GIBBS MEASURES ON PERMUTATIONS
OVER ONE-DIMENSIONAL DISCRETE POINT SETS

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Abstract: We consider Gibbs distributions on permutations of a locally-finite infinite set \( X \subset \mathbb{R} \), where a permutation \( \sigma \) of \( X \) is assigned (formal) energy \( \sum_{x \in X} V(\sigma(x) - x) \). This is motivated by Feynman’s path representation of the quantum Bose gas; the choice \( X := \mathbb{Z} \) and \( V(x) := \alpha x^2 \) is of principal interest. Under suitable regularity conditions on the set \( X \) and the potential \( V \), we establish existence and a full classification of the infinite-volume Gibbs measures for this problem, including a result on the number of infinite cycles of typical permutations. Unlike earlier results, our conclusions are not limited to small densities and/or high temperatures.

1. INTRODUCTION

1.1 Motivation.

One of the principal difficulties underlying quantum statistical mechanics is the non-commutative nature of the relevant observables. For some systems, the said difficulty can sometimes be reduced by developing a suitable classical, and often probabilistic, representation of the problem at hand. Interestingly, this can be done for quite a few examples of interest — namely, the quantum rotator, the quantum Heisenberg model, the Ising model in a transversal field and, most notably, the Bose gas. The classical representation is still hard to analyze but some results often follow. See Tóth [22] and Aizenman and Nachtergaele [3] for early studies of such representations.

In this paper we take up a model that is derived from the classical representation of interacting Bose gas. This representation is originally due to Feynman [10] to whom it served as a mathematical tool for the analysis of the onset of Bose-Einstein condensation in \(^3\)He. Feynman’s representation yields a model on \( N \) classical particles in positions \( x_1, \ldots, x_N \in \mathbb{R}^d \) that are given the weight

\[
\sum_{\sigma \in \Omega_X} \int W^{\beta, \sigma}_X(d\mathcal{B}) \exp\left\{ -\sum_{x \in X} V(\sigma(x) - x) - \sum_{1 \leq i < j \leq N} \int_0^\beta w(B^{(i)}_t - B^{(j)}_t) dt \right\}.
\]

(1.1)

Here \( X := \{x_1, \ldots, x_N\} \) is the set of all particles, \( \Omega_X \) is the set of all permutations of \( X \) (i.e., all one-to-one maps of \( X \) onto \( X \)), \( \beta \) is (twice) the inverse temperature and \( V(x) := \frac{1}{2} \alpha |x|^2 \) so that the first sum in the exponent yields independent Gaussian factors. The probability measure \( W^{\beta, \sigma}_X \) is over collections of independent Brownian bridges \( \mathcal{B} = (B^{(i)}_t : 0 \leq t \leq \beta, 1 \leq i \leq N) \), where the \( i \)-th bridge starts at \( x_i \) and terminates at \( \sigma(x_i) \). The function \( w: \mathbb{R}^d \rightarrow \mathbb{R} \) is the two-body
interaction potential between the bosons. The a priori measure on the positions \(x_1, \ldots, x_N\) is Lebesgue over a finite set; integrating the weight (1.1) defines the normalizing constant called the partition function.

A natural first case to explore is that of no interaction, i.e., \(w := 0\). Averaging the positions over, say, a torus can then be exactly carried out with the help of Fourier representation and the sum over permutations can then also be performed. An outcome of this, envisioned already by Feynman [10], is as follows: For \(d = 1, 2\), any finite density of particles and any \(\beta > 0\), a typical \(\sigma\) will decompose entirely into finite cycles (i.e., of sizes not growing with \(N\)). On the other hand, for \(d \geq 3\) and each \(\beta > 0\) there is a critical density above which a particle is contained in a cycle of length of order \(N\) with positive probability. A mathematical proof of this has been given only recently by Sütő [20, 21]; the critical density turns out to coincide with that for the appearance of the Bose-Einstein condensate.

It has subsequently been observed by Betz and Ueltschi [8] that a similar calculation to the one just mentioned can be carried out for \(V(x) = |x|^2\) replaced by more general potential functions. The principal next challenge from the mathematics point of view is thus to either allow for non-zero interactions, \(w \neq 0\), or to drop the integration over the positions \(x_i\). The former choice is that of the prime interest for physics; unfortunately, at this moment we do not see any tangible way to tackle it. The latter option is nonetheless interesting as well; it leads to natural measures on partitions of point sets in \(\mathbb{R}^d\). This motivation was the basis of an earlier article of Fichtner [11]. More recently, Gandolfo, Ruiz and Ueltschi [12] and Betz and Ueltschi [8] proposed a similar model with the particles placed at the vertices of the integer lattice \(\mathbb{Z}^d\).

1.2 Main questions.

We will henceforth focus on the latter case and formalize it as the following problem: For a given locally-finite set \(X \subset \mathbb{R}^d\) of positions and a potential function \(V : \mathbb{R}^d \to \mathbb{R}\), we wish to consider a probability measure \(\mu\) that is formally given by

\[
\mu(\\{\sigma\}) = \frac{1}{Z} \exp \left(-\sum_{x \in X} V(\sigma(x) - x)\right), \quad \sigma \in \Omega_X.
\]  

(1.2)

If \(X\) is finite, this corresponds to the measure arising from (1.1) with \(w := 0\). If \(X\) is infinite, the expression (1.2) is generally ill defined. In order to extend it to infinite volume (i.e., infinite number of particles), one either has to appeal to limits — an approach previously used in this context by Fichtner [11] and Gandolfo, Ruiz and Ueltschi [12] — or go directly by prescribing infinite-volume Gibbs measures via a family of specifications (Georgii [13]). This will be our choice, so the first question to ask is:

(1) Can we define a consistent family of specifications corresponding to (1.2)?

As we will see later, already this represents a departure from the standard theory. Naturally, one thus immediately adds:

(2) Under what conditions are there (infinite-volume) Gibbs measures for the specifications defined in (1)?

In an approach via limits from finite volume, this boils down to controlling tightness — the issue is that in the limiting measure, no points get mapped to/from infinity. This has, so far, only been accomplished under the assumption of low density/high temperature (cf conditions (V,3) or (5.8))
in Fichtner [11]) or for interaction with a finite-range cutoff (Betz and Ueltschi [8, p. 478]). In addition, all of this is only for the free boundary condition (Fichtner [11, Theorems 2.2 and 3.1]).

Note: When this manuscript was very near its completion, Betz [7] posted a proof of tightness for the measures with periodic boundary conditions over $X := \mathbb{Z}^d$ in all $d \geq 1$ assuming the summability condition of the kind $\sum_{x \in \mathbb{Z}^d} e^{-\delta V(x)} < \infty$ for some $\delta \in (0, 1)$.

Once the setting for Gibbs theory is fixed, the natural follow-up questions concern the structure of the permutations that are typical samples from these measures:

3. Characterize the Gibbs measures that are trivial on the tail sigma field (i.e., the extremal Gibbs measures).

4. Under what conditions does $\sigma$ contain finite cycles only, and when do infinite cycles occur with positive probability?

The mathematical results that are available at present all pertain to the regime of low densities/high temperatures: the measure defined by the free boundary condition contains only finite cycles almost surely (Fichtner [11, Theorem 3.2]). Interesting numerical simulations were performed by Gandolfo, Ruiz and Ueltschi [12] and Grosskinsky, Lovisolo and Ueltschi [15] of the model with $X := \mathbb{Z}^d$ (the integer lattice) and $V(x) := \alpha |x|^2$. These indicated a similar dichotomy as for the ideal Bose gas: only microscopic (finite) cycles in dimensions $d = 1, 2$ and macroscopic cycles in $d = 3$ for $\beta := \frac{1}{\beta}$ sufficiently large.

The principal goal of this paper is to answer the above questions in the case of one-dimensional point sets, $X \subset \mathbb{R}$, subject to natural homogeneity conditions, and a fairly rich class of potentials $V$. Explicitly, we show how to define Gibbs specifications, establish the existence of a family of Gibbs measures and prove that the extremal ones are in one-to-one correspondence with an integer parameter called the flow. This is the quantity whose absolute value gives the (a.s.-constant) number of infinite cycles while its sign tells us their asymptotic direction (it turns out that all infinite cycles necessarily "flow" in the same direction).

1.3 Further related work.

Apart from the abovementioned works that are focused on the connection to the Bose gas, there are numerous other studies of combinatorial nature that deal with similar problems. One line of research is that on compositions of random transpositions (e.g., Schramm [19], Berestycki and Durrett [6], Berestycki [4, 5]) which deal with situations without underlying geometry. Another direction concerns the random stirring process where transpositions "arrive" randomly but only over edges of an underlying graph. For the graph being an infinite regular tree, progress has recently been made concerning the existence and uniqueness of a transition from a regime without infinite cycles to a regime with infinite cycles (Hammond [16, 17]). A recent review by Golschmidt, Ueltschi and Windridge [14] gives further connections between the combinatorial models and quantum systems.

There are also several alternative approaches to statistical mechanics of Bose gases to the one proposed by Feynman. Recently, much progress has been achieved in the analysis of the so-called Gross-Pitaevski limit; this density-function approach is summarized in the monograph by Lieb, Seiringer, Solovej and Yngvason [18]. Other studies were put forward that expand on the ideas of Bogoliubov [9], see, for instance, a review article by Zagrebnov and Bru [24]. The jury is still out on which of these approaches is best suited for understanding the physics, although some
connection of the present problem to the others has also been made, see, e.g., Ueltschi [23] and
the work of Adams, Bru and König [1, 2].

1.4 Outline.

The remainder of the paper is organized as follows: In the next section we define a suitable notion
of Gibbs measures for a given set of points $X$ and potential $V$ satisfying suitable assumptions, we
introduce the main concepts for our analysis and formulate a series of lemmas containing our
findings for Gibbs measures in the given context, concluding with a summarizing theorem. The
following sections contain the proofs for these results.

2. Definitions and results

Here we develop the mathematical framework of our problem and give statements of the results.
Our approach is based on the theory of Gibbs measures but, since we work in a somewhat non-
standard setting, we will be rather pedantic in introducing all necessary notations.

2.1 Permutations on point sets and their flow.

Our aim is to construct a measure of the form (1.2) on permutations on a given countably-infinite
set of points $X \subset \mathbb{R}$. One may want to think of a regular point set such as the set of all integers
$\mathbb{Z}$, but at this point we only assume that $X$ is

(i) locally finite (i.e., any bounded subset of $X$ is finite) and

(ii) bi-infinite (i.e., $X$ is unbounded from above and below).

In order to be able to identify particular points of $X$ with respect to a given position, we use
the following notation: For $a \in \mathbb{R}$ and $n \geq 1$ let $a_n, a_{-n}$ be the unique points of $X$ such that
$
\#([a_n, a] \cap X) = n = \#([a_{-n}, a) \cap X),
$
where $\#(A)$ is our notation for the cardinality of $A$. In other
words, $a_n, a_{-n}$ are the $n$-th point of $X$ lying (strictly) to the right and left of $a$, respectively. If $a \in X$
we also write $a_0 := a$. We set $\Lambda^c := X \setminus \Lambda$ for $\Lambda \subset X$, and write $\Lambda \subseteq X$ if $\Lambda$ is a finite subset of $X$. We
will also write

$$X^* := \left\{ \frac{x_0 + x_1}{2} : x \in X \right\},$$  \quad (2.1)

to denote the dual set of points of $X$.

The configuration space $\Omega_X$ is the set of all permutations (i.e., bijections) on $X$. For given $\sigma \in \Omega_X$ it will be useful to think of a pair $(x, \sigma(x))$ with $x \in X$ as a jump from $x$ to $\sigma(x)$ with the notations

$$x \rightarrow y :\Leftrightarrow \sigma(x) = y \quad \text{and} \quad x \leftrightarrow y :\Leftrightarrow \sigma(x) = y \text{ or } \sigma(y) = x.$$  \quad (2.2)

We will say that $(x, \sigma(x))$ is a jump over $a \in \mathbb{R}$ if $x < a < \sigma(x)$ or $\sigma(x) < a < x$, and it is a jump to the right, resp., left if $\sigma(x) > x$, resp., $\sigma(x) < x$. The quantity $|\sigma(x) - x|$ will be referred to as the length of the jump.

An important tool in our analysis will be the flow $F_a(\sigma)$ of a permutation $\sigma$ through $a \in X^*$. To define this object, we set

$$F_a^+(\sigma) := \# \{ x \in X : x < a < \sigma(x) \} \quad \text{and} \quad F_a^-(\sigma) := \# \{ x \in X : \sigma(x) < a < x \}$$  \quad (2.3)
and define
\[ F_a(\sigma) := \begin{cases} 
F_a^+(\sigma) - F_a^-(\sigma), & \text{if } F_a^+(\sigma), F_a^-(\sigma) < \infty, \\
\infty, & \text{otherwise.} 
\end{cases} \]  
(2.4)

As we will see, the formal (second) value is a proviso that will turn out to be irrelevant for the typical permutations to be considered later. A key fact is that \( F_a(\sigma) \) does not depend on \( a \):

**Lemma 2.1** For locally finite \( X \) and \( \sigma \in \Omega_X \), \( F_a(\sigma) \) has the same value for all \( a \in X^* \).

Therefore, we may (and will) drop \( a \) from the notation and define \( F : \Omega_X \to \mathbb{Z} \cup \{\infty\} \) to be the common value of \( F_a \) for all \( a \in X^* \).

Every permutation \( \sigma \in \Omega_X \) can be decomposed into disjoint cycles, some of which may be infinite. If \( x \in X \) belongs to an infinite cycle and \( \sigma^n(x) \to -\infty \) for \( n \to -\infty \) and \( \sigma^n(x) \to \infty \) for \( n \to \infty \), we say that the cycle is going from \(-\infty \) to \( \infty \). Similarly we may have cycles going from \( \infty \) to \( -\infty \), \(-\infty \) to \( -\infty \), \(-\infty \) to \(-\infty \). If one or both of the above limits do not exist, we say that the infinite cycle is indeterminate. Not too surprisingly, the value of the flow gives some information on the number of such infinite cycles:

**Lemma 2.2** Assume that \( X \) is locally finite. Any \( \sigma \in \Omega_X \) with \( F(\sigma) = n \in \mathbb{Z} \) does not have indeterminate infinite cycles, and it has at least \( |n| \) infinite cycles from \(-\infty \) to \( \infty \) if \( n > 0 \) and at least \( |n| \) infinite cycles from \( \infty \) to \(-\infty \) if \( n < 0 \).

We note that this lemma still leaves the possibility of having additional infinite cycles that do not contribute to the flow, e.g., an infinite cycle from \( \infty \) to \( \infty \) or a pair of cycles, one from \(-\infty \) to \( \infty \) and one from \( \infty \) to \(-\infty \). These will be effectively ruled out in Theorem 2.15.

### 2.2 Energy of permutations.

The energy of a permutation \( \sigma \) will be defined in terms of a given potential function \( V : \mathbb{R} \to \mathbb{R} \). The most interesting choice is \( V(x) := \alpha x^2 \) for \( \alpha > 0 \), but for now we will only assume \( V \) to be symmetric \((V(x) = V(-x))\) for all \( x \in \mathbb{R} \) and strictly convex. The formal Hamiltonian
\[ H(\sigma) := \sum_{x \in X} V(\sigma(x) - x) \]  
(2.5)

does not converge to a finite limit for most permutations \( \sigma \). As usual this problem can be avoided by considering a local version of the energy for a given boundary configuration. For \( \Lambda \subset X \) we define a compatibility relation \( \sim_\Lambda \) between configurations \( \sigma, \eta \in \Omega_X \) by setting
\[ \sigma \sim_\Lambda \eta \quad \iff \quad \forall x \in \Lambda^c : \sigma(x) = \eta(x), \sigma^{-1}(x) = \eta^{-1}(x). \]  
(2.6)

In particular, if \( \sigma \sim_\Lambda \eta \), then \( \sigma \) maps \( \Lambda \cap \eta^{-1}(\Lambda) = \Lambda \cap \sigma^{-1}(\Lambda) \) bijectively onto \( \Lambda \cap \eta(\Lambda) = \Lambda \cap \sigma(\Lambda) \). The Hamiltonian of \( \sigma \in \Omega_X \) in \( \Lambda \subset X \) is defined by
\[ H_\Lambda(\sigma) := \sum_{x \in \sigma^{-1}(\Lambda) \cap \Lambda} V(\sigma(x) - x). \]  
(2.7)

As usual, a configuration will be called a ground state of \( H \) if its energy is smaller than that of any local perturbation thereof:

**Definition 2.3** \( \tau \in \Omega_X \) is said to be a ground state of \( H \) iff
\[ H_\Lambda(\tau) \leq H_\Lambda(\sigma) \quad \text{for all} \ \Lambda \subset X \ \text{and all} \ \sigma \sim_\Lambda \tau. \]  
(2.8)
It turns out that the ground states can be explicitly described:

**Lemma 2.4** If $X$ is locally finite and bi-infinite, and $V$ is strictly convex, then the ground states of $H$ form the set $\{\tau_n: n \in \mathbb{Z}\}$, where $\tau_n$ is the $n$-shift permutation defined by

$$\tau_n(x) := x_n \quad \text{for all } x \in X. \quad (2.9)$$

Note that the ground states $\{\tau_n: n \in \mathbb{Z}\}$ are precisely the increasing bijections of $X$. (We say that $\sigma$ is strictly increasing if $x < y$ implies $\sigma(x) < \sigma(y)$.) Moreover, they are completely parametrized by their flow, $F(\tau_n) = n$, which (as we will see later) will be true even for the Gibbs measures. The proof of Lemma 2.4 and many of the following results rely on the energy comparison of a permutation and its perturbation at exactly two points:

**Definition 2.5** For $x, y \in X$ and $\sigma \in \Omega_X$ we define $\sigma_{xy} \in \Omega_X$ by setting

$$\sigma_{xy}(x) := \sigma(y), \quad \sigma_{xy}(y) := \sigma(x) \quad \text{and} \quad \sigma_{xy}(z) := \sigma(z) \quad \text{for all } z \neq x, y. \quad (2.10)$$

We will sometimes refer to the transformation $\sigma \mapsto \sigma_{xy}$ as a swap. Note that, for $x, y, \sigma(x), \sigma(y) \in \Lambda \Subset X$ we have $\sigma_{xy} \sim_{\Lambda} \sigma$ and

$$H_{\Lambda}(\sigma) - H_{\Lambda}(\sigma_{xy}) = V(\sigma(x) - x) + V(\sigma(y) - y) - V(\sigma(x) - y) - V(\sigma(y) - x). \quad (2.11)$$

This relation is the reason why strict convexity of $V$ is crucial for the validity of Lemma 2.4.

### 2.3 Specifications and Gibbs measures.

We continue to suppose that $X, V$ satisfy the above assumptions. Before defining Gibbs measures on $\Omega_X$, we need to impose topological and measurable structures on $\Omega_X$. We endow $\Omega_X$ with the smallest topology under which all the projections

$$\sigma \mapsto P^+_{\sigma} := \sigma(x) \quad \text{and} \quad \sigma \mapsto P^-_{\sigma} := \sigma^{-1}(x) \quad (x \in X) \quad (2.12)$$

are continuous. (Identifying $\sigma$ with $(\sigma(x), \sigma^{-1}(x))_{x \in X}$ this topology on $\Omega_X$ coincides with the product of the discrete topologies on $X \times X$.) This topology is metrizable, e.g., by

$$d(\sigma, \eta) := \inf \{2^{-r}: \sigma = \eta \quad \text{and} \quad \sigma^{-1} = \eta^{-1} \quad \text{on } X \cap [-r, r]\}, \quad (2.13)$$

and $\Omega_X$ is thus a complete separable metric space.

Let $\mathcal{F}_X := \sigma(P^+_{\sigma}, P^-_{\sigma} : x \in X)$ denote the Borel-$\sigma$-algebra on $\Omega$, and for $\Lambda \subset X$ we use

$$\mathcal{F}_{\Lambda} := \sigma(P^+_{\sigma}, P^-_{\sigma} : x \in \Lambda) \quad (2.14)$$

to denote the $\sigma$-algebra of events depending on $\Lambda$ only. A function $f: \Omega_X \to \mathbb{R}$ is called local if it is measurable with respect to $\mathcal{F}_\Lambda$ for some $\Lambda \Subset X$; an event is called local if its indicator is a local function. Since every local event is open and closed, every local function is continuous. As usual we can use these to define

$$\mathcal{T} := \bigcap_{\Lambda \Subset X} \mathcal{F}_{\Lambda^c}, \quad (2.15)$$

the tail-$\sigma$-algebra of all events that do not depend on what a permutation looks like on any bounded set.

In order to construct (infinite volume) Gibbs measures for the above Hamiltonian $H$, we will use the method of specifications. Here, the specification corresponding to boundary configuration
\( \eta \in \Omega_X \) and volume \( \Lambda \subseteq X \) is the discrete probability measure \( \gamma_{\Lambda}(\cdot | \eta) \) on \( (\Omega_X, \mathcal{F}_X) \) defined by
\[
\gamma_{\Lambda}(\{ \sigma \} | \eta) := \frac{1}{Z_{\Lambda}(\eta)} e^{-H_{\Lambda}(\sigma)} 1_{\{ \sigma \sim_{\Lambda} \eta \}}, \quad \text{where } Z_{\Lambda}(\eta) := \sum_{\sigma : \sigma \sim_{\Lambda} \eta} e^{-H_{\Lambda}(\sigma)}. \tag{2.16}
\]
We note that \( Z_{\Lambda}(\eta) > 0 \) (since \( \sigma := \eta \) gives a non-zero contribution) and \( Z_{\Lambda}(\eta) < \infty \) (since only finitely many permutations are \( \sim_{\Lambda} \)-compatible to \( \eta \) if \( \Lambda \) is finite). Obviously, \( \eta \mapsto \gamma_{\Lambda}(A|\eta) \) is measurable for each \( A \in \mathcal{F}_X \).

**Definition 2.6** A Gibbs measure \( \mu \) with respect to the above Hamiltonian \( H \) is a probability measure on \( (\Omega_X, \mathcal{F}_X) \) that is compatible with all specifications:
\[
E_{\mu}(\gamma_{\Lambda}(A|\cdot)) = \mu(A) \quad \text{for all } A \in \mathcal{F}_X \text{ and all } \Lambda \subseteq X, \tag{2.17}
\]
where \( E_{\mu} \) denotes expectation with respect to \( \mu \). We will write \( \mathcal{G} \) to denote the set of all Gibbs measures for the (implicit) Hamiltonian \( H \).

**Remark 2.7** Our setup differs from that in Georgii [13], which is versed in terms of spin systems. Such a description is permissible in our case as well. A natural attempt would be to proclaim \( \sigma(x) \) to be an \( X \)-valued spin at \( x \), but that choice does not lead to quasilocal specifications. Taking a pair of values \( (\sigma(x), \sigma^{-1}(x)) \) for the spin at \( x \) solves the quasilocality problem, but only at the cost of introducing hard-core restrictions.

### 2.4 Conditions on the point set and the potential.

In order to show the existence (and discuss further properties) of Gibbs measures, we have to impose conditions on \( X \) and \( V \) that are stronger than those considered so far. We did not aim at finding the most general conditions under which our results hold, but rather gave conditions under which the proofs remain transparent and which still include the main examples of interest.

For the formulation of the conditions on \( X \) we introduce constants that control the spacings between the points of \( \Lambda(a,n) := \{a_{-|a|-1}, \ldots, a_{|a|+1}\} \) \( (n \in \mathbb{Z}) \) and the growth rate of the number of particles of \( X \) as seen from \( a \):
\[
c_{X}^s(a, n) := \min \{c \geq 1 : \text{consecutive points of } \Lambda(a,n) \text{ keep distances in } [c^{-1}, c]\} \tag{2.18}
\]
\[
c_{X}^g(a) := \inf \{c \geq 0 : \# \{x \in X : 0 < |x-a| \leq t\} \leq ct \text{ for all } t > 0\}. \tag{2.19}
\]
(The subindices “\( s \)”, resp., “\( g \)” stand for “separation,” resp., “growth.”) Below we will consider the following, progressively restrictive, properties of \( X \subseteq \mathbb{R} \) (and \( n \geq 0 \)):
- (X1) \( X \) is locally finite and bi-infinite (just as before).
- (X2) \( c_{X}^g(a) < \infty \) for some \( a \in \mathbb{R} \).
- (Xn) For some \( c_n < \infty \), bi-infinitely many points \( a \in X^* \) satisfy
\[
c_{X}^s(a, n) \leq c_n \text{ and } c_{X}^g(a) \leq c_n. \tag{2.20}
\]
We note that \( X := \mathbb{Z} \) satisfies the given conditions (since \( c_{X}^g(a, n) = 1 \) and \( c_{X}^g(a) = 2 \) for all \( a \in X^* \)), but the conditions also allow for point sets such that the distance of consecutive points is not bounded from above or below. In particular, we can consider point sets produced by a Poisson point process or other rather general shift-invariant processes:

**Lemma 2.8** Consider a point process (i.e., an \( \mathbb{N} \cup \{\infty\} \)-valued random purely-atomic Borel measure) \( X \) on \( \mathbb{R} \) that has the following properties:
Then the set of points corresponding to a.e. sample of $X$ satisfies (XI) and (Xn) for all $n \in \mathbb{Z}$.

Concerning the potential function $V: \mathbb{R} \to \mathbb{R}$, we will consider the following properties:

(V1) $V$ is symmetric and strictly convex (just as before).

(V2) $V$ satisfies the following growth condition: For all $d > 0$ we have

$$
\psi_d(x) := \frac{V(x) + V(0) - 2V(\frac{x+d}{2})}{x \log x} \to \infty \quad \text{for } x \to \infty.
$$

We note that $V(x) := \alpha |x|^{1+\varepsilon}$ satisfies both (V1) and (V2) for all $\alpha > 0$ and all $\varepsilon > 0$. In particular, this includes the most interesting case $V(x) := \alpha x^2$ for $\alpha > 0$. Linearly growing potentials, e.g., $V(x) := \alpha |x|$, satisfy neither (V1) nor (V2), and indeed we are unable to extend our conclusions to these cases.

### 2.5 Existence of Gibbs measures.

We note that without any assumptions on $V$ it may be the case that there are no Gibbs measures. This can already be seen when $V := 0$, which corresponds to the physically interesting case of zero temperature (i.e., $\alpha := 0$).

**Lemma 2.9** For infinite $X$ and $V := 0$ we have $\mathcal{G} = \emptyset$.

On the other hand, existence can be shown under the conditions introduced in the previous section. At this point the weaker condition (X2) is sufficient.

**Lemma 2.10** Let $X$ and $V$ satisfy (XI),(X2) and (V1),(V2) respectively. For every $n \in \mathbb{Z}$ there is a Gibbs measure $\mu \in \mathcal{G}$ such that $\mu(F = n) = 1$.

So there are in fact many Gibbs measures, at least one for each value of the flow. In all cases the flow is finite and this turns out to be no accident:

**Lemma 2.11** Let $X$ and $V$ satisfy (XI),(X2) and (V1),(V2) respectively. Then $F$ is $\mathcal{T}$-measurable and finite $\mu$-a.s. for every $\mu \in \mathcal{G}$.

As in the classical Gibbs measure theory, $\mathcal{G}$ is a closed convex set. Its extremal elements are called extremal Gibbs measures, and they are precisely those that trivialize on the tail sigma algebra $\mathcal{T}$. Consequently, every Gibbs measure can be decomposed into extremal Gibbs measures by conditioning on $\mathcal{T}$. The previous lemma implies that $F$ is a.s. constant with respect to every extremal Gibbs measure. We may thus filter the Gibbs measures according to the flow and focus attention on sets

$$
\mathcal{G}_n := \{ \mu \in \mathcal{G} : \mu(F = n) = 1 \}.
$$

A key challenge now is to relate the value of the flow to the number of infinite cycles.

### 2.6 Infinite cycles and classification of Gibbs measures.

Our analysis of infinite cycles requires introduction of an additional technical tool, called a cut. This is defined as follows:
Definition 2.12 Let $\sigma \in \Omega_X$ be a permutation with flow $F(\sigma) = n \geq 0$ and $a \in X^*$. Then $a$ is called a cut for $\sigma$ if $F_a^+(\sigma) = n$ (and so $F_a^-(\sigma) = 0$) and $\sigma$ contains the following $n$ jumps

$$a_{n-1} \rightarrow a_1, \ldots, a_{n-2} \rightarrow a_n.$$  

(2.23)

For $n < 0$ similar notions are defined by reversing the directions of the jumps.

The cuts are helpful for the following reason: If $a$ is a cut for $\sigma$, then $\sigma$ cannot have cycles jumping over $a$ apart from the $|n|$ infinite ones described in Lemma 2.2. In particular, we can make the following conclusions:

Lemma 2.13 Let $X, V$ and $n \in \mathbb{Z}$ satisfy (X1),(Xn) and (V1),(V2), and let $\mu \in G_n$.

(a) There are bi-infinitely many cuts for $\mu$-a.e. permutation.

(b) $\mu$-a.e. permutation has exactly $|n|$ infinite cycles.

A duplication argument yields a uniqueness result. For this we call $a \in X^*$ a cut for a pair of permutations $(\sigma, \sigma')$ if it is a cut for both $\sigma$ and $\sigma'$.

Lemma 2.14 Let $X, V$ and $n \in \mathbb{Z}$ satisfy (X1),(Xn) and (V1),(V2), and let $\mu, \mu' \in G_n$.

(a) There are bi-infinitely many cuts for $\mu \otimes \mu'$-a.e. pair of permutations.

(b) $\mu = \mu'$.

We conclude this section by collecting all of our previous findings and thus providing a complete description of the set of all Gibbs measures.

Theorem 2.15 Let $X, V$ and $n \in \mathbb{Z}$ satisfy (X1),(Xn) and (V1),(V2).

(a) $\mathcal{G}_n$ contains a single Gibbs measure $\mu_n$.

(b) $\mu_n$-a.e. permutation has exactly $|n|$ infinite cycles.

(c) For any $\eta \in \Omega$ with $F(\eta) = n$ and any sequence of increasing subsets $\Lambda \uparrow X$, $\mu_n$ is the weak limit of specifications $\gamma_\Lambda(\cdot | \eta)$.

In particular, if (Xn) is satisfied for every $n \in \mathbb{Z}$ (and (V1),(V2) hold), then

$$\mathcal{G} = \left\{ \sum_{n \in \mathbb{Z}} c_n \mu_n : c_n \geq 0, \sum_n c_n = 1 \right\}$$

(2.24)

We finish with a few remarks:

Remark 2.16 As already mentioned, our assumptions on $V$ are by far not optimal. In fact, we believe that a majority of our results — that is, with the exception of the characterization of ground states in Lemma 2.4 — carry over when $V : \mathbb{R} \to \mathbb{R}$ is perturbed by adding a continuous, even function that decays sufficiently fast to zero at infinity.

Remark 2.17 Our main theorem states that the specifications $\gamma_\Lambda(\cdot | \eta)$ have a weak limit along volumes increasing to $X$ for any boundary condition $\eta$ that has a finite flow. If $\eta$ has infinite flow, this is not necessarily the case. To illustrate different behaviors we give two examples, assuming that (V1) and (V2) are satisfied and that $X := \mathbb{Z}$ for simplicity.

Example 1: Let $\eta \in \Omega_{\mathbb{Z}}$ be defined by $\eta(0) := 0$ and $\eta(x) := x + 2p_x$, where $p_x$ is the maximal power of 2 dividing $|x|$. Writing $x := (2n+1)2^k$ for $x \neq 0$ and some $n \in \mathbb{Z}$ and $k \in \mathbb{N}$, we get $\eta(x) = (2n+3)2^k$ and so $\eta$ is a permutation. It is easy to check that $\eta$ has infinitely many cycles from $-\infty$ to $\infty$. Here an argument underlying the proof of Lemma 5.1 can be used to show that
for all \( x, y \in \mathbb{Z} \) we have \( \gamma_\Lambda(x \rightarrow y|\eta) \to 0 \) whenever \( \Lambda \uparrow \mathbb{Z} \). (Indeed, with increasing \( \Lambda \) there is an increasing number of jumps over \([x, y]\) with both endpoints in \( \Lambda \); any of these can be swapped with the jump \( x \rightarrow y \) while gaining energy.) So in this case no sequence of specifications (with \( \Lambda \uparrow \mathbb{Z} \)) is tight and no weak limits can be extracted.

**Example 2:** Let \( \eta \in \Omega_\mathbb{Z} \) be defined by \( \eta(x) := -x \) for all \( x \in \mathbb{Z} \). Then \( \eta \) has infinite flow, but no infinite cycles. Since the restriction of \( \gamma_\Lambda(\cdot|\eta) \) to \( \Lambda \cap (-\Lambda) \) is the same as \( \gamma_{\Lambda \cap (-\Lambda)}(\cdot|\tau_0) \), here we have \( \gamma_\Lambda(\cdot|\eta) \to \mu_0 \) weakly whenever \( \Lambda \uparrow \mathbb{X} \).

**Remark 2.18** The one-dimensional nature of the underlying set \( \mathbb{X} \) has been crucial for our reasoning, mainly due to the concept of the flow. Utilizing this concept we are able to show that for any given jump of sufficient length it is possible to find another jump such that the corresponding swap (see Definition 2.5) decreases the energy significantly. In two dimensions it is not clear how to define a useful quantity analogous to the flow, and indeed the idea of reducing the energy by a suitable swap fails: If a long jump is entirely surrounded by jumps of the same length and in the same direction, swapping this jump with another one may in fact increase the energy. As a consequence our technique does not even give existence of Gibbs measures in two or more dimensions.

### 3. Proofs: Preliminary Observations

We are now ready to commence the exposition of the proofs. Here we begin with some preliminary observations. Certain technical aspects that feed into the main arguments are then discussed in Section 4; the main results are proved in Sections 5 and 6. This section gives the proofs of Lemmas 2.1, 2.2, 2.4 and 2.8.

#### 3.1 Permutations and their flow

We begin by giving the proofs of the lemmas dealing with existential facts and properties of the flow of a permutation.

**Proof of Lemma 2.1.** Let \( \sigma \in \Omega_\mathbb{X} \). It suffices to show that \( F_a(\sigma) = F_b(\sigma) \) whenever \( a < b \in X^* \) with only one point \( x \in X \) in between. Then every jump except the one to and the one from \( x \) contributes to \( F_a^+(\sigma) \) and \( F_b^+(\sigma) \) exactly the same way (and likewise for \( F_a^-(\sigma) \) and \( F_b^-(\sigma) \)). This already implies that \( F_a(\sigma) = \infty \) iff \( F_b(\sigma) = \infty \). The jumps to and from \( x \) decompose into five cases: (1) \( \sigma(x) = x \), (2) \( \sigma^{-1}(x) < x < \sigma(x) \), (3) \( \sigma(x) < x < \sigma^{-1}(x) \), (4) \( x > \sigma^{-1}(x), \sigma(x) \) and (5) \( x < \sigma(x), \sigma^{-1}(x) \). In each of these it is easy to check that the contribution of the combined jumps to \( F_a(\sigma) \) is the same as the one to \( F_b(\sigma) \). Thus \( F_a(\sigma) = F_b(\sigma) \) as desired. \( \square \)

**Proof of Lemma 2.2.** Pick \( \sigma \in \Omega_\mathbb{X} \) with \( n := F(\sigma) \) finite and let \( x \in X \) be such that \( \{ \sigma^k(x) : k \in \mathbb{Z} \} \) is an infinite cycle. Since \( X \) is locally finite, and \( \sigma^k(x) \) are all distinct, \( \sigma^k(x) \) eventually leaves every bounded interval for both \( k \to \pm \infty \). Moreover, as \( F_a(\sigma) = n \) for any \( a \in X^* \), the sequence \( \sigma^k(x) \) can jump over \( a \) at most finitely many times and \( \sigma^k(x) - a \) changes sign only finitely often. Hence, the limits of \( \sigma^k(x) \) as \( k \to \pm \infty \) exist in \( \{ \pm \infty \} \). This rules out indeterminate infinite cycles.

In order to compare the number of infinite cycles of \( \sigma \) with \( n = F_a(\sigma) \), consider all cycles of \( \sigma \) and note that finite cycles, cycles from \( \infty \) to \( \infty \) and cycles from \( -\infty \) to \( -\infty \) do not contribute...
to $F_\sigma(\sigma)$ since they jump over $a$ equally (finitely) often to the left and to the right. Cycles from $-\infty$ to $\infty$ contribute $1$ to $F_\sigma(\sigma)$ since they jump to the right one more times than to the left. Similarly, cycles from $\infty$ to $-\infty$ contribute $-1$ to $F_\sigma(\sigma)$. It follows that there have to be at least $|n|$ infinite cycles of the same orientation as the sign of $n$. 

\section*{3.2 Ground states.}

Using the local perturbation from Definition 2.5 we easily get the following monotonicity result for local minima of the energy:

\begin{lemma}
Let $X$ be locally finite and $V$ be strictly convex. Let $\Lambda \subseteq X$ and $\sigma \in \Omega$. The minimum \(\min\{H_\Lambda(\tau) : \tau \sim_\Lambda \sigma\}\) is attained at the unique $\tau \sim_\Lambda \sigma$ such that $\tau : \Lambda \cap \sigma^{-1}(\Lambda) \to \Lambda \cap \sigma(\Lambda)$ is a (strictly) increasing function.
\end{lemma}

\begin{proof}
There are only a finite number of permutations $\tau$ with $\tau \sim_\Lambda \sigma$ and so the minimum of $H_\Lambda$ is attained. Moreover, any $\tau \sim_\Lambda \sigma$ maps $\Lambda \cap \sigma^{-1}(\Lambda) = \Lambda \cap \tau^{-1}(\Lambda)$ bijectively onto $\Lambda \cap \sigma(\Lambda)$ and it coincides with $\sigma$ elsewhere. If $\tau : \Lambda \cap \sigma^{-1}(\Lambda) \to \Lambda \cap \sigma(\Lambda)$ is not an increasing function, there are $x, y \in \Lambda \cap \tau^{-1}(\Lambda)$ such that $x < y$ and $\tau(x) > \tau(y)$. As $x, y, \tau(x), \tau(y) \in \Lambda$, we have $\tau_{xy} \sim_\Lambda \tau \sim_\Lambda \sigma$. We also note that

\begin{equation}
H_\Lambda(\tau) - H_\Lambda(\tau_{xy}) = V(\tau(x) - x) + V(\tau(y) - y) - V(\tau(x) - y) - V(\tau(y) - x) > 0,
\end{equation}

where the inequality follows from the strict convexity of $V$ and the assumed ordering of $x, y, \tau(x)$ and $\tau(y)$ — see Lemma 4.2. Thus $H_\Lambda$ does not take its minimum at a non-increasing $\tau$. The minimizing $\tau$ is unique because it is a bijection of $\Lambda \cap \sigma^{-1}(\Lambda)$ onto $\Lambda \cap \sigma(\Lambda)$, and is determined by $\sigma$ elsewhere.
\end{proof}

\begin{proof}[Proof of Lemma 2.4]
First, $\tau_n$ is a ground state by Lemma 3.1, since $\tau_n$ is increasing on $\Lambda \cap \tau_n^{-1}(\Lambda)$ for every $\Lambda \subseteq X$. On the other hand, if $\tau \neq \tau_n$ for all $n$, then $\tau$ is not increasing, i.e., there are $x, y$ such that $x < y$ and $\tau(x) > \tau(y)$. Considering $\Lambda \subseteq X$ such that $x, \tau(x), y, \tau(y) \in \Lambda$, Lemma 3.1 shows that $H_\Lambda(\tau)$ is not minimal and so $\tau$ is not a ground state.
\end{proof}

\section*{3.3 Point sets arising from point processes.}

Here we show that samples from point processes fulfilling the premises of Lemma 2.8 automatically satisfy the requirements (X1) and (Xn) for any $n \in \mathbb{Z}$. Recall that a point process $X$ on $\mathbb{R}$ is a random purely-atomic $\mathbb{N} \cup \{\infty\}$-valued Borel measure on $\mathbb{R}$. Assuming that compact sets receive finite mass almost surely, $X$ can be interpreted as a sum of unit point masses. If, furthermore, no points are degenerate a.s., which technically means that $X(\{x\}) \in \{0, 1\}$ for every $x$, then samples from $X$ can be identified with point sets.

\begin{proof}[Proof of Lemma 2.8]
Let us first prove (X1). Condition (1) ensures that $X$ is locally finite a.s. Since also $X(\mathbb{R}) = \infty$ a.s., $X$ contains infinitely many points a.s. We cannot have $X([0, \infty)) < \infty$ with a positive probability, because the shift invariance of the law of $X$ would also imply $X([-L, \infty)) \xrightarrow{\text{law}} X([0, \infty)) < \infty$ with the same probability for all $L \in \mathbb{N}$ and, taking $L \to \infty$, also $X(\mathbb{R}) < \infty$ with the same probability, in contradiction with (1). Similarly $X((-\infty, 0]) = \infty$, and we conclude that (X1) holds a.s.
Moving over to \((Xn)\), we fix \(n \in \mathbb{N}\), pick constants \(K_1, K_2\) and introduce the events
\[
A^{(1)}_k := \left\{ \sup_{\ell \geq 1} \mathcal{X}(k - \ell, k + \ell) \leq K_1 \right\} \quad \text{and} \quad A^{(2)}_k := \left\{ c_{\mathcal{X}}(k, n) \leq K_2 \right\}
\]
for \(k \in \mathbb{Z}\). We claim that \(K_1, K_2\) can be chosen so large that each of these events has probability in excess of \(\frac{1}{2}\). For the first we observe that \(\mathcal{X}(k - \ell, k + \ell) \to \mathbb{E}\mathcal{X}(k, k + 1)\) a.s. by the Birkhoff Pointwise Ergodic Theorem (using the additivity of \(\mathcal{X}\) and shift-ergodicity of the law of \(\mathcal{X}\)) and the limit is finite by assumption (1). For the second we note that a.s. there are \(n + 1\) points of \(\mathcal{X}\) to the right of \(k\) and \(n + 1\) points to the left of \(k\) since \(\mathcal{X}\) is bi-infinite, and together with the non-degeneracy of points from condition (2) this implies that the distances between the points of \(\Lambda(k, n)\) are positive and finite.

Now set \(A_k := A^{(1)}_k \cap A^{(2)}_k\) and note that \(A_k\) has positive probability. So, by ergodicity and the Birkhoff Theorem again, \(A_k\) occurs at bi-ininitely many — in fact, a positive density of — \(k\)’s almost surely. Suppose \(A_k\) occurs and set \(a_k := \frac{k_1 + k_2}{2}\) if \(k \notin \mathcal{X}\) and \(a_k := \frac{k_1 + k_0}{2}\) if \(k \in \mathcal{X}\) so that \(a_k \in \mathcal{X}^\ast\). We observe that \(c_{\mathcal{X}}(a_n, n) \leq K_2\) by containment in \(A_k\), so it suffices to bound \(c_{\mathcal{X}}(a_k)\).

Since \(|k - a_k| \leq \frac{K_2}{2}\) we have
\[
\#\{x \in \mathcal{X}: |x - a_k| \leq t\} \leq \#\{x \in \mathcal{X}: |x - k| \leq t + \frac{K_2}{2}\} \leq (t + \frac{K_2}{2} + 1)K_1
\]
using containment in \(A_k\). Thus for \(t \geq \frac{1}{\set{2}K_2}\) we can estimate
\[
\frac{1}{t} \#\{x \in \mathcal{X}: 0 < |x - a| \leq t\} \leq \frac{t + \frac{K_2}{2} + 1}{t}K_1 \leq 2K_2\left(\frac{1}{2K_2} + \frac{K_2}{2} + 1\right)K_1
\]
using monotonicity. For \(t < \frac{1}{\set{2}K_2}\) the expression on the left is 0 by the lower bound on the gap between successive points. This shows \(c_{\mathcal{X}}(a_k) \leq 2K_2\left(\frac{1}{2K_2} + \frac{K_2}{2} + 1\right)K_1\). These bounds apply at bi-ininitely many \(a \in \mathcal{X}^\ast\), so the condition \((Xn)\) is successfully verified.

\[\square\]

4. TECHNICAL ISSUES

In this section we collect some straightforward consequences of our assumptions on \(X\) and \(V\). Then we proceed to discuss how to obtain energy estimates and how these can be converted to bounds on probabilities of events.

4.1 Consequences of our assumptions.

As the set \(X\) will be always clear from the context, we will henceforth write \(c_s(a, n)\) for \(c_{\mathcal{X}}(a, n)\). We note that, for any discrete bi-infinite \(X \subset \mathbb{R}\), automatically \(c_s(a, n) < \infty\), since this quantity is determined by only finitely many non-zero and finite distances. We also observe that \((X1)\) and \((X2)\) imply that \(c_{\mathcal{X}}(a) < \infty\) for every \(a \in \mathbb{R}\).

**Lemma 4.1** If \(c_{\mathcal{X}}(a) < \infty\) and \(f: [0, \infty) \to [0, \infty)\) is non-increasing, then
\[
\sum_{x \in \mathcal{X}: x > a} f(x - a) \leq c_{\mathcal{X}}(a) \int_{0}^{\infty} \mathrm{d}s f(s),
\]
\[
\sum_{x \in \mathcal{X}: x < a} f(a - x) \leq c_{\mathcal{X}}(a) \int_{0}^{\infty} \mathrm{d}s f(s)
\]
Lemma 4.3

Many of the following results rely on energy estimates, comparing a permutation \( \sigma \) with jumps \( v \to w, y \to z \) to the swapped configuration \( \sigma_{vy} \). The following estimate will be helpful:

**Lemma 4.2** Suppose \( V : \mathbb{R} \to \mathbb{R} \) is strictly convex and symmetric. Then

\[
V \text{ is strictly increasing in } [0, \infty).
\]

In addition, for all \( t_1 < s_1 < s_2 < t_2 \in \mathbb{R} \) such that \( s_1 + s_2 = t_1 + t_2 \) we have

\[
V(t_1) + V(t_2) - V(s_1) - V(s_2) \geq 2 \frac{t_2 - s_2}{t_2 - t_1} \left( V(t_2) + V(t_1) - 2V\left( \frac{t_1 + t_2}{2} \right) \right)
\]

and thus, in particular,

\[
V(t_1) + V(t_2) - V(s_1) - V(s_2) > 0.
\]

**Proof.** The first conclusion is trivial. For (4.6), comparing the slopes of secant lines yields

\[
\frac{V(s_1) - V(t_1)}{s_1 - t_1} \leq \frac{V(\frac{s_1 + t_1}{2}) - V(t_1)}{\frac{s_1 + t_1}{2} - t_1} \leq \frac{V(t_2) - V(\frac{s_2 + t_2}{2})}{\frac{s_2 + t_2}{2} - t_2} \leq \frac{V(t_2) - V(s_2)}{t_2 - s_2}.
\]

The bound follows by estimating the difference of the outer terms against the difference of the inner terms and invoking \( t_2 - s_2 = s_1 - t_1 \). \( \square \)

If \( V \) satisfies (V2), it will be convenient to denote \( c_{\psi}(d,N) := \sup \{ x \in \mathbb{R} : \psi_d(x) < N \} \). Then

\[
\psi_d(x) \geq N \quad \text{for all } x \geq c_{\psi}(d,N).
\]

This relation will be quite useful in what follows.

### 4.2 Energy estimates.

Many of the following results rely on energy estimates, comparing a permutation \( \sigma \) with jumps \( v \to w, y \to z \) to the swapped configuration \( \sigma_{vy} \). The following estimate will be helpful:

**Lemma 4.3** Suppose \( V : \mathbb{R} \to \mathbb{R} \) is strictly convex and even. Let \( \Lambda \in X \) and \( v, w, y, z \in \Lambda \) and suppose that \( y', z' \in \mathbb{R} \) are such that \( v < y' \leq z' < w \), \( y' \leq y \) and \( z \leq z' \). If \( \sigma \in \Omega_X \) contains jumps \( v \to w, y \to z \), then

\[
H_\Lambda(\sigma) - H_\Lambda(\sigma_{vy}) \geq 2 \min \{ y' - v, w - z' \} \psi_{y'-y}(w - v) \log(w - v).
\]
Proof. Let $\Delta H := H_\Lambda(\sigma) - H_\Lambda(\sigma_{\eta})$ denote the energy difference to be estimated. Then
\[
\Delta H = V(w-v) + V(z-y) - V(z-v) - V(w-y) \\
\geq V(w-v) + V(z-y') - V(z-v) - V(w-y') \tag{4.11} \\
\geq V(w-v) + V(z'-y') - V(z'-v) - V(w-y'),
\]
where for the inequalities we used (4.7) along with $\sigma$ with $y$ second jump and then apply $V$. Since the terms in the large parentheses non-negative, we now estimate
\[
\Delta H \geq 2 \min\{y'-v, w-z'\} \left( V(w-v) + V(z'-y') - 2V\left(\frac{w-v+z'-y'}{2}\right) \right) \tag{4.12}
\]
and then apply $V(z'-y') \geq V(0)$, as implied by (4.5), to obtain (4.10).

We only have control over one jump $v \to w$ of a permutation $\sigma$, we need to find a suitable second jump $y \to z$ to make the above energy estimate work. This is achieved by:

Lemma 4.4 Let $a \in \mathbb{R}$ and let $\sigma \in \Omega_X$ with $|F(\sigma)| =: n \in \mathbb{N}$ contain a jump $(v,w)$ over $a_1, \ldots, a_n$ to the right. Then $\sigma$ contains a jump $(y,z)$ such that $y \geq a_1$ and $z \leq a_n$.

Proof. We consider the $n$ jumps starting from $a_1, \ldots, a_n$. If one of them does not jump over $b := \frac{1}{2}(a_n + a_{n+1}) \in X^*$, it is of the desired type. Otherwise all of them jump over $b$ to the right. Together with the jump $(v,w)$ this gives $F_+^b \geq n + 1$, and thus $F_{-}^b \geq 1$, i.e., there is a jump over $b$ to the left, as desired.

We note that, thanks to the reflection symmetry, versions of the above lemmas hold also for jumps to the left.

4.3 General facts from Gibbs measure theory.

Here we review some techniques from the theory of Gibbs measures. Most of these are well known; our aim is to have these presented in one block for easier later reference. We start with the general fact that, in the context of Gibbs measures, energy estimates yield probabilistic bounds:

Lemma 4.5 Let $\Lambda \in X$, $\eta \in \Omega$, $A,A'_1 \in \mathcal{F}$, $c \in \mathbb{R}$ and $N \geq 0$. Suppose that for every $\sigma \in A$ with $\sigma \sim_\Lambda \eta$, we can find a $\sigma' \in A'_1$ with $\sigma' \sim_\Lambda \eta$ such that $H_\Lambda(\sigma') - H_\Lambda(\sigma) \leq c$ and every $\sigma'$ can be attributed to at most $N$ distinct $\sigma$. Then
\[
\gamma_\Lambda(A|\eta) \leq Ne^c \gamma_\Lambda(A'|\eta). \tag{4.14}
\]
In particular, for $A'_1 := \Omega$ and $N := 1$, which requires $\sigma \mapsto \sigma'$ to be injective, we get $\gamma_\Lambda(A|\eta) \leq e^c$.

Proof. Using the two assumptions we get
\[
\sum_{\sigma \in A : \sigma \sim_\Lambda \eta} e^{-H_\Lambda(\sigma)} \leq e^c \sum_{\sigma \in A : \sigma \sim_\Lambda \eta} e^{-H_\Lambda(\sigma')} \leq Ne^c \sum_{\sigma' \in A'_1 : \sigma' \sim_\Lambda \eta} e^{-H_\Lambda(\sigma')}. \tag{4.15}
\]
Dividing both sides by $Z_\Lambda(\eta)$ gives a corresponding inequality for the specifications. □
An estimate of specification probabilities such as the one obtained in Lemma 4.5 implies a corresponding estimate of probabilities:

**Lemma 4.6** Let $A, A' \in \mathcal{F}$, $c > 0$ and suppose that for every $\eta \in \Omega$ we have $\gamma_\Lambda(A|\eta) \leq c \gamma_\Lambda(A'|\eta)$ for all sufficiently large $\Lambda \subseteq X$ (where “large” is allowed to depend on $\eta$). Then $\mu(A) \leq c \mu(A')$ for all $\mu \in \mathcal{G}$.

**Proof.** By the definition of Gibbs measure, for any increasing sequence of finite $\Lambda_m \uparrow X$ we have

$$c \mu(A') - \mu(A) = \lim_{m \to \infty} E_\mu(c \gamma_{\Lambda_m}(A'|\cdot) - \gamma_{\Lambda_m}(A|\cdot))$$

$$\geq E_\mu(\liminf_{m \to \infty} (c \gamma_{\Lambda_m}(A'|\cdot) - \gamma_{\Lambda_m}(A|\cdot))) \geq 0,$$

(4.16)

where we used Fatou’s Lemma to get the middle inequality. \qed

Our next observation is that Gibbs measures can be obtained as weak limits of specifications.

**Lemma 4.7** Let $\Lambda, \Lambda' \in X$. For every $\mathcal{F}_{\Lambda'}$-measurable function $f$, the expectation $\gamma_\Lambda(f|\eta)$ is a $\mathcal{F}_{\Lambda \cup \Lambda'}$-measurable, and thus continuous, function of $\eta$.

**Proof.** Let $\eta \in \Omega$. The partition function can be written as

$$Z_\Lambda(\eta) = \sum_{\sigma: \sigma \sim \Lambda|\eta} e^{-H_\Lambda(\sigma)} = \sum_{\sigma'} e^{-\sum_i V(\sigma'(x) - x)},$$

(4.17)

where the exterior sum extends over all bijections $\sigma': \Lambda \cap \eta^{-1}(\Lambda) \to \Lambda \cap \eta(\Lambda)$ and the interior sum is over all $x \in \Lambda \cap \eta^{-1}(\Lambda)$. Thus $Z_\Lambda(\eta)$ depends on $\eta$ only through $\eta(\Lambda)$ and $\eta^{-1}(\Lambda)$, and thus is $\mathcal{F}_{\Lambda}$-measurable. The measurability of $\gamma_\Lambda(f|\eta)$ follows similarly. \qed

**Lemma 4.8** Let $\eta \in \Omega$ and $\Lambda_m \subseteq X$ ($m \geq 1$) be such that $\gamma_{\Lambda_m}(\cdot|\eta)$ converges weakly, as $m \to \infty$, to some probability measure $\mu$. Then $\mu \in \mathcal{G}$.

**Proof.** Since the specifications are consistent, for every local event $A$ and $\Lambda \subseteq X$ we have

$$E_\mu \gamma_\Lambda(A|\cdot) = \lim_{m \to \infty} E_{\gamma_{\Lambda_m}(\cdot|\eta)} \gamma_\Lambda(A|\cdot) = \lim_{m \to \infty} \gamma_{\Lambda_m}(A|\eta) = \mu(A)$$

(4.18)

since both $1_A$ and $\gamma_\Lambda(A|\cdot)$ are local and thus continuous functions. This proves equality of $\mu$ with the measure on the extreme left for $A$ local; an extension to general events is unique thanks to, e.g., the $\pi$-$\lambda$ theorem. \qed

5. PROOFS: EXISTENCE

In this section we address a.s. finiteness of the flow in all Gibbs measures, non-existence of Gibbs measures in the absence of interactions and tightness of the family of specifications leading to the proof of existence of Gibbs measures. In particular, we provide formal proofs of Lemmas 2.9, 2.10 and 2.11. We assume throughout that $X$ and $V$ satisfy (X1), (V1) and (V2); all other assumptions will be mentioned explicitly whenever needed.
5.1 Finiteness of the flow.

We begin by proving Lemma 2.11. First we dismiss the issue of measurability:

**Proof of Lemma 2.11, tail measurability.** In order to show that $F$ is $\mathcal{T}$-measurable, let $a \in X^*$ and $\Lambda_a := \{x \in X : x > a\}$. Based on its definition, $F^+_{a}$ is measurable with respect to $\sigma(\{P^+_{x} : x \in \Lambda_a\})$ and $F^-_{a}$ is measurable with respect to $\sigma(\{P^-_{x} : x \in \Lambda_a\})$. Since $F$ is, modulo a proviso when both $F^+_{a} = \infty$, the difference $F^+_{a} - F^-_{a}$, it follows that also $F$ is $\mathcal{T}_{\Lambda_a}$-measurable. Since this holds for all $a$, we get that $F$ is $\mathcal{T}$-measurable. \qed

To show that $F$ is finite $\mu$-a.s. for every Gibbs measure, we have to prove that long jumps are unlikely. This follows from a suitable energy estimate for permutations having two nested jumps.

**Lemma 5.1** Let $x, y \in X$ obey $x \leq y$ and let $\mu \in \mathcal{G}$. Then there is an $l_0 = l_0(y-x) \geq 0$ such that

$$\mu(x \rightarrow y, v \rightarrow w) \leq \frac{1}{|w-v|^5}$$

(5.1)

holds for all $v, w \in X$ that obey $v \leq x - 1$, $w \geq y + 1$ and $w - v \geq l_0$.

**Proof.** For $x, y, v, w$ as above, set $l_0 := c_\psi(y-x, 3)$ where $c_\psi$ is as in (4.9). Let $\eta \in \Omega_X$ and pick $\Lambda \subseteq X$ such that $x, y, v, w \in \Lambda$. Let $\sigma \sim_\Lambda \eta$ be such that $\sigma(x) = y$ and $\sigma(v) = w$. By our choice of $\Lambda$ we have $\sigma_{\Lambda} \sim_\Lambda \sigma$ and the energy estimate (4.10) implies

$$H_\Lambda(\sigma) - H_\Lambda(\sigma_{\Lambda}) \geq 2 \min\{x - v, w - y\} \psi_{y-x}(w-v) \log(w-v) \geq 5 \log(w-v)$$

(5.2)

thanks to our choice of $l_0$. We also note that $\sigma_{\Lambda}$ uniquely determines $\sigma$ for given $x, y, v, w$. By Lemma 4.5 we thus get

$$\gamma_\Lambda(x \rightarrow y, v \rightarrow w|\eta) \leq \frac{1}{|w-v|^5}.$$  

The conclusion follows by integrating with respect to $\mu$. \qed

**Lemma 5.2** Let $a \in X^*$ satisfy $c_g(a) < \infty$. Then every $\mu \in \mathcal{G}$ obeys $\mu(F^+_{a} = \infty) = 0$.

**Proof.** Any $\sigma \in \Omega_X$ such that $F^+_{a}(\sigma) = \infty$ contains a jump $(x, y)$ with $x < a < y$. Fix $l_0 \geq 0$ as in the previous lemma. In addition, for any $l > 0$, $\sigma$ also contains a jump $(v, w)$ such that $v \leq x - 1$, $w \geq y + 1$ and $w - v \geq l, l_0$. (This is because only finitely many jumps can start or end in a given finite set.) A union bound and Lemma 5.1 then give

$$\mu(F^+_{a} = \infty) \leq \sum_{x, y, v, w \in X \atop v+1 \leq x < a \leq y-1; \atop w-v \geq l, l_0} \mu(x \rightarrow y, v \rightarrow w) \leq \sum_{x, y, v, w \in X \atop v+1 \leq x < a \leq y-1; \atop w-v \geq l, l_0} \frac{1}{|w-v|^5}.$$  

(5.4)

We first sum over $x$ and $y$ using $\#\{x \in X : v < x < a\} \leq c_g(a)(a-v) \leq c_g(a)(w-v)$ and similarly for $\#\{y \in X : a < y < w\}$. In order to sum over $v$ and $w$ we use (4.2) with the result

$$\mu(F^+_{a} = \infty) \leq \sum_{v < a < w; \atop w-v \geq l} \frac{c_g(a)^2}{|w-v|^3} \leq \sum_{v < a < w} \frac{c_g(a)^2}{\max\{w-v, l\}^3}$$

$$\leq c_g(a)^4 \int_0^\infty ds \frac{s}{\max\{s, l\}^3} = \frac{3c_g(a)^4}{2l}.$$  

(5.5)

As $l$ can be arbitrarily large, we are done. \qed
Proof of Lemma 2.11, a.s. finiteness. Let $\mu \in \mathcal{G}$. Since (V1), (V2) and (X1), (X2) are assumed, all previous derivations are at our disposal. The previous lemma gave $\mu(F_a^+ = \infty) = 0$ and, thanks to obvious symmetry considerations and the fact that $c_\mathcal{G}(a)$ is defined symmetrically, we similarly have $\mu(F_a^- = \infty) = 0$. Therefore, the flow is finite $\mu$-a.s., as desired. \hfill \Box

5.2 No Gibbs measures without interaction.

Here we address the fact that, in the absence of “interactions” the set of Gibbs measures is empty.

Proof of Lemma 2.9. Assume $V := 0$ and suppose that there is some $\mu \in \mathcal{G}$. Fix $x, y \in X$, let $N \in \mathbb{N}$ and pick $\eta \in \Omega$. Let $\Lambda_N \subseteq X$ be such that $x, y \in \Lambda_N$ and $|\Lambda_N| \geq N$. For any finite $\Lambda \subseteq X$ that obeys $\Lambda_N \cup \eta(\Lambda_N) \cup \eta^{-1}(\Lambda_N) \subseteq \Lambda$ we observe that, since $\gamma_\Lambda(\cdot|\eta)$ is the uniform distribution on all $\sigma \sim_\Lambda \eta$ and $\sigma(x)$ can be any point of $\Lambda_N$ with equal probability,

$$
\gamma_\Lambda(x \rightarrow y|\eta) = \frac{1}{|\Lambda_N|} \leq \frac{1}{N}. \tag{5.6}
$$

By Lemma 4.6 this implies $\mu(x \rightarrow y) \leq \frac{1}{N}$ and, since $N$ was arbitrary $\mu(x \rightarrow y) = 0$ for all $y \in X$. As this cannot hold, we have $\mathcal{G} = \emptyset$ after all. \hfill \Box

5.3 Existence of Gibbs measures.

Now assuming (X2) in addition to conditions (X1), (V1) and (V2), we proceed to establish Lemma 2.10 dealing with existence of Gibbs measures. Not unexpectedly, this will be done by proving tightness for sequences of specifications over increasing volumes. We begin by deriving an estimate on the probability of a long jump.

Lemma 5.3 Let $\eta \in \Omega_X$ be such that $|F(\eta)| =: n \in \mathbb{N}$, and fix $x \in X$. There are $\Lambda_0 \subseteq X$ and $l_0 \in \mathbb{R}$ (both depending on $n, x$ only) such that for all $\Lambda$ with $\Lambda_0 \subseteq \Lambda \subseteq X$ and all $v, w \in X$ with $v \leq x \leq w$ and $w - v \geq l_0$ we have

$$
\gamma_\Lambda(v \leftrightarrow w|\eta) \leq \frac{2}{|w - v|^3}. \tag{5.7}
$$

Proof. Let $\Lambda_0 \subseteq X$ be so large that it contains all jumps of $\eta$ to, from and over any point of the set $\{x_i: -n \leq i \leq n\}$. This is possible because $\eta$ has only finitely many jumps over these points, in light of finiteness of $F(\eta)$. Define, assuming (V1) and (V2),

$$
l_0 := \max \{x_n - x_{-n} + 1, c_\psi(x_n - x_{-n}, 2c_\psi(x, n))\}, \tag{5.8}
$$

and suppose that $v, w \in X$ are such that $v \leq x \leq w$ and $w - v \geq l_0$. Note that if $\Lambda$ obeys $\Lambda_0 \subseteq \Lambda \subseteq X$ and if $\sigma \sim_\Lambda \eta$ is such that $\sigma(v) = w$, the definition of $l_0$ implies $w > x_n$ or $v < x_{-n}$. We will only consider the first case (the other one can be done similarly).

By Lemma 4.4 we can choose $y, z \in X$ such that $\sigma(y) = z$ and $y \geq x_1$ and $z \leq x_n$ hold. Then $y, z, v, w \in \Lambda_0$ and so $y, z, v, w \in \Lambda$. It follows that $\sigma_{xy} \sim_\Lambda \sigma \sim_\Lambda \eta$ and we can use the energy estimate (4.10) from Lemma 4.3 to obtain

$$
H(\sigma) - H(\sigma_{xy}) \geq 2 \min \{x_1 - v, w - x_n\} \psi_{x_n - x_1}(w - v) \log(w - v)
$$

\[\geq \frac{2}{c_\psi(x, n)} \psi_{x_n - x_{-n}}(w - v) \log(w - v) \geq 3 \log(w - v). \tag{5.9}\]
We also note that \( \sigma_{xy} \) uniquely determines \( \sigma \) for given \( v, w \) (since \( \sigma_{xy}^{-1}(w) = y \)). By Lemma 4.5 we thus get \( \gamma_x(v \to w|\eta) \leq \frac{1}{|w-v|^3} \), as desired, and by symmetry we get the same estimate for jumps \( w \to v \) to the left.

Lemma 5.4 Suppose \( \eta \in \Omega_X \) obeys \(|F(\eta)| < \infty \) and let \( \Lambda_m \subseteq X \) be sets such that \( \Lambda_m \uparrow X \) for \( m \to \infty \). Then \( \{\gamma_{\Lambda_m}(\cdot|\eta)\}_{m \geq 1} \) is tight.

Proof. Let \( \epsilon > 0 \) and choose \( a(x) > 0 \) such that \( \sum_{x \in X} a(x) \leq 1 \). For \( x \in X \) let

\[
K_f(x) := \{ \sigma \in \Omega_X : |x - \sigma(x)|, |x - \sigma^{-1}(x)| \leq f(x) \},
\]

(5.10)

where \( f(x) \geq 0 \) is chosen so that the following holds true: First we need \( f(x) \geq l_0 \), where \( l_0 = l_0(n, x) \) with \( n := |F(\eta)| \) are as in previous lemma. Next we will require that

\[
f(x) \geq \sqrt{\frac{6c_g(x)}{\varepsilon a(x)}}
\]

(5.11)

and also

\[
\gamma_{\Lambda_m}(K_f(x)^c|\eta) \leq \varepsilon a(x)
\]

(5.12)

for every \( m \) such that \( \Lambda_m \not\supset \Lambda_0 \), where \( \Lambda_0 = \Lambda_0(x, \eta) \) is from the previous lemma. The latter is possible since there are only finitely many such \( \Lambda_m \) and \( K_f(x) \uparrow \Omega_X \) for \( f(x) \to \infty \). For \( m \) such that \( \Lambda_m \supset \Lambda_0(x, \eta) \) and \( w \geq x \) such that \( w - x \leq f(x) \) we can use Lemma 5.3 to estimate \( \gamma_{\Lambda_m}(x \leftrightarrow w|\eta) \leq \frac{2}{|w-v|^3} \), and using (4.1) gives

\[
\sum_{w > x + f(x)} \left( \frac{2}{w-x} \right)^3 \leq \sum_{w > x} \left( \frac{2}{\max\{w-x, f(x)\}} \right)^3 \leq \int_0^{\infty} ds \frac{2c_g(x)}{\max\{s, f(x)\}^3} = \frac{3c_g(x)}{f(x)^2}.
\]

(5.13)

By symmetry we have the same estimate for jumps from and to \( x \) from the left. Due to the choice of \( f(x) \) and the definition of \( K_f(x) \), the bound (5.12) thus holds for all \( m \geq 1 \).

Now set \( K_f := \bigcap_{x \in X} K_f(x) \) and note that

\[
\gamma_{\Lambda_m}(K_f^c|\eta) \leq \sum_x \gamma_{\Lambda_m}(K_f(x)^c|\eta) \leq \sum_x \varepsilon a(x) \leq \varepsilon
\]

(5.14)

for all \( m \geq 1 \). Since \( K_f \) is compact, which can be seen by a diagonal argument using the completeness of \( \Omega_X \), we obtain tightness for the sequence \( \{\gamma_{\Lambda_m}(\cdot|\eta)\}_{m \geq 1} \).

Our last item of concern is whether or not subsequential limits of specifications are Gibbs measures with the correct value of the flow.

Lemma 5.5 Let \( \eta \in \Omega_X \) such that \( n := F(\eta) \in \mathbb{Z} \). Let \( \Lambda_m \subseteq X \) be sets such that \( \Lambda_m \uparrow X \) and such that \( \gamma_{\Lambda_m}(\cdot|\eta) \) converges, as \( m \to \infty \), weakly to a probability measure \( \mu \). Then \( \mu \in \mathcal{G}_n \).

Proof. Lemma 4.8 ensures \( \mu \in \mathcal{G} \) so we only have to show that \( \mu(F_a = n) = 1 \) for some \( a \in X^* \). For this it is convenient to define a localized version of the flow: For a given \( l \geq 0 \) let \( F^{l \pm}_a(\sigma) \) be the analogues of the quantities \( F^{l \pm}_a(\sigma) \), respectively, but counting only jumps over \( a \) of length at most \( l \). (The specific choice of “length” is not important here, one can, e.g., use Euclidean distance.) Define \( F^+_a(\sigma) := F_{a^+}(\sigma) - F_{a^-}(\sigma) \), where no provisos are necessary because all
quantities are finite. Although $F^l_a(\sigma)$ may, unlike $F_a(\sigma)$, depend on $a$, we have
\[ |F_a(\sigma)| < \infty \implies F^l_a(\sigma) \xrightarrow{l \to \infty} F_a(\sigma) \]  
(5.15)
This is because the finiteness of $F_a(\sigma)$ implies that all jumps over $a$ have a bounded length (depending only on $\sigma$).

Returning to the main line of the proof, we note that
\[ |\mu(F_a = n) - 1| = |\mu(F_a = n) - \gamma_m(F_a = n|\eta)| \]
\[ \leq \mu(F_a \neq F^l_a) + |\mu(F^l_a = n) - \gamma_m(F^l_a = n|\eta)| + \gamma_m(F^l_a \neq F_a|\eta). \]  
(5.16)
If $F_a(\sigma) \neq F^l_a(\sigma)$, then $\sigma$ contains a jump to, from or over $x := a_k$ of length $\geq l$, so if $m$ and $l$ are sufficiently large (depending on $\eta, a$ only), Lemma 5.3 and (4.1) gives
\[ \gamma_m(F_a \neq F^l_a|\eta) \leq \sum_{v < w < l: w - v \geq 1} \gamma_m(v \leftrightarrow w|\eta) \leq \sum_{v < w < l: w - v \geq 1} \frac{2}{(w - v)^3} \]
\[ \leq \frac{2}{\max\{l, w - v\}^3} \leq \int_{0}^{\infty} ds \frac{2c_g(a)^2s}{\max\{l, s\}^3} = \frac{3c_g(a)^2}{l}. \]  
(5.17)
Since $\{F^l_a = n\}$ is a local event and thus its indicator is a continuous function, we definitely have $\gamma_m(F^l_a = n|\eta) \to \mu(F^l_a = n)$ as $m \to \infty$, so letting $m \to \infty$ in (5.16) yields
\[ |\mu(F_a = n) - 1| \leq \mu(F_a \neq F^l_a) + \frac{3c_g(a)^2}{l}, \]  
(5.18)
once $l$ is sufficiently large. Since $\mu \in \mathcal{G}$, Lemma 2.11 implies $|F_a| < \infty \mu$-a.s. and so, by (5.15), also $F_a \to F_a \mu$-a.s. as $l \to \infty$. In particular, $\mu(F_a \neq F^l_a) \to 0$ in this limit as well and so the claim follows by taking $l \to \infty$. \[ \square \]

**Proof of Lemma 2.10.** This is now a direct consequence of the three preceding lemmas. \[ \square \]

### 6. PROOFS: INFINITE CYCLES AND UNIQUENESS

In this final section we develop the desired level of control over the number of infinite cycles in permutations sampled from a Gibbs measure. The key notion is that of a cut, introduced in Definition 2.12. Cuts will allow us to give full classification of all Gibbs measures leading to the proof of Theorem 2.15. Naturally, we will also provide formal proofs of Lemmas 2.13 and 2.14. Throughout this section we assume the validity of (V1), (V2), (X1) and (Xn), where $n$ will be clear from context.

#### 6.1 Pre-cuts.

As already mentioned, cuts permit us to control the exact number of infinite cycles crossing over a given point of $X^*$. In order to exercise this control throughout $X^*$, we need to identify a bi-infinite sequence of cuts in a.e. permutation. This will be achieved by relaxing to the notion of $k$-pre-cut, defined as follows:

**Definition 6.1** For $k \in \mathbb{N}$, $\sigma \in \Omega_X$ and $n := F(\sigma) \in \mathbb{Z}$, the point $a \in X^*$ is called a $k$-pre-cut for $\sigma$ if all jumps of $\sigma$ over $a$, from $a_{-n}, \ldots, a_{-1}$ or to $a_1, \ldots, a_n$ are completely contained in $\{a_{-k}, \ldots, a_k\}$. 

The role of \( k \)-pre-cuts is that of candidates for \( k \)-cuts, since any \( k \)-pre-cut \( a \) can be made into a cut by modifying the permutation locally near \( a \).

In this section we prepare the proof of Lemma 2.13 by showing that, in a.e. permutation, there are sufficiently many \( k \)-pre-cuts. Let \( n \geq 0 \) be fixed throughout this section, and let \( c_n \) be a sequence sufficiently large that \( c_n \geq n+1 \) and such that

\[
X_n^* := \{ a \in X^* : c_k(a,n) \leq c_n, c_k(a) \leq c_n \},
\]

is bi-infinite. The latter is possible in light of condition (\( X_n \)). Let \( k \) be the smallest natural number such that

\[
k \geq \max \{ n+1, c_n c_\psi(nc_n,2c_n), c_n c_\psi(0,2c_n) + nc_n^2, 24c_n^3, 33nc_n^2 + 1 \}. \tag{6.2}
\]

We first consider possible configurations in which a point \( a \in X^* \) is not a \( k \)-pre-cut. For this we call a jump \( v \rightarrow w \) (with \( v, w \in X \), of course) \( a \)-relevant, if it is a jump over \( a \) that is going either from \( a_{-\eta}, \ldots, a_{-1} \) or to \( a_1, \ldots, a_n \).

**Lemma 6.2** Let \( \mu \in \mathcal{M}_n \), \( a \in X_n^* \) and \( k \) as above. For \( \Delta \subset X \) let \( C(\Delta,a) \) be the event that all \( a \)-relevant jumps are contained in \( \Delta \) and let \( B \in \mathcal{F}_X \). Then for all \( a \)-relevant jumps \( v \rightarrow w \) (\( v, w \in X \)) that are not contained in \( \{ a_{-k}, \ldots, a_k \} \) we have

\[
\mu(C(\Delta,a) \cap B \cap \{ v \rightarrow w \}) \leq \frac{1}{|w-v|^3} \mu(B). \tag{6.3}
\]

The proof is easier to present when we deal separately with the case when the said jump \( v \rightarrow w \) is to the right and to the left.

**Proof of Lemma 6.2**, the case \( w > v \). Here we suppose that \( w > v \) which means that \( v \rightarrow w \) is a jump to the right. In this case, either \( v < a < a_k < w \) or \( v < a_{-k} < a < w \). From these two possibilities we will address only the former since the latter is quite analogous.

Let \( \eta \in \Omega_X \) be such that \( F(\eta) = n \). Let \( \Lambda_0 \subset X \) be so large that it contains all jumps of \( \eta \) over \( a \) or to \( a_1, \ldots, a_n \) (which is possible since \( F(\eta) \) is finite), and let \( \Lambda \subset X \) be such that \( \Lambda_0 \subset \Lambda \). Suppose \( \sigma \sim_\Lambda \eta \) is such that \( \sigma(v) = w \) and \( \sigma \in C(\Delta,a) \cap B \). By Lemma 4.4 we can find a jump \( y \rightarrow z \) such that \( y \geq a_1, z \leq a_n \). Thanks to the containment \( \Lambda_0 \subset \Lambda \) we have \( v, w, y, z \in \Lambda \), so \( \sigma_y \sim_\Lambda \sigma \sim_\Lambda \eta \) and the energy estimate from Lemma 4.3 (with \( y' := a_1 \) and \( z' := a_n \)) gives

\[
H_\Lambda(\sigma) - H_\Lambda(\sigma_y) \geq 2 \min\{a_1 - v, w - a_n\} \psi_{a_n-a_1}(w-v) \log(w-v) \\
\geq \frac{2}{c_n} \psi_{nc_n}(w-v) \log(w-v) \geq 3 \log(w-v). \tag{6.4}
\]

Here in the second bound we used that \( a \in X_n^* \) in order to estimate minimal and maximal distances against \( c_n \) and then applied the natural monotonicity of \( d \rightarrow \psi_d(x) \). The last inequality holds because \( w - v \geq a_k - a \geq \frac{k}{c_n} \geq c \psi(nc_n,2c_n) \).

Note that for \( \sigma \in C(\Delta,a) \) we have \( v, w, y, z \in \Delta \), so \( \sigma \in B \) implies \( \sigma_y \in B \). The swapped permutation \( \sigma_y \) uniquely determines \( \sigma \) for given \( v, w \) (since \( \sigma_y^{-1}(w) = y \)). By Lemma 4.5 we thus get

\[
\gamma_\Lambda(C(\Delta,a) \cap B \cap \{ v \rightarrow w \} | \eta) \leq \frac{1}{|w-v|^3} \gamma_\Lambda(B | \eta). \tag{6.5}
\]

As this holds for all \( \Lambda \) large, Lemma 4.6 implies the desired estimate. \( \square \)
Proof of Lemma 6.2, the case $v > w$. Now suppose that $v \rightarrow w$ is a jump to the left, i.e., either $w \leq a_n < a_k < v$ or $w < a_{k-1} < a_n \leq v$. We will henceforth assume the former as the latter can be dealt with similarly.

Let $\eta \in \Omega_X$ such that $F(\eta) = n$. Let $b := \frac{1}{2}(a_n + a_{n+1}) \in X^*$ and let $\Lambda_0 \subseteq X$ be so large that it contains all jumps of $\eta$ over $b$ (which is possible since $F(\eta)$ is finite), and let again $\Lambda \subseteq X$ obey $\Lambda_0 \subseteq \Lambda$. Let $\sigma \sim_\Lambda \eta$ such that $\sigma(v) = w$ and $\sigma \in C(\Delta, a) \cap B$. Since $F_b(\eta) = n \geq 0$ and $(v,w)$ is a jump over $b$ to the left, there are at least $n+1$ jumps over $b$ to the right, so one of them, say $(y,z)$ satisfies $y \leq a$ and $z \geq a_{n+1}$. By our choice of $\Lambda$ we again have $v,w,y,z \in \Lambda$, so $\sigma_{vy} \sim_\Lambda \sigma \sim_\Lambda \eta$ and the energy estimate (4.10) with reversed directions gives

$$H_\Lambda(\sigma) - H_\Lambda(\sigma_{vy}) \geq 2 \min\{v-a, a_{n+1} - w\} \psi_{a-a_{n+1}}(v-w) \log(v-w) \geq 3 \log|w-v|. \quad (6.6)$$

Here we have estimated the minimal distance by $\frac{1}{c_n}$ and used

$$v-w \geq a_k-a_n \geq (a_k-a) - (a_n-a) \geq \frac{k}{c_n} - nc_n \geq c_y(0,2c_n), \quad (6.7)$$

which implies $\psi_{a-a_{n+1}}(v-w) \geq \psi_0(v-w) \geq 2c_n$. All estimates involving $c_n$ use that $a \in X^*_n$.

Thanks to the choice of $\sigma \in C(\Delta,a)$ we have $z,y,v,w \in \Delta$, so $\sigma \in B$ implies $\sigma_{vy} \in B$. Moreover, $\sigma_{vy}$ uniquely determines $\sigma$ for given $v,w$ (since $\sigma_{vy}^{-1}(w) = y$). As above, a combination of Lemmas 4.5 and 4.6 then implies the desired estimate. \(\square\)

We can now move on to the main conclusion of this subsection:

**Lemma 6.3** Let $\mu \in \mathcal{G}_n$ and let $Y_n \subset X^*_n$ be bi-infinite. Then $Y_n$ contains bi-ininitely many $k$-pre-cuts $\mu$-a.s.

**Proof.** The proof proceeds by a sort of renewal argument: We examine a subsequence of points from $Y_n$ in an ordered fashion and note that the probability of not seeing a $k$-pre-cut in the first $m$ of them decays exponentially with $m$. We will have to do this relative to any position (marked by an integer $N$) and for any number of consecutive points (marked by a natural $M$).

Let us now proceed with details. Fix any $N \in \mathbb{Z}$, let $M \in \mathbb{N}$ and consider points $a_1,\ldots,a_M \in Y_n$ be such that $a_1 \geq N$ and

$$(a_m)_{-n} - (a_{m-1})_{n+k} + l(m) := c_n^2 2^{m+4} \quad \text{for all } 1 < m \leq M. \quad (6.8)$$

Let $A_m$ denote the event that $a_m$ is not a $k$-pre-cut, set $B_m := \bigcap_{i \leq m} A_i$ (with $B_0 := \Omega_X$) and write $C_m$ for the set of all configurations containing an $a_m$-relevant jump that is not contained in $\Delta_{m-1} := \{(a_m-1)_i: i > n\}$ (with $\Delta_0 := X$). Our goal is to derive an inductive bound on $\mu(B_m)$.

We will begin by deriving a bound on $\mu(C_m)$. Fix some $m$ with $1 \leq m \leq M$ for the time being and simplify notation by setting $a := a_m$ and $a' := a_{m-1}$. We note that

$$\mu(C_m) \leq \sum_{v < a, w > a} \mu(v \rightarrow w) + \sum_{v > a, w < a'} \mu(v \rightarrow w) \leq \sum_{v < a, w > l(m)} \frac{2}{|w-v|^3} + \sum_{l \leq n, w < a'} \frac{1}{|w-a'|^3}, \quad (6.9)$$

where we have estimated each probability using Lemma 6.2 with $\Delta := X$ and $B := \Omega_X$ noting that each jump considered is $a$-relevant and not contained in $\{a_{-k},\ldots,a_k\}$. In the last display the first
sum can be estimated using (4.1) and \( a \in X_n^* \) by

\[
\sum_{v < a < w: |w - v| > l(m)} \frac{2}{w - v} \leq \int_0^\infty \frac{c_n^2 2s}{\max \{s, l(m)\}^3} ds \leq \frac{3c_n^2}{l(m)} \leq \frac{1}{2^{m+2}}.
\] (6.10)

Similarly the second sum can be estimated by

\[
\sum_{w < a < B} \frac{1}{(a - n - w)^3} \leq \sum_{w < a} \frac{n}{\max \{a - nc_n - w, l(m)\}^3} \leq \int_0^\infty \frac{nc_n}{\max \{s - nc_n, l(m)\}^3} ds
\]

\[
= nc_n \left( \frac{1}{l(m)^3} + \frac{1}{2l(m)^2} \right) \leq \frac{2nc_n}{l(m)^2} \leq \frac{1}{2^{m+2}}.
\] (6.11)

Combining the above estimates gives \( \mu(C_m) \leq \frac{1}{2^{m+2}} \).

Moving over to an inductive bound on \( \mu(B_m) \), here we employ \( B_m \subseteq C_m \cup (B_m \cap C_m) \). The probability of the first event has just been estimated, the second we further decompose as

\[
B_m \cap C_m \subseteq B_{m-1} \cap C(\Delta_{m-1}, a_m) \cap \bigcup \{ v \rightarrow w \},
\] (6.12)

where the union is taken over all \( a \)-relevant jumps \( v \rightarrow w \) that not contained in \( \{a_{-k}, \ldots, a_k\} \).

Since \( B_{m-1} \subseteq \mathcal{F}_{X_{m-1}} \), Lemma 6.2 implies

\[
\mu(B_m \cap C_m) \leq \mu(B_{m-1} \cap C(\Delta, a_m) \cap \{ v \rightarrow w \}) \leq \mu(B_{m-1}) \sum_{w < a < B} \frac{1}{(a - n - w)^3},
\] (6.13)

so it remains to bound the sum on the right (which is over pairs \( v \rightarrow w \) as specified above).

Every jump \( v \rightarrow w \) considered is either a jump over \( a \) of length \( |w - v| \geq |a_{+k} - a| \geq \frac{k}{c_n} \) or a jump from \( v > a_k \) to \( w \in \{a_1, \ldots, a_n\} \) of length \( v - w = v - a - (w - a) \geq \frac{k}{c_n} - nc_n \) or a jump from \( v \in \{a_{-1}, \ldots, a_{-k}\} \) to \( w < a_{-k} \) of length \( v - w \geq v - a - (w - a) \geq \frac{k}{c_n} - nc_n \). The corresponding contributions to the above sum can be estimated using (4.1). For the first case

\[
\sum_{v < a < w: |v - w| \geq \frac{k}{c_n}} \frac{1}{w - v} \leq \frac{3c_n^2}{k/c_n} \leq \frac{1}{8},
\] (6.14)

thanks to our choice of \( k \). For the second case (and similarly for the third case)

\[
\sum_{a < w < d_k, l > d_k} \frac{1}{w - v} \leq n \sum_{v > d_k} \frac{1}{v - a - nc_n} \leq nc_n \int_0^\infty \frac{1}{(\max\{s, \frac{k}{c_n}\} - nc_n)^3} ds
\]

\[
= nc_n \left( \frac{k}{c_n} + \frac{\frac{k}{c_n} - nc_n}{2(\frac{k}{c_n} - nc_n)^2} \right) \leq nc_n \frac{2}{\frac{k}{c_n} - nc_n} \leq \frac{1}{16},
\] (6.15)

again by our choice of \( k \).

Combining the above bounds we get \( \mu(B_m \cap C_m) \leq \frac{1}{4} \mu(B_{m-1}) \) and thus

\[
\mu(B_m) \leq \mu(C_m) + \mu(B_m \cap C_m) \leq \frac{1}{4} \mu(B_{m-1}) + \frac{1}{2^{m+1}},
\] (6.16)

for arbitrary \( m \geq 1 \). By induction, this implies \( \mu(B_M) \leq \frac{1}{M} \). Since \( M \geq 1 \) was arbitrary, the probability that there is no \( k \)-pre-cut right of \( N \) is zero. But \( N \) was arbitrary (integer) too so, as a moment’s thought shows, there are infinitely many \( k \)-pre-cuts in positive direction \( \mu \)-a.s. A completely analogous argument proves the same for negative direction as well. □
6.2 Cuts.

We now turn from pre-cuts to cuts. The way things are set up, every pre-cut has a chance to be a cut with a uniformly positive probability. This implies that a positive proportion of pre-cuts are actually cuts, although for us it will be enough to show that the set of cuts is bi-infinite. Throughout we keep \( n \geq 0 \) and \( k \) exactly as in the previous section.

Lemma 6.4 Let \( \mu \in \mathcal{G}_n \). Then \( X^*_n \) contains bi-infinitely many cuts \( \mu \)-a.s.

**Proof.** Let \( N \in \mathbb{Z} \) and let \( Y_n(N) \) be a an infinite subset of \( \{ x \in X^*_n : x \geq N \} \) such that between any two successive points of \( Y_n(N) \) there are at least \( 2k \) points of \( X \). Let \( a_1 < a_2 < ... \) be an enumeration of \( Y_n(N) \). For \( M \in \mathbb{N} \) and \( I \subset \mathbb{N} \) with \( |I| = M \), let \( A_M(I) \) denote the event that \( a_i, i \in I \), are the first \( M \) \( k \)-pre-cuts in \( Y_n(N) \). For \( i \in I \) let \( C_i \) be the event that \( a_i \) is a cut and set

\[
B_m := A_M(I) \cap C_1^\circ \cap \ldots \cap C_m^\circ, \quad 0 \leq m \leq M. \tag{6.17}
\]

We again aim at estimating \( \mu(B_m) \) by an exponentially decaying factor.

For fixed \( M, m \) and \( I \), let \( a := a_m \), pick \( \eta \in \Omega \) with \( F(\eta) = n \) and set \( \Lambda := [a_{-k}, a_k] \cap X \). Let \( \sigma \in B_{m+1} \) such that \( \sigma \sim_\Lambda \eta \). Let \( \sigma' \sim_\Lambda \eta \) be the permutation minimizing \( H_\Lambda \). Since \( a \) is a \( k \)-pre-cut with respect to \( \sigma \), all jumps over \( a \), to \( a_1, ..., a_n \) and from \( a_{-1}, ..., a_{-n} \) are contained in \( \Lambda \) and Lemma 3.1 implies that \( a \) is a cut in permutation \( \sigma' \). We note that \( \sigma' \in B_{m-1} \) since a change on \( \Lambda \) does not affect the \( k \)-pre-cut-status or cut-status of the points of \( Y_n(N) \) other than \( a = a_m \). As \( H_\Lambda(\sigma') - H_\Lambda(\sigma) \leq 0 \) and as there are at most \((2k)! \) different \( \sigma \) that give the same \( \sigma' \) (we change only jumps contained within \( \Lambda \), Lemma 4.5 implies

\[
\gamma_\Lambda(B_{m-1}|\eta) \leq (2k)! \gamma_\Lambda(B_{m-1} \cap C_m|\eta). \tag{6.18}
\]

Integrating with respect to \( \mu \) gives \( \mu(B_{m-1}) \leq (2k)! \mu(B_{m-1} \cap C_m) \), which yields

\[
\mu(B_m) = \mu(B_{m-1} \cap C_m^\circ) \leq \left(1 - \frac{1}{(2k)!}\right) \mu(B_{m-1}). \tag{6.19}
\]

Inductively we thus get \( \mu(B_M) \leq \left(1 - \frac{1}{(2k)!}\right)^M \mu(A_M(I)) \). Since \( \mu \)-a.s. there are infinitely many \( k \)-pre-cuts in \( Y_n(N) \), we can sum over all admissible \( I \), and thus get that the \( \mu \)-probability to have no cuts in \( Y_n(N) \) is at most \( (1 - \frac{1}{(2k)!})^M \). Letting \( M \to \infty \), this shows that \( Y_n(N) \) contains a cut \( \mu \)-a.s. Since this is true for any \( N \), the set \( Y_n \) contains infinitely many cuts \( \mu \)-a.s. in the positive direction; the negative direction is then handled similarly. \( \square \)

We are now able to also give:

**Proof of Lemma 2.13.** The last lemma proves part (a) for \( n \geq 0 \); symmetry then extends this to \( n \leq 0 \). For part (b) with \( n \geq 0 \) it remains to note that occurrence of a single cut restricts the number of infinite cycles from \( -\infty \) to \( \infty \) to (exactly) \( n \) and rules out cycles from \( \infty \) to \( -\infty \) altogether. Having infinitely many such cuts excludes cycles from \( \infty \) to \( \infty \) and \( -\infty \) to \( -\infty \) as well, so the statement in (b) follows. The case \( n \leq 0 \) is completely analogous by symmetry. \( \square \)
6.3 Uniqueness of Gibbs measures.

The sole purpose of this subsection is to give:

**Proof of Lemma 2.14.** Let \( \mu, \bar{\mu} \in \mathcal{G}_n \) for \( n \geq 0 \); the case of negative \( n \) is handled by symmetry. Our principal observation is that Lemmas 6.3 and 6.4 from the preceding subsections still hold when \( \mu \)-a.s.\)” is replaced by \( \mu \otimes \bar{\mu} \)-a.s.\)” and \( k \) is increased somewhat.

To see this for Lemmas 6.3 we first note that \( \mu \otimes \bar{\mu} \) is a Gibbs measure with respect to the product specification \( \gamma_A(\cdot | \eta, \bar{\eta}) := \gamma_A(\cdot | \eta) \otimes \gamma_A(\cdot | \bar{\eta}) \). Also, \( a \) is (defined to be) a pre-cut, resp., cut in \((\sigma, \bar{\sigma})\) if and only if it is a pre-cut, resp., cut in both \( \sigma \) and \( \bar{\sigma} \). Thus the only change to be made in the proof is that in all probability estimates for bad jumps we have to consider two cases: the jump is bad either for \( \sigma \) or for \( \bar{\sigma} \). This leads to an additional factor of 2 in all probability estimates. By increasing \( k \) and \( l(m) \) accordingly, this factor can be easily absorbed.

Concerning Lemma 6.4, the only change required to the proof is that \((2k)! \) has to be replaced by \([2k]!^2 \) since we now have to take into account all possible local rearrangements of \( \sigma \) and \( \bar{\sigma} \) that make a given \( k \)-precut into a cut. This does not affect the argument and, in particular, part (a) of Lemma 2.14 thus holds.

It remains to show that \( \mu = \bar{\mu} \). Consider a cylinder event \( A \in \mathcal{F}_{[-N,N]} \) for some \( N \geq 1 \). Let \( C(a,b) \) be the event that \( a \) is the last cut before \( -N \) and \( b \) the first cut after \( N \). (Because of (a) these cuts exist a.s.) Let \( \Lambda := \{ x \in X : a < x < b \} \). We note that

\[
\gamma_A((A \times \Omega) \cap C(a,b) | \eta, \bar{\eta}) = \gamma_A((\Omega \times A) \cap C(a,b) | \eta, \bar{\eta})
\]

(6.20)

for all \( \eta, \bar{\eta} \in \Omega \). Indeed, if \( a \) or \( b \) are not cuts with respect to \((\eta, \bar{\eta})\) then both sides are 0; otherwise both boundary conditions can be replaced by \( \tau_n \) without changing the probabilities and then the equality follows from the fact that the product measure \( \gamma_A(\cdot | \tau_n, \tau_n) \) and the event \( C(a,b) \) are invariant under \((\sigma, \bar{\sigma}) \mapsto (\bar{\sigma}, \sigma)\). Integrating w.r.t. \( \mu \otimes \bar{\mu} \), we now get

\[
\mu(A) = \mu \otimes \bar{\mu}(A \times \Omega) = \sum_{a,b} \mu \otimes \bar{\mu}( (A \times \Omega) \cap C(a,b) )
\]

(6.21)

\[
= \sum_{a,b} \mu \otimes \bar{\mu}((\Omega \times A) \cap C(a,b)) = \mu \otimes \bar{\mu}(\Omega \times A) = \bar{\mu}(A).
\]

Since \( \bigcup_{N \in \mathbb{N}} \mathcal{F}_{[-N,N]} \) forms a \( \cap \)-stable generator of \( \mathcal{F} \) the above implies \( \mu = \bar{\mu} \). \( \square \)

6.4 Classification of Gibbs measures.

It is formally present the proof of our main result:

**Proof of Theorem 2.15.** All conclusions except (c) follow already from the preceding lemmas. To get also (c), let \( \Lambda_N \uparrow X \) be an increasing sequence of sets and, given \( n \in \mathbb{Z} \), pick \( \eta \in \Omega \) with \( F(\eta) = n \). Combining Lemmas 5.4, 5.5 and part (a) of the theorem we see that every subsequence of \( \{ \gamma_{\Lambda_N}(\cdot | \eta) : N \geq 1 \} \) has a subsequence converging to \( \mu_n \). By a standard argument, this implies \( \gamma_{\Lambda_N}(\cdot | \eta) \rightharpoonup \mu_n \) weakly as \( n \to \infty \). (Otherwise one could find a local event \( A \) and a subsequence \( N_k \) such that \( \gamma_{\Lambda_{N_k}}(A | \eta) \) stays away from \( \mu_n(A) \) by a positive factor.) \( \square \)

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REFERENCES


