

Thermodynamic Formalism for one-dimensional spin-lattices: entropy and pressure

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Let (M, d_M) be a compact metric space. We consider the metric in $M^{\mathbb{N}}$:

$$d(x, y) = \sum_{n=1}^{\infty} \frac{1}{2^n} d_M(x_n, y_n),$$

where $x = (x_1, x_2, \dots)$ and $y = (y_1, y_2, \dots)$. Denote $\mathcal{B} := M^{\mathbb{N}}$ which is compact.

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We also denote by F_α the set of Holder functions $A : \mathcal{B} \rightarrow \mathbb{R}$ with the norm $\|A\|_\alpha = \|A\| + |A|_\alpha$.

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For a fixed potential $A \in F_\alpha$ we define a **Transfer (Ruelle) operator** $\mathcal{L}_A : \mathcal{C} \rightarrow \mathcal{C}$ by the rule

$$\mathcal{L}_A(\varphi)(x) = \int_M e^{A(ax)} \varphi(ax) d\nu(a),$$

where $a \in M$, $x \in \mathcal{B}$ and we will use the following notation
 $ax = (a, x_1, x_2, \dots)$, is a pre-image of x .

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(b) the measure $\rho_A = \frac{1}{\psi_A} \mu_A$ satisfies $\mathcal{L}_A^*(\rho_A) = \lambda_A \rho_A$.
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We call μ_A the **Gibbs state** for A . We will leave the term **equilibrium state** for the one which maximizes pressure.

Definition Consider ν a fixed a-priori probability on M . Denote $\mathcal{G} = \mathcal{G}_\nu$ the set of Gibbs measures, that means, $\mathcal{L}_B^*(\mu) = \mu$ for some normalized potential $B \in F_\alpha$. We define the entropy of $\mu \in \mathcal{G}$ as

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If our a-priori probability is $\nu = \frac{1}{d} \sum_{j=1}^d \delta_j$, to get the above entropy $h(\mu)$ you have just to add $-\log d$ to the classical one $H(\mu)$. Therefore, the above definition of $h(\mu)$ results in a number between $-\log(d)$ and 0.

We point out that at **level 2 the Deviation Function for the maximal entropy probability** is $I(\mu) = \log d - H(\mu) \geq 0$.

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Proposition If $\mu \in \mathcal{G}$, we have $h(\mu) \leq 0$.

Proof: Let μ be a probability on \mathcal{G} with associated normalized potential B . Then, we have

$$\begin{aligned} h(\mu) &= - \int_{\mathcal{B}} B(x) d\mu(x) = \int_{\mathcal{B}} \log e^{-B(x)} d\mu(x) \leq \log \int_{\mathcal{B}} e^{-B(x)} d\mu(x) = \\ &= \log \int_{\mathcal{B}} e^{-B(x)} d\mathcal{L}_B^*(\mu)(x) = \log \int_{\mathcal{B}} \mathcal{L}_B e^{-B(x)} d\mu(x) = 0, \end{aligned}$$

where we have used Jensen inequality and also $\mathcal{L}_B e^{-B(x)} = 1$.

Definition Let $\mu \in \mathcal{M}_\sigma$ be an invariant measure. We define the entropy of μ as

$$h^\nu(\mu) = h(\mu) = \inf_{A \in \mathcal{F}_\alpha} \left\{ - \int_{\mathcal{B}} A d\mu + \log \lambda_A \right\}$$

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The Entropy of a measure supported on a fixed point.

Suppose $M = [0, 1]$, and $A_c : M^{\mathbb{N}} \rightarrow \mathbb{R}$ is given by

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Then, we have $h(\mu) \leq - \int A_c d\mu = -A_c(0^\infty) = -\log \left(\frac{c}{1-e^{-c}} \right) \rightarrow -\infty$ when $c \rightarrow \infty$.

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This shows that $h(\mu) = -\infty$.

Lemma Consider a Holder continuous potential A and $\mu \in \mathcal{G}$ associated to B normalized, and call \mathcal{C}^+ the space of continuous bounded positive functions on \mathcal{B} . Then,

$$h(\mu) + \int_{\mathcal{B}} A(x) d\mu(x) = \inf_{u \in \mathcal{C}^+} \left\{ \int_{\mathcal{B}} \log \left(\frac{\mathcal{L}_A u(x)}{u(x)} \right) d\mu(x) \right\},$$

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Lemma Consider a Holder continuous potential A and $\mu \in \mathcal{G}$ associated to B normalized, and call \mathcal{C}^+ the space of continuous bounded positive functions on B . Then,

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Theorem: Given a Lipchitz potential A consider the associated normalized potential \bar{A} , then, the Gibbs probability for \bar{A} is the equilibrium state for A , that is, attains the supremum value of $h(\mu) + \int_B A(x) d\mu(x)$ among invariant probabilities for σ .

The differentiable structure and the involution kernel

We consider now the **general XY model**. This is the case where $M = S^1$ and the **a priori measure is the Lebesgue measure on the circle**. $(S^1)^{\mathbb{N}}$ has a differentiable C^∞ structure.

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Let $\mathcal{B}^* = \{y = (\dots y_2, y_1) \in (S^1)^{\mathbb{N}}\}$, and, we denote by the pair

$$\langle y|x \rangle = \langle \dots, y_2, y_1 | x_1, x_2 \dots \rangle,$$

the general element of $\hat{\mathcal{B}} := \mathcal{B}^* \times \mathcal{B} = (S^1)^{\mathbb{Z}}$.

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We denote $\hat{\sigma}$ the shift on $\hat{\mathcal{B}}$, and, therefore

$$\hat{\sigma}(\langle \dots, y_2, y_1 | x_1, x_2, \dots \rangle) = \langle \dots, y_2, y_1, x_1 | x_2, x_3, \dots \rangle.$$

Definition Let $A : \mathcal{B} \rightarrow \mathbb{R}$ be a **continuous potential** (considered as a function on $\hat{\mathcal{B}}$). A continuous function $W : \hat{\mathcal{B}} \rightarrow \mathbb{R}$ is called a **involution kernel** if for all $\langle y, x \rangle \in \hat{\mathcal{B}} := \mathcal{B}^* \times \mathcal{B} = (S^1)^{\mathbb{Z}}$

$$A^* := A \circ \hat{\sigma}^{-1} + W \circ \hat{\sigma}^{-1} - W$$

depends only on **the variable y** . Above $A(x)$ and $A^*(y)$.

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Suppose c is such that $\int \int e^{W(y|x) - c} d\rho_{A^*}(y) d\rho_A(x) = 1$.

Proposition Denote $Z(y|x) = e^{W(y|x) - c}$, then,

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$\hat{\mu}_A = Z(y|x) d\rho_{A^*}(y) d\rho_A(x)$ is invariant for $\hat{\sigma}$ and is the **natural extension of the Gibbs measure μ_A** .

Moreover, $\psi_A(x) = \int_{\mathcal{B}^*} Z(y|x) d\rho_{A^*}(y)$ is the **main eigenfunction** for \mathcal{L}_A .

Suppose that A is differentiable in each coordinate of $x \in \mathcal{B}$, and, that given $\epsilon > 0$, there exists $H_\epsilon > 0$, such that, if $|h| < H_\epsilon$, then

$\left| \frac{A(x+he_j) - A(x)}{h} - D_j A(x) \right| \leq \frac{\epsilon}{2^j}$, $\forall j \in \mathbb{N}$, where $D_j A(x)$ is the derivative of A w. r. to the j -th coordinate.

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$$\frac{\partial}{\partial x_j} \psi_A(x) = \int_{\mathcal{B}^*} Z(y|x) \sum_{n \geq 1} D_{n+j} A(y_n \dots y_1 x) d\rho_{A^*}(y).$$

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In the case where A depends only on the two first coordinates

$$\psi_A(x_1) = \frac{1}{\lambda_A} \int_{S^1} e^{A(y_1, x_1)} \psi_A(y_1) d\nu(y_1).$$

Suppose that A is differentiable in each coordinate of $x \in \mathcal{B}$, and, that given $\epsilon > 0$, there exists $H_\epsilon > 0$, such that, if $|h| < H_\epsilon$, then

$\left| \frac{A(x+he_j) - A(x)}{h} - D_j A(x) \right| \leq \frac{\epsilon}{2^j}$, $\forall j \in \mathbb{N}$, where $D_j A(x)$ is the derivative of A w. r. to the j -th coordinate.

Proposition Suppose A is differentiable. Then, the eigenfunction ψ_A is differentiable, and

$$\frac{\partial}{\partial x_j} \psi_A(x) = \int_{\mathcal{B}^*} Z(y|x) \sum_{n \geq 1} D_{n+j} A(y_n \dots y_1 x) d\rho_{A^*}(y).$$

In the case where A depends only on the two first coordinates

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Hence, ψ_A satisfies the equation

$$\frac{\partial}{\partial x_1} \psi_A(x_1) = \frac{1}{\lambda_A} \int_{\mathcal{S}^1} e^{A(y_1, x_1)} D_2 A(y_1, x_1) \psi_A(y_1) d\nu(y_1).$$

The zero temperature limit

We would like to investigate general properties of the limits of $\mu_{\beta_n A}$, and, of $\psi_{\beta_n A}$ when $\beta_n \rightarrow \infty$.

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In this case in the XY model the Large Deviation function for $\mu_{\beta A}$, when $\beta \rightarrow \infty$ is considered in

[LM] A. Lopes and J. Mengue, Selection of measure and a Large Deviation Principle for the general one-dimensional XY model (2011)

The calibrated subaction plays an essential role on this proof.

Definition A continuous function $u : \mathcal{B} \rightarrow \mathbb{R}$ is called a **calibrated subaction** for $A : \mathcal{B} \rightarrow \mathbb{R}$, if, for any $y \in \mathcal{B}$, we have

$$u(y) = \max_{\sigma(x)=y} [A(x) + u(x) - m(A)]. \quad (1)$$

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Proposition - Existence of calibrated subaction - For any uniformly convergent subsequence $\beta_n \rightarrow \infty$, if

$$V := \lim_{n \rightarrow \infty} \frac{1}{\beta_n} \log(\psi_{\beta_n A}),$$

then **V is a calibrated subaction for A**. In the case we have uniqueness of μ_∞ , then selection.

An application to the non-compact case

Suppose $M_0 = \{z_i, i \in \mathbb{N}\}$ is an increasing infinite sequence of points in $[0, 1)$, and suppose $z_\infty := 1 = \lim_{i \rightarrow \infty} z_i$. We will assume $z_1 = 0$. Then, $M = M_0 \cup \{1\}$ is a compact set.

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Therefore, we have a state space M_0 that can be identified with \mathbb{N} , and M has a special point $z_\infty = 1$ playing the role of the infinity. Let $\mathcal{B}_0 = M_0^{\mathbb{N}}$ and $\mathcal{B} = M^{\mathbb{N}}$.

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Theorem Suppose that $A : \mathcal{B}_0 \rightarrow \mathbb{R}$ is a Holder continuous potential. Then it can be extended as a Holder continuous function to $A : \mathcal{B} \rightarrow \mathbb{R}$. If the extension satisfies

$A(x_1, \dots, x_{n-1}, 1, x_{n+1}, x_{n+2}, \dots) < A(x_1, \dots, x_{n-1}, 0, x_{n+1}, x_{n+2}, \dots)$ for any $n \in \mathbb{N}$ and $x_i \in M$ then:

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a) A has a calibrated subaction V on \mathcal{B}_0 : for any $x \in \mathcal{B}_0$,

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b) Any maximizing measure for A has support on \mathcal{B}_0 .

Markov Chains with values on the interval $[0, 1]$

Suppose $M = [0, 1]$.

Let $K : M^2 \rightarrow \mathbb{R}$, $\theta : M \rightarrow \mathbb{R}$ satisfying

$$\int_M K(x_1, x_2) d\nu(x_2) = 1, \forall x_1 \quad \text{and} \quad \int_M \theta(x_1) K(x_1, x_2) d\nu(x_1) = \theta(x_2), \forall x_2$$

We call K a transition kernel and θ the stationary measure for K .

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We define the Markov measure associated to K and θ as

$$\mu(A_1 \dots A_n \times [0, 1]^{\mathbb{N}}) := \int_{A_1 \dots A_n} \theta(x_1) K(x_1, x_2) \dots K(x_{n-1}, x_n) d\nu(x_n) \dots d\nu(x_1)$$

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Proposition Given $A(x_1, x_2)$ a continuous potential that depends on two coordinates, there exists a Markov measure that is Gibbs for A .

Proposition Let μ be the Markov measure defined by a transition kernel K and the stationary measure θ . The definition of **penalized entropy** given in [LMST]:

$$S(\theta K) = - \int_{M^2} \theta(x_1) K(x_1, x_2) \log(K(x_1, x_2)) d\nu(x_1) d\nu(x_2) \leq 0$$

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For the general problem for a metric space M one can consider an a priori probability ν_z on M which depends on $z \in M$ and define an Ruelle operator

$\mathcal{L}_A : \mathcal{C} \rightarrow \mathcal{C}$ by the rule: given $x = (x_0, x_1, x_2, \dots)$

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Under some mild assumptions on $\nu_z, z \in M$, one can show several of the above properties.

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