] and quantum cryptographic schemes [9–12].

Much progress has been made in developing a quantitative theory of entanglement [5, 6]. The bipartite entanglement between simple systems can be unambiguously defined in terms of the von Neumann entropy or the entanglement of formation [4, 13, 14]. On the other hand, an exhaustive characterization of multipartite entanglement is more elusive [5, 6] and...
different definitions \[15–19\] often do not agree with each other, essentially because they tend to capture different aspects of the phenomenon. More to this, a complete evaluation of global entanglement measures bears serious computational difficulties, because states endowed with large entanglement typically involve exponentially many coefficients.

We proposed in [20] that multipartite entanglement shares many characteristic traits of complex systems and can therefore be analyzed in terms of the probability density function of an entanglement measure (say purity) of a subsystem over all (balanced) bipartitions of the total system [21]. A state has a large multipartite entanglement if its average bipartite entanglement is large. In addition, if the entanglement distribution has a small standard deviation, bipartite entanglement is essentially independent of the bipartition and can be considered as being fairly ‘shared’ among the elementary constituents (qubits) of the system. Clearly, average and standard deviation are but the first two moments of a distribution function. A full characterization of the multipartite entanglement of a quantum state must therefore take into account higher moments and/or the whole distribution function, in particular if the latter is not bell shaped or is endowed with unusual and/or irregular features.

The idea that complicated phenomena cannot be summarized in a single (or a few) number(s), but rather require a large number of measures (or even a whole function), is not novel in the context of complex systems [22] and even in the study of quantum entanglement [23]. In this paper we shall pursue this idea even further and shall study the bipartite and multipartite entanglement of a system of qubits by making full use of the tools and techniques of classical statistical mechanics: we shall explore the features of a partition function, expressed in terms of the average purity of a subset of the qubits; this will be viewed as a cost function, that plays the role of the Hamiltonian. Interestingly, this approach brings to light the presence of frustration in the system [24], highlighting the complexity inherent in the phenomenon of multipartite entanglement.

This paper is organized as follows. We introduce notation and define maximally bipartite and maximally multipartite entangled states in section 2. Multipartite entanglement is characterized in terms of the distribution function of bipartite entanglement in section 3. The statistical mechanical approach and the partition function are introduced in section 4. The high temperature expansion and its first three cumulants are computed in section 5. Section 6 contains our conclusions and an outlook.

2. From bipartite to multipartite entanglement

2.1. Bipartite purity

We consider an ensemble \( S = \{1, 2, \ldots, n\} \) of \( n \) qubits in the Hilbert space \( \mathcal{H}_S = (\mathbb{C}^2)^\otimes n \) and focus on pure states
\[
|\psi\rangle = \sum_{k \in \mathbb{Z}_2^n} z_k |k\rangle, \quad z_k \in \mathbb{C}, \quad \sum_{k \in \mathbb{Z}_2^n} |z_k|^2 = 1, \tag{1}
\]
where \( k = (k_i)_{i \in S} \), with \( k_i \in \mathbb{Z}_2 = \{0, 1\} \), and
\[
|k\rangle = \bigotimes_{i \in S} |k_i\rangle_i, \quad |k_i\rangle_i \in \mathbb{C}^2, \quad \langle k_i | k_j \rangle = \delta_{ij}. \tag{2}
\]

For the sake of simplicity, in this paper we will focus on pure states of qubits and will not discuss additional phenomena such as bound entanglement [25, 26]. Consider a bipartition \((A, \bar{A})\) of the system, where \( A \subset S \) is a subset of \( n_A \) qubits and \( \bar{A} = S \setminus A \) its complement, with \( n_A + n_{\bar{A}} = n \). We set \( n_A \leq n_A \) with no loss of generality. The total Hilbert space factorizes into \( \mathcal{H}_S = \mathcal{H}_A \otimes \mathcal{H}_{\bar{A}} \), with \( \mathcal{H}_A = \bigotimes_{i \in A} \mathbb{C}^2_i \), of dimensions \( N_A = 2^{n_A} \) and \( N_{\bar{A}} = 2^{n_{\bar{A}}} \), respectively.
\((N_A N_{\bar{A}} = N)\). As a measure of the bipartite entanglement between the two subsets, we consider the purity of subsystem \(A\):

\[
\pi_A = \text{tr}_A \rho_A^2, \quad \rho_A = \text{tr}_A |\psi\rangle \langle \psi|,
\]

with \(\text{tr}_X\) being the partial trace over \(X = A\) or \(\bar{A}\). We note that \(\pi_A = \pi_{\bar{A}}\) and

\[
1/N_A \leq \pi_A \leq 1.
\]

State (1) can be written according to the bipartition \((A, \bar{A})\) as

\[
|\psi\rangle = \sum_{k \in \mathbb{Z}^n} z_k |k_A\rangle_A \otimes |k_{\bar{A}}\rangle_{\bar{A}},
\]

where \(k_A = (k_i)_{i \in A}\) and \(|l\rangle_A = \bigotimes_{i \in A} |l_i\rangle \in \mathcal{H}_A\). By plugging equation (5) into equation (3), we obtain

\[
\rho_A = \sum_{k,l \in \mathbb{Z}^n} z_k z_l \delta_{k,l} |k_A\rangle \langle l_A|,
\]

and

\[
\pi_A = \sum_{k,k',l,l' \in \mathbb{Z}^n} z_k z_{k'} z_l z_{l'} \delta_{k,l} \delta_{k',l'} \delta_{k',l'} \delta_{k,l},
\]

which is a quartic function of the coefficients of expansion (1). If, for example, the system is partitioned into two blocks of contiguous qubits \((C, \bar{C})\), namely \(C = \{1, 2, \ldots, n_A\}\), then

\[
\pi_C = \sum_{l,l' \in \mathbb{Z}^n} \sum_{m,m' \in \mathbb{Z}^n} z(l,m) z(l',m') z(l,m'),
\]

where \((l, m) = (l_1, \ldots, l_{n_A}, m_1, \ldots, m_{n_A}) \in \mathbb{Z}_2^n\).

### 2.2. Minimal bipartite purity

For a given bipartition it is very easy to saturate the lower bound \(1/N_A\) of (7). For example,

\[
z_k = N_A^{-1/2} \delta_{k,k'},
\]

which represents a \textit{maximally bipartite entangled state}

\[
|\psi\rangle = N_A^{-1/2} \sum_{l \in \mathbb{Z}^n} |l\rangle_A \otimes |l\rangle_{\bar{A}},
\]

yields \(\rho_A = 1/N_A\) and \(\pi_A = 1/N_A\). In fact, the general minimizer is a maximally bipartite entangled state whose Schmidt basis is not the computational basis, namely,

\[
z_k = N_A^{-1/2} \sum_{l \in \mathbb{Z}^n} U^A_{k,l} U_{k',l'},
\]

where \(U^A_{l,l'} = \langle l | U^A | l' \rangle\) with \(U^A\) a local unitary operator in \(\mathcal{H}_A\) that transforms the computational bases into the Schmidt one, that is

\[
|\psi\rangle = N_A^{-1/2} \sum_{l \in \mathbb{Z}^n} U^A |l\rangle_A \otimes U^{\bar{A}} |l\rangle_{\bar{A}}.
\]

For this state, \(\pi_A(|\psi\rangle) = 1/N_A\). The information contained in a maximally bipartite entangled state with \(n_A = n_{\bar{A}}\) is not locally accessible by party \(A\) or \(\bar{A}\), because their partial density matrices are maximally mixed, but rather is totally shared by them.
2.3. Average purity and MMES

Entanglement, in very few words, embodies the impossibility of factorizing a state of the total quantum system in terms of the states of its constituents. Most measures of bipartite entanglement (for pure states) exploit the fact that when a (pure) quantum state is entangled, its constituents do not have (pure) states of their own. This is, for instance, what we did in the previous section. We wish to generalize the above distinctive trait to the case of multipartite entanglement, by requiring that this feature be valid for all bipartitions.

Let $|\psi\rangle \in \mathcal{H}_S$ and consider the average purity [27, 29]

$$\pi^{(n)}_{\text{ME}}(|\psi\rangle) = \mathbb{E} \left[ \pi_A \right] = \left( \frac{n}{n_A} \right)^{-1} \sum_{|A|=n_A} \pi_A,$$

(13)

where $\mathbb{E}$ denotes the expectation value, $|A|$ is the cardinality of $A$ and the sum is over balanced bipartitions $n_A = [n/2]$, where $[x]$ denotes the integer part of $x$. Since we are focusing on balanced bipartitions, and any bipartition can be brought into any other bipartition by applying a permutation of the qubits, the sum over balanced bipartitions in (13) is equivalent to a sum over the permutations of the qubits. The quantity $\pi_{\text{ME}}$ measures the average bipartite entanglement over all possible balanced bipartitions and inherits the bounds (4)

$$\frac{1}{N_A} \leq \pi^{(n)}_{\text{ME}}(|\psi\rangle) \leq 1.$$  

The average purity introduced in equation (13) is related to the average linear entropy $S_L = \frac{N_A}{N_A - 1} (1 - \pi_{\text{ME}})$ [27] and extends ideas put forward in [18, 28].

A maximally multipartite entangled state (MMES) [29] $|\varphi\rangle$ is a minimizer of $\pi_{\text{ME}},$

$$\pi^{(n)}_{\text{ME}}(|\varphi\rangle) = E^{(n)}_0,$$

with $E^{(n)}_0 = \min \pi^{(n)}_{\text{ME}}(|\psi\rangle) | |\psi\rangle \in \mathcal{H}_S, \langle \psi | \psi \rangle = 1.$$

(15)

The meaning of this definition is clear: the density matrix of each subsystem $A \subset S$ of a MESS is as mixed as possible (given the constraint that the total system is in a pure state), so that the information contained in a MESS is as distributed as possible.

2.4. Perfect MMES and the symptoms of frustration

For small values of $n$ one can tackle the minimization problem (15) both analytically and numerically. For $n = 2, 3, 5, 6$ the average purity saturates its minimum in (14): this means that purity is minimal for all balanced bipartitions. In this case, we shall say that the MESS is perfect.

For $n = 2$ (perfect) MESS are Bell states up to local unitary transformations, while for $n = 3$ they are equivalent to the GHZ states [30]. For $n = 4$ one numerically obtains $E^{(4)}_0 = \min \pi^{(4)}_{\text{ME}} = 1/3 > 1/4 = 1/N_A$ [29, 31–33]. For $n = 5$ and 6 one can find several examples of perfect MESS, some of which can be expressed in terms of binary strings of coefficients ($z_k = \pm 1$ in equation (1)).

The case $n = 7$ is still open, our best estimate being $E^{(7)}_0 \approx 0.133877 > 1/8 = 1/N_A.$ Most interestingly, perfect MMES do not exist for $n \geq 8$ [27]. These findings are summarized in table 1. This brings to light an intriguing feature of multipartite entanglement: we observed in section 2.2 that it is always possible to saturate the lower bound in (4)

$$\pi^{(n)}_A = 1/N_A$$

(16)

for a given bipartition $(A, \bar{A})$. However, in order to saturate the lower bound

$$E^{(n)}_0 = 1/N_A$$

(17)
Table 1. Perfect MMES for different number $n$ of qubits.

<table>
<thead>
<tr>
<th>$n$</th>
<th>Perfect MMES</th>
</tr>
</thead>
<tbody>
<tr>
<td>2,3</td>
<td>Exist</td>
</tr>
<tr>
<td>4</td>
<td>Do not exist</td>
</tr>
<tr>
<td>5,6</td>
<td>Exist</td>
</tr>
<tr>
<td>$\geq 8$</td>
<td>Do not exist</td>
</tr>
</tbody>
</table>

In equation (14), it must happen that (16) be valid for any bipartition in average (13). As we have seen, this requirement can be satisfied only for very few ‘special’ values of $n$. For all other values of $n$ this is impossible: different bipartitions ‘compete’ with each other, and the minimum $E_0^{(n)}$ of $\pi_{ME}^{(n)}$ is strictly larger than $1/N_A$. We view this ‘competition’ among different bipartitions as a phenomenon of frustration: it is already present for $n$ as small as 4 [24]. (Interestingly, an analogous phenomenon exists also for ‘Gaussian MMES’, see [34].)

This frustration is the main reason for the difficulties one encounters in minimizing $\pi_{ME}$ in (13). Note that the dimension of $\mathcal{H}_S$ is $N = 2^n$ and the number of partitions scales like $2^N$. We therefore need to define a viable strategy for the characterization of MMES, when $n \geq 8$.

3. Probability distribution of bipartite entanglement

We now introduce the distribution function of purity over all bipartitions, $p(\pi_A)$, that will induce a probability–density–function characterization of multipartite entanglement. For rather regular (i.e. bell-shaped) distributions the first few moments already yield a good characterization: in particular, the average will measure the amount of entanglement of the state when the bipartitions are varied, while the variance will quantify how uniformly is bipartite entanglement distributed among balanced bipartitions.

The calculation of the properties of $\pi_A$ is particularly simple for an important class of states. Consider the set

$$C = \left\{ (z_1, z_2, \ldots, z_N) \in \mathbb{C}^N | \sum_k |z_k|^2 = 1 \right\},$$

(18)
corresponding to normalized vectors in $\mathcal{H}_S$. This set is left invariant under the natural action of the unitary group $\mathcal{U}(\mathcal{H}_S)$. A typical state is obtained by sampling with respect to the action of $\mathcal{U}(\mathcal{H}_S)$ on this set. Typical states have been extensively studied in the literature [35–40] and can be (efficiently) generated by a chaotic dynamics [41, 42].

For large $N$, the $\pi_A$’s have a bell-shaped distribution over the bipartitions with mean and variance [21]

$$\mu^{(n)} = \langle \pi_A \rangle_0 = \frac{N_A + N_{\bar{A}}}{N + 1},$$

(19)

$$\sigma^2 = \langle (\pi_A - \mu)^2 \rangle_0 = \frac{2(N_A^2 - 1)(N_{\bar{A}}^2 - 1)}{(N + 1)^2(N + 2)(N + 3)},$$

(20)

respectively, where the brackets $\langle \cdots \rangle_0$ denote the average with respect to the unitarily invariant measure over pure states

$$d\mu_C(z) = \frac{(N - 1)!}{\pi^N} \delta \left( 1 - \sum_k |z_k|^2 \right) \prod_k dz_k d\bar{z}_k,$$

(21)
induced by the Haar measure over $\mathcal{U}(\mathcal{H}_S)$ through the mapping $|\psi\rangle = \sum z_j |j\rangle = U |\psi_0\rangle$, for a given reference state $|\psi_0\rangle$ [38]. Here, $dz_k \, dq_k = dx_k \, dy_k$, with $x_k = \Re z_k$ and $y_k = \Im z_k$, denotes the Lesbegue measure on $\mathbb{C}$.

Given a state $|\psi\rangle \in \mathcal{H}_S$, the potential of multipartite entanglement has the following expression in terms of its Fourier coefficients $z_i$:

$$\pi_{\text{ME}} = \sum_{k,k',l,l' \in \mathbb{Z}^2} \Delta(k,k';l,l') z_k z_{k'} \bar{z}_l \bar{z}_{l'},$$

(22)

with a coupling function

$$\Delta(k,k';l,l') = \left( \frac{n_A}{2} \right)^{-1} \sum_{|a|=n_A} \frac{1}{2} \left( \delta_{k,i} \delta_{j,l} \delta_{k',i'} \delta_{j',l'} + \delta_{k,i} \delta_{j,l} \delta_{k',i'} \delta_{j',l'} \right),$$

(23)

with $n_A = \lfloor n/2 \rfloor$ (balanced bipartitions). The result follows by plugging expression (7) of $\pi_A$ into equation (13), and by symmetrizing under the exchange $k \leftrightarrow k'$ (or, equivalently, $A \leftrightarrow \bar{A}$). The coupling function $\Delta$ has the following expression (see appendix A for details):

$$\Delta(k,k';l,l') = g((k \oplus l) \vee (k' \oplus l'), (k \oplus l') \vee (k' \oplus l)),$$

(24)

where

$$g(a,b) = \delta_{a\oplus b,0} \hat{g}(|a|, |b|),$$

(25)

with $|a| = \sum_{i \in S} a_i$, $|b| = \sum_{i \in S} b_i$, $a \oplus b = (a_i + b_i \mod 2)_{i \in S}$ the XOR operation, $a \vee b = (a_i + b_i - a_i b_i)_{i \in S}$ the OR operation, $a \wedge b = (a_i b_i)_{i \in S}$ the AND operation and

$$\hat{g}(s,t) = \frac{1}{2} \left( \frac{n}{n_A} \right)^{-1} \left( \frac{n - s - t}{n_A - s} + \frac{n - s - t}{n_A - t} \right).$$

(26)

Using the definitions we notice the following symmetries of the coupling function:

$$\begin{cases}
\Delta(k,k';l,l') = \Delta(k', k; l, l') \\
\Delta(k,k';l,l') = \Delta(l,l';k,k') \\
\Delta(k,k';l,l') = \Delta(k', k; l', l)
\end{cases}$$

(27)

4. Partition function

In order to study the minimization problem, we will reformulate it in terms of classical statistical mechanics: in particular, the minimum $E_0$ of $\pi_{\text{ME}}$ will be recovered in the zero temperature limit of a suitable classical system.

The main quantity we are interested in is the average bipartite entanglement between balanced bipartitions, $\pi_{\text{ME}}$ in equation (13). This quantity will play the role of energy in the statistical mechanical approach. We therefore start by viewing $\pi_{\text{ME}}$ in equation (13) as a cost function (potential of multipartite entanglement) and write

$$H(z) = \pi_{\text{ME}}(|\psi\rangle),$$

(28)

where $z$ are the Fourier coefficients of expansion (1). We consider an ensemble of $|m_j\rangle$ of $M$ vectors (states), where $m_j$ is the number of vectors with purity $H = \epsilon_j$. In the standard ensemble approach to statistical mechanics, one seeks the distribution that maximizes the number of states $\Omega = M! / \prod_j m_j!$ under the constraints that $\sum_j m_j = M$ and $\sum_j m_j \epsilon_j = ME$. For $M \to \infty$, the above optimization problem yields the canonical ensemble and its partition function

$$Z(\beta) = \int d\mu_C(z) \, e^{-\beta H(z)} = c_N \int d\mu_H(U) \exp(-\beta \mathbb{E}[\Tr_A(U |\psi_0\rangle \langle\psi_0|U^{\dagger})^2]),$$

(29)
where the expectation value $E$ was introduced in equation (13) and the Lagrange multiplier $\beta$, which plays the role of an inverse temperature, fixes the average value of purity $E$. In the first integral we have used the measure (21) and taken into account the normalization condition (18). In the last (base-independent) expression $\mu_H$ denotes the Haar measure over $U(H)$, $|\psi_0\rangle$ is any given vector and the (unimportant) constant $c_N$ is proportional to the ratio $\mu_C(C^N)/\mu_H(U(H))$ between the area of the $(N-1)$-dimensional sphere (18) and the volume of the unitary group. In conclusion, the potential of multipartite entanglement can now be considered as the Hamiltonian of a classical statistical mechanical system.

4.1. Comments

In order to clarify the rationale behind our analysis, a few comments are necessary.

(1) Although our interest is focused on the microcanonical features of the system, namely on ‘isoentangled’ manifolds [43], we find it convenient to define a canonical ensemble and a temperature. This makes the analysis easier to handle and is at the very foundations of statistical mechanics, when one discusses the equivalence in the description of large systems between the microcanonical ensemble (in which energy is fixed) and the canonical ensemble (in which temperature is fixed).

(2) One can view the multipartite system as an ensemble for the collection of all balanced bipartitions. However, what makes the problem intricate and interesting is the fact that there is a nontrivial interaction among different bipartitions, which in general provokes frustration.

(3) From a physical point of view, the measure of typical states is a uniform measure over the whole projective space. This would be consistent with ergodicity. However, our analysis is purely static and we do not consider the time evolution generated by the (purity) Hamiltonian. The relaxation to equilibrium, as well as its ergodic properties, deserves a deeper study and would probably uncover additional features with respect to the equilibrium situation. This aspect will be investigated in the future.

(4) Temperature is a Lagrange multiplier for the optimization parameter. It is the variable that is naturally conjugate to $H$, in exactly the same way as inverse temperature is conjugate to energy: $\beta$ fixes, with an uncertainty that becomes smaller for a larger system, the level of the purity of the subset of vectors under consideration, and thus an isoentangled submanifold. The use of a temperature is a common expedient in problems that can be recast in terms of classical statistical mechanics. One can find examples of this kind in the stochastic approach to optimization processes (for instance simulated annealing) [44, 45].

4.2. Some limits

We start by looking at some interesting limits and give a few preliminary remarks. For $\beta \to 0$, equation (29) clearly yields the distribution of the typical states (21). For $\beta \to +\infty \ (T \to 0^+)$, only those configurations that minimize the Hamiltonian survive, namely the MMES. There is a physically appealing interpretation even for negative temperatures: for $\beta \to -\infty \ (T \to 0^-)$, those configurations are selected that maximize the Hamiltonian, that is separable (factorized and non-entangled) states.

The energy distribution function at arbitrary $\beta$ can be obtained from the partition function

$$Z(\beta) = \int d\mu_C(z) e^{-\beta H(z)} = \int_{E_0}^{1} dE e^{-\beta E} \int d\mu_C(z) \delta(H(z) - E), \quad (30)$$
where \( E \in [E_0, 1] \), \( E_0 \) being the minimum of the spectrum of \( H \) and \( \delta \) the Dirac function. Incidentally, note that equations (14) and (19) yield

\[
\lim_{n \to \infty} E_0^{(n)} \leq \lim_{n \to \infty} \mu^{(n)} = 0, \quad E_0^{(n)} \geq 2^{-[n/2]}, \quad \mu^{(n)} = \frac{2^{[n/2]} + 2^{([n+1]/2)}}{2^n + 1}.
\]

The energy distribution function reads

\[
P\beta(E) = \frac{e^{-\beta E}}{Z(\beta)} \int d\mu C(z) \delta(H(z) - E)
\]

which, for \( \beta = 0 \), simply reads

\[
P_0(E) = \frac{1}{Z(0)} \int d\mu C(z) \delta(H(z) - E).
\]

In figure 1 we show the probability density function \( P_0(E) \) for \( n = 4 \). As emphasized in section 2.4, this is one of those cases in which frustration appears, as for \( n = 4 \) qubits one (numerically) finds \( E_0^{(4)} = \min \pi_{ME}^{(4)} = 1/3 > 1/4 = 1/NA \) [29, 31–33]. We clearly observe the asymmetry of the curve, denoting a positive value of the skewness. This deformation becomes less evident for larger values of \( n \). As we will see, in the thermodynamic limit, \( n \to \infty \), \( P_0(E) \) will become more and more symmetric and will tend to a Gaussian.

Using equations (30)–(33) we obtain the expression of the energy distribution function at arbitrary \( \beta \) in terms of its infinite temperature limit:

\[
P\beta(E) = \frac{e^{-\beta E} P_0(E)}{\int_E^{\infty} dE e^{-\beta E} P_0(E)}.
\]

Note that this equation is valid at fixed \( n \).

By multiplying and dividing the last equation by \( |\beta| e^\beta \) and \( \beta e^{\beta E_0} \), respectively, and remembering that

\[
\frac{1}{\epsilon} e^{-x/\epsilon} \xrightarrow{\epsilon \to 0} \delta(x)
\]

we find

\[
P_{-\infty}(E) = \delta(E - 1), \quad P_{\infty}(E) = \delta(E - E_0).
\]
These limits are the counterparts of those discussed for the partition function and are reflected in the asymptotic behavior of the average energy as a function of $\beta$

$$\langle H \rangle_{\beta} = \frac{1}{Z(\beta)} \int d\mu C(z) H e^{-\beta H} = \int_{E_0}^1 dE \ E P_{\beta}(E) = -\frac{\partial}{\partial \beta} \ln Z(\beta). \quad (37)$$

Indeed,

$$\langle H \rangle_{\beta \to -\infty} = 1, \quad \langle H \rangle_{\beta \to +\infty} = E_0. \quad (38)$$

More generally, the $m$th cumulant of $H$ reads

$$\kappa_m^{(\beta)}[H] = (-)^m \frac{\partial^m}{\partial \beta^m} \ln Z(\beta) = (-)^{m-1} \frac{\partial^{m-1}}{\partial \beta^{m-1}} \langle H \rangle_{\beta}. \quad (39)$$

We find

$$\frac{\partial}{\partial \beta} \langle H \rangle_{\beta} = -\kappa_2^{(2)}[H] = -\langle H^2 \rangle_{\beta} + \langle H \rangle_{\beta}^2 \equiv -\bar{\sigma}^2 \leq 0, \quad (40)$$

which is non-positive. In particular

$$\bar{\sigma}^2 = \bar{\sigma}^2_0 = \kappa_2^{(2)}[H]. \quad (41)$$

The average energy is a non-increasing function of $\beta$ and has at least one inflexion point as a function of $\beta$. Moreover,

$$\kappa_3^{(3)}[H] = \frac{\partial^2}{\partial \beta^2} \langle H \rangle_{\beta} = \frac{1}{2} \frac{\partial}{\partial \beta} \bar{\sigma}^2. \quad (42)$$

From the qualitative behavior of $\kappa_3^{(3)}$ one can obtain information about the width of the distribution. For $\beta \to +\infty$ the curvature of $\langle H \rangle_{\beta}$ is positive and therefore $\bar{\sigma}_\beta^2$ is a decreasing function.

From a qualitative point of view, one expects the behavior sketched in figure 2: for $\beta \to 0^+$ ($T \to +\infty$), the distribution is bell-shaped (typical states); when $\beta \to +\infty$ ($T \to 0^+$) the distribution tends to become more concentrated around $E_0$. The energy distribution (34) at sufficiently high temperatures (how high will be discussed in section 5.4, see equation (134)) can be obtained by observing that from equation (40) $P_{\beta}(E) \sim P_0(E + \beta \bar{\sigma}^2)$.

For larger values of $\beta$ the left tail of the distribution starts ‘feeling’ the wall at $E_0$. The value of $P_0(E_0)$ influences the behavior of $P_\beta(E)$. In general, $P_0(E_0)$ can vanish or not, yielding the behavior sketched in figures 2(a) and (b), respectively. One finds

$$P_\beta(E) \sim \frac{\beta^{r+1}}{r!} (E - E_0)^r e^{-\beta(E - E_0)}, \quad (44)$$

where $r$ is the order of the first nonvanishing derivative of $P_0(E)$ at $E_0$. (Figures 2(a), (b) display the case $r = 0, 1$, respectively.) Note that the only relic of $P_0(E)$ in (44) is $r$ and $P_{\beta \to \infty}(E)$ yields the second equation in (36). Actually if $r = 0$, equation (44) yields a pure exponential converging to

$$P_{\infty}(E) = \delta(E - E_0). \quad (45)$$

If $r \geq 1$ the probability for finite $\beta$ has an initial polynomial increase but still converges to a Dirac $\delta$ in $E_0$, corresponding to MMES. The analysis for $\beta \to -\infty$ is analogous (we expand $P_\beta(E)$ around $E = 1$, which is the maximum of $H$); it yields the first equation in (36). In this limit we obtain the separable states.
Figure 2. Qualitative sketch of equation (34), at fixed $N_A$, in arbitrary units. The energy density function is distributed around $\mu$ with standard deviation $\bar{\sigma}$ at $\beta = 0$ (first bell-shaped curve at the right in both panels) and moves toward $E_0$ when $\beta$ increases. In each panel, from right to left, $\beta$ changes in constant steps. The probability density rigidly shifts with $\beta$, for $\beta \lesssim N^{-7/2} - \log 3$: see equation (133) and following discussion. Note that both $E_0$ and $\mu$ are $O(N^{-1/2})$. In $(a)$ $P_0(E_0) \neq 0$; in $(b)$ $P_0(E_0) = 0$.

5. High temperature expansion

This section is devoted to the study of the cumulants of $P_0(E)$. This will enable us to look at some properties of the high temperature expansion of the distribution function of the potential of multipartite entanglement. We recall that for $\beta \to 0$ one obtains the typical states.

The high temperature expansion originates from the Taylor series

$$
\kappa^{(m)}_\beta[H] = (-\beta)^m \sum_{j=0}^{\infty} \frac{\bar{\sigma}^j}{j!} \left( \frac{\partial^{m+j}}{\partial \beta^{m+j}} \ln Z(\beta) \right) \bigg|_{\beta=0} = \sum_{j=0}^{\infty} \frac{(-\beta)^j}{j!} \kappa^{(m+j)}_{\beta=0}[H].
$$

(46)

The average energy reads

$$
\langle H \rangle_\beta = \sum_{m=1}^{\infty} \frac{(-\beta)^{m-1}}{(m-1)!} \kappa^{(m)}_{\beta=0}[H]
\sim \langle H \rangle_0 - \beta \langle (H - \langle H \rangle_0)^2 \rangle_0 + \frac{\beta^2}{2} \langle (H - \langle H \rangle_0)^3 \rangle_0.
$$

(47)

while the free energy takes the form

$$
F(\beta) = \frac{1}{\beta} \ln Z(\beta)
\sim \frac{\ln Z(0)}{\beta} - \langle H \rangle_0 + \frac{\beta}{2} \langle (H - \langle H \rangle_0)^2 \rangle_0 - \frac{\beta^2}{6} \langle (H - \langle H \rangle_0)^3 \rangle_0.
$$

(48)

In the following three subsections we will evaluate the first three cumulants of the distribution for $\beta = 0$ in order to characterize the high temperature expansion of the energy distribution function.
5.1. First cumulant

The joint probability density of \( z = (z_k) \in \mathbb{C}^N \) associated with the measure of typical states (21) is

\[
p_N(z_1, z_2, \ldots, z_N) = \frac{(N - 1)!}{\pi^N} \delta \left( 1 - \sum_{1 \leq k \leq N} |z_k|^2 \right).
\] (49)

By integrating out \( N-M \) variables, one obtains

\[
p_N(z_1, z_2, \ldots, z_M) = \frac{(N - 1)!}{(N - M - 1)! \pi^M} \left( 1 - \sum_{1 \leq k \leq M} |z_k|^2 \right)^{N-M-1},
\] (50)

for \( 1 \leq M < N \). In particular, the probability density of an arbitrary element of \( z \) is

\[
p_N(z_1) = \frac{N - 1}{\pi} (1 - |z_1|^2)^{N-2}.
\] (51)

Since \( \langle e^{i \alpha \arg z_j} \rangle_0 = 0 \), the only nonvanishing averages of the type

\[
\langle \prod_{k \in X} z_k \prod_{l \in Y} \bar{z}_l \rangle_0,
\] (52)

with \( X, Y \subset \mathbb{Z}_+^N \), are obtained when the variables \( \{z_k\} \) and \( \{\bar{z}_l\} \) are equal pair by pair, that is when the sets of indices are equal, \( X = Y \). The nonvanishing correlation functions are given by

\[
\langle \prod_{j=1}^k |z_{q_j}|^{2m_j} \rangle_0 = \int \prod_{j=1}^k |z_j|^{2m_j} p_N(z_1, \ldots, z_N) \prod_j dz_j d\bar{z}_j
\]

\[
= \frac{(N - 1)! \prod_{j=1}^k m_j!}{(N - 1 + \sum_{j=1}^k m_j)!}.
\] (53)

A simple proof goes as follows. Extend the product to all \( N \) variables by letting some \( m_j \) vanish and consider the quantity, with \( \alpha_i > 0 \),

\[
\langle \prod_{j=1}^N |z_j|^{2m_j} e^{-\sum_i \alpha_i |z_i|^2} \rangle_0 = \int_\mathbb{R}^N \prod_j (dx_j x_j^{m_j}) e^{-\sum_i \alpha_i x_i} (N - 1)! \int_\mathbb{R} \frac{d\omega}{2\pi} e^{-i\omega (1 - \sum_i x_i)}
\]

\[
= (N - 1)! \int_\mathbb{R} \frac{d\omega}{2\pi} e^{-i\omega} \prod_k J_{m_k} (\alpha_k - i\omega),
\] (54)

where

\[
J_m(z) = \int_\mathbb{R}^+ \frac{\chi^m e^{-zx}}{x} \, dx, \quad (\text{Re} \ z > 0).
\] (55)

Now, we have \( J_1(z) = 1/z \) and \( J_m(z) = (-1)^m \frac{d^m J_0}{dz^m} = m! / z^{m+1} \), and thus

\[
\langle \prod_{j} |z_j|^{2m_j} e^{-\sum_i \alpha_i |z_i|^2} \rangle_0 = (N - 1)! \int_\mathbb{R} \frac{d\omega}{2\pi} e^{-i\omega} \prod_j \frac{(m_j)!}{(\alpha_j - i\omega)^{m_j+1}}.
\] (56)

By setting \( \alpha_j = \alpha \) for all \( j \) we obtain

\[
\langle \prod_{j} |z_j|^{2m_j} e^{-\alpha \sum_i |z_i|^2} \rangle_0 = e^{-\alpha} \langle \prod_{j} |z_j|^{2m_j} \rangle_0
\]

\[
= (N - 1)! \prod_j (m_j)! \int_\mathbb{R} \frac{d\omega}{2\pi} e^{-i\omega} \frac{1}{(\alpha - i\omega)^{N+\sum_j m_j}},
\] (57)
which when \( m_j = 0 \) for all \( j \) reads
\[
e^{-\alpha} = (N - 1)! \frac{1}{\pi} \int_{-\infty}^{\infty} e^{-\alpha x^2} dx.
\]

Therefore,
\[
e^{-\alpha} \left( \prod_{j} |z_j|^{2m_j} \right)_0 = \frac{(N - 1)! \prod_j (m_j)!}{(N + \sum_j m_j - 1)} e^{-\alpha}
\]
and (53) follows.

The average energy at \( \beta = 0 \) can be easily evaluated and is equal to the average purity \( \mu \) defined in (19):
\[
\langle H \rangle_0 = \langle \mathbb{E}[\pi_A] \rangle_0 = \mathbb{E} \{ \langle \pi_A \rangle_0 \} = \langle \pi_A \rangle_0 = \mu^{(n)}.
\]

Let us check the above result by direct computation, through (22). We obtain
\[
\langle H \rangle_0 = \sum_{k, j \in \mathbb{Z}_2^n} \Delta(k_1, k_2; k, k) \langle z_{k_1, k_2, l_1, l_2} \rangle_0
\]
and thus
\[
\langle H \rangle_0 = 2 \langle |z_1|^2 |z_2|^2 \rangle_0 \sum_{k, k_2 \in \mathbb{Z}_2^n} \Delta(k_2; k_1, k_2)
\]
and
\[
\langle z_{k_1, k_2, l_1, l_2} \rangle_0 = \langle |z_1|^2 |z_2|^2 \rangle_0 \left( \delta_{k_1, k_2} \delta_{l_1, l_2} + \delta_{k_1, l_2} \delta_{k_2, l_1} \right)
\]
and hence
\[
\langle H \rangle_0 = 2 \langle |z_1|^2 |z_2|^2 \rangle_0 \sum_{k, k_2 \in \mathbb{Z}_2^n} \Delta(k, k; k, k).
\]

where the symmetry (27) was used.

By using (24) and by setting \( k = k_1 \oplus k_2 \), we obtain
\[
\sum_{k_1, k_2} \Delta(k_1, k_2; k_1, k_2) = \sum_{k_1, k_2} g(0, k_1 \oplus k_2) = \sum_{k_1, k_2} g(k, 0) = N \sum_k g(k, 0).
\]

Since \( \delta_{k, 0} = 1 \) and \( \sum_{0 \leq s \leq n} \delta_{k, s} = 1 \), by using (25) we can write
\[
\sum_k g(k, 0) = \sum_k \hat{g}(k, 0) = \sum_k \hat{g}(k, 0) \sum_{0 \leq s \leq n} \delta_{k, s} = \sum_k \hat{g}(s, 0) \sum_k \delta_{k, |s|}.
\]
The number of strings containing \( s \) ones is
\[
\sum_{k \in \mathbb{Z}_2^n} \delta_{k, |s|} = \binom{n}{s},
\]
and from (26)
\[
\binom{n}{s} \hat{g}(s, 0) = \frac{1}{2} \binom{n}{s} \binom{n_A}{n_A}^{-1} \left[ \binom{n - s}{n_A} + \binom{n - s}{n_A} \right] = \frac{1}{2} \left[ \binom{n_A}{s} + \binom{n_A}{n_A} \right],
\]
whence
\[
\sum_k g(k, 0) = \frac{1}{2} \sum_s \left[ \binom{n_A}{s} + \binom{n_A}{n_A} \right] = \frac{1}{2} (N_A + N_A),
\]
The second cumulant is defined as

$$\sum_{k_1, k_2} \Delta(k_1, k_2; k_1, k_2) = N \sum_{k} g(k, 0) = \frac{N(N_A + N_A)}{2}. \quad (69)$$

On the other hand,

$$\sum_{k} \Delta(k, k; k) = \sum_{k} g(0, 0) = \sum_{k} g(0, 0) = N,$$  \quad (70)

because $$g(0, 0) = 1$$. Summing up, we obtain

$$\langle H \rangle_0 = N(N_A + N_A)\langle |z_1|^2 |z_2|^2 \rangle_0 + N\langle |z_1|^4 \rangle_0 - 2\langle |z_1|^2 |z_2|^2 \rangle_0, \quad (71)$$

and since (see equation (53))

$$\langle |z_1|^2 |z_2|^2 \rangle_0 = \frac{1}{2} \langle |z_1|^4 \rangle_0 = \frac{1}{N(N + 1)}, \quad (72)$$

we obtain

$$\langle H \rangle_0 = \frac{N_A + N_A}{N + 1}, \quad (73)$$

which equals the value (19) of the average purity $$\mu^{(m)}$$. In the thermodynamic limit, $$N \to \infty$$, with $$N_A = N_A = \sqrt{N}$$

$$\langle H \rangle_0 \sim \frac{2}{\sqrt{N}}. \quad (74)$$

5.2. Second cumulant

The second cumulant is defined as

$$\tilde{\sigma}^2 = \kappa_0^{(2)} [H] = \langle H^2 \rangle_0 - \langle H \rangle_0^2. \quad (75)$$

In order to evaluate this quantity we will use a diagrammatic technique based on the definition of the coupling function $$\Delta$$ and its properties (equation (27)). We start considering

$$\langle H^2 \rangle_0 = \sum_{k, l \in \mathbb{Z}_n^+} \Delta(k_1, k_2; l_1, l_2) \Delta(k_3, k_4; l_3, l_4)\langle z_6, z_6, z_6, z_6, z_6, z_6 \rangle_0. \quad (76)$$

We must have $$\{k_i\} = \{l_j\}$$ as sets, that is

$$l_i = k_{p(i)}, \quad p \in \mathcal{P}_4$$  \quad (77)

with $$1 \leq i \leq 4$$, where $$\mathcal{P}_4$$ is the permutation group of $$\{1, 2, 3, 4\}$$. Therefore,

$$\langle H^2 \rangle_0 = \sum_{k \in \mathbb{Z}_n^+} \sum_{p \in \mathcal{P}_4} \Delta(k_1, k_2; k_{p(1)}, k_{p(2)}) \Delta(k_3, k_4; k_{p(3)}, k_{p(4)})\langle |z_6|^2 |z_6|^2 |z_6|^2 |z_6|^2 \rangle_0. \quad (78)$$

where

$$\langle |z_1|^{2m} |z_2|^{2n} |z_3|^{2s} |z_4|^{2t} \rangle_0 = \frac{1}{m!n!s!t!} \langle |z_1|^{2m} |z_2|^{2n} |z_3|^{2s} |z_4|^{2t} \rangle_0. \quad (79)$$

The above normalization takes into account the fact that if $$k_i = k_j$$ for some $$i \neq j$$ the sum over the permutation group $$\mathcal{P}_4$$ overcounts the number of different terms. For example, if $$k_1 = k_2$$ and is different from the others, we get $$m = 2, n = 0, s = t = 1$$, and there is a factor 1/2!, while, if $$k_1 = k_2 = k_3 \neq k_4$$, we get $$m = 3, n = s = 0, t = 1$$, and there is a factor 1/3!.

Since $$m + n + s + t = 4$$, from equation (53) we observe that

$$\langle |z_1|^{2m} |z_2|^{2n} |z_3|^{2s} |z_4|^{2t} \rangle_0 = \frac{(N - 1)!}{(N + 3)!} = \frac{1}{N(N + 1)(N + 2)(N + 3)}. \quad (80)$$
is independent of $k \in \mathbb{Z}_2^{2n}$. Therefore,
\[
\langle H^2 \rangle_0 = \langle |z_1|^2 |z_2|^2 |z_3|^2 |z_4|^2 \rangle_0 \sum_{p \in \mathcal{P}} [p(1) p(2), p(3) p(4)],
\] (81)
with the notation
\[
[p(1) p(2), p(3) p(4)] = \sum_{k \in \mathbb{Z}_2^{2n}} \Delta(k_1, k_2; k_{p(1)}, k_{p(2)}) \Delta(k_3, k_4; k_{p(3)}, k_{p(4)}).
\] (82)

Note that by the symmetries (27) of $\Delta$, we can swap $p(1) \leftrightarrow p(2)$ or $p(3) \leftrightarrow p(4)$, as well as $1 \leftrightarrow 2$ or $3 \leftrightarrow 4$, so that
\[
[w, x, y, z] = [x, w, y, z] = [w, x, z, y],
\] if $w, y \in \{1, 2\}$, or $w, y \in \{3, 4\}$. (83)

Using these symmetries we can give a graphical representation of the quantity in equation (82). Let us consider figure 3(a). Each vertex represents a pair $(k_i, k_j)$ in the summation. The edges between vertices and the loops on the same vertex fix the value of $p(i)$. For instance, a double loop on $(k_1, k_2)$ and $(k_3, k_4)$ (see figure 3(b)) yields
\[
[2, 1, 3, 4] = \sum_k \Delta(k_1, k_2; k_1, k_1) \Delta(k_3, k_4; k_3, k_4).
\] (84)

Each vertex has order 4 with two incoming and two outgoing edges. Each graph is oriented. However, for simplicity, in the graphs of figure 3 we have not indicated the orientations since in this case, as it is easy to see, they do not yield different contributions. As we shall see, this will not be the case for higher cumulants, where graphs with different orientations represent nonequivalent contributions.

We start considering graphs with no links between the left and right pairs; see figure 3(b). The sum of this class of graphs is
\[
[0 - \text{link}] = [1, 2, 3, 4] + [1, 2, 4, 3] + [2, 1, 3, 4] + [2, 1, 4, 3] = 4 [1, 2, 3, 4].
\] (85)

We have
\[
[1, 2, 3, 4] = \sum_k \Delta(k_1, k_2; k_1, k_2) \Delta(k_3, k_4; k_3, k_4) = \sum_k g(0, k_1 \oplus k_2) g(0, k_3 \oplus k_4).
\] (86)
By setting \( l_2 = k_1 \oplus k_2 \) and \( l_3 = k_1 \oplus k_3 \), we obtain

\[
[1, 2, 3, 4] = \sum_{l_2, l_3} g(0, l_2) g(0, l_3) = N^2 \sum_{l_2, l_3} g(0, l_2) g(0, l_3) = N^2 \left( \sum_k g(0, k) \right)^2.
\]

(87)

Therefore, by using (68), we obtain

\[
[0 \text{- link}] = 4 \left[1, 2, 3, 4\right] = N^2 (N_A + N_A)^2.
\]

(88)

Let us now consider the graphs with two links between left and right pairs in figure 3(c). The sum of this class of graphs is

\[
[2 \text{- link}] = [1, 3, 2, 4] + [1, 3, 4, 2] + [3, 1, 2, 4] + [3, 1, 4, 2]
+ [2, 3, 1, 4] + [2, 3, 4, 1] + [3, 2, 1, 4] + [3, 2, 4, 1]
+ [1, 4, 2, 3] + [1, 4, 3, 2] + [4, 1, 2, 3] + [4, 1, 3, 2]
+ [2, 4, 1, 3] + [2, 4, 3, 1] + [4, 2, 1, 3] + [4, 2, 3, 1]
= 16 [1, 3, 2, 4].
\]

(89)

One obtains

\[
[1, 3, 2, 4] = \sum_k \Delta(k_1, k_2; k_1, k_3) \Delta(k_3, k_4; k_2, k_4)
= \sum_k g(k_2 \oplus k_3, (k_1 \oplus k_3) \lor (k_1 \oplus k_2)) g(k_2 \oplus k_3, (k_2 \oplus k_4) \lor (k_3 \oplus k_4)).
\]

(90)

By setting \( l_1 = k_1 \oplus k_3, l_2 = k_2 \oplus k_3 \) and \( l_4 = k_3 \oplus k_4 \), we obtain

\[
[1, 3, 2, 4] = \sum_{k_2, l_2, l_4} g(l_2, l_1 \lor (l_1 \oplus l_2)) g(l_2, (l_2 \oplus l_4) \lor l_4)
= N \sum_{l_1, l_2, l_4} g(l_2, l_1 \lor l_2) g(l_2, l_2 \lor l_4).
\]

(91)

where we have used the (easy to prove) useful relation

\[
l_1 \lor (l_1 \oplus l_2) = l_1 \lor l_2.
\]

(92)

We get

\[
l_2 \land (l_1 \lor l_2) = (l_1 \land l_2) \lor l_2,
\]

(93)

so that the constraint of the function \( g, l_2 \land (l_1 \lor l_2) = 0 \), implies that \( l_2 = 0 \). Therefore, by using (68), we obtain

\[
[2 \text{- link}] = 16 [1, 3, 2, 4] = 16 N \left( \sum_k g(k, 0) \right)^2 = 4 N (N_A + N_A)^2.
\]

(94)

The contribution of the graphs with four links between left and right pairs (see figure 3(d)) has the form

\[
[4 \text{- link}] = [3, 4, 1, 2] + [4, 3, 2, 1] + [4, 3, 1, 2] + [3, 4, 2, 1] = 4[3, 4, 1, 2].
\]

(95)

We have

\[
[3, 4, 1, 2] = \sum_k \Delta(k_1, k_2; k_3, k_4) \Delta(k_3, k_4; k_1, k_2) = \sum_k \Delta(k_1, k_2; k_3, k_4)^2
= \sum_k g((k_1 \oplus k_3) \lor (k_2 \oplus k_4), (k_1 \oplus k_4) \lor (k_2 \oplus k_3))^2.
\]

(96)
By setting $l_1 = k_1 \oplus k_3$, $l_4 = k_1 \oplus k_4$ and $l_2 = k_1 \oplus k_2 \oplus k_3 \oplus k_4$, we obtain
\[ [3, 4, 1, 2] = \sum_{l_1, l_2, l_4} g(l_1 \lor l_2, l_4 \lor l_2)^2 = N \sum_{l_1, l_4} g(l_1, l_4)^2, \] (97)
where we used relation (92) and the constraint $l_2 = 0$ implied by $(l_1 \lor l_2) \land (l_4 \lor l_2) = 0$. Therefore, we obtain
\[ [4 - \text{link}] = 4[3, 4, 1, 2] = N f_2(N), \] (98)
where
\[ f_2(N) = 4 \sum_{k,l \in \mathbb{Z}} g(k, l)^2. \] (99)
Note that if
\[ d = |A \cap \bar{B}| = |B \cap \bar{A}| \in [0, n_A]. \] (100)
is the distance between bipartitions $(A, \bar{A})$ and $(B, \bar{B})$, defined as the number of qubits belonging to $A$ and not to $B$, then
\[ f_2(N) = 2 \binom{n}{n_A}^{-1} \sum_{0 \leq d \leq n_A} \binom{n_A}{d} \binom{n_A}{d} 2^{n/2} [4^{n/4 - d} + 4^{-(n/4 - d)}]. \] (101)
See appendix B. Summing up, we obtain
\[
\sum_{p \in P_3} [p(1) \ p(2), \ p(3) \ p(4)] = [0 - \text{link}] + [2 - \text{link}] + [4 - \text{link}]
= 4[1, 2, 3, 4] + 16[1, 3, 2, 4] + 4[3, 4, 1, 2]
= N(N + 4)(N_A + N_{\bar{A}})^2 + N f_2(N).
\] (102)
Therefore,
\[ \langle H^2 \rangle_0 = \frac{f_2(N) + (N + 4)(N_A + N_{\bar{A}})^2}{(N + 1)(N + 2)(N + 3)}, \] (103)
and
\[ \bar{\sigma}^2 = \frac{(N + 1) f_2(N) - 2(N_A + N_{\bar{A}})^2}{(N + 1)^2(N + 2)(N + 3)}. \] (104)
We have checked that the above analytic expression of the second cumulant, with $f_2$ given by equation (99), agrees very well (within 1% up to $n = 8$) with the numerical estimates based on the probability density function (obtained by sampling $5 \times 10^4$ typical states for each value of $n$).

Finally, one proves that (see appendix B), in the limit $N \to \infty$,
\[ f_2(N) \sim 3\sqrt{2}N^\alpha, \] (105)
with
\[ \alpha = \log_2 3 - 1 \simeq 0.5850. \] (106)
Therefore, for $N \to \infty$ we have
\[ \bar{\sigma}^2 \sim \frac{f_2(N)}{N^3} = \frac{3\sqrt{2}}{N^{4 - \log_2 3}} = O(N^{-2.415}). \] (107)
Incidentally, note that
\[ \bar{\sigma}^2 \equiv \binom{n}{n_A}^{-2} \sum_{A,B} [\langle \pi_A \pi_B \rangle_0 - \langle \pi_A \rangle_0 \langle \pi_B \rangle_0]. \] (108)
so that if the bipartitions were independent, we would have obtained

$$\bar{\sigma}_2^2 = \left( \frac{n}{n_A} \right)^2 \sum_{A} \left[ \langle \pi_A^2 \rangle_0 - \langle \pi_A \rangle_0^2 \right] = \left( \frac{n}{n_A} \right)^2 \sigma^2 \sim N^{-3}. \quad (109)$$

Thus, the result in equation (107) detects an interference among different bipartitions. We stress that the asymptotic estimate is very accurate even for small values of \( n \). In figure 4 we plot the difference between the analytic value of second cumulant, obtained using equation (104) with \( f_2 \) given by equation (99), and its asymptotic limit, obtained by substituting (105) into equation (104). We notice an oscillatory behavior: the asymptotic expression systematically overestimates (underestimates) the second cumulant for even (odd) values of \( n \). On the other hand, this approximation is very good even for small values of \( n \).

5.3. Third cumulant

The third cumulant is defined as

$$\kappa_3^{(3)}[H] = \langle (H - \langle H \rangle_0)^3 \rangle_0 = \langle H^3 \rangle_0 - 3\langle H^2 \rangle_0\langle H \rangle_0 + 2\langle H^3 \rangle_0. \quad (110)$$

In analogy with the evaluation of the second cumulant, we have

$$\langle H^3 \rangle_0 = \langle |z_1|^2|z_2|^2|z_3|^2|z_4|^2|z_5|^2|z_6|^2 \rangle_0 \sum_{p \in \mathcal{P}_6} [p(1)\ p(2)\ p(3)\ p(4)\ p(5)\ p(6)], \quad (111)$$

with

$$[p(1)\ p(2)\ p(3)\ p(4)\ p(5)\ p(6)] = \sum_{k \in \mathbb{Z}_6^+} \Delta(k_1, k_2; k_{p(1)}, k_{p(2)}) \times \Delta(k_3, k_4; k_{p(3)}, k_{p(4)}) \Delta(k_5, k_6; k_{p(5)}, k_{p(6)}). \quad (112)$$

and \( \mathcal{P}_6 \) the permutation group of \( \{1, 2, 3, 4, 5, 6\} \). From equation (53) we easily obtain

$$\langle |z_1|^2|z_2|^2|z_3|^2|z_4|^2|z_5|^2|z_6|^2 \rangle_0 = \frac{1}{N(N+1)(N+2)(N+3)(N+4)(N+5)}. \quad (113)$$

We start by considering connected graphs with three ears. A representative of this equivalence class is depicted in figure 5(a). We have
Figure 5. Non-oriented connected graphs used for the evaluation of the third cumulant. (a) Three ears. (b) Two ears. (c) One ear.

\[1 6, 3, 2, 5, 4\] = \(\sum_{k} \Delta(k_1, k_2; k_1, k_6) \Delta(k_3, k_4; k_3, k_2) \Delta(k_5, k_6; k_5, k_4)\)

\[= \sum_{k} g(k_2 \oplus k_6, (k_1 \oplus k_6)) \cdot g(k_2 \oplus k_4, (k_2 \oplus k_3)) \cdot (k_3 \oplus k_4)) \times g(k_4 \oplus k_6, (k_4 \oplus k_5) \vee (k_5 \oplus k_6))\]

\[= \sum_{k_1, k_2, k_3, k_5, k_6} g(0, k_1 \oplus k_2) \cdot g(0, k_2 \oplus k_3) \cdot g(0, k_2 \oplus k_5)\]

\[= N \left( \sum_{k} g(0, k) \right)^3 = N \left( \frac{N_A + N_A}{8} \right)^3, \tag{114} \]

where the constraint in the definition of the function \(g\) has implied \(k_2 = k_4 = k_6\) and we used equation (68). The degeneracy of this class of graphs is 128.

We now consider the class of connected graphs with two ears represented in figure 5(b). We obtain

\[1 3, 2, 5, 4, 6\] = \(\sum_{k} \Delta(k_1, k_2; k_1, k_3) \Delta(k_3, k_4; k_2, k_5) \Delta(k_5, k_6; k_4, k_6)\)

\[= \sum_{k_1, k_2, k_4, k_6} g(k_1 \oplus k_2, 0) \cdot g(k_2 \oplus k_4, 0) \cdot g(k_4 \oplus k_6, 0)\]

\[= N \left( \sum_{k} g(k, 0) \right)^3 = N \left( \frac{N_A + N_A}{8} \right)^3, \tag{115} \]

where we have imposed \(k_2 = k_3\) and \(k_4 = k_5\). The degeneracy of the class is 192.

The final class of non-oriented connected graphs is represented in figure 5(c). Its explicit calculation yields
where we have used the constraints $k_2 = k_6$ and the function $f_2(N)$ defined in equation (99). The degeneracy of this graph is 192.

In order to take into account the contribution of connected graphs with no ears, it is necessary to consider two different classes of oriented graphs whose representatives are shown in figures 6(a) and (b), respectively. For the first class (figure 6(a), nonvanishing ‘current’) we have

$$[1 6, 2 5, 3 4] = \sum_k \Delta(k_1, k_2; k_1, k_6) \Delta(k_3, k_4; k_2, k_3) \Delta(k_5, k_6; k_3, k_4)$$

$$= \sum_k g(k_2 \oplus k_6, (k_1 \oplus k_6) \lor (k_1 \oplus k_2))$$

$$\times g((k_2 \oplus k_3) \lor (k_4 \oplus k_3), (k_3 \oplus k_5) \lor (k_2 \oplus k_4))$$

$$\times g((k_3 \oplus k_5) \lor (k_4 \oplus k_6), (k_4 \oplus k_5) \lor (k_3 \oplus k_6))$$

$$= \sum_{k_1, \ldots, k_5} g(0, k_1 \oplus k_2) g((k_2 \oplus k_3) \lor (k_4 \oplus k_5), (k_3 \oplus k_5) \lor (k_2 \oplus k_4))^2$$

$$= \sum_{k_1, \ldots, k_5} g(0, k_1 \oplus k_2) g(k_2 \oplus k_3, k_2 \oplus k_4)^2$$

$$= N \sum_k g(0, k) \sum_{l_1, l_2} g(l_1, l_2)^2 = N \frac{N_A + N_\bar{A}}{2} \sum_{l_1, l_2} g(l_1, l_2)^2$$

$$= N \frac{N_A + N_\bar{A}}{8} f_2(N),$$

(116)

where we have used the constraint $k_2 = k_6$ and the function $f_2(N)$ defined in equation (99). The degeneracy of this graph is 192.

For the second class (figure 6(b), vanishing ‘current’). We obtain

$$[3 6, 5 2, 1 4] = \sum_k \Delta(k_1, k_2; k_3, k_6) \Delta(k_3, k_4; k_5, k_2) \Delta(k_5, k_6; k_1, k_4)$$

$$= N \sum_{k_1, k_2, k_3} g(k_1, k_2) g(k_3, k_2) g(k_1 \oplus k_3, k_2) = N f_3^{(0)}(N)$$

(119)
with
\[ f_3^{(0)}(N) = \sum_{k_1, k_2, k_3} g(k_1, k_2) g(k_3, k_2) g(k_1 \oplus k_3, k_2). \tag{120} \]

In this case, the degeneracy is 64.

The contribution of disconnected graphs (figure 7) can be computed by considering the results obtained for the first and second cumulants. For the class of graphs represented in figure 7(a), we have
\[ [1, 2, 3, 4, 5 6] = N^3 \frac{(N_A + N_A)^3}{8}, \tag{121} \]
with degeneracy 8. In the case of the graph shown in figure 7(b), the result is
\[ [1, 3, 2, 4, 5 6] = N \left(\frac{N_A + N_A}{4}\right)^2 N \left(\frac{N_A + N_A}{2}\right) = N^2 \frac{(N_A + N_A)^3}{8}, \tag{122} \]
with degeneracy 96. Finally from the disconnected graphs with two loops (figure 7(c)), we obtain
\[ [3, 4, 1, 2, 5 6] = N^2 \frac{(N_A + N_A)}{8} f_2(N), \tag{123} \]
with degeneracy 24. In conclusion, we find

$$\langle H^3 \rangle_0 = \frac{1}{(N+1)(N+2)(N+3)(N+4)(N+5)} \left( 16 f_3^{(1)}(N) + 64 f_3^{(0)}(N) + 3(N+8) \right) \times (N_A + N_A) f_2(N) + (N_A + N_A)^3 (N^2 + 12N + 40)$$

(124)

and, therefore,

$$\kappa_0^{(3)}[H] = \frac{1}{(N+1)^3(N+2)(N+3)(N+4)(N+5)} \left( 16(N+1)^2 f_3^{(1)}(N) + 64(N+1)^3 f_3^{(0)}(N) - 36(N+1)(N_A + N_A) f_2(N) - 8(N_A + N_A)^3 (N - 5) \right).$$

(125)

We have checked that the above analytic expression of the third cumulant, with $f_2$, $f_3^{(0)}$ and $f_3^{(1)}$ given by (99), (120) and (118), respectively, agrees very well (less than 1% for $n = 1 \div 7$ and a few % for $n = 8$) with the numerical estimates based on the probability density function (obtained by sampling $5 \times 10^4$ typical states for each value of $n$).

Finally, in the limit $N \to \infty$, one can prove that (see appendix C)

$$f_3^{(0)}(N) \sim c N^{5-\gamma},$$

(126)

with $c \simeq 1.05385$ and $\gamma \simeq 4.1583$ given by (C.30) and (C.31), and that (see appendix D)

$$f_3^{(1)}(N) \sim N^\alpha,$$

(127)

with $\alpha \simeq 0.5850$ given by (106). Therefore, by recalling the asymptotic expression for $f_2(N)$ (105), we find that the graph shown in figure 6(b) dominates over that in figure 6(a) and

$$\kappa_0^{(3)}[H] \sim 64 c N^{-\gamma} \simeq 67.443 N^{-4.1583}.$$  

(128)

In figure 8, we plot the difference between the analytic value of third cumulant and its asymptotic limit, obtained by substituting (105), (126), (127) into equation (125). We again observe an oscillating behavior. The approximation is good for $n \geq 4$.

5.4. Gaussian approximation

We can now summarize the results obtained for the first three cumulants and try to get a broader picture. Equations (73), (104) and (125) are all exact. Their asymptotic expansions for large
$N$ are given in equations (74), (107) and (128). By plugging these results into equations (47) and (48), we obtain the asymptotic expressions of the average energy

$$\langle H \rangle_\beta \sim \frac{2}{\sqrt{N}} - \frac{\beta}{N^{3-a}} + \frac{\beta^2}{2N^\gamma} + \frac{64c}{2N^{4.1583}}$$

and the free energy

$$F(\beta) \sim \frac{\ln Z(0)}{\beta} - \frac{2}{\sqrt{N}} + \frac{\beta}{2N^{3-a}} - \frac{64c}{6N^\gamma} + \frac{205.7}{2N^{4.1583}}$$

where $c \simeq 1.054$, $\alpha = \log_2 3 - 1 \simeq 0.5850$, $\gamma = 4.1583$. (131)

See equations (C.30) and (C.31).

If $N$ is large enough and the first two cumulants at $\beta = 0$ suffice, the energy distribution (33) can be taken to be Gaussian

$$P_0(E) \sim \frac{1}{\sqrt{2\pi \sigma^2}} \exp\left(-\frac{(E - \mu)^2}{2\sigma^2}\right),$$

where $\mu$ and $\sigma$ are given in (74) and (107), respectively. The energy distribution at arbitrary temperature is then (see equation (43))

$$P_\beta(E) \sim \frac{1}{\sqrt{2\pi \sigma^2}} \exp\left(-\frac{(E - \mu + \beta \sigma^2)^2}{2\sigma^2}\right).$$

This is valid for relatively small $\beta$:

$$\mu - \beta \sigma^2 - \bar{\sigma} \gtrsim 0 \iff \beta \lesssim \mu / \sigma^2 \sim N^{7/2 - \log_2 3}.$$  (134)

Up to this value the probability density rigidly shifts with $\beta$, as is apparent in figure 2, which was obtained by numerically solving equation (34).

5.5. A few comments

The behavior of the cumulants derived in this section is very peculiar. The second and third cumulants follow a nontrivial power dependence, with transcendental exponents (see equations (129)). Interestingly, close scrutiny of the calculation in section 5.3 shows also that $3 - \alpha$, the exponent that governs the $N$-dependence of $\bar{\sigma}^2$, is found in a class of (nondominant) graphs that appear in the evaluation of the third cumulant: the exponent $5 - \alpha$ stems from the graph shown in figure 6(a) (the dominant exponent $\gamma$ stemming from the graph shown in figure 6(b)). This might suggest a possible recursion of the exponent $\alpha$ at all orders in the cumulant expansion. At this stage, we are unable to say if at higher orders the dominant graph for $\zeta(3)$ in figure 6(b) cancels, yielding a series in $N^{q(\alpha)}$ with $q$ a function of $\alpha$.

It would be important to go beyond the Gaussian approximation in order to evaluate the behavior of the left tail of the probability density function, close to $\pi_{ME} = E \simeq E_0$. See figures 1 and 2. This would give us some precious information about the features of MMES and the very structure of entanglement frustration [24]. In particular, it would be interesting to understand the role played by the interference among the bipartitions, in connection with the appearance of frustration in MMESs. See for instance the asymptotic behavior of the second cumulant in equations (107)–(109) and the short discussion that follows. Additional investigation is necessary in order to elucidate these intriguing issues.
6. Concluding remarks and outlook

We have built a statistical mechanical approach to multipartite entanglement, by introducing a partition function in order to tackle a complex optimization problem, whose solutions are the maximally multipartite entangled states that appear as minimal energy configurations.

The scheme adopted here is general. In classical statistical mechanics, temperature is used to fix the energy to a given value in the thermodynamic limit. Analogously, the fictitious temperature introduced here localizes the measure on a set of states whose entanglement (energy) is fixed, and can be larger or smaller than the entanglement associated with typical states.

Remarkably, a strategy like the one adopted in this paper, when applied to the simpler case of bipartite entanglement (at a fixed bipartition) [47], brings to light an involved landscape of phase transitions for the purity. Clearly, the multipartite version of the problem is much more involved, as the picture that emerges is complex and unearths a remarkable interplay between multipartite entanglement and frustration. It would therefore be of great interest to understand whether the phase transition that occurs in the bipartite situation, when there is no average over the bipartitions, survives and has a counterpart in the multipartite scenario. This possibility will be explored in the future.

One important property that we have not investigated here and that is often used to characterize multipartite entanglement is the so-called monogamy of entanglement [15, 48] that essentially states that entanglement cannot be freely shared among the parties. Interestingly, although monogamy is a typical property of multipartite entanglement, it is expressed in terms of a bound on a sum of bipartite entanglement measures. This is reminiscent of the approach taken in this paper. The curious fact that bipartite sharing of entanglement is bounded might have interesting consequences in the present context. It would be worth understanding whether monogamy of entanglement generates frustration.

Finally, we think that the characterization of multipartite entanglement proposed here can be important for the analysis of the entanglement features of many-body systems, such as spin systems and systems close to criticality.

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Appendix A.

We derive here expression (24) of the coupling function. See [46]. We start from definition (23) that can be rewritten as

$$\Delta(k, k'; l, l') = \frac{1}{2} \binom{n}{n_A}^{-1} (\tilde{\Delta}(k, k'; l, l'; [n/2]) + \tilde{\Delta}(k', k; l, l'; [n/2])), \quad (A.1)$$

where

$$\tilde{\Delta}(k, k'; l, l'; n_A) = \sum_{|A|=n_A} \delta_{k_A, l_A} \delta_{k'_A, l'_A} \delta_{k_A, l'_A} \delta_{k'_A, l_A}. \quad (A.2)$$

Let us fix a quadruple of binary strings $(k, k', l, l')$ and a dimension $n_A$. See figure A1. A bipartition $(A, \overline{A})$, with $|A| = n_A$, yields a nonvanishing contribution to sum $(A.2)$ when

$$\delta_{k_A, l_A} \delta_{k'_A, l_A} \delta_{k_A, l'_A} \delta_{k'_A, l_A} = 1, \quad (A.3)$$
that is when
\[ k_A = I_A', \quad k_A' = I_A, \quad k_{\bar{A}} = I_{\bar{A}}, \quad k_{\bar{A}}' = I_{\bar{A}}', \quad (A.4) \]
where we recall that \( k_A = I_A \) means that the substrings of \( k \) and \( l \) are equal, namely \( k_i = l_i \) for all \( i \in A \). By noting that two bits \( k_i \) and \( l_i \) are equal when \( k_i \oplus l_i = 0 \), the above condition can be rephrased as
\[ k_A \oplus I_A' = 0, \quad k_A' \oplus I_A = 0, \quad k_{\bar{A}} \oplus I_{\bar{A}} = 0, \quad k_{\bar{A}}' \oplus I_{\bar{A}}' = 0, \quad (A.5) \]
that is
\[ (k_A \oplus I_A') \lor (k_A' \oplus I_A) = 0, \quad (k_{\bar{A}} \oplus I_{\bar{A}}) \lor (k_{\bar{A}}' \oplus I_{\bar{A}}') = 0. \quad (A.6) \]
Summarizing, a bipartition \((A, \bar{A})\) yields a nonvanishing contribution to (23) if and only if the following substrings are zero
\[ a_\bar{A} = 0 \quad \text{and} \quad b_A = 0, \quad (A.7) \]
where
\[ a = (k \oplus I) \lor (k' \oplus I') \quad \text{and} \quad b = (k \oplus I') \lor (k' \oplus I). \quad (A.8) \]
Note that equation (A.7) implies that
\[ a \land b = 0, \quad (A.9) \]
since \((a \land b)_A = a_A \land 0 = 0 \) and \((a \land b)_{\bar{A}} = 0 \land b_{\bar{A}} = 0 \). On the other hand, the substrings \( a_A \) and \( b_{\bar{A}} \) are totally free, whence
\[ |a| = |a_A| \leq |A| = n_A, \quad |b| = |b_{\bar{A}}| \leq |\bar{A}| = n_{\bar{A}}. \quad (A.10) \]
It is easy to see that (A.9) and (A.10) are also sufficient conditions for the existence of a partition \((A, \bar{A})\) that satisfies (A.7).

In conclusion, \( \tilde{\Delta}(k, k'; l, l'; n_A) \neq 0 \) when
\[ a \land b = 0, \quad \text{with} \quad |a| \leq n_A, \quad |b| \leq n_{\bar{A}}. \quad (A.11) \]
Therefore,
\[ \tilde{\Delta}(k, k'; l, l'; n_A) = \delta_{a \land b = 0} \chi_{|0, n_A|}(|a|) \chi_{|0, n_{\bar{A}}|}(|b|) \#(k, k', l, l'), \quad (A.12) \]
where \( \chi_G \) is the characteristic function of set \( G \) and \#(k, k', l, l') is the number of terms in the sum in (A.1) that contribute to the function \( \tilde{\Delta} \).
Note that, since by (A.11) both the strings $a$ and $b$ cannot be 1 at the same position, the set $S$ is partitioned into three disjoint subsets (see figure A1)

$$S = S_0 \cup A_1 \cup B_1,$$

where

$$S_0 = \{ i \in S \mid a_i = 0, \; b_i = 0 \},$$
$$A_1 = \{ i \in S \mid a_i = 1, \; b_i = 0 \},$$
$$B_1 = \{ i \in S \mid a_i = 0, \; b_i = 1 \}.$$  \hfill (A.14)

Obviously, $|A_1| = |a| \leq n_A$ and $|B_1| = |b| \leq n_{\bar{A}}$.

The number of terms $\#(k, k', l, l')$ is given by the number of bipartitions $(A, \bar{A})$ with $|A| = n_A$ such that

$$A_1 \subset A \quad \text{and} \quad B_1 \subset \bar{A}. \hfill (A.15)$$

Since $A \cap \bar{A} = \emptyset$, parties $A$ and $\bar{A}$ contend for $S_0$, namely

$$A = A_1 \cup (S_0 \cap A) \quad \text{and} \quad \bar{A} = B_1 \cup (S_0 \cap \bar{A}). \hfill (A.16)$$

Thus, the number of bipartition is the number of ways of picking $|A \setminus A| \leq n_A - |a|$ unordered outcomes from $|S_0|$ possibilities. Since $|A \setminus A| = |A| - |A_1| = n_A - |a|$ and $|S_0| = |S| - |A_1| - |B_1| = n - |a| - |b|$, one obtains

$$\#(k, k', l, l') = \binom{n - |a| - |b|}{n_A - |a|}. \hfill (A.17)$$

Substituting equation (A.17) into equation (A.12) and by defining the binomial coefficient to be identically zero when its arguments are negative, we note that the characteristic functions in (A.12) yield always 1, and obtain equation (24).

Appendix B.

We derive here the asymptotic (for large $N$) behavior of the function $f_2(N)$ defined in equation (105). Let us define the distance between bipartitions $(A, \bar{A})$ and $(B, \bar{B})$ as the number of qubits belonging to $A$ and not to $B$

$$d = |A \cap \bar{B}| = |B \cap \bar{A}| \in [0, n_A]. \hfill (B.1)$$

The number of pairs of bipartitions at a distance $d$ is

$$n_d = \binom{n}{n_A} \binom{n_A}{d} \binom{n_{\bar{A}}}{d}. \hfill (B.2)$$

Therefore, the sum over the bipartitions can be rewritten as a sum over $d$

$$\sum_{|A|, |\bar{A}| = n_A} \cdot \cdot \cdot = \binom{n}{n_A} \sum_{d=0}^{n_A} \binom{n_A}{d} \binom{n_{\bar{A}}}{d} \cdot \cdot \cdot. \hfill (B.3)$$

Let us consider for instance

$$\sum_{k,l \in \mathbb{Z}_2^n} g(k, l)^2 = \frac{1}{N} \sum_{k \in \mathbb{Z}_2^n} \Delta(k_1, k_2; k_3, k_4)^2$$

$$= \frac{1}{N} \binom{n}{n_A}^{-2} \sum_k \sum_{|A|, |\bar{B}| = n_A} \frac{1}{4} \{ \delta(k; A)\delta(k; B) + \delta(k; A)\delta(k; \bar{B})$$

$$+ \delta(k; \bar{A})\delta(k; B) + \delta(k; \bar{A})\delta(k; \bar{B}) \}$$

$$= \frac{1}{N} \binom{n}{n_A}^{-2} \sum_k \sum_{|A|, |\bar{B}| = n_A} \frac{1}{2} \{ \delta(k; A)\delta(k; B) + \delta(k; A)\delta(k; \bar{B})$$

$$+ \delta(k; \bar{A})\delta(k; B) + \delta(k; \bar{A})\delta(k; \bar{B}) \}, \hfill (B.4)$$

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where
\[ \delta(k, k', l, l'; A) = \delta_{k_{x}, l_{x}} \delta_{k_{y}, l_{y}} \delta_{k_{z}, l_{z}}. \]  
(\text{B.5})

Let us start by showing that
\[ h(A, B) = \sum_{k} \delta(k; A) \delta(k; B) = \sum_{k, k', l, l'} \delta_{k_{x}, l_{x}} \delta_{k_{y}, l_{y}} \delta_{k_{z}, l_{z}}. \]
(\text{B.6})
depends only on \( d = |A \cap B| \). When \( A = B \), i.e. \( d = 0 \),
\[ h(A, A) = \sum_{k} \delta(k; A)^2 = \sum_{k} \delta_{k_{x}, l_{x}} \delta_{k_{y}, l_{y}} \delta_{k_{z}, l_{z}} = N^2, \]
(\text{B.7})
while, when \( d \neq 0 \), we obtain
\[ h(A, B) = \left( \frac{N}{2^d} \right)^2 = 4^{-d} N^2. \]
(\text{B.8})

Therefore, we obtain
\[ \sum_{k} \sum_{|A|, |B| = n_k} \delta(k; A) \delta(k; B) = \sum_{|A|, |B| = n_k} h(A, B) \]
\[ = \sum_{0 \leq d \leq n_k} n_d 4^{-d} N^2 = \left( \frac{n}{n_k} \right) \sum_{0 \leq d \leq n_k} 4^{-d} \left( \frac{n_{\bar{A}}}{d} \right) \left( \frac{n_{\bar{A}}}{d} \right). \]
(\text{B.9})

Analogously, we find
\[ \sum_{k} \sum_{|A|, |B| = n_k} \delta(k; A) \delta(k; \bar{B}) = \left( \frac{n}{n_k} \right) \sum_{0 \leq d \leq n_k} 4^{-d} \left( \frac{n_{\bar{A}}}{d} \right) \left( \frac{n_{\bar{A}}}{d} \right). \]
(\text{B.10})

Putting together equations (\text{B.9}) and (\text{B.10}), we obtain
\[ f_2(N) = 4 \sum_{k,l} g(k, l)^2 = 4 \left( \frac{n}{n_k} \right)^{-1} \sum_{0 \leq d \leq n_k} \left( \frac{n_{\bar{A}}}{d} \right) \left( \frac{n_{\bar{A}}}{d} \right) (4^{n/2-d} + 4^d) \]
\[ = 2 \left( \frac{n}{n_k} \right)^{-1} \sum_{0 \leq d \leq n_k} \left( \frac{n_{\bar{A}}}{d} \right) \left( \frac{n_{\bar{A}}}{d} \right) 2^{n/2}[4^{n/4-d} + 4^{-(n/4-d)}]. \]
(\text{B.11})

We note that the terms in the summation strongly depend on the ratio \( 2d/n \). In the limit \( n \to +\infty \) only the terms with \( d = n/6 \) and \( d = n/3 \) give a significant contribution to the summation (see figure B1). Let us consider the case of even \( n \) (in the thermodynamic limit the result for an odd number of qubits is the same)
\[ n_{\bar{A}} = n_{\bar{A}} = n/2. \]
(\text{B.12})

By Stirling’s approximation \( n! \sim (n/e)^n \sqrt{2\pi n} \) (for \( n \) large) and by defining the new variable
\[ x = \frac{2d}{n}, \]
(\text{B.13})
after a straightforward calculation, we obtain
\[ f_2(N) \sim 2 \sqrt{\frac{\pi n}{2}} 2^{-n} \sum_{d} \frac{1}{\pi n x (1 - x)} \exp \left[ n S(x) \right] 2^{n/2}[4^{\frac{1}{2}(1-x)} + 4^{-\frac{1}{2}(1-x)}] \]
\[ \sim 2 \sqrt{2\pi n} \int_{0}^{1} dx \frac{1}{2 \pi x (1 - x)} \exp \left[ n [S(x) - x \ln 2] \right] + 2^{-n} \exp \left[ n [S(x) + x \ln 2] \right]. \]
(\text{B.14})
Figure B1. $d$-Dependence of the terms $\left(\binom{n_A}{d}\right)^22^{n/2}\left[4^{n/4-d}+4^{-n/4-d}\right]$ in the sum (B.11).

where

$$S(x) = -x \ln x - (1-x) \ln (1-x)$$  \hspace{1cm} \text{(B.15)}

is the Shannon entropy. Using the saddle point approximation in the integrand, we obtain

$$f_2(N) \sim \sqrt{2\pi n} \int_0^1 dx \frac{9}{4\pi} \exp \left\{ n \left[ S\left(\frac{1}{3}\right) - \frac{1}{3} \ln 2 + \frac{1}{2} S''\left(\frac{1}{3}\right) \left(x - \frac{1}{3}\right)^2\right] \right\}$$

$$= \sqrt{2\pi n} \int_0^1 dx \frac{9}{4\pi} \exp \left\{ n \ln \frac{3}{2} \right\} \times \left[ \exp \left\{ -n \frac{9}{4} \left(x - \frac{1}{3}\right)^2\right\} + \exp \left\{ -n \frac{9}{4} \left(x - \frac{2}{3}\right)^2\right\} \right]$$

$$\sim \sqrt{2\pi} \left(\frac{3}{2}\right)^n \int_{-\infty}^{+\infty} dx \exp \left\{ -\frac{9}{4} x^2\right\} = 3\sqrt{2} N^\alpha,$$  \hspace{1cm} \text{(B.16)}

where

$$\alpha = \log_2 3 - 1 \simeq 0.584 \ 963.$$  \hspace{1cm} \text{(B.17)}

This is the asymptotic expression (105) used in equation (107).

Appendix C.

We evaluate here the asymptotic behavior of the function $f_3^{(0)}(N)$ defined in (120). By using definition (25), (120) can be written as

$$f_3^{(0)}(N) = \sum_{k_1, k_2, k_3} g(k_1, k_2) g(k_3, k_2) g(k_1 \oplus k_3, k_2)$$

$$= \sum_{k_1, k_2, k_3} \delta_{k_1 \wedge k_2, 0} \delta_{k_2 \wedge k_3, 0} \delta_{k_2 \wedge (k_1 \oplus k_3), 0} \delta_{|k_1|, |k_2|} \delta_{|k_3|, |k_2|} \delta_{|k_1 \oplus k_3|, |k_2|}$$

$$= \sum_{s_0, s_1, s_2, s_3} f(s_0, s_1, s_2, s_3) \delta_{s_1, s_2} \delta_{s_3, s_2} \delta_{s_1 + s_3 - 2s_0, s_2},$$  \hspace{1cm} \text{(C.1)}

where

$$f(s_0, s_1, s_2, s_3) = \sum_{k_1, k_2, k_3} \delta_{k_1 \wedge k_2, 0} \delta_{k_2 \wedge k_3, 0} \delta_{k_1 |k_1|, |k_2|} \delta_{s_2 |k_2|} \delta_{s_3 |k_3|} \delta_{s_0 |k_1 \wedge k_3|}$$  \hspace{1cm} \text{(C.2)}
and we have used
\[ k_1 \wedge k_2 = 0, \quad k_2 \wedge k_3 = 0 \quad \Rightarrow \quad k_2 \wedge (k_1 \oplus k_3) = 0, \]  
(C.3)

\[ |k_1 \oplus k_3| = |k_1| + |k_3| - 2|k_1 \wedge k_3|. \]  
(C.4)

It is straightforward to count the number of terms in (C.2) and obtain
\[
f(s_0, s_1, s_2, s_3) = \binom{n}{s_2} \binom{n-s_2}{s_1} \binom{s_1}{s_0} \binom{n-s_2-s_1}{s_3-s_0} n!
\]
\[= \frac{n!}{s_0!s_1!s_2!s_3!(n-s_1-s_2-s_3-s_0)!} \times \hat{g}(s_1+s_0, s_2) \hat{g}(s_1+s_0, s_2) \hat{g}(s_1+s_3, s_2).
\]  
(C.5)

By substituting \(s_1 \rightarrow s_1 + s_0\) and \(s_3 \rightarrow s_3 + s_0\) in equation (C.1), we obtain
\[
f_3^{(0)}(N) = \sum_s \binom{n}{s_0, s_1, s_2, s_3} \hat{g}(s_1+s_0, s_0) \hat{g}(s_2+s_0, s_0) \hat{g}(s_3+s_1, s_0).
\]  
(C.6)

where
\[
\binom{n}{i_1, i_2, \ldots, i_k} = \frac{n!}{\prod_{j=1}^k (i_j)! (n - \sum_{j=1}^k i_i)!}
\]  
(C.7)

denotes the multinomial coefficient. A relabeling of the dummy variables yields
\[
f_3^{(0)}(N) = \sum_s \binom{n}{s_0, s_1, s_2, s_3} \hat{g}(s_1+s_2, s_0) \hat{g}(s_2+s_0, s_0) \hat{g}(s_3+s_1, s_0).
\]  
(C.8)

From definition (26), one easily obtains
\[
\hat{g}(s, t) = \frac{1}{2} \left( \binom{n}{s, t} - \binom{n}{t, s} \right) \left[ \binom{n_A}{s} \binom{n_A}{t} + \binom{n_A}{t} \binom{n_A}{s} \right].
\]  
(C.9)

Therefore, for \(n_A = n_A = n/2\), we finally obtain
\[
f_3^{(0)}(N) = \sum_s \binom{n}{s_0, s_1, s_2, s_3} \left( \frac{n/2}{s_0} \right)^3 \prod_{1 \leq i \leq 3} \binom{n}{s_{0i}+s_i+s_{i+1}}^{-1} \left( \frac{n/2}{s_{0i}+s_i+s_{i+1}} \right).
\]  
(C.10)

Now, by using the Stirling approximation and scaling the variables
\[
s_0 = \frac{s_0}{n}, \quad s_1 = \frac{s_1}{n}, \quad s_2 = \frac{s_2}{n}, \quad s_3 = \frac{s_3}{n}
\]  
(C.11)

we obtain, after some algebra, the asymptotic form
\[
f_3^{(0)}(N) \sim \frac{1}{(2\pi n)^{3/2}} \sum_4 A(s_0, s_1, s_2, s_3) \exp[nS(s_0, s_1, s_2, s_3)].
\]  
(C.12)

In equation (C.12), we have set (with the implicit convention that the indices are cyclical)
\[
A = \sqrt{\frac{\prod_{i=1}^3 (1-\sigma_0-\sigma_i-\sigma_{i+1})}{\sigma_0(1-2\sigma_0)^3(1-\sigma_0-\sigma_0-\sigma_2-\sigma_3)} \prod_{i=1}^3 \sigma_i (1-2\sigma_i-2\sigma_{i+1})}
\]  
(C.13)
and

\[ S(\sigma_0, \sigma_1, \sigma_2, \sigma_3) = S_1(\sigma_0, \sigma_1, \sigma_2, \sigma_3) - S_2(\sigma_0, \sigma_1 + \sigma_2) - S_3(\sigma_0, \sigma_2 + \sigma_3) \\
- S_4(\sigma_0, \sigma_3 + \sigma_1) + \frac{1}{2} S_1(2\sigma_0) + \frac{1}{2} S_1(2\sigma_1 + 2\sigma_2) \\
+ \frac{1}{2} S_1(2\sigma_2 + 2\sigma_3) + \frac{1}{2} S_1(2\sigma_3 + 2\sigma_1), \tag{C.14} \]

with

\[ S_n(x_1, \ldots, x_n) = - \sum_{i=1}^{n} x_i \log x_i - \left(1 - \sum_{i=1}^{n} x_i\right) \log \left(1 - \sum_{i=1}^{n} x_i\right). \tag{C.15} \]

By noting that

\[ S_n(x_1, \ldots, x_n) = \sum_{i=1}^{n} x_i \frac{\partial S_n}{\partial x_i} - \log \left(1 - \sum_{i=1}^{n} x_i\right), \tag{C.16} \]

one easily obtains

\[ S = \sigma_0 \frac{\partial S}{\partial \sigma_0} + \sum_{i=1}^{3} \sigma_i \frac{\partial S}{\partial \sigma_i} + S_0, \tag{C.17} \]

with

\[ S_0 = \frac{1}{2} \log \frac{\prod_{i=1}^{3} (1 - \sigma_0 - \sigma_i - \sigma_{i+1})^2}{(1 - \sigma_0 - \sigma_1 - \sigma_2 - \sigma_3)^3 (1 - 2\sigma_0)^3} \prod_{i=1}^{3} (1 - 2\sigma_i - 2\sigma_{i+1}). \tag{C.18} \]

In the limit \( n \to \infty \), the main contribution comes from the saddle point \((\sigma_0^*, \sigma_1^*, \sigma_2^*, \sigma_3^*)\), solution to the system

\[ \frac{\partial S}{\partial \sigma_i} = 0, \quad \text{with} \quad i = 0, 1, 2, 3, \tag{C.19} \]

that reads

\[ (1 - \sigma_0 - \sigma_1 - \sigma_2 - \sigma_3)(1 - 2\sigma_0)^3 = 8\sigma_0 \prod_{i=1}^{3} (1 - \sigma_0 - \sigma_i - \sigma_{i+1}), \tag{C.20} \]

\[ (1 - \sigma_0 - \sigma_1 - \sigma_2 - \sigma_3)(1 - 2\sigma_i - 2\sigma_{i+1}) (1 - 2\sigma_i - 2\sigma_{i+2}) = 4\sigma_i(1 - \sigma_0 - \sigma_i - \sigma_{i+1})(1 - \sigma_0 - \sigma_i - \sigma_{i+2}), \tag{C.21} \]

with \( i = 1, 2, 3 \). In the limit \( n \to \infty \), we obtain

\[ f_3^{(0)}(N) \sim \left( \frac{n}{2\pi} \right)^{2} A^* e^{S_n} \int_{\mathbb{R}^+}^{\mathbb{R}^+} \exp \left( \frac{n}{2} \sum_{i,j=0}^{3} \frac{\partial^2 S^*}{\partial \sigma_i \partial \sigma_j} (\sigma_i - \sigma_i^*)(\sigma_j - \sigma_j^*) \right) \mathrm{d}\sigma_0 \mathrm{d}\sigma_1 \mathrm{d}\sigma_2 \mathrm{d}\sigma_3 \]

\[ = A^* \det \left( \frac{\partial^2 S^*}{\partial \sigma_i \partial \sigma_j} \right)^{-1/2} \exp(nS_n^*) \tag{C.22} \]

where the starred functions \( A^*, S^* \) and \( \partial^2 S^*/\partial \sigma_i \partial \sigma_j \) are evaluated at the saddle point \((\sigma_0^*, \sigma_1^*, \sigma_2^*, \sigma_3^*)\).

The symmetry of the equations suggests to look at a symmetric solution of (C.21) with

\[ \sigma_i = \sigma \quad \text{with} \quad i = 1, 2, 3, \tag{C.23} \]

which yields

\[ (1 - \sigma_0 - 3\sigma)(1 - 2\sigma_0)^3 = 8\sigma_0(1 - \sigma_0 - 2\sigma)^3, \]

\[ (1 - \sigma_0 - 3\sigma)(1 - 4\sigma)^2 = 4\sigma(1 - \sigma_0 - 2\sigma)^2. \tag{C.24} \]
We get

\[ A^* = \sqrt{\frac{(1 - \sigma_0^* - 2\sigma^*)^3}{\sigma_0^*\sigma^3(1 - 2\sigma_0^*)^3(1 - 4\sigma^*)^3(1 - \sigma_0^* - 3\sigma^*)}} = \sqrt{\frac{1}{8\sigma_0^2\sigma^3(1 - 4\sigma^*)^3}}, \]  

(C.25)

\[ S_0^* = \frac{1}{2} \log \frac{(1 - \sigma_0^* - 2\sigma^*)^6}{(1 - \sigma_0^* - 3\sigma^*)^2(1 - 2\sigma_0^*)^3(1 - 4\sigma^*)^3} \equiv \frac{1}{2} \log \frac{(1 - 2\sigma_0^*)^3}{64\sigma_0^2(1 - 4\sigma^*)^3}, \]  

(C.26)

and

\[ \det \left( \frac{\partial ^2 S^*}{\partial \sigma_i \partial \sigma_j} \right) = \frac{(1 - \sigma_0^* - 5\sigma^* + 2\sigma^*(4\sigma^* + \sigma_0^*))^2}{\sigma_0^*\sigma^3(1 - 2\sigma_0^*)^3(1 - 4\sigma^*)^3(1 - \sigma_0^* - 3\sigma^*)(1 - \sigma_0^* - 2\sigma^*)^3} \times (1 - 2\sigma^* - \sigma_0^2 - 4\sigma^2 + 4\sigma^*\sigma_0^8(8\sigma^* + 2\sigma_0^* - 3)). \]  

(C.27)

The solution of the system that gives the largest contribution is

\[ \sigma_0^* = \frac{13}{36} - \frac{13}{36\sqrt{197 - 18\sqrt{113}}} - \frac{1}{36} \sqrt{197 - 18\sqrt{113}} \approx 0.108955, \]

\[ \sigma^* = -\frac{5043923}{144(197 - 18\sqrt{113})^{7/3}} + \frac{158161\sqrt{113}}{48(197 - 18\sqrt{113})^{7/3}} \]

\[ + \frac{980473}{72(197 - 18\sqrt{113})^2} - \frac{2561\sqrt{113}}{2(197 - 18\sqrt{113})^2} \]

\[ - \frac{18119}{144(197 - 18\sqrt{113})^{5/3}} + \frac{563\sqrt{113}}{48(197 - 18\sqrt{113})^{5/3}} \]

\[ \approx 0.104767. \]  

(C.28)

Plugging these results into equation (C.22), we obtain

\[ f_3^{(0)}(N) \sim c N^{5-\gamma} \]  

(C.29)

where

\[ c = A^* \det \left( \frac{\partial ^2 S^*}{\partial \sigma_i \partial \sigma_j} \right)^{-1/2} \]

\[ = \frac{(1 - \sigma_0^* - 2\sigma^*)^3}{1 - 2\sigma_0^*} \]

\[ \times \frac{1}{(1 - \sigma_0^* - 5\sigma^* + 2\sigma^*(4\sigma^* + \sigma_0^*))\sqrt{1 - 2\sigma^* - \sigma_0^2 - 4\sigma^2 + 4\sigma^*\sigma_0^8(8\sigma^* + 2\sigma_0^* - 3)}} \]

\[ \approx 1.05385, \]  

(C.30)

and

\[ \gamma = 5 - S_0^* \log_2 e = 5 - \frac{1}{2} \log_2 \left[ \frac{(1 - 2\sigma_0^*)^3}{64\sigma_0^2(1 - 4\sigma^*)^3} \right] \approx 4.1583. \]  

(C.31)
Appendix D.

We evaluate here the asymptotic behavior of the function \( f_3^{(1)}(N) \) defined in (118). By using definition (25), we have

\[
f_3^{(1)}(N) = \sum_{k_1, k_2, k_3} g(k_1, k_2 \oplus k_3) g(k_2, k_1 \oplus k_1) g(k_3, k_1 \oplus k_2)
\]

\[
= \sum_{k_1, k_2, k_3} \delta_{k_1, k_3} \delta_{k_2, k_3} \delta_{k_1 \oplus k_2, 0} \times \hat{g}(|k_1|, |k_2|, |k_1 \oplus k_2|) \delta_{k_1, k_1} \delta_{k_2, k_2} \delta_{k_3, k_3} \delta_{k_1 \oplus k_2, k_1 \oplus k_2} + \hat{g}(s_1, s_2 + s_3 - 2s_0) \hat{g}(s_2, s_1 + s_2 - 2s_0)
\]

\[
= \sum_{n_0, n_1, n_2} h(n_0, n_1, n_2) g(s_1, s_2 + s_3 - 2s_0) \hat{g}(s_2, s_1 + s_2 - 2s_0)
\]

\[
= \sum_{n_0, n_1, n_2} h(n_0, n_1, n_2) \delta_{n_1, n_2} \delta_{s_1, s_2} \delta_{s_1, s_3} \delta_{s_2, s_1} \delta_{s_2, s_3} \delta_{s_3, s_2} \delta_{s_3, s_1} \delta_{s_1, s_1} \delta_{s_2, s_2} \delta_{s_3, s_3} \delta_{s_1, s_1} \delta_{s_2, s_2} \delta_{s_3, s_3}
\]

\[
= \sum_{n_0, n_1, n_2} h(n_0, n_1, n_2) \delta_{n_1, n_2} \delta_{s_1, s_2} \delta_{s_1, s_3} \delta_{s_2, s_1} \delta_{s_2, s_3} \delta_{s_3, s_2} \delta_{s_3, s_1} \delta_{s_1, s_1} \delta_{s_2, s_2} \delta_{s_3, s_3} \delta_{s_1, s_1} \delta_{s_2, s_2} \delta_{s_3, s_3}
\]

where

\[
h(n_0, n_1, n_2) = \sum_{k_1, k_2, k_3} \delta_{n_1, n_2} \delta_{s_1, s_2} \delta_{s_1, s_3} \delta_{s_2, s_1} \delta_{s_2, s_3} \delta_{s_3, s_2} \delta_{s_3, s_1} \delta_{s_1, s_1} \delta_{s_2, s_2} \delta_{s_3, s_3} \delta_{s_1, s_1} \delta_{s_2, s_2} \delta_{s_3, s_3}
\]

and we have used

\[
k_1 \wedge (k_2 \oplus k_3) = 0, \quad k_2 \wedge (k_1 \oplus k_3) = 0, \quad k_3 \wedge (k_1 \oplus k_2) = 0
\]

\[
\implies k_1 \wedge k_2 = k_1 \wedge k_3 = k_2 \wedge k_3,
\]

\[
k_1 \oplus k_j = |k_1| + |k_j| - 2|k_i \wedge k_j| \quad \forall \ i, j = 1, 2, 3.
\]

We find

\[
h(n_0, n_1, n_2) = \frac{n!}{s_0! (s_1 - s_0)! (s_2 - s_0)! (s_3 - s_0)! (n - s_2 - s_1 - s_3 + 2s_0)!}
\]

Using the substitution \( s_1 \rightarrow s_1 + s_0, s_2 \rightarrow s_2 + s_0 \) and \( s_3 \rightarrow s_3 + s_0 \) in equation (D.1), we obtain

\[
f_3^{(1)}(N) = \sum_s \left( \frac{n}{s_0, s_1, s_2, s_3} \right) \hat{g}(s_1 + s_0, s_2 + s_1) \hat{g}(s_2 + s_0, s_1 + s_3) \hat{g}(s_3 + s_0, s_1 + s_2)
\]

in terms of the multinomial coefficient (C.7). Using (C.9) for \( n_A = n_A = n/2 \), we finally obtain

\[
f_3^{(1)}(N) = \sum_s \left( \frac{n}{s_0, s_1, s_2, s_3} \right) \prod_{1 \leq i < j \leq 3} \left( s_0 + s_1 + s_{i+1} + s_{j+1} \right)^{-1} \left( \frac{n/2}{s_o + s_1} \right) \left( \frac{n/2}{s_j + s_{i+1}} \right)
\]

Using the Stirling approximation and equation (C.11), we obtain

\[
f_3^{(1)}(N) \sim \frac{1}{(2\pi n)^2} \sum_s A(s_0, s_1, s_2, s_3) \exp[nS(s_0, s_1, s_2, s_3)]
\]

where

\[
A(s_0, s_1, s_2, s_3) = \frac{(1 - s_0 - s_1 - s_2) \sqrt{s_0 s_1 s_2 s_3 \prod_{1 \leq i < j \leq 3} (1 - 2s_0 - 2s_i)(1 - 2s_i + 2s_{i+2})}}{s_0 s_1 s_2 s_3}
\]

and

\[
S(s_0, s_1, s_2, s_3) = S_1(s_0, s_1, s_2, s_3) - S_2(s_0 + s_1, s_2 + s_3)
\]

\[
= -S_2(s_0 + s_2, s_3 + s_1) - S_2(s_0 + s_3, s_3 + s_1) - S_2(s_0 + s_1, s_2 + s_3)
\]

\[
= \frac{1}{2} S_1(2s_0 + 2s_1) + \frac{1}{2} S_1(2s_0 + 2s_2) + \frac{1}{2} S_1(2s_0 + 2s_3) + \frac{1}{2} S_1(2s_0 + 2s_1) + \frac{1}{2} S_1(2s_3 + 2s_2).
\]
with the entropies defined in equation (C.15). By (C.16) one obtains (C.17) with
\[ S_0 = \frac{1}{2} \log \prod_{1 \leq i \leq 3} (1 - 2\sigma_0 - 2\sigma_i)(1 - 2\sigma_{i+1} - 2\sigma_{i+2}) . \tag{D.10} \]

In the limit \( n \to +\infty \), we can use the saddle point approximation. The saddle point is solution to the set of equations
\[ \frac{\partial S}{\partial \sigma_i} = 0, \quad \text{with} \quad i = 0, 1, 2, 3, \tag{D.11} \]
that reads
\[ (1 - 2\sigma_i - 2\sigma_{i+1})(1 - 2\sigma_i - 2\sigma_{i+2})(1 - 2\sigma_i - 2\sigma_{i+3}) = 8\sigma_i(1 - \sigma_0 - \sigma_1 - \sigma_2 - \sigma_3)^2. \tag{D.12} \]
The symmetric solution, \( \sigma_i = \sigma \) for all \( i \), corresponds to the largest contribution and is given by
\[ \sigma^* = \frac{1}{12}. \tag{D.13} \]
As in (C.22), in the limit \( n \to \infty \) we obtain
\[ f_3^{(1)}(N) \sim A^* \det \left( \frac{\partial^2 S^*}{\partial \sigma_i \partial \sigma_j} \right)^{-1/2} \exp \left( nS_0^* \right), \tag{D.14} \]
with
\[ A^* = \frac{1}{\sigma^*(1 - 4\sigma^*)^2} = 324, \quad \det \left( \frac{\partial^2 S^*}{\partial \sigma_i \partial \sigma_j} \right) = \frac{1}{\sigma^*(1 - 4\sigma^*)^4} = 324^2, \tag{D.15} \]
and
\[ S_0^* = \log \frac{1}{(1 - 4\sigma^*)} = \log \frac{3}{2}. \tag{D.16} \]
We finally obtain
\[ f_3^{(1)}(N) \sim e^{n \log(3/2)} = N^\alpha, \quad \text{for} \quad N \to \infty, \tag{D.17} \]
with
\[ \alpha = \log_2 3 - 1 \simeq 0.584963. \tag{D.18} \]

References
