

On the dimension of stationary measures for random piecewise affine interval homeomorphisms

KRZYSZTOF BARAŃSKI[†] and ADAM ŚPIEWAK^{‡§}

[†] *Institute of Mathematics, University of Warsaw, ul. Banacha 2,
02-097 Warszawa, Poland
(e-mail: baranski@mimuw.edu.pl)*

[‡] *Department of Mathematics, Bar-Ilan University, Ramat-Gan 5290002, Israel*

[§] *Institute of Mathematics, Polish Academy of Sciences, ul. Śniadeckich 8,
00-656 Warszawa, Poland
(e-mail: ad.spiewak@gmail.com)*

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Abstract. We study stationary measures for iterated function systems (considered as random dynamical systems) consisting of two piecewise affine interval homeomorphisms, called Alsedà–Misiurewicz (AM) systems. We prove that for an open set of parameters, the unique non-atomic stationary measure for an AM system has Hausdorff dimension strictly smaller than 1. In particular, we obtain singularity of these measures, answering partially a question of Alsedà and Misiurewicz [Random interval homeomorphisms. *Publ. Mat.* **58**(suppl.) (2014), 15–36].

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1. Introduction

In recent years, a growing interest in low-dimensional random dynamics has led to an intensive study of random one-dimensional systems given by (semi)groups of interval and circle homeomorphisms, both from the stochastic and geometric points of view (see e.g. [1, 7, 8, 10–12, 16–18, 23, 24]). This can be seen as an extension of the research on the well-known case of groups of smooth circle diffeomorphisms (see e.g. [13, 19]).

Let f_1, \dots, f_m , $m \geq 2$, be homeomorphisms of a 1-dimensional compact manifold X (a closed interval or a circle). The transformations f_i generate a semigroup consisting of iterates $f_{i_n} \circ \dots \circ f_{i_1}$, where $i_1, \dots, i_n \in \{1, \dots, m\}$, $n \in \{0, 1, 2, \dots\}$. For a probability vector (p_1, \dots, p_m) , such a system defines a Markov process on X which, by the

Krylov–Bogolyubov theorem, admits a (non-necessarily unique) *stationary measure*, i.e. a Borel probability measure μ on X satisfying

$$\mu(A) = \sum_{i=1}^m p_i \mu(f_i^{-1}(A))$$

for every Borel set $A \subset X$. In many cases, it can be shown that the stationary measure is unique (at least within some class of measures) and is either absolutely continuous or singular with respect to the Lebesgue measure. It is usually a non-trivial problem to determine which of the two cases occurs (see e.g. [20, §7]), and the question has been solved only in some particular cases.

This paper is a continuation of the research started in [2] on singular stationary measures for so-called Alsedà–Misiurewicz systems (AM systems), defined in [1]. These are random systems generated by two piecewise affine increasing homeomorphisms f_-, f_+ of the unit interval $[0, 1]$, such that $f_i(0) = 0, f_i(1) = 1$ for $i = -, +$, each f_i has exactly one point of non-differentiability $x_i \in (0, 1)$ and $f_-(x) < x < f_+(x)$ for $x \in (0, 1)$. For a detailed description of AM systems, refer to [2]. The dynamics of AM systems and related ones has already gained some interest in recent years, being studied in e.g. [1–6, 25]. Within the class of uniformly contracting iterated function systems, piecewise linear maps and the dimension of their attractors were recently studied in [22].

In this paper, as explained below, we study stationary measures for symmetric AM systems with positive endpoint Lyapunov exponents.

Definition 1.1. A symmetric AM system is the system $\{f_-, f_+\}$ of increasing homeomorphisms of the interval $[0, 1]$ of the form

$$f_-(x) = \begin{cases} ax & \text{for } x \in [0, x_-], \\ 1 - b(1 - x) & \text{for } x \in (x_-, 1], \end{cases} \quad f_+(x) = \begin{cases} bx & \text{for } x \in [0, x_+], \\ 1 - a(1 - x) & \text{for } x \in (x_+, 1], \end{cases}$$

where $0 < a < 1 < b$ and

$$x_- = \frac{b - 1}{b - a}, \quad x_+ = \frac{1 - a}{b - a}.$$

See Figure 1.

We consider $\{f_-, f_+\}$ as a random dynamical system, which means that iterating the maps, we choose them independently with probabilities p_-, p_+ , where (p_-, p_+) is a given probability vector (i.e. $p_-, p_+ > 0, p_- + p_+ = 1$). Formally, this defines the *step skew product*

$$\mathcal{F}^+ : \Sigma_2^+ \times [0, 1] \rightarrow \Sigma_2^+ \times [0, 1], \quad \mathcal{F}^+(\underline{i}, x) = (\sigma(\underline{i}), f_{i_1}(x)), \tag{1.1}$$

where $\Sigma_2^+ = \{-, +\}^{\mathbb{N}}, \underline{i} = (i_1, i_2, \dots) \in \Sigma_2^+$ and $\sigma : \Sigma_2^+ \rightarrow \Sigma_2^+$ is the left-side shift.

The *endpoint Lyapunov exponents* of an AM system $\{f_-, f_+\}$ are defined as

$$\Lambda(0) = p_- \log f'_-(0) + p_+ \log f'_+(0), \quad \Lambda(1) = p_- \log f'_-(1) + p_+ \log f'_+(1).$$

It is known (see [1, 12]) that if the endpoint Lyapunov exponents are both positive, then the AM system exhibits the *synchronization* property, i.e. for almost all paths

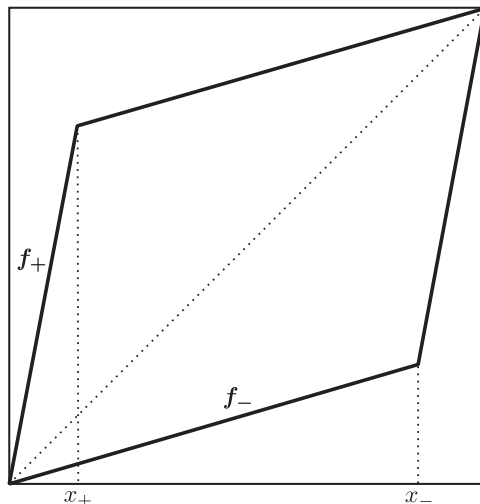


FIGURE 1. An example of a symmetric AM system.

$(i_1, i_2, \dots) \in \{-, +\}^{\mathbb{N}}$ (with respect to the (p_-, p_+) -Bernoulli measure), we have $|f_{i_n} \circ \dots \circ f_{i_1}(x) - f_{i_n} \circ \dots \circ f_{i_1}(y)| \rightarrow 0$ as $n \rightarrow \infty$ for every $x, y \in [0, 1]$. Moreover, in this case, there exists a unique stationary measure μ without atoms at the endpoints of $[0, 1]$, i.e. a Borel probability measure μ on $[0, 1]$, such that

$$\mu = p_- (f_-)_* \mu + p_+ (f_+)_* \mu,$$

with $\mu(\{0, 1\}) = 0$ (see [1], [11, Proposition 4.1], [12, Lemmas 3.2–3.4] and, for a more general case, [7, Theorem 1]). From now on, by a stationary measure for an AM system, we will always mean the measure μ . It is known that μ is non-atomic and is either absolutely continuous or singular with respect to the Lebesgue measure (see [2, Propositions 3.10 and 3.11]).

In [1], Alsedà and Misiurewicz conjectured that the stationary measure μ for an AM system should be singular for typical parameters. In our previous paper [2], we showed that there exist parameters $a, b, (p_-, p_+)$, for which μ is singular with Hausdorff dimension smaller than 1 (see [2, Theorems 2.10 and 2.12]). These examples can be found among AM systems with *resonant* parameters, that is, those with $\log a / \log b \in \mathbb{Q}$. In most of the examples, the measure μ is supported on an *exceptional minimal set*, which is a Cantor set of dimension smaller than 1 (although we also have found examples of singular stationary measures with the support equal to the unit interval, see [2, Theorem 2.16]).

In this paper, already announced in [2], we make a subsequent step to answer the Alsedà and Misiurewicz question, showing that the stationary measure μ is singular for an open set of parameters (a, b) and probability vectors (p_-, p_+) . In particular, we find non-resonant parameters (i.e. those with $\log a / \log b \notin \mathbb{Q}$), for which the corresponding stationary measure is singular (note that non-resonant AM systems necessarily have stationary measures with support equal to $[0, 1]$, see [2, Proposition 2.6]). To prove the result, we present another method to verify singularity of stationary measures for AM systems.

Namely, instead of constructing a measure supported on a set of small dimension, we use the well-known bound on the dimension of stationary measure

$$\dim_H \mu \leq -\frac{H(p_-, p_+)}{\chi(\mu)},$$

in terms of its entropy

$$H(p_-, p_+) = -p_- \log p_- - p_+ \log p_+$$

and the Lyapunov exponent

$$\chi(\mu) = \int_{[0,1]} (p_- \log f'_-(x) + p_+ \log f'_+(x)) d\mu(x),$$

proved in [15] in a very general setting. We find an open set of parameters for which the Lyapunov exponent is small enough (hence the average contraction is strong enough) to guarantee $\dim_H \mu < 1$. The upper bound on the Lyapunov exponent follows from estimates of the expected return time to the interval

$$M = [x_+, x_-].$$

Remark 1.1. One should note that the question of Alsedà and Misiurewicz has been answered when considered within a much broader class of general random interval homeomorphisms with positive endpoint Lyapunov exponents [3, 5] and minimal random homeomorphisms of the circle [4]. More precisely, Czernous and Szarek considered in [5] the closure $\overline{\mathcal{G}}$ of the space \mathcal{G} of all random systems $((g_-, g_+), (p_-, p_+))$ of absolutely continuous, increasing homeomorphisms g_-, g_+ of $[0, 1]$, taken with probabilities p_-, p_+ , such that g_-, g_+ are C^1 in some fixed neighbourhoods of 0 and 1, have positive endpoint Lyapunov exponents and satisfy $g_-(x) < x < g_+(x)$ for $x \in (0, 1)$. In [5, Theorem 10], they proved that for a generic system in $\overline{\mathcal{G}}$ (in the Baire category sense under a natural topology), the unique non-atomic stationary measure is singular. This result was extended by Bradík and Roth in [3, Theorem 6.2], where they allowed the functions to be only differentiable at 0, 1, and showed that in addition to being singular, the stationary measure has typically full support. Similar results were obtained by Czernous [4] for minimal systems on the circle. However, as the finite-dimensional space of AM systems is meagre as a subset of the spaces considered in [3–5], these results give no information on the singularity of stationary measures for typical AM systems.

2. Results

We adopt a convenient notation

$$b = a^{-\gamma}$$

for $a \in (0, 1)$, $\gamma > 0$ and

$$\mathcal{I}: [0, 1] \rightarrow [0, 1], \quad \mathcal{I}(x) = 1 - x,$$

so that a symmetric AM system has the form

$$f_{-}(x) = \begin{cases} ax & \text{for } x \in [0, x_{-}], \\ \mathcal{I}(a^{-\gamma}\mathcal{I}(x)) & \text{for } x \in (x_{-}, 1], \end{cases} \quad f_{+}(x) = \begin{cases} a^{-\gamma}x & \text{for } x \in [0, x_{+}], \\ \mathcal{I}(a\mathcal{I}(x)) & \text{for } x \in (x_{+}, 1], \end{cases} \quad (2.1)$$

where

$$x_{-} = \frac{a^{-\gamma} - 1}{a^{-\gamma} - a}, \quad x_{+} = \frac{1 - a}{a^{-\gamma} - a}.$$

By definition, we have

$$f_{\pm} = \mathcal{I} \circ f_{\mp} \circ \mathcal{I}^{-1} = \mathcal{I} \circ f_{\mp} \circ \mathcal{I}. \quad (2.2)$$

Under this notation, the endpoint Lyapunov exponents for the system in equation (2.1) and a probability vector (p_{-}, p_{+}) are given by

$$\Lambda(0) = (p_{-} - \gamma p_{+}) \log a, \quad \Lambda(1) = (p_{+} - \gamma p_{-}) \log a.$$

Throughout the paper, we assume that $\Lambda(0)$ and $\Lambda(1)$ are positive, which is equivalent to

$$\gamma > \max\left(\frac{p_{-}}{p_{+}}, \frac{p_{+}}{p_{-}}\right). \quad (2.3)$$

In particular, we have $\gamma > 1$. Note that this implies

$$x_{+} < x_{-}. \quad (2.4)$$

Indeed, if $\gamma > 1$, then the endpoint Lyapunov exponents for $p_{-} = p_{+} = 1/2$ are positive, so equation (2.4) follows from [2, Lemma 4.1].

The aim of this paper is to prove the following theorem.

THEOREM 2.1. *Consider a space of symmetric AM systems $\{f_{-}, f_{+}\}$ of the form in equation (2.1) with positive endpoint Lyapunov exponents. Then there is a non-empty open set of parameters $(a, \gamma) \in (0, 1) \times (1, \infty)$ and probability vectors (p_{-}, p_{+}) , such that the corresponding stationary measure μ for the system $\{f_{-}, f_{+}\}$ is singular with Hausdorff dimension smaller than 1. More precisely, there exists $\delta > 0$ such that for every (p_{-}, p_{+}) with $p_{-}, p_{+} < \frac{1}{2} + \delta$, there is a non-empty open interval $J_{p_{-}, p_{+}} \subset (1, 3/2)$, depending continuously on (p_{-}, p_{+}) , such that for $\gamma \in J_{p_{-}, p_{+}}$ and $a \in (0, a_{max})$ for some $a_{max} = a_{max}(\gamma) > 0$, depending continuously on γ , we have*

$$\dim_H \mu \leq \frac{p \log p + (1 - p) \log(1 - p)}{(1 - (1 + \gamma)p^2(p + \gamma)/(\gamma - p(1 - p))) \log a} < 1,$$

where $p = \max(p_{-}, p_{+})$.

In particular, in the case $(p_{-}, p_{+}) = (\frac{1}{2}, \frac{1}{2})$, we have

$$\dim_H \mu \leq \frac{(1 - 4\gamma) \log 2}{(\gamma - 1)(3/2 - \gamma) \log a} < 1$$

for $\gamma \in (1, 3/2)$, $a \in (0, 2^{(1-4\gamma)/((\gamma-1)(3/2-\gamma))})$.

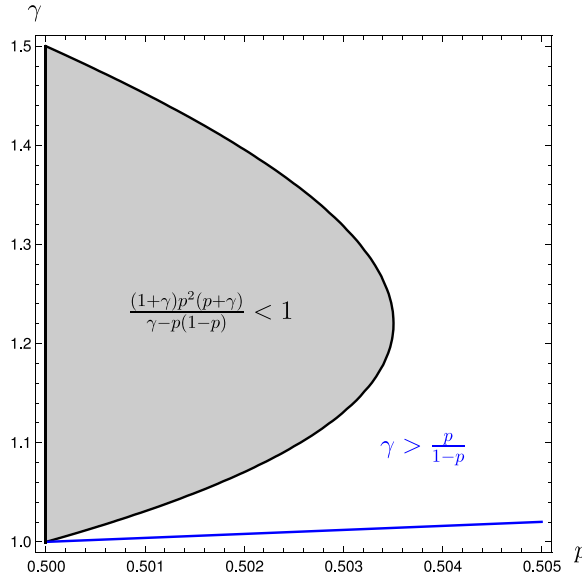


FIGURE 2. The range of parameters $p = \max(p_-, p_+)$ and γ , for which the stationary measure μ for the system in equation (2.1) is singular for sufficiently small $a > 0$.

Remark 2.2. The range of probability vectors (p_-, p_+) for which we obtain the singularity of μ for a non-empty open set of parameters a, γ , is rather small. As the proof of Theorem 2.1 shows, suitable conditions for the possible values of $p = \max(p_-, p_+)$ are given by the inequalities in equations (4.2) and (4.4). Solving them, we obtain $p \in [\frac{1}{2}, p_0)$, where $p_0 = 0.503507\dots$ is the smaller of the two real roots of the polynomial $p^6 - 2p^5 + 5p^4 - 6p^3 - 2p^2 + 1$. As p varies from $\frac{1}{2}$ to p_0 , the range of allowable parameters γ shrinks from the interval $(1, 3/2)$ to a singleton. For such values of p and γ , the measure μ is singular for sufficiently small $a > 0$. See Figure 2.

Remark 2.3. Every system of the form in equation (2.1) with $a < \frac{1}{2}$ is of disjoint type in the sense of [2, Definition 2.3], i.e. $f_-(x_-) < f_+(x_+)$. Indeed, for $a < \frac{1}{2}$, we have

$$2a^{1-\gamma} < a^{-\gamma} < a^{-\gamma} + a,$$

so $a^{1-\gamma} - a < a^{-\gamma} - a^{1-\gamma}$ and

$$f_-(x_-) = a \frac{a^{-\gamma} - 1}{a^{-\gamma} - a} < a^{-\gamma} \frac{1 - a}{a^{-\gamma} - a} = f_+(x_+).$$

Since a simple calculation shows $2^{(1-4\gamma)/((\gamma-1)(3/2-\gamma))} < \frac{1}{2}$ for $\gamma \in (1, 3/2)$, we see that all the systems with the probability vector $(p_-, p_+) = (\frac{1}{2}, \frac{1}{2})$ covered by Theorem 2.1 are of disjoint type.

Remark 2.4. Since the conditions used in the proof of Theorem 2.1 to obtain the singularity of μ define open sets in the space of system parameters, it follows that the singularity of the stationary measure holds also for non-symmetric AM systems with

parameters close enough to those covered by Theorem 2.1. We leave the details to the reader.

3. Preliminaries

We state some standard results from probability and ergodic theory, which we will use within the proofs.

THEOREM 3.1. (Hoeffding’s inequality) *Let X_1, \dots, X_n be independent bounded random variables and let $S_n = X_1 + \dots + X_n$. Then for every $t > 0$,*

$$\mathbb{P}(S_n - \mathbb{E}S_n \geq t) \leq \exp\left(-\frac{2t^2}{\sum_{j=1}^n (\sup X_j - \inf X_j)^2}\right).$$

THEOREM 3.2. (Wald’s identity) *Let X_1, X_2, \dots be independent identically distributed random variables with finite expected value and let N be a stopping time with $\mathbb{E}N < \infty$. Then,*

$$\mathbb{E}(X_1 + \dots + X_N) = \mathbb{E}N \mathbb{E}X_1.$$

THEOREM 3.3. (Kac’s lemma) *Let $F: X \rightarrow X$ be a measurable μ -invariant ergodic transformation of a probability space (X, μ) and let $A \subset X$ be a measurable set with $\mu(A) > 0$. Then,*

$$\int_A n_A d\mu_A = \frac{1}{\mu(A)},$$

where

$$n_A: X \rightarrow \mathbb{N} \cup \{\infty\}, \quad n_A(x) = \inf\{n \geq 1 : F^n(x) \in A\}$$

is the first return time to A and $\mu_A = 1/\mu(A)\mu|_A$.

For the proofs of these results, refer respectively to [14, Theorem 2], [9, Ch. XII, Theorem 2] and [21, Theorem 4.6].

4. Proofs

As noted in the introduction, the proof of Theorem 2.1 is based on an upper bound on the Hausdorff dimension of a stationary measure in terms of its entropy and Lyapunov exponent, in a version proved by Jaroszewska and Rams in [15, Theorem 1]. Consider a symmetric AM system $\{f_-, f_+\}$ of the form in equation (2.1) with positive endpoint Lyapunov exponents for some probability vector (p_-, p_+) , and its stationary measure μ . Recall that the entropy of (p_-, p_+) is defined by

$$H(p_-, p_+) = -p_- \log p_- - p_+ \log p_+,$$

while

$$\chi(\mu) = \int_{[0,1]} (p_- \log f'_-(x) + p_+ \log f'_+(x)) d\mu(x)$$

is the Lyapunov exponent of μ . As μ is non-atomic (see [2, Proposition 3.11]) and f_-, f_+ are differentiable everywhere except for the points x_-, x_+ , the Lyapunov exponent $\chi(\mu)$ is well defined. Moreover, μ is ergodic (see [12, Lemmas 3.2 and 3.4]). It follows that we can use [15, Theorem 1] which asserts that

$$\dim_H \mu \leq -\frac{H(p_-, p_+)}{\chi(\mu)} \tag{4.1}$$

as long as $\chi(\mu) < 0$.

Now we proceed with the details. Let

$$M = [x_+, x_-], \quad L = [x_+, f_-^{-1}(x_+)], \quad R = \mathcal{I}(L) = (f_+^{-1}(x_-), x_-].$$

It follows from equation (2.4) that these intervals are well defined. Note that M, L, R depend on parameters a and γ , but we suppress this dependence in the notation. To estimate the Hausdorff dimension of μ , we find an upper bound for $\chi(\mu)$ in terms of $\mu(M)$ and estimate $\mu(M)$ from below. To this aim, we need the disjointness of the intervals L, R . The following lemma provides the range of parameters for which this condition holds.

LEMMA 4.1. *The following assertions are equivalent:*

- (a) $\bar{L} \cap \bar{R} = \emptyset$;
- (b) $x_+ < f_-(\frac{1}{2})$;
- (c) $\gamma > 1 - \log(a^2 - 2a + 2)/\log a$.

Proof. By equation (2.4) and the fact that $x_+ = \mathcal{I}(x_-)$, we have $\frac{1}{2} < x_-$, so $f_-(\frac{1}{2}) = a/2$ and condition (b) becomes $x_+ < a/2$. Then a direct computation yields the equivalence of conditions (b) and (c). Furthermore, by equation (2.4), condition (a) holds if and only if $f_-^{-1}(x_+) < f_+^{-1}(x_-)$. As $f_- \circ \mathcal{I} = \mathcal{I} \circ f_+$, this is equivalent to $f_-^{-1}(x_+) < \mathcal{I}(f_+^{-1}(x_-))$, which is the same as $f_-^{-1}(x_+) < 1/2$. Applying f_- to both sides, we arrive at condition (b). □

Remark 4.2. The condition in Lemma 4.1(c) can be written as $\gamma - 1 > -\log((1 - a)^2 + 1)/\log a$. As $\log((1 - a)^2 + 1) < \log 2$ for $a \in (0, 1)$, we see that the condition is satisfied provided $\gamma > 1, a \in (0, 2^{1/(1-\gamma)})$.

We can now estimate the measure of M . It is convenient to use the notation

$$p = \max(p_-, p_+).$$

Obviously, $p \in [\frac{1}{2}, 1)$. Note that the condition in equation (2.3) for the positivity of the endpoint Lyapunov exponents can be written as

$$\gamma > \frac{p}{1 - p} \tag{4.2}$$

and the entropy of (p_-, p_+) is equal to

$$H(p) = -p \log p - (1 - p) \log(1 - p).$$

The following lemma provides a lower bound for $\mu(M)$.

LEMMA 4.3. Let $a \in (0, 1)$, $\gamma > 1$ and $p \in [\frac{1}{2}, 1)$ satisfy the conditions in equation (4.2) and Lemma 4.1(c). Then,

$$\mu(M) \geq \frac{\gamma(1-p) - p}{\gamma - p(1-p)}.$$

Before giving the proof of Lemma 4.3, let us explain how it implies Theorem 2.1. Suppose the lemma is true. Then we can estimate the Lyapunov exponent $\chi(\mu)$ in the following way.

COROLLARY 4.4. Let $a \in (0, 1)$, $\gamma > 1$ and $p \in [\frac{1}{2}, 1)$ satisfy the conditions in equation (4.2) and Lemma 4.1(c). Then,

$$\chi(\mu) \leq \left(1 - \frac{(1 + \gamma)p^2(p + \gamma)}{\gamma - p(1 - p)}\right) \log a.$$

Proof. By definition, we have

$$\chi(\mu) = (\mu(M) + (p_- - \gamma p_+) \mu([0, x_+]) + (p_+ - \gamma p_-) \mu([x_-, 1])) \log a. \tag{4.3}$$

Computing the maximum of this expression under the condition $\mu([0, x_+]) + \mu([x_-, 1]) = 1 - \mu(M)$, we obtain

$$\chi(\mu) \leq (1 - (1 + \gamma)p(1 - \mu(M))) \log a.$$

Then Lemma 4.3 provides the required estimate by a direct computation. □

Proof of Theorem 2.1. Let $a \in (0, 1)$, $\gamma > 1$ and $p \in [\frac{1}{2}, 1)$ satisfy the conditions in equation (4.2) and Lemma 4.1(c). By Corollary 4.4, we have $\chi(\mu) < 0$ provided

$$\frac{(1 + \gamma)p^2(p + \gamma)}{\gamma - p(1 - p)} < 1. \tag{4.4}$$

Hence, applying equation (4.1) and Corollary 4.4, we obtain

$$\dim_H \mu \leq \frac{p \log p + (1 - p) \log(1 - p)}{(1 - (1 + \gamma)p^2(p + \gamma)/(\gamma - p(1 - p))) \log a} \tag{4.5}$$

as long as equation (4.4) is satisfied. If, additionally,

$$p \log p + (1 - p) \log(1 - p) > \left(1 - \frac{(1 + \gamma)p^2(p + \gamma)}{\gamma - p(1 - p)}\right) \log a, \tag{4.6}$$

then equation (4.5) provides $\dim_H \mu < 1$. We conclude that the conditions required for $\dim_H \mu < 1$ are equations (4.2), (4.4), Lemma 4.1(c) and equation (4.6).

To find the range of allowable parameters, consider first the case $p_- = p_+ = \frac{1}{2}$ (which corresponds to $p = \frac{1}{2}$). Then the condition in equation (4.2) is equivalent to $\gamma > 1$, while the inequality in equation (4.4) takes the form $2\gamma^2 - 5\gamma + 3 < 0$ and is satisfied for $\gamma \in (1, 3/2)$. Furthermore, by Remark 4.2, the condition in Lemma 4.1(c) is fulfilled for $\gamma > 1$, $a \in (0, 2^{1/(1-\gamma)})$. The condition in equation (4.6) can be written as $(1 - 4\gamma) \log 2 / ((\gamma - 1)(3/2 - \gamma) \log a) < 1$, which is equivalent to

$$a < 2^{(1-4\gamma)/((\gamma-1)(3/2-\gamma))}.$$

A direct computation shows $2^{(1-4\gamma)/((\gamma-1)(3/2-\gamma))} < 2^{1/(1-\gamma)}$ for $\gamma \in (1, 3/2)$. By equation (4.5), we conclude that in the case $p_- = p_+ = \frac{1}{2}$, we have

$$\dim_H \mu \leq \frac{(1 - 4\gamma) \log 2}{(\gamma - 1)(3/2 - \gamma) \log a} < 1$$

for $\gamma \in (1, 3/2)$, $a \in (0, 2^{(1-4\gamma)/((\gamma-1)(3/2-\gamma))})$.

Suppose now that (p_-, p_+) is a probability vector with $p < \frac{1}{2} + \delta$ for a small $\delta > 0$. Note that the functions appearing in equations (4.2) and (4.4) are well defined and continuous for $\gamma \in (1, 3/2)$ and p in a neighbourhood of $\frac{1}{2}$. Hence, equations (4.2) and (4.4) are fulfilled for $\gamma \in J_{p_-, p_+} = J_p$, where $J_p \subset (1, 3/2)$ is an interval slightly smaller than $(1, 3/2)$, depending continuously on $p \in [\frac{1}{2}, \frac{1}{2} + \delta)$. Furthermore, if $\gamma \in J_p$, then the conditions in Lemma 4.1(c) and equation (4.6) hold for sufficiently small $a > 0$, where an upper bound for a can be taken to be a continuous function of γ , which does not depend on p . By equation (4.5), we have

$$\dim_H \mu \leq \frac{p \log p + (1 - p) \log(1 - p)}{(1 - (1 + \gamma)p^2(p + \gamma)/(\gamma - p(1 - p))) \log a} < 1$$

for $p \in [\frac{1}{2}, \frac{1}{2} + \delta)$, $\gamma \in J_p$ and sufficiently small $a > 0$ (with a bound depending continuously on γ). In fact, analysing the inequalities in equations (4.2) (4.4), Lemma 4.1(c) and equation (4.6), one can obtain concrete ranges of parameters a, γ, p , for which $\dim_H \mu < 1$ (cf. Remark 2.2 and Figure 2). □

To complete the proof of Theorem 2.1, it remains to prove Lemma 4.3.

Proof of Lemma 4.3. The proof is based on Kac’s lemma (see Theorem 3.3) and the observation that outside of the interval M , the system $\{f_-, f_+\}$ (after a logarithmic change of coordinates) acts like a random walk with a drift. Note first that $\mu(M) > 0$. Indeed, we have

$$f_+^{-1}(x_-) > x_+, \tag{4.7}$$

as it is straightforward to check that this inequality is equivalent to $a^{1-\gamma} > 1$, which holds since $a \in (0, 1)$ and $\gamma > 1$. This means that the sets M and $f_+^{-1}(M)$ are not disjoint. By symmetry, M and $f_-^{-1}(M)$ are also not disjoint. As $\lim_{n \rightarrow \infty} f_+^{-n}(x_-) = 0$ and $\lim_{n \rightarrow \infty} f_-^{-n}(x_+) = 1$, we see that $\bigcup_{n=0}^{\infty} f_+^{-n}(M) \cup f_-^{-n}(M) = (0, 1)$ and hence $\mu(M) > 0$, as μ is stationary and $\mu(\{0, 1\}) = 0$.

We will apply Kac’s lemma to the step skew product in equation (1.1) and the set $\Sigma_2^+ \times M$. Let $n_M : \Sigma_2^+ \times M \rightarrow \mathbb{N} \cup \{\infty\}$ be the first return time to $\Sigma_2^+ \times M$, that is,

$$n_M(\underline{i}, x) = \inf\{n \geq 1 : (\mathcal{F}^+)^n(\underline{i}, x) \in \Sigma_2^+ \times M\}.$$

Set $\mathbb{P} = \text{Ber}_{p_-, p_+}^+$ to be the (p_-, p_+) -Bernoulli measure on Σ_2^+ . Since $\mathbb{P} \otimes \mu$ is invariant and ergodic for \mathcal{F}^+ (cf. [12, Lemmas 3.2 and A.2]) and $(\mathbb{P} \otimes \mu)(\Sigma_2^+ \times M) = \mu(M) > 0$,

Kac’s lemma implies

$$\int_{\Sigma_2^+ \times M} n_M \, d\nu = \frac{1}{\mu(M)}, \tag{4.8}$$

where

$$\nu = \frac{1}{\mu(M)} (\mathbb{P} \otimes \mu)|_{\Sigma_2^+ \times M}.$$

Recall that we assume the condition in Lemma 4.1(c), so $\bar{L} \cap \bar{R} = \emptyset$. Let

$$C = [\sup L, \inf R],$$

so that $M = L \cup C \cup R$ with the union being disjoint. By the definitions of L , C and R ,

$$f_-(L) \subset [0, x_+), \quad f_-(C \cup R) \cup f_+(L \cup C) \subset M, \quad f_+(R) \subset (x_-, 1]. \tag{4.9}$$

Let

$$E = \{(\underline{i}, x) \in \Sigma_2^+ \times M : f_{i_1}(x) \notin M\} = \{(\underline{i}, x) \in \Sigma_2^+ \times M : n_M > 1\}.$$

It follows from equation (4.9) that

$$E = \{i_1 = -\} \times L \cup \{i_1 = +\} \times R, \tag{4.10}$$

so

$$\nu(E) = \frac{p_- \mu(L) + p_+ \mu(R)}{\mu(M)} \tag{4.11}$$

and as L, R are disjoint subsets of M ,

$$\nu(E) \leq p \frac{\mu(L) + \mu(R)}{\mu(M)} \leq p \tag{4.12}$$

for $p = \max(p_-, p_+)$. By equation (4.10),

$$\int_{\Sigma_2^+ \times M} n_M \, d\nu = 1 - \nu(E) + \int_E n_M \, d\nu = 1 - \nu(E) + \int_{\{i_1 = -\} \times L} n_M \, d\nu + \int_{\{i_1 = +\} \times R} n_M \, d\nu. \tag{4.13}$$

Note that it follows from equation (4.7) that $f_+(x_+) < x_-$, and hence a trajectory $\{f_{i_n} \circ \dots \circ f_{i_1}(x)\}_{n=0}^\infty$ of a point $x \in [0, 1]$ cannot jump from $[0, x_+)$ to $(x_-, 1]$ (or *vice versa*) without passing through M . Combining this observation with the fact that the transformations f_- and f_+ are increasing, we conclude that

$$\begin{aligned} n_M(\underline{i}, x) &\leq n_M(\underline{i}, x_+) && \text{for } (\underline{i}, x) \in \{i_1 = -\} \times L, \\ n_M(\underline{i}, x) &\leq n_M(\underline{i}, x_-) && \text{for } (\underline{i}, x) \in \{i_1 = +\} \times R. \end{aligned} \tag{4.14}$$

Therefore, we can apply equation (4.14) together with equation (4.11) to obtain

$$\begin{aligned} & \int_{\{i_1=-\} \times L} n_M \, d\nu + \int_{\{i_1=+\} \times R} n_M \, d\nu \leq \int_{\{i_1=-\} \times L} n_M(\underline{i}, x_+) \, d\nu + \int_{\{i_1=+\} \times R} n_M(\underline{i}, x_-) \, d\nu \\ & = \frac{\mu(L)}{\mu(M)} \int_{\{i_1=-\}} n_M(\underline{i}, x_+) \, d\mathbb{P}(\underline{i}) + \frac{\mu(R)}{\mu(M)} \int_{\{i_1=+\}} n_M(\underline{i}, x_-) \, d\mathbb{P}(\underline{i}) \\ & = p_- \frac{\mu(L)}{\mu(M)} \mathbb{E}_- N_- + p_+ \frac{\mu(R)}{\mu(M)} \mathbb{E}_+ N_+ \leq \nu(E) \max(\mathbb{E}_- N_-, \mathbb{E}_+ N_+), \end{aligned} \tag{4.15}$$

where

$$N_{\pm}(\underline{i}) = \inf\{n \geq 1 : f_{i_n} \circ \dots \circ f_{i_1}(x_{\mp}) \in M\}$$

and \mathbb{E}_{\pm} is the expectation taken with respect to the conditional measure

$$\mathbb{P}_{\pm} = \frac{1}{\mathbb{P}(i_1 = \pm)} \mathbb{P}|_{\{i_1=\pm\}} = \frac{1}{p_{\pm}} \mathbb{P}|_{\{i_1=\pm\}}.$$

Using equations (4.13), (4.15) and (4.12), we obtain

$$\int_{\Sigma_2^+ \times M} n_M \, d\nu \leq 1 + \nu(E)(\max(\mathbb{E}_- N_-, \mathbb{E}_+ N_+) - 1) \leq 1 + p(\max(\mathbb{E}_- N_-, \mathbb{E}_+ N_+) - 1). \tag{4.16}$$

Define random variables $X_j^{\pm} : \Sigma_2^+ \rightarrow \mathbb{R}, j \in \mathbb{N}$, by

$$X_j^-(i) = \begin{cases} 1 & \text{if } i_j = -, \\ -\gamma & \text{if } i_j = +, \end{cases} \quad X_j^+(i) = \begin{cases} -\gamma & \text{if } i_j = -, \\ 1 & \text{if } i_j = +. \end{cases}$$

Then X_2^-, X_3^-, \dots is an independent and identically distributed sequence of random variables with $\mathbb{P}_-(X_j^- = 1) = p_-$, $\mathbb{P}_-(X_j^- = -\gamma) = p_+$. To estimate $\mathbb{E}_- N_-$, note that for $\underline{i} \in \{i_1 = -\}$, we have

$$N_-(\underline{i}) = \inf\{n \geq 1 : a^{1+X_2^-+\dots+X_n^-} x_+ \geq x_+\} = \inf\{n \geq 2 : X_2^- + \dots + X_n^- \leq -1\},$$

as for $n < N_-(\underline{i})$, we have $f_{i_n} \circ \dots \circ f_{i_1}(x_+) < x_+$ and $f_-(x) = ax$, $f_+(x) = a^{-\gamma}x$ on $[0, x_+]$. Consequently, N_- is a stopping time for $\{X_j^-\}_{j=2}^{\infty}$. We show that $\mathbb{E}_- N_- < \infty$. To do this, note that by Hoeffding’s inequality (see Theorem 3.1) and equation (2.3),

$$\begin{aligned} \mathbb{P}_-(N_- > n + 1) & \leq \mathbb{P}_-\left(\sum_{j=2}^{n+1} X_j^- > -1\right) \\ & = \mathbb{P}_-\left(\sum_{j=2}^{n+1} X_j^- - n\mathbb{E}_- X_2^- \geq -1 - n(p_- - \gamma p_+)\right) \\ & \leq \exp\left(-\frac{2(1 + n(p_- - \gamma p_+))^2}{n(\gamma + 1)^2}\right) \leq \exp(-cn) \end{aligned}$$

for some constant $c > 0$ and $n \in \mathbb{N}$ large enough. We have used here the fact that $t := -1 - n(p_- - \gamma p_+)$ is positive for n large enough, following from equation (2.3). As $\mathbb{E}_- N_- = \sum_{n=0}^\infty \mathbb{P}_-(N > n)$, the above inequality implies $\mathbb{E}_- N_- < \infty$.

Let

$$S_{N_-}(i) = \sum_{n=2}^{N_-(i)} X_n^-(i).$$

This random variable is well defined, since $2 \leq N_- < \infty$ holds \mathbb{P}_- -almost surely. As $\mathbb{E}_- N_- < \infty$, we can apply Wald's identity (see Theorem 3.2) to obtain

$$\mathbb{E}_- S_{N_-} = \mathbb{E}_- X_2^-(\mathbb{E}_- N_- - 1) = (p_- - \gamma p_+)(\mathbb{E}_- N_- - 1). \tag{4.17}$$

To estimate $\mathbb{E}_- S_{N_-}$, we condition on X_2^- and note that $S_{N_-} \geq -1 - \gamma$ almost surely and, by equation (2.3), $-\gamma < -1$. This gives

$$\begin{aligned} \mathbb{E}_- S_{N_-} &= p_- \mathbb{E}_-(S_{N_-} | X_2^- = 1) + p_+ \mathbb{E}_-(S_{N_-} | X_2^- = -\gamma) \\ &\geq p_- (-1 - \gamma) - p_+ \gamma \\ &= -p_- - \gamma. \end{aligned} \tag{4.18}$$

Combining this with equations (2.3) and (4.17), we get

$$\mathbb{E}_- N_- - 1 \leq \frac{p_- + \gamma}{\gamma p_+ - p_-}. \tag{4.19}$$

By the symmetry in equation (2.2) and $x_+ = \mathcal{I}(x_-)$, we can estimate $\mathbb{E}_+ N_+$ in the same way, exchanging the roles of p_- and p_+ , obtaining

$$\mathbb{E}_+ N_+ - 1 \leq \frac{p_+ + \gamma}{\gamma p_- - p_+}. \tag{4.20}$$

Applying equations (4.19) and (4.20) to equation (4.16), we see that

$$\int_{\Sigma_2^+ \times M} n_M(i, x) \, d\nu(i, x) \leq 1 + p \max\left(\frac{p_- + \gamma}{\gamma p_+ - p_-}, \frac{p_+ + \gamma}{\gamma p_- - p_+}\right) = 1 + p \frac{p + \gamma}{\gamma(1 - p) - p}.$$

Invoking equation (4.8), we obtain

$$\mu(M) \geq \frac{1}{1 + p(p + \gamma)/(\gamma(1 - p) - p)} = \frac{\gamma(1 - p) - p}{\gamma - p(1 - p)},$$

which ends the proof. □

We finish the paper with some remarks on the limitations of our method for proving singularity of the measure μ .

Remark 4.5. One should be aware that, in general, the upper bound $-H(p_-, p_+)/\chi(\mu)$ does not coincide with the actual value of $\dim_H \mu$ for AM systems. Indeed, for $(p_-, p_+) = (\frac{1}{2}, \frac{1}{2})$, we have $H(p_-, p_+) = \log 2$ and by equation (4.3),

$$\chi(\mu) = \left(\frac{1 + \gamma}{2} \mu(M) + \frac{1 - \gamma}{2}\right) \log a \geq \log a.$$

However, [2, Theorems 2.10 and 2.12] provide an exact value of the dimension of μ in the resonance case $\gamma = k \in \mathbb{N}, k \geq 2$, yielding

$$\dim_H \mu = \dim_H(\text{supp } \mu) = \frac{\log \eta}{\log a},$$

where $\eta \in (\frac{1}{2}, 1)$ is the unique solution of the equation $\eta^{k+1} - 2\eta + 1 = 0$. Therefore,

$$\dim_H \mu = \frac{\log \eta}{\log a} < \frac{\log 1/2}{\log a} \leq -\frac{H(p_-, p_+)}{\chi(\mu)}.$$

Remark 4.6. It is natural to ask what is a possible range of parameters for which the method presented in this paper could be applied. Let us discuss this in the basic case $p_- = p_+ = \frac{1}{2}$. Following the proof of Theorem 2.1 in this case, we see that by equations (4.1) and (4.3), if for a given $\gamma > 1$ we have

$$\mu(M) > \frac{\gamma - 1}{\gamma + 1},$$

then the measure μ is singular for $a < 1$ small enough (depending on γ). However, combining equations (4.8), (4.16), (4.17), and noting that $\mathbb{E}_- N_- = \mathbb{E}_+ N_+$ and $\mathbb{E}_- S_{N_-} = \mathbb{E}_+ S_{N_+}$ for $p_- = p_+ = \frac{1}{2}$, we see that

$$\mu(M) \geq \frac{\gamma - 1}{\gamma - 1 - \mathbb{E}_- S_{N_-}},$$

provided that the condition of Lemma 4.1(c) is satisfied (which for a fixed $\gamma > 1$ holds for small enough $a \in (0, 1)$). Therefore, if for fixed $\gamma > 1$ inequality

$$\mathbb{E}_- S_{N_-} > -2 \tag{4.21}$$

is satisfied, then μ is singular for $a \in (0, 1)$ small enough. The proof of Theorem 2.1 shows that equation (4.21) holds for $\gamma \in (1, 3/2)$. Figure 3 presents computer simulated values of $\mathbb{E}_- S_{N_-}$ for γ in the interval $(1, 3)$. It suggests that the range of parameters γ for which the singularity of μ holds with a small enough could be extended from $(1, 3/2)$ to a larger set of the form $(1, \gamma_1) \cup (2, \gamma_2)$, for some $\gamma_1 \in (1, 2), \gamma_2 \in (2, 3)$. It is easy to see that one can obtain equation (4.21) for some $\gamma > 3/2$ by conditioning on a larger number of steps in equation (4.18). We do not pursue the task of finding a wider set of possible parameters in this work. One should note, however, that equation (4.21) cannot hold for $\gamma \geq 3$, as the formula from the first line of equation (4.18) can be used together with an obvious bound $S_{N_-} \leq -1$ to obtain (for $p_- = p_+ = \frac{1}{2}$)

$$\mathbb{E}_- S_{N_-} \leq -p_- - p_+ \gamma = \frac{-1 - \gamma}{2},$$

yielding $\mathbb{E}_- S_{N_-} \leq -2$ for $\gamma \geq 3$. This shows that the method used in this paper cannot be (directly) applied for $\gamma \geq 3$ (even though there do exist AM systems with $\gamma \geq 3$ for which μ is singular—see [2, Theorems 2.10 and 2.12]). To obtain an optimal range of γ satisfying equation (4.21), one should compute $\mathbb{E}_- S_{N_-}$ explicitly in terms of γ . This

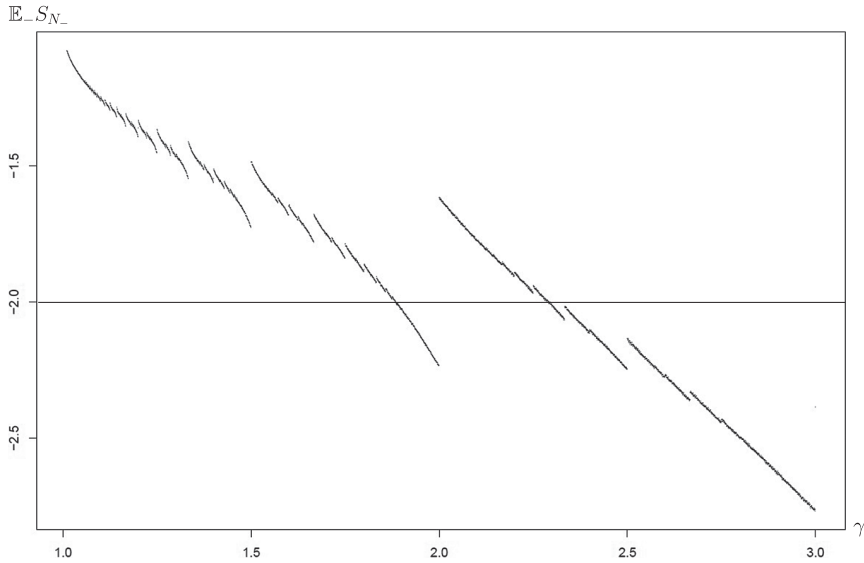


FIGURE 3. Simulated values of $\mathbb{E}_- S_N$ as a function of γ . The values of γ are presented on the x -coordinate axis, while the y -coordinate gives the corresponding value of $\mathbb{E}_- S_N$. Simulations were performed for 4000 values of γ , uniformly spaced in the interval $(1, 3)$. For each choice of γ , we performed 40 000 simulations of 3000 steps of the corresponding random walk.

however seems to be complicated and Figure 3 suggests that one should not expect a simple analytic formula.

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