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Isomorphism of a Piecewise Linear Transformation to a Markov Automorphism

By

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D. Ornstein and others proved important isomorphism theorems for a large class of automorphisms, i.e. measure preserving transformations. However, it is also interesting to explicitely determine an isomorphic mapping in each concrete case. R. Adler and B. Weiss [1] explicitely constructed an isomorphism between an ergodic group automorphism of a 2-dimensional torus and a Markov automorphism. Y. Takahashi [8] gave an isomorphism between a β -automorphism and a Markov automorphism. The crucial point of their argument lies in the fact that the metrical entropy coincides with the topological entropy for these automorphisms. Using this fact and Parry's result [3], they showed that the representation mapping of the automorphism in consideration to a Markov subshift (of a symbolic dynamics) is actually an isomorphism in each case.

In this paper, we will explicitly construct an isomorphism of a piecewise linear transformation (a generalization of β -automorphism) to a Markov automorphism. Since the metrical entropy of such a transformation does not always attain its topological entropy, we cannot use the method mentioned above, so instead of topological entropy we use free energy as our main tool.

In §1 we define the free energy of a Markov subshift and show under certain conditions that a shift invariant measure with the minimal free energy is unique. This is a generalization of Parry's result about topological entropy [3]. In §2 we define a piecewise linear transformation and investigate its properties. In §3 we construct an isomorphic mapping from a piecewise linear transformation to a Markov automor-

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§1. Markov Potential Function and Its Free Energy

Let I^{z} be an infinite product of state space *I*, where $I = \{0, 1, \dots, s\}$ or $\{0, 1, 2, \dots\}$ and $z = \{\dots, -1, 0, 1, \dots\}$. Let $\omega(n)$ be the *n*-th coordinate of ω in I^{z} and let σ be a shift operator on I^{z} such that

$$(\sigma\omega)(n) = \omega(n+1)$$

Definition 1-1. A function $U: I^{\mathbb{Z}} \to [0, \infty]$ is called a (simple) Markov potential function, if $U(\omega) = U(\omega(0), \omega(1))$ for all $\omega \in I^{\mathbb{Z}}$.

Definition 1-2. A Markov potential function U is called irreducible, if, for any $i, j \in I$, there exists a finite sequence (i_0, i_1, \dots, i_n) such that $i_0 = i$, $i_n = j$, and $U(i_k, i_{k+1}) < \infty$ for all $0 \leq k \leq n-1$.

Definition 1-3. A Markov potential function U is called of Perron-Frobenius type if

- (1) U is irreducible
- (2) Q(i,j) defined by $Q(i,j) = e^{-U(i,j)}$

has positive right and left eigenvectors, $r = (r_i)$ and $l = (l_i)$ respectively, with a common positive eigenvalue λ , such that

$$\sum_{j \in I} Q(i, j) r_j = \lambda r_i \quad \text{for all} \quad i \in I,$$
$$\sum_{i \in I} Q(i, j) l_i = \lambda l_j \quad \text{for all} \quad j \in I,$$
$$\sum_{i \in I} l_i r_i < \infty.$$

Definition 1-4. Let λ , $r = (r_i)$ and $l = (l_i)$ be the ones defined by Perron-Frobenius type potential U as above. The Gibbs measure $\nu = \nu(U)$ is an ergodic σ -invariant Markov measure with initial distribution

$$\pi_i = l_i r_i / (\sum_{j \in I} l_j r_j) \quad \text{for} \quad i \in I,$$

2

and transition probabilities

$$P_{ij} = \frac{r_j Q(i,j)}{\lambda r_i}$$
 for $i, j \in I$.

When $I = \{0, 1, \dots, s\}$ $(s < \infty)$, free energy $f_{\upsilon}(\mu)$ is defined by (cf. Spitzer [7]), $f_{\upsilon}(\mu) = \int U d\mu - h_{\mu}(\sigma)$, where μ is a σ -invariant probability measure on (I^{z}, σ) . Now we extend this definition.

Definition 1-5. Let $I = \{0, 1, 2, \dots\}$ and let μ be a σ -invariant probability measure on (I^z, σ) with finite metrical entropy $h_{\mu}(\sigma)$. Then free energy $f_{\sigma}(\mu)$ is defined by,

$$f_{\mathcal{U}}(\mu) = \overline{\lim_{m \to \infty}} \, \overline{\lim_{n \to \infty}} \, \sum_{x \in I_m^n} \mu([x]) \, U_n^m(x) - h_\mu(\sigma) \,,$$

where $I_m = \{0, 1, \dots, m, m^*\}, m^* = \{m+1, m+2, \dots\}, [x]$ is the cylinder set generated by $x \in I_m^n$ and U_n^m , a function on I_m^n , is defined in the following way. For $x, y \in I_m' = \{0, 1, \dots, m\}, k \ge 1$, we put

$$\begin{split} U_{2}^{m}(x, y) &= U(x, y), \\ U_{k+1}^{m}(x, \underbrace{m^{*}, \cdots, m^{*}}_{k}) \\ &= -\frac{1}{k} \log \sum_{\substack{i_{1} \geq m+1 \\ i_{k} \geq m+1}} r_{i_{k}} \exp\left[-U(x, i_{1}) - \sum_{j=1}^{k-1} U(i_{j}, i_{j+1})\right] \\ U_{k+2}^{m}(x, \underbrace{m^{*}, \cdots, m^{*}}_{k}, y) \\ &= -\frac{1}{k+1} \log \sum_{\substack{i_{1} \geq m+1 \\ i_{k} \geq m+1}} \exp\left[-U(x, i_{1}) - \sum_{j=1}^{k-1} U(i_{j}, i_{j+1}) - U(i_{k}, y)\right] \\ U_{k+1}^{m}(\underbrace{m^{*}, \cdots, m^{*}}_{k+1}) \\ &= -\frac{1}{k} \log \sum_{\substack{i_{1} \geq m+1 \\ i_{k+1} \geq m+1}} \frac{l_{i_{1}} r_{i_{k+1}}}{\sum_{j} l_{j} r_{j}} \exp\left[-\sum_{j=1}^{k} U(i_{j}, i_{j+1})\right], \end{split}$$

and

$$U_{k+1}^{m}(\underbrace{m^{*}, \cdots, m^{*}}_{k}, y) = -\frac{1}{k} \log \sum_{\substack{i_{1} \ge m+1 \\ i_{k} \ge m+1}} \frac{l_{i_{1}}}{\sum_{j} l_{j}r_{j}} \exp\left[-\sum_{j=1}^{k-1} U(i_{j}, i_{j+1}) - U(i_{k}, y)\right]$$

(we define $\log 0 = -\infty$). Moreover we define U_n^m inductively by

$$(n-1) U_{i}^{m}(x_{1}, \dots, x_{n})$$

= $(i-1) U_{i}^{m}(x_{1}, \dots, x_{i}) + (n-i) U_{n-i+1}^{m}(x_{i}, \dots, x_{n})$

 $\text{if } x_{j} \! \in \! I_{m} \hspace{0.1 cm} (1 \! \leq \! j \! \leq \! n) \hspace{0.1 cm} \text{and} \hspace{0.1 cm} x_{i} \! \in \! I_{m}'.$

Remark 1. For $x_i \in I_m'$ $(1 \leq i \leq n)$,

$$(n-1) U_n^m(x_1, \cdots, x_n) = \sum_{i=1}^{n-1} U(x_i, x_{i+1}).$$

We get, for $x \in I_m^n$,

$$U_n^{m}(x) = -\log \lambda - \frac{1}{n-1}\log P_n^{m}(x) - \frac{1}{n-1}\log r_{x_1}^{m} + \frac{1}{n-1}\log r_{x_n}^{m}$$

(including $\infty = \infty$)

where $r_i^m = r_i$, $i \in I_m'$, $r_{m^*}^m = 1$, and $P_n^m(x)$, a function on I_m^n , is defined by the following; for $x, y \in I_m'$, $k \ge 1$

$$P_{2}^{m}(x, y) = P_{x,y}$$

$$P_{k+1}^{m}(x, \underbrace{m^{*}, \cdots, m^{*}}_{k}) = \sum_{\substack{i_{1} \ge m+1 \\ i_{k} \ge m+1}} P_{x,i_{1}}P_{i_{1},i_{2}} \cdots P_{i_{k-1},i_{k}}$$

$$P_{k+2}^{m}(x, \underbrace{m^{*}, \cdots, m^{*}}_{k}, y) = \sum_{\substack{i_{1} \ge m+1 \\ i_{k} \ge m+1}} P_{x,i_{1}}P_{i_{1},i_{2}} \cdots P_{i_{k-1},i_{k}}P_{i_{k},y}$$

$$P_{k+1}^{m}(\underbrace{m^{*}, \cdots, m^{*}}_{k}, y) = \sum_{\substack{i_{1} \ge m+1 \\ i_{k} \ge m+1}} \pi_{i_{1}}P_{i_{1},i_{2}} \cdots P_{i_{k},i_{k+1}}$$

$$P_{k+1}^{m}(\underbrace{m^{*}, \cdots, m^{*}}_{k}, y) = \sum_{\substack{i_{1} \ge m+1 \\ i_{k+1} \ge m+1}} \pi_{i_{1}}P_{i_{1},i_{2}} \cdots P_{i_{k-1},i_{k}}P_{i_{k},y}$$

and

$$P_n^{m}(x_1, \dots, x_n) = P_i^{m}(x_1, \dots, x_i) P_{n-i+1}^{m}(x_i, \dots, x_n)$$

where $x_j \in I_m$ (1 $\leq j \leq n$), $x_i \in I_m'$.

Remark. If
$$x_i \in I_m'$$
 $(1 \leq j \leq n)$,
 $P_n^m(x_1, \dots, x_n) = \prod_{i=1}^{n-1} P_{x_i, x_{i+1}}$

Theorem 1. Let U be a Markov potential of Perron-Frobenius type on (I^z, σ) , and let λ be an eigenvalue of $Q = \exp(-U)$ as in Definition 1-3. Then, for any σ -invariant probability measure μ with finite entropy, we have

$$f_{v}(\mu) \ge -\log \lambda \tag{1.1}$$

In particular, (1.1) holds with equality if and only if $\mu = \nu$ and $h_{\nu}(\sigma) < \infty$, where $\nu = \nu(U)$ is the Gibbs measure defined by U.

Proof. 1st stage: We will show (1.1) holds with equality if $\mu = \nu$ and $h_{\nu}(\sigma) < \infty$. We denote by φ_m the σ -algebra generated by the cylinder sets of I_m^z , and by $h_{\mu}(\sigma|\varphi_m)$ we mean the entropy of the factor of σ on φ_m . Evidently,

$$\begin{split} h_{\mu}(\sigma) &= \lim_{m \to \infty} h_{\mu}(\sigma|_{\mathscr{G}m}) \\ &= \lim_{m \to \infty} \lim_{n \to \infty} \frac{1}{n} \sum_{x \in I_m n} \mu([x]) \log \mu([x]). \end{split}$$

Therefore,

$$f_{\mathcal{U}}(\boldsymbol{\nu}) = \overline{\lim_{m \to \infty}} \, \overline{\lim_{n \to \infty}} \Big[-\log \lambda - \frac{1}{n-1} \sum_{x \in I_m^n} \boldsymbol{\nu}\left([x]\right) \log \boldsymbol{\nu}\left([x]\right) \\ + \frac{1}{n-1} \sum_{i \in I_m'} \pi_i \log \pi_i \Big] + \lim_{m \to \infty} \lim_{n \to \infty} \frac{1}{n} \sum_{x \in I_m^n} \boldsymbol{\nu}\left([x]\right) \log \boldsymbol{\nu}\left([x]\right) \\ = -\log \lambda \,.$$

2nd stage: Let μ be a σ -invariant probability measure with finite entropy which does not equal to ν . Let $\{\tilde{\pi}_i\}$ be a distribution such that

(1)
$$\sum_{i \in I} \tilde{\pi}_i = 1$$

(2)
$$-\sum_{i \in I} \mu([i]) \log \tilde{\pi}_i < \infty$$

(3)
$$\tilde{\pi}_i > 0 \text{ if and only if } \pi_i > 0.$$

We define $\tilde{\nu}$ the Markov measure with initial distribution $\tilde{\pi}_i$ and transition probability P_{ii} . (Notice that $\tilde{\nu}$ is not always σ -invariant.) Then it is easy to see that $\tilde{\nu} \sim \nu$. From the σ -invariance of μ , we get

$$\frac{1}{n-1}\sum_{x\in I_m^n}\mu([x])\log P_n^m(x) \leq \sum_{x,y\in I_m'}\mu([x,y])\log P_{x,y}.$$

Therefore,

$$\begin{split} &\lim_{n\to\infty} \frac{1}{n-1} \sum_{x\in I_m^n} \mu([x]) \log P_n^m(x) \leq \sum_{x,y\in I'_m} \mu([x,y]) \log P_{x,y} \, . \\ & \text{If } \lim_{m\to\infty} \sum_{x,y\in I'_m} \mu([x,y]) \log P_{x,y} = -\infty, \text{ then} \\ & f_U(\mu) = \overline{\lim_{m\to\infty}} \overline{\lim_{n\to\infty}} \sum_{x\in I_m^n} \mu([x]) U_n^m(x) - h_\mu(\sigma) = \infty > -\log \lambda \, . \\ & \text{If } \lim_{m\to\infty} \sum_{x,y\in I_m'} \mu([x,y]) \log P_{x,y} > -\infty, \text{ then} \\ & \overline{\lim_{m\to\infty}} \overline{\lim_{n\to\infty}} \frac{1}{n-1} \sum_{x\in I_m^n} \mu([x]) \log P_n^m(x) \\ & \leq \sum_{x,y\in I} \mu([x,y]) \log P_{x,y} \\ & = \frac{1}{n-1} \sum_{x\in I_m} \mu([x]) \log P_{x,y} \\ & = \frac{1}{n-1} \sum_{x\in I^m} \mu([x]) \log \tilde{\gamma}([x]) - \frac{1}{n-1} \sum_{i\in I} \mu([i]) \log \tilde{\pi}_i \\ & \leq \overline{\lim_{m\to\infty}} \frac{1}{n-1} \sum_{x\in I_m} \mu([x]) \log \tilde{\gamma}([x]) - \frac{1}{n-1} \sum_{i\in I} \mu([i]) \log \tilde{\pi}_i \, . \end{split}$$

Thus,

$$\begin{split} f_{\mathcal{U}}(\mu) &= \lim_{m \to \infty} \lim_{n \to \infty} \sum_{x \in I_m^n} \mu([x]) U_n^m(x) - \lim_{m \to \infty} h_\mu(\mathcal{O}|_{\mathcal{G}_m}) \\ &\geq -\log \lambda - \lim_{m \to \infty} \frac{1}{n-1} \sum_{x \in I_m^n} \mu([x]) \log \tilde{\nu}([x]) \\ &+ \frac{1}{n-1} \sum_{i \in I} \mu([i]) \log \tilde{\pi}_i + \lim_{m \to \infty} \frac{1}{n} \sum_{x \in I_m^n} \mu([x]) \log \mu([x]) \\ &\geq \frac{1}{n-1} \left[\lim_{m \to \infty} \sum_{x \in I_m^n} \mu([x]) \log \left(\frac{\mu([x])}{\tilde{\nu}([x])} \right) + \sum_{i \in I} \mu([i]) \log \tilde{\pi}_i \right] \\ &- \log \lambda. \end{split}$$

On the other hand, since ν is ergodic and μ is σ -invariant, it follows $\mu \not\sim \nu$. We cite the followings.

Lemma 1 ([5]). $\lim_{m \to \infty} \sum_{x \in I_m^n} \mu([x]) \log \frac{\mu([x])}{\widetilde{\nu}([x])} \ge \sum_{x \in I^n} \mu([x]) \log \frac{\mu([x])}{\widetilde{\nu}([x])}.$ **Lemma 2.** (Dobrushin, Gelfand, Yaglom and Perez [5] (p. 20)). If μ_1 and μ_2 are probability measures on I^z and $\mu_1 \not\prec \mu_2$, then

$$\sum_{x \in I^n} \mu_1([x]) \log \left(\frac{\mu_1([x])}{\mu_2([x])} \right) \to \infty \quad as \quad n \to \infty$$

By the above lemmas, we can take $n \ge 1$ such that

$$\sum_{i\in I} \mu([i])\log \tilde{\pi}_i + \sum_{x\in I^n} \mu([x])\log\left(\frac{\mu([x])}{\tilde{\nu}([x])}\right) > 0,$$

and thus we can conclude that

$$f_v(\mu) > -\log \lambda$$
.

Remark 1-6. If a Markov potential U satisfies the condition in Theorem 1 and takes values only 0 and ∞ , then $Q = e^{-U}$ becomes a structure matrix of a Markov subshift [2]. Hence $\inf -f_{\upsilon}(\mu) = -f_{\upsilon}(\nu)$ and we see that ν is the maximal Markov measure [3].

Let U be a Markov potential on (I^{z}, σ) which satisfies the conditions in Theorem 1 and let ν be the Gibbs measure defined by U.

Main Lemma. Let (X, T, μ) be an automorphism with finite entropy. Then (X, T, μ) is isomorphic to (I^z, σ, ν) , if there exist a 1-1 (μ -a.e.) bi-Borel mapping $\varphi: X \rightarrow I^z$, μ -integrable functions $V(\omega')$ and $h(\omega')$ ($\omega' \in X$) such that

(1) $\sigma \cdot \varphi = \varphi \cdot T$, μ -a.e. (2) $U(\varphi(\omega')) = V(\omega') + h(\omega')$, μ -a.e. $\int h(\omega') d\mu(\omega') = 0$, (3) $f_U(\nu) = \int d\mu(\omega') V(\omega') - h_{\mu}(T)$, (4) $\lim_{m \to \infty} \lim_{n \to \infty} \sum_{x \in I_m^n} \mu \cdot \varphi^{-1}([x]) U_n^m(x) = \int d\mu \cdot \varphi^{-1} U$.

Proof. Let $\mu \cdot \varphi^{-1} = \mu'$, then from the condition (1) μ' is a σ -invariant measure with finite entropy $h_{\mu}(T)$. Therefore to show that they are isomorphic it is sufficient to prove that $f_{\upsilon}(\mu') = f_{\upsilon}(\nu)$.

$$f_{\mathcal{U}}(\mu') = \lim_{m \to \infty} \lim_{n \to \infty} \sum \mu'([x]) U_n^m(x) - h_{\mu'}(\sigma)$$
$$= \int d\mu(\omega') U(\varphi(\omega')) - h_{\mu}(T)$$
$$= \int d\mu(\omega') V(\omega') - h_{\mu}(T)$$
$$= f_{\mathcal{U}}(\nu).$$

§ 2. A Definition and Properties of a Piecewise Linear Transformation

We define a piecewise linear transformation according to [6].

Let $\beta = (\beta_0, \beta_1, \dots, \beta_N)$ is an N+1-tuple vector such that $\beta_k > 1$ for $0 \leq k \leq N$ and $\sum_{k=0}^{N} \beta_k^{-1} \geq 1 > \sum_{k=0}^{N-1} \beta_k^{-1}$. A partition $\{A_i\}_{i=0,\dots,N}$ of interval [0, 1) is defined by

$$\begin{split} &A_{0} = \begin{bmatrix} 0, \beta_{0}^{-1} \end{bmatrix} \\ &A_{i} = (\sum_{k=0}^{i-1} \beta_{k}^{-1}, \sum_{k=0}^{i} \beta_{k}^{-1}) \quad i = 1, \dots, N-1 \\ &A_{N} = (\sum_{k=0}^{N-1} \beta_{k}^{-1}, 1) . \end{split}$$

We define mappings $T:[0,1) \rightarrow [0,1)$ and $\pi;[0,1) \rightarrow S^N$ by

$$Tx = \beta_i (x - \sum_{k=0}^{i-1} \beta_k^{-1})$$
 if $x \in A_i, i = 0, 1, ..., N$,

and

$$\pi(x) = (a_0(x), a_1(x), \cdots, a_n(x), \cdots),$$

 $a_n(x) = i$ if $T^n x \in A_i$, respectively, where $S = \{0, 1, \dots, N\}$ and S^N is its one-sided product. Then the mapping T can be realized as the shift σ on S^N by the mapping π , namely

$$\pi \cdot T = \sigma \cdot \pi$$

For convenience we define a sequence corresponding to 1 by

$$\pi(1) = (a_0(1), a_1(1), \dots, a_n(1), \dots)$$

= $\sup_{x \in [0,1]} (a_0(x), a_1(x), \dots, a_n(x), \dots)$

where supremum is taken over with respect to the lexicographical order. Then it is easy to see that for $x \in [0, 1)$

$$x = \sum_{k=0}^{a_0(x)^{-1}} \beta_k^{-1} + \beta_{a_0(x)}^{-1} \sum_{k=0}^{a_1(x)^{-1}} \beta_k^{-1} + \cdots + \beta_{a_0(x)}^{-1} \cdots \beta_{a_n(x)}^{-1} \left(\sum_{k=0}^{a_{n+1}(x)^{-1}} \beta_k^{-1} \right) + \cdots$$

We put

$$T^{j} 1 = \sum_{k=0}^{a_{j}(1)-1} \beta_{k}^{-1} + \beta_{a_{j}(1)}^{-1} (\sum_{k=0}^{a_{j+1}(1)-1} \beta_{k}^{-1}) + \cdots.$$

We already know the results that the endomorphism T has an invariant probability measure ν_{β}^{+} equivalent to Lebesgue measure, and its unnormalized density f(x) is given by

$$f(x) = \sum_{n=0}^{\infty} \frac{X_{[0, T^{n}]}}{\beta_{a_0(1)}\beta_{a_1(1)}\cdots\beta_{a_n(1)}}(x),$$

where X_A is the characteristic function of A[6].

Let $\pi[0, 1) = X_{\beta}^{+} \subset S^{\mathbf{N}}$ and let μ_{β}^{+} be an induced measure of ν_{β}^{+} by π . Then $[X_{\beta}^{+}, \sigma, \mu_{\beta}^{+}]$ becomes an endomorphism which is homeomorphic to $([0, 1), T, \nu_{\beta}^{+})$. We call this endomorphism a piecewise linear endomorphism.

Proposition 2.1.

(1) For any $\omega \in X_{\beta}$

 $\sigma^n \omega < (a_0(1), a_1(1), \dots, a_n(1), \dots), \text{ for } n \ge 0.$

(2) Let $X_{\beta}^{n} = \{ [a_{0}(x), \dots, a_{n}(x)] : x \in [0, 1) \}, then$

$$\max\{[b_0, \dots, b_n] \in X_{\beta}^n\} = [a_0(1), \dots, a_n(1)].$$

The proofs of this proposition are very easy and we omitt it (c.f. [6]).

The piecewise linear transformation $(X_{\beta}, \sigma, \mu_{\beta})$ which is the natural extension of $(X_{\beta}^{+}, \sigma, \mu_{\beta}^{+})$ is defined by

$$X_{\beta} = \{ \omega \in S^{\mathbf{z}}; (\omega(n), \omega(n+1), \cdots) \in X_{\beta}^{+}, \text{ for all } n \},$$
$$\mu_{\beta}([b_{0}, \cdots, b_{n}]) = \mu_{\beta}^{+}([b_{0}, \cdots, b_{n}]),$$

where $[b_0, \dots, b_n] \in X_{\beta}^n$. Note that μ_{β} is well-defined because of σ -invariance of μ_{β}^+ .

§ 3. Isomorphism of a Piecewise Linear Transformation to a Markov Automorphism

In this section we will prove the following Theorem 2 by appealing to Main Lemma in Section 1 and a mapping φ which is similar to the mapping introduced by Y. Takahashi for β -transformation ([8]).

Theorem 2. A piecewise linear transformation $(X_{\beta}, \sigma, \mu_{\beta})$ is isomorphic to a Markov automorphism.

To prove this Theorem we begin with the construction of a mapping $\varphi: X_{\beta} \rightarrow I^{z}$ where $I = \{-N+1, \dots, -1, 0, 1, \dots\}$.

For $\omega \in X_{\beta}$, $\tau(\omega)$ is defined by

$$\tau(\omega) = \begin{cases} \sup \{i : i \ge 1, \ \omega \in B_i\}, \\ 0, \quad \text{if} \quad \omega \in X_\beta \setminus \bigcup_{i \ge 1} B_i, \end{cases}$$

where $B_i = \{ \omega \in X_{\beta} : (\omega(-i), \dots, \omega(-1)) = (a_0(1), \dots, a_{i-1}(1)) \}.$

Let
$$C_i = \{ \omega \in X_{\beta} : \tau(\omega) = i \}$$
, for $i = 1, 2, \cdots$,
= $\{ \omega \in X_{\beta} : \tau(\omega) = 0, \omega(-1) = -i \}$, for $-N+1 \leq i \leq 0$,
 $C_{\infty} = \omega \in X_{\beta} : \tau(\omega) = \infty \}$.

It is easy to see that $\mu_{\beta}(C_{\infty}) = 0$ and $\sigma C_{\infty} = C_{\infty}$. Define a mapping $\varphi: X_{\beta} \setminus C_{\infty} \to I^{z}$ by

$$(\varphi(\omega))(n) = i$$
 if $\omega \in \sigma^n C_i, n \in \mathbb{Z}, i \in I$.

Then, it is easy to see that the mapping φ is a Borel injection and satisfies

$$\varphi \cdot \sigma = \sigma^{-1} \cdot \varphi$$
.

Lemma 3-1. (1) If $\omega \in C_i$, $1 \leq i \leq \infty$, then $\sigma^{-1} \omega \in C_{i-1}$.

(2) If
$$\omega \in C_i$$
, $i \leq 0$, and $\sigma^{-1} \omega \in C_j$, $j \geq 1$, then
 $a_j(1) > -i = |i|$.

Proof. (1) Suppose $\sigma^{-1}\omega \in C_j$ $j \geq i$, then $(\omega(-j-1), \dots, \omega(-1)) < (a_0(1), \dots, a_j(1))$

and

$$(\omega(-j-1), \dots, \omega(-2)) = (a_0(1), \dots, a_{j-1}(1)).$$

Hence

$$\omega(-1) < a_j(1)$$

and so

$$(\omega(-i), \cdots, \omega(-1)) < (\omega(-i), \cdots, \omega(-2), a_j(1)).$$

This contradicts to the maximality of $(\omega(-i), \dots, \omega(-1))$.

(2) By the assumption, the j-tuple

$$(\omega(-j-1), \dots, \omega(-2)) = (a_0(1), \dots, a_{j-1}(1))$$

is maximum. From $\omega \! \in \! C_i$, $(i \leq \! 0)$, it follows that $\omega(-1) = -i \! = \! |i|$ and

$$(\omega(-j-1), \dots, \omega(-1)) < (a_0(1), \dots, a_j(1)).$$

Therefore

$$a_j(1) > -i = |i|.$$

We define a Markov potential U on I^{z} by

$$U(i,j) = \begin{cases} \log \beta_{-i}, & -N+1 \leq i \leq 0, & -N+1 \leq j \leq 0, \\ \log \beta_N, & i=1, & -N+1 \leq j \leq 0, \\ \log \beta_{a_{i-1}(1)}, & i \geq 2, & j=i-1, \\ \log \beta_{-i}, & -N+1 \leq i \leq 0, & j>0 \text{ and } a_j(1) > -i, \\ \infty, & \text{otherwise.} \end{cases}$$

It is easy to see that this potential U is irreducible. We show that U is of the Perron-Frobenius type by finding an eigenvalue λ of $Q = e^{-v}$ and corresponding eigenvectors to λ .

Put λ , $l = (l_i)_{i \ge -N+1}$ and $r = (r_i)_{i \ge -N+1}$ as follows, $\lambda = 1$, $l_i = 1$, $-N+1 \le i \le 0$, $= T^i 1$, $i \ge 1$, $r_i = \beta_{a_0(1)}^{-1} \beta_{a_1(1)}^{-1} \cdots \beta_{a_{i-1}(1)}^{-1}$, $i \ge 1$, $= \beta_{-i}^{-1} (1 + \sum_{j \ge 1} M_{ij} \beta_{a_0(1)}^{-1} \cdots \beta_{a_{j-1}(1)}^{-1})$, $-N+1 \le i \le 0$,

where

$$M_{ij} = 1$$
, if $Q(i, j) > 0$,
= 0, if $Q(i, j) = 0$.

Then it is easy to see that

$$\begin{split} &\sum_{i \in I} l_i Q(i, j) = l_j = \lambda l_j, & \text{for all } j, \\ &\sum_{j \in I} Q(i, j) r_j = r_i, & \text{for all } i, \\ &\sum_{i \in I} r_i l_i < \infty. \end{split}$$

Therefore U is of the Perron-Frobenius type. Let ν be the Gibbs measure corresponding to U. From the fact $-\sum \pi_i \log \pi_i < \infty$,

 $h_{\nu}(\sigma) < \infty$.

Hence Theorem 1 in Section 1 implies

$$f_v(\mathbf{v}) = -\log 1 = 0 \, .$$

Now we will define $V(\omega')$ on $(X_{\beta}, \sigma, \mu_{\beta})$ to apply the Main Lemma. Let

$$V(\omega') = \sum_{i \in S} X_i(\omega') \log \beta_i,$$

where $X_i(\omega')$ is the characteristic function of $\{\omega': \omega'(-1) = i\}$. Then

$$\int d\mu_{\beta}(\omega') V(\omega') = \int d\mu_{\beta}(\omega') V(\sigma^{-1}\omega')$$
$$= \int_{i \in \mathcal{S}} X_{[\omega':\omega'(0)=i]}(\omega') \log \beta_i d\mu_{\beta}(\omega').$$

On the other hand by [4],

12

$$h_{\mu_{\beta}}(\sigma) = \int d\mu_{\beta} \log \phi'(\omega'),$$

where $\phi'(\omega')$ is the derivative of the graph of ϕ -expansion. Hence

$$\int d\mu_{\beta}(\omega') V(\omega') - h_{\mu_{\beta}}(\sigma) = 0 = f_{U}(\nu).$$

We have already proved φ satisfies the conditions (1) and (3) in Main Lemma, and hence it suffices to show only the conditions (2) and (4).

1) In case
$$\omega' \in C_i$$
, $2 \leq i < \infty$, it holds $\omega'(-1) = a_{i-1}(1)$ and hence
 $V(\omega') = \log \beta_{a_{i-1}(1)}$.

On the other hand by Lemma 3-1, it follows

$$(\varphi(\omega'))(0) = i, \quad (\varphi(\omega'))(1) = i-1.$$

Thus

$$U(\varphi(\omega')) = U((\varphi(\omega'))(0), (\varphi(\omega'))(1)),$$
$$= \log \beta_{\alpha_{t-1}(0)}.$$

2) In case $\omega' \in C_i$, $-N+1 \leq i \leq 0$, it holds $\omega'(-1) = -i$, and hence $V(\omega') = \log \beta_{-i}$.

From Lemma 3-1,

$$(\varphi(\omega'))(0) = i, \quad (\varphi(\omega'))(1) = j, \text{ for some } j,$$

 $U(\varphi(\omega')) = \log \beta_{-i}.$

3) In case $\omega' \in C_1$, it holds $\omega'(-1) = N$, and hence

$$V(\omega') = \log \beta_N.$$

and

$$(\varphi(\omega'))(0) = i$$
, $(\varphi(\omega'))(1) = j$, $-N+1 \leq j \leq 0$,
 $U(\varphi(\omega')) = \log \beta_{Y}$.

This implies that

$$U(\varphi(\omega')) = V(\omega'), \quad \mu_{\beta}\text{-a.e.}$$

Therefore (2) of the Main Lemma is satisfied for $h(\omega') = 0$, $\omega' \in X_{\beta}$. To prove (4) of the Main Lemma, put $\mu_{\beta}' = \mu_{\beta} \cdot \varphi^{-1}$, then

$$(*) \sum_{x \in I_{m^{n}}} \mu_{\beta}'([x]) U_{n}^{m}(x) \\ = \sum_{x, y \in I_{m'}} \mu_{\beta}'([x, y]) U(x, y) \\ + \sum_{i=1}^{n-2} \frac{(i+1)(n-1-i)}{n-1} \sum_{x, y \in I_{m'}} \mu_{\beta}'([x, \underbrace{m^{*}, \cdots, m^{*}, y}]) \\ \times U_{i+2}^{m}(x, \underbrace{m^{*}, \cdots, m^{*}, y}) \\ + \sum_{i=1}^{n-1} \frac{i}{n-1} \sum_{x \in I_{m'}} [\mu_{\beta}'(x, \underbrace{m^{*}, \cdots, m^{*}}]) U_{i+1}^{m}(x, \underbrace{m^{*}, \cdots, m^{*}}_{i}) \\ + \mu_{\beta}'([\underbrace{m^{*}, \cdots, m^{*}}_{n}, x]) U_{i+1}^{m}(\underbrace{m^{*}, \cdots, m^{*}}_{n}, x)] \\ + \mu_{\beta}'([\underbrace{m^{*}, \cdots, m^{*}}_{n}]) U_{n}^{m}(\underbrace{m^{*}, \cdots, m^{*}}_{n}).$$

On the support of $\mu_{\beta}{}'$

$$U(x, y) \leq \log \beta_{max}$$

where $\beta_{max} = \max\{\beta_0, \beta_1, \cdots, \beta_N\}$. For $i \ge 1$ we get

$$|U_{i+2}^{m}(x, \underbrace{m^{*}, \cdots, m^{*}}_{i}, m)| = \frac{1}{i+1} [U(x, m+i) + \sum_{j=0}^{i-1} U(m+j+1, m+j)]$$

$$\leq \log \beta_{max},$$

$$|U_{i+1}^{m}(x, \underbrace{m^{*}, \cdots, m^{*}}_{i})|$$

$$= \left| -\frac{1}{i} \log \sum_{j \ge i} r_{m+j} \exp[-U(x, m+i+j) - \sum_{k=1}^{i-1} U(m+k+j+1, m+k+j)] \right| \le \log \beta_{max} - \frac{1}{i} \log \frac{\beta_{max}^{-m}}{1-\beta_{max}},$$

$$|U_{i+1}^{m}(\underbrace{m^{*}, \cdots, m^{*}}_{i}, m)|$$

$$= |-\frac{1}{i} \log \frac{l_{m+i}}{\sum_{j} l_{j}r_{j}} \exp[-\sum_{k=0}^{i-1} U(m+k+1, m+k)]|$$

$$\le \log \beta_{max} + \left| \frac{1}{i} \log \frac{l_{m+i}}{\sum_{j} l_{j}r_{j}} \right|,$$

$$|U_{i+1}^{m}(\underbrace{m^{*}, \cdots, m^{*}}_{i+1})|$$

14

$$= \left| -\frac{1}{i} \log \sum_{j \ge 1} \frac{l_{m+j+i+1}r_{m+j}}{\sum_{k} l_{k}r_{k}} \exp\left[-\sum_{k=1}^{i} U(m+k+j+1, m+k+j)\right] \right|$$

$$\leq \log \beta_{max} + \left| \frac{1}{i} \log \frac{\sum_{j \ge 1} l_{m+j+i+1}r_{m+j}}{\sum_{k} l_{k}r_{k}} \right|.$$

Moreover for $j \ge 1$

$$\begin{split} l_{j} &= T^{j} \mathbf{1} \\ &= \sum_{k=0}^{a_{j}(1)-1} \beta_{k}^{-1} + \beta_{a_{j}(1)}^{-1} \left(\sum_{k=0}^{a_{j+1}(1)-1} \beta_{k}^{-1} \right) + \cdots \\ &= \{ \beta_{a_{0}(1)}^{-1} \cdots \beta_{a_{j-1}(1)}^{-1} \sum_{k=0}^{a_{j}(1)-1} \beta_{k}^{-1} + \beta_{a_{0}(1)}^{-1} \cdots \beta_{a_{j}(1)}^{-1} \sum_{k=0}^{a_{j+1}(1)-1} \beta_{k}^{-1} + \cdots \} \\ &\times \beta_{a_{0}(1)} \cdots \beta_{a_{j-1}(1)} \\ &\geq C \cdot \mu_{\beta}' \left([j] \right) \beta_{a_{0}(1)} \cdots \beta_{a_{j-1}(1)} , \end{split}$$

for some constant C, and

$$\mu_{\beta}'([k]) \leq K \beta_{min}^{-k}$$

for some constants K, $k \ge 1$ and $\beta_{min} = \min{\{\beta_0, \dots, \beta_N\}}$. Therefore the right-hand terms of (*) become

$$\begin{split} &\lim_{m \to \infty} \overline{\lim_{n \to \infty}} \text{ (the first term)} = \sum_{x, y \in I} \mu_{\beta}' ([x, y]) U(x, y) = \int d\mu_{\beta}' \cdot U \,, \\ &\lim_{m \to \infty} \overline{\lim_{n \to \infty}} \lim_{n \to \infty} |\text{the second term}| \\ & \leq \lim_{m \to \infty} \sum_{i=1}^{\infty} (i+1) \mu_{\beta}' ([\underline{m^*, \cdots, m^*}_{i}, m]) \log \beta_{max} \\ & \leq \lim_{m \to \infty} \sum_{i=1}^{\infty} (i+1) K \beta_{min}^{-m-i} \log \beta_{max} = 0 \,, \end{split}$$

 $\lim_{m\to\infty} \overline{\lim} |\text{the third term}|$

$$\leq \underbrace{\overline{\lim}}_{m \to \infty} \max_{i} \mu_{\beta}'([\underbrace{m^{*}, \cdots, m^{*}}_{i}]) \left(i \log \beta_{max} - \log \frac{\beta_{max}^{-m}}{1 - \beta_{max}}\right)$$
$$\leq \underbrace{\overline{\lim}}_{m \to \infty} \max_{i} \sum_{k=m+1}^{\infty} K \beta_{min}^{-k} \left(i \log \beta_{max} - \log \frac{\beta_{max}^{-m}}{1 - \beta_{max}}\right) = 0,$$

 $\overline{\lim_{m\to\infty}} \ \overline{\lim_{n\to\infty}} \ |\text{the fourth term}|$

$$\leq \underbrace{\lim_{m \to \infty} \max_{i} \, \mu_{\beta}'([\underbrace{m^*, \cdots, m^*}_{i}, m]) \left(i \log \beta_{max} + \left| \log \frac{\mu_{\beta}'([\underline{m}+i])}{\sum_{j} l_j r_j} \right| \right) }_{= \underbrace{\lim_{m \to \infty} \max_{i} \, \mu_{\beta}'([\underline{m}+i]) \log \, \mu_{\beta}'([\underline{m}+i]) \mid = 0},$$

 $\lim_{m\to\infty} \lim_{n\to\infty} |\text{the fifth term}|$

$$\leq \overline{\lim_{m \to \infty} \lim_{n \to \infty}} \left(\sum_{k \geq m+n+1} \mu_{\beta}'([k]) \left(\log \beta_{max} + \left| \frac{1}{n-1} \log \frac{\sum_{s \geq 1} l_{m+s+n} r_{m+s}}{\sum_{j} l_{j} r_{j}} \right| \right) \\ \leq \overline{\lim_{m \to \infty} \lim_{n \to \infty}} \frac{1}{n-1} \left(\sum_{k \geq m+n+1} \mu_{\beta}'([k]) \right) \left| \log \left(\sum_{s \geq 1} \mu_{\beta}'([m+s+n]) \right) \right| = 0 .$$

Thus (4) of the Main Lemma is proved. Therefore, applying the Main Lemma, we complete the proof of Theorem 2.

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