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| Citation | 0saka Journal of Mathematics. 1999, 36(4), p. <br> $993-1010$ |
| Version Type | VoR |
| URL | https://doi.org/10.18910/10471 |
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# $L^{q}$-SPECTRUM OF BERNOULLI CONVOLUTIONS ASSOCIATED WITH P.V. NUMBERS 

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(Received October 20, 1997)

## 1. Introduction

Let $\mu$ be a positive bounded Borel measure on $\mathbb{R}^{d}$ with bounded support and let $\operatorname{supp}(\mu)$ denote the support of $\mu$. For $\delta>0$ and $q \in \mathbb{R}$, the $L^{q}$-(moment) spectrum of $\mu$ is defined as

$$
\tau(q)=\varliminf_{\delta \rightarrow 0^{+}} \frac{\ln \sup \sum_{i} \mu\left(B_{\delta}\left(x_{i}\right)\right)^{q}}{\ln \delta}
$$

where $\left\{B_{\delta}\left(x_{i}\right)\right\}_{i}$ is a disjoint family of $\delta$-balls with center $x_{i} \in \operatorname{supp}(\mu)$ and the supremum is taken over all such families. For $q>1$, the $L^{q}$-dimension (or generalized Rényi dimension, see e.g. [10], [22]) of $\mu$ is defined as

$$
\underline{\operatorname{dim}}_{q}(\mu)=\frac{\tau(q)}{q-1}
$$

The spectra $\tau(q)$ and $\underline{\operatorname{dim}}_{q}(\mu)$ play a central role in studying the multifractal structure of the measure $\mu$ (e.g., the multifractal formalism [6], [9], [10]) and it is of great interest to compute them. There is a simple formula for $\tau(q)$ if $\mu$ is a self-similar measure defined by an iterated function system of contractive similitudes satisfying the open set condition (OSC) ([3], [4], [8], [18], [19]).

The OSC is a separation condition on the similitudes. In the absence of this condition, the dynamics of the iteration is not clear and very few results are known. In [14] the authors introduced a weak separation condition to study some interesting selfsimilar measures defined by similitudes that do not satisfy the OSC. An important class of examples comes from the self-similar measure $\mu$ satisfying the identity

$$
\begin{equation*}
\mu=\frac{1}{2} \mu \circ \psi_{1}^{-1}+\frac{1}{2} \mu \circ \psi_{2}^{-1} \tag{1.1}
\end{equation*}
$$

where $\psi_{1} x=\rho x, \psi_{2} x=\rho x+(1-\rho), 1 / 2<\rho<1$ ([12], [13], [14]). It is called an infinitely convolved Bernoulli measure (ICBM) because it can be identified (up to a scalar

[^0]multiple) with the random variable $\boldsymbol{X}=\sum_{k=0}^{\infty} \rho^{k} \boldsymbol{X}_{k}$ where $\left\{\boldsymbol{X}_{k}\right\}$ is a sequence of i.i.d. random variables each taking values 0 or 1 with probability $1 / 2$. If $\rho^{-1}$ is a P.V. number, then $\left\{\psi_{1}, \psi_{2}\right\}$ satisfies the above mentioned weak separation condition and $\mu$ is singular ([5], [20]). (Recall that an algebraic integer $\beta>1$ is a Pisot-Vijayaraghavan (P.V.) number if all of its conjugates are in modulus strictly less than 1 . The golden ratio $(\sqrt{5}+1) / 2$ is such a number.)

It is interesting to calculate the exact values of $\tau(q), q \in \mathbb{R}$, for the class of measures in (1.1) because the OSC fails. Only some partial results to this problem are known. For $\rho^{-1}$ equal to a P.V. number, the value of $\tau(2)$ for the associated measure has been calculated in [12] and [13]. For the special case $\rho^{-1}$ equal to the golden ratio, the entropy dimension (corresponding to the $L^{1}$-dimension) has been studied and estimated by a number of authors (e.g., [1], [2], [11], [17]). For this particular measure, an explicit formula defining $\tau(q)$ for $q>0$ was given in [15] recently.

In this paper we continue our study of the ICBM associated with P.V. numbers. Our goal is to obtain a simple algorithm to calculate the $L^{q}$-spectrum $\tau(q)$ for such measures when $q \geq 2$ is an integer. Note that for any bounded positive Borel measure with bounded support, it is known that $\tau(1)=0$ and $-\tau(0)$ is the box dimension of the support of the measure (see e.g. [14]). The basic idea to construct the algorithm can be summarized as follows: First, we observe that for $q>0$,

$$
\begin{equation*}
\tau(q)=\inf \left\{\alpha: \varlimsup_{h \rightarrow 0^{+}} \frac{1}{h^{1+\alpha}} \int_{-\infty}^{\infty} \mu\left(B_{h}(x)\right)^{q} d x>0\right\} \tag{1.2}
\end{equation*}
$$

where $B_{h}(x)$ is the interval $[x-h, x+h$ ) (see [12], [15], [22]). For $q$ equal to a positive integer we let $s=\left(s_{1}, \ldots, s_{q}\right) \in \mathbb{R}^{q}$ and let

$$
\begin{equation*}
\Phi_{s}^{(\alpha)}(h)=\frac{1}{h^{1+\alpha}} \int_{-\infty}^{\infty} \mu\left(B_{h}\left(t+s_{1}\right)\right) \cdots \mu\left(B_{h}\left(t+s_{q}\right)\right) d t . \tag{1.3}
\end{equation*}
$$

By using the self-similar identity (1.1) we can introduce a dynamics on a suitable parameter set of the $s$ including 0 . When $\rho^{-1}$ is a P.V. number, there are only finitely many $s$ 's involved (i.e., $\Phi_{s}^{(\alpha)}(h) \neq 0$ ) and we can represent this dynamics in terms of a sub-Markov matrix. The maximal eigenvalue of the matrix will give the desired $\tau(q)$. This technique is a simplification of that used in [13]. For P.V. numbers that are solutions of the polynomials $x^{n}-x^{n-1}-\cdots-x-1=0, n \geq 2$ (including the golden ratio), the matrix can be reduced to a very simple form.

We organize this paper as follows. In Section 2, we give some algebraic preliminaries and set up the involved matrix. The main result is proved in Section 3. In Section 4 we present techniques to reduce the size of the matrix and describe an algorithm to generate it. In Section 5, we derive an explicit expression for the matrix corresponding to the special class of P.V. numbers mentioned above. Finally in Section 6 , we make some remarks on the more general case when $\psi_{1}, \psi_{2}$ are allowed to take different probability weights.

## 2. Algebraic preliminaries

Let $1 / 2<\rho<1$. For $q \geq 2$ an integer, we define a set of $q$-dimensional vectors by letting $s_{0}=(0, \ldots, 0)$, and for $n \geq 1$,

$$
\begin{equation*}
\boldsymbol{s}_{n}=\rho^{-1}\left(\boldsymbol{s}_{n-1}+(1-\rho) \boldsymbol{\epsilon}\right) \tag{2.1}
\end{equation*}
$$

where $\boldsymbol{\epsilon}=\left(\epsilon_{1}, \ldots, \epsilon_{q}\right), \epsilon_{i}=0$ or 1 .
We first make an important identification for the set of $s_{n}$ 's generated by this iteration process. We identify $\boldsymbol{s}_{n}$ and $\boldsymbol{s}_{m}$ whenever $\boldsymbol{s}_{n}=\boldsymbol{s}_{m}+\boldsymbol{c}$ for some $\boldsymbol{c}=c(1, \ldots, 1)$ (i.e., whenever $s_{n}$ and $s_{m}$ both lie on the same straight line in $\mathbb{R}^{q}$ parameterized by $\left.x_{i}=t+a_{i}, 1 \leq i \leq q\right)$. Let $S$ be the quotient set under such identification. Intuitively, we think of each $s \in S$ as a "line" in $\mathbb{R}^{q}$. It follows from (2.1) that each element in $S$ has a representation of the form

$$
\frac{1-\rho}{\rho} \sum_{k=0}^{n} \rho^{-(n-k)} \epsilon_{k}
$$

Let

$$
S_{1}=\left\{\left(s_{1}, \ldots, s_{q}\right) \in S:\left|s_{i}-s_{j}\right| \leq 1 \text { for all } 1 \leq i, j \leq q\right\} .
$$

Geometrically, $S_{1}$ consists of those lines in $S$ that intersect the unit cube $[0,1]^{q}$.
We consider $S$ to be a set of states that spans some vector space $\langle S\rangle$ (i.e., $S$ is a basis for the vector space). Define a Markov matrix $T$ on $S$ by

$$
\begin{equation*}
T(s)=\frac{1}{2^{q}} \cdot \sum_{\epsilon}^{\prime} \boldsymbol{s}^{\epsilon} \tag{2.2}
\end{equation*}
$$

where $\left(s^{\boldsymbol{\epsilon}}\right)_{i}=\rho^{-1}\left(s_{i}+(1-\rho) \epsilon_{i}\right), \boldsymbol{\epsilon}=\left(\epsilon_{1}, \ldots, \epsilon_{q}\right), \epsilon_{i}=0$ or 1 , and the summation is taken over all such $\epsilon$. We emphasize that the operations • and $\sum^{\prime}$ in (2.2) are respectively scalar multiplication and addition in the vector space $\langle S\rangle$. They should be distinguished from the linear combinations in $\mathbb{R}^{d}$ by regarding the $s$ as usual vectors, as those in (2.1). We also note that the sum of the entries of each column of $T$ is 1 .

Proposition 2.1. $T$ is invariant on the subspace of $\langle S\rangle$ spanned by $S \backslash S_{1}$.
Proof. Let $s \in S \backslash S_{1}$ with $\left|s_{i}-s_{j}\right|>1$ and let $\boldsymbol{t}=\boldsymbol{s}^{\boldsymbol{\epsilon}}$. Then

$$
\left|t_{i}-t_{j}\right|=\rho^{-1}\left|\left(s_{i}-s_{j}\right)+(1-\rho)\left(\epsilon_{i}-\epsilon_{j}\right)\right|>\rho^{-1}(1-(1-\rho))=1
$$

This implies that $t \in S \backslash S_{1}$. The assertion follows since $T(s)$ is a linear combination, in $\langle S\rangle$, of the $s^{\epsilon}$.

Proposition 2.2. If $\rho^{-1}$ is a P.V. number, then $S_{1}$ is a finite set.
Proof. The proof of this result can be found in [13]. We give another proof using a lemma of Garsia ([7, Lemma 1.51]): Let $\beta>1$ be an algebraic integer, let $\beta_{1}, \ldots, \beta_{\ell}$ be the algebraic conjugates of $\beta$ and let $\sigma$ be the number of $\beta_{i}$ such that $\left|\beta_{i}\right|=1$. For an $n$-th degree polynomial $L$ with integer coefficients $a_{i}$ and height $M:=\max \left\{\left|a_{i}\right|: i=1, \ldots, n\right\}$, if $L(\beta) \neq 0$, then

$$
|L(\beta)| \geq \frac{\prod_{\left|\beta_{i}\right|=1}| | \beta_{i}|-1|}{(n+1)^{\sigma}\left(\prod_{\beta_{i}>1}\left|\beta_{i}\right|\right)^{n+1} \dot{M}^{\ell}}
$$

Now if $\rho^{-1}=\beta$ is a P.V. number, then the above reduces to

$$
\begin{equation*}
|L(\beta)| \geq M^{-\ell} \prod_{\left|\beta_{i}\right| \neq 1}| | \beta_{i}|-1|:=C . \tag{2.3}
\end{equation*}
$$

We observe that for $s \in S, s_{j}=\frac{1-\rho}{\rho} \sum_{k=0}^{n} \beta^{n-k} \epsilon_{k}^{(j)}$. Hence $s \in S_{1}$ if and only if

$$
\left|\sum_{k=0}^{n} \beta^{n-k}\left(\epsilon_{k}^{(i)}-\epsilon_{k}^{(j)}\right)\right| \leq \frac{1}{\beta-1}, \quad \text { for all } 1 \leq i, j \leq q
$$

Therefore to show that $S_{1}$ is finite, it suffices to show that the set $B$ defined below is finite:

$$
B=\bigcup_{n=0}^{\infty}\left\{y_{n}=\sum_{k=0}^{n} \beta^{n-k} \eta_{k}: \eta_{k}=0 \text { or } \pm 1,\left|y_{n}\right| \leq \frac{1}{\beta-1}\right\}
$$

If $y_{n} \neq y_{m}$ are two elements in $B$ with $n \geq m$, then

$$
y_{n}-y_{m}=\sum_{k=0}^{n} \beta^{n-k} \eta_{k}-\sum_{k=0}^{m} \beta^{m-k} \eta_{k}^{\prime} .
$$

We use this to define a polynomial $L$ with coefficients $\eta_{k}-\eta_{k}^{\prime}$ (letting $\eta_{k}^{\prime}=0$ for $m<$ $k \leq n$ ). In this case, $L$ has height at most 2. It follows from (2.3) that $\left|y_{n}-y_{m}\right| \geq C$. Since all elements of $B$ are bounded in between $\pm 1 /(\beta-1), B$ must be a finite set and hence $S_{1}$ must also be finite.

It follows from Propositions 2.1 and 2.2 that if $\rho^{-1}$ is a P.V. number, then the matrix $T$ is of the form

$$
T=\left[\begin{array}{cc}
T_{1} & \mathbf{0}  \tag{2.4}\\
Q & T_{2}
\end{array}\right]
$$

where $T_{1}$ corresponds to the states $S_{1}$ and is a finite sub-Markov matrix. The matrix $T_{1}$ is the one we need to calculate the spectrum $\tau(q)$.

We will make a further identification for the states in $S_{1}$. For $s=\left(s_{1}, \ldots, s_{q}\right) \in$ $S_{1}$, we let $\sigma$ (depending on $s$ ) be a particular permutation on $\{1,2, \ldots, q\}$ such that the state $s_{\sigma}=\left(s_{\sigma(1)}, \ldots, s_{\sigma(q)}\right) \in S_{1}$ satisfies $s_{\sigma(1)} \geq s_{\sigma(2)} \geq \cdots \geq s_{\sigma(q)}$. Let $S_{1}^{\sigma}=\left\{s_{\sigma}: s \in S_{1}\right\}$ and define

$$
\pi: S_{1} \rightarrow S_{1}^{\sigma} \quad \text { by } \pi(s)=s_{\sigma} \quad \text { and } \quad T_{1}^{\sigma}: S_{1}^{\sigma} \rightarrow S_{1}^{\sigma} \quad \text { by } T_{1}^{\sigma}=\pi \circ T_{1}
$$

Extend $\pi$ and $T_{1}^{\sigma}$ linearly to $\left\langle S_{1}\right\rangle$ and $\left\langle S_{1}^{\sigma}\right\rangle$ respectively. (Note that we have used $T_{1}$ both as a matrix and an operator. This slight abuse of notation will also apply to other matrices and operators throughout the paper.)

It is easy to see that the entry $(s, t) \in S_{1}^{\sigma} \times S_{1}^{\sigma}$ of the matrix $T_{1}^{\sigma}$ is given by

$$
\frac{1}{2^{q}} \#\left\{s^{\epsilon} \in S_{1}:\left(s^{\epsilon}\right)_{\sigma}=\boldsymbol{t}, \boldsymbol{\epsilon}=\left(\epsilon_{1}, \ldots, \epsilon_{q}\right), \epsilon_{i}=0, \text { or } 1\right\}
$$

where $s^{\epsilon}$ is defined as in (2.2) and we use the notation $\# E$ to denote the cardinality of a set $E$.

Proposition 2.3. Let $1 / 2<\rho<1$ such that $\rho^{-1}$ is a P.V. number. Then the maximal eigenvalues of $T_{1}$ and $T_{1}^{\sigma}$ are equal.

Proof. It follows directly from (2.2) that $T_{1}^{\sigma} \circ \pi=\pi \circ T_{1}$ on $\left\langle S_{1}\right\rangle$. Let $\lambda$ and $\lambda^{\sigma}$ be the maximal eigenvalues of $T_{1}$ and $T_{1}^{\sigma}$ respectively. Suppose $s$ is a nonnegative $\lambda$-eigenvector of $T_{1}$. Then $s=\sum_{i}^{\prime} c_{i} \cdot s_{i}$, where $c_{i} \geq 0$ and not all $c_{i}$ are zero. This implies that $\pi(s) \neq 0$. Since

$$
T_{1}^{\sigma}(\pi(\boldsymbol{s}))=\pi\left(T_{1}(\boldsymbol{s})\right)=\pi(\lambda \cdot \boldsymbol{s})=\lambda \cdot \pi(\boldsymbol{s})
$$

we conclude that $\lambda$ is an eigenvalue of $T_{1}^{\sigma}$, and $\lambda \leq \lambda^{\sigma}$.
On the other hand taking the adjoint of the identity $T_{1}^{\sigma} \circ \pi=\pi \circ T_{1}$, we have $\pi^{*} \circ\left(T_{1}^{\sigma}\right)^{*}=T_{1}^{*} \circ \pi^{*}$. The eigenvalues of $T_{1}{ }^{*}$ and $\left(T_{1}^{\sigma}\right)^{*}$ are unchanged, and the same argument as above implies that $\lambda^{\sigma} \leq \lambda$. (We can also consider the left eigenvector instead of using the adjoints.)

## 3. The basic theorem.

Let $\mu$ be the ICBM as defined in (1.1). For each $s \in S, \alpha \geq 0$ and $h>0$, we define

$$
\begin{equation*}
\Phi_{\boldsymbol{s}}^{(\alpha)}(h)=\frac{1}{h^{1+\alpha}} \int_{-\infty}^{\infty} \mu\left(B_{h}\left(t+s_{1}\right)\right) \cdots \mu\left(B_{h}\left(t+s_{q}\right)\right) d t \tag{3.1}
\end{equation*}
$$

It is straight forward to verify the following proposition, which justifies the identifications made in the previous section.

Proposition 3.1. Let $\Phi_{s}^{(\alpha)}(h)$ be defined as in (3.1). Then the following hold:
(a)If $s^{\prime}$ is another representation of $s$, i.e., $s^{\prime}=s+c(1, \ldots, 1)$ for some $c \in \mathbb{R}$, then $\Phi_{\boldsymbol{s}^{\prime}}^{(\alpha)}(h)=\Phi_{\boldsymbol{s}}^{(\alpha)}(h)$.
(b)Let $s_{\sigma}$ be the decreasing rearrangement of the coordinates of $s$ as defined in the previous section. Then $\Phi_{\boldsymbol{s}_{\sigma}}^{(\alpha)}(h)=\Phi_{\boldsymbol{s}}^{(\alpha)}(h)$.

Proposition 3.2. $\Phi_{\boldsymbol{s}}^{(\alpha)}(h) \neq 0$ for all $h>0$ if and only if $s \in S_{1}$.

Proof. The proposition is a simple consequence of the following observation: $s \in$ $S_{1}$ if and only if the line $\boldsymbol{t}+\boldsymbol{s}$ with $\boldsymbol{t}=(t, \ldots, t),-\infty<t<\infty$ has nonvoid intersection with $[0,1]^{q}$. Using the fact that $\operatorname{supp}(\mu)=[0,1]$, it is easy to show that this is equivalent to

$$
\int_{-\infty}^{\infty} \mu\left(B_{h}\left(t+s_{1}\right)\right) \cdots \mu\left(B_{h}\left(t+s_{q}\right)\right) d t \neq 0 \quad \text { for all } h>0
$$

i.e., $\Phi_{\boldsymbol{s}}^{(\alpha)}(h) \neq 0$ for all $h>0$.

For $s \in\langle S\rangle$ with $s=\sum^{\prime} c_{i} \cdot s_{i}, s_{i} \in S$, we define

$$
\Phi_{\boldsymbol{s}}^{(\alpha)}(h)=\sum c_{i} \Phi_{\boldsymbol{s}_{i}}^{(\alpha)}(h)
$$

Hence $\Phi_{T \boldsymbol{s}}^{(\alpha)}(h)=2^{-q} \sum_{\epsilon} \Phi_{\boldsymbol{s}^{\epsilon}}^{(\alpha)}(h)$. The Markov matrix $T$ has the following important invariance property.

Proposition 3.3. Let $s \in\langle S\rangle$. Then for any $\alpha \geq 0$ and any $h>0$,

$$
\begin{equation*}
\Phi_{\boldsymbol{s}}^{(\alpha)}(h)=\frac{1}{\rho^{\alpha}} \Phi_{T \boldsymbol{s}}^{(\alpha)}\left(\frac{h}{\rho}\right) \tag{3.2}
\end{equation*}
$$

Proof. By linearity, it suffices to show that this holds for all $s \in S$. Using the self-similar identity (1.1) followed by a change of variables, we have

$$
\begin{aligned}
\Phi_{s}^{(\alpha)}(h) & =\frac{1}{2^{q} h^{1+\alpha}} \int_{-\infty}^{\infty} \prod_{i=1}^{q}\left(\mu\left(B_{\frac{h}{\rho}}\left(\frac{t}{\rho}+\frac{s_{i}}{\rho}\right)\right)+\mu\left(B_{\frac{h}{\rho}}\left(\frac{t}{\rho}+\frac{s_{i}}{\rho}-\frac{1-\rho}{\rho}\right)\right)\right) d t \\
& =\frac{1}{2^{q} h^{1+\alpha}} \sum_{\epsilon} \int_{-\infty}^{\infty} \prod_{i=1}^{q} \mu\left(B_{\frac{h}{\rho}}\left(\frac{t}{\rho}+\frac{s_{i}}{\rho}-\epsilon_{i} \frac{1-\rho}{\rho}\right)\right) d t
\end{aligned}
$$

$$
\begin{aligned}
& =\frac{1}{2^{q} \rho^{\alpha}\left(\frac{h}{\rho}\right)^{1+\alpha}} \sum_{\epsilon} \int_{-\infty}^{\infty} \prod_{i=1}^{q} \mu\left(B_{\frac{h}{\rho}}\left(t+\frac{s_{i}}{\rho}+\epsilon_{i} \frac{1-\rho}{\rho}\right)\right) d t \\
& =\frac{1}{\rho^{\alpha}} \Phi_{T s}^{(\alpha)}\left(\frac{h}{\rho}\right)
\end{aligned}
$$

where the summation $\sum_{\epsilon}$ is over all $\epsilon=\left(\epsilon_{1}, \ldots, \epsilon_{q}\right), \epsilon_{i}=0$ or 1 .
Proposition 3.4. Let $\mu$ be the self-similar measure defined by (1.1) with $\rho^{-1}$ equal to a P.V. number, let $T_{1}$ be defined as in (2.4) and let $\lambda$ be its maximal eigenvalue. Then $\rho^{q-1}<\lambda<1$.

Proof. In view of (2.4) and the fact that $\left\langle S_{1}\right\rangle$ contains no invariant subspaces of $T$, it is easy to show that the maximal eigenvalue of $T_{1}$ is strictly less than 1 (see e.g. [21]).

To prove the lower bound estimate for $\lambda$, we first claim that if $\alpha$ is such that $\rho^{\alpha}=\lambda$, then for any $\eta<\alpha$,

$$
\begin{equation*}
\varlimsup_{h \rightarrow 0^{+}} \frac{1}{h^{1+\eta}} \int_{-\infty}^{\infty} \mu\left(B_{h}(t)\right)^{q} d t=0 \tag{3.3}
\end{equation*}
$$

For this we first consider the case $T_{1}$ is irreducible. Let $s=\sum_{i}^{\prime} c_{i} \cdot s_{i}$ be a positive eigenvector associated with $\lambda$. Then Propositions 3.2 and 3.3 imply that for $h>0$ sufficiently small,

$$
\Phi_{\boldsymbol{s}}^{(\eta)}(h)=\frac{1}{\rho^{\eta}} \Phi_{T_{1} \boldsymbol{s}}^{(\eta)}\left(\frac{h}{\rho}\right)=\frac{\lambda}{\rho^{\eta}} \Phi_{\boldsymbol{s}}^{(\eta)}\left(\frac{h}{\rho}\right)
$$

Inductively, for all $m \in \mathbb{N}$ and for all $h>0$ sufficiently small,

$$
\begin{equation*}
\Phi_{\boldsymbol{s}}^{(\eta)}\left(\rho^{m} h\right)=\left(\frac{\lambda}{\rho^{\eta}}\right)^{m} \Phi_{\boldsymbol{s}}^{(\eta)}(h) \tag{3.4}
\end{equation*}
$$

Since $\lambda / \rho^{\eta}<1$ by assumption, we have $\lim _{h \rightarrow 0^{+}} \Phi_{s}^{(\eta)}(h)=0$. The irreducibility of $T_{1}$ implies that each $c_{i}$ is positive (see e.g. [21]). Hence, for $s_{0}=(0, \ldots, 0)$, $\lim _{h \rightarrow 0^{+}} \Phi_{s_{0}}^{(\eta)}(h)=0$, which proves (3.3). In the case $T_{1}$ is reducible, by re-arranging the basis elements, we can assume that

$$
T_{1}=\left[\begin{array}{ccccc}
E_{\ell} & 0 & 0 & \ldots & 0 \\
\times & E_{\ell-1} & 0 & \ldots & 0 \\
\vdots & \ddots & \ddots & \ddots & \vdots \\
\times & & & & 0 \\
\times & \times & \cdots & \times & E_{1}
\end{array}\right]
$$

where $E_{i}, 1 \leq i \leq \ell$ are irreducible. An inductive argument will yield the same conclusion [13, Lemma 4.4]. This proves the claim.

Now suppose $\lambda<\rho^{q-1}$. Then $\lambda=\rho^{\alpha}$ for some $\alpha>q-1$. By taking $\eta=q-1$ in the above claim, we have

$$
\begin{equation*}
\varlimsup_{h \rightarrow 0^{+}} \frac{1}{h^{q}} \int_{-\infty}^{\infty} \mu\left(B_{h}(t)\right)^{q} d t=0 \tag{3.5}
\end{equation*}
$$

which implies that

$$
\sup _{h>0} \frac{1}{h^{q}} \int_{-\infty}^{\infty} \mu\left(B_{h}(t)\right)^{q} d t<\infty
$$

By using the same argument as in Corollary 4.5 in [13], we have

$$
\varlimsup_{h \rightarrow 0^{+}} \frac{1}{h^{q}} \int_{-\infty}^{\infty} \mu\left(B_{h}(t)\right)^{q} d t>0
$$

contradicting (3.5). Hence $\lambda \geq \rho^{q-1}$.
It remains to show that $\lambda \neq \rho^{q-1}$. Let $q(x)$ be the characteristic polynomial of $T_{1}$. Suppose, on the contrary, $\lambda=\rho^{q-1}$. Let $\beta=\rho^{-1}$ and let $p(x)=\sum_{k=0}^{n} a_{k} x^{k}$ be the minimal polynomial of $\beta^{q-1}$, which is also a P.V. number (see e.g. [20, p.4, Theorem A]). Let $\beta_{1}$ be a conjugate of $\beta^{q-1}$. Then both $\rho^{q-1}$ and $\beta_{1}^{-1}$ are roots of the polynomial $\tilde{p}(x)=\sum_{k=0}^{n} a_{n-k} x^{k}$. This implies that $\beta_{1}^{-1}$ is also a root of $q(x)$. But $\left|\beta_{1}^{-1}\right|>1>\rho^{q-1}$, contradicting the maximality of $\lambda$.

Theorem 3.5. Let $q \geq 2$ be a positive integer, let $1 / 2<\rho<1$ such that $\beta=\rho^{-1}$ is a P.V. number and let $\mu$ be the self-similar measure defined by (1.1). Then $\tau(q)=$ $\ln \lambda / \ln \rho$ where $\lambda$ is the the maximal eigenvalue of $T_{1}$.

Proof. Let $\alpha=\ln \lambda / \ln \rho$. For $s \in\left\langle S_{1}\right\rangle$ a $\lambda$-eigenvector of $T_{1}$, a similar derivation as that for (3.4) yields

$$
\Phi_{\boldsymbol{s}}^{(\alpha)}\left(\rho^{m} h\right)=\Phi_{\boldsymbol{s}}^{(\alpha)}(h), \quad \text { for } m \in \mathbb{N} \text { and for all } h>0 \text { sufficiently small, }
$$

i.e., $\Phi_{\boldsymbol{s}}^{(\alpha)}(h)$ is multiplicatively periodic on $h$. Observe also that $\Phi_{\boldsymbol{s}}^{(\alpha)}(h)$ is strictly positive because $s=\sum_{i}^{\prime} c_{i} \cdot s_{i}$, where $c_{i} \geq 0$ and $s_{i} \in S_{1}$ (Proposition 3.2). Hence there exists $s_{i}$ such that $\varlimsup_{h \rightarrow 0^{+}} \Phi_{\boldsymbol{s}_{i}}^{(\alpha)}(h)>0$. By using Hölder's inequality we have

$$
\varlimsup_{h \rightarrow 0^{+}} \frac{1}{h^{1+\alpha}} \int_{-\infty}^{\infty} \mu\left(B_{h}(t)\right)^{q} d t>0
$$

It follows that $\tau(q) \leq \alpha=\ln \lambda / \ln \rho$. On the other hand, for any $\eta<\alpha$, the claim in the proof of Proposition 3.4 implies that

$$
\varlimsup_{h \rightarrow 0^{+}} \frac{1}{h^{1+\eta}} \int_{-\infty}^{\infty} \mu\left(B_{h}(t)\right)^{q} d t=0
$$

Hence $\tau(q) \geq \alpha$ and the proof is complete.

## 4. Simplification of the matrix $T_{1}$.

To actually compute the maximal eigenvalue of $T_{1}$, it is desirable to replace $T_{1}$ by a matrix of smaller order but has the same maximal eigenvalue. First, by making use of Proposition 2.3, we can reduce $T_{1}$ to $T_{1}^{\sigma}$. Recall that $T_{1}^{\sigma}$ is defined on the span of the set of states $S_{1}^{\sigma}$, which consists only of those $s \in S_{1}$ with $s_{1} \geq s_{2} \geq \cdots \geq s_{q}$.

In this section we will identify each state $s=\left(s_{1}, \ldots, s_{q}\right) \in S_{1}^{\sigma}$ uniquely with the point $\left(s_{1}-s_{q}, \ldots, s_{q-1}-s_{q}\right) \in \mathbb{R}^{q-1}$. Geometrically this corresponds to identifying the "line" $s \in \mathbb{R}^{q}$ with its "point" of intersection with the hyperplane $\mathbb{R}^{q-1}$. We then describe an algorithm to generate the set $W_{1}^{\sigma}$ of all such points and construct the matrix $A_{1}$ induced by $T_{1}^{\sigma}$ and such identification. $A_{1}$ and $T_{1}^{\sigma}$ have the same maximal eigenvalue (Proposition 4.1). In Proposition 4.2 we will further simplify the matrix $A_{1}$.

Let $\tau$ be the projection of $\mathbb{R}^{q}$ onto $\mathbb{R}^{q-1}$ defined by $\tau(s)=\left(s_{1}-s_{q}, \ldots, s_{q-1}-s_{q}\right)$ and define $W_{1}^{\sigma}=\tau\left(S_{1}^{\sigma}\right)$. Let $A_{1}$ be the matrix that is defined on the states $W_{1}^{\sigma}$ and is induced by $T_{1}^{\sigma}$ and $\tau$, i.e., $A_{1}$ is defined by the identity

$$
\begin{equation*}
A_{1} \circ \tau=\tau \circ T_{1}^{\sigma} \quad \text { on } S_{1}^{\sigma} . \tag{4.1}
\end{equation*}
$$

Proposition 4.1. Assume the same hypotheses of Theorem 3.5. Then

$$
\begin{array}{r}
W_{1}^{\sigma}=\left\{t \in \mathbb{R}^{q-1}: t_{i}=(\beta-1) \sum_{k=0}^{n} \beta^{n-k}\left(\epsilon_{k}^{(i)}-\epsilon_{k}^{(q)}\right), \epsilon_{k}^{(j)}=0 \text { or } 1 \text { for } 1 \leq j \leq q\right. \\
\text { and } \left.0 \leq k \leq n, 1 \geq t_{1} \geq \cdots \geq t_{q-1} \geq 0, \text { and } n \in \mathbb{N}\right\}
\end{array}
$$

Moreover, $A_{1}$ and $T_{1}^{\sigma}$ have the same maximal eigenvalue.
Proof. The first part is a direct consequence of the explicit form of the states in $S_{1}^{\sigma}$ and the definition of $\tau$. The second part follows by using the identity in (4.1) and the same argument as that in the proof of Proposition 2.3.

Proposition 4.1 provides us with a convenient algebraic criterion to determine whether a state in $\tau\left(S^{\sigma}\right)$ belongs to $W_{1}^{\sigma}$. Summarizing the previous arguments, we have the following

## Algorithm to construct $A_{1}$ :

(I) Starting from $\mathbf{0} \in \mathbb{R}^{q}$, suppose we have constructed $\boldsymbol{t} \in W_{1}^{\boldsymbol{\sigma}}$ in the ( $n-1$ )-th step. Let $\boldsymbol{s}=\rho^{-1}((\boldsymbol{t}, 0)+(1-\rho) \boldsymbol{\epsilon}), \epsilon_{i}=0$ or $1,1 \leq i \leq q$. Rearrange $\boldsymbol{s}$ to $\boldsymbol{s}_{\sigma}$ so that $s_{\sigma(1)} \geq s_{\sigma(2)} \geq \cdots \geq s_{\sigma(q)}$ and let

$$
\boldsymbol{t}^{\prime}=\left(s_{\sigma(1)}-s_{\sigma(q)}, \ldots, s_{\sigma(q-1)}-s_{\sigma(q)}\right) .
$$

Keep those $t^{\prime}$ in $W_{1}^{\sigma}$ that are distinct from those previously chosen. (The process terminates when no more new members are generated.)
(II) For the column of the matrix $A_{1}$ corresponding to $\boldsymbol{t}$, the entry corresponding to $\boldsymbol{t}^{\prime} \in W_{1}^{\sigma}$ is given by

$$
\frac{1}{2^{q}}\left(\text { number of appearances of the } s \text { that gives } t^{\prime}\right) .
$$

We can further reduce the set of states in $W_{1}^{\sigma}$ by discarding those states of the form $\left(1, \ldots, 1, t_{i+1}, \ldots, t_{q-1}\right)$. These states correspond to those lines $s \in S_{1}^{\sigma}$ that intersect only the boundary of the unit cube $[0,1]^{q}$.

Proposition 4.2. Assume the same hypotheses of Theorem 3.5. Let

$$
W_{0}^{\sigma}=\left\{\left(t_{1}, t_{2}, \ldots, t_{q-1}\right) \in W_{1}^{\sigma}: t_{1}<1\right\}
$$

and let $A_{0}$ be the restriction of $A_{1}$ on $W_{0}^{\sigma}$. Then $A_{0}$ and $A_{1}$ have the same maximal eigenvalue.

Proof. We can decompose $W_{1}^{\sigma} \backslash W_{0}^{\sigma}$ into the following disjoint sets

$$
U_{i}=\left\{\left(1, \ldots, 1, t_{i+1}, \ldots, t_{q-1}\right) \in W_{1}^{\sigma}: t_{i+1}<1\right\}, \quad 1 \leq i \leq q-2
$$

Let $A_{1, i}$ be the matrix obtained by restricting $A_{1}$ on $U_{i}$. It is easy to check that $A_{1}\left(U_{i}\right) \subseteq\left\langle U_{i} \bigcup \cdots \bigcup U_{q-2}\right\rangle$. Hence we can represent the matrix $A_{1}$ on $W_{1}^{\sigma}$ as

$$
A_{1}=\left[\begin{array}{ccccc}
A_{0} & 0 & 0 & \cdots & 0 \\
\times & A_{1,1} & 0 & \cdots & 0 \\
\vdots & \ddots & \ddots & \ddots & \vdots \\
& & & & 0 \\
\times & \times & \cdots & \times & A_{1, q-2}
\end{array}\right]
$$

To compare the maximal eigenvalues of $A_{0}$ and $A_{1,1}$, we observe that the action of the iteration $\boldsymbol{s}=\rho^{-1}((\boldsymbol{t}, 0)+(1-\rho) \boldsymbol{\epsilon})$ (as in part (I) of the above algorithm) on

$$
\left\{\boldsymbol{t}=\left(t_{1}, t_{2}, \ldots, t_{q-2}, 0\right): \boldsymbol{t} \in W_{0}^{\sigma}\right\} \subseteq W_{0}^{\sigma}
$$

is the same as that on $U_{1}=\left\{\boldsymbol{t}=\left(1, t_{2}, \ldots, t_{q-1}\right): t_{2}<1\right\}$. This implies that $A_{1,1}$ is a principal submatrix of $A_{0}$, so the maximal eigenvalue $\lambda_{0}$ of $A_{0}$ is larger. Inductively we can use the same method to compare $A_{i}$ and $A_{i+1}$ and conclude that $\lambda_{0}$ is the maximal eigenvalue of the matrix $A_{1}$.

By using this proposition we can modify the above algorithm by replacing $W_{1}^{\sigma}$ and $A_{1}$ with $W_{0}^{\sigma}$ and $A_{0}$ respectively. We end this section by giving the following commutative diagram which shows the relationships among the different vector spaces and the associated operators. (Here $\left.\mathrm{Id}\right|_{E}$ denotes the restriction of the identity map to the space $E$.)


## 5. A special family of P.V. numbers.

In this section we will consider the measure associated with the special family of P.V. numbers $\rho^{-1}=\beta_{n}$ defined by the algebraic equations

$$
\begin{equation*}
x^{n}-x^{n-1}-\cdots-x-1=0, \quad n \geq 2 . \tag{5.1}
\end{equation*}
$$

This family includes the golden ratio $\beta_{2}=(\sqrt{5}+1) / 2$. For this class of numbers, we will derive a formula for the matrix $A_{0}$ for each integer $q \geq 2$. For completeness we include a proof of the following known result:

Proposition 5.1. The $\beta_{n}>1$ satisfying (5.1) is a P.V. number.
Proof. We can write (5.1) as $x^{n}=\left(x^{n}-1\right) /(x-1)$ and get

$$
\begin{equation*}
x^{n+1}-2 x^{n}+1=0 \tag{5.2}
\end{equation*}
$$

Except for the extra root $x=1$, equation (5.2) has exactly the same set of roots as (5.1). Let $f(z)=z^{n+1}-2 z^{n}+1$ and $g(z)=-2 z^{n}+1$. For $\epsilon>0$ let $C_{\epsilon}$ denote the circle $\{z:|z|=1+\epsilon\}$. Then for $\epsilon>0$ sufficiently small we have

$$
|f(z)-g(z)|=\left|z^{n+1}\right|<\left|-2 z^{n}+1\right|=|g(z)|, \quad z \in C_{\epsilon} .
$$

Hence by Rouché's theorem $f(z)$ and $g(z)$ have the same number of zeros inside $C_{\epsilon}$. Clearly $g(z)$ has $n$ zeros inside $C_{\epsilon}$, and therefore so does $f(z)$. By letting $\epsilon \rightarrow 0$, it follows that $f(z)$ must have $n$ zeros in $\{z:|z| \leq 1\}$. It is easy to see that 1 is the only zero of $f(z)$ on the unit circle and we conclude that (5.1) has $n-1$ roots of modulus less than one. Lastly by writing (5.1) as $x=1+x^{-1}+\cdots+x^{-(n-1)}$, it is obvious that it has a root $\beta_{n}>1$. This completes the proof.

Proposition 5.2. Let $n \geq 2$ and let $\beta_{n}$ be the P.V. number defined by (5.1). Then (a) $1<\beta_{n}<2$; (b) $\left\{\beta_{n}\right\}_{n=2}^{\infty}$ is an increasing sequence and $\lim _{n \rightarrow \infty} \beta_{n}=2$.

Proof. Write $\beta=\beta_{n}$. (5.1) implies that

$$
\begin{equation*}
\beta=1+\beta^{-1}+\cdots+\beta^{-(n-1)}=\frac{1-\beta^{-n}}{1-\beta^{-1}} \tag{5.3}
\end{equation*}
$$

That $\beta>1$ is obvious. Multiplying both sides by $1-\beta^{-1}$, we get $\beta=2-\beta^{-n}<2$ and (a) follows.

To prove (b) we let $g_{n}(x)=1+x^{-1}+\cdots+x^{-(n-1)}$. Then $\beta_{n}$ is a solution of $x=g_{n}(x)$. Observe that since $g_{n}(x)<g_{n+1}(x)$ for all $x>0$, we have

$$
0=g_{n}\left(\beta_{n}\right)-\beta_{n}<g_{n+1}\left(\beta_{n}\right)-\beta_{n}
$$

Moreover, $g_{n+1}(x)-x$ is a strictly decreasing function of $x$ on $[0, \infty)$ with a unique zero at $x=\beta_{n+1}$. It follows that $\beta_{n}<\beta_{n+1}$. That $\lim _{n \rightarrow \infty} \beta_{n}=2$ now follows from (5.3).

For fixed $\beta=\beta_{n}$ with $n \geq 2$, we let $v_{0}=0, v_{1}=1$, and

$$
v_{m}=\beta^{m-1}-\beta^{m-2}-\cdots-\beta-1 \quad \text { for } 2 \leq m \leq n
$$

Lemma 5.3. The finite sequence $\left\{v_{m}\right\}_{m=1}^{n}$ is strictly decreasing and

$$
\begin{equation*}
\frac{1}{\beta-1}-1 \leq v_{m} \leq 1, \quad \text { for } 1 \leq m \leq n \tag{5.4}
\end{equation*}
$$

Proof. Since $v_{m+1}-v_{m}=\beta^{m}-2 \beta^{m-1}=\beta^{m-1}(\beta-2)<0,\left\{v_{m}\right\}_{m=1}^{n}$ is strictly decreasing and the upper bound in (5.4) follows. For the lower bound, we notice that by (5.1), $v_{n}=1 / \beta$. Hence for $1 \leq m \leq n$,

$$
v_{m}-\left(\frac{1}{\beta-1}-1\right) \geq v_{n}-\left(\frac{1}{\beta-1}-1\right)=\frac{\beta(\beta-1)-1}{\beta(\beta-1)} \geq 0
$$

(The last inequality is because $\beta(\beta-1)-1 \geq \beta_{2}\left(\beta_{2}-1\right)-1=0$.) This completes the proof.

Let $q \geq 2$ be a positive integer and $n \geq 2$ be the degree of the polynomial in (5.1). For $0 \leq k \leq q-1$ and $0 \leq m \leq n$, let

$$
\boldsymbol{v}_{m, k}=(\beta-1)(\underbrace{v_{m}, \ldots, v_{m}}_{k}, 0, \ldots, 0) \in \mathbb{R}^{q-1} .
$$

Note that for $m=0$ or $k=0$, all the $\boldsymbol{v}_{m, k}$ equal $(0, \ldots, 0)$; we will simply denote them by $\boldsymbol{v}_{0}$.

Theorem 5.4. Fix some $\beta=\beta_{n}, n \geq 2$, and a positive integer $q \geq 2$. Let $W_{0}^{\sigma}$ be the set of states as defined in Proposition 4.2. Then $W_{0}^{\sigma}=\left\{\boldsymbol{v}_{m, k}: 0 \leq k \leq\right.$ $q-1$ and $0 \leq m \leq n\}$, which has $n(q-1)+1$ elements.

Proof. We will make use of the modified iteration algorithm described at the end of last section:

$$
\boldsymbol{s}=\rho^{-1}((\boldsymbol{t}, 0)+(1-\rho) \boldsymbol{\epsilon})=\beta(\boldsymbol{t}, 0)+(\beta-1) \boldsymbol{\epsilon}, \quad \boldsymbol{t} \in W_{0}^{\sigma}, \epsilon_{i}=0 \text { or } 1 .
$$

We divide our proof into the following cases:
(i) $\boldsymbol{t}=\boldsymbol{v}_{0}=(0, \ldots, 0)$. Then $\boldsymbol{s}=(\beta-1) \boldsymbol{\epsilon} \in \mathbb{R}^{q}$. Denoting by $\boldsymbol{t}^{\prime}$ the projection of the decreasing rearrangement of $\boldsymbol{s}$ onto $\mathbb{R}^{q-1}$ by $\tau$ (i.e., $\boldsymbol{t}^{\prime}=\tau\left(\boldsymbol{s}_{\sigma}\right)$ ), we see that $\boldsymbol{t}^{\prime}$ is of the form $\boldsymbol{v}_{1, k}, 0 \leq k \leq q-1$.
(ii) $\boldsymbol{t}=\boldsymbol{v}_{m, k}$, where $1 \leq m \leq n-1$ and $1 \leq k \leq q-1$. Then

$$
\begin{equation*}
\boldsymbol{s}=(\beta-1)\left(\beta v_{m}+\epsilon_{1}, \ldots, \beta v_{m}+\epsilon_{k}, \epsilon_{k+1}, \ldots, \epsilon_{q}\right) . \tag{5.5}
\end{equation*}
$$

By Lemma 5.3,

$$
\beta v_{m}=v_{m+1}+1 \geq \frac{1}{\beta-1}
$$

and therefore $(\beta-1) \beta v_{m} \geq 1$. Note that for $\tau\left(s_{\sigma}\right)$ to belong to $W_{0}^{\sigma}$, the condition $\left|s_{i}-s_{j}\right|<1$ must be satisfied. This forces $\epsilon_{1}=\cdots=\epsilon_{k}=0$ and $\epsilon_{k+1}=\cdots=\epsilon_{q}=1$. Hence $s=(\beta-1)(\underbrace{\beta v_{m}, \ldots, \beta v_{m}}_{k}, 1, \ldots, 1)$, and by projecting it to $\mathbb{R}^{q-1}$ by $\tau$, we have $\boldsymbol{t}^{\prime}=\boldsymbol{v}_{m+1, k} \in W_{0}^{\boldsymbol{\sigma}}$.
(iii) $\boldsymbol{t}=\boldsymbol{v}_{n, k}$, where $1 \leq k \leq q-1$. Then by (5.1), $\beta v_{n}=1$ and the analogue of expression (5.5) is

$$
\boldsymbol{s}=(\beta-1)\left(1+\epsilon_{1}, \ldots, 1+\epsilon_{k}, \epsilon_{k+1}, \cdots, \epsilon_{q}\right), \quad 1 \leq k \leq q-1 .
$$

Consider the following two subcases.
Case 1. $\epsilon_{i}=1$ for some $1 \leq i \leq k$. Then for $k+1 \leq j \leq q$, the condition $\left|s_{i}-s_{j}\right|<1$ becomes $(\beta-1)\left(2-\epsilon_{j}\right)<1$. This implies that all $\epsilon_{j}=1$ for $k+1 \leq j \leq q$ and $s=(\beta-1)\left(1+\epsilon_{1}, \ldots, 1+\epsilon_{k}, 1, \ldots, 1\right)$. If $\ell(1 \leq \ell \leq k)$ of the $\epsilon_{i}$ are equal to 1 , then the corresponding $t^{\prime} \in W_{0}^{\sigma}$ is of the form

$$
\boldsymbol{t}^{\prime}=(\beta-1)(\underbrace{1, \ldots, 1}_{\ell}, 0, \ldots, 0)=\boldsymbol{v}_{1, \ell} .
$$

Case 2. $\epsilon_{i}=0$ for all $1 \leq i \leq k$. Then

$$
s=(\beta-1)\left(1, \ldots, 1, \epsilon_{k+1}, \ldots, \epsilon_{q}\right)
$$

where $\epsilon_{j}=0$ or 1 for $k+1 \leq j \leq q$. If $\ell(0 \leq \ell \leq q-k-1)$ of the $\epsilon_{j}$ are equal to 1 , then $\boldsymbol{t}^{\prime}=\boldsymbol{v}_{1, k+\ell}$. If $\epsilon_{j}=1$ for all $k+1 \leq j \leq q$, then $\boldsymbol{t}^{\prime}=\boldsymbol{v}_{0}$.

The above enumerates all the possible iterations, and hence $W_{0}^{\sigma}$ is as described. It is direct to see that there are $n(q-1)+1$ distinct $\boldsymbol{v}_{m, k}$.

We now describe the construction of the matrix $A_{0}$ based on the proof of Theorem 5.4. For an integer $q \geq 2$, we define
and let $M_{q}=M_{q}^{(0)}+M_{q}^{(1)}$. Also, we let $I_{m}$ be the $m \times m$ identity matrix and let $D_{m}$ be the $m \times m$ matrix of the form

$$
\left[\begin{array}{cccc}
0 & \ldots & 0 & 1 \\
0 & \ldots & 1 & 0 \\
\vdots & \vdots & & \vdots \\
1 & \ldots & 0 & 0
\end{array}\right]
$$

(i.e., the $i j$-entry of $D_{m}$ is 1 if $i+j=m+1$ and 0 otherwise.)

Theorem 5.5. Suppose we arrange the above $\boldsymbol{v}_{m, k}$ in the order

$$
\left\{\boldsymbol{v}_{1,1}, \ldots, \boldsymbol{v}_{1, q-1}, \boldsymbol{v}_{0}, \boldsymbol{v}_{n, q-1}, \ldots, \boldsymbol{v}_{n, 1}, \ldots, \boldsymbol{v}_{2, q-1}, \ldots, \boldsymbol{v}_{2,1}\right\}
$$

Then the matrix $A_{0}$ in Proposition 4.2 is given by

$$
A_{0}=\frac{1}{2^{q}}\left[\begin{array}{ccc}
\mathbf{0} & M_{q} & \mathbf{0}  \tag{5.6}\\
\mathbf{0} & \mathbf{0} & I_{(n-2)(q-1)} \\
D_{q-1} & \mathbf{0} & \mathbf{0}
\end{array}\right] .
$$

Proof. The submatrices $D_{q-1}$ and $I_{(n-2)(q-1)}$ correspond to case (ii) in the proof of Theorem 5.4. To see the construction of $M_{q}$ we re-examine the proofs in cases (i) and (iii). Fix a $\boldsymbol{v}_{n, k}, 1 \leq k \leq q-1$. In the first subcase of (iii), for each $1 \leq \ell \leq k$, there are $\binom{k}{l}$ of the $\epsilon_{i}(1 \leq i \leq k)$ equal to 1 , and hence $\binom{k}{l}$ of the $s$ can be rearranged and projected by $\tau$ to $\boldsymbol{t}^{\prime}=\boldsymbol{v}_{1, \ell}$. This gives rise to the corresponding columns of $M_{q}^{(0)}$. Similarly the second subcase of (iii) determines the corresponding columns of $M_{q}^{(1)}$. Lastly, it is easy to see that the first columns of $M_{q}^{(0)}$ and $M_{q}^{(1)}$ are determined by case (i). This completes the proof.

For the golden ratio $\beta_{2}$ and for $q=2,3, A_{0}$ equals respectively

$$
\frac{1}{4}\left[\begin{array}{lll}
0 & 2 & 2 \\
0 & 2 & 1 \\
1 & 0 & 0
\end{array}\right], \quad \frac{1}{8}\left[\begin{array}{lllll}
0 & 0 & 3 & 2 & 2 \\
0 & 0 & 3 & 2 & 2 \\
0 & 0 & 2 & 1 & 1 \\
0 & 1 & 0 & 0 & 0 \\
1 & 0 & 0 & 0 & 0
\end{array}\right]
$$

We remark that the algorithm to compute $\tau(q), q \geq 2$ an integer by using the matrix in Theorem 5.5 is much faster than the one used in Theorem 4.1 of [15], which requires computing the inverse of a $q \times q$ matrix with each of its entries containing the unknown defining $\tau(q)$. A complete description of $\tau(q)$ for $0<q<\infty$ is given in [15].

As a simple application of Theorem 5.5 we have
Corollary 5.6. Let $\mu_{\rho_{n}}$ be the ICBM corresponding to $\beta_{n}=\rho_{n}^{-1}$ as defined in (1.1) and let $\tau_{n}(q)$ be its $L^{q}$-spectrum. Then

$$
\lim _{n \rightarrow \infty} \tau_{n}(2)=\lim _{n \rightarrow \infty} \underline{\operatorname{dim}}_{2}\left(\mu_{\rho_{n}}\right)=1
$$

Proof. For $q=2$ the matrix $A_{n}$ in Theorem 5.5 is

$$
A_{n}=\frac{1}{4}\left[\begin{array}{cccccc}
0 & 2 & 2 & 0 & \cdots & 0 \\
0 & 2 & 1 & 0 & \cdots & 0 \\
0 & 0 & 0 & 1 & \cdots & 0 \\
\vdots & & & \vdots & \ddots & \vdots \\
0 & 0 & 0 & 0 & \cdots & 1 \\
1 & 0 & 0 & 0 & \cdots & 0
\end{array}\right]:=\frac{1}{4} B_{n}
$$

A direct calculation shows that $\operatorname{det}\left(\lambda I-B_{n}\right)=\lambda^{n}(\lambda-2)-2(\lambda-1)$. If we let $\lambda_{n}$ denote the maximal eigenvalue of $B_{n}$, then

$$
\begin{equation*}
\lambda_{n}^{n}\left(\lambda_{n}-2\right)-2\left(\lambda_{n}-1\right)=0 . \tag{5.7}
\end{equation*}
$$

Observe that $\lambda_{n}>1$ because the column sums of the irreducible matrix $B_{n}$ are at least 1 and not all equal (see e.g. [21]). Moreover, (5.7) forces $\underline{\lim }_{n \rightarrow \infty} \lambda_{n} \geq c>1$ for some constant $c$. By rewriting (5.7) as $\left(\lambda_{n}-2\right)-2\left(\lambda_{n}-1\right) / \lambda_{n}^{n}=0$, we conclude that $\lim _{n \rightarrow \infty} 2\left(\lambda_{n}-1\right) / \lambda_{n}^{n}=0$ and hence $\lim _{n \rightarrow \infty} \lambda_{n}=2$. Consequently, using Theorem 3.5 , we have

$$
\lim _{n \rightarrow \infty} \tau_{n}(2)=\lim _{n \rightarrow \infty} \frac{\ln \left(\lambda_{n} / 4\right)}{\ln \rho_{n}}=\frac{\ln (1 / 2)}{\ln (1 / 2)}=1 .
$$

The following is a list of $\underline{\operatorname{dim}}_{2}\left(\mu_{\rho_{n}}\right)$, rounded off to the 10th decimal place. Note that the smallest value of $\underline{\operatorname{dim}}_{2}\left(\mu_{\rho_{n}}\right)$ occurs at $n=3$.

| $n$ | $\beta_{n}$ | $\operatorname{dim}_{2}\left(\mu_{\rho_{n}}\right)$ |
| :---: | :---: | :---: |
| 2 | 1.6180338997 | 0.9923994336 |
| 3 | 1.8392867552 | 0.9642200274 |
| 4 | 1.9275619755 | 0.9733294764 |
| 5 | 1.9659482366 | 0.9835653645 |
| 6 | 1.9835828434 | 0.9906789642 |
| 7 | 1.9919641966 | 0.9949638696 |
| 8 | 1.9960311797 | 0.9973606068 |
| 9 | 1.9980294703 | 0.9986428460 |
| 10 | 1.9990186327 | 0.9993102630 |
|  | $\vdots$ | $\vdots$ |
| 15 | 1.9999694754 | 0.9999780091 |
|  | $\vdots$ | $\vdots$ |
| 20 | 1.9999990463 | 0.9999993121 |

## 6. A remark

All the results in the previous sections can be generalized to allow arbitrary probability weights on the contractive similitudes $\psi_{1}$ and $\psi_{2}$. More precisely, for $1 / 2<\rho<$ $1,0<a<1$ we can consider the self-similar measure $\mu_{a}$ defined by

$$
\begin{equation*}
\mu_{a}=a \mu_{a} \circ \psi_{1}^{-1}+(1-a) \mu_{a} \circ \psi_{2}^{-1} . \tag{6.1}
\end{equation*}
$$

For $\epsilon=\left(\epsilon_{1}, \ldots, \epsilon_{q}\right), \epsilon_{i}=0$ or 1 , we let $|\boldsymbol{\epsilon}|:=\sum_{i=1}^{q} \epsilon_{i}$. We modify the Markov matrix $T$ in (2.2) by

$$
T(s):=\sum_{\epsilon}^{\prime} a^{|\epsilon|}(1-a)^{q-|\epsilon|} \cdot s^{\epsilon},
$$

and define $T_{1}$ and $T_{1}^{\sigma}$ exactly the same way as in Section 2 . Then the theory goes through without change.

We conclude this section with the following proposition:
Proposition 6.1. Let $\mu_{a}$ be the self-similar measure defined as in (6.1) and let $D(a)=\underline{\operatorname{dim}}_{2}\left(\mu_{a}\right)$. Then $D(a)$ attains its maximum at $a=1 / 2$.

Proof. We will use the random variable setup of the measure $\mu_{a}$, i.e., $\mu_{a}$ is the distribution measure of the random variable $\boldsymbol{X}=\sum_{k=0}^{\infty} \rho^{k} \boldsymbol{X}_{k}$ where $\boldsymbol{X}_{k}$ takes values 1 and -1 with probabilities $a$ and $1-a$. Let $\mu_{a, k}$ be the distribution measure of $\rho^{k} \boldsymbol{X}_{k}$. A direct calculation shows that its Fourier transform is

$$
\left|\hat{\mu}_{a, k}(\xi)\right|^{2}=(2 a-1)^{2}+4 a(1-a) \cos ^{2}\left(\rho^{k} \xi\right),
$$

which is minimum when $a=1 / 2$ (as a function of $a$ ). It follows that for each $\xi$, $\left|\hat{\mu}_{a}(\xi)\right|^{2}$ is minimum when $a=1 / 2$. It is known that

$$
\varlimsup_{h \rightarrow 0^{+}} \frac{1}{h^{d+\alpha}} \int_{\mathbb{R}^{d}} \mu\left(B_{h}(x)\right)^{2} d x \approx \varlimsup_{M \rightarrow \infty} \frac{1}{M^{d-\alpha}} \int_{|\xi|<M}|\hat{\mu}(\xi)|^{2} d \xi
$$

for any bounded Borel measure $\mu$ on $\mathbb{R}^{d}$ (see [16]). (Here $\approx$ means each quantity dominates the other by a positive constant.) Using this and the definition of $\tau(2)$ given in (1.2), we conclude that $D(a)$ attains its maximum at $a=1 / 2$. In fact, the above proof shows that $D(a)$ is symmetric about $a=1 / 2$ and is increasing from $a=0$ to $a=1 / 2$.

Acknowledgment. The authors would like to thank the referee for some valuable comments and suggestions.

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[^0]:    ** Research supported in part by a postdoctoral fellowship from the Chinese University of Hong Kong.

