Abstract: We study the thermodynamic formalism for generalized Gibbs measures, such as renormalization group transformations of Gibbs measures or joint measures of disordered spin systems. We first show existence of the relative entropy density and obtain a familiar expression in terms of entropy and relative energy for "almost Gibbsian measures" (almost sure continuity of conditional probabilities). We also describe these measures as equilibrium states and establish an extension of the usual variational principle. As a corollary, we obtain a full variational principle for quasilocal measures. For the joint measures of the random field Ising model, we show that the weak Gibbs property holds, with an almost surely rapidly decaying translation invariant potential. For these measures we show that the variational principle fails as soon as the measures loses the almost Gibbs property. These examples suggest that the class of weakly Gibbsian measures is too broad from the perspective of a reasonable thermodynamic formalism.

Keywords: Gibbs vs non-Gibbs, generalized Gibbs measures, variational principle, renormalization group, disordered systems, random field Ising model.

AMS Subject Classification: 60G60 (Primary), 82B20, 82B30 (Secondary).
1 Introduction

Since the discovery of the Griffiths-Pearce singularities of renormalization group transformations \[10, 5\], it has been a challenging question whether the classical Gibbs formalism can be extended in such a way as to incorporate renormalized low temperature phases, so that renormalizing the measure can really be viewed as a transformation on the level of Hamiltonians. Later on, many other examples of “non-Gibbsian” measures appeared in the context of joint measures of disordered spin systems \[13\], time evolution of Gibbs measures \[4\], and dynamical systems \[17\], providing further motivation for the construction of a generalized Gibbs formalism.

As soon as the first examples of non-Gibbsian measures appeared, Dobrushin proposed a program of “Gibbsian restoration of non-Gibbsian fields”, arguing that the phenomenon of non-Gibbsianness is caused by “exceptional” configurations, which are negligible in the measure-theoretic sense. He thus proposed the notion of a “weakly Gibbsian” measure, where the existence of the finite volume Hamiltonian is not required uniformly in the boundary condition, but only for boundary conditions in a set of measure one. This is clearly enough to define the Gibbsian form of the conditional probabilities, and Gibbs measures via the DLR equations. Since \[3\], many papers have been written showing the “weak Gibbs” property of renormalized low temperature phases, see e.g., \[2, 18, 20, 21\], and of joint measures of disordered spin systems, \[13, 14\]. Parallel to this, Fernández and Pfister \[8\] developed ideas about generalized regularity properties of the conditional probabilities. They proved that the decimation of the low temperature plus phase of the Ising model is consistent with a monotone right-continuous system of conditional probabilities. In the framework of investigating regularity of the conditional probabilities in \[18\], the notion of “almost Gibbs” was introduced. A measure \(\mu\) is called almost Gibbs if its conditional probabilities have a version which is continuous on a set of \(\mu\)-measure one. If one does not insist on “absolute” convergence of the sums of potentials constituting finite volume Hamiltonians, then almost Gibbs implies weak Gibbs, but the converse is not true, see \[16, 18\]. In \[7\] it is proved that e.g. the decimation of the plus phase of the low temperature Ising model is almost Gibbs, and the criterion to characterize an essential point of discontinuity of the conditional probabilities given in \[5\] strongly suggests that many other examples of renormalized low temperature phases are almost Gibbs. The investigation of generalized Gibbs properties of the non-Gibbsian measures which appears e.g. as transformations of Gibbs measures is called the “first part of the Dobrushin program”.

The “second part of the Dobrushin program” then consists in building a thermodynamic formalism within the new class of “generalized Gibbs measures”. Already in \[2\] the question is raised whether in the context of weakly Gibbsian measures there is a reasonable notion of “physical equivalence”, i.e., if two systems of conditional probabilities share a Gibbs measure, then they are equal. In the classical Gibbs formalism, physical equivalence corresponds to zero relative entropy density, or zero “information distance”. Generally speaking, one would like to obtain a relation between vanishing relative entropy density and conditional probabilities. For Gibbs measures with a translation invariant uniformly absolutely convergent potential, a translation invariant probability measure \(\mu\) has zero relative entropy density \(h(\mu|\nu)\) with respect to a Gibbs measure \(\nu\) if and only if \(\mu\) is Gibbs with the same potential. Physically speaking, this means that the only minimizers of the free energy are the equilibrium phases. In complete generality, i.e. without any locality requirements, \(h(\mu|\nu) = 0\) does not imply that \(\mu\) and \(\nu\) have anything in common, see e.g. the example in \[26\] where a measure \(\nu\) is constructed such that for any translation
invariant probability measure $h(\mu|\nu) = 0$.

In this paper we investigate the relation between $h(\mu|\nu) = 0$ and the property of having a common system of conditional probabilities for general quasilocal measures, almost Gibbsian measures and weakly Gibbsian measures. We will work in the context of lattice spin systems with a single-site spin taking a finite number of values. Let $\gamma$ denote a translation invariant system of conditional probabilities, and $G_{\text{inv}}(\gamma)$ the set of all translation invariant probability measures having $\gamma$ as a version of their conditional probabilities. If $\gamma$ is continuous then, for $\nu \in G_{\text{inv}}(\gamma)$, we obtain $h(\mu|\nu) = 0$ if and only if $\mu \in G_{\text{inv}}(\gamma)$. If $\gamma$ is continuous $\mu$-almost everywhere, then we obtain that $h(\mu|\nu) = 0$ and $\nu \in G_{\text{inv}}(\gamma)$ implies $\mu \in G_{\text{inv}}(\gamma)$. More generally, for $\nu \in G_{\text{inv}}(\gamma)$ and $\mu \in \mathcal{M}_1^+$ concentrating on a set of “good configurations”, we obtain the existence of $h(\mu|\nu)$, an explicit expression for it where $\nu$ enters only through its conditional probabilities and that $h(\mu|\nu) = 0$ implies $\mu \in G_{\text{inv}}(\gamma)$.

The “good configurations” here are defined such that a telescoping procedure - inspired by the method of Sullivan [25] - almost surely converges. These results, together with some examples of non-Gibbsian measures to which they apply show that almost Gibbsian measures exhibit a reasonable thermodynamic formalism. The fact that some concentration properties of the measures are required is reminiscent of the situation in unbounded spin systems, an analogy already pointed out by Dobrushin, [23].

The context of joint measures of disordered spin systems provides a good source of examples for validity and failure of the variational principle. Here by joint measure we mean the joint distribution of both the spins and the disorder. In these examples (especially for the random field Ising model) there is a precise criterion separating the almost Gibbsian case from the weakly Gibbsian case. In particular, for the random field Ising model, the joint measure is always weakly Gibbs, and at low temperatures we prove here that it even admits a translation invariant potential which decays almost surely as a stretched exponential (so in particular converges absolutely a.s.). If there is no phase transition, then the joint measure for the random field Ising model is almost Gibbs (but not Gibbs in dimension two at low temperature). In the almost Gibbsian regime we obtain the variational principle whereas in the weakly but not-almost Gibbsian regime we show the invalidity of the variational principle. More precisely, in that case the joint measure for the minus phase ($K^-$) is not consistent with the (weakly Gibbsian) system of conditional probabilities of the plus phase ($K^+$), but one easily obtains that the relative entropy densities $h(K^-|K^+) = h(K^+|K^-) = 0$. Physically speaking, this means that we are in the pathological situation where a minimizer of the free energy is not a “phase” (in the DLR sense). At the same time, we also treat the joint measures in a very broad sense, i.e., for possibly non-i.i.d. disorder, prove existence of relative entropy density and give an explicit representation in terms of the defining potentials.

Our paper is organized as follows: in section 2 we introduce basic definitions and notations, discuss the different generalized Gibbs measures and define the variational principle. In section 3 we prove the variational principle for some class of almost Gibbsian measures, using the technique of “relative energies” of [25]. In section 4 we prove the variational principle for measures with translation invariant continuous system of conditional probabilities. In section 5 we give the example of the GriSing random field and the decimation of the low-temperature plus phase of the Ising model. In section 6 we discuss the examples of joint measures of disordered spin systems.
2 Preliminaries

2.1 Configuration space

The configuration space is an infinite product space $\Omega = E^{\mathbb{Z}^d}$ with $E$ a finite set. Its Borel-$\sigma$-field is denoted by $\mathcal{F}$. We denote by $S = \{ \Lambda \subset \mathbb{Z}^d, |\Lambda| < \infty \}$ the set of the finite subsets of $\mathbb{Z}^d$ and for any $\Lambda \in S$, $\Omega_\Lambda = E^\Lambda$. $\mathcal{F}_\Lambda$ denotes the $\sigma$-algebra generated by $\{ \sigma(x) : x \in \Lambda \}$. For all $\sigma, \omega \in \Omega$, we denote $\sigma_\Lambda, \omega_\Lambda$ the projections on $\Omega_\Lambda$ and also write $\sigma_\Lambda \omega_{\Lambda^c}$ for the configuration which agrees with $\sigma$ in $\Lambda$ and with $\omega$ in $\Lambda^c$. The set of probability measures on $(\Omega, \mathcal{F})$ is denoted by $\mathcal{M}^1_\Lambda$. A function $f$ is said to be local if there exists $\Delta \in S$ such that $f$ is $\mathcal{F}_\Delta$-measurable. We denote by $L$ the set of all local functions. The uniform closure of $L$ is $C(\Omega)$, the set of continuous functions on $\Omega$.

On $\Omega$, translations $\{ \tau_x : x \in \mathbb{Z}^d \}$ are defined via $(\tau_x \omega)(y) = \omega(x + y)$, and similarly on functions: $\tau_x f(\omega) = f(\tau_x \omega)$, and on measures $\int f d\tau_x \mu = \int (\tau_x f) d\mu$. The set of translation invariant probability measures on $\Omega$ is denoted by $\mathcal{M}^1_{\text{inv}}$.

On $\Omega$, we have the partial order defined as follows $\eta \leq \zeta$ if and only if for all $x \in \mathbb{Z}^d$ $\eta(x) \leq \zeta(x)$. A function $f : \Omega \to \mathbb{R}$ is called monotone if $\eta \leq \zeta$ implies $f(\eta) \leq f(\zeta)$. This order induces stochastic domination on $\mathcal{M}^1_\Lambda$: $\mu \preceq \nu$ if and only if $\mu(f) \leq \nu(f)$ for all $f$ monotone.

2.2 Specification and quasilocality

**Definition 2.1** A specification on $(\Omega, \mathcal{F})$ is a family $\gamma = \{ \gamma_\Lambda, \Lambda \in S \}$ of probability kernels from $\Omega_{\Lambda^c}$ to $\mathcal{F}$ that are

(I) Proper: For all $B \in \mathcal{F}_{\Lambda^c}$, $\gamma_\Lambda(B|\omega) = 1_B(\omega)$.

(II) Consistent: If $\Lambda \subset \Lambda'$ are finite sets, then $\gamma_{\Lambda'} \gamma_\Lambda = \gamma_{\Lambda'}$.

The notation $\gamma_{\Lambda'} \gamma_\Lambda$ refers to the composition of probability kernels: for $A \in \mathcal{F}$, $\omega \in \Omega$:

$$(\gamma_{\Lambda'} \gamma_\Lambda)(A|\omega) = \int_\Omega \gamma_\Lambda(A|\omega') \gamma_{\Lambda'}(d\omega'|\omega).$$

These kernels also acts on bounded measurable functions $f$:

$$\gamma_\Lambda f(\omega) = \int f(\sigma) \gamma_\Lambda(d\sigma|\omega)$$

and on measures $\mu$:

$$\mu \gamma_\Lambda(f) \equiv \int f d\mu \gamma_\Lambda = \int (\gamma_\Lambda f) d\mu.$$ 

A specification is a strengthening of the notion of a system of proper regular conditional probabilities. Indeed, in the former, the consistency condition (II) is required to hold for every configuration $\omega \in \Omega$, and not only for almost every $\omega \in \Omega$. This is because the notion of specification is defined without any reference to a particular measure. A specification $\gamma$ is translation invariant if for all $A \in \mathcal{F}$, $\Lambda \in S$, $\omega \in \Omega$:

$$\gamma_{\Lambda+x}(A|\omega) = \gamma_\Lambda(\tau_x A|\tau_x \omega)$$
In this paper we will always restrict to the case of non-null specifications, i.e., for any \( \Lambda \in \mathcal{S} \), there exist \( 0 < a_\Lambda < b_\Lambda < 1 \) such that
\[
a_\Lambda < \inf_{\sigma, \eta} \gamma_\Lambda(\sigma | \eta) \leq \sup_{\sigma, \eta} \gamma_\Lambda(\sigma | \eta) < b_\Lambda.
\]

**Definition 2.2** A probability measure \( \mu \) on \((\Omega, \mathcal{F})\) is said to be consistent with a specification \( \gamma \) (or specified by \( \gamma \)) if the latter is a realization of its finite-volume conditional probabilities, that is, if for all \( A \in \mathcal{F} \) and \( \Lambda \in \mathcal{S} \),
\[
\text{for } \mu - \text{a.e. } \omega, \quad \mu[A | \mathcal{F}_\Lambda](\omega) = \gamma_\Lambda(A | \omega). \tag{2.3}
\]

Equivalently, \( \mu \) is consistent with \( \gamma \) if
\[
\int (\gamma_f) d\mu = \int f d\mu
\]
for all \( f \in C(\Omega) \). We denote by \( \mathcal{G}(\gamma) \) the set of measures consistent with \( \gamma \). For a translation invariant specification \( \mathcal{G}_{\text{inv}}(\gamma) \) is the set of translation invariant elements of \( \mathcal{G}(\gamma) \).

**Definition 2.4**
1. A specification \( \gamma \) is quasilocal if for each \( \Lambda \in \mathcal{S} \) and each \( f \) local, \( \gamma_\Lambda f \in C(\Omega) \).
2. A probability measure \( \mu \) is quasilocal if it is consistent with some quasilocal specification.

### 2.3 Potentials and Gibbs measures

Examples of quasilocal measures are Gibbs measures defined via potentials.

**Definition 2.5**
1. A potential is a family \( \Phi = \{ \Phi_A : A \in \mathcal{S} \} \) of local functions such that for all \( A \in \mathcal{S} \), \( \Phi_A \) is \( \mathcal{F}_A \)-measurable.
2. A potential is translation invariant if for all \( A \in \mathcal{S} \), \( x \in \mathbb{Z}^d \) and \( \omega \in \Omega \):
\[
\Phi_{A+x}(\omega) = \Phi_A(\tau_x \omega)
\]

**Definition 2.6**
A potential is said to be
1. Convergent at the configuration \( \omega \) if for all \( \Lambda \in \mathcal{S} \) the sum
\[
\sum_{A \cap \Lambda \neq \emptyset} \Phi_A(\omega) \tag{2.7}
\]
is convergent.
2. Uniformly convergent if convergence in (2.7) is uniform in \( \omega \).
3. Uniformly absolutely convergent (UAC) if for all \( \Lambda \in \mathcal{S} \)
\[
\sum_{A \cap \Lambda \neq \emptyset} \sup_{\omega} |\Phi_A(\omega)| < \infty.
\]
For a general potential $\Phi$, we define the measurable set of its points of convergence:

$$\Omega_\Phi = \{ \omega \in \Omega : \Phi \text{ is convergent at } \omega \}.$$  

In order to define Gibbs measures, we consider a UAC potential and define its finite-volume Hamiltonian for $\Lambda \in S$ and boundary condition $\omega \in \Omega$ by

$$H^\Phi_\Lambda(\sigma | \omega) = \sum_{A \cap \Lambda \neq \emptyset} \Phi_A(\sigma_{\Lambda A})$$

**Definition 2.8** Let $\Phi$ be UAC. The Gibbs specification $\gamma^\Phi$ with potential $\Phi$ is defined by

$$\gamma^\Phi_\Lambda(\sigma | \omega) = \frac{1}{Z^\Phi_\Lambda(\omega)} e^{-H^\Phi_\Lambda(\sigma | \omega)}$$

where the partition function $Z^\Phi_\Lambda(\omega)$ is the normalizing constant.

A measure $\mu$ is a Gibbs measure if there exists a UAC potential $\Phi$ such that $\mu \in \mathcal{G}(\gamma^\Phi)$. Gibbs measures are quasilocal and conversely, any non-null quasilocal measure can be written in a Gibbsian way (see [12] and more details in section 4).

**2.4 Generalized Gibbs measures**

**Definition 2.9** A measure $\nu$ is weakly Gibbs if there exists a potential $\Phi$ such that $\nu(\Omega_\Phi) = 1$ and

$$\nu[\sigma_\Lambda | \mathcal{F}_{\Lambda^c}](\omega) = \frac{e^{-H^\Phi_\Lambda(\sigma | \omega)}}{Z^\Phi_\Lambda(\omega)}$$

for $\nu$-almost every $\omega$.

**Remark 2.10** Some authors insist on the almost surely absolute convergence of the sums defining $H^\Phi_\Lambda$. However, for the definition of the weakly Gibbsian specification there is no reason to prefer absolute convergence.

**Definition 2.11** Let $\gamma$ be a specification. A configuration $\omega$ is said to be a point of continuity for $\gamma$ if for all $\Lambda \in S$, $f \in L$, $\gamma_\Lambda f$ is continuous at $\omega$.

For a given $\gamma$, $\Omega_\gamma$ denotes its measurable set of points of continuity.

**Definition 2.12** A measure $\nu$ is called almost Gibbs if there exists a specification $\gamma$ such that $\nu \in \mathcal{G}(\gamma)$ and $\nu(\Omega_\gamma) = 1$.

If $\nu$ is almost Gibbs, then there exists an almost surely convergent potential $\Phi$ such that $\nu$ is weakly Gibbsian for $\Phi$, and thus almost Gibbsianness implies weak Gibbsianness. The converse is not true: a measure can be weakly Gibbs and for the associated potential $\Phi$, $\Omega_{\gamma^\Phi}$ is of measure zero [16, 18]. If a measure is almost Gibbs and translation invariant, then the corresponding potential can be chosen to be translation invariant.
2.5 Relative entropy and variational principle

For $\mu, \nu \in \mathcal{M}_{1,\text{inv}}^+$, the finite relative entropy at volume $\Lambda \in \mathcal{S}$ of $\mu$ relative to $\nu$ is defined as

$$h_\Lambda(\mu|\nu) = \begin{cases} \int_\Omega \frac{d\mu_\Lambda}{d\nu_\Lambda} \log \frac{d\mu_\Lambda}{d\nu_\Lambda} d\nu_\Lambda & \text{if } \mu_\Lambda \ll \nu_\Lambda \\ +\infty & \text{otherwise} \end{cases}$$

(2.13)

The notation $\mu_\Lambda$ refers to the distribution of $\omega_\Lambda$ when $\omega$ is distributed according to $\mu$. By Jensen’s inequality, $h_\Lambda(\mu|\nu) \geq 0$. The relative entropy of $\mu$ relative to $\nu$ is the limit

$$h(\mu|\nu) = \lim_{n \to \infty} \frac{1}{|\Lambda_n|} h_{\Lambda_n}(\mu|\nu)$$

(2.14)

where $\Lambda_n = [n,n]^d \cap \mathbb{Z}^d$ is a sequence of cubes (this can be replaced by a Van Hove sequence). In what follows, if we write $\lim_{\Lambda \uparrow \mathbb{Z}^d} f(\Lambda)$ we mean that the limit is taken along a Van Hove sequence. The defining limit (2.14) is known to exist if $\nu \in \mathcal{M}_{1,\text{inv}}^+$ is a translation invariant Gibbs measure with a translation invariant UAC potential and $\mu \in \mathcal{M}_{1,\text{inv}}^+$ arbitrary. The Kolmogorov-Sinai entropy $h(\mu)$ is defined for $\mu \in \mathcal{M}_{1,\text{inv}}^+$:

$$h(\mu) = -\lim_{n \to \infty} \frac{1}{|\Lambda_n|} \sum_{\sigma_{\Lambda_n}} \mu(\sigma_{\Lambda_n}) \log \mu(\sigma_{\Lambda_n}).$$

(2.15)

We are now ready to state the variational principle for specifications and measures which gives a relation between zero relative entropy and equality of conditional probabilities.

Definition 2.16 Let $\gamma$ be a specification, $\nu \in \mathcal{G}_{\text{inv}}(\gamma)$ and $\mathcal{M} \subset \mathcal{M}_{1,\text{inv}}^+$. We say that a variational principle holds for the triple $(\gamma, \nu, \mathcal{M})$ if

(0) $h(\mu|\nu)$ exists for all $\mu \in \mathcal{M}$.

(1) $\mu \in \mathcal{G}_{\text{inv}}(\gamma) \cap \mathcal{M}$ implies $h(\mu|\nu) = 0$.

(2) $h(\mu|\nu) = 0$ and $\mu \in \mathcal{M}$ implies $\mu \in \mathcal{G}_{\text{inv}}(\gamma)$.

Items (1) and (2) are called the first and second part of the variational principle. The second part is true for any translation invariant quasilocal measure $\nu$ [9] (with $\mathcal{M} = \mathcal{M}_{1,\text{inv}}^+$). The first part is proved for translation invariant Gibbs measures associated with a translation invariant UAC potential (with $\mathcal{M} = \mathcal{M}_{1,\text{inv}}^+$ also). We extend this result to any translation invariant quasilocal measure in section 4. In [7], the second part has been proved for some renormalized non-Gibbsian FKG measures. In general, the set $\mathcal{M}$ will be a set of translation invariant probability measures concentrating on “good configurations” (e.g., points of continuity of conditional probabilities corresponding to $\gamma$).

3 Variational principle for generalized Gibbs measures

We study the variational principle for generalized Gibbs measures. We first prove the second part for almost Gibbsian measures, which is a rather straightforward technical extension of [9], chapter 15.
3.1 Second part of the variational principle for almost Gibbsian measures

**Theorem 3.1** Let \( \gamma \) be a translation invariant specification on \((\Omega, F)\) and \( \nu \in \mathcal{G}_{\text{inv}}(\gamma) \). For all \( \mu \in \mathcal{M}^+_{1, \text{inv}} \),

\[
\begin{align*}
    h(\mu | \nu) &= 0 \\
    \mu(\Omega_\gamma) &= 1
\end{align*}
\]

and thus such a measure \( \mu \) is almost Gibbs w.r.t. \( \gamma \).

**Proof.** Choose \( \nu \in \mathcal{G}_{\text{inv}}(\gamma) \) and \( \mu \) such that \( h(\mu | \nu) = 0 \). We have to prove that for any \( g \in \mathcal{L}, \Lambda \in \mathcal{S} \):

\[
\mu(\gamma_{\Lambda} g - g) = 0. \tag{3.2}
\]

Fix \( g \in \mathcal{L} \) and \( \Delta \in \mathcal{S} \) such that \( g \) is \( \mathcal{F}_{\Lambda} \)-measurable. The hypothesis

\[
\begin{align*}
    h(\mu | \nu) &= \lim_{\Lambda \uparrow \Omega} \frac{1}{|\Lambda|} h(\mu | \nu) = 0 \tag{3.3}
\end{align*}
\]

implies that for every \( \Lambda \in \mathcal{S} \), the density \( f_{\Lambda} = \frac{d\mu_{\Lambda}}{d\nu_{\Lambda}} \) exists and is a bounded positive \( \mathcal{F}_{\Lambda} \)-measurable function. Introduce local approximations of \( \gamma_{\Lambda} g \):

\[
\begin{align*}
    g_n^- (\sigma) &= \inf_{\omega \in \Omega} \gamma_{\Lambda} g(\sigma_{\Lambda_n} \omega_{\Lambda_n}) \\
    g_n^+ (\sigma) &= \sup_{\omega \in \Omega} \gamma_{\Lambda} g(\sigma_{\Lambda_n} \omega_{\Lambda_n}).
\end{align*}
\]

In the quasilocal case, we have \( g_n^+ - g_n^- \to 0 \) uniformly when \( n \) goes to infinity, whereas here we have \( g_n^+ - g_n^- \to 0 \) on the set \( \Omega_\mu \) of \( \mu \)-measure one, and hence, by dominated convergence in \( L^1(\mu) \). To obtain (3.2) decompose:

\[
\begin{align*}
    \mu(\gamma_{\Lambda} g - g) &= A_n + B_n + C_n + D_n \tag{3.4}
\end{align*}
\]

where

\[
\begin{align*}
    A_n &= \mu(\gamma_{\Lambda} g - g_n^-) \\
    B_n &= \nu((g_n^- - \gamma_{\Lambda} g)f_{\Lambda_n \setminus \Lambda}) \\
    C_n &= \nu(f_{\Lambda_n \setminus \Lambda}(\gamma_{\Lambda} g - g)) \\
    D_n &= \nu((f_{\Lambda_n \setminus \Lambda} - f_{\Lambda_n})g).
\end{align*}
\]

Using

\[
0 \leq \gamma_{\Lambda} g - g_n^- \leq g_n^+ - g_n^-
\]

for \( A_n \to 0 \) as \( n \) goes to infinity. For \( B_n \), use

\[
0 \leq |B_n| = \nu((\gamma_{\Lambda} g - g_n^-)f_{\Lambda_n \setminus \Lambda}) \leq \nu(f_{\Lambda_n \setminus \Lambda}(g_n^+ - g_n^-)) = \mu(g_n^+ - g_n^-),
\]

to obtain \( B_n \to 0 \) as \( n \to \infty \).

Since \( \nu \in \mathcal{G}(\gamma) \), and \( f_{\Lambda_n \setminus \Lambda} \in \mathcal{F}_{\Lambda'} \), \( C_n \to 0 \). The fact that \( D_n \to 0 \) follows from the assumption of zero relative entropy density: see Georgii, p 324.

\[
\]

**Remark 3.5**

1. The role of \( \mathcal{M} \) in Definition 2.16 is played here by the set of measures concentrating on the points of continuity \( \omega \) (\( \mu \in \mathcal{M} \) if and only if \( \mu(\Omega_\gamma) = 1 \)).

2. Remark that in Theorem 3.1, we do not ask any concentration properties of \( \nu \).
3.2 First part of the variational principle for some almost Gibbsian measures

To obtain the first part of the variational principle, it will turn out that concentration of \( \mu \) on the set \( \Omega_\gamma \) is not the right condition. We need that some particular class of “telescoping configurations” are points of continuity of the specification. This reminds of asking continuity properties of the one-sided conditional probabilities. In the case of (uniformly) continuous specifications, this distinction between one-sided and two-sided is of course not visible.

We choose a particular value written +1 in the state space \( E \) and denote by “+” the configuration whose value is +1 everywhere. To any configuration \( \sigma \in \Omega \), we associate the configuration \( \sigma^+ \) defined by

\[
\sigma^+(x) = \begin{cases} 
\sigma(x) & \text{if } x \leq 0 \\
+1 & \text{if } x > 0.
\end{cases}
\]

Here, the order \( \leq \) is lexicographic. We define then \( \Omega_\gamma^{<0} \) to be the subset of \( \Omega \) of the configurations \( \sigma \) such that the new configuration \( \sigma^+ \) is a good configuration for \( \gamma \):

\[
\Omega_\gamma^{<0} = \{ \sigma \in \Omega, \sigma^+ \in \Omega_\gamma \}.
\]

This set will be described in different examples in section 5.

3.2.1 Results

We consider a pair \( (\gamma, \nu) \) of a specification \( \gamma \) and a \( \mu \) which satisfies the following condition:

**Condition C1**

\[
\mu(\Omega_\gamma^{<0}) = 1.
\]

We also introduce

\[
e_\nu^+ := -\lim_{\Lambda \uparrow \mathbb{Z}^d} \frac{1}{|\Lambda|} \log \nu(+\Lambda)
\]

whenever it exists.

**Theorem 3.6** Under the condition C1:

1. \( h(\mu|\nu) \) exists if and only if \( e_\nu^+ \) exists and then

\[
h(\mu|\nu) = e_\nu^+ - h(\mu) - \int_{\Omega} \log \frac{\gamma_0(\sigma^+|\sigma^+)}{\gamma_0(+|\sigma^+)} \mu(d\sigma).
\]

where \( h(\mu) \) is the Kolmogorov-Sinai entropy of \( \mu \).

2. If moreover \( \mu \in \mathcal{G}_{\text{inv}}(\gamma) \) and \( e_\nu^+ \) exists, then

\[
h(\mu|\nu) = \lim_{\Lambda \uparrow \mathbb{Z}^d} \frac{1}{|\Lambda|} \log \frac{\mu(+\Lambda)}{\nu(+\Lambda)}.
\]
To get the more usual expression of the variational principle, we add an extra condition to the condition C1:

**Condition C2**

\[ \mu \in \mathcal{G}_{\text{inv}}(\gamma) \text{ is such that } \lim_{\Lambda \to \mathbb{Z}^d} \frac{1}{|\Lambda|} \log \frac{\mu(+\Lambda)}{\nu(+\Lambda)} = 0. \]

**Theorem 3.9** Assume that conditions C1 and C2 are fulfilled. Then

1. \( h(\mu|\nu) = 0 \).
2. \( e^+_\nu \) exists and \( e^+_\nu = e^+_\mu \).
3. \( h(\alpha|\nu) \) exists for all \( \alpha \in \mathcal{M}^+_{1,\text{inv}} \) satisfying C1.

As a corollary of these theorems, we obtain the usual first part of the variational principle.

**Theorem 3.10** Let \( \mu \in \mathcal{M}^+_{1,\text{inv}} \) and \( \nu \in \mathcal{G}_{\text{inv}}(\gamma) \) such that conditions C1 and C2 hold and \( e^+_\nu \) exists. Then

1. \( h(\mu|\nu) \) exists.
2. \( \mu \in \mathcal{G}_{\text{inv}}(\gamma) \) implies \( h(\mu|\nu) = 0 \).

**Remark 3.11** The existence of the limit defining \( e^+_\nu \) is guaranteed for e.g. renormalization group transformations of Gibbs measures, and for \( \nu \) with positive correlations (by subadditivity). Moreover, in the case of transformations of Gibbs measures, condition C2 is also easy to verify. See section 5 below.

### 3.3 Proofs

**Proof of Theorem 3.6**

First we need the following

**Lemma 3.12** If \( \mu(\Omega^<_0) = 1 \), then

1. Uniformly in \( \omega \in \Omega \),
\[
\lim_{n \to \infty} \frac{1}{|\Lambda_n|} \int_{\Omega} \log \frac{\gamma_{\Lambda_n}(\sigma|\omega)}{\gamma_{\Lambda_n}(+|\omega)} \mu(d\sigma) = \int_{\Omega} \log \frac{\gamma_0(+|\sigma^+)}{\gamma_0(\sigma^+|\sigma^+)} \mu(d\sigma).
\]
2. For \( \nu \in \mathcal{G}(\gamma) \),
\[
\lim_{n \to \infty} \frac{1}{|\Lambda_n|} \int_{\Omega} \log \frac{\nu(\sigma_{\Lambda_n})}{\nu(+_{\Lambda_n})} \mu(d\sigma) = \int_{\Omega} \log \frac{\gamma_0(\sigma^+|\sigma^+)}{\gamma_0(+|\sigma^+)} \mu(d\sigma).
\]
In particular, the limit depends only on the pair \((\gamma, \mu)\).
Remark 3.13 If $\mu$ is ergodic under translations, we have a slightly stronger statement for item 1: \[
\frac{1}{|\Lambda_n|} \int_\Omega \log \frac{\gamma_{\Lambda, \omega}(\sigma)}{\gamma_{\Lambda_n, \omega}(\sigma)} \mu(d\sigma) \text{ converges in } L^1(\mu) \text{ to } \int_\Omega \log \frac{\gamma_{\Lambda_n}(\sigma^+)}{\gamma_{\Omega}(\sigma^+)} \mu(d\sigma), \]
uniformly in $\omega \in \Omega$.

Proof.

1. The proof uses relative energies as in Sullivan [25]. For all $\Lambda \in \mathcal{S}$, $\sigma, \omega \in \Omega$, we define,

$E^+_{\Lambda}(\sigma|\omega) = \log \frac{\gamma_{\Lambda, \omega}(\sigma \mid \omega)}{\gamma_{\Lambda_n, \omega}(\sigma \mid \omega)}$ and $D(\sigma) = E^+_{\{0\}}(\sigma \mid \sigma) = \log \frac{\gamma_0(\sigma)}{\gamma_0(\sigma)}$.

We consider an approximation of $\sigma^+$ at finite volume $\Lambda$ with boundary condition $\omega$ and define the telescoping configuration $T^n_{\Lambda}[x, \sigma, +]$:

\[
T^n_{\Lambda}[x, \sigma, +](y) = \begin{cases} 
\omega(y) & \text{if } y \in \Lambda^c \\
\sigma(y) & \text{if } y \leq x, y \in \Lambda \\
+1 & \text{if } y > x, y \in \Lambda.
\end{cases}
\]

Using the consistency property of $\gamma$, we have by telescoping,

$E^+_{\Lambda}(\sigma|\omega) = \sum_{x \in \Lambda} E^+_{x}(\sigma|T^n_{\Lambda}[x, \sigma, +])$.

By translation invariance of $\gamma$,

$E^+_{\Lambda}(\sigma|\omega) = \sum_{x \in \Lambda} D(\tau_{-x}T^n_{\Lambda}[x, \sigma, +])$.

By translation invariance of $\mu$,

\[
\int_\Omega E^+_{\Lambda_n}(\sigma|\omega) \mu(d\sigma) = \sum_{x \in \Lambda_n} \int_\Omega D(\tau_{-x}T^n_{\Lambda}[x, \tau_x \sigma, +]) \mu(d\sigma).
\]

Therefore, we have to prove that, uniformly in $\omega$,

\[
\lim_{n \to \infty} \frac{1}{|\Lambda_n|} \left( \sum_{x \in \Lambda_n} \int_\Omega \left[ D(\tau_{-x}T^n_{\Lambda}[x, \tau_x \sigma, +]) - D(\sigma^+) \right] \mu(d\sigma) \right) = 0.
\]

By definition,

\[
\tau_{-x}T^n_{\Lambda}[x, \tau_x \sigma, +] = \begin{cases} 
\tau_{-x}\omega(y) & \text{if } y + x \in \Lambda^c_n \\
+ & \text{if } 0 < y, y + x \in \Lambda_n \\
\sigma(y) & \text{if } y \leq 0, y + x \in \Lambda_n.
\end{cases}
\]

Now, pick $\epsilon > 0$, $\omega \in \Omega$ and $\sigma \in \Omega^\leq_{0}$. Using the fact that $\sigma^+$ is a point of continuity of $D$, we choose $n_0$ such that $\xi|_{\Lambda_{n_0}} = \sigma^+|_{\Lambda_{n_0}}$ implies $|D(\xi) - D(\sigma^+)| \leq \epsilon$. Remark
that $\tau_\omega T^\omega_n [x, \tau_\omega, +]$ and $\sigma^+$ are differ only on the set \( \{ y \in \mathbb{Z}^d : x + y \in \Lambda^c_n \} \). Therefore, the difference \( |D(\sigma^+) - D(\tau_\omega T^\omega_n [x, \tau_\omega, +])| \) can only be bigger than $\epsilon$ for $x$ such that $(\Lambda_n - x) \cap \Lambda^c_n \neq \emptyset$.

Therefore,

\[
\frac{1}{|\Lambda_n|} \sum_{x \in \Lambda_n} \left| D(\tau_\omega T^\omega_n [x, \tau_\omega, +]) - D(\sigma^+) \right| \\
\leq \epsilon + 2 \| D \|_{\infty} \frac{1}{|\Lambda_n|} \sum_{x \in \Lambda_n} \left| \{ x \in \Lambda_n : (\Lambda_n - x) \cap \Lambda^c_n \neq \emptyset \} \right|
\]

and this is less or equal to $2\epsilon$ for $n$ big enough. So we obtain that

\[
\frac{1}{|\Lambda_n|} \sum_{x \in \Lambda_n} \left| D(\tau_\omega T^\omega_n [x, \tau_\omega, +]) - D(\sigma^+) \right|
\]

converges to zero on the set of $\Omega_\gamma^{<0}$ of full $\mu$-measure, uniformly in $\omega$. By dominated convergence, we then obtain

\[
\lim_{n \to \infty} \sup_{\omega} \frac{1}{|\Lambda_n|} \int_{\Omega} \left| \sum_{x \in \Lambda_n} \left[ D(\tau_\omega T^\omega_n [x, \tau_\omega, +]) - D(\sigma^+) \right] \right| \mu(d\sigma) = 0
\]

which implies statement 1 of the lemma.

2. Denote

\[
F_{\Lambda_n}(\mu, \nu) = \frac{1}{|\Lambda_n|} \int_{\Omega} \log \frac{\nu(\sigma_{\Lambda_n})}{\nu(+\Lambda_n)} \mu(d\sigma).
\]

Using $\nu \in \mathcal{G}(\gamma)$, we obtain

\[
F_{\Lambda_n}(\mu, \nu) = \frac{1}{|\Lambda_n|} \int_{\Omega} \log \frac{\gamma_{\Lambda_n}(\sigma|\omega)}{\gamma_{\Lambda_n}(+|\omega)} \nu(d\omega) \mu(d\sigma).
\]

Use

\[
\inf_{\omega \in \Omega} \frac{\gamma_{\Lambda_n}(\sigma|\omega)}{\gamma_{\Lambda_n}(+|\omega)} \leq \frac{\int_{\Omega} \gamma_{\Lambda_n}(\sigma|\omega) \nu(d\omega)}{\int_{\Omega} \gamma_{\Lambda_n}(+|\omega) \nu(d\omega)} \leq \sup_{\omega \in \Omega} \frac{\gamma_{\Lambda_n}(\sigma|\omega)}{\gamma_{\Lambda_n}(+|\omega)}
\]

Let $\epsilon > 0$ be given and $\omega = \omega(n, \sigma, \epsilon)$, $\omega' = \omega'(n, \sigma, \epsilon)$ such that

\[
\int_{\Omega} \log \frac{\gamma_{\Lambda_n}(\sigma|\omega)}{\gamma_{\Lambda_n}(+|\omega)} \mu(d\sigma) \geq \int_{\Omega} \log \frac{\gamma_{\Lambda_n}(\sigma|\omega(n, \sigma, \epsilon))}{\gamma_{\Lambda_n}(+|\omega(n, \sigma, \epsilon))} - \epsilon
\]

and

\[
\int_{\Omega} \log \frac{\gamma_{\Lambda_n}(\sigma|\omega)}{\gamma_{\Lambda_n}(+|\omega)} \mu(d\sigma) \leq \int_{\Omega} \log \frac{\gamma_{\Lambda_n}(\sigma|\omega'(n, \sigma, \epsilon))}{\gamma_{\Lambda_n}(+|\omega'(n, \sigma, \epsilon))} + \epsilon.
\]

Now use the first item of the lemma and choose $N$ such that for all $n \geq N$,

\[
\sup_{\omega} \left| \frac{1}{|\Lambda_n|} \int_{\Omega} \log \frac{\gamma_{\Lambda_n}(\sigma|\omega)}{\gamma_{\Lambda_n}(+|\omega)} \mu(d\sigma) - \int_{\Omega} D(\sigma^+) \mu(d\sigma) \right| \leq \epsilon.
\]

For $n \geq N$, we obtain

\[
\int_{\Omega} D(\sigma^+) \mu(d\sigma) - 2\epsilon \leq F_{\Lambda_n}(\mu|\nu) \leq \int_{\Omega} D(\sigma^+) \mu(d\sigma) + 2\epsilon.
\]
Proof of Theorem 3.6

1. Denote

\[ h_n(\mu|\nu) := \frac{1}{|\Lambda_n|} \sum_{\sigma \Lambda_n} \mu(\sigma \Lambda_n) \log \frac{\mu(\sigma \Lambda_n)}{\nu(\sigma \Lambda_n)}. \]

We recall that for \( \mu \in \mathcal{M}^{+}_{\text{inv}}(\Omega) \), the limit of \( h_n(\mu) := -\frac{1}{|\Lambda_n|} \sum_{\sigma \Lambda_n} \mu(\sigma \Lambda_n) \log \mu(\sigma \Lambda_n) \) is the Kolmogorov-Sinai entropy of \( \mu \) denoted \( h(\mu) \). We write

\[ h_n(\mu|\nu) = -h_n(\mu) - \frac{1}{|\Lambda_n|} \sum_{\sigma \Lambda_n} \mu(\sigma \Lambda_n) \log \frac{\nu(\sigma \Lambda_n)}{\nu(\sigma \Lambda_n)} - \frac{1}{|\Lambda_n|} \log \nu(+\Lambda_n). \]

When condition C1 holds, the asymptotic behavior of the second term of the r.h.s. is given by Lemma 3.12. Hence, the relative entropy exists if and only if \( e^+\nu \) exists, and it is given by (3.7).

2. We consider \( \mu \in \mathcal{G}_{\text{inv}}(\gamma) \) such that \( \mu(\Omega^0) = 1 \) and use the following decomposition of the finite volume relative entropy:

\[ h_n(\mu|\nu) = \frac{1}{|\Lambda_n|} \sum_{\sigma \Lambda_n} \mu(\sigma \Lambda_n) \log \frac{\mu(\sigma \Lambda_n)}{\mu(+\Lambda_n)} - \frac{1}{|\Lambda_n|} \sum_{\sigma \Lambda_n} \mu(\sigma \Lambda_n) \log \frac{\nu(\sigma \Lambda_n)}{\nu(\sigma \Lambda_n)} + \frac{1}{|\Lambda_n|} \log \mu(+\Lambda_n). \]

By Lemma 3.12, in the limit \( n \to \infty \), the first two terms of the r.h.s. are functions of \( \gamma \) rather than functions of \( \mu, \nu \in \mathcal{G}_{\text{inv}}(\gamma) \) and cancel out. Hence, the relative entropy exists if and only if the third term converges. Using Item 1 (existence of relative entropy), we obtain the existence of the limit (3.8) and the equality

\[ h(\mu|\nu) = \lim_{n \to \infty} \frac{1}{|\Lambda_n|} \log \frac{\mu(+\Lambda_n)}{\nu(+\Lambda_n)}. \]

Proof of Theorem 3.9

1. This is direct consequence by Theorem 3.6 and (3.8): under the conditions C1 and C2, \( h(\mu|\nu) = 0 \).

2. The existence of the relative entropy proves that \( e^+\nu \) exists and is given by

\[ e^+\nu = h(\mu) + \int \log \frac{\gamma_0(\sigma^+|\sigma^+)}{\gamma_0(+|\sigma^+)} \mu(d\sigma). \]

Combined with C2 this proves \( e^+\mu = e^+\nu \).

3. Consider any other measure \( \alpha \in \mathcal{M}^{+}_{\text{inv}} \) such that C1 holds. The existence of the relative entropy \( h(\alpha|\mu) \) follows by combining the existence of \( e^+\nu \) with Theorem 3.6, and

\[ h(\alpha|\nu) = e^+\nu - h(\alpha) - \int \log \frac{\gamma_0(\sigma^+|\sigma^+)}{\gamma_0(+|\sigma^+)} \alpha(d\sigma). \]

If moreover \( \alpha \) satisfies C2, we also obtain that \( e^+\alpha \) exists and equals \( e^+\nu \).
3.4 Generalization

In the hypothesis of the theorems above, the plus-configuration plays a particular role of telescoping reference configuration. Without too much effort, we obtain the following generalization where we telescope w.r.t a random configuration \( \xi \) chosen from some translation invariant measure \( \lambda \). Results of the previous section are recovered by choosing \( \lambda = \delta_+ \). The generalization to a random telescoping configuration will be natural in the context of joint measures of disordered spin systems in section 6.

For any \( \xi, \sigma \in \Omega \), we define the concatenated configuration \( \sigma^\xi \):

\[
\forall x \in \mathbb{Z}^d, \sigma^\xi(x) = \begin{cases} 
\sigma(x) & \text{if } x \leq 0 \\
\xi(x) & \text{if } x > 0.
\end{cases}
\]

and the set \( \Omega_\gamma^{\xi,<0} \) to be the subset of \( \Omega \times \Omega \) of the configurations \( (\sigma, \xi) \) such that the new configuration \( \sigma^\xi \) is a good configuration for \( \gamma \):

\[
\Omega_\gamma^{\xi,<0} = \{ (\sigma, \xi) \in \Omega \times \Omega, \sigma^\xi \in \Omega_\gamma \}.
\]

We also generalize \( e^+_{\nu} \) and denotes

\[
e^\lambda_{\nu} = - \lim_{|A| \to \infty} \frac{1}{|A|} \int_{\Omega} \log(\nu|\xi) \lambda(d\xi)
\]

provided this limit exists.

We consider a specification \( \gamma \), measures \( \nu \in \mathcal{G}_{\text{inv}}(\gamma) \), \( \mu, \lambda \in \mathcal{M}_{\text{inv}}^{+} \), and the following conditions:

\begin{itemize}
  \item \textbf{C'1} \( \lambda \otimes \mu(\Omega^{\xi,<0}_\gamma) = 1. \)
  \item \textbf{C'2} \( \lim_{|A| \to \infty} \frac{1}{|A|} \int_{\Omega} \left( \log \frac{d\mu_{\Lambda}}{d\lambda}\right)(\xi) \lambda(d\xi_{\Lambda}) = 0. \)
\end{itemize}

The following theorems are the straightforward generalizations of Theorem 3.6 and 3.10, and their proofs follow the same lines.

\textbf{Theorem 3.15} Under the condition \textbf{C'1},

1. \( h(\mu|\nu) \) exists if and only if \( e^\lambda_{\nu} \) exists and then

\[
h(\mu|\nu) = e^\lambda_{\nu} - h(\mu) - \int_{\Omega \times \Omega} \log \frac{\gamma_{0}(\sigma_{\xi_1}|\sigma_{\xi_2})}{\gamma_{0}(\xi_{\sigma_1}|\sigma_{\xi_2})} \mu(d\sigma_{\xi_1}) \lambda(d\xi_{\sigma_1}). \tag{3.16}
\]

2. If moreover \( \mu \in \mathcal{G}_{\text{inv}}(\gamma) \) and \( e^\lambda_{\nu} \) exists, then

\[
h(\mu|\nu) = \lim_{|A| \to \infty} \frac{1}{|A|} \int_{\Omega} \left( \log \frac{d\mu_{\Lambda}}{d\lambda}\right)(\xi) \lambda(d\xi_{\Lambda}).
\]

\textbf{Theorem 3.17} Consider \( \mu \in \mathcal{M}_{\text{inv}}^{+}, \gamma \) a specification, \( \nu \in \mathcal{G}_{\text{inv}}(\gamma) \) such that \( e^\lambda_{\nu} \) exists and conditions \textbf{C'1} and \textbf{C'2} are true. Then

1. \( h(\mu|\nu) \) exists and is given by (3.16).

2. \( \nu \in \mathcal{G}_{\text{inv}}(\gamma) \) implies \( h(\mu|\nu) = 0. \)
4 Variational principle for quasilocal measures

The usual way to prove $\mu \in G_{\text{inv}}(\gamma) \iff h(\mu|\nu) = 0$ in the Gibbsian context uses that $\gamma$ is a specification associated with a translation invariant and UAC potential $\Phi$, and goes via existence and boundary condition independence of pressure (see [9]). Since for a general quasilocal specification $\gamma$ we cannot rely on the existence of such a potential (see [12] and open problem in [5]), we show here that the weaker property of uniform convergence of the vacuum potential which can be associated to the quasilocal $\gamma$ (see [12]) suffices to obtain zero relative entropy.

**Theorem 4.1** Let $\gamma$ be a translation invariant quasilocal specification, $\nu \in G_{\text{inv}}(\gamma)$ and $\mu \in M_{1,\text{inv}}^+$. Then $h(\mu|\nu)$ exists for all $\mu \in M_{1,\text{inv}}^+$ and

$$\mu \in G_{\text{inv}}(\gamma) \iff h(\mu|\nu) = 0.$$ 

**Proof.** The implication of the left (the second part) is proved in [9]. To prove the first part, we need the following lemma to check hypothesis of Theorem 3.10. Condition C2 is trivially true when $\gamma$ is quasilocal ($\Omega < 0 \gamma = \Omega$).

**Lemma 4.2** $\forall \mu, \nu \in G_{\text{inv}}(\gamma)$ with $\gamma$ translation invariant and quasilocal, $e_\nu^+, e_\mu^+$ exist and

$$\lim_{n \to \infty} \frac{1}{|\Lambda_n|} \log \frac{\mu(\Lambda_n^c)}{\nu(\Lambda_n^c)} = 0.$$

**Proof.** In [12], Kozlov proves that to any translation invariant quasilocal specification $\gamma$ there corresponds a translation invariant uniformly convergent vacuum potential $\Phi$ such that $\gamma = \gamma^\Phi$.

By uniform convergence, we have

$$\lim_{|\Lambda| \to \infty} \sup_{\sigma, \eta, \sigma} \left| \sum_{A \supset \Lambda \Lambda \neq \emptyset} \Phi_A(\sigma) \right| = 0. \quad (4.3)$$

Remark that in (4.3), the absolute value is outside the sum, i.e., (4.3) means that the series $\sum_{A \supset \emptyset} \Phi_A(\sigma)$ is convergent in the sup-norm topology on $C(\Omega)$, but not necessarily absolutely convergent\(^1\). We can defined a Hamiltonian and a partition function for any $\Lambda \in \mathcal{S}, \eta, \sigma \in \Omega$, as usual:

$$H_\Lambda^\eta(\sigma) = \sum_{A \subset \Lambda \Lambda \neq \emptyset} \Phi_A(\sigma)\eta_{A^c}$$

and

$$Z_\Lambda(\omega) = \sum_{\sigma \in \Omega} e^{-H_\Lambda^\eta(\sigma)}.$$ 

(4.4)

Lemma 4.2 is now a direct consequence of the following

**Lemma 4.5** $1.$

$$\lim_{n \to \infty} \sup_{\omega, \eta, \sigma} \frac{1}{|\Lambda_n|} \left| H_\Lambda^\eta(\sigma) - H_\Lambda^\omega(\sigma) \right| = 0. \quad (4.6)$$

\(^1\)The so-called *telescoping potential* introduced by Kozlov [12] is UAC but is not translation invariant in general.
\[ \lim_{n \to \infty} \sup_{\omega, \eta} \frac{1}{|\Lambda_n|} \log \frac{Z_{\Lambda_n}(\omega)}{Z_{\Lambda_n}(\eta)} = 0. \]  

(4.7)

**Proof.** We follow the standard line of the argument used by [11] to prove existence and boundary condition independence of the pressure for a UAC potential, but we detail it because the vacuum potential is only uniformly convergent. Clearly, (4.6) implies (4.7): for all \( n \in \mathbb{N}, \)

\[
\exp \left\{ - \sup_{\omega, \eta, \sigma} \left| H_{\Lambda_n}^\eta (\sigma) - H_{\Lambda_n}^\omega (\sigma) \right| \right\} \leq \sup_{\omega, \eta} \frac{Z_{\Lambda_n}(\omega)}{Z_{\Lambda_n}(\eta)} \leq \exp \left\{ \sup_{\omega, \eta, \sigma} \left| H_{\Lambda_n}^\eta (\sigma) - H_{\Lambda_n}^\omega (\sigma) \right| \right\}.
\]

To prove (4.6), we write

\[
H_{\Lambda_n}^\eta (\sigma) - H_{\Lambda_n}^\omega (\sigma) = \sum_{A \cap \Lambda_n \neq \emptyset, A \cap \Lambda_n^c \neq \emptyset} \left[ \Phi_A(\sigma, \eta, \Lambda_n^c) - \Phi_A(\sigma, \omega, \Lambda_n^c) \right].
\]

and we first remark:

\[
\frac{1}{|\Lambda_n|} \sum_{A \cap \Lambda_n \neq \emptyset, A \cap \Lambda_n^c \neq \emptyset} \left| \Phi_A(\sigma, \eta, \Lambda_n^c) - \Phi_A(\sigma, \omega, \Lambda_n^c) \right| \leq \frac{2}{|\Lambda_n|} \sum_{\tau \in \Lambda_n} \sup_{\sigma} \left| \sum_{A \ni \tau, A \cap \Lambda_n^c \neq \emptyset} \Phi_A(\sigma) \right|.
\]

We obtain

\[
\sup_{\sigma} \left| \sum_{A \ni \tau, A \cap \Lambda_n^c \neq \emptyset} \Phi_A(\sigma) \right| = \left| \sum_{A \ni \tau, A \subset \Lambda_n} \Phi_A(\tau_x \sigma) - \sum_{A \ni \tau, A \subset \Lambda_n} \Phi_A(\tau_x \sigma) \right| \leq \sup_{\xi} \left| \sum_{A \ni \tau, \Lambda_n - \tau} \Phi_A(\xi) \right|.
\]

Pick \( \epsilon > 0 \) and choose \( \Delta \) such that

\[
\sup_{\xi} \left| \sum_{A \ni \tau, A \cap \Delta^c \neq \emptyset} \Phi_A(\xi) \right| \leq \epsilon
\]

then

\[
\left| \sum_{A \ni \tau, A \cap (\Lambda_n - \tau)^c \neq \emptyset} \Phi_A(\xi) \right| \leq \begin{cases} \epsilon & \text{ if } (\Lambda_n - \tau) \supset \Delta \\ C & \text{ if } (\Lambda_n - \tau) \cap \Delta^c = \emptyset \end{cases}
\]

where

\[
C = \sup_{\xi} \left| \sum_{A \ni \tau} \Phi_A(\xi) \right| < \infty.
\]

Since for any \( \Delta \subset \mathbb{Z}^d \) finite,

\[
\lim_{n \to \infty} \epsilon \left| \frac{\{x : \Delta + x \cap \Lambda_n^c \neq \emptyset\}}{|\Lambda_n|} \right| = 0
\]

we obtain

\[
\limsup_{n} \frac{1}{|\Lambda_n|} \sum_{\tau \in \Lambda_n} \sup_{\xi} \left| \sum_{A \ni \tau, A \cap \Lambda_n^c \neq \emptyset} \Phi_A(\xi) \right| \leq \epsilon
\]

15
which by the arbitrary choice of \( \epsilon > 0 \) proves (4.6) and the statement of the lemma.

To derive Lemma 4.5 from Lemma 4.2, we only have to prove that for all \( \nu \in \mathcal{G}_{\text{inv}}(\gamma) \), \( \epsilon^+_\nu \) exists and is independent of \( \gamma \). For such a measure \( \nu \), write

\[
\nu(+\Lambda) = \int_{\Omega} e^{-H^\eta_{\Lambda_n}(+)} \frac{Z_{\Lambda_n}(\eta)}{Z_{\Lambda_n}(\eta)} \nu(d\eta)
\]

where \( H^\eta_{\Lambda_n} \) is defined via the vacuum potential of \( \gamma \) in (4.4). We use Lemma 4.5 to write

\[
\nu(+\Lambda) \approx \int_{\Omega} e^{-H^\Lambda(+)} Z^+_{\Lambda} \nu(d\eta)
\]

where \( a_{\Lambda} \equiv b_{\Lambda} \) means \( \lim_{\Lambda \to \infty} \frac{1}{|\Lambda|} \log \frac{a_{\Lambda}}{b_{\Lambda}} = 0 \). Since \( \Phi \) is the vacuum potential with vacuum state +, \( H^\Lambda(+) = 0 \) and hence

\[
\nu(+\Lambda) = (Z^+_{\Lambda})^{-1} = (Z^\text{free}_{\Lambda})^{-1} = \left( \prod_{\sigma \in \Omega_{\Lambda}} \exp(-\sum_{A \subset \Lambda} \Phi_A(\sigma)) \right)^{-1}
\]

where \( Z^+_{\Lambda} \) (resp. \( Z^\text{free}_{\Lambda} \)) is the partition function with + (resp. free) boundary condition, which in our case coincide. Fix \( R > 0 \) and put

\[
\Phi^{(R)}_A(\sigma) := \Phi_A(\sigma) \text{ if diam}(A) \leq R
\]

\[
= 0 \text{ if diam}(A) > R.
\]

Then, using existence of pressure for finite range potentials, cf. Israel [11],

\[
\lim_{\Lambda \to \infty} \frac{1}{|\Lambda|} \log Z^\text{free}_{\Lambda}(\Phi^{(R)}_A) := P(\Phi^{(R)}_A) \text{ exists.}
\]

Now use

\[
\log \frac{\sum_{\sigma} \exp(-\sum_{A \subset \Lambda} \Phi_A(\sigma))}{\sum_{\sigma} \exp(-\sum_{A \subset \Lambda} \Phi^{(R)}_A(\sigma))} \leq \sup_{\sigma} \left| \sum_{A \subset \Lambda, \text{diam}(A) > R} \Phi_A(\sigma) \right|
\]

\[
\leq \sup_{\sigma} \sum_{x \in \Lambda} \left| \sum_{A \ni x, \text{diam}(A) > R} \Phi_A(\sigma) \right|
\]

\[
\leq \sum_{x \in \Lambda} \sup_{\sigma} \left| \sum_{A \ni x, \text{diam}(A) > R} \Phi_A(\sigma) \right|
\]

\[
= |\Lambda| \sup_{\sigma} \left| \sum_{A \ni 0, \text{diam}(A) > R} \Phi_A(\sigma) \right|
\]

and

\[
\frac{\sum_{\sigma} \exp(-\sum_{A \subset \Lambda} \Phi^{(R)}_A(\sigma))}{\sum_{\sigma} \exp(-\sum_{A \subset \Lambda} \Phi^{(R)}_A(\sigma))} \leq |\Lambda| \sup_{\sigma} \left| \sum_{A \ni 0, \text{diam}(A) > R} \Phi_A(\sigma) \right|
\]

to conclude that \( \{P(\Phi^{(R)}_A), R > 0\} \) is a Cauchy net with limit

\[
\lim_{R \to \infty} P(\Phi^{(R)}_A) = \lim_{\Lambda \to \infty} \frac{1}{|\Lambda|} \log Z^\text{free}_{\Lambda} = \epsilon^+_\nu.
\]
which depends only on the vacuum potential (hence on the specification $\gamma$). This proves that $e_\mu^+$ and $e_\nu^+$ exist for all $\mu, \nu \in \mathcal{G}_{\text{inv}}(\gamma)$, and depends of $\gamma$ only. This proves 

$$\lim_{\Lambda \uparrow \mathbb{Z}^d} \frac{1}{|\Lambda|} \log \frac{\mu^{-}(+\Lambda)}{\nu^{-}(+\Lambda)} = e_\nu^+ - e_\mu^+ = 0$$

and Lemma 4.2.

A direct consequence of this lemma is that in the framework of Theorem 4.1, $e_\nu^+$ exists and condition C1 and C2 are true. We obtain the theorem by applying Theorem 3.10.

5 Examples

5.1 The GriSing random field

The GriSing random field is an example of joint measure of disordered systems, studied in more details in section 6. It has been studied in [6] and provides an example of a non-Gibbsian random fields which fits in the framework of our theorems. The random field is constructed as follows. Sites are empty or occupied according a Bernoulli product measure of parameter $p < p_c$ where $p_c$ is the percolation threshold for site percolation on $\mathbb{Z}^d$. For any realization $\eta$ of occupancies where all occupied clusters are finite, we have the Gibbs measure on configurations $\sigma \in \{-1, +1\}^{\mathbb{Z}^d}$

$$\mu^\eta(\sigma) = \frac{1}{Z_\Lambda} e^{-\beta \sum_{xy \subset C} \sigma(x)\sigma(y)}.$$

The GriSing random field is then defined as:

$$\xi(x) = \sigma(x)\eta(x).$$

In words, $\xi(x) = 0$ for unoccupied sites and equal to the spin $\sigma(x)$ at occupied sites.

We denote by $K_{p,\beta}$ the law of the random field $\xi$.

It is known that for any $p \in (0, 1)$, $\beta$ large enough, $K_{p,\beta}$ is not a Gibbs measure (see [6] for $p < p_c$ and [13] for any $p \in (0, 1)$). The points of essential discontinuity of the conditional probabilities $K_{p,\beta}(\sigma(0)|\xi_{\mathbb{Z}^d \setminus \{0\}})$ are a subset of

$$D = \{\xi : \xi \text{ contains an infinite cluster of occupied sites}\}.$$

Since $p < p_c$, there exists a specification $\gamma$ such that $\{K_{p,\beta}\} = \mathcal{G}(\gamma)$ and such that for the continuity points $\Omega_\gamma$, we have $K_{p,\beta}(\Omega_\gamma) = 1$, i.e., $K_{p,\beta}$ is almost Gibbs. Moreover, if we choose $\xi_0 \equiv 0$ as a telescoping reference configuration, then clearly $\sigma \in D^c$ implies $\sigma^{\xi_0} \in D^c$, i.e., in this case $\Omega_\gamma \subset \Omega^{\xi_0}_\gamma$. Therefore, in this example condition C1 is satisfied as soon as $\mu$ concentrates on $D^c$. Using $\{K_{p,\beta}\} = \mathcal{G}(\gamma)$, and

$$\lim_{\Lambda \uparrow \mathbb{Z}^d} \frac{1}{|\Lambda|} \log K_{p,\beta}(0_\Lambda) = \log(1 - p),$$

we obtain the following proposition:

**Proposition 5.1** If $\mu(D) = 0$ then $h(\mu|K_{p,\beta})$ exists and is zero if and only if $\mu = K_{p,\beta}$. 

17
5.2 Decimation

Let \( \mu^\beta_+ \) (resp. \( \mu^\beta_- \)) be the low-temperature \( (\beta > \beta_c) \) plus (resp. minus) phase of the Ising model on \( \mathbb{Z}^d \). For \( b \in \mathbb{N} \), \( \nu^\beta_+ \) denotes its decimation, i.e., the distribution of \( \{ \sigma(bx) : x \in \mathbb{Z}^d \} \) when \( \sigma \) is distributed according to \( \mu^\beta_+ \). It is known that \( \nu^\beta_+ \) is not a Gibbs measure \([5]\). In \([8]\) it is proved that there exists a monotone specification \( \gamma^+ \) such that \( \nu^\beta_+ \in \mathcal{G}(\gamma^+) \). In \([7]\) it is proved that the points of continuity \( \Omega_{\gamma^+} \) satisfy \( \nu^\beta_+(\Omega_{\gamma^+}) = 1 \), i.e., \( \nu^\beta_+ \) is almost Gibbs. The point of continuity of \( \gamma^+ \) can be described as those configurations \( \eta \) for which the “internal spins” do not exhibit a phase transition when the decimated spins are fixed to be \( \eta \). E.g., the all plus and the all minus configurations are elements of \( \Omega_{\gamma^+} \), but the alternating configuration is not.

The first part of the variational principle \( (\gamma^+, \nu^\beta_+, \mathcal{M}) \), where \( \mathcal{M} \) is the set of the translation invariant measures which concentrate on \( \Omega_{\gamma^+} \), has already been proved in \([7]\) (and is direct by Theorem 3.1). Here we complete this result by adding a second part:

**Theorem 5.2** For any \( \mu \in \mathcal{M}^+_{\text{inv}} \) satisfying CI for \( \gamma^+ \),

1. \( h(\mu|\nu) \) exists.
2. 
   \[ \mu \in \mathcal{G}_{\text{inv}}(\gamma^+) \iff h(\mu|\nu^+) = 0. \]

We first use a lemma.

**Lemma 5.3** \( \mu \in \mathcal{G}(\gamma^+) \) and \( \mu(\Omega_{\gamma^+}) = 1 \) implies

\[ \nu^\beta_+ \preceq \mu \preceq \nu^\beta_+ \tag{5.4} \]

**Proof.** Consider \( f \) monotone. Then for all \( \Lambda \in \mathcal{S} \), by monotonicity of \( \gamma^+ \) \([8]\),

\[ \int fd\mu = \int_{\Omega} (\gamma^+_\Lambda f)(\omega) \mu(d\omega) \leq \int_{\Omega} (\gamma^+_\Lambda f)(+) \mu(d\omega) = (\gamma^+_\Lambda f)(+). \]

Taking the limit \( \Lambda \uparrow \mathbb{Z}^d \) gives, and using \( \gamma^+_\Lambda (\cdot|+) \) goes to \( \nu^\beta_+ \),

\[ \int fd\mu \leq \int fd\nu^\beta_+. \]

Similarly, using \( \mu(\Omega_{\gamma}) = 1 \), and the expression of \( \Omega_{\gamma} \) in \([8]\), we have \( \gamma^-_\Lambda (f) = \gamma^-_\Lambda (f) \), \( \mu \)-a.s. and hence

\[ \int fd\mu = \int \gamma^-_\Lambda (f) d\mu \geq \gamma^-_\Lambda f(-) \]

which gives

\[ \int fd\mu \geq \int fd\nu^-_. \]

\( \blacksquare \)

The following corollary proves Theorem 5.2 using Theorem 3.10.

**Proposition 5.5** 1. \( e^+_\nu^\beta = -\lim_{\Lambda \uparrow \mathbb{Z}^d} \frac{1}{|\Lambda|} \log \nu^\beta_(+) \) exists.
2. For any \( \mu \in G(\gamma^+) \),
\[
\lim_{\Lambda \uparrow \mathbb{Z}^d} \frac{1}{|\Lambda|} \log \frac{\mu(\Lambda)}{\nu^+_\beta(\Lambda)} = 0.
\]

**Proof.**

1. Follows from subadditivity and positive correlations.

2. Follows from stochastic domination (5.4) and
\[
\lim_{\Lambda \uparrow \mathbb{Z}^d} \frac{1}{|\Lambda|} \log \frac{\nu^+_\beta(\Lambda)}{\nu^+_\beta(\Lambda)} = \lim_{\Lambda \uparrow \mathbb{Z}^d} \frac{1}{|\Lambda|} \log \frac{\mu^+_\beta(\Lambda)}{\mu^+_\beta(\Lambda)} = 0
\]

where, to obtain the last equality, we used that \( \mu^+_\beta, \mu^-_\beta \) are the Ising plus and minus phases.

**Remark 5.6** We conjecture that \( C1 \) is satisfied for all ergodic measure \( \mu \in G(\gamma^+) \) in dimension \( d = 2 \). This amounts of proving that the internal spins do not show a phase transition, given a “typical configuration of \( \mu \)” on \( b\mathbb{Z}^d \) on the left of the origin, and all + on \( b\mathbb{Z}^d \) on the right. Fixing these decimated spins acts as a magnetic field, pushing the spins on the right of the origin into a “plus-like” phase and the spins on the left of the origin in a “plus-like” or “minus-like” phase, depending on \( \mu \). The location of the interface between “right and left” should not depend on the boundary condition in \( d = 2 \) (no Basuev transition). However, we do not have a rigorous proof of this fact.

6 Examples II: Joint measure of random spin systems

We consider the joint measures of disordered spins-systems on the product of spin-space and disorder-space defined in terms of a quenched absolutely convergent Gibbs-interaction and an a priori-distribution of the disorder variables. They were treated before in [13, 14] and provide a broad class of examples of generalized Gibbs measures. A specific example of this, the GriSing field, was already considered in section 5.1.

First we prove that, for the same quenched potential, the relative entropy density between corresponding, possibly different joint measures is always zero. Next we prove in generality that these measures are asymptotically decoupled whenever the a-priori distribution of the disorder is. The useful notion of asymptotic decoupled was recently coined by Pfister [22], and provides a broad class of measures, including local transformations of Gibbs measures, for which the existence of relative entropy density and the large deviation principle holds. Using results of this paper, we easily obtain existence of the relative entropy density. Next we specialize to the specific example of the random field Ising model in section 6.3. We focus on the interesting region of the parameter space when there is a phase transition for the spin-variables, for almost any configuration of disorder variables. Here we show on the basis of [14] that the joint plus and the joint minus state for the same quenched potential are not compatible with the same interaction potential. In [14] it was already shown that there is always a translation-invariant convergent potential, or a possibly non-translation-invariant absolutely convergent potential for the corresponding joint
measure. Here we discuss this in more detail and sketch a proof on the basis of [14] and the RG-analysis of [1] that shows that there is a translation-invariant joint potential that even decays like a stretched exponential. This provides an explicit example of a weakly (but not almost) Gibbsian measure for which the variational principle fails.

6.1 Setup

We consider disordered models of the following general type. We assume that the configuration space of the quenched model is again as detailed in section 2.1 and we denote the spin variables by $\sigma$. Additionally we assume that there are also disorder variables $\eta = (\eta_x)_{x \in \mathbb{Z}^d}$ entering the game, taking values in an infinite product space $(E')^\mathbb{Z}^d$, where again $E'$ is a finite set. We denote the joint variables by $\xi = (\xi_x)_{x \in \mathbb{Z}^d} = (\sigma, \eta) = (\sigma_x, \eta_x)_{x \in \mathbb{Z}^d}$. It will be convenient later also to write simply $(\sigma, \eta)$ to denote the pair $(\sigma_x, \eta_x)$.

One essential ingredient of the model is given by the defining potential $\Phi = (\Phi_A)_{A \subset \mathbb{Z}^d}$ depending on the joint variables $\xi = (\sigma, \eta)$. $\Phi_A(\xi)$ depends on $\xi$ only through $\xi_A$. We assume that $\Phi$ is finite range. When we fix a realization of the disorder $\eta$, we have a potential for the spin-variables $\sigma$ that is typically non-translation invariant. We then define the corresponding quenched Gibbs specification by Definition 2.8 using the notation

$$\mu_{\Lambda}^\sigma(\eta)(B) := \frac{1}{Z_{\Lambda}^{\eta}} \sum_{\sigma_{\Lambda}} 1_B(\sigma_A \bar{\sigma}_{\mathbb{Z}^d \setminus \Lambda}) e^{-\sum_{A: A \subset \Lambda \neq \emptyset} \Phi_A(\sigma_A \bar{\sigma}_{\mathbb{Z}^d \setminus \Lambda}, \eta)}.$$  \hspace{1cm} (6.1)

Here we do not make the defining potential explicit anymore in order not to overburden notation. The measures (6.1) are also called more loosely quenched finite volume Gibbs measures. Obviously, the finite-volume summation is over $\sigma_{\Lambda} \in E^\Lambda$. The symbol $\sigma_{\Lambda} \bar{\sigma}_{\mathbb{Z}^d \setminus \Lambda}$ denotes the configuration in $\Omega$ that is given by $\sigma_x$ for $x \in \Lambda$ and by $\bar{\sigma}_x$ for $x \in \mathbb{Z}^d \setminus \Lambda$.

The second ingredient of the quenched model is the distribution of the disorder variables $P(d\eta)$. Most of the times in the theory of disordered systems one considers the case of i.i.d. variables, but we can and will be more general here.

The objects of interest will then be the infinite volume joint measures $K^\sigma(\eta)(d\xi)$, by which we understand any limiting measure of $\lim_{\Lambda \uparrow \mathbb{Z}^d} P(d\eta) \mu_{\Lambda}^\sigma(\eta)(d\sigma)$ in the product topology on the space of joint variables. Of course, there are examples for different joint measures of the same quenched Gibbs specification for different spin boundary conditions $\bar{\sigma}$. In principle there can even be different ones for the same spin-boundary condition $\bar{\sigma}$, depending on the sub-sequence, though this is not to be expected in reasonable models.

For all of this the reader might think of the concrete example of the Random Field Ising model. Here the spin variables $\sigma_x$ take values in $\{-1, 1\}$. The disorder variables are given by the random fields $\eta_x$ that are i.i.d. with single-site distribution $P_0$ that is supported on a finite set $\mathcal{H}_0$ and assumed to be symmetric. The defining potential $\Phi(\sigma, \eta)$ is given by $\Phi_{\{x,y\}}(\sigma, \eta) = -\beta \sigma_x \sigma_y$ for nearest neighbors $x, y \in \mathbb{Z}^d$, $\Phi_{\{x\}}(\sigma, \eta) = -h \eta_x \sigma_x$, and $\Phi_A = 0$ else.

6.2 Relative entropy for joint measures

For the first result we do not need the independence of the disorder field. In fact, without any decoupling assumption on $P$ we have the following.
**Theorem 6.2** Denote by $K^\sigma$ and $K^{\sigma'}$ two joint measures for the same quenched Gibbs specification $\mu_{\Lambda}[\eta](d\eta)$, obtained with any two spin boundary conditions $\sigma$ (and $\sigma'$ respectively), along any subsequences $\Lambda_N$ (and $\Lambda'_N$ respectively). Then their relative entropy density vanishes, i.e., $h(K^\sigma|K^{\sigma'}) = 0$.

**Remark 6.3** Note that we are more general than in the usual set up and we do not need to assume translation invariance, not even of the defining potential $\Phi$.

**Remark 6.4** This result is neither directly related to the first part nor to the second part of the variational principle. It does not yield the first part (which will be proved differently) because it is not clear that every measure that is compatible with the same specification as $K^\sigma$ can be written in terms of $K^{\sigma'}$. Applied to the random field Ising model in section 6.3, this result will disprove the second part of the variational principle for weakly but not almost Gibbs measures.

**Proof.** We have from the definition of the joint measures as limit points with suitable sequences of volumes

$$
\frac{K^\sigma(\sigma_{\Lambda}\eta_\Lambda)}{K^{\sigma'}(\sigma_{\Lambda}\eta_\Lambda)} = \lim_N \frac{K_N^\sigma(\sigma_{\Lambda_N}\eta_{\Lambda_N})}{K_N^{\sigma'}(\sigma_{\Lambda_N}\eta_{\Lambda_N})} = \lim_N \frac{\int \mathbb{P}(d\tilde{\eta})1_{\eta_{\Lambda_N}[\tilde{\eta}](\sigma_{\Lambda_N})}}{\int \mathbb{P}(d\tilde{\eta})1_{\eta_{\Lambda_N}[\tilde{\eta}](\sigma_{\Lambda_N})}}.
$$

(6.5)

Here and later we will write in short $1_{\eta_A}$ for the indicator function of the event that the integration variable $\tilde{\eta}$ coincides with the fixed configuration $\eta$ on $\Lambda$. We have from the finite range of the disordered potential that

$$
\sup_{\eta^{\sigma}} \left| \sum_{A} (\Phi_A(\sigma) - \Phi_A(\sigma')) \right| \leq C_1 |\partial\Lambda|
$$

for cubes $\Lambda$ with some finite constant $C_1$. By $\partial\Lambda$ we mean the $r$-boundary of $\Lambda$, where $r$ is the range of $\Phi$. So we get that for $N$ large enough

$$
e^{-2C_1|\partial\Lambda|} \mu_{\Lambda}[\eta_{\Lambda_N}[\tilde{\eta}_{\Lambda_N}\setminus\Lambda](\sigma_{\Lambda})] \leq \mu_{\Lambda}[\eta_{\Lambda_N}[\tilde{\eta}_{\Lambda_N}\setminus\Lambda](\sigma_{\Lambda})] \leq e^{2C_1|\partial\Lambda|} \mu_{\Lambda}[\eta_{\Lambda_N}[\tilde{\eta}_{\Lambda_N}\setminus\Lambda](\sigma_{\Lambda})]
$$

for any joint reference configuration $\tilde{\sigma}\tilde{\eta}$. But this gives the upper bound $e^{4C_1|\partial\Lambda|}$ on the r.h.s. of (6.5), by application of the last inequalities on numerator and denominator of (6.5) for the same reference configuration.

This implies for the finite volume relative entropy an upper bound of the order of the boundary, i.e.,

$$
h_{\Lambda}(K^\sigma|K^{\sigma'}) = \sum_{\sigma_{\Lambda}\eta_\Lambda} K^\sigma(\sigma_{\Lambda}\eta_\Lambda) \log \frac{K^\sigma(\sigma_{\Lambda}\eta_\Lambda)}{K^{\sigma'}(\sigma_{\Lambda}\eta_\Lambda)} \leq 4C_1|\partial\Lambda|.
$$

From that clearly follows the claim $h(K^\sigma|K^{\sigma'}) \leq \limsup_{n\to\infty} \frac{1}{|\Lambda_n|} h_{\Lambda_n}(K^\sigma|K^{\sigma'}) = 0$ for $\Lambda_n$ = sequence of cubes and this finishes the proof.

Also the next theorem can be proved in a natural way when we relax the independence assumption of the a priori distribution $\mathbb{P}$ of the disorder variables. It says that the property of being *asymptotically decoupled* carries over from the distribution of the disorder fields to any corresponding joint distribution. Following [22], we give the following
Definition 6.6 A probability measure $\mathbb{P} \in \mathcal{M}_{1,\text{inv}}^+$ is called asymptotically decoupled (AD) if there exists sequences $g_n$, $c_n$ such that
\[
\lim_{n \to \infty} \frac{c_n}{|\Lambda_n|} = 0, \quad \lim_{n \to \infty} \frac{g_n}{n} = 0
\]
and for all $A \in \mathcal{F}_{\Lambda_n}$, $B \in \mathcal{F}_{\Lambda_n^c + g_n}$ with $\mathbb{P}(A)\mathbb{P}(B) \neq 0$:
\[
e^{-c_n} \leq \frac{\mathbb{P}(A \cap B)}{\mathbb{P}(A)\mathbb{P}(B)} \leq e^{c_n}. \quad (6.7)
\]

Theorem 6.8 Suppose $\mathbb{P}$ is asymptotically decoupled with functions $g_n$ and $c_n$. Assume that $K_{\hat{\sigma}}$ is a corresponding translation invariant joint measure of a quenched random system, with a defining finite range potential. Then $K_{\hat{\sigma}}$ is asymptotically decoupled with functions $g'_n = g_n$ and $c'_n = c_n + C|\partial \Lambda_n|$, where $C$ is a finite constant.

Proof. It suffices to show that for any finite $V \subset \Lambda_n^c + g'(n)$ we have
\[
e^{-c'_n} \leq \frac{K(\xi_{\Lambda_n} \xi_V)}{K(\xi_{\Lambda_n})K(\xi_V)} = \frac{K(\sigma_{\Lambda_n} \eta_{\Lambda_n} \sigma_V \eta_V)}{K(\sigma_{\Lambda_n} \eta_{\Lambda_n})K(\sigma_V \eta_V)} \leq e^{c'_n}. \quad (6.9)
\]
We only show the upper bound. It suffices to show
\[
\limsup_n \frac{K_{\hat{\sigma}_{\Lambda_n}}(\sigma_{\Lambda_n} \eta_{\Lambda_n} \sigma_V \eta_V)}{K_{\hat{\sigma}_{\Lambda_n}}(\sigma_{\Lambda_n})K_{\hat{\sigma}_{\Lambda_n}}(\sigma_V \eta_V)} \leq e^{c_n}
\]
for any sequence $\hat{\Lambda}_n$. The quantity under the limsup equals
\[
\int \mathbb{P}(d\tilde{\eta}_1) \eta_{\Lambda_n} 1_{\eta_V} \mu_{\hat{\sigma}_{\Lambda_n}}^\sigma [\tilde{\eta}](\sigma_V | \sigma_{\Lambda_n}) \int \mathbb{P}(d\tilde{\eta}_2) \eta_{\Lambda_n} 1_{\eta_V} \mu_{\hat{\sigma}_{\Lambda_n}}^\sigma [\tilde{\eta}_2](\sigma_V | \sigma_{\Lambda_n}). \quad (6.10)
\]
Look at the term under the disorder-integral in the numerator. We have by the compatibility of the quenched kernels that
\[
\mu_{\hat{\sigma}_{\Lambda_n}}^\sigma [\eta_{\Lambda_n} \eta_V \tilde{\eta}_{\Lambda_n^c \setminus (\Lambda_n \cup V)}](1_{\sigma_{\Lambda_n}} 1_{\sigma_V}) = \int \mu_{\hat{\sigma}_{\Lambda_n}}^\sigma [\eta_{\Lambda_n} \eta_V \tilde{\eta}_{\Lambda_n^c \setminus (\Lambda_n \cup V)}](d\tilde{\sigma}) 1_{\sigma_V} \mu_{\hat{\sigma}_{\Lambda_n}}^\sigma [\eta_{\Lambda_n} \eta_V \tilde{\eta}_{\Lambda_n^c \setminus (\Lambda_n \cup V)}](1_{\sigma_{\Lambda_n}}) \\
\leq e^{2C_i|\partial \Lambda_n|} \mu_{\hat{\sigma}_{\Lambda_n}}^\sigma [\eta_{\Lambda_n} \tilde{\eta}_{\Lambda_n^c \setminus \Lambda_n}](\sigma_{\Lambda_n}) \times \mu_{\hat{\sigma}_{\Lambda_n}}^\sigma [\eta_{\Lambda_n} \eta_V \tilde{\eta}_{\Lambda_n^c \setminus (\Lambda_n \cup V)}](1_{\sigma_V})
\]
where the inequality follows from the uniform absolute convergence of the quenched potential, for any reference configuration $\hat{\sigma} \tilde{\eta}$.

We use that
\[
\mu_{\hat{\sigma}_{\Lambda_n}}^\sigma [\eta_{\Lambda_n} (\tilde{\eta})_{\Lambda_n^c \setminus \Lambda_n}](\sigma_{\Lambda_n}) \geq e^{-2C_i|\partial \Lambda_n|} \mu_{\hat{\sigma}_{\Lambda_n}}^\sigma [\eta_{\Lambda_n} \tilde{\eta}_{\Lambda_n^c \setminus \Lambda_n}](\sigma_{\Lambda_n})
\]
and the similar lower bound on the first disorder-integral in the denominator of (6.10) with the same reference joint reference configuration $\hat{\sigma} \tilde{\eta}$. From this we get an upper bound on (6.10) in the form of
Last we need to control the influence of the variation of the random fields inside the finite volume \( \eta_{\Lambda_n} \) on the Gibbs-expectation outside. We have that

\[
\mu_{\Lambda_n}^\sigma [\eta_{\Lambda_n} \tilde{\eta}_{\mathbb{Z}^d \setminus \Lambda_n}] (\sigma_V) \leq e^{2C_1 |\partial \Lambda_n|} \mu_{\Lambda_n}^\sigma [\eta_{\Lambda_n}^{-1} \tilde{\eta}_{\mathbb{Z}^d \setminus \Lambda_n}] (\sigma_V)
\]

for any configurations \( \eta \) and \( \eta^{-1} \) inside \( \Lambda_n \). This gives the following upper bound on (6.11)

\[
e^{8C_1 |\partial \Lambda_n|} \frac{\int \mathbb{P}(d\tilde{\eta})1_{\eta_{\Lambda_n}1_{\eta_V} \mu_{\Lambda_n}^\sigma [\eta_{\Lambda_n}^{-1} \tilde{\eta}_{\mathbb{Z}^d \setminus \Lambda_n}] (\sigma_V)}}{\int \mathbb{P}(d\tilde{\eta}_1)1_{\eta_{\Lambda_n}1_{\eta_V}} \int \mathbb{P}(d\tilde{\eta}_2)1_{\eta_V}}.
\]

But this, by the property of asymptotic decoupled of the disorder field is bounded by \( e^{8C_1 |\partial \Lambda_n|} + c_n \) and the proof of the upper bound in (6.9) is done. The proof of the lower bound is similar.

Applying Pfister’s theory [22], we have

**Corollary 6.12** Suppose \( \mathbb{P} \) is asymptotically decoupled and that \( K^\sigma \) is a corresponding translation invariant joint measure of a quenched random system, with a defining finite range potential. Then \( h(K|K^\sigma) \) exists for all translation invariant probability measures \( K \).

Moreover we have the following explicit formula:

**Theorem 6.13** Suppose that the defining potential \( \Phi(\sigma, \eta) \) is translation invariant and that \( \mathbb{P} \) is asymptotically decoupled. Suppose that \( K^\sigma \) is translation invariant joint measure constructed with the boundary condition \( \bar{\sigma} \). Suppose that \( K \) is a translation-invariant measure on the product space. Denote by \( K_d \) its marginal on the disorder variables \( \eta \). Then

\[
h(K|K^\sigma) = h(K_d|\mathbb{P}) - h(K) - h(K_d) + \sum_{A \ni 0} \frac{1}{|A|} K(A(\eta | \sigma) = \cdot) + K\left( \lim_{\Lambda} \frac{1}{|\Lambda|} \log Z^\sigma(\eta = \cdot) \right)
\]

where \( h(K) \) is the Kolmogorov-Sinai entropy (2.15).

**Remark 6.14** The third term has the meaning of the \( K \)-expectation of the ‘joint energy’. The last term is the \( K \)-mean of the “quenched pressure”. Note that it is boundary condition \( \bar{\sigma} \)-independent, of course.

**Remark 6.15** In the case that \( \mathbb{P} \) is a Gibbs distribution, the existence of the relative entropy density is obtained directly, i.e., without relying on Pfister’s theory.
We have
\[
\frac{1}{|\Lambda|} h_{\Lambda}(K|K^\sigma) = \frac{1}{|\Lambda|} \sum_{\sigma \in \tilde{\Lambda}} K(\sigma \Lambda \eta_{\Lambda}) \log K(\sigma \Lambda \eta_{\Lambda}) - \frac{1}{|\Lambda|} \sum_{\sigma \in \tilde{\Lambda}} K(\sigma \Lambda \eta_{\Lambda}) \log K^\sigma(\sigma \Lambda \eta_{\Lambda})
\]
where the first term converges to \(-h(K)\). For the second term we use the approximation
\[
\sup_{\sigma, \tilde{\sigma}, \hat{\sigma}} \left| \log \left( \frac{K^\sigma(\sigma \Lambda \eta_{\Lambda})}{P(\eta_{\Lambda})\mu^\sigma[\eta_{\Lambda} \eta_{\tilde{\Lambda}}]}(\sigma_{\Lambda})} \right) \right| \leq 2C_1|\partial|\Lambda|.
\]
First have
\[
\frac{1}{|\Lambda|} \sum_{\sigma \in \tilde{\Lambda}} K(\sigma \Lambda \eta_{\Lambda}) \log P(\eta_{\Lambda}) = \frac{1}{|\Lambda|} h_{\Lambda}(K_d|P) - \frac{1}{|\Lambda|} \sum_{\eta_{\Lambda}} K_d(\eta_{\Lambda}) \log K_d(\eta_{\Lambda}).
\]
The second term converges to \(h(K_d)\). The first term converges to \(h(K_d|P)\). This is clear either by the classical theory for the case that \(P\) is Gibbs or even independent, or by Pfister’s theory if \(P\) is asymptotically decoupled. Next, by definition
\[
\log \mu^\sigma[\eta_{\tilde{\Lambda}}|\Lambda](\sigma_{\Lambda}) = - \sum_{A: A \subseteq \Lambda \neq \emptyset} \Phi_A(\sigma A \tilde{\sigma}\Lambda \eta_{\tilde{\Lambda}}] - \log Z^\sigma_A(\eta_{\tilde{\Lambda}}|\Lambda).
\]
Using translation-invariance of the measure \(K\) we get that the application of \(1/|\Lambda| \int K(d\sigma \Lambda d\eta_{\Lambda})\) over the first sum of the r.h.s. converges to \(-\sum_{A: A \subseteq \Lambda \neq \emptyset} \int K(\Phi_A(\eta \sigma = \cdot)).\) To see that the average over the last term converges we use the ergodic decomposition of \(K_d\) to write \(K_d(\eta) = \int \rho(d\kappa)\kappa(\eta)\) where \(\rho(d\kappa)\) is a probability measure that is concentrated on the ergodic measures on \(\eta\). Fix any ergodic \(\kappa\). For \(\kappa\)-a.e. disorder configuration \(\eta\) we have the existence of the limit \(-\lim_{A: A \subseteq \Lambda \neq \emptyset} \frac{1}{|\Lambda|} \log Z^\sigma_A(\eta = \cdot),\) by standard arguments [24]. The convergence is also in \(L^1\), by dominated convergence. So we may integrate over \(\rho\) to see the statement of the theorem. 

### 6.3 Discussion of the first part of the variational principle for joint measures

To discuss the first part of the variational principle we will use an explicit representation of the conditional expectations of the joined measures. For this we need to restrict to the case that \(P\) is a product measure. First, in the situation detailed below, we prove the first part of the variational principle by direct arguments. Next, we illustrate the criteria given in the general theory of section 3.4 by showing that they can be verified in the context of joined measures in the almost Gibbsian case, giving then an alternative proof of the variational principle.

We start with the following proposition of [14].

**Proposition 6.16** Assume that \(P\) is a product measure. Assume that there is a set of realizations of \(\eta\)’s of \(P\)-measure one such that the quenched infinite-volume Gibbs measure \(\mu[\eta]\) is a weak limit of the quenched finite volume measures (6.1). Then, a version of the

\[
\text{Proof.}
\]
infinite-volume conditional expectation of the corresponding joint measure \(K^\mu(\sigma, \eta) = \mathbb{P}(d\eta) \mu[\eta](d\sigma)\) is given by the formula

\[
K^\mu[\xi_\Lambda | \xi_{\Lambda^c}] = \frac{\mu^\text{ann,\xi_\Lambda}(\xi_\Lambda)}{\int \mu^\text{ann,\xi_\Lambda}(d\tilde{\eta}_\Lambda)Q^\mu_{\Lambda}(\eta_\Lambda, \tilde{\eta}_\Lambda, \eta_{\Lambda^c})}.
\]  

(6.17)

Here \(\mu^\text{ann,\xi_\Lambda}(\xi_\Lambda)\) is the trivial annealed local specification given by in terms of the potential \(U^\text{triv}_A(\sigma, \eta) = \Phi_A(\sigma, \eta) - 1_{A=\{x\}} \log \mathbb{P}_0(\eta_x)\) w.r.t counting measure on the product space.

Further we have put

\[
Q^\mu_{\Lambda}(\eta_\Lambda^1, \eta_\Lambda^2, \eta_{\Lambda^c}) = \mu[\eta_\Lambda^2 \eta_{\Lambda^c}] (e^{-\Delta H_\Lambda(\eta_\Lambda^1, \eta_\Lambda^2, \eta_{\Lambda^c})})
\]

where

\[
\Delta H_\Lambda(\eta_\Lambda^1, \eta_\Lambda^2, \eta_{\Lambda^c})(\sigma) = \sum_{A \cap \Lambda \neq \emptyset} \left( \Phi_A(\eta_\Lambda^1 \eta_{\Lambda^c} \sigma) - \Phi_A(\eta_\Lambda^2 \eta_{\Lambda^c} \sigma) \right).
\]

According to our assumption on the measurability on \(\mu[\eta]\), \(Q_x\) depends measurably on \(\eta_{\not\in\Lambda \setminus \Lambda^c}\). We fix a version of the map and define the r.h.s. of (6.17) to be the specification \(\gamma^\mu\). Note that for the random field Ising model, this specification exists for all random field configurations by monotonicity.

In this context we always have the first part of the variational principle. Note that we do not need any further assumption about almost Gibbsianness.

**Theorem 6.18** Assume that \(\mathbb{P}\) is a product measure. There exists a constant \(C\) depending only on \(\Phi, \mathbb{P}\) such that for any \(K, K' \in \mathcal{G}(\gamma^\mu)\) one has

\[
\sup_{\xi} \left| \log \frac{K(\xi_\Lambda)}{K'(\xi_\Lambda)} \right| \leq C|\partial \Lambda|.
\]

In particular \(h(K|K') = h(K'|K) = 0\).

**Proof.** Using \(K, K' \in \mathcal{G}(\gamma^\mu)\), it suffices to show that we have the estimate

\[
\frac{\gamma^\mu(\xi_\Lambda|\xi_{\Lambda^c})}{\gamma^\mu(\xi_\Lambda'|\xi_{\Lambda^c'})} \leq e^{C|\partial \Lambda|}
\]

where the constant \(C\) is independent of \(\Lambda, \xi, \xi'\). From the explicit representation (6.17) we obtain

\[
\frac{\gamma^\mu(\xi_\Lambda|\xi_{\Lambda^c})}{\gamma^\mu(\xi_\Lambda'|\xi_{\Lambda^c'})} = \frac{\mu^\text{ann,\xi_\Lambda}(\xi_\Lambda)}{\mu^\text{ann,\xi_\Lambda'}(\xi_\Lambda')} \frac{\int \mu^\text{ann,\xi_\Lambda}(d\tilde{\eta}_\Lambda)Q^\mu_{\Lambda}(\eta_\Lambda, \tilde{\eta}_\Lambda, \eta_{\Lambda^c})}{\int \mu^\text{ann,\xi_\Lambda'}(d\tilde{\eta}_\Lambda')Q^\mu_{\Lambda'}(\eta_\Lambda, \tilde{\eta}_\Lambda', \eta_{\Lambda^c'})}.
\]

(6.19)

Using the definition of \(\mu^\text{ann,\xi_\Lambda}\) and using the finite range assumption on \(\Phi\), we obtain the bound \(e^{C|\partial \Lambda|}\) for the first factor on the r.h.s. of (6.19). The second factor on the r.h.s. of (6.19) is bounded by

\[
\left( \sup_{\tilde{\eta}_\Lambda} \frac{Q^\mu_{\Lambda}(\eta_\Lambda, \tilde{\eta}_\Lambda, \eta_{\Lambda^c})}{Q^\mu_{\Lambda'}(\eta_\Lambda, \tilde{\eta}_\Lambda', \eta_{\Lambda^c'})} \right) \frac{\int \mu^\text{ann,\xi_\Lambda}(d\tilde{\eta}_\Lambda)Q^\mu_{\Lambda}(\eta_\Lambda, \tilde{\eta}_\Lambda, \eta_{\Lambda^c})}{\int \mu^\text{ann,\xi_\Lambda}(d\tilde{\eta}_\Lambda)Q^\mu_{\Lambda}(\eta_\Lambda, \tilde{\eta}_\Lambda, \eta_{\Lambda^c})}.
\]
Using the same argument on $\mu_\Lambda^{ann,\xi_0}$ again, we see that the second factor is bounded by $e^{c|\partial\Omega|}$. To estimate the first factor, remind the explicit expression

$$Q^\mu_\Lambda(\eta_\Lambda, \bar{\eta}_\Lambda, \eta_{\Lambda^c}) = \mu_\Lambda[\bar{\eta}_\Lambda \eta_{\Lambda^c}] \left( e^{-\Delta H_\Lambda(\eta_\Lambda, \bar{\eta}_\Lambda, \eta_{\Lambda^c})} \right) \leq e^{c|\partial\Omega|} \mu_\Lambda[\bar{\eta}_\Lambda \eta_{\Lambda^c}] \left( e^{-\Delta H_\Lambda(\eta_\Lambda, \bar{\eta}_\Lambda, \eta_{\Lambda^c})} \right).$$

Here the inequality follows from the definition of $H_\Lambda$ and the finite range property of $\Phi$. Now use the definition of the quenched kernels and once again the finite range of $\Phi$ to see that the last expectation is bounded from above by

$$e^{c|\partial\Omega|} \mu_\Lambda[\bar{\eta}_\Lambda \eta_{\Lambda^c}'] \left( e^{-\Delta H_\Lambda(\eta_\Lambda, \bar{\eta}_\Lambda, \eta_{\Lambda^c}')} \right) = Q^\mu_\Lambda(\eta_\Lambda, \bar{\eta}_\Lambda, \eta_{\Lambda^c}).$$

This finishes the proof. \hfill \qed}

Let us now check what can be said about the criteria for the first part of the variational principle for joint measures. It turns out that it is natural to use the criteria given in section 3.4 with a measure $\lambda$ that is not a Dirac measure. Instead, let us take any translation invariant configuration $\sigma^0$ and put $\lambda := P \otimes \delta_{\sigma^0}$.

First, using the arguments given in the proof of Theorem 6.13 it is simple to see the following.

**Proposition 6.20** Suppose that the defining potential $\Phi(\eta)$ is translation invariant. Suppose that $K^\sigma$ is translation invariant joint measure constructed with the boundary condition $\bar{\sigma}$. Then

$$e_{K^\sigma}^\lambda = -h(P) + \sum_{A \ni 0} P(d\eta) \frac{\Phi_A(\eta \sigma^0)}{|A|} + \int P(d\eta) \lim_{\Lambda \uparrow \mathbb{Z}^d} \frac{1}{|A|} \log Z^\Lambda_\lambda(\eta)$$

exists.

Put

$$\mathcal{H}_\mu := \{ \eta \in \mathcal{H}, \eta \mapsto Q^\mu_\Lambda(\eta_x, \eta_x^1, \eta_x^2) \text{ is continuous } \forall x, \eta_x^1, \eta_x^2 \},$$

then we have that $\sigma \eta \in \Omega_\gamma \Leftrightarrow \eta \in \mathcal{H}_\mu$. Assume that $P[\mathcal{H}_\mu] = 1$. Then any joint measure is almost Gibbs. This was pointed out and discussed in the papers [13, 14] and is apparent from the above representation of the conditional expectation.

Let us remark that, whenever $K$ is a translation-invariant probability measure on the product space and $K^\sigma$ is any joint measure with $K^\sigma(d\eta) = P(d\eta)$ we have that $K(d\eta) \neq P(d\eta) \Rightarrow h(K|K^\sigma) > 0$. This is clear from the monotonicity of the relative entropy w.r.t. to the filtration. ([9], Proposition 15.5 c). So, $h(K|K^\sigma) = 0$ would imply that $h(K_\eta|P) = 0$ which again would imply $K_\eta = P$, by the classical variational principle applied to the product measure $P$. So, given a joint measure $K^\sigma$, the class of interesting measures is reduced to the ones having the same $\eta$-marginal.

**Proposition 6.21** Suppose that $P$ is a product measure and that $\gamma^\mu$ is the above specification for a translation-invariant joint measure $K^\mu$. Suppose that $P[\mathcal{H}_\mu] = 1$. Take $K$ a translation-invariant measure with marginal $K_d = P$.

Then condition $C'1$ holds for the measure $K$, for the above choice of $\lambda$. 

26
Proof. We have to check that \( \lambda (d\sigma^1 d\eta^1) K(d\sigma^2 d\eta^2) \)-a.s a configuration \( \sigma^1 _{<0} \eta^1 _{<0} \sigma^2 _{>0} \eta^2 _{>0} \) is in \( \Omega_{\lambda,\mu} \), where for a configuration \( \sigma \) we have written \( \sigma _{<0} = (\sigma _x)_{x<0} \), etc. This is equivalent to \( \eta^1 _{<0} \eta^2 _{>0} \in \mathcal{H}_\mu \) for \( \mathbb{P} \otimes \mathbb{P} \)-a.e. \( \eta^1, \eta^2 \), since both \( \lambda \) and \( K \) have marginal \( \mathbb{P} \), and the later is immediate because it is a product measure.

To illustrate the general theory of section 3.4 we note the following

**Corollary 6.22** Suppose that \( \mathbb{P} \) is a product measure and that \( \gamma^\mu \) is the above specification for a translation-invariant joint measure \( K^\mu \). Suppose that \( \mathbb{P}(\mathcal{H}_\mu) = 1 \). Take \( K \in \mathcal{G}^{\text{inv}}(\gamma^\mu) \) with marginal \( K_d = \mathbb{P} \).

Then condition C’2 of Theorem 3.17 is true and hence

\[
\lim_{\lambda} \frac{1}{|\Lambda|} \int \mathbb{P}(d\eta) \log \frac{K(\eta \lambda \sigma^\lambda_\Lambda)}{K^\mu(\eta \lambda \sigma^\lambda_\Lambda)} = 0
\]

for any translation invariant spin-configuration \( \sigma^0 \).

### 6.4 Random field Ising model: failure of the second part of the variational principle

Let us now specialize to the random field Ising model.

For all what follows we will denote by \( K^+(d\sigma d\eta) = \mathbb{P}(d\eta) \mu^+[\eta](d\sigma) \) the ‘plus-joint measure’. Here we clearly mean by \( \mu^+[\eta](d\sigma) = \lim_{A \uparrow \Lambda} \mu^+[\eta](d\sigma) \) the random infinite volume Gibbs measure on the Ising spins. The limit exists for any arbitrary fixed \( \eta \), by monotonicity. Similarly we write \( K^- (d\sigma d\eta) = \mathbb{P}(d\eta) \mu^-[\eta](d\sigma) \). In this situation we have

**Proposition 6.23** Assume that the quenched random field Ising model has a phase transition in the sense that \( \mu^+[\eta](\sigma_x = +) > \mu^-[\eta](\sigma_x = +) \) for \( \mathbb{P} \)-a.e. \( \eta \). Then the joint measures \( K^+ \) and \( K^- \), obtained with the same defining potential are not compatible with the same specification.

**Remark 6.24** We already know by Theorem 6.2 that the relative entropy \( h(K^+|K^-) \) is zero, and thus we prove here that the second part of the variational principle is not valid in case of phase transition for the quenched random field Ising model.

**Proof.** The proof relies on the explicit representation of proposition 6.16 for the conditional expectations of \( K^+ \) (resp. \( K^- \)) in terms of \( \mu^+ \) (resp. \( \mu^- \)). We will show that

\[
\int K^+(d\xi_x) K^-(-|\xi_x) \neq K^+(\cdot). \quad \text{Let us evaluate both sides on the event } B := \{ \eta_x = +, \sum_{y:|y-x|=1} \sigma_y = 0 \}.
\]

Using proposition 6.16, it is simple to see that we have in particular for the local event \( \eta_x = + \) for any configuration \( \sigma \) with \( \sum_{y:|y-x|=1} \sigma_y = 0 \) the formula

\[
K^+(\eta_x = +|\sigma_x, \eta_x) = \left( 1 + \int \mu^+[\eta_x = -, \eta_x](d\bar{\sigma}_x) e^{2\bar{h}\bar{\sigma}} \right)^{-1} =: r^+(\eta_x).
\]

So we get that

\[
K^+(B) = \int \mathbb{P}(d\eta) \mu^+[\eta] \left( \sum_{y:|y-x|=1} \sigma_y = 0 \right) \times r^+(\eta_x).
\]

27
Define \( r^-(\eta_{x^c}) \) as above, but with the Gibbs measure \( \mu^- \). Then we have

\[
\int K^+(d\xi_{x^c})K^-(\cdot|\xi_{x^c})(B) = \int \mathbb{P}(d\tilde{\eta})\mu^+ |\tilde{\eta}\left( \sum_{y:y=x^c=1} \sigma_y = 0 \right) \times r^-(\tilde{\eta}_{x^c}).
\]

Now it follows from our assumption that, for \( \mathbb{P}\text{-a.e. configuration } \tilde{\eta} \) we have the strict inequality \( r^+(\tilde{\eta}_{x^c}) > r^-(\tilde{\eta}_{x^c}) \). But this shows that both measures give different expectations of \( B \) and finishes the claim. 

In the following we show from the weakly Gibbsian point of view that \( K^+ \) and \( K^- \) have a “good” (rapidly decaying) almost surely converging translation invariant potential. This strengthens the results in [14], where the a.s. absolutely convergent potential is not translation invariant.

**Theorem 6.25** Assume that \( d \geq 3 \), \( \beta \) is large enough, the random fields \( \eta_x \) are i.i.d. with symmetric distribution that is concentrated on finitely many values, and that \( h\mathbb{E}\eta_x^2 \) is sufficiently small.

There exists an absolutely convergent potential that is translation invariant for the plus joint measure \( K^+(d\eta|\eta) \) for sufficiently low temperature and small disorder. It decays like a stretched exponential.

**Proof.** Applying the remark given after (5.5) that rely on Theorem 2.4. of [14] we have the following.

**Fact proved in [14].**

Assume that \( K^\mu(d\xi) = \mathbb{P}(d\eta)\mu[\eta](d\sigma) \) is a joint measure for the random field Ising model. Denote the disorder average of the quenched spin-spin correlation by

\[
c(m) := \sup_{x,y:|x-y|=m} \int \mathbb{P}(d\eta) |\mu[\eta]|(\sigma_x \sigma_y) - \mu[\eta](\sigma_x) \mu[\eta](\sigma_y)|.
\]

Suppose we give ourselves any nonnegative translation invariant function \( w(A) \) giving weight to a subset \( A \subset \mathbb{Z}^d \).

Then there is a potential \( \bar{U}^\mu(\eta) \) on the disorder space satisfying the decay property

\[
\sum_{A: A \ni x_0} w(A) \int \mathbb{P}(d\eta) |\bar{U}_A^\mu(\eta)| \leq \bar{C}_1 + \bar{C}_2 \sum_{m=2}^\infty m^{2d-1} \bar{w}(m)c(m)
\]

if the r.h.s. is finite. Here \( \bar{w}(m) := w(\{z \in \mathbb{Z}^d; z \geq 0, |z| \leq m\}) \) where \( \geq \) denotes the lexicographic order. \( \bar{C}_1, \bar{C}_2 \) are constants, depending on \( \beta, h \). If \( K^\mu \) is translation invariant, then \( \bar{U}^\mu(\eta) \) is translation invariant, too. The total potential \( U^{triv}(\sigma, \eta) + \bar{U}^\mu(\eta) \) is a potential for \( K^\mu \). Here \( U^{triv} \) is a potential for the formal Hamiltonian \( -\beta \sum_{<i,j>} \sigma_i \sigma_j - h \sum_i \eta_i \sigma_i - \sum_i \log \mathbb{P}_0(\eta_i) \).

It was already stated in [14] that we expect a superpolynomial decay of the quantity \( c(m) \) with \( m \), when \( m \) tends to infinity. We remark first that in [1] it was already stated and proved that \( |\mu[\eta]|(\sigma_x \sigma_y) - \mu[\eta](\sigma_x) \mu[\eta](\sigma_y)| \leq C(\eta)e^{-C|\beta|d(x,y)} \) with a random constant
\( C(\eta) \) that is finite for \( \mathbb{P} \)-a.e. \( \eta \). The problem is that integrability of the constant is not to be expected. Unfortunately, [1] do not control explicitly in their paper the decay of the disorder average \( c(m) \). Now we will reenter their renormalization group proof and sketch how stretched exponential decay is obtained for \( c(m) \). Obviously, we cannot repeat the details of the RG analysis here.

\textbf{Corollary 6.26} (From [1]) There is an exponent \( \alpha > 0 \) such that, for all \( m \) sufficiently large we have that

\[ c(m) \leq e^{-m^\alpha}. \]  

(6.27)

\textit{Sketch of proof based on RG:}

For the first part we follow Bricmont-Kupiainen, page 750, 8.3 ‘Exponential Decay of Correlations.’ Fix \( x \) and \( y \). We will be interested in sending their distance to infinity. Let us denote by \( H \subset \mathbb{Z}^d \) the half space \( H := \{ z \in \mathbb{Z}^d, e \cdot z \leq a \} \) for \( a > 0 \), where \( e \) is a fixed unit vector. Let us denote by \( \mu_H^+[\eta] := \lim_{\Lambda \uparrow H} \mu^+_\Lambda[\eta] \). By monotonicity we have for any configuration of random fields \( \eta \) that the quenched expectation of the spin at the origin in the measure \( \mu_a^+[\eta] \) is bigger than that in the measure \( \mu^+[\eta] \).

Repeating the FKG-arguments given in the first steps of [1] Chapter 8.3., it is sufficient to show stretched exponential decay of the quantity

\[ \int \mathbb{P}(d\eta)(\mu_H^+[\eta](\sigma_0) - \mu^+[\eta](\sigma_0)) \]

as a function of \( d(H^c, 0) \) to prove (6.27). As in [1] we denote by \( E_H \) the “good” event in spin-space in all of \( \mathbb{Z}^d \) that there is no Peierls contour around 0 in that touches the complement of \( H \). Then, in the same configuration \( \eta \), we have that the r.h.s. is bounded by

\[ \mu_H^+[\eta](\sigma_0) - \mu^+[\eta](\sigma_0) \leq \mu^+[\eta](E_H^c). \]

Now, we can always estimate this expectation as a sum over probabilities of Peierls contours

\[ \mu^+[\eta](E_H^c) \leq \sum_{\gamma \in \text{int} \gamma^0 \cap H^c \neq \emptyset} \mu^+[\eta](\gamma). \]

The problem is that there is no uniform Peierls estimate for all configurations of the disorder. There is however a “good event” in disorder space \( G = G_H \) such that there really is a Peierls estimate for all the “long” contours appearing in the above sum. The \( \mathbb{P} \)-probability of the complement of this event is small and controlled (in a very-nontrivial way) by the renormalization group construction. For \( \eta \in G_H \) we really have that

\[ \sum_{\gamma \in \text{int} \gamma^0 \cap H^c \neq \emptyset} \mu^+[\eta](\gamma) \leq e^{-C\beta d(H^c, 0)}. \]

This is stated as (8.34) in [1]. So we have that

\[ \int \mathbb{P}(d\eta)\mu^+[\eta](E_H^c) \leq \mathbb{P}(G^c) + e^{-C\beta d(H^c, 0)}. \]
From the construction of the renormalization group in Bricmont-Kupiainen we can see that $G$ is expressable in the so-called bad fields $N^k_x(\eta)$ in the form $G = \{\eta, N^k_x(\eta) = 0 \quad \forall x < L, \forall k > \frac{\log d(x, H^c)}{\log L} \}$. $L$ is a fixed finite length scale (the block-length suitably chosen in the construction of the RG). It appears here just as a constant. $x \in \mathbb{Z}^d$ runs over sites in the lattice and $k$ is a natural number denoting the $k$-the application of the renormalization group transformation. The renormalization group gives the probabilistic control of the form

$$P(N^k_x(\eta) \neq 0) \leq e^{-Lr_1k}$$

with some $r_1 > 0$ (this follows from Lemma 1 and Lemma 2, page 563 in [1]) and so we have

$$P(G^c_H) \leq L^d \sum_{k > \log d(0,H^c)} e^{-Lr_1k} \leq L^d e^{-d(0,H^c)r_2}$$

for $d(0,H^c)$ sufficiently large with $r_1 > r_2 > 0$. This proves the claim. \hfill \Box

References


