# Two special invariant measures for the random $\beta$-transformation 

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(1) Definition of $K_{\beta}$, the random $\beta$-transformation
(2) Unique measure of maximal entropy
(3) A $K_{\beta}$-invariant measure of product type
(4) A special Pisot case

## Terminology

Let $\beta>1$ be a non-integer. By a $\beta$-expansion we mean an expression of the form

$$
x=\sum_{i=1}^{\infty} \frac{b_{i}}{\beta^{i}}
$$

with $b_{i} \in\{0,1, \ldots,\lfloor\beta\rfloor\}$.

## Definition of $K_{\beta}$

Roughly, $K_{\beta}$ is obtained by randomizing the greedy map, and the lazy map.

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Figure 1: The greedy map $T_{\beta}$ (left), and lazy map $L_{\beta}$ (right). Here $\beta=\pi$.

## Definition of $K_{\beta}$

If we take the common refinement, or simply superimpose the two maps, we get the following picture on $[0,\lfloor\beta\rfloor /(\beta-1)]$.


Figure 1: The greedy and lazy maps, and their switch regions.

## Definition of $K_{\beta}$ : a special partition

We get a partition of the interval $\left[0, \frac{\lfloor\beta\rfloor}{\beta-1}\right]$ into $\lfloor\beta\rfloor$ switch regions, $S_{1}, \ldots, S_{\lfloor\beta\rfloor}$, and $\lfloor\beta\rfloor+1$ uniqueness regions, $E_{0}, \ldots, E_{\lfloor\beta\rfloor}$, where

$$
\begin{aligned}
& E_{0}=\left[0, \frac{1}{\beta}\right), \quad E_{\lfloor\beta\rfloor}=\left(\frac{\lfloor\beta\rfloor}{\beta(\beta-1)}+\frac{\lfloor\beta\rfloor-1}{\beta}, \frac{\lfloor\beta\rfloor}{\beta-1}\right], \\
& E_{k}=\left(\frac{\lfloor\beta\rfloor}{\beta(\beta-1)}+\frac{k-1}{\beta}, \frac{k+1}{\beta}\right), \quad k=1, \ldots,\lfloor\beta\rfloor-1, \\
& S_{k}=\left[\frac{k}{\beta}, \frac{\lfloor\beta\rfloor}{\beta(\beta-1)}+\frac{k-1}{\beta}\right], \quad k=1, \ldots,\lfloor\beta\rfloor .
\end{aligned}
$$

## Definition of $K_{\beta}$ : a special partition

- On $S_{k}$, the greedy map assigns the digit $k$, while the lazy map assigns the digit $k-1$. On $E_{k}$ both maps assign the same digit $k$.
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## The map $K_{\beta}$

Consider $\Omega=\{0,1\}^{\mathbb{N}}$ with product $\sigma$-algebra. Let $\sigma: \Omega \rightarrow \Omega$ be the left shift. Define $K=K_{\beta}: \Omega \times[0,\lfloor\beta\rfloor /(\beta-1)] \rightarrow \Omega \times[0,\lfloor\beta\rfloor /(\beta-1)]$ by

$$
K(\omega, x)= \begin{cases}(\omega, \beta x-\ell) & x \in E_{\ell}, \ell=0,1, \ldots,\lfloor\beta\rfloor \\ (\sigma(\omega), \beta x-\ell) & x \in S_{\ell} \text { and } \omega_{1}=1, \ell=1, \ldots,\lfloor\beta\rfloor \\ (\sigma(\omega), \beta x-\ell+1) & x \in S_{\ell} \text { and } \omega_{1}=0, \ell=1, \ldots,\lfloor\beta\rfloor\end{cases}
$$

## Random Digits

Let
$d_{1}=d_{1}(\omega, x)= \begin{cases}\ell & \text { if } x \in E_{\ell}, \ell=0,1, \ldots,\lfloor\beta\rfloor, \\ & \text { or }(\omega, x) \in\left\{\omega_{1}=1\right\} \times S_{\ell}, \ell=1,2, \ldots,\lfloor\beta\rfloor, \\ \ell-1 & \text { if }(\omega, x) \in\left\{\omega_{1}=0\right\} \times S_{\ell}, \ell=1,2, \ldots,\lfloor\beta\rfloor,\end{cases}$
then

$$
K_{\beta}(\omega, x)= \begin{cases}\left(\omega, \beta x-d_{1}\right) & \text { if } x \in E \\ \left(\sigma(\omega), \beta x-d_{1}\right) & \text { if } x \in S\end{cases}
$$

## Random $\beta$-expansions

Set $d_{n}=d_{n}(\omega, x)=d_{1}\left(K_{\beta}^{n-1}(\omega, x)\right)$, and let $\pi_{2}: \Omega \times[0,\lfloor\beta\rfloor /(\beta-1)] \rightarrow[0,\lfloor\beta\rfloor /(\beta-1)]$ be the canonical projection onto the second coordinate. Then

$$
\pi_{2}\left(K_{\beta}^{n}(\omega, x)\right)=\beta^{n} x-\beta^{n-1} d_{1}-\cdots-\beta d_{n-1}-d_{n}
$$

rewriting gives

$$
x=\frac{d_{1}}{\beta}+\frac{d_{2}}{\beta^{2}}+\cdots+\frac{d_{n}}{\beta^{n}}+\frac{\pi_{2}\left(K_{\beta}^{n}(\omega, x)\right)}{\beta^{n}}
$$

Since $\pi_{2}\left(K_{\beta}^{n}(\omega, x)\right) \in[0,\lfloor\beta\rfloor /(\beta-1)]$, it follows that

$$
\left|x-\sum_{i=1}^{n} \frac{d_{i}}{\beta^{i}}\right|=\frac{\pi_{2}\left(K_{\beta}^{n}(\omega, x)\right)}{\beta^{n}} \rightarrow 0 \quad \text { as } n \rightarrow \infty
$$

## Random $\beta$-expansions

## Theorem

(D. + de Vries) Suppose $x \in[0,\lfloor\beta\rfloor /(\beta-1)]$ can be written as

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x=\frac{b_{1}}{\beta}+\frac{b_{2}}{\beta^{2}}+\cdots+\frac{b_{n}}{\beta^{n}}+\cdots
$$

with $b_{i} \in\{0,1, \cdots,\lfloor\beta\rfloor\}$. Then, there exists $\omega \in \Omega$ such that $b_{n}=d_{n}(\omega, x)=d_{1}\left(K_{\beta}^{n-1}(\omega, x)\right)$ for all $n \geq 1$.

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The proof relies on the behavior of the sequence $\left\{x_{n}=\sum_{i=1}^{\infty} \frac{b_{n-1+i}}{\beta^{i}}: n \geq 1\right\}$. If the set $N(x)=\left\{n: x_{n} \in S\right\}$ is infinite, then there is a unique $\omega \in \Omega$ such that $b_{n}=d_{n}(\omega, x)$. If $N(x)$ is finite, then there are uncountably many $\omega \in \Omega$ such that $b_{n}=d_{n}(\omega, x)$.

## Unique measure of maximal entropy

The measure of maximal entropy is basically obtained by identifying $K_{\beta}$ with the full shift on $(\lfloor\beta\rfloor+1)$ symbols with the uniform product measure. It is easy to see that the full shift is a factor (no measure yet) of $K_{\beta}$

## Unique measure of maximal entropy

- Consider the Bernoulli shift $\left(D, \mathcal{F}, \mathbb{P}, \sigma^{\prime}\right)$, where $D=\{0,1, \cdots,\lfloor\beta\rfloor\}^{\mathbb{N}}, \mathcal{F}$ the product $\sigma$-algebra, $\mathbb{P}$ the uniform product measure, and $\sigma^{\prime}$ the left shift.
- Define $\phi: \Omega \times[0,\lfloor\beta\rfloor /(\beta-1)] \rightarrow D$ by

- Then, $\phi$ is (i) measurable, (ii) surjective and (iii)
- $\phi$ is not invertible.


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\phi(\omega, x)=\left(d_{1}(\omega, x), d_{2}(\omega, x), \cdots,\right) .
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## Unique measure of maximal entropy

- Let

$$
Z=\left\{(\omega, x) \in \Omega \times[0,\lfloor\beta\rfloor /(\beta-1)]: \pi\left(K_{\beta}^{n}(\omega, x)\right) \in S \text { i.o. }\right\}
$$

$$
D^{\prime}=\left\{\left(b_{1}, b_{2}, \ldots\right) \in D: \sum_{i=1}^{\infty} \frac{b_{j+i-1}}{\beta^{i}} \in S \text { for infinitely many } j \text { 's }\right\}
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\end{aligned}
$$

- Then, (i) $\phi(Z)=D^{\prime}$, (ii) $K_{\beta}^{-1}(Z)=Z$, and (iii) $\left(\sigma^{\prime}\right)^{-1}\left(D^{\prime}\right)=D^{\prime}$.


## Unique measure of maximal entropy

## Lemma

(D. de Vries) Let $\phi^{\prime}$ be the restriction of $\phi$ to $Z$, then $\phi^{\prime}: Z \rightarrow D^{\prime}$ is a measurable bijection, and $\mathbb{P}\left(D^{\prime}\right)=1$.

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Define the $K_{\beta}$-invariant measure $\nu_{\beta}$ by $\nu_{\beta}(A)=\mathbb{P}\left(\phi^{\prime}(Z \cap A)\right)$.

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Define the $K$-invariant measure $\nu_{\beta}$ by $\nu\left({ }_{\beta} A\right)=\mathbb{P}\left(\phi^{\prime}(Z \cap A)\right)$. Then,

## Theorem

(D.+ de Vries) Let $\beta>1$ be a non-integer. Then the dynamical systems $\left(\Omega \times[0,\lfloor\beta\rfloor /(\beta-1)], \nu_{\beta}, K_{\beta}\right)$ and $\left(D, \mathbb{P}, \sigma^{\prime}\right)$ are measurably isomorphic.

## Unique measure of maximal entropy

A consequence of the above theorem is that among all the $K_{\beta}$-invariant measures with support $Z, \nu_{\beta}$ has the largest entropy, namely $\log (1+\lfloor\beta\rfloor)$. It is the only one with support $Z$, and this value of the entropy.

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## Lemma

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This leads to

## Theorem

(D. + de Vries) $\nu_{\beta}$ is the unique $K_{\beta}$-invariant measure of maximal entropy.

## The marginals

The projection of $\nu_{\beta}$ in the second coordinate is the Erdős measure:

$$
\begin{array}{ccc}
\Omega \times[0,\lfloor\beta\rfloor /(\beta-1)] & \xrightarrow{\pi_{2}} & {[0,\lfloor\beta\rfloor /(\beta-1)]} \\
\searrow \phi & \uparrow_{h} \\
& & \{0,1, \ldots,\lfloor\beta\rfloor\}^{\mathbb{N}}
\end{array}
$$

where $h\left(b_{1}, b_{2}, \ldots\right)=\sum_{i=1}^{\infty} \frac{b_{i}}{\beta^{i}}$, i.e. $\nu_{\beta} \circ \pi_{2}^{-1}$ gives the distribution of the random variable $h$.

## The marginals

Projection in the first coordinate need not be product measure, only in special cases, namely if the greedy expansion of 1 has the form

$$
1=\frac{a_{1}}{\beta}+\frac{a_{2}}{\beta^{2}}+\ldots+\frac{a_{n}}{\beta^{n}},
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However, the projection is symmetric:

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\nu_{\beta} \circ \pi_{1}^{-1}\left(\left\{\omega_{1}=i_{1}, \ldots, \omega_{n}=i_{n}\right\}\right)=\nu_{\beta} \circ \pi_{1}^{-1}\left(\left\{\omega_{1}=1-i_{1}, \ldots, \omega_{n}=1-i_{n}\right\}\right) .
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As a consequence,

$$
\nu_{\beta} \circ \pi_{1}^{-1}\left(\left\{\omega_{1}=1\right\}\right)=\nu_{\beta} \circ \pi_{1}^{-1}\left(\left\{\omega_{1}=0\right\}\right)=1 / 2
$$

## The invariant measure $m_{p} \times \mu_{\beta, p}$

For $0<p<1$, let $m_{p}$ be the $(p, 1-p)$ product measure on $\Omega$.

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the measure $m_{p} \times \mu_{\beta, p}$ is $K_{\beta}$-invariant, ergodic and is equivalent to $m_{p}$
11., p has density bounded away from 0 $\mu_{\beta, p}$ satisfies
Let $\nu_{\beta}$ be the measure of maximal entropy, then $\nu_{\beta} \neq m_{p} \times \mu_{\beta, p}$. As a consequence the two measures are mutually singular.

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(ii) $\mu_{\beta, p}$ has density bounded away from 0 .

$$
\begin{aligned}
& \mu_{\beta, p} \text { satisfies } \\
& \qquad \mu_{\beta, p}=p \mu_{\beta, p} \circ T_{\beta}^{-1}+(1-p) \mu_{\beta, p} \circ L_{\beta}^{-1} \\
& \text { Let } \nu_{\beta} \text { be the measure of maximal entropy, then } \nu_{\beta} \neq m_{p} \times \mu_{\beta, p} \text {. As } \\
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(iv) Let $\nu_{\beta}$ be the measure of maximal entropy, then $\nu_{\beta} \neq m_{p} \times \mu_{\beta, p}$. As a consequence the two measures are mutually singular.

## Comments on the construction

To construct $m_{p} \times \mu_{\beta, p}$, we use an intermediate transformation, namely the genuine skew product

$$
R_{\beta}(\omega, x)=\left(\sigma \omega, T_{\omega_{1} x}\right)
$$

where $T_{0}=L_{\beta}$ and $T_{1}=T_{\beta}$.

## Some observations

Let $\mu$ be any Borel probability measure on $[0,\lfloor\beta\rfloor /(\beta-1)]$. The following are equivalent:

- The measure $m_{p} \times \mu$ is $R_{\beta}$-invariant
- $\mu$ satisfies $\mu=p \mu \circ T_{\beta}^{-1}+(1-p) \mu \circ L_{\beta}^{-1}$
- The measure $m_{p} \times \mu$ is $K_{\beta}$-invariant


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## An $R_{\beta}$-invariant measure of product type

We consider a randomized version of the Perron-Frobenius operator defined for probability density functions:

$$
P f=p P_{T_{\beta}}+(1-p) P_{L_{\beta}},
$$

where $P_{T_{\beta}}$, and $P_{L_{\beta}}$ are the Perron-Frobenius operator of $T_{\beta}$ and $L_{\beta}$ respectively.

## An $R_{\beta}$-invariant measure of product type

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(Pelikan) For any probability density $f$, the limit

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exists in $L^{1}$, and $P f^{*}=f^{*}$.
In particular, if we take $f=1$, then $P 1^{*}=1^{*}$, and the measure defined by

$$
\mu_{\beta, p}(A)=\int_{A} 1^{*} d \lambda
$$

satisfies

$$
\mu_{\beta, p}=p \mu_{\beta, p} \circ T_{\beta}^{-1}+(1-p) \mu_{\beta, p} \circ L_{\beta}^{-1}
$$

i.e. $m_{p} \times \mu_{\beta, p}$ is $R_{\beta}$-invariant, and ergodic (follows from the fact that $T_{\beta}$ and $L_{\beta}$ are ergodic w.r.t. an absolutely continuous probability measure).

## Ergodicity w.r.t. $K_{\beta}$

The measure $m_{p} \times \mu$ is $K_{\beta}$-invariant. Ergodicity follows from the following.

- Define $F: \Omega \times[0,\lfloor\beta\rfloor /(\beta-1)] \rightarrow D$ by

$$
F(\omega, x)=\left(d_{1}(\omega, x), d_{1}\left(R_{\beta}(\omega, x)\right), d_{1}\left(R_{\beta}^{2}(\omega, x)\right), \ldots,\right)
$$

Then, $\sigma^{\prime} \circ F=F \circ R_{\beta}$. Hence the measure $m_{p} \times \mu_{\beta, p} \circ F^{-1}$ is $\sigma^{\prime}$-invariant and ergodic.

- The measure $m_{p} \times \mu_{\beta, p} \circ F^{-1}$ is concentrated on $\phi(Z)=D^{\prime}$.
- Therefore, the measure $\rho$ defined by $\rho(A)=m_{p} \times \mu_{\beta, p} \circ F^{-1}(\phi(A \cap Z))$ is $K_{\beta}$-invariant and ergodic.
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- $\rho=m_{p} \times \mu_{\beta, p}$.


## Ergodicity w.r.t. $K_{\beta}$

The measure $m_{p} \times \mu$ is $K_{\beta}$-invariant. Ergodicity follows from the following.

- Define $F: \Omega \times[0,\lfloor\beta\rfloor /(\beta-1)] \rightarrow D$ by

$$
F(\omega, x)=\left(d_{1}(\omega, x), d_{1}\left(R_{\beta}(\omega, x)\right), d_{1}\left(R_{\beta}^{2}(\omega, x)\right), \ldots,\right) .
$$

Then, $\sigma^{\prime} \circ F=F \circ R_{\beta}$. Hence the measure $m_{p} \times \mu_{\beta, p} \circ F^{-1}$ is $\sigma^{\prime}$-invariant and ergodic.

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## Comments on the mutual singularity of $\nu_{\beta}$ and $m_{p} \times \mu_{\beta, p}$

Since $K_{\beta}$ is ergodic w.r.t. $\nu_{\beta}$ and $m_{p} \times \mu_{\beta, p}$, we only need to show that $\nu_{\beta} \neq m_{p} \times \mu_{\beta, p}$.

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The answer is clear for $p \neq 1 / 2$ since

$$
\nu_{\beta}\left(\left\{\omega_{1}=1\right\} \times[0,\lfloor\beta\rfloor /(\beta-1)]\right)=1 / 2,
$$

while

$$
m_{p} \times \mu_{\beta, p}\left(\left\{\omega_{1}=1\right\} \times[0,\lfloor\beta\rfloor /(\beta-1)]\right)=p .
$$

## Comments on the mutual singularity of $\nu_{\beta}$ and $m_{p} \times \mu_{\beta, p}$

Assume $p=1 / 2$. Choose $n$ large enough so that $\left.\left[1 / \beta, 1 / \beta+1 / \beta^{n}\right]\right) \subset S_{1}$. By symmetry of the measure $\nu_{\beta}$ we have

$$
\nu_{\beta}\left(\left\{\omega_{1}=1\right\} \times[0,\lfloor\beta\rfloor /(\beta-1)] \mid \Omega \times\left[1 / \beta, 1 / \beta+1 / \beta^{n}\right)\right)=1 / 2
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On the other hand, if $\nu_{\beta}=m_{p} \times \mu_{\beta, p}$, then using the fact that $\nu_{\beta}$ is the uniform Bernoulli measure on the (random) digits, and that $\mu_{\beta, p}$ is bounded away from 0, we get

$$
\nu_{\beta}\left(\left\{\omega_{1}=1\right\} \times[0,\lfloor\beta\rfloor /(\beta-1)] \mid \Omega \times\left[1 / \beta, 1 / \beta+1 / \beta^{n}\right)\right) \leq C \frac{\beta^{n}}{(1+\lfloor\beta\rfloor)^{n}}
$$

which tends to 0 as $n \rightarrow \infty$, leading to a contradiction.

## Some consequences: uncountably many expansions

Ergodicity of $m_{p} \times \mu_{\beta, p}$ gives

$$
m_{p} \times \mu_{\beta, p}\left(\left\{(\omega, x): K_{\beta}^{i}(\omega, x) \in \Omega \times S \text { i.o. }\right\}\right)=1
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Thus, the set $G=\{x: x$ has a unique $\beta$ - expansion $\}$ has Lebesgue measure 0 . By non-singularity of the greedy and lazy maps, we get that the set

$$
F=\bigcup_{n=1}^{\infty}\left\{x: T_{u_{0}} \circ T_{u_{1}} \circ \ldots \circ T_{u_{n}} \in G \text { for some } u_{1}, \ldots, u_{n}\right\}
$$

has Lebesgue measure zero, where $T_{0}=L_{\beta}$ and $T_{1}=T_{\beta}$.
For $x \notin F$ different elements of $\Omega$ lead to different expansions. Hence a.e. $x$ has uncountably many $\beta$-expansions.

## Some consequences: universal expansions

An expansion of $x$,

$$
x=\sum_{i=1}^{\infty} \frac{b_{i}}{\beta^{i}},
$$

with digits in $A=\{0,1, \cdots,\lfloor\beta\rfloor\}$ is called universal if every possible block of digits in $A$ appears somewhere in the the above expansion of $x$.

Erdős and Komornik (1998) proved that there exists a $\beta_{0} \in(1,2)$ such that for each $\beta \in\left(1, \beta_{0}\right)$, every $x$ has a universal expansion in base $\beta$.

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Sidorov (2003) showed that for each $\beta \in(1,2)$, Lebesgue a.e. point has a universal expansion.

## Some consequences: universal expansions

Using the ergodicity of the map $K_{\beta}$ w.r.t the measure $m_{p} \times \mu_{\beta, p}$, together with the equivalence of the measure $\mu_{\beta, p}$ w.r.t. Lebesgue measure $\lambda$, one can show (using the Ergodic Theorem and Fubini) the following result.

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## Theorem

(D. de Vries) For any non-integer $\beta>1$, and for $\lambda$ a.e.
$x \in[0,\lfloor\beta\rfloor /(\beta-1)]$, there exists a continuum of universal expansions of $x$ in base $\beta$.

## Underlying Markov partition

For the rest of this talk, we assume that the greedy expansion of 1 has the form

$$
1=\frac{a_{1}}{\beta}+\frac{a_{2}}{\beta^{2}}+\ldots+\frac{a_{n}}{\beta^{n}},
$$

with $a_{i}>0$ for $i=1, \ldots, n$.

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with $a_{i}>0$ for $i=1, \ldots, n$.
The underlying dynamics of $K_{\beta}$ is given by a simple Markov chain.

## Underlying Markov partition

Start with the partition $\left\{E_{0}, S_{1}, \ldots, S_{\lfloor\beta\rfloor}, E_{\lfloor\beta\rfloor}\right\}$.

## Underlying Markov partition

Start with the partition $\left\{E_{0}, S_{1}, \ldots, S_{\lfloor\beta\rfloor}, E_{\lfloor\beta\rfloor}\right\}$.
Refine using the orbit of 1 and $\frac{\lfloor\beta\rfloor}{\beta-1}-1$. The refinement gives the desired Markov partition

$$
\left\{C_{0}, C_{1}, \ldots, C_{L}\right\}
$$

where $C_{i}$ is either $S_{j}$ for some $j$, or is a subset of $E_{k}$ for some $k$.

## Measure of maximal entropy

We consider the associated topological Markov chain and its corresponding adjacency matrix. We use the Parry recipe to find the (Markov) measure $Q$ of maximal entropy.
An easy calculation shows that

$$
Q\left(\left[j_{1}, \ldots, j_{\ell}\right]\right)=\frac{v_{j_{\ell}}}{(1+\lfloor\beta\rfloor)^{\ell-1}},
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where the probability vector $v=\left(v_{1}, \ldots, v_{L}\right)$ is a right Perron eigenvalue. When viewed as a measure on $\Omega \times\left[0, \frac{\lfloor\beta\rfloor}{\beta-1}\right]$, one can show that the projection in the first coordinate is the uniform Bernoulli measure (the proof uses the strong Markov property, and the fact the elements of $\Omega$ depend on the times the Markov chain is in the $S$-region)

## The measure $m_{p} \times \mu_{\beta, p}$

To identify the measure $\mu_{\beta, p}$, we consider the transition matrix $P=\left(p_{i, j}\right)$, given by

$$
p_{i, j}= \begin{cases}\lambda\left(C_{i} \cap T_{\beta}^{-1} C_{j}\right) / \lambda\left(C_{i}\right) & \text { if } C_{i} \subseteq E_{k} \text { for some } k, \\ p & \text { if } C_{i} \subseteq S_{k} \text { for some } k \text { and } j=0, \\ 1-p & \text { if } C_{i} \subseteq S_{k} \text { for some } k \text { and } j=L\end{cases}
$$

Denote by $\pi=\left(\pi_{1}, \ldots, \pi_{L}\right)$ the stationary distribution of $P$.

## The measure $m_{p} \times \mu_{\beta, p}$

An easy calculation shows that

$$
\mu_{\beta, p}(B)=\sum_{j=0}^{L} \frac{\lambda\left(B \cap C_{j}\right)}{\lambda\left(C_{j}\right)} \cdot \pi(j) \quad[B \in \mathcal{B}]
$$

and $\mu_{\beta, p}$ has density

$$
\mathbf{1}^{*}=\sum_{i=0}^{L} \frac{\pi_{i}}{\lambda\left(C_{i}\right)} \mathbb{I}_{C_{i}} .
$$

