

MULTIFRACTAL ANALYSIS FOR BEDFORD-MCMULLEN CARPETS

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ABSTRACT. In this paper we compute the multifractal analysis for local dimensions of Bernoulli measures supported on the self-affine carpets introduced by Bedford-McMullen. This extends the work of King where the multifractal analysis is computed with strong additional separation assumptions.

1. INTRODUCTION

The multifractal properties of local dimensions of fractal measures have been studied for more than twenty years. Some problems are already completely solved (for example, the local dimension spectra for Gibbs measures on a conformal repeller, see [9]). However non-conformal systems turn out to be much less tractable and only some specific examples have been solved.

By the local dimension spectrum of a measure we mean the function $\alpha \rightarrow \dim_{\mathcal{H}} X_{\alpha}$ where X_{α} is the set of all points with local dimension α . There exists a well-developed technique for calculating the local dimension spectra of invariant measures for dynamical systems. One begins by introducing a symbolic description on the attractor and defining a suitable *symbolic* local dimension of the measure. The first step is usually first to compute the multifractal spectrum for the symbolic local dimension. This is done by constructing a suitable auxiliary measure which is exact-dimensional and only supported on the set of points where the symbolic local dimension of the original measure takes a prescribed value, say α . If the auxiliary measure has been correctly chosen then the Hausdorff dimension of points with symbolic local dimension α will be the dimension of the auxiliary measure. The final step is to show that the multifractal spectrum for the symbolic local dimension of the original measure is the same as the spectrum for the

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real local dimension. Once again, we refer the reader to [9], where this technique is explained in detail.

In this paper we are interested in the local dimension spectra for Bernoulli measures on Bedford-McMullen carpets. This problem has already been studied in several papers. King calculated the symbolic local dimension spectrum in [5], this result was then generalized to Gibbs measures by Barral and Mensi [2] and to higher dimensional generalized Sierpiński carpets by Olsen [7] (see also [8]). However, in all those papers authors were unable to make the last step, from symbolic to real local dimension. For that reason, these papers had to assume some strong additional assumptions (e.g. a very strong separation property) guaranteeing the equality of symbolic and real local dimension spectra. However in [2] it is shown the symbolic and real local dimension spectra are the same for the decreasing part of the spectrum. Naturally, it was conjectured (for example by Olsen in [8]) that these additional assumptions are not necessary for the increasing part of the spectra also to be the same.

The purpose of this paper is to present this missing argument. Unfortunately our arguments only apply to the two-dimensional case. For simplicity we just look at the Bernoulli measures studied in [5].

2. STATEMENT OF RESULTS

We now proceed to formally state our results. To define the Bedford-McMullen carpets [3, 6] we introduce a digit set

$$D \subseteq \{0, \dots, m-1\} \times \{0, \dots, n-1\},$$

where $m < n$. For each $(i, j) \in D$ we define $T_{i,j} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ by $T_{i,j}(x, y) = (n^{-1}(x+j), m^{-1}(y+i))$. We let Λ be the unique non-empty compact set which satisfies $\cup_{(i,j) \in D} T_{i,j}(\Lambda) = \Lambda$. We will also let $\sigma = \frac{\log m}{\log n}$. The Hausdorff and box counting dimension of Λ were calculated by McMullen and Bedford in [6] and [3]. We introduce a positive probability vector \underline{p} with element p_{ij} for each $(i, j) \in D$. We also define the related probability vector \underline{q} where $q_i = \sum_{j:(i,j) \in D} p_{ij}$. Thus we can define a self-affine measure μ which is the unique probability measure satisfying

$$\mu(A) = \sum_{(i,j) \in D} p_{ij} \mu(T_{i,j}^{-1}A)$$

for any Borel subset of \mathbb{R}^2 . For any $x \in \mathbb{R}^2$ the local dimension of μ at x is defined by

$$d_\mu(x) = \lim_{r \rightarrow 0} \frac{\log \mu(B(x, r))}{\log r}$$

if this limit exists. For $\alpha \in \mathbb{R}$ the level sets X_α are defined by

$$X_\alpha = \{x \in \Lambda : d_\mu(x) = \alpha\}.$$

In King [5] the function $f(\alpha) = \dim_{\mathcal{H}} X_\alpha$ was calculated under the condition that for any pair $\{(i, j), (i', j')\} \in D$ we have that $|i - i'| \neq 1$ and if $i = i'$ then $|j - j'| > 1$. In [2] this assumption is weakened: they are also able to allow digit sets D where $D \cap \{(0, 0), \dots, (0, n-1)\} = \emptyset$ or $D \cap \{(m-1, 0), \dots, (m-1, n-1)\} = \emptyset$. Furthermore arbitrary digit sets can be considered as long as the probability vector p is chosen so that $\sum_{j:(1,j) \in D} p_{1j}^t = \sum_{j:(m,j) \in D} p_{mj}^t$ for all $t > 0$. We will only assume that the digit set has elements in more than one row and more than one column, without this assumption the measure μ is effectively a self-similar measure on the line and the singularity spectrum is computed in [1].

The formula we obtain for $\dim_{\mathcal{H}} X_\alpha$ is exactly the formula obtained in [5]. To recall the definition we fix $t > 0$ and let $\gamma_i = \sum_{j:(i,j) \in D} p_{ij}^t$. We then define $\beta(t)$ to be the unique solution to

$$m^{\beta(t)} \sum_{(i,j) \in D} p_{ij}^t q_i^{(1-\sigma)t} \gamma_i^{\sigma-1} = 1.$$

We will let

$$\alpha_{\min} = \min_{(i,j) \in D} \frac{-\sigma \log p_{ij} + (\sigma - 1) \log q_i}{\log m}$$

and

$$\alpha_{\max} = \max_{(i,j) \in D} \frac{-\sigma \log p_{ij} + (\sigma - 1) \log q_i}{\log m}.$$

Our main result is that

Theorem 1. *For any $\alpha \in (\alpha_{\min}, \alpha_{\max})$ we have that*

$$f(\alpha) = \dim_{\mathcal{H}} X_\alpha = \inf_t (\alpha t + \beta(t)).$$

In other words f is the Legendre transform of β . Furthermore f is differentiable with respect to α and is concave.

We are not going to rewrite all of the King's paper [5], so we will frequently make use of his partial results, referring the reader to his paper. In particular, the properties of $f(\alpha)$ can be found in section 4 of [5].

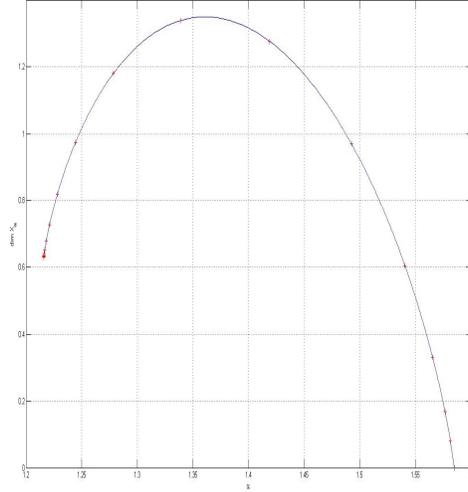


FIGURE 1. The graph of $\alpha \rightarrow \dim_{\mathcal{H}} X_{\alpha}$ where $m = 2$, $n = 3$, $D = \{(0, 0), (2, 0), (1, 1)\}$ and $\underline{p} = (\frac{1}{3}, \frac{1}{3}, \frac{1}{3})$. The endpoints of the graph are $(\frac{\log 3}{\log 2} + \frac{\log 2}{\log 3} - 1, \frac{\log 2}{\log 3})$ and $(\frac{\log 3}{\log 2}, 0)$.

The rest of paper is divided as follows. In Section 3 we will show how to obtain the lower bound for arbitrarily digit sets. The main argument, calculating of the upper bound, is presented in Section 4.

3. SYMBOLIC CODING AND THE LOWER BOUND

The lower bound we need was obtained by Barral and Mensi in [2] however for completeness we give a proof here. The lower bound can be proved following the method of King, [5], with the addition of just one simple lemma. First of all we need to introduce of a natural symbolic coding. If we let $\Sigma = D^{\mathbb{N}}$ it is possible to define a natural projection $\Pi : \Sigma \rightarrow \Lambda$. Initially we let $\pi : D \rightarrow \{0, \dots, m - 1\}$ be defined by $\pi((i, j)) = i$ and $\bar{\pi} : D \rightarrow \{0, \dots, n - 1\}$ by $\bar{\pi}(i, j) = j$. This allows us to define Π by

$$\Pi(\underline{i}) = \sum_{j=0}^{\infty} (\bar{\pi}(i_j)n^{-j}, \pi(i_j)m^{-j}),$$

where i_j is the j -th element of the sequence \underline{i} .

It is usual in the study of self-similar sets or conformal systems to study cylinder sets. In this self-affine setting there is a related idea of

approximate squares. To construct them we let $l(k) = [\sigma k]$ (where $[\cdot]$ denotes the integer part) and define for any $\underline{i} \in \Sigma$

$$R_k(\underline{i}) = \left[\sum_{j=1}^{l(k)} \bar{\pi}(i_j) n^{-j}, \sum_{j=1}^{l(k)} \bar{\pi}(i_j) n^{-j} + n^{-l(k)} \right] \times \left[\sum_{j=1}^k \pi(i_j) m^{-j}, \sum_{j=1}^k \pi(i_j) m^{-j} + m^{-k} \right].$$

The shorter side of $R_k(\underline{i})$ is always of length m^{-k} . As the ratio between the sides of this rectangle is clearly between $1 : 1$ and $1 : n$, we can let $D_1 = \sqrt{n^2 + 1}$ and note that for all $\underline{i} \in \Sigma$ we have that $|R_k(\underline{i})| \leq D_1 m^{-k}$. An approximate square $R_k(\underline{i})$ can contain elements $x = \Pi(\underline{j})$ where $(j_{l+1}, \dots, j_k) \neq (i_{l+1}, \dots, i_k)$. To help keep track of these elements we denote the set of their possible initial segments of symbolic expansions:

$$\Gamma_k(\underline{i}) = \{(j_1, \dots, j_k) : (i_1, \dots, i_l) = (j_1, \dots, j_l) \text{ and } \pi(j_{l+1}) = \pi(i_{l+1}), \dots, \pi(j_k) = \pi(i_k)\}.$$

For each $\underline{i} \in \Sigma$ we will define

$$\delta_\mu(\underline{i}) = \lim_{k \rightarrow \infty} - \frac{\log \mu(R_k(\underline{i}))}{k \log m}$$

if this limit exists. Under the separation conditions imposed in [5] if $\Pi \underline{i} = x$ then $\delta_\mu(\underline{i}) = d_\mu(x)$ as long as one of these two limits exists. However without these assumptions this is not always the case. The following lemma shows that in terms of calculating the lower bound this poses no problem. We fix $t \in R$ and let ν_t be the Bernoulli measure defined by the probability vector $\{P_{ij}\}_{(i,j) \in D}$ where for $(i, j) \in D$

$$P_{ij} = p_{ij}^t m^{\beta(t)} q_i^{(1-\sigma)t} \gamma_i^{\sigma-1}.$$

We also let $Q_i = \sum_{j:(i,j) \in D} P_{ij}$ and

$$\alpha(t) = \frac{-\sigma \sum_{(i,j) \in D} P_{ij} \log p_{ij} - (1-\sigma) \sum_{i=0}^{m-1} Q_i \log q_i}{\log m}.$$

Lemma 1. *For μ_t almost all \underline{i} we have that*

$$\delta_\mu(\underline{i}) = d_\mu(\Pi \underline{i}) = \alpha(t)$$

Proof. The fact that $\delta_\mu(\underline{i}) = \alpha(t)$ for μ_t almost all \underline{i} follows from Lemma 4 of [5] (in the formula given for $\alpha(t)$ in lemma 5 in [5] $-\sum_{(i,j) \in D} P_{ij}$ should read $-\sum_{(i,j) \in D} P_{ij} \log P_{ij}$). Thus we only need to show that $\delta_\mu(\underline{i}) = d_\mu(\Pi \underline{i})$ for μ_t almost all \underline{i} . We let

$$Z_k(\underline{i}) = \max \left\{ \min \{ \eta : \pi(i_{k+\eta}) \neq \pi(i_{k+1}) \}, \sigma^{-1} \min \{ \eta : \bar{\pi}(i_{l(k)+\eta}) \neq \bar{\pi}(i_{l(k)}) \} \right\}$$

and note that by the Borel-Cantelli Lemma

$$\limsup_{k \rightarrow \infty} \frac{Z_k(\underline{i})}{k} = 0$$

for μ_t almost all \underline{i} (here we use the assumption that the digit set has elements in two distinct rows and two distinct columns). We now fix $\underline{i} \in \Sigma$ such that $\limsup_{k \rightarrow \infty} \frac{Z_k(\underline{i})}{k} = 0$ and let $x = \Pi(\underline{i})$. We have that for any k

$$\mu(B(x, D_1 m^{-k})) \geq \mu(R_k(\underline{i})) \geq \mu(B(x, m^{-k-Z_k(\underline{i})})).$$

The result now follows by considering k sufficiently large. \square

We can now conclude that

$$\dim_{\mathcal{H}}(X_{\alpha(t)}) \geq \dim_{\mathcal{H}} \mu_t \circ \Pi^{-1}.$$

However in [5] it is shown in Lemma 5 that $\alpha(t) = -\beta'(t)$ and that $\dim_{\mathcal{H}} \mu_t \circ \Pi^{-1} = t\alpha(t) + \beta(t)$. Technically this result in King is proved with additional assumptions about D however the formula for the dimension of a Bernoulli measure is still valid without these assumption (see for example [3] or [2]). We can now deduce that

$$\dim_{\mathcal{H}}(X_{\alpha(t)}) \geq t\alpha(t) + \beta(t)$$

and the proof of the lower bound is complete.

4. UPPER BOUND

In this section we will find efficient coverings of X_{α} by considering appropriate sets of approximate squares. In particular we want to show that $\dim_{\mathcal{H}} X_{\alpha} \leq \alpha t + \beta(t)$ for any $t \in \mathbb{R}$. Combining this with the lower bound completes the proof of Theorem 1. We will fix $t \in \mathbb{R}$ for the rest of this section. For any $\alpha \in (\alpha_{\min}, \alpha_{\max})$ and $\epsilon > 0$ we will denote

$$Y(\alpha, \epsilon, k) = \{R_k(\underline{i}) : m^{-k\alpha(1+\epsilon)} \leq \mu(R_k) \leq m^{-k\alpha(1-\epsilon)}\}.$$

For a sequence $\underline{i} \in \Sigma$ we will define

$$V_k(\underline{i}) := \inf \{l > k + 1 : \pi(i_l) \notin \{0, m-1\} \text{ or } \pi(i_l) \neq \pi(i_{k+1})\} - k - 2.$$

This function will be used to bound the distance between x and the horizontal boundary of $R_k(\underline{i})$.

Lemma 2. *For any $x \in X_{\alpha}$ and $\underline{j} \in \Sigma$ where $\Pi \underline{j} = x$ there exists $K \in \mathbb{N}$ such that for any $k \geq K$ there exists $\underline{i} \in \Sigma$ such that:*

- (1) $d(R_k(\underline{i}), x) \leq \frac{D_1 m^{-k}}{2}$,
- (2) $R_k(\underline{i}) \in Y(\alpha, \epsilon, k)$,
- (3) If $V_k(\underline{j}) \leq \frac{\epsilon k}{2}$ then $(\pi(i_1), \dots, \pi(i_k)) = (\pi(j_1), \dots, \pi(j_k))$.

Proof. Fix $\epsilon > 0$. If $x \in X_\alpha$ then we can find $R > 0$ such that if $r < R$ we have that $\frac{\log \mu(B(x,r))}{\log r} \in [\alpha(1 - \epsilon)/3, \alpha(1 + \epsilon)/3]$. We choose k such that $D_1 m^{-k} \leq \frac{R}{2}$ and let $\underline{j} \in \Sigma$ satisfy $d(R_k(\underline{j}), x) \leq \frac{D_1 m^{-k}}{2}$. It then follows that $\mu(R_k(\underline{j})) \leq (D_1 m^{-k})^{\alpha(1-\epsilon/3)}$. Furthermore there are at most 9 sequences \underline{j}^k where $d(R_k(\underline{j}), x) \leq m^{-k}$ which means that one of these sequences \underline{j}^k must satisfy $\mu(R_k(\underline{j})) \geq \frac{m^{-k\alpha(1+\epsilon/3)}}{9}$. Parts 1 and 2 of the assertion now follow easily.

For the part 3 we fix k such that $V_k(\underline{j}) \leq \frac{\epsilon k}{2}$ and note that for any sequence \underline{i} where $(\pi(i_1), \dots, \pi(i_k)) \neq (\pi(j_1), \dots, \pi(j_k))$ we have that

$$d(x, R_k(\underline{i})) \geq m^{-k-V_k(\underline{j})} \geq m^{-k(1+\epsilon/2)}.$$

This means that any $\underline{i} \in \Sigma$ which satisfies $B(x, m^{-k(1+\epsilon/2)}) \cap R_k(\underline{i}) \neq \emptyset$ must also satisfy $(\pi(i_1), \dots, \pi(i_k)) = (\pi(j_1), \dots, \pi(j_k))$. Hence the ball $B(x, m^{-k(1+\epsilon/2)})$ satisfying

$$\mu(B(x, m^{-k(1+\epsilon/2)})) \geq m^{-k\alpha(1+\epsilon/2)(1+\epsilon/3)} \geq 2m^{-k\alpha(1+\epsilon)}$$

must be contained in the union of two approximate squares $R_k(\underline{i})$ which both satisfy $(\pi(i_1), \dots, \pi(i_k)) = (\pi(j_1), \dots, \pi(j_k))$. This implies that one of these approximate squares has measure not smaller than $m^{-k\alpha(1+\epsilon)}$. This together with the upper bound proved for the measure in part (2) means that one of these approximate squares is contained in $Y(\alpha, \epsilon, k)$. This completes the proof. \square

It should be noted that while this lemma indicates that

$$X_\alpha \subset \bigcup_{R_k(\underline{i}) \in Y(\alpha, \epsilon, k)} B(R_k(\underline{i}), D_1 m^{-k}/2),$$

this would not provide an efficient cover in terms of Hausdorff dimension. To get the efficient cover we define a function $\omega : D \rightarrow \mathbb{R}$ by

$$\omega(i, j) = q_i^t \gamma_i^{-1}.$$

It is important that ω only depends on the vertical coordinates. For $\underline{i} \in \Sigma$ we will denote

$$B_l(\underline{i}) = \frac{1}{l} \sum_{r=1}^l \log \omega_{i_r}$$

and

$$A_k(\underline{i}) = B_{l(k)}(\underline{i}) - B_k(\underline{i}).$$

A_k is essentially the logarithm of the function f_k defined on page 6 of [5]. The upper bound in [5] uses covering of approximate squares,

$R_k(\underline{i})$, where $A_k(\underline{i}) > -\epsilon$ for some small ϵ . More precisely for any $\epsilon > 0$ we let

$$G(\alpha, \epsilon, k) = Y(\alpha, \epsilon, k) \cap \{R_k(\underline{i}) : A_k(\underline{i}) \geq -\log(1 + \epsilon)\}$$

and note that in [5] it is shown that for any $\underline{i} \in \Sigma$ where $\delta_\mu(\underline{i}) = \alpha$, there exist infinitely many k such that $R_k(\underline{i}) \in G(\alpha, \epsilon, k)$. We are going to show that these covers can be modified so that the result holds for any $x \in X_\alpha$ even if $\underline{i} \in \Sigma$ with $\Pi(\underline{i}) = x$ does not satisfy $\delta_\mu(\underline{i}) = \alpha$. To be able to do this we need the following proposition:

Proposition 1. *For any $\epsilon > 0$ and $x \in X_\alpha$ there exist infinitely many $k \in \mathbb{N}$ for which there is a sequence $\underline{i} \in \Sigma$ such that*

1. $d(R_k(\underline{i}), x) \leq D_1 m^{-k}$,
2. $A_k(\underline{i}) \geq -\epsilon$,
3. $R_k(\underline{i}) \in Y(\alpha, \epsilon, k)$.

Before we prove Proposition 1, we will show how this proposition and the following simple lemma imply the upper bound for $\dim_{\mathcal{H}} X_\alpha$.

Lemma 3. *If $R_k(\underline{i}) \in G(\alpha, \epsilon, k)$ and k is sufficiently large then*

$$\mu(R_k(\underline{i}))^t \leq (1 + \epsilon)^k \sum_{(j_1, \dots, j_k) \in \Gamma_k(\underline{i})} (p_{j_1} \cdots p_{j_k})^t (\omega_{j_1} \cdots \omega_{j_k})^{1-\sigma}.$$

Proof. If we fix $k \in \mathbb{N}$, \underline{i} and write $l = l(k)$ then

$$\mu(R_k(\underline{i}))^t = \sum_{(j_1, \dots, j_k) \in \Gamma_k(\underline{i})} (p_{j_1} \cdots p_{j_k})^t \omega_{j_{l+1}} \cdots \omega_{j_k}.$$

However for any $(j_1, \dots, j_k) \in \Gamma_k(\underline{i})$ we have that $\omega_{j_{l+1}} \cdots \omega_{j_k} = \omega_{i_{l+1}} \cdots \omega_{i_k}$. Thus since $A_k(\underline{i}) \geq -\log(1 + \epsilon)$ we have that for $(j_1, \dots, j_k) \in \Gamma_k(\underline{i})$

$$\frac{\omega_{j_{l+1}} \cdots \omega_{j_k}}{(\omega_{j_1} \cdots \omega_{j_k})^{1-\sigma}} = \frac{(\omega_{i_1} \cdots \omega_{i_k})^\sigma}{\omega_{i_1} \cdots \omega_{i_l}} \leq e^{\sigma k B_k(\underline{i}) - l B_l(\underline{i})} \leq e^{-B_k(\underline{i})} (1 + \epsilon)^l.$$

To complete the proof simply note that $e^{-B_k(\underline{i})}$ is uniformly bounded by some constant C and thus it is enough to choose k large enough so that $(1 + \epsilon)^{k-l} \geq C$. \square

Proposition 1 shows that for any $K \in \mathbb{N}$ and $\epsilon > 0$ we have

$$X_\alpha \subset \bigcup_{k > K} \bigcup_{R_k(\underline{i}) \in G(\alpha, \epsilon, k)} \widehat{R}_k(\underline{i}),$$

where $\widehat{R}_k(\underline{i})$ stands for a rectangle with the same center as $R_k(\underline{i})$ but $2D_1$ times greater.

We now choose $\epsilon > 0$. As $m \geq 2$, we have $\log(1 + \epsilon) < 2\epsilon \log m$. For any $\delta > \epsilon(\alpha|t| + 2)$ and $K \in \mathbb{N}$ we have that by using the definition of $G(\alpha, \epsilon, k)$, using Lemma 3 and applying the multinomial theorem

$$\begin{aligned}
 & \sum_{k \geq K} \sum_{R_k(\underline{i}) \in G(\alpha, \epsilon, k)} |\widehat{R}_k(\underline{i})|^{\alpha t + \beta(t) + \delta} \\
 & \leq (2D_1)^{\alpha t + \beta(t) + \delta} \sum_{k \geq K} \sum_{R_k(\underline{i}) \in G(\alpha, \epsilon, k)} |R_k(\underline{i})|^{\beta(t) + 2\epsilon} \mu(R_k(\underline{i}))^t \\
 & \leq D_2 \sum_{k \geq K} \sum_{R_k(\underline{i}) \in G(\alpha, \epsilon, k)} m^{-k(\beta(t) + 2\epsilon)} (1 + \epsilon)^k \sum_{(j_1, \dots, j_k) \in \Gamma_k(\underline{i})} (p_{j_1} \cdots p_{j_k})^t (\omega_{j_1} \cdots \omega_{j_k})^{1-\sigma} \\
 & \leq D_2 \sum_{k \geq K} m^{-2\epsilon k} (1 + \epsilon)^k \sum_{(i_1, \dots, i_k) \in D^k} m^{-k\beta(t)} (p_{i_1} \cdots p_{i_k})^t (\omega_{i_1} \cdots \omega_{i_k})^{1-\sigma} \\
 & = D_2 \sum_{k \geq K} m^{-2\epsilon k} (1 + \epsilon)^k \left(\sum_{(i, j) \in D} m^{-\beta(t)} p_{ij}^t \omega_{ij}^{1-\sigma} \right)^k \\
 & = D_2 \sum_{k \geq K} m^{-2\epsilon k} (1 + \epsilon)^k < \infty
 \end{aligned}$$

(where $D_2 = 2^{\alpha t + \beta(t) + \delta} D_1^{\alpha t + 2\beta(t) + \delta + 2\epsilon}$). It follows immediately that

$$\dim_{\mathcal{H}} X_\alpha \leq t\alpha + \beta(t) + \delta$$

and δ can be arbitrarily small.

We now proceed to prove Proposition 1.

Proof of Proposition 1. We start with the case where $\underline{i} \in \Sigma$, $\Pi \underline{i} \in X_\alpha$ and for some k , $V_k(\underline{i}) = \infty$. We assume, without loss of generality, that $\pi(i_m) = 0$ for all $m > k$ and that $\pi(i_k) \neq 0$ and let $\epsilon > 0$. If we fix $\eta \in \mathbb{N}$ and consider $k + \eta$ level approximate squares $R_{k+\eta}(j)$ which have points within $m^{-k-\eta}/2$ of x then j must satisfy $\pi(j_u) = \pi(i_u)$ for all $u \leq k + \eta$ or $\pi(j_u) = m - 1$ for all $k < u \leq k + \eta$. In both of these cases if η is sufficiently large then $A_{k+\eta}(j) \geq -\epsilon$ ($B_{k+\eta}$ is a converging sequence at such points). It follows from parts 1 and 2 of Lemma 2 that if η is sufficiently large one of these sequences j must satisfy that $R_{k+\eta}(j) \in Y(\alpha, \epsilon, k + \eta)$.

We now turn to the case where $V_k(\underline{i}) < \infty$. In this case Proposition 1 will follow from the following lemma.

Lemma 4. *For any $\epsilon > 0$ if $\underline{i} \in \Sigma$ and $V_u(\underline{i}) < \infty$ for all u then we can find infinitely many $k \in \mathbb{N}$ such that*

- (1) $A_k(\underline{i}) > -\epsilon$,
- (2) $V_k(\underline{i}) = 0$.

Proof. Let $\epsilon > 0$ and $\underline{i} \in \Sigma$ such that $V_k(\underline{i}) < \infty$ for all \underline{i} . It is possible to find bounds $C_1, C_2 \in \mathbb{R}$ such that $C_1 \leq B_k(\underline{i}) \leq C_2$. We now prove the assertion by contradiction. We assume that \underline{i} does not satisfy the assertion and for each $k \in \mathbb{N}$ let $\sigma_k = \frac{[\sigma(k+V_k(\underline{i}))]}{k+V_k(\underline{i})}$ and note that $\lim_{k \rightarrow \infty} \sigma_k = \sigma$. We now fix $K \in \mathbb{N}$ such that for all $k \geq K$ we have that $A_k(\underline{i}) \leq -\epsilon$ or $V_k(\underline{i}) > 0$ and that $\frac{1-\sigma_k}{\sigma_k} \leq \frac{2-\sigma}{\sigma}$.

Firstly for $k \geq K$ let us consider how large $V_k(\underline{i})$ can be. If $V_k(\underline{i})/k \leq (1-\sigma_k)/\sigma_k$ then we already have an upper bound on $V_k(\underline{i})$, so we will assume that $V_k(\underline{i})/k \geq (1-\sigma_k)/\sigma_k$. We then have that $k < [\sigma(k+V_k(\underline{i}))]$ and thus we have that if $\eta \in \mathbb{N}$ satisfies $[\sigma(k+V_k(\underline{i}))] \leq \eta \leq k+V_k(\underline{i})$ then $\omega_{i_\eta} = \omega_{i_{k+1}}$. By definition, $B_l(\underline{i})$ is the sequence of Birkhoff averages for the locally constant function $\underline{\tau} \rightarrow \log \omega_{\tau_1}$. Thus we can calculate

$$\begin{aligned}
A_{k+V_k(\underline{i})}(\underline{i}) &= B_{[\sigma(k+V_k(\underline{i}))]}(\underline{i}) - B_{k+V_k(\underline{i})}(\underline{i}) \\
&= \frac{kB_k(\underline{i}) + ([\sigma(k+V_k(\underline{i}))] - k) \log \omega_{i_{k+1}}}{[\sigma(k+V_k(\underline{i}))]} - \frac{kB_k(\underline{i}) + V_k(\underline{i}) \log \omega_{i_{k+1}}}{k+V_k(\underline{i})} \\
&= \frac{kB_k(\underline{i}) + (\sigma_k(k+V_k(\underline{i})) - k) \log \omega_{i_{k+1}} - \sigma_k(kB_k(\underline{i}) + V_k(\underline{i}) \log \omega_{i_{k+1}})}{\sigma_k(k+V_k(\underline{i}))} \\
&= \frac{k(1-\sigma_k)B_k(\underline{i}) + k(\sigma_k - 1) \log \omega_{i_{k+1}}}{\sigma_k(k+V_k(\underline{i}))} \\
&= \frac{(\sigma_k - 1)k}{\sigma_k(k+V_k(\underline{i}))} (B_k(\underline{i}) - \log \omega_{i_{k+1}}).
\end{aligned}$$

As $B_k(\underline{i}) - \log \omega_{i_{k+1}} \geq C_1 - C_2$ and the left hand side is at most $-\epsilon$ (because $V_{k+V_k(\underline{i})}(\underline{i}) = 0$), we can estimate

$$\frac{1}{k}V_k(\underline{i}) \leq \frac{1-\sigma_k}{\sigma_k \epsilon} (C_2 - C_1).$$

Thus if we let

$$V = \max \left\{ \frac{2-\sigma}{\sigma \epsilon} (C_2 - C_1), \frac{2-\sigma}{\sigma} \right\}$$

then

$$\frac{1}{k}V_k(\underline{i}) \leq V$$

for all $k \geq K$. In particular, there exists $K' \geq K$ such that if $k+V_k(\underline{i}) \geq (V+1)K'$ then $V_k(\underline{i}) \leq Vk$.

We continue with the proof. We choose $u \in \mathbb{N}$ such that $u\epsilon > C_2 - C_1$. Denote

$$(1) \quad W = (V+1)\sigma^{-1}.$$

We can choose $n_0 > K'W^{u+1}$ such that $V_{n_0}(\underline{i}) = 0$ (and hence $A_{n_0}(\underline{i}) \leq -\epsilon$). We will now define a sequence n_j inductively. We assume that $A_{n_j}(\underline{i}) \leq -\epsilon$ and follow the following inductive procedure.

- i) If $n_j \leq K'W$, we stop the construction.
- ii) If $A_{[\sigma n_j]} \leq -\epsilon$ then let $n_{j+1} = [\sigma n_j]$ and note that $B_{n_{j+1}} \leq B_{n_j} - \epsilon$.
- iii) If $A_{[\sigma n_j]} > -\epsilon$ then since $n_j \geq K'W > K$ we know that $V_{[\sigma n_j]}(\underline{i}) > 0$. We let

$$a = \min\{\eta \geq [\sigma n_j] : V_\eta(\underline{i}) = 0\} \text{ and } b = \max\{\eta \leq [\sigma n_j] : V_\eta(\underline{i}) = 0\}.$$

Now choose

$$n_{j+1} = \begin{cases} a & \text{if } B_a(\underline{i}) \leq B_b(\underline{i}) \\ b & \text{if } B_b(\underline{i}) < B_a(\underline{i}) \end{cases}.$$

Since we have that ω_{i_η} is constant for $a < \eta \leq b$ it follows that $B_\eta(\underline{i})$ is monotonic for $a \leq \eta \leq b$. Hence at either a or b the value of B is going to be not greater than $B_{[\sigma n_j]}(\underline{i})$. Thus it follows that $B_{n_{j+1}}(\underline{i}) \leq B_{n_j}(\underline{i}) - \epsilon$.

In the construction, n_{j+1} can either be equal to $[\sigma n_j]$ (in case ii) or be between $[\sigma n_j]/(V+1)$ and $(V+1)[\sigma n_j]$ (in case iii). This, together with (1), gives us

$$n_{j+1} \geq n_j W^{-1}$$

as long as $n_j > K'W$. As $n_0 > K'W^{u+1}$, the construction will go on for at least u steps. Moreover, we have

$$B_{n_j}(\underline{i}) \leq B_{n_{j-1}}(\underline{i}) - \epsilon$$

for each j . We get

$$\begin{aligned} B_{n_u}(\underline{i}) &\leq B_{n_0}(\underline{i}) - u\epsilon \\ &\leq C_2 - u\epsilon < C_1. \end{aligned}$$

This contradicts the definition of C_1 . □

We now fix $x \in X_\alpha$ and let $\underline{i} \in \Sigma$ satisfy $\Pi \underline{i} = x$. The proof of Proposition 1 is completed by combining Lemma 4 with part 3 of Lemma 2 (recall that A only depends on the vertical coordinates).

REFERENCES

- [1] M. Arbeiter and N. Patzschke, Random self-similar multifractals, *Math. Nachr.*, 181 (1996), 5–42.
- [2] J. Barral and M. Mensi, Gibbs measures on self-affine Sierpiński carpets and their singularity spectrum. *Ergodic Theory Dynam. Systems* 27 (2007), no. 5, 1419–1443.
- [3] T. Bedford, Crinkly curves, Markov partitions and box dimensions in self-similar sets, PhD Thesis, The University of Warwick, 1984.
- [4] R. Cawley and R.D. Mauldin, Multifractal decompositions of Moran fractals., *Adv. Math.* 92 (1992), no. 2, 196–236.
- [5] J. King, The singularity spectrum for general Sierpiński carpets, *Adv. Math.* 116 (1995), 1–8.
- [6] C. McMullen, The Hausdorff dimension of general Sierpiński carpets, *Nagoya Math. J.* 96 (1984), 1–9.
- [7] L. Olsen, Self-affine multifractal Sierpiński sponges in \mathbf{R}^d , *Pacific J. Math.* 183 (1998), no. 1, 143–199.
- [8] L. Olsen, Symbolic and geometric local dimensions of self-affine multifractal Sierpinski sponges in \mathbb{R}^d , *Stoch. Dyn.* 7 (2007), no. 1, 37–51.
- [9] Y. Pesin, *Dimension Theory in Dynamical Systems: Contemporary Views and Applications*, University of Chicago Press, Chicago 1998.

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