

## The absolute continuity of the invariant measure of random iterated function systems with overlaps

by

Balázs Bárány (Budapest) and Tomas Persson (Warszawa)

**Abstract.** We consider iterated function systems on the interval with random perturbation. Let  $Y_\varepsilon$  be uniformly distributed in  $[1-\varepsilon, 1+\varepsilon]$  and let  $f_i \in C^{1+\alpha}$  be contractions with fixpoints  $a_i$ . We consider the iterated function system  $\{Y_\varepsilon f_i + a_i(1 - Y_\varepsilon)\}_{i=1}^n$ , where each of the maps is chosen with probability  $p_i$ . It is shown that the invariant density is in  $L^2$  and its  $L^2$  norm does not grow faster than  $1/\sqrt{\varepsilon}$  as  $\varepsilon$  vanishes.

The proof relies on defining a piecewise hyperbolic dynamical system on the cube with an SRB-measure whose projection is the density of the iterated function system.

**1. Introduction and statements of results.** Let  $\{f_1, \dots, f_l\}$  be an iterated function system (IFS) on the real line, where the maps are applied according to the probabilities  $(p_1, \dots, p_l)$ , with the choice of the map random and independent at each step. We assume that for each  $i$ ,  $f_i$  maps  $[-1, 1]$  into itself so that the image is bounded away from  $-1$  and  $1$ , and  $f_i \in C^{1+\alpha}([-1, 1])$ . Let  $\nu$  be the invariant measure of our IFS, i.e.

$$(1.1) \quad \nu = \sum_{i=1}^l p_i \nu \circ f_i^{-1}.$$

Let  $\mu = (p_1, \dots, p_l)^{\mathbb{N}}$  be a Bernoulli measure on the space  $\Sigma = \{1, \dots, l\}^{\mathbb{N}}$ . Let  $h(\underline{p}) = -\sum_{i=1}^l p_i \log p_i$  be the entropy of the left-shift operator with respect to the Bernoulli measure  $\mu$ . It was proved in [7], for non-linear, contracting on average, iterated function systems (and later extended in [3]) that

$$\dim_{\text{H}}(\nu) \leq h/|\chi|,$$

where  $\dim_{\text{H}}(\nu)$  is the Hausdorff dimension of the measure  $\nu$ , and  $\chi$  is the Lyapunov exponent of the IFS associated to the Bernoulli measure  $\mu$ .

---

2010 *Mathematics Subject Classification*: Primary 37C40; Secondary 37H15.

*Key words and phrases*: iterated function system, absolute continuity, random perturbation.

One can expect that, at least “typically”, the measure  $\nu$  is absolutely continuous when  $h/|\chi| > 1$ . Essentially the only known approach to this is transversality. For example, for the linear case with uniform contraction ratios, see [8] and [10]. For the linear case and non-uniform contraction ratios, see [5] and [6]. For the non-linear case, see for example [14] and [1]. We note that there is another direction in the study of iterated function systems with overlaps, which is concerned with concrete, but non-typical systems, often of arithmetic nature, for which there is a dimension drop (see for example [4]).

Throughout this paper we are interested in studying absolute continuity with density in  $L^2$ . We will study a modification of the problem, namely we consider a random perturbation of the functions. The linear case was studied by Peres, Simon and Solomyak in [9]. They proved absolute continuity for random linear IFS, with non-uniform contraction ratios and also  $L^2$  and continuous density in the uniform case. We extend this result by proving  $L^2$  density with non-uniform contraction ratios and in the non-linear case.

We consider two cases. First let us suppose that for each  $i \in \{1, \dots, l\}$ ,  $f_i$  maps  $[-1, 1)$  into itself,  $f_i([-1, 1))$  is bounded away from  $-1$  and  $1$ ,  $f_i \in C^{1+\alpha}([-1, 1))$  and

$$(1.2) \quad 0 < \lambda_{i,\min} \leq |f'_i(x)| \leq \lambda_{i,\max} < 1$$

for every  $x \in [-1, 1)$ . Moreover suppose that for every  $i$  the fixed point of  $f_i$  is  $a_i \in (-1, 1)$ , and

$$(1.3) \quad i \neq j \Rightarrow a_i \neq a_j.$$

Let  $Y_\varepsilon$  be uniformly distributed on  $[1 - \varepsilon, 1 + \varepsilon]$ . Denote the probability measure of  $Y_\varepsilon$  by  $\eta_\varepsilon$ . Let

$$(1.4) \quad f_{i,Y_\varepsilon}(x) = Y_\varepsilon f_i(x) + a_i(1 - Y_\varepsilon)$$

for every  $i \in \{1, \dots, l\}$ . Then  $f_{i,Y_\varepsilon}(x)$  is in  $[-1, 1)$  for all values of  $x \in [-1, 1)$  and  $Y_\varepsilon$ , provided  $\varepsilon$  is sufficiently small. The iterated maps are applied randomly according to the stationary measure  $\mu$ , with the sequence of independent and identically distributed errors  $y_1, y_2, \dots$  distributed as  $Y_\varepsilon$ , independent of the choice of the function. The Lyapunov exponent of the IFS is defined by

$$\chi(\mu, \eta_\varepsilon) = \mathbb{E}(\log(Y_\varepsilon f'))$$

and it is easy to see that

$$\chi(\mu, \eta_\varepsilon) < \sum_{i=1}^l p_i \log((1 + \varepsilon)\lambda_{i,\max}) < 0$$

for sufficiently small  $\varepsilon > 0$ . Let  $Z_\varepsilon$  be the random variable

$$(1.5) \quad Z_\varepsilon := \lim_{n \rightarrow \infty} f_{i_1, y_{1,\varepsilon}} \circ \dots \circ f_{i_n, y_{n,\varepsilon}}(0),$$

where the numbers  $i_k$  are i.i.d., with distribution  $\mu$  on  $\{1, \dots, l\}$ , and  $y_{k,\varepsilon}$  are pairwise independent with the distribution of  $Y_\varepsilon$  and also independent of the choice of  $i_k$ . Let  $\nu_\varepsilon$  be the distribution of  $Z_\varepsilon$ .

One can easily prove the following theorem.

**THEOREM 1.1.** *The measure  $\nu_\varepsilon$  converges weakly as  $\varepsilon \rightarrow 0$  to the measure  $\nu$  satisfying (1.1).*

**THEOREM 1.2.** *Let  $\nu_\varepsilon$  be the distribution of the limit (1.5). Assume that (1.2) and (1.3) hold, and*

$$(1.6) \quad \sum_{i=1}^l p_i^2 \frac{\lambda_{i,\max}}{\lambda_{i,\min}^2} < 1.$$

*Then for every sufficiently small  $\varepsilon > 0$ ,  $\nu_\varepsilon$  is absolutely continuous with respect to Lebesgue measure, with density in  $L^2$ , and there exists a constant  $C$  such that the density of  $\nu_\varepsilon$  satisfies*

$$\|\nu_\varepsilon\|_2 \leq C/\sqrt{\varepsilon}.$$

**REMARK 1.** Let

$$C'_\varepsilon = \sqrt{\frac{32}{(1 - \sum_{i=1}^l p_i^2 \frac{(1+\varepsilon)\lambda_{i,\max}}{((1-\varepsilon)\lambda_{i,\min})^2}) C''_\varepsilon}},$$

$$C''_\varepsilon = \min_{i \neq j} \left\{ \frac{|a_i - a_j| + \varepsilon(-|a_i + a_j| - 2)}{1 - \varepsilon^2} \right\}.$$

The proof of Theorem 1.2 will show that  $\|\nu_\varepsilon\|_2 \leq C'_\varepsilon/\sqrt{\varepsilon}$ . Hence we can choose any  $C > \lim_{\varepsilon \rightarrow 0} C'_\varepsilon$ .

**REMARK 2.** Actually the proof of Theorem 1.2 shows that  $Z_\varepsilon$  conditioned on the perturbations  $y_{1,\varepsilon}, y_{2,\varepsilon}, \dots$  has density in  $L^2$  for almost all  $y_{1,\varepsilon}, y_{2,\varepsilon}, \dots$ .

We can state an easy corollary of the theorem.

**COROLLARY 1.3.** *Let  $\{\lambda_i Y_\varepsilon x + a_i(1 - \lambda_i Y_\varepsilon)\}_{i=1}^l$  be a random iterated function system. Assume that (1.3) holds, and*

$$(1.7) \quad \sum_{i=1}^l \frac{p_i^2}{\lambda_i} < 1.$$

*Then for every sufficiently small  $\varepsilon > 0$ ,  $\nu_\varepsilon$  is absolutely continuous with respect to Lebesgue measure with density in  $L^2$ , and there exists a constant  $C$  such that*

$$\|\nu_\varepsilon\|_2 \leq C/\sqrt{\varepsilon}.$$

We study another case of random perturbation, namely let  $\tilde{\lambda}_{i,\varepsilon}$  be uniformly distributed on  $[\lambda_i - \varepsilon, \lambda_i + \varepsilon]$ . Let  $\{\tilde{\lambda}_{i,\varepsilon} x + a_i(1 - \tilde{\lambda}_{i,\varepsilon})\}_{i=1}^l$  be our random

iterated function system, where  $a_i \neq a_j$  for every  $i \neq j$ . Let  $\underline{\lambda} = (\lambda_1, \dots, \lambda_l)$ , and  $X_{\underline{\lambda}, \varepsilon}$  be the random variable

$$(1.8) \quad X_{\underline{\lambda}, \varepsilon} = \sum_{k=1}^{\infty} (a_{i_k} (1 - \tilde{\lambda}_{i_k, \varepsilon})) \prod_{j=1}^{k-1} \tilde{\lambda}_{i_j, \varepsilon}$$

where the numbers  $i_k$  are i.i.d. with distribution  $\mu$  on  $\{1, \dots, l\}$ , and  $\tilde{\lambda}_{i_k, \varepsilon}$  are pairwise independent. Let  $\nu_{\underline{\lambda}, \varepsilon}$  denote the distribution of  $X_{\underline{\lambda}, \varepsilon}$ . Moreover let  $\nu_{\underline{\lambda}}$  be the invariant measure of the iterated function system  $\{\lambda_i x + a_i(1 - \lambda_i)\}_{i=1}^l$  according to  $\mu$ .

**THEOREM 1.4.** *The measure  $\nu_{\underline{\lambda}, \varepsilon}$  converges weakly to  $\nu_{\underline{\lambda}}$  as  $\varepsilon \rightarrow 0$ .*

To have a statement similar to Theorem 1.2 we need a technical assumption

$$(1.9) \quad \min_{i \neq j} \left| \frac{a_j \lambda_i - a_i \lambda_j}{\lambda_i - \lambda_j} \right| > 1.$$

**THEOREM 1.5.** *Suppose that (1.9) and (1.3) hold, and moreover*

$$(1.10) \quad \sum_{i=1}^l \frac{p_i^2}{\lambda_i} < 1.$$

*Then for every sufficiently small  $\varepsilon > 0$ , the measure  $\nu_{\underline{\lambda}, \varepsilon}$  is absolutely continuous with respect to Lebesgue measure, with density in  $L^2$ , and there exists a constant  $C$  such that*

$$\|\nu_{\underline{\lambda}, \varepsilon}\|_2 \leq C/\sqrt{\varepsilon}.$$

**REMARK 3.** Let

$$C'_\varepsilon = \sqrt{\frac{32}{(1 - \sum_{i=1}^l p_i^2 \frac{\lambda_i + \varepsilon}{(\lambda_i - \varepsilon)^2}) C''_\varepsilon}},$$

$$C''_\varepsilon = \sigma \min_{i \neq j} \frac{|a_i \lambda_j - a_j \lambda_i| - |\lambda_i - \lambda_j|}{\lambda_i \lambda_j},$$

where  $0 < \sigma < 1$ . As in Remark 1, the proof of Theorem 1.5 will show that  $\|\nu_{\underline{\lambda}, \varepsilon}\|_2 \leq C'_\varepsilon/\sqrt{\varepsilon}$  for small  $\varepsilon$ .

The main difference between Theorem 1.5 and Corollary 1.3 is the random perturbation. Namely, in Theorem 1.5 we choose the contraction ratio uniformly in the  $\varepsilon$ -neighborhood of  $\lambda_i$ , while in Corollary 1.3 we choose the contraction ratio uniformly in the  $\lambda_i \varepsilon$ -neighborhood of  $\lambda_i$ .

Throughout this paper we will use the method of [11].

**2. Proof of Theorem 1.2.** Let  $Q = [-1, 1]^3$  and  $m \in \mathbb{N}$ . We partition the cube  $Q$  into the rectangles  $\{Q_{1,k}, \dots, Q_{l,k}\}_{k=0}^{2^m-1}$ , where

$$Q_{i,k} = \left\{ (x, y, z) \in Q : -1 + 2 \sum_{j=1}^{i-1} p_j \leq y < -1 + 2 \sum_{j=1}^i p_j, \right. \\ \left. -1 + k2^{-m+1} \leq z < -1 + (k+1)2^{-m+1} \right\},$$

where we use the convention that an empty sum is 0. Hence we slice  $Q$  into  $2^m$  slices along the  $z$ -axis and  $l$  slices along the  $y$ -axis. We thereby get  $2^m l$  pieces which we call  $Q_{i,k}$ , according to the definition above.

Let

$$Q_i = \bigcup_{k=0}^{2^m-1} Q_{i,k}.$$

We define a map  $g_{\varepsilon,m} : Q \rightarrow Q$  so that each slice  $Q_{i,k}$  is expanded as much as possible in the second and third coordinates. In the first coordinate it is mapped according to a perturbation of  $f_i$ , and hence contracted. Which perturbation is chosen depends on the third coordinate. There is a picture of this in Figure 1.

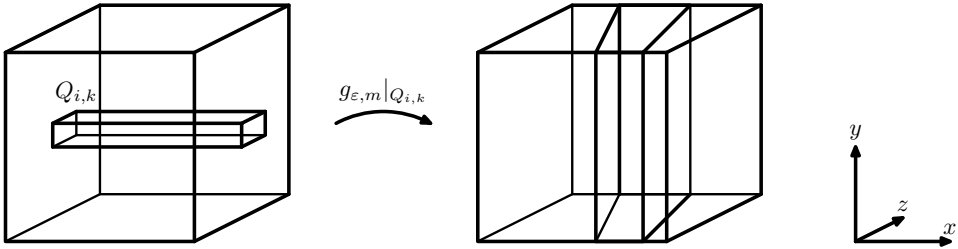


Fig. 1. The action of  $g_{\varepsilon,m}$  restricted to  $Q_{i,k}$

More precisely, we define  $g_{\varepsilon,m} : Q \rightarrow Q$  by setting, for  $(x, y, z) \in Q_{i,k}$ ,

$$g_{\varepsilon,m} : (x, y, z) \mapsto \left( d(z)f_i(x) + a_i(1 - d(z)), \frac{1}{p_i}y + b(y), 2^m z + c(z) \right),$$

where

$$d(z) = 1 + 2^m \varepsilon (z - (-1 + (k+1/2)2^{-m+1})) \quad \text{for } (x, y, z) \in Q_{i,k},$$

$$b(y) = 1 - \frac{1}{p_i} \left( -1 + 2 \sum_{j=1}^i p_j \right) \quad \text{for } (x, y, z) \in Q_{i,k},$$

$$c(z) = 2^m - 2k - 1 \quad \text{for } (x, y, z) \in Q_{i,k}.$$

Hence  $g_{\varepsilon,m}$  maps each of the pieces  $Q_{i,j}$  so that it is contracted in the  $x$ -direction and fully expanded in the  $y$ - and  $z$ -directions.

Let  $\mathcal{L}_3$  be the normalised Lebesgue measure on  $Q$ . The measures

$$\gamma_{\varepsilon,m,n} = \frac{1}{n} \sum_{k=0}^{n-1} \mathcal{L}_3 \circ g_{\varepsilon,m}^{-k}$$

converge weakly to an SRB-measure  $\gamma_{\varepsilon,m}$  as  $n \rightarrow \infty$  (see [12] and [13]). The measure  $\gamma_{\varepsilon,m}$  is ergodic by the Hopf argument, since  $g_{\varepsilon,m}$  is hyperbolic and the stable and unstable manifolds are parallel to the coordinate axes and have maximal extension in the box  $Q$ . Moreover, let  $\nu_{\varepsilon,m}$  be the projection of  $\gamma_{\varepsilon,m}$  onto the first coordinate. More precisely, if  $E \subset [-1, 1)$  is a measurable set, then we define  $\nu_{\varepsilon,m}(E) = \gamma_{\varepsilon,m}(E \times [-1, 1) \times [-1, 1))$ .

The measure  $\nu_{\varepsilon,m}$  is the distribution of the limit

$$\lim_{n \rightarrow \infty} f_{i_1, y_{1,\varepsilon}} \circ \cdots \circ f_{i_n, y_{n,\varepsilon}}(0),$$

where  $y_{i,\varepsilon}$  are uniformly distributed on  $[1 - \varepsilon, 1 + \varepsilon]$ , but not independent. However, one can easily prove the following lemma.

LEMMA 2.1. *The measure  $\nu_{\varepsilon,m}$  converges weakly to  $\nu_\varepsilon$  as  $m \rightarrow \infty$ .*

Let

$$A_i = \{(i, 0), (i, 1), \dots, (i, 2^m - 1)\} \quad \text{and} \quad A = \bigcup_{i=1}^l A_i.$$

If  $a = (i, k) \in A$  we will write  $\hat{Q}_a$  for  $Q_{i,k}$ . With this notation we have

$$Q = \bigcup_{a \in A} \hat{Q}_a \quad \text{and} \quad Q_i = \bigcup_{a \in A_i} \hat{Q}_a, \quad i = 0, 1, \dots, l.$$

Let  $\Theta_0 = A^{\mathbb{N} \cup \{0\}}$ . If  $p \in Q$  then there is a unique sequence  $\rho_0(p) = \{\rho_0(p)_k\}_{k=0}^\infty \in \Theta_0$  such that

$$g_{\varepsilon,m}^k(p) \in Q_{\rho_0(p)_k}, \quad k = 0, 1, \dots$$

The map  $\rho_0: Q \rightarrow \Theta_0$  is not injective. We have  $\rho_0(x, y, z) = \rho_0(x', y', z')$  if  $y = y'$  and  $z = z'$ , but  $\rho_0(x, y, z) \neq \rho_0(x', y', z')$  otherwise. Hence we can (and will) use the notation  $\rho_0(y, z)$  instead of  $\rho_0(x, y, z)$ .

We will denote elements in  $\Theta_0$  by  $\mathbf{a}$ ,  $\mathbf{b}$  and so on. We let  $\sigma$  denote the left shift on  $\Theta_0$ , defined in the usual way.

We can transfer the measures  $\gamma_{\varepsilon,m}$  to a measure  $\gamma_{\Theta_0}$  by  $\gamma_{\Theta_0} = \gamma_{\varepsilon,m} \circ \rho_0^{-1}$ .

We let  $\Theta$  denote the natural extension of  $\Theta_0$ . That is,  $\Theta$  is the set of all two-sided infinite sequences such that any one-sided infinite subsequence of a sequence in  $\Theta$  is a sequence in  $\Theta_0$ . The measure  $\gamma_{\Theta_0}$  defines an ergodic measure  $\gamma_\Theta$  on  $\Theta$  in a natural way. If  $\xi: \Theta \rightarrow \Theta_0$  is defined by  $\xi(\{i_k\}_{k \in \mathbb{Z}}) = \{i_k\}_{k \in \mathbb{N} \cup \{0\}}$ , then we define  $\gamma_\Theta(\xi^{-1}E) = \gamma_{\Theta_0}(E)$ . We can define a map  $\rho^{-1}: \Theta \rightarrow Q$  such that  $\rho^{-1}(\sigma(\mathbf{a})) = g_{\varepsilon,m}(\rho^{-1}(\mathbf{a}))$  for any sequence  $\mathbf{a} \in \Theta$ .

We note that the  $L^2$  norm of the density  $\nu_{\varepsilon,m}$  is not larger than twice that of the density of  $\gamma_{\varepsilon,m}$ . If  $h_{\nu_{\varepsilon,m}}(x)$  and  $h_{\gamma_{\varepsilon,m}}(x, y, z)$  denote the densities of  $\nu_{\varepsilon,m}$  and  $\gamma_{\varepsilon,m}$  respectively, then by Cauchy–Schwarz’s inequality

$$\begin{aligned} \|\nu_{\varepsilon,m}\|_2^2 &\leq \int_{-1}^1 h_{\nu_{\varepsilon,m}}(x)^2 dx = 32 \int_{-1}^1 \left( \int_{-1}^1 \int_{-1}^1 h_{\gamma_{\varepsilon,m}}(x, y, z) \frac{dy}{2} \frac{dz}{2} \right)^2 \frac{dx}{2} \\ &\leq 32 \int_{-1}^1 \int_{-1}^1 \int_{-1}^1 h_{\gamma_{\varepsilon,m}}(x, y, z)^2 \frac{dy}{2} \frac{dz}{2} \frac{dx}{2} = 4\|\gamma_{\varepsilon,m}\|_2^2. \end{aligned}$$

This proves that if  $\gamma_{\varepsilon,m}$  has  $L^2$  density, then so has  $\nu_{\varepsilon,m}$ , and

$$(2.1) \quad \|\nu_{\varepsilon,m}\|_2 \leq 2\|\gamma_{\varepsilon,m}\|_2.$$

If  $p$  is a point in  $Q$ , then we let  $T_pQ$  denote the tangent space at  $p$ . For each  $p$  in  $Q$  we define the following cone in the tangent space  $T_pQ$ :

$$C_p = \left\{ (u, v, w) \in T_pQ : \left| \frac{u}{w} \right|, \left| \frac{v}{w} \right| < \frac{2^{m+1}\varepsilon}{2^m - \lambda_{\max,\max}(1 + \varepsilon)} \right\},$$

where  $\lambda_{\max,\max} = \max_i \lambda_{i,\max} = \max_i \sup_{x \in [-1,1]} |f'_i(x)|$ . The following lemma states that the set of cones  $C_p$  defines a family of unstable cones, and that images of certain curves intersect transversally. There is an illustration of the transversality in Figure 2.

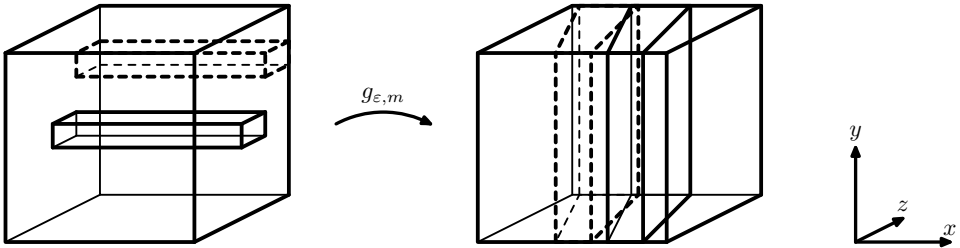


Fig. 2. Any two different  $Q_{i,k}$  and  $Q_{j,l}$  on the same height ( $i = j$ ) share the same image, but in the case when  $i \neq j$  their images have transversal intersection if they intersect.

LEMMA 2.2. *The cones  $C_p$  make up a family of unstable cones, that is,  $d_p g_{\varepsilon,m}(C_p) \subset C_{g_{\varepsilon,m}(p)}$ .*

Moreover, for sufficiently large  $m$  and every  $0 < \varepsilon < \min_{i \neq j} \frac{|a_i - a_j|}{2 + |a_i + a_j|}$ , if  $\zeta_1 \subset Q_{\xi_1}$  and  $\zeta_2 \subset Q_{\xi_2}$  are two curve segments with tangents in  $C_p$  such that  $\xi_1 \in A_i$  and  $\xi_2 \in A_j$ ,  $i \neq j$ , then if  $g_{\varepsilon,m}(\zeta_1)$  and  $g_{\varepsilon,m}(\zeta_2)$  intersect, and if  $(u_1, v_1, 1)$  and  $(u_2, v_2, 1)$  are tangents to  $g_{\varepsilon,m}(\zeta_1)$  and  $g_{\varepsilon,m}(\zeta_2)$  respectively, it follows that  $|u_1 - u_2| > C_{\varepsilon,m}\varepsilon$ , where

$$C_{\varepsilon,m} = \min_{i \neq j} \left\{ \frac{|a_i - a_j| + \varepsilon(-|a_i + a_j| - 2)}{1 - \varepsilon^2} - \frac{4(1 + \varepsilon)\lambda_{\max,\max}}{2^m - \lambda_{\max,\max}(1 + \varepsilon)} \right\}.$$

*Proof.* The Jacobian of  $g_{\varepsilon,m}$  is

$$d_p g_{\varepsilon,m} = \begin{pmatrix} d(z)f'_i(x) & 0 & 2^m \varepsilon (f_i(x) - a_i) \\ 0 & 1/p_i & 0 \\ 0 & 0 & 2^m \end{pmatrix}$$

where  $p = (x, y, z) \in Q_{i,k}$ . If  $(u, v, w) \in C_p$ , then

$$d_p g_{\varepsilon,m}(u, v, w) = \begin{pmatrix} d(z)f'_i(x)u + 2^m \varepsilon (f_i(x) - a_i)w \\ (1/p_i)v \\ 2^m w \end{pmatrix}$$

We just need to check that this vector is in  $C_p$ , provided that  $m$  is large. This is easily checked, using that  $|d(z)| \leq 1 + \varepsilon$ ,  $|f'_i(x)| \leq \lambda_{i,\max}$  and  $|f_i(x) - a_i| \leq 2$ . Indeed,

$$\begin{aligned} \frac{|d(z)f'_i(x)u + 2^m \varepsilon (f_i(x) - a_i)w|}{|2^m w|} &\leq \frac{(1 + \varepsilon)\lambda_{i,\max}}{2^m} \frac{|u|}{|w|} + 2\varepsilon \\ &\leq \frac{(1 + \varepsilon)\lambda_{i,\max}}{2^m} \frac{2^{m+1}\varepsilon}{2^m - (1 + \varepsilon)\lambda_{\max,\max}} + 2\varepsilon \leq \frac{2^{m+1}\varepsilon}{2^m - (1 + \varepsilon)\lambda_{\max,\max}} \end{aligned}$$

and

$$\frac{|(1/p_i)v|}{|2^m w|} \leq \frac{1}{p_i 2^m} \frac{2^{m+1}\varepsilon}{2^m - (1 + \varepsilon)\lambda_{\max,\max}} \leq \frac{2^{m+1}\varepsilon}{2^m - (1 + \varepsilon)\lambda_{\max,\max}}$$

proves that  $d_p g_{\varepsilon,m}(C_p) \subset C_{g_{\varepsilon,m}(p)}$  if  $m$  is sufficiently large, so that  $2^m - (1 + \varepsilon)\lambda_{\max,\max} > 0$  and  $p_i 2^m > 1$ .

To prove the other statement of the lemma, assume that  $p = (x_p, y_p, z_p) \in Q_i$  and  $q = (x_q, y_q, z_q) \in Q_j$ ,  $i \neq j$ , are such that  $g_{\varepsilon,m}(p) = g_{\varepsilon,m}(q) = (x, y, z)$ . Then, if  $p \in Q_i$ ,

$$d_p g_{\varepsilon,m} : (u, v, 1) \mapsto 2^m \left( \frac{d(z_p)f'_i(x_p)}{2^m} u + (f_i(x_p) - a_i)\varepsilon, \frac{v}{p_i}, 1 \right).$$

Then

$$f_i(x_p) = \frac{x - a_i(1 - d(z_p))}{d(z_p)} \quad \text{and} \quad f_j(x_q) = \frac{x - a_j(1 - d(z_q))}{d(z_q)}.$$

Without loss of generality, assume that  $a_i > a_j$ . For simplicity we study the case  $x \geq a_i > a_j$ . The proofs for  $a_i \geq x \geq a_j$  and  $a_i > a_j \geq x$  are similar. Then

$$d_p g_{\varepsilon,m}(C_p) \subset \left\{ w(u, v, 1) : \frac{x - a_i}{1 + \varepsilon} \varepsilon - \Delta_i \varepsilon \leq u \leq \frac{x - a_i}{1 - \varepsilon} \varepsilon + \Delta_i \varepsilon \right\},$$



where  $\Delta_i = \frac{2(1+\varepsilon)\lambda_{i,\max}}{2^m - \lambda_{\max,\max}(1+\varepsilon)}$ . Therefore

$$\begin{aligned} |u_2 - u_1| &\geq \frac{x - a_j}{1 + \varepsilon} \varepsilon - \frac{x - a_i}{1 - \varepsilon} \varepsilon - (\Delta_i + \Delta_j) \varepsilon \\ &\geq \left( \frac{a_i - a_j + \varepsilon(a_i + a_j - 2)}{1 - \varepsilon^2} - 2 \max_i \Delta_i \right) \varepsilon \end{aligned}$$

for every  $x \geq a_i > a_j$ . Let  $\Delta_{\max} = \max_i \Delta_i$ . Since  $0 < \varepsilon < \min_{i \neq j} \frac{|a_i - a_j|}{2 + |a_i + a_j|}$ , we have

$$\frac{a_i - a_j + \varepsilon(a_i + a_j - 2)}{1 - \varepsilon^2} > 0.$$

Therefore

$$\frac{a_i - a_j + \varepsilon(a_i + a_j - 2)}{1 - \varepsilon^2} - 2\Delta_{\max} > 0,$$

for sufficiently large  $m$ . By similar methods we have for  $a_i \geq x \geq a_j$ ,

$$|u_2 - u_1| \geq \left( \frac{a_i - a_j}{1 + \varepsilon} - 2\Delta_{\max} \right) \varepsilon,$$

and for  $a_i > a_j \geq x$ ,

$$|u_2 - u_1| \geq \left( \frac{a_i - a_j - \varepsilon(a_i + a_j + 2)}{1 - \varepsilon^2} - 2\Delta_{\max} \right) \varepsilon.$$

Therefore we can choose

$$C_{\varepsilon,m} = \min_{i \neq j} \left\{ \frac{|a_i - a_j| + \varepsilon(-|a_i + a_j| - 2)}{1 - \varepsilon^2} - 2\Delta_{\max} \right\}. \blacksquare$$

The rest of the section follows Tsujii's article [15].

*Proof of Theorem 1.2.* For any  $r > 0$  we define the bilinear form  $(\cdot, \cdot)_r$  of signed measures on  $\mathbb{R}$  by

$$(\rho_1, \rho_2)_r = \int_{\mathbb{R}} \rho_1(B_r(x)) \rho_2(B_r(x)) dx$$

where  $B_r(x) = [x - r, x + r]$ . It is easy to see that if

$$\liminf_{r \rightarrow 0} \frac{1}{r^2} (\rho, \rho)_r < \infty$$

then the measure  $\rho$  has density in  $L^2$  (see [15]). Moreover

$$\|\rho\|_2^2 \leq \liminf_{r \rightarrow 0} \frac{1}{r^2} (\rho, \rho)_r.$$

Let  $\gamma_z$  denote the conditional measure of  $\gamma_{\varepsilon,m}$  on the set  $R_z = \{(u, v, w) \in Q : v = y, w = z\}$ . Since the one-dimensional Lebesgue measure is invariant under the action of  $g_{\varepsilon,m}$  projected to the second coordinate, we conclude

that  $\gamma_z$  is independent of  $y$  almost everywhere. It follows that

$$(2.2) \quad \|\gamma_{\varepsilon,m}\|_2^2 = \int_{-1}^1 \|\gamma_z\|_2^2 dz.$$

Let

$$J(r) := \frac{1}{r^2} \int_{-1}^1 (\gamma_z, \gamma_z)_r dz.$$

By the invariance of  $\gamma_{\varepsilon,m}$  it follows that

$$(2.3) \quad \gamma_z = 2^{-m} \sum_{i=1}^l p_i \sum_{a \in A_i} \gamma_{g_{\varepsilon,m}^{-a}(z)} \circ g_{\varepsilon,m}^{-a},$$

where  $g_{\varepsilon,m}^{-a}$  denotes the inverse branch of  $g_{\varepsilon,m}$  with image in  $\hat{Q}_a$ . Recall that  $a \in A_i$  means that  $a = (i, k)$  for some  $k$ , so that  $\hat{Q}_a = Q_{i,k}$  for some  $k$ . We denote the measure  $\gamma_{g_{\varepsilon,m}^{-a}(z)} \circ g_{\varepsilon,m}^{-a}$  by  $\sigma_{a,z}$ . Then by (2.3) and the definition of  $J(r)$ ,

$$(2.4) \quad J(r) = \frac{1}{2^{2m} r^2} \sum_{i=1}^l \sum_{j=1}^l p_i p_j \sum_{a \in A_i} \sum_{b \in A_j} \int_{-1}^1 (\sigma_{a,z}, \sigma_{b,z})_r dz.$$

For fixed  $a, b \in A_i$ ,

$$(2.5) \quad \begin{aligned} (\sigma_{a,z}, \sigma_{b,z})_r &\leq (\sigma_{a,z}, \sigma_{a,z})_r^{1/2} (\sigma_{b,z}, \sigma_{b,z})_r^{1/2} \\ &\leq (1 + \varepsilon) \lambda_{i,\max} (\gamma_{g_{\varepsilon,m}^{-a}(z)}, \gamma_{g_{\varepsilon,m}^{-a}(z)})^{1/2} \frac{r}{(1-\varepsilon)\lambda_{i,\min}} \times (\gamma_{g_{\varepsilon,m}^{-b}(z)}, \gamma_{g_{\varepsilon,m}^{-b}(z)})^{1/2} \frac{r}{(1-\varepsilon)\lambda_{i,\min}} \\ &\leq (1 + \varepsilon) \lambda_{i,\max} \frac{(\gamma_{g_{\varepsilon,m}^{-a}(z)}, \gamma_{g_{\varepsilon,m}^{-a}(z)})^{1/2} \frac{r}{(1-\varepsilon)\lambda_{i,\min}} + (\gamma_{g_{\varepsilon,m}^{-b}(z)}, \gamma_{g_{\varepsilon,m}^{-b}(z)})^{1/2} \frac{r}{(1-\varepsilon)\lambda_{i,\min}}}{2}. \end{aligned}$$

Moreover, if  $a \in A_i$  and  $b \in A_j$ ,  $i \neq j$ , then

$$\begin{aligned} (\sigma_{a,z}, \sigma_{b,z})_r &= \int \sigma_{a,z}(B_r(x)) \sigma_{b,z}(B_r(x)) dx \\ &= \iiint \mathbb{I}_{\{s: |s-x| < r\}}(s) \mathbb{I}_{\{t: |t-x| < r\}}(t) d\sigma_{a,z}(s) d\sigma_{b,z}(t) dx \\ &\leq \iint 2r \mathbb{I}_{\{(s,t): |s-t| < 2r\}}(s, t) d\sigma_{a,z}(s) d\sigma_{b,z}(t) \\ &= \iint \mathbb{I}_{\{(\mathbf{c}, \mathbf{d}): |\rho^{-1}(\dots c_{-2} c_{-1} a \rho_0(z)) - \rho^{-1}(\dots d_{-2} d_{-1} b \rho_0(z))| < 2r\}}(\mathbf{c}, \mathbf{d}) \\ &\quad d\gamma_{\Theta}(\mathbf{c}) d\gamma_{\Theta}(\mathbf{d}). \end{aligned}$$

Let us comment on the notation  $\rho_0(z)$ . Actually  $\rho_0(z)$  is not defined, but rather  $\rho_0(x, y, z)$ . Recall that  $\rho_0(x, y, z)$  is independent of  $x$  and that we therefore have introduced the notation  $\rho_0(y, z)$ . Moreover, as noticed above, the measures  $\gamma_z$ , and therefore also  $\sigma_{a,z}$ , are independent of  $y$ . Hence we can choose arbitrary  $x, y$  and let  $\rho_0(z)$  denote  $\rho_0(x, y, z) = \rho_0(y, z)$ . Since all

the estimates below will be independent of this choice of  $y$ , we will use the notation  $\rho_0(z)$  instead of  $\rho_0(x, y, z)$ .

By a change of order of integration we get

$$(2.6) \quad \int_{-1}^1 (\sigma_{a,z}, \sigma_{b,z})_r dz \leq 2r \iint \mathcal{L}_1(\{z : |\rho^{-1}(\cdots c_{-2}c_{-1}a\rho_0(z)) - \rho^{-1}(\cdots d_{-2}d_{-1}b\rho_0(z))| < 2r\}) d\gamma_\Theta(\mathbf{c}) d\gamma_\Theta(\mathbf{d}).$$

We will now apply Lemma 2.2 to (2.6). Note that

$$z \mapsto \rho^{-1}(\cdots c_{-2}c_{-1}a\rho_0(z)), \quad \text{and} \quad z \mapsto \rho^{-1}(\cdots d_{-2}d_{-1}b\rho_0(z))$$

are two curves with tangents in the cones  $C_p$ . Lemma 2.2 states that these curves have a transversal intersection, if they intersect, so that

$$\mathcal{L}_1(\{z : |\rho^{-1}(\cdots c_{-2}c_{-1}a\rho_0(z)) - \rho^{-1}(\cdots d_{-2}d_{-1}b\rho_0(z))| < 2r\}) \leq 4r/C_{\varepsilon,m}.$$

Hence

$$(2.7) \quad \int_{-1}^1 (\sigma_{a,z}, \sigma_{b,z})_r dz \leq \frac{8r^2}{C_{\varepsilon,m}\varepsilon}.$$

By using (2.4) we have

$$(2.8) \quad J(r) = \frac{1}{2^{2m}r^2} \sum_{i=1}^l p_i^2 \sum_{a,b \in A_i} \int_{-1}^1 (\sigma_{a,z}, \sigma_{b,z})_r dz + \frac{1}{2^{2m}r^2} \sum_{i \neq j} p_i p_j \sum_{a \in A_i} \sum_{b \in A_j} \int_{-1}^1 (\sigma_{a,z}, \sigma_{b,z})_r dz.$$

We first give an upper bound for the first summand in (2.8), using (2.5) and an integral transformation. By (2.5) we have

$$\begin{aligned} & \sum_{a,b \in A_i} \int_{-1}^1 (\sigma_{a,z}, \sigma_{b,z})_r dz \\ & \leq (1 + \varepsilon)\lambda_{i,\max} 2^m \sum_{a \in A_i} \int_{-1}^1 (\gamma_{g_{\varepsilon,m}^{-a}}(z), \gamma_{g_{\varepsilon,m}^{-a}}(z)) \frac{r}{(1-\varepsilon)\lambda_{i,\min}} dz \\ & = (1 + \varepsilon)\lambda_{i,\max} 2^m \sum_{k=0}^{2^m-1} 2^m \int_{-1+k2^{-m+1}}^{-1+(k+1)2^{-m+1}} (\gamma_z, \gamma_z) \frac{r}{(1-\varepsilon)\lambda_{i,\min}} dz. \end{aligned}$$

Hence

$$(2.9) \quad \begin{aligned} & \frac{1}{2^{2m}r^2} \sum_{i=1}^l p_i^2 \sum_{a,b \in A_i} \int_{-1}^1 (\sigma_{a,z}, \sigma_{b,z})_r dz \\ & \leq \frac{1}{2^{2m}r^2} \sum_{i=1}^l p_i^2 (1 + \varepsilon)\lambda_{i,\max} 2^m \sum_{k=0}^{2^m-1} 2^m \int_{-1+k2^{-m+1}}^{-1+(k+1)2^{-m+1}} (\gamma_z, \gamma_z) \frac{r}{(1-\varepsilon)\lambda_{i,\min}} dz \end{aligned}$$

$$\begin{aligned}
&\leq \sum_{i=1}^l p_i^2 \frac{(1+\varepsilon)\lambda_{i,\max}}{((1-\varepsilon)\lambda_{i,\min})^2} \frac{1}{\left(\frac{r}{(1-\varepsilon)\lambda_{i,\min}}\right)^2} \int_{-1}^1 (\gamma_z, \gamma_z)_{\frac{r}{(1-\varepsilon)\lambda_{i,\min}}} dz \\
&\leq \max_i J\left(\frac{r}{\lambda_{i,\min}(1-\varepsilon)}\right) \sum_{i=1}^l p_i^2 \frac{(1+\varepsilon)\lambda_{i,\max}}{((1-\varepsilon)\lambda_{i,\min})^2}.
\end{aligned}$$

For the second summand in (2.8), we use (2.7) to prove that it is bounded by

$$\begin{aligned}
(2.10) \quad \frac{1}{2^{2m} r^2} \sum_{i \neq j} p_i p_j \sum_{a \in A_i} \sum_{b \in A_j} \int_{-1}^1 (\sigma_{a,z}, \sigma_{b,z})_r dz \\
\leq \frac{1}{2^{2m} r^2} \sum_{i \neq j} p_i p_j \sum_{a \in A_i} \sum_{b \in A_j} \frac{8r^2}{C_{\varepsilon,m}\varepsilon} \leq \frac{8}{C_{\varepsilon,m}\varepsilon}.
\end{aligned}$$

By combining (2.9) and (2.10) we have

$$(2.11) \quad J(r) \leq \frac{8}{C_{\varepsilon,m}\varepsilon} + \beta \max_i J\left(\frac{r}{\lambda_{i,\min}(1-\varepsilon)}\right)$$

where  $\beta = \sum_{i=1}^l p_i^2 \frac{(1+\varepsilon)\lambda_{i,\max}}{((1-\varepsilon)\lambda_{i,\min})^2}$  is less than 1 by (1.6).

We define recursively a strictly decreasing sequence  $r_k$ . Let  $r_0 < 1/2$  be fixed. Assume that  $r_{k-1}$  has been defined. Then we define  $r_k = (1-\varepsilon)\lambda_{i_k,\min} r_{k-1}$ , where  $i_k$  is chosen such that

$$\max_i J\left(\frac{r_k}{(1-\varepsilon)\lambda_{i,\min}}\right) = J\left(\frac{r_k}{(1-\varepsilon)\lambda_{i_k,\min}}\right) = J(r_{k-1}).$$

Hence  $r_k = r_0(1-\varepsilon)^k \prod_{n=1}^k (\lambda_{i_n,\min})$ .

We note that  $r_k$  is a well defined sequence. By induction and (2.11), we have

$$(2.12) \quad J(r_k) \leq \frac{8}{C_{\varepsilon,m}\varepsilon} \frac{1-\beta^k}{1-\beta} + \beta^k J(r_0)$$

for every  $k \geq 1$ . Hence by (2.1), (2.2) and (2.12) we get

$$\begin{aligned}
(2.13) \quad \|\nu_{\varepsilon,m}\|_2^2 &\leq 4 \liminf_{r \rightarrow 0} J(r) \leq 4 \liminf_{k \rightarrow \infty} J(r_k) \\
&\leq \frac{32}{C_{\varepsilon,m}\varepsilon} \frac{1}{1 - \sum_{i=1}^l p_i^2 \frac{(1+\varepsilon)\lambda_{i,\max}}{((1-\varepsilon)\lambda_{i,\min})^2}}.
\end{aligned}$$

We now use the fact that a closed ball in the Hilbert space  $L^2$  is compact in the weak topology. (See for instance Theorem V.2.1 in Yosida's book [16].) Hence, if  $h_{\nu_{\varepsilon,m}}$  is the density of  $\nu_{\varepsilon,m}$ , then  $h_{\nu_{\varepsilon,m}}$  is in  $L^2$ , and from the above we know that there is a constant  $C'_\varepsilon$  such that  $\|h_{\nu_{\varepsilon,m}}\|_2 \leq C'_\varepsilon/\sqrt{\varepsilon}$ .

By the compactness statement above, there is an  $h$  with  $\|h\|_2 \leq C'_\varepsilon/\sqrt{\varepsilon}$  such that some subsequence of  $h_{\nu_{\varepsilon,m}}$  converges weakly to  $h$ . Moreover  $h$  defines a probability measure since  $1 = \int 1 \cdot h_{\nu_{\varepsilon,m}} d\mathcal{L}_3 \rightarrow \int 1 \cdot h d\mathcal{L}_3$ .

Since  $\nu_{\varepsilon,m}$  converges weakly to  $\nu_\varepsilon$  it follows that  $\nu_\varepsilon$  has density in  $L^2$  and

$$(2.14) \quad \|\nu_\varepsilon\|_2 \leq \frac{1}{\sqrt{\varepsilon}} C'_\varepsilon,$$

where

$$C'_\varepsilon = \sqrt{\frac{32}{(1 - \sum_{i=1}^l p_i^2 \frac{(1+\varepsilon)\lambda_{i,\max}}{((1-\varepsilon)\lambda_{i,\min})^2}) C''_\varepsilon}},$$

$$C''_\varepsilon = \lim_{m \rightarrow \infty} C_{\varepsilon,m} = \min_{i \neq j} \left\{ \frac{|a_i - a_j| + \varepsilon(-|a_i + a_j| - 2)}{1 - \varepsilon^2} \right\}. \blacksquare$$

**3. Proof of Theorem 1.5.** We do not give the whole proof of Theorem 1.5, because it is similar to the proof of Theorem 1.2. We only prove a modification of Lemma 2.2, which is important as it proves transversality.

First we define a new dynamical system  $\tilde{g}_{\varepsilon,m}: Q \rightarrow Q$ , similar to the dynamical system  $g_{\varepsilon,m}: Q \rightarrow Q$ . Let  $Q_{i,k}$  and  $A_{i,k}$  be as in Section 2. Let  $\tilde{g}_{\varepsilon,m}: Q \rightarrow Q$  be defined by

$$\tilde{g}_{\varepsilon,m}: (x, y, z) \mapsto \left( \tilde{d}(z)x + a_i(1 - \tilde{d}(z)), \frac{1}{p_i}y + b(y), 2^m z + c(z) \right)$$

for  $(x, y, z) \in Q_i$ , where

$$\tilde{d}(z) = \lambda_i + 2^m \varepsilon (z - (-1 + (k + 1/2)2^{-m+1})) \quad \text{for } (x, y, z) \in Q_{i,k},$$

$$b(y) = 1 - \frac{1}{p_i} \left( -1 + 2 \sum_{j=1}^i p_j \right) \quad \text{for } (x, y, z) \in Q_{i,k},$$

$$c(z) = 2^m - 2k - 1 \quad \text{for } (x, y, z) \in Q_{i,k}.$$

Hence the only difference between  $\tilde{g}_{\varepsilon,m}$  and  $g_{\varepsilon,m}$  is in the first coordinate, where the perturbation of  $f_i$  is made. Figure 1 also serves to visualise the action of  $\tilde{g}_{\varepsilon,m}$ .

We define the cones

$$C_p = \left\{ (u, v, w) \in T_p Q : \left| \frac{u}{w} \right|, \left| \frac{v}{w} \right| < \frac{2^{m+1}\varepsilon}{2^m - \lambda_{\max} - \varepsilon} \right\},$$

where  $p \in Q$  and  $\lambda_{\max} = \max_i \lambda_i$ . Similar to Lemma 2.2, we show that these cones define a family of unstable cones, and that a certain transversality property holds.

LEMMA 3.1. *Suppose that (1.9) holds. The cones  $C_p$  form a family of unstable cones, that is,  $d_p \tilde{g}_{\varepsilon,m}(C_p) \subset C_{\tilde{g}_{\varepsilon,m}(p)}$ .*

Moreover, for sufficiently large  $m$  and every sufficiently small  $\varepsilon > 0$ , if  $\zeta_1 \in Q_{\xi_1}$  and  $\zeta_2 \in Q_{\xi_2}$  are two line segments with tangents in  $C_p$  such that  $\xi_1 \in A_i$  and  $\xi_2 \in A_j$ ,  $i \neq j$ , then if  $\tilde{g}_{\varepsilon,m}(\zeta_1)$  and  $\tilde{g}_{\varepsilon,m}(\zeta_2)$  intersect, and if  $(u_1, v_1, 1)$  and  $(u_2, v_2, 1)$  are tangents to  $\tilde{g}_{\varepsilon,m}(\zeta_1)$  and  $\tilde{g}_{\varepsilon,m}(\zeta_2)$  respectively, there exists a constant  $C_{\varepsilon,m}$ , depending on  $\varepsilon$  and  $m$ , but bounded away from 0 and infinity, such that  $|u_1 - u_2| > C_{\varepsilon,m}\varepsilon$ .

*Proof.* The Jacobian of  $\tilde{g}_{\varepsilon,m}$  is

$$d_p \tilde{g}_{\varepsilon,m} = \begin{pmatrix} \tilde{d}(z) & 0 & 2^m \varepsilon (x - a_i) \\ 0 & 1/p_i & 0 \\ 0 & 0 & 2^m \end{pmatrix},$$

where  $p = (x, y, z) \in Q_{i,k}$ . If  $(u, v, w) \in C_p$ , then

$$d_p \tilde{g}_{\varepsilon,m}(u, v, w) = \begin{pmatrix} \tilde{d}(z)u + 2^m \varepsilon (x - a_i)w \\ (1/p_i)v \\ 2^m w \end{pmatrix}.$$

The estimate

$$\begin{aligned} \frac{|\tilde{d}(z)u + 2^m \varepsilon (x - a_i)w|}{|2^m w|} &\leq \frac{\tilde{d}(z)|u|}{2^m |w|} + 2\varepsilon \\ &\leq \frac{\lambda_i + \varepsilon}{2^m} \frac{2^{m+1}\varepsilon}{2^m - \lambda_{\max} - \varepsilon} + 2\varepsilon \leq \frac{2^{m+1}\varepsilon}{2^m - \lambda_{\max} - \varepsilon} \end{aligned}$$

shows that  $d_p \tilde{g}_{\varepsilon,m}(C_p) \subset C_{\tilde{g}_{\varepsilon,m}(p)}$ . Now we prove the other statement of the lemma. Assume that  $p = (x_p, y_p, z_p) \in Q_i$  and  $q = (x_q, y_q, z_q) \in Q_j$ ,  $i \neq j$ , are such that  $\tilde{g}_{\varepsilon,m}(p) = \tilde{g}_{\varepsilon,m}(q) = (x, y, z)$ . Then

$$p \in Q_i \Rightarrow d_p \tilde{g}_{\varepsilon,m} : (u, v, 1) \mapsto 2^m \left( \frac{\tilde{d}(z_p)}{2^m} u + (x_p - a_i)\varepsilon, \frac{v}{p_i}, 1 \right),$$

and

$$x_p = \frac{x - a_i(1 - \tilde{d}(z_p))}{\tilde{d}(z_p)}, \quad x_q = \frac{x - a_j(1 - \tilde{d}(z_q))}{\tilde{d}(z_q)}.$$

Let  $\tilde{\Delta}_i = \frac{2(\lambda_i + \varepsilon)}{2^m - \lambda_{\max} - \varepsilon}$ . Then

$$d_p \tilde{g}_{\varepsilon,m}(C_p) \subset \left\{ w(u, v, 1) : \frac{x - a_i}{\tilde{d}(z_p)} \varepsilon - \tilde{\Delta}_i \varepsilon \leq u \leq \frac{x - a_i}{\tilde{d}(z_p)} \varepsilon + \tilde{\Delta}_i \varepsilon \right\}.$$

Therefore

$$|u_2 - u_1| \geq \left( \left| \frac{x - a_i}{\tilde{d}(z_p)} - \frac{x - a_j}{\tilde{d}(z_q)} \right| - (\tilde{\Delta}_i + \tilde{\Delta}_j) \right) \varepsilon.$$

The term

$$\left| \frac{x - a_i}{\tilde{d}(z_p)} - \frac{x - a_j}{\tilde{d}(z_q)} \right|$$

can be estimated by

$$\left| \frac{x - a_i}{\tilde{d}(z_p)} - \frac{x - a_j}{\tilde{d}(z_q)} \right| \geq \left| \frac{|\tilde{d}(z_p) - \tilde{d}(z_q)| |x| - |a_j \tilde{d}(z_p) - a_i \tilde{d}(z_q)|}{\tilde{d}(z_p) \tilde{d}(z_q)} \right|.$$

Hence, this term is positive provided that

$$|a_j \tilde{d}(z_p) - a_i \tilde{d}(z_q)| > |\tilde{d}(z_p) - \tilde{d}(z_q)|.$$

Since  $\lambda_i - \varepsilon \leq \tilde{d}(z_p) \leq \lambda_i + \varepsilon$  and  $\lambda_j - \varepsilon \leq \tilde{d}(z_q) \leq \lambda_j + \varepsilon$ , this is implied by (1.9) if  $\varepsilon$  is sufficiently small.

If we let

$$C_{\varepsilon, m} = \frac{1}{2} \min_{i \neq j} \frac{|a_i \lambda_j - a_j \lambda_i| - |\lambda_i - \lambda_j|}{\lambda_i \lambda_j},$$

then

$$|u_2 - u_1| \geq C_{\varepsilon, m} \varepsilon,$$

provided that  $\varepsilon$  is small and  $m$  large.

In fact we can let

$$C_{\varepsilon, m} = \sigma \min_{i \neq j} \frac{|a_i \lambda_j - a_j \lambda_i| - |\lambda_i - \lambda_j|}{\lambda_i \lambda_j}$$

for some  $0 < \sigma < 1$ . ■

**Acknowledgments.** Research of Bárány was supported by the EU FP6 Research Training Network CODY.

### References

- [1] B. Bárány, M. Pollicott and K. Simon, *Stationary measures for projective transformations: the Blackwell and Furstenberg measures*, preprint, 2009.
- [2] P. Diaconis and D. A. Freedman, *Iterated random functions*, SIAM Rev. 41 (1999), 45–76.
- [3] A. H. Fan, K. Simon and H. Tóth, *Contracting on average random IFS with repelling fixed point*, J. Statist. Phys. 122 (2006), 169–193.
- [4] K.-S. Lau, S.-M. Ngai and H. Rao, *Iterated function systems with overlaps and self-similar measures*, J. London Math. Soc. (2) 63 (2001), 99–116.
- [5] J. Neunhuserer, *Properties of some overlapping self-similar and some self-affine measures*, Acta Math. Hungar. 92 (2001), 143–161.
- [6] S.-M. Ngai and Y. Wang, *Self-similar measures associated with IFS with non-uniform contraction ratios*, Asian J. Math. 9 (2005), 227–244.
- [7] M. Nicol, N. Sidorov and D. Broomhead, *On the fine structure of stationary measures in systems which contract on average*, J. Theoret. Probab. 15 (2002), 715–730.
- [8] Y. Peres and W. Schlag, *Smoothness of projections, Bernoulli convolutions, and the dimension of exceptions*, Duke Math. J. 102 (2000), 193–251.

- [9] Y. Peres, K. Simon and B. Solomyak, *Absolute continuity for random iterated function systems with overlaps*, J. London Math. Soc. (2) 74 (2006), 739–756.
- [10] Y. Peres and B. Solomyak, *Self-similar measures and intersections of Cantor sets*, Trans. Amer. Math. Soc. 350 (1998), 4065–4087.
- [11] T. Persson, *On random Bernoulli convolutions*, Dynam. Systems, to appear.
- [12] Ya. Pesin, *Dynamical systems with generalized hyperbolic attractors: hyperbolic, ergodic and topological properties*, Ergodic Theory Dynam. Systems 12 (1992), 123–151.
- [13] J. Schmeling and S. Troubetzkoy, *Dimension and invertibility of hyperbolic endomorphisms with singularities*, Ergodic Theory Dynam. Systems 18 (1998), 1257–1282.
- [14] K. Simon, B. Solomyak and M. Urbański, *Invariant measures for parabolic IFS with overlaps and random continued fractions*, Trans. Amer. Math. Soc. 353 (2001), 5145–5164.
- [15] M. Tsujii, *Fat solenoidal attractor*, Nonlinearity 14 (2001), 1011–1027.
- [16] K. Yosida, *Functional Analysis*, Springer, Berlin, 1980.

Balázs Bárány  
Department of Stochastics  
Institute of Mathematics  
Technical University of Budapest  
P.O. Box 91  
1521 Budapest, Hungary  
E-mail: balubsheep@gmail.com

Tomas Persson  
Institute of Mathematics  
Polish Academy of Sciences  
Śniadeckich 8  
P.O. Box 21  
00-956 Warszawa, Poland  
E-mail: tomasp@impan.pl

*Received 24 February 2009;  
in revised form 28 June 2010*