#### **Gradient Gibbs measures**

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### **Abstract**

In the talk we define (gradient) Gibbs measures of physical models with a countable set of spin values. For SOS (solid-on-solid) model, with spin values from the set of all integers, on a Cayley tree we give some gradient Gibbs measures (GGMs) of the model. Such a measure corresponds to a boundary law (a function defined on vertices of Cayley tree) satisfying a functional equation. In the ferromagnetic SOS case we give several concrete GGMs which correspond to periodic boundary laws.

## **Preliminaries**

 $\sigma$ -algebra, Hamiltonian. The study of random functions  $\xi_X$  from a lattice  $\mathbb{L}$  (usually  $\mathbb{Z}^d$  or  $\Gamma^k$ ) to a measure space  $(E,\mathcal{E})$  is a central component of ergodic theory and statistical physics. In many classical models from physics (e.g., the Ising model, the Potts model), E is a finite set (i.e., with a finite underlying measure  $\lambda$ ), and  $\xi_X$  has a physical interpretation as the spin of a particle at location X in a crystal lattice.

Since 2018 we interested to the models, where  $(E, \mathcal{E})$  is a space with an infinite underlying measure  $\lambda$  (i.e.  $\mathbb{L}$  with counting measure) where  $\mathcal{E}$  is the Borel  $\sigma$ -algebra of E and  $\xi_X$  usually has a physical interpretation as the spatial position of a particle at location x in a lattice.

First such models were considered in<sup>1</sup>.

The prime examples of unbounded spin systems are harmonic oscillators. Another example is the Ginzburg-Landau interface model; which is obtained from the anharmonic oscillators<sup>2</sup>

<sup>&</sup>lt;sup>1</sup>Funaki, T., Spohn, H. **Comm. Math. Phys**. 185 (1997), no. 1, 1–36.

<sup>&</sup>lt;sup>2</sup>H.O. Georgii, Gibbs Measures and Phase Transitions, Berlin, 2011

Denote by  $\Omega$  the set of functions from  $\mathbb{L}$  to E, such a function also is called a configuration.

Assume random field  $(\xi_x)_{x\in\mathbb{L}}$  on  $\Omega$  given as the projection onto the coordinate  $x\in\mathbb{L}$ :

$$\xi_{\mathsf{X}}(\omega) = \omega(\mathsf{X}) = \omega_{\mathsf{X}}, \ \omega \in \Omega.$$

If  $\Lambda \subset \mathbb{L}$ , we denote by  $\mathcal{F}_{\Lambda}$  the smallest  $\sigma$ -algebra with respect to which  $\xi_X$  is measurable for all  $X \in \Lambda$ . We write  $\mathcal{T}_{\Lambda} = \mathcal{F}_{\mathbb{L} \setminus \Lambda}$ .

A subset of  $\Omega$ , is called a cylinder set if it belongs to  $\mathcal{F}_{\Lambda}$  for some finite set  $\Lambda \subset \mathbb{L}$ .

Let  $\mathcal F$  be the smallest  $\sigma$ -algebra on  $\Omega$  containing the cylinder sets.

We write  $\mathcal{T}$  for the tail- $\sigma$ -algebra, i.e., intersection of  $\mathcal{T}_{\Lambda}$  over all finite subsets  $\Lambda$  of  $\mathbb{L}$  the sets in  $\mathcal{T}$  are called tail-measurable sets.



Assume that we are given a family of measurable potential functions  $\Phi_{\Lambda}:\Omega\to\mathbb{R}\cup\{\infty\}$  (one for each finite subset  $\Lambda$  of  $\mathbb{L}$ ) each  $\Phi_{\Lambda}$  is  $\mathcal{F}_{\Lambda}$  measurable.

For each finite subset  $\Lambda$  of  $\mathbb{L}$  we also define a Hamiltonian:

$$\mathcal{H}_{\Lambda}(\sigma) = \sum_{\stackrel{\Delta \subset \mathbb{L}:}{\Delta \cup \Lambda 
eq \emptyset}} \Phi_{\Delta}(\sigma),$$

where the sum is taken over finite subsets  $\Delta$ .

# Gibbs Measures.

To define Gibbs measures and gradient Gibbs measures, we will need some additional notation<sup>3</sup>.

Let  $(X, \mathcal{X})$  and  $(Y, \mathcal{Y})$  be general measure spaces.

A function  $\pi: \mathcal{X} \times Y \to [0, \infty]$  is called a probability kernel from  $(Y, \mathcal{Y})$  to  $(X, \mathcal{X})$  if

- 1.  $\pi(\cdot|y)$  is a probability measure on  $(X, \mathcal{X})$  for each fixed  $y \in Y$ , and
- 2.  $\pi(A|\cdot)$  is  $\mathcal{Y}$ -measurable for each fixed  $A \in \mathcal{X}$ .

Such a kernel maps each measure  $\mu$ , on  $(Y, \mathcal{Y})$  to a measure  $\mu\pi$  on  $(X, \mathcal{X})$  by

$$\mu\pi(A) = \int \pi(A|\cdot)d\mu$$

<sup>&</sup>lt;sup>3</sup>S. Sheffield, Random surfaces: Large deviations principles and gradient Gibbs measure classifications. Thesis (Ph.D.) Stanford University: 2003.

The following is a probability kernel from  $(\Omega, \mathcal{T}_{\Lambda})$  to  $(\Omega, \mathcal{F})$ :

$$\gamma_{\Lambda}(A,\omega) = Z_{\Lambda}(\omega)^{-1} \int \exp(-H_{\Lambda}(\sigma_{\Lambda}\omega_{\Lambda^c})) \mathbf{1}_{A}(\sigma_{\Lambda}\omega_{\Lambda^c}) \nu^{\otimes \Lambda}(d\sigma_{\Lambda}),$$

where  $\nu = \{\nu(i) > 0, i \in E\}$  is a counting measure. We say  $\sigma$  has finite energy if  $\Phi_{\Lambda}(\sigma) < \infty$  for all finite  $\Lambda$ . We say  $\sigma$  is  $\Phi$ -admissible if each  $Z_{\Lambda}(\sigma)$  is finite and non-zero. Given a measure  $\mu$  on  $(\Omega, \mathcal{F})$ , we define a new measure  $\mu\gamma_{\Lambda}$  by

$$\mu \gamma_{\Lambda}(A) = \int \gamma_{\Lambda}(A, \cdot) d\mu$$

**Definition 1.** A probability measure  $\mu$  on  $(\Omega, \mathcal{F})$  is called a Gibbs measure if  $\mu$  is supported on the set of  $\Phi$ -admissible configurations in  $\Omega$  and for all finite subset  $\Lambda$  we have

$$\mu \gamma_{\Lambda} = \mu$$
.



### **Gradient Gibbs measure**

For any configuration  $\omega = (\omega(x))_{x \in \mathbb{L}} \in E^{\mathbb{L}}$  and edge  $b = \langle x, y \rangle$  of  $\mathbb{L}$  the *difference* along the edge b is given by  $\nabla \omega_b = \omega_y - \omega_x$  and we also call  $\nabla \omega$  the *gradient field* of  $\omega$ .

The gradient spin variables are now defined by  $\eta_{\langle x,y\rangle}=\omega_y-\omega_x$  for each  $\langle x,y\rangle$ .

The space of *gradient configurations* denoted by  $\Omega^{\nabla}$ . The measurable structure on the space  $\Omega^{\nabla}$  is given by  $\sigma$ -algebra

$$\mathcal{F}^{\nabla} := \sigma(\{\eta_{b} \mid b \in \mathbb{L}\}).$$

Note that  $\mathcal{F}^{\nabla}$  is the subset of  $\mathcal{F}$  containing those sets that are invariant under translations  $\omega \to \omega + c$  for  $c \in \mathcal{E}$ . Similarly, we define

$$\mathcal{T}^\nabla_\Lambda = \mathcal{T}_\Lambda \cap \mathcal{F}^\nabla, \ \mathcal{F}^\nabla_\Lambda = \mathcal{F}_\Lambda \cap \mathcal{F}^\nabla.$$



Let  $\Phi$  be an translation invariant gradient potential. Since, given any  $A \in \mathcal{F}^{\nabla}$ , the kernels  $\gamma_{\Lambda}^{\Phi}(A,\omega)$  are  $\mathcal{F}^{\nabla}$ -measurable functions of  $\omega$ , it follows that the kernel sends a given measure  $\mu$  on  $(\Omega, \mathcal{F}^{\nabla})$  to another measure  $\mu\gamma_{\Lambda}^{\Phi}$  on  $(\Omega, \mathcal{F}^{\nabla})$ .

**Definition 2.** A measure  $\mu$  on  $(\Omega, \mathcal{F}^{\nabla})$  is called a gradient Gibbs measure if for all finite subset  $\Lambda$  we have

$$\mu \gamma_{\Lambda}^{\Phi} = \mu.$$

Note that, if  $\mu$  is a Gibbs measure on  $(\Omega, \mathcal{F})$ , then its restriction to  $\mathcal{F}^{\nabla}$  is a gradient Gibbs measure.

A gradient Gibbs measure is said to be localized or smooth if it arises as the restriction of a Gibbs measure in this way. Otherwise, it is non-localized or rough.

It is known<sup>4</sup> that many natural Gibbs measures are rough when  $d \in \{1, 2\}$ .

<sup>&</sup>lt;sup>4</sup>H.O. Georgii, *Gibbs Measures and Phase Transitions*, Berlin, 2011

# Construction of gradient Gibbs measure on Cayley tree.

Following<sup>5</sup> we consider models where spin-configuration  $\omega$  is a function from the vertices of the Cayley tree  $\Gamma^k = (V, \vec{L})$  to the set  $E = \mathbb{Z}$ .

For nearest-neighboring interaction potential  $\Phi = (\Phi_b)_b$ , where bonds are denoted  $b = \langle x, y \rangle$ , define symmetric transfer matrices  $Q_b$  by

$$Q_b(\omega_b) = e^{-\left(\Phi_b(\omega_b) + |\partial x|^{-1}\Phi_{\{x\}}(\omega_x) + |\partial y|^{-1}\Phi_{\{y\}}(\omega_y)\right)}.$$

Define the Markov (Gibbsian) specification as

$$\gamma_{\Lambda}^{\Phi}(\sigma_{\Lambda} = \omega_{\Lambda}|\omega) = (Z_{\Lambda}^{\Phi})(\omega)^{-1} \prod_{b \cap \Lambda \neq \emptyset} Q_b(\omega_b).$$

<sup>&</sup>lt;sup>5</sup>C. Külske, P. Schriever. *Gradient Gibbs measures and fuzzy transformations on trees*, **Markov Process. Relat. Fields**, 23, (2017), 553-590.

If for any bond  $b=\langle x,y\rangle$  the transfer operator  $Q_b(\omega_b)$  is a function of gradient spin variable  $\zeta_b=\omega_y-\omega_x$  then the underlying potential  $\Phi$  is called a *gradient interaction potential*.

Introduce a notion of *boundary laws* that allows to describe the Gibbs measures that are Markov chains on trees.

**Definition 3.** A family of vectors  $\{I_{xy}\}_{\langle x,y\rangle\in\vec{L}}$  with  $I_{xy}=\{I_{xy}(i):i\in\mathbb{Z}\}\in(0,\infty)^{\mathbb{Z}}$  is called a *boundary law for the transfer operators*  $\{Q_b\}_{b\in\vec{L}}$  if for each  $\langle x,y\rangle\in\vec{L}$  there exists a constant  $c_{xy}>0$  such that the consistency equation

$$I_{xy}(i) = c_{xy} \prod_{z \in \partial x \setminus \{y\}} \sum_{j \in \mathbb{Z}} Q_{zx}(i,j) I_{zx}(j)$$
 (1)

holds for every  $i \in \mathbb{Z}$ .



A boundary law is called *q-periodic* if  $I_{xy}(i+q) = I_{xy}(i)$  for every oriented edge  $\langle x, y \rangle \in \vec{L}$  and each  $i \in \mathbb{Z}$ .

It is known that there is a one-to-one correspondence between boundary laws and tree-indexed Markov chains if the boundary laws are *normalisable* in the sense of Zachary<sup>6</sup>:

**Definition 4.** A boundary law *I* is said to be *normalisable* if and only if

$$\sum_{i\in\mathbb{Z}} \left( \prod_{z\in\partial x} \sum_{j\in\mathbb{Z}} Q_{zx}(i,j) I_{zx}(j) \right) < \infty$$
 (2)

at any  $x \in V$ .

<sup>&</sup>lt;sup>6</sup>S. Zachary, Countable state space Markov random fields and Markov chains on trees, Ann. Probab. 11(4) (1983), 894–903□ → ⟨⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩ → ⟨⟨⟨⟩⟩⟩⟩ → ⟨⟨⟨⟩⟩

The correspondence now reads the following:

**Theorem 1** (Zachary) For any Markov specification  $\gamma$  with associated family of transfer matrices  $(Q_b)_{b\in L}$  we have

• Each *normalisable* boundary law  $(I_{xy})_{x,y}$  for  $(Q_b)_{b\in L}$  defines a unique tree-indexed Markov chain  $\mu \in \mathcal{G}(\gamma)$  via the equation given for any connected set  $\Lambda \in \mathcal{S}$ 

$$\mu(\sigma_{\Lambda \cup \partial \Lambda} = \omega_{\Lambda \cup \partial \Lambda}) = (Z_{\Lambda})^{-1} \prod_{y \in \partial \Lambda} I_{yy_{\Lambda}}(\omega_y) \prod_{b \cap \Lambda \neq \emptyset} Q_b(\omega_b), (3)$$

where for any  $y \in \partial \Lambda$ ,  $y_{\Lambda}$  denotes the unique n.n. of y in  $\Lambda$ .

② Conversely, every tree-indexed Markov chain  $\mu \in \mathcal{G}(\gamma)$  admits a representation of the form (3) in terms of a normalisable boundary law (unique up to a constant positive factor).



#### Remark

The Markov chain  $\mu$  defined in (3) has the transition probabilities

$$P_{xy}(i,j) = \mu(\sigma_y = j \mid \sigma_x = i) = \frac{I_{yx}(j)Q_{yx}(j,i)}{\sum_{s} I_{yx}(s)Q_{yx}(s,i)}.$$
 (4)

The expressions (4) may exist even in situations where the underlying boundary law  $(I_{xy})_{x,y}$  is not normalisable. However, the Markov chain given by 4, in general, does not have an invariant probability measure. Thus, there is no obvious extension of Theorem 1 to non-normalisable boundary laws.

Therefore in<sup>7</sup> the non-normalisable boundary laws are used to give gradient Gibbs measures.

- C. Külske, P. Schriever. Markov Process. Relat. Fields, 23, (2017), 553-590.
- F. Henning, C. Külske, A. LeNy, U. Rozikov. Elect. Jour. Probab. 24, (2019), 1–23.



### SOS model.

Now we give some results of above mentioned papers. Consider a model on Cayley tree  $\Gamma^k = (V, \vec{L})$ , where the spin takes values in the set of all integer numbers  $\mathbb{Z}$ . The set of all configurations is  $\Omega := \mathbb{Z}^V$ .

The (formal) Hamiltonian of the SOS model is

$$H(\omega) = -J \sum_{\{x,y\} \in L} |\omega_x - \omega_y|, \ \omega \in \Omega,$$
 (5)

where  $J \in \mathbb{R}$  is a constant and  $\{x, y\}$  denotes nearest neighbor vertices.

Then fix a site  $w \in \Lambda$ . If the boundary law I is assumed to be q-periodic, then take  $s \in \mathbb{Z}_q$  and define probability measure  $\nu_{w,s}$  on  $\mathbb{Z}^{\{b \in \vec{L} | b \subset \Lambda\}}$  by

$$\nu_{w,s}(\eta_{\Lambda\cup\partial\Lambda} = \zeta_{\Lambda\cup\partial\Lambda}) =$$

$$Z_{w,s}^{\Lambda} \prod_{y \in \partial\Lambda} I_{yy_{\Lambda}} \left( T_q(s + \sum_{b \in \Gamma(w,y)} \zeta_b) \right) \prod_{b \cap \Lambda \neq \emptyset} Q_b(\zeta_b),$$

where  $Z_{w,s}^{\Lambda}$  is a normalization constant,  $\Gamma(w,y)$  is the unique path from w to y and  $T_q: \mathbb{Z} \to \mathbb{Z}_q$  denotes the coset projection. In<sup>8</sup> the following theorem is proved:

<sup>&</sup>lt;sup>8</sup>C. Külske, P. Schriever. **Markov Process. Relat. Fields**, 23, (2017), 553-590.

**Theorem 2.** (Külske, Schriever) Let  $(I_{\langle xy \rangle})_{\langle x,y \rangle \in \vec{L}}$  be any q -periodic boundary law to some gradient interaction potential. Fix any site  $w \in V$  and any class label  $s \in \mathbb{Z}_q$ . Then

$$\nu_{w,s}(\eta_{\Lambda\cup\partial\Lambda} = \zeta_{\Lambda\cup\partial\Lambda}) =$$

$$Z_{w,s}^{\Lambda} \prod_{y \in \partial\Lambda} I_{yy_{\Lambda}} \left( T_q(s + \sum_{b \in \Gamma(w,y)} \zeta_b) \right) \prod_{b \cap \Lambda \neq \emptyset} Q_b(\zeta_b), \tag{6}$$

gives a consistent family of probability measures on the gradient space  $\Omega^{\nabla}$ . Here  $\Lambda$  with  $w \in \Lambda \subset V$  is any finite connected set,  $\zeta_{\Lambda \cup \partial \Lambda} \in \mathbb{Z}^{\{b \in \vec{L} | b \subset (\Lambda \cup \partial \Lambda)\}}$  and  $Z_{w,s}^{\Lambda}$  is a normalization constant. The measures  $\nu_{w,s}$  will be called pinned gradient measures.

If q-periodic boundary law and the underlying potential are translation invariant then it is possible to obtain probability measure  $\nu$  on the the gradient space by mixing the pinned gradient measures:

**Theorem 3.**(Külske, Schriever) Let a q-periodic boundary law l and its gradient interaction potential are translation invariant. Let  $\Lambda \subset V$  be any finite connected set and let  $w \in \Lambda$  be any vertex. Then the measure  $\nu$  with marginals given by

$$\nu(\eta_{\Lambda\cup\partial\Lambda}=\zeta_{\Lambda\cup\partial\Lambda})=Z_{\Lambda}\left(\sum_{s\in\mathbb{Z}_{q}}\prod_{y\in\partial\Lambda}I(s+\sum_{b\in\Gamma(w,y)}\zeta_{b})\right)\prod_{b\cap\Lambda\neq\emptyset}Q(\zeta_{b}),$$
(7)

where  $Z_{\Lambda}$  is a normalisation constant, defines a translation invariant gradient Gibbs measure on  $\Omega^{\nabla}$ .

Now using Theorem 3 and following<sup>9</sup> we give some gradient Gibbs measures.

Let  $\beta>0$  be inverse temperature and  $\theta:=\exp(-\beta)<1$ . The transfer operator Q then reads  $Q(i-j)=\theta^{|i-j|}$  for any  $i,j\in\mathbb{Z}$ , and a translation invariant boundary law, denoted by  $\mathbf{z}$ , is any positive function on  $\mathbb{Z}$  solving the consistency equation, whose values we will denote by  $z_i$  instead of z(i). By definition of the boundary law it is only unique up to multiplication with any positive prefactor. Hence we may choose this constant in a way such that we have  $z_0=1$ .

Set  $\mathbb{Z}_0:=\mathbb{Z}\setminus\{0\}.$  Then the boundary law equation reads

$$z_{i} = \left(\frac{\theta^{|i|} + \sum_{j \in \mathbb{Z}_{0}} \theta^{|i-j|} z_{j}}{1 + \sum_{j \in \mathbb{Z}_{0}} \theta^{|j|} z_{j}}\right)^{k}, \quad i \in \mathbb{Z}_{0}.$$
 (8)

<sup>&</sup>lt;sup>9</sup>F. Henning, C. Külske, A. LeNy, U. Rozikov. **Elect. Jour. Probab.** 24, (2019), 1–23.

Let  $\mathbf{z}(\theta) = (z_i = z_i(\theta), i \in \mathbb{Z}_0)$  be a solution to (8). Denote

$$I_{i} \equiv I_{i}(\theta) = \sum_{j=-\infty}^{-1} \theta^{|i-j|} z_{j}, \quad r_{i} \equiv r_{i}(\theta) = \sum_{j=1}^{\infty} \theta^{|i-j|} z_{j}, \quad i \in \mathbb{Z}_{0}.$$
 (9)

Note that each  $I_i$  and  $r_i$  can be a finite positive number or  $+\infty$ . **Lemma.** For each  $i \in \mathbb{Z}_0$  we have

- $I_i < +\infty$  if and only if  $I_0 < +\infty$ ;
- $r_i < +\infty$  if and only if  $r_0 < +\infty$ .

In what follows, we will assume that  $l_0 < +\infty$  and  $r_0 < +\infty$ .

Denote  $u_i = \sqrt[k]{z_i}$  and assume  $u_0 = 1$ .

Then we have the following

**Proposition.** If  $z_0 = 1$  (i.e.  $u_0 = 1$ ) then the equation (8) is equivalent to the following

$$u_i^k = \frac{u_{i-1} + u_{i+1} - \tau u_i}{u_{-1} + u_1 - \tau}, \quad i \in \mathbb{Z},$$
 (10)

where  $\tau = \theta^{-1} + \theta$ .

The following theorem is proved for k = 2 and 4-periodic boundary laws:

**Theorem 4.** For the SOS model (5) on the binary tree (i.e. k = 2) with parameter  $\tau = \theta + \theta^{-1}$  the following assertions hold

- 1. If  $\tau \leq$  4 then there is precisely one GGM associated to a 4-periodic boundary law.
- 2. If  $4 < \tau \le 6$  then there are precisely two GGMs.
- 3. If  $6 < \tau < 2 + 2\sqrt{5}$  then there are precisely three GGMs.
- 4. If  $\tau \ge 2 + 2\sqrt{5}$  then there are precisely four such measures.



The following theorem is proved for any  $k \ge 2$  and 3-periodic boundary laws.

Denote

$$\tau_0:=\frac{2k+1}{k-1}.$$

**Theorem 5.** For the SOS-model on the k-regular tree,  $k \geq 2$ , with parameter  $\tau$  there is  $\tau_c$  such that  $0 < \tau_c < \tau_0$  and the following holds:

- 1. If  $\tau < \tau_c$  then there are no GGM corresponding to nontrivial 3-periodic boundary.
- 2. At  $\tau = \tau_c$  there is a unique GGM corresponding to a nontrivial 3-periodic boundary law.
- 3. For  $\tau > \tau_c$ ,  $\tau \neq \tau_0$  (resp.  $\tau = \tau_0$ ) there are exactly two such (resp. one) GGMs.

The GGMs described above are all different from the GGMs mentioned in Theorem 4.



Thank you!