

Position dependent random maps in one and higher dimensions

by

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Abstract. A random map is a discrete-time dynamical system in which one of a number of transformations is randomly selected and applied on each iteration of the process. We study random maps with position dependent probabilities on the interval and on a bounded domain of \mathbb{R}^n . Sufficient conditions for the existence of an absolutely continuous invariant measure for a random map with position dependent probabilities on the interval and on a bounded domain of \mathbb{R}^n are the main results.

1. Introduction. Let τ_1, \dots, τ_K be a collection of transformations from X to X . Usually, the random map T is defined by choosing τ_k with constant probability p_k , $p_k > 0$, $\sum_{k=1}^K p_k = 1$. The ergodic theory of such dynamical systems was studied in [9] and in [8] (see also [7]).

There is a rich literature on random maps with position dependent probabilities with τ_1, \dots, τ_K being continuous contracting transformations (see [10]).

In this paper, we deal with piecewise monotone transformations τ_1, \dots, τ_K and position dependent probabilities $p_k(x)$, $k = 1, \dots, K$, where $p_k(x) > 0$, $\sum_{k=1}^K p_k(x) = 1$, i.e., the p_k 's are functions of position. We point out that studying such dynamical systems was begun in [5], where sufficient conditions for the existence of an absolutely continuous invariant measure were given. The conditions in [5] are applicable only when τ_1, \dots, τ_K are C^2 expanding transformations (see [5] for details). In this paper, we prove the existence of an absolutely continuous invariant measure for a random map T on $[a, b]$ under milder conditions (see Section 4, Conditions (A) and (B)). Moreover, we prove the existence of an absolutely continuous invariant measure for a random map T on a bounded domain of \mathbb{R}^n (see Section 6, Condition (C)).

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The paper is organized in the following way: In Section 2, following the ideas of [5], we formulate the definition of a random map T with position dependent probabilities and introduce its Perron–Frobenius operator. In Section 3, we prove some properties of the Perron–Frobenius operator of T . In Section 4, we prove the existence of an absolutely continuous invariant measure for T on $[a, b]$. In Section 5, we give an example of a random map T which does not satisfy the conditions of [5]; yet, it preserves an absolutely continuous invariant measure under conditions (A) and (B). In Section 6, we prove the existence of an absolutely continuous invariant measure for T on a bounded domain of \mathbb{R}^n . In Section 7, we give an example of a random map in \mathbb{R}^n that preserves an absolutely continuous invariant measure.

2. Preliminaries. Let $(X, \mathfrak{B}, \lambda)$ be a measure space, where λ is an underlying measure. Let $\tau_k : X \rightarrow X, k = 1, \dots, K$, be piecewise one-to-one, non-singular transformations on a common partition \mathcal{P} of X : $\mathcal{P} = \{I_1, \dots, I_q\}$ and $\tau_{k,i} = \tau_k|_{I_i}, i = 1, \dots, q, k = 1, \dots, K$ (\mathcal{P} can be found by considering finer partitions). We define the *transition function* for the random map $T = \{\tau_1, \dots, \tau_K; p_1(x), \dots, p_K(x)\}$ as follows:

$$\mathbb{P}(x, A) = \sum_{k=1}^K p_k(x) \chi_A(\tau_k(x)),$$

where A is any measurable set and $\{p_k(x)\}_{k=1}^K$ is a set of position dependent measurable probabilities, i.e., $\sum_{k=1}^K p_k(x) = 1, p_k(x) \geq 0$ for any $x \in X$, and χ_A denotes the characteristic function of the set A . We define $T(x) = \tau_k(x)$ with probability $p_k(x)$ and $T^N(x) = \tau_{k_N} \circ \tau_{k_{N-1}} \circ \dots \circ \tau_{k_1}(x)$ with probability $p_{k_N}(\tau_{k_{N-1}} \circ \dots \circ \tau_{k_1}(x)) \cdot p_{k_{N-1}}(\tau_{k_{N-2}} \circ \dots \circ \tau_{k_1}(x)) \cdot \dots \cdot p_{k_1}(x)$. The transition function \mathbb{P} induces an operator \mathbb{P}_* on measures on (X, \mathfrak{B}) defined by

$$\begin{aligned} \mathbb{P}_* \mu(A) &= \int \mathbb{P}(x, A) d\mu(x) = \sum_{k=1}^K \int p_k(x) \chi_A(\tau_k(x)) d\mu(x) \\ &= \sum_{k=1}^K \int_{\tau_k^{-1}(A)} p_k(x) d\mu(x) = \sum_{k=1}^K \sum_{i=1}^q \int_{\tau_{k,i}^{-1}(A)} p_k(x) d\mu(x). \end{aligned}$$

We say that the measure μ is T -invariant iff $\mathbb{P}_* \mu = \mu$, i.e.,

$$\mu(A) = \sum_{k=1}^K \int_{\tau_k^{-1}(A)} p_k(x) d\mu(x), \quad A \in \mathfrak{B}.$$

If μ has density f with respect to λ , then $\mathbb{P}_* \mu$ also has a density which we denote by $P_T f$. By change of variables, we obtain

$$\begin{aligned} \int_A P_T f(x) d\lambda(x) &= \sum_{k=1}^K \sum_{i=1}^q \int_{\tau_{k,i}^{-1}(A)} p_k(x) f(x) d\lambda(x) \\ &= \sum_{k=1}^K \sum_{i=1}^q \int_A p_k(\tau_{k,i}^{-1}x) f(\tau_{k,i}^{-1}x) \frac{1}{J_{k,i}(\tau_{k,i}^{-1})} d\lambda(x), \end{aligned}$$

where $J_{k,i}$ is the Jacobian of $\tau_{k,i}$ with respect to λ . Since this holds for any measurable set A we obtain an a.e. equality

$$(P_T f)(x) = \sum_{k=1}^K \sum_{i=1}^q p_k(\tau_{k,i}^{-1}x) f(\tau_{k,i}^{-1}x) \frac{1}{J_{k,i}(\tau_{k,i}^{-1})} \chi_{\tau_k(I_i)}(x)$$

or

$$(P_T f)(x) = \sum_{k=1}^K P_{\tau_k}(p_k f)(x),$$

where P_{τ_k} is the Perron–Frobenius operator corresponding to the transformation τ_k (see [1] for details). We call P_T the *Perron–Frobenius operator* of the random map T . It is main tool in this paper, with very useful properties.

3. Properties of the Perron–Frobenius operator of T . The properties of P_T resemble the properties of the classical Perron–Frobenius operator of a single transformation.

LEMMA 3.1. P_T has the following properties:

- (i) P_T is linear;
- (ii) P_T is non-negative; i.e., $f \geq 0 \Rightarrow P_T f \geq 0$;
- (iii) $P_T f = f \Leftrightarrow \mu = f \cdot \lambda$ is T -invariant;
- (iv) $\|P_T f\|_1 \leq \|f\|_1$, where $\|\cdot\|_1$ denotes the L^1 norm;
- (v) $P_{T \circ R} = P_R \circ P_T$. In particular, $P_T^N f = P_{T^N} f$.

Proof. The proofs of (i)–(iv) are analogous to those for a single transformation. For the proof of (v), let T and R be two random maps corresponding to $\{\tau_1, \dots, \tau_K; p_1, \dots, p_K\}$ and $\{\zeta_1, \dots, \zeta_L; r_1, \dots, r_L\}$ respectively. We define $\{\tau_k\}_{k=1}^K$ and $\{\zeta_l\}_{l=1}^L$ on a common partition \mathcal{P} . We have

$$\begin{aligned} P_R(P_T f) &= P_R\left(\sum_{k=1}^K P_{\tau_k}(p_k f)\right) \\ &= \sum_{l=1}^L \sum_{k=1}^K P_{\zeta_l}(r_l P_{\tau_k}(p_k f)) \\ &= \sum_{l=1}^L \sum_{k=1}^K \sum_{i=1}^q r_l(\zeta_{l,i}^{-1}) [P_{\tau_k}(p_k f)](\zeta_{l,i}^{-1}) \frac{1}{J_{\zeta_{l,i}}(\zeta_{l,i}^{-1})} \chi_{\zeta_{l,i}(I_i)} \end{aligned}$$

$$\begin{aligned}
 &= \sum_{k=1}^K \sum_{l=1}^L \sum_{j=1}^q \sum_{i=1}^q r_l(\zeta_{l,i}^{-1}) p_k(\tau_{k,j}^{-1} \circ \zeta_{l,i}^{-1}) f(\tau_{k,j}^{-1} \circ \zeta_{l,i}^{-1}) \\
 &\quad \times \frac{1}{J_{\tau_{k,j}^{-1}}(\tau_{k,j}^{-1} \circ \zeta_{l,i}^{-1})} \frac{1}{J_{\zeta_{l,i}^{-1}}(\zeta_{l,i}^{-1})} \chi_{\tau_k(I_j)}(\zeta_{l,i}^{-1}) \chi_{\zeta_{l,i}(I_i)} \\
 &= \sum_{k=1}^K \sum_{l=1}^L P_{\tau_k \circ \zeta_l} (p_k(\zeta_l) r_l f) = P_{T \circ R} f. \blacksquare
 \end{aligned}$$

4. The existence of an absolutely continuous invariant measure on $[a, b]$. Let $(I, \mathfrak{B}, \lambda)$ be a measure space, where λ is normalized Lebesgue measure on $I = [a, b]$. Let $\tau_k : I \rightarrow I, k = 1, \dots, K$, be piecewise one-to-one and differentiable, non-singular transformations on a partition \mathcal{P} of I : $\mathcal{P} = \{I_1, \dots, I_q\}$ and $\tau_{k,i} = \tau_k|_{I_i}, i = 1, \dots, q, k = 1, \dots, K$. Denote by $V(\cdot)$ the standard one-dimensional variation of a function, and by $\text{BV}(I)$ the space of functions of bounded variation on I equipped with the norm $\|\cdot\|_{\text{BV}} = V(\cdot) + \|\cdot\|_1$.

Let $g_k(x) = p_k(x)/|\tau'_k(x)|, k = 1, \dots, K$. We assume the following conditions:

CONDITION (A). $\sum_{k=1}^K g_k(x) < \alpha < 1, x \in I$.

CONDITION (B). $g_k \in \text{BV}(I), k = 1, \dots, K$.

Under the above conditions our goal is to prove

$$(4.1) \quad V_I P_T^n f \leq AV_I f + B\|f\|_1$$

for some $n \geq 1$, where $0 < A < 1$ and $B > 0$. The inequality (4.1) guarantees the existence of a T -invariant measure absolutely continuous with respect to Lebesgue measure and the quasi-compactness of the operator P_T with all the consequences of this fact (see [1]). We will need a number of lemmas:

LEMMA 4.1. *Let $f \in \text{BV}(I)$. Suppose $\tau : I \rightarrow J$ is differentiable and $\tau'(x) \neq 0, x \in I$. Set $\phi = \tau^{-1}$ and let $g(x) = p(x)/|\tau'(x)| \in \text{BV}(I)$. Then*

$$V_J(f(\phi)g(\phi)) \leq (V_I f + \sup_I f)(V_I g + \sup_I g).$$

Proof. First, note that we have dropped all the k, i indices to simplify the notation. The proof follows in the same way as in Lemma 3 of [9]. \blacksquare

LEMMA 4.2. *Let T satisfy conditions (A) and (B). Then for any $f \in \text{BV}(I)$,*

$$V_I P_T f \leq AV_I f + B\|f\|_1,$$

where

$$A = 3\alpha + \max_{1 \leq i \leq q} \sum_{k=1}^K V_{I_i} g_k, \quad B = 2\beta\alpha + \beta \max_{1 \leq i \leq q} \sum_{k=1}^K V_{I_i} g_k,$$

with $\beta = \max_{1 \leq i \leq q} (\lambda(I_i))^{-1}$.

Proof. First, we will refine the partition \mathcal{P} to satisfy an additional condition. Let $\eta > 0$ be such that $\sum_{k=1}^K (g_k(x) + \varepsilon_k) < \alpha$ whenever $|\varepsilon_k| < \eta$, $k = 1, \dots, K$. Since g_k , $k = 1, \dots, K$, are of bounded variation we can find a finite partition \mathcal{K} such that for any $k = 1, \dots, K$,

$$|g_k(x) - g_k(y)| < \eta$$

for x, y in the same element of \mathcal{K} . Without loss of generality, we can assume that our original partition is the join $\mathcal{P} \vee \mathcal{K}$. Then

$$\max_{1 \leq i \leq q} \sum_{k=1}^K \sup_{x \in I_i} g_k(x) < \alpha.$$

We have $V_I(P_T f) = V_I(\sum_{k=1}^K P_{\tau_k}(p_k f))$. We will estimate this variation. Let $\phi_{k,i} = \tau_{k,i}^{-1}$, $k = 1, \dots, K$, $i = 1, \dots, q$. We have

$$\begin{aligned} (4.2) \quad V_I \left(\sum_{k=1}^K P_{\tau_k}(p_k f) \right) &= V_I \left(\sum_{k=1}^K \sum_{i=1}^q f(\phi_{k,i}) g_k(\phi_{k,i}) \chi_{\tau_k(I_i)} \right) \\ &\leq \sum_{k=1}^K \sum_{i=1}^q [|f(a_{i-1})| |g_k(a_{i-1})| + |f(a_i)| |g_k(a_i)|] \\ &\quad + \sum_{k=1}^K \sum_{i=1}^q V_{\tau_k(I_i)} [f(\phi_{k,i}) g_k(\phi_{k,i})]. \end{aligned}$$

First, we estimate the first sum on the right hand side of (4.2):

$$\begin{aligned} (4.3) \quad &\sum_{k=1}^K \sum_{i=1}^q [|f(a_{i-1})| |g_k(a_{i-1})| + |f(a_i)| |g_k(a_i)|] \\ &= \sum_{i=1}^q \left[|f(a_{i-1})| \left(\sum_{k=1}^K |g_k(a_{i-1})| \right) + |f(a_i)| \left(\sum_{k=1}^K |g_k(a_i)| \right) \right] \\ &\leq \alpha \left(\sum_{i=1}^q (|f(a_{i-1})| + |f(a_i)|) \right) \\ &\leq \alpha \left(\sum_{i=1}^q \left(V_{I_i} f + (\lambda(I_i))^{-1} \int_{I_i} f d\lambda \right) \right) = \alpha (V_I f + \beta \|f\|_1). \end{aligned}$$

We now estimate the second term on the right hand side of (4.2). Using Lemma 4.1 we obtain

$$\begin{aligned}
 (4.4) \quad & \sum_{k=1}^K \sum_{i=1}^q V_{\tau_k(I_i)} [f(\phi_{k,i})g_k(\phi_{k,i})] \\
 & \leq \sum_{k=1}^K \sum_{i=1}^q (V_{I_i}f + \sup_{I_i} f)(V_{I_i}g_k + \sup_{I_i} g_k) \\
 & \leq \sum_{i=1}^q \left(2V_{I_i}f + \beta \int_{I_i} f d\lambda \right) \left(\max_{1 \leq i \leq q} \sum_{k=1}^K (V_{I_i}g_k + \sup_{I_i} g_k) \right) \\
 & \leq (2V_I f + \beta \|f\|_1) \left(\max_{1 \leq i \leq q} \sum_{k=1}^K V_{I_i}g_k + \alpha \right).
 \end{aligned}$$

Thus, using (4.3) and (4.4), we obtain

$$V_I P_T f \leq \left(3\alpha + \max_{1 \leq i \leq q} \sum_{k=1}^K V_{I_i}g_k \right) V_I f + \left(2\beta\alpha + \beta \max_{1 \leq i \leq q} \sum_{k=1}^K V_{I_i}g_k \right) \|f\|_1. \quad \blacksquare$$

In the following two lemmas we show that the constants α as well as $\max_{1 \leq i \leq q} \sum_{k=1}^K V_{I_i}g_k$ decrease when we consider higher iterations T^n instead of T . The constant β obviously increases, but this is not important.

LEMMA 4.3. *Let T be a random map which satisfies condition (A). Then, for $x \in I$,*

$$(4.5) \quad \sum_{w \in \{1, \dots, K\}^N} \frac{p_w(x)}{|T'_w(x)|} < \alpha^N,$$

where $T_w(x) = \tau_{k_N} \circ \tau_{k_{N-1}} \circ \dots \circ \tau_{k_1}(x)$ and $p_w(x) = p_{k_N}(\tau_{k_{N-1}} \circ \dots \circ \tau_{k_1}(x)) \cdot p_{k_{N-1}}(\tau_{k_{N-2}} \circ \dots \circ \tau_{k_1}(x)) \cdots p_{k_1}(x)$ define the random map T^N .

Proof. We have

$$T^N(x) = \tau_{k_N} \circ \tau_{k_{N-1}} \circ \dots \circ \tau_{k_1}(x)$$

with probability

$$p_{k_N}(\tau_{k_{N-1}} \circ \dots \circ \tau_{k_1}(x)) \cdot p_{k_{N-1}}(\tau_{k_{N-2}} \circ \dots \circ \tau_{k_1}(x)) \cdots p_{k_1}(x).$$

The maps defining T^N may be indexed by $w \in \{1, \dots, K\}^N$. Set

$$T_w(x) = \tau_{k_N} \circ \tau_{k_{N-1}} \circ \dots \circ \tau_{k_1}(x),$$

where $w = (k_1, \dots, k_N)$, and

$$p_w(x) = p_{k_N}(\tau_{k_{N-1}} \circ \dots \circ \tau_{k_1}(x)) \cdot p_{k_{N-1}}(\tau_{k_{N-2}} \circ \dots \circ \tau_{k_1}(x)) \cdots p_{k_1}(x).$$

Then

$$T'_w(x) = \tau'_{k_N}(\tau_{k_{N-1}} \circ \dots \circ \tau_{k_1}(x)) \tau'_{k_{N-1}}(\tau_{k_{N-2}} \circ \dots \circ \tau_{k_1}(x)) \cdots \tau'_{k_1}(x).$$

Suppose that T satisfies condition (A). We will prove (4.5) using induction on N . For $N = 1$, we have

$$\sum_{w \in \{1, \dots, K\}} \frac{p_w(x)}{|T'_w(x)|} < \alpha$$

by condition (A). Assume (4.5) is true for $N - 1$. Then

$$\begin{aligned} \sum_{w \in \{1, \dots, K\}^N} \frac{p_w(x)}{|T'_w(x)|} &= \sum_{\bar{w} \in \{1, \dots, K\}^{N-1}} \sum_{k=1}^K \frac{p_k(x)p_{\bar{w}}(\tau_k(x))}{|\tau'_k(x)||T'_{\bar{w}}(\tau_k(x))|} \\ &\leq \left(\sum_{k=1}^K \frac{p_k(x)}{|\tau'_k(x)|} \right) \left(\sum_{\bar{w} \in \{1, \dots, K\}^{N-1}} \frac{p_{\bar{w}}(\tau_k(x))}{|T'_{\bar{w}}(\tau_k(x))|} \right) < \alpha \cdot \alpha^{N-1} = \alpha^N. \blacksquare \end{aligned}$$

LEMMA 4.4. Let $g_w = p_w/|T'_w|$, where T_w and p_w are defined in Lemma 4.3 and $w \in \{1, \dots, K\}^n$. Define

$$W_1 \equiv \max_{1 \leq i \leq q} \sum_{k=1}^K V_{I_i} g_k, \quad W_n \equiv \max_{J \in \mathcal{P}^{(n)}} \sum_{w \in \{1, \dots, K\}^n} V_J g_w,$$

where $\mathcal{P}^{(n)}$ is the common monotonicity partition for all T_w . Then, for all $n \geq 1$,

$$W_n \leq n\alpha^{n-1}W_1,$$

where α is defined in condition (A).

Proof. We prove the lemma by induction on n . For $n = 1$ the assertion is true by definition of W_n . Assume that it is true for n , i.e.,

$$W_n \leq n\alpha^{n-1}W_1.$$

Let $J \in \mathcal{P}^{(n+1)}$ and $x_0 < x_1 < \dots < x_l$ be a sequence of points in J . Then

$$\begin{aligned} \sum_w \sum_{j=0}^{l-1} |g_w(x_{j+1}) - g_w(x_j)| &= \sum_{j=0}^{l-1} \sum_{w \in \{1, \dots, K\}^{n+1}} |g_w(x_{j+1}) - g_w(x_j)| \\ &\leq \sum_{j=0}^{l-1} \sum_{\bar{w} \in \{1, \dots, K\}^n} \sum_{k=1}^K |g_{\bar{w}}(\tau_k(x_{j+1}))g_k(x_{j+1}) - g_{\bar{w}}(\tau_k(x_j))g_k(x_j)| \\ &\leq \sum_{j=0}^{l-1} \sum_{\bar{w} \in \{1, \dots, K\}^n} \sum_{k=1}^K |g_{\bar{w}}(\tau_k(x_{j+1}))g_k(x_{j+1}) - g_{\bar{w}}(\tau_k(x_{j+1}))g_k(x_j)| \\ &\quad + \sum_{j=0}^{l-1} \sum_{\bar{w} \in \{1, \dots, K\}^n} \sum_{k=1}^K |g_{\bar{w}}(\tau_k(x_{j+1}))g_k(x_j) - g_{\bar{w}}(\tau_k(x_j))g_k(x_j)| \end{aligned}$$

$$\begin{aligned}
 &\leq \sum_{j=0}^{l-1} \sum_{k=1}^K |g_k(x_{j+1}) - g_k(x_j)| \sum_{\bar{w} \in \{1, \dots, K\}^n} g_{\bar{w}}(\tau_k(x_{j+1})) \\
 &\quad + \sum_{j=0}^{l-1} \sum_{k=1}^K g_k(x_j) \sum_{\bar{w} \in \{1, \dots, K\}^n} |g_{\bar{w}}(\tau_k(x_{j+1})) - g_{\bar{w}}(\tau_k(x_j))| \\
 &\leq \alpha^n \sum_{j=0}^{l-1} \sum_{k=1}^K |g_k(x_{j+1}) - g_k(x_j)| \\
 &\quad + \alpha \sum_{j=0}^{l-1} \sum_{\bar{w} \in \{1, \dots, K\}^n} |g_{\bar{w}}(\tau_k(x_{j+1})) - g_{\bar{w}}(\tau_k(x_j))| \\
 &\leq \alpha^n W_1 + \alpha W_n \leq \alpha^n W_1 + n\alpha^n W_1 = (n + 1)\alpha^n W_1.
 \end{aligned}$$

We used condition (A) and Lemma 4.3. ■

THEOREM 4.5. *Let T be a random map which satisfies conditions (A) and (B). Then T preserves a measure which is absolutely continuous with respect to Lebesgue measure. The operator P_T is quasi-compact on $BV(I)$ (see [1]).*

Proof. Let N be such that $A_N = 3\alpha^N + W_N < 1$. Then, by Lemma 4.3,

$$\sum_{w \in \{1, \dots, K\}^N} g_w(x) < \alpha^N, \quad x \in I.$$

We refine the partition $\mathcal{P}^{(N)}$ as in the proof of Lemma 4.2, to have

$$\max_{J \in \mathcal{P}^N} \sum_{w \in \{1, \dots, K\}^N} \sup_J g_w < \alpha^N.$$

Then, by Lemma 4.2, we get

$$\|P_T^N f\|_{BV} \leq A_N \|f\|_{BV} + B_N \|f\|_1,$$

where $B_N = \beta_N(2\alpha^N + W_N)$, $\beta_N = \max_{J \in \mathcal{P}^N} (\lambda(J))^{-1}$. The theorem follows by the standard technique (see [1]). ■

REMARK 4.6. It is enough to assume that condition (A) is satisfied for some iterate T^m , $m \geq 1$.

REMARK 4.7. The number of absolutely continuous invariant measures for random maps has been studied in [4]. The proof of [4], which uses graph theoretic methods, goes through analogously in our case, i.e., when T is a random map with position dependent probabilities.

5. Example. We present an example of a random map T which does not satisfy the conditions of [5], yet it preserves an absolutely continuous invariant measure under conditions (A) and (B).

EXAMPLE 5.1. Let T be a random map which is given by $\{\tau_1, \tau_2; p_1(x), p_2(x)\}$, where

$$\begin{aligned} \tau_1(x) &= \begin{cases} 2x & \text{for } 0 \leq x \leq 1/2, \\ x & \text{for } 1/2 < x \leq 1, \end{cases} & p_1(x) &= \begin{cases} 2/3 & \text{for } 0 \leq x \leq 1/2, \\ 1/3 & \text{for } 1/2 < x \leq 1, \end{cases} \\ \tau_2(x) &= \begin{cases} x + 1/2 & \text{for } 0 \leq x \leq 1/2, \\ 2x - 1 & \text{for } 1/2 < x \leq 1, \end{cases} & p_2(x) &= \begin{cases} 1/3 & \text{for } 0 \leq x \leq 1/2, \\ 2/3 & \text{for } 1/2 < x \leq 1. \end{cases} \end{aligned}$$

Then $\sum_{k=1}^2 g_k(x) = 2/3 < 1$. Therefore, T satisfies conditions (A) and (B). Consequently, by Theorem 4.5, T preserves an invariant measure absolutely continuous with respect to Lebesgue measure. Notice that τ_1, τ_2 are piecewise linear Markov maps defined on the same Markov partition $\mathcal{P}: \{[0, 1/2], [1/2, 1]\}$. For such maps the Perron–Frobenius operator reduces to a matrix (see [1]). The corresponding matrices are:

$$P_{\tau_1} = \begin{pmatrix} 1/2 & 1/2 \\ 0 & 1 \end{pmatrix}, \quad P_{\tau_2} = \begin{pmatrix} 0 & 1 \\ 1/2 & 1/2 \end{pmatrix}.$$

Their invariant densities are $f_{\tau_1} = [0, 2]$ and $f_{\tau_2} = [2/3, 4/3]$. The Perron–Frobenius operator of the random map T is given by

$$P_T = \begin{pmatrix} 2/3 & 0 \\ 0 & 1/3 \end{pmatrix} \begin{pmatrix} 1/2 & 1/2 \\ 0 & 1 \end{pmatrix} + \begin{pmatrix} 1/3 & 0 \\ 0 & 2/3 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1/2 & 1/2 \end{pmatrix} = \begin{pmatrix} 1/3 & 2/3 \\ 1/3 & 2/3 \end{pmatrix}.$$

If the invariant density of T is $f = [f_1, f_2]$, normalized by $f_1 + f_2 = 2$ and satisfying the equation $fP_T = f$, then $f_1 = 2/3$ and $f_2 = 4/3$.

6. The existence of an absolutely continuous invariant measure in \mathbb{R}^n . Let S be a bounded region in \mathbb{R}^n and λ_n be Lebesgue measure on S . Let $\tau_k : S \rightarrow S$, $k = 1, \dots, K$, be piecewise one-to-one and C^2 , non-singular transformations on a partition \mathcal{P} of S , $\mathcal{P} = \{S_1, \dots, S_q\}$ and $\tau_{k,i} = \tau_k|_{S_i}$, $i = 1, \dots, q$, $k = 1, \dots, K$. Suppose each S_i is a bounded closed domain having a piecewise C^2 boundary of finite $(n - 1)$ -dimensional measure. We assume that the faces of ∂S_i meet at angles bounded uniformly away from 0. We will also assume that the probabilities $p_k(x)$ are piecewise C^1 functions on the partition \mathcal{P} . Let $D\tau_{k,i}^{-1}(x)$ be the derivative matrix of $\tau_{k,i}^{-1}$ at x . We assume:

CONDITION (C).

$$\max_{1 \leq i \leq q} \sum_{k=1}^K p_k(x) \|D\tau_{k,i}^{-1}(\tau_{k,i}(x))\| < \sigma < 1.$$

Let $\sup_{x \in \tau_{k,i}(S_i)} \|D\tau_{k,i}^{-1}(x)\| =: \sigma_{k,i}$ and $\sup_{x \in S_i} p_k(x) =: \pi_{k,i}$. Using the smoothness of $D\tau_{k,i}^{-1}$'s and p_k 's we can refine the partition \mathcal{P} to satisfy

CONDITION (C').

$$\sum_{k=1}^K \max_{1 \leq i \leq q} \sigma_{k,i} \pi_{k,i} < \sigma < 1.$$

Under this condition, our goal is to prove the existence of an a.c.i.m. for the random map $T = \{\tau_1, \dots, \tau_K; p_1, \dots, p_K\}$. The main tool of this section is the multidimensional notion of variation defined using derivatives in the distributional sense (see [3]):

$$V(f) = \int_{\mathbb{R}^n} \|Df\| = \sup \left\{ \int_{\mathbb{R}^n} f \operatorname{div}(g) \, d\lambda_n : g = (g_1, \dots, g_n) \in C_0^1(\mathbb{R}^n, \mathbb{R}^n) \right\},$$

where $f \in L_1(\mathbb{R}^n)$ has bounded support, Df denotes the gradient of f in the distributional sense, and $C_0^1(\mathbb{R}^n, \mathbb{R}^n)$ is the space of continuously differentiable functions from \mathbb{R}^n into \mathbb{R}^n having a compact support. We will use the following property of variation which is derived from [3, Remark 2.14]: If $f = 0$ outside a closed domain A whose boundary is Lipschitz continuous, $f|_A$ is continuous, $f|_{\operatorname{int}(A)}$ is C^1 , then

$$V(f) = \int_{\operatorname{int}(A)} \|Df\| \, d\lambda_n + \int_{\partial A} |f| \, d\lambda_{n-1},$$

where λ_{n-1} is the $n - 1$ -dimensional measure on the boundary of A . In this section we shall consider the Banach space (see [3, Remark 1.12])

$$\operatorname{BV}(S) = \{f \in L_1(S) : V(f) < \infty\},$$

with the norm $\|f\|_{\operatorname{BV}} = V(f) + \|f\|_1$. We adapt the following two lemmas from [6]. Their proofs are exactly the same as in [6].

LEMMA 6.1. *Consider $S_i \in \mathcal{P}$. Let x be a point in ∂S_i and $y = \tau_k(x)$ a point in $\partial(\tau_k(S_i))$. Let $J_{k,i}$ be the Jacobian of $\tau_k|_{S_i}$ at x and $J_{k,i}^0$ be the Jacobian of $\tau_k|_{\partial S_i}$ at x . Then*

$$\frac{J_{k,i}^0}{J_{k,i}} \leq \sigma_{k,i}. \blacksquare$$

Fix $1 \leq i \leq q$. Let Z denote the set of singular points of ∂S_i . Let us construct for any $x \in Z$ the largest cone with vertex at x and which lies

completely in S_i . Let $\theta(x)$ denote the vertex angle of this cone. Then define

$$\beta(S_i) = \min_{x \in Z} \theta(x).$$

Since the faces of ∂S_i meet at angles bounded away from 0, we have $\beta(S_i) > 0$. Let $\alpha(S_i) = \pi/2 + \beta(S_i)$ and

$$a(S_i) = |\cos(\alpha(S_i))|.$$

Now we will construct a C^1 field of segments $L_y, y \in \partial S_i$, every L_y being a central ray of a regular cone contained in S_i , with vertex angle at y greater than or equal to $\beta(S_i)$.

We start at points $y \in Z$ where the minimal angle $\beta(S_i)$ is attained, defining L_y to be central rays of the largest regular cones contained in S_i . Then we extend this field of segments to the C^1 field we want, making L_y short enough to avoid overlapping. Let $\delta(y)$ be the length of $L_y, y \in \partial S_i$. By the compactness of ∂S_i we have

$$\delta(S_i) = \inf_{y \in \partial S_i} \delta(y) > 0.$$

Now, we shorten the L_y of our field, making them all of length $\delta(S_i)$.

LEMMA 6.2. For any $S_i, i = 1, \dots, q$, if f is a C^1 function on S_i , then

$$\int_{\partial S_i} f(y) d\lambda_{n-1}(y) \leq \frac{1}{a(S_i)} \left(\frac{1}{\delta(S_i)} \int_{S_i} f d\lambda_n + V_{\text{int}(S_i)}(f) \right). \blacksquare$$

Our main technical result is the following:

THEOREM 6.3. If T is a random map which satisfies condition (C), then

$$V(P_T f) \leq \sigma(1 + 1/a)V(f) + \left(M + \frac{\sigma}{a\delta} \right) \|f\|_1,$$

where $a = \min\{a(S_i) : i = 1, \dots, q\} > 0, \delta = \min\{\delta(S_i) : i = 1, \dots, q\} > 0$, and

$$M_{k,i} = \sup_{x \in S_i} \left(Dp_k(x) - \frac{DJ_{k,i}}{J_{k,i}} p_k(x) \right), \quad M = \sum_{k=1}^K \max_{1 \leq i \leq q} M_{k,i}.$$

Proof. We have $V(P_T f) \leq \sum_{k=1}^K V(P_{\tau_k}(p_k f))$. To estimate $V(P_{\tau_k}(p_k f))$, let

$$F_{k,i} = \frac{f(\tau_{k,i}^{-1})p_k(\tau_{k,i}^{-1})}{J_{k,i}(\tau_{k,i}^{-1})}, \quad R_{k,i} = \tau_{k,i}(S_i), \quad i = 1, \dots, q, k = 1, \dots, K.$$

Then

$$\begin{aligned} \int_{\mathbb{R}^n} \|DP_{\tau_k}(p_k f)\| d\lambda_n &\leq \sum_{i=1}^q \int_{\mathbb{R}^n} \|D(F_{k,i}\chi_{R_i})\| d\lambda_n \\ &\leq \sum_{i=1}^q \left(\int_{\mathbb{R}^n} \|D(F_{k,i})\chi_{R_i}\| d\lambda_n + \int_{\mathbb{R}^n} \|F_{k,i}(D\chi_{R_i})\| d\lambda_n \right). \end{aligned}$$

Now, for the first integral we have

$$\begin{aligned}
 \int_{\mathbb{R}^n} \|D(F_{k,i})\chi_{R_i}\| d\lambda_n &= \int_{R_i} \|D(F_{k,i}p_k)\| d\lambda_n \\
 &\leq \int_{R_i} \left\| D(f(\tau_{k,i}^{-1})) \frac{p_k(\tau_{k,i}^{-1})}{J_{k,i}(\tau_{k,i}^{-1})} \right\| d\lambda_n + \int_{R_i} \left\| f(\tau_{k,i}^{-1}) D\left(\frac{p_k(\tau_{k,i}^{-1})}{J_{k,i}(\tau_{k,i}^{-1})}\right) \right\| d\lambda_n \\
 &\leq \int_{R_i} \|Df(\tau_{k,i}^{-1})\| \|D\tau_{k,i}^{-1}\| \frac{p_k(\tau_{k,i}^{-1})}{J_{k,i}(\tau_{k,i}^{-1})} d\lambda_n + \int_{R_i} \|f(\tau_{k,i}^{-1})\| \frac{M_k}{J_{k,i}(\tau_{k,i}^{-1})} d\lambda_n \\
 &\leq \sigma_{k,i}\pi_{k,i} \int_{S_i} \|Df\| d\lambda_n + M_k \int_{S_i} \|f\| d\lambda_n.
 \end{aligned}$$

For the second integral we have

$$\int_{\mathbb{R}^n} \|F_{k,i}(D\chi_{R_i})\| d\lambda_n = \int_{\partial R_i} |f(\tau_{k,i}^{-1})| \frac{p_k(\tau_{k,i}^{-1})}{J_{k,i}(\tau_{k,i}^{-1})} d\lambda_{n-1} = \int_{\partial S_i} |f| p_k \frac{J_{k,i}^0}{J_{k,i}} d\lambda_{n-1}.$$

By Lemma 4.3, $J_{k,i}^0/J_{k,i} \leq \sigma_{k,i}$. Using Lemma 4.2, we get

$$\begin{aligned}
 \int_{\mathbb{R}^n} \|F_{k,i}(D\chi_{R_i})\| d\lambda_n &\leq \sigma_{k,i}\pi_{k,i} \int_{\partial S_i} |f| d\lambda_{n-1} \\
 &\leq \frac{\sigma_{k,i}\pi_{k,i}}{a} V_{S_i}(f) + \frac{\sigma_{k,i}\pi_{k,i}}{a\delta} \int_{S_i} |f| d\lambda_n.
 \end{aligned}$$

Using Condition (C'), summing first over i , we obtain

$$\begin{aligned}
 V(P_{\tau_k}(p_k f)) &\leq (\max_{1 \leq i \leq q} \sigma_{k,i}\pi_{k,i})(1 + 1/a)V(f) \\
 &\quad + \left(\max_{1 \leq i \leq q} M_{k,i} + \frac{\max_{1 \leq i \leq q} \sigma_{k,i}\pi_{k,i}}{a\delta} \right) \|f\|_1.
 \end{aligned}$$

Then summing over k yields the assertion. ■

THEOREM 6.4. *Let T be a random map which satisfies condition (C). If $\sigma(1 + 1/a) < 1$, then T preserves a measure which is absolutely continuous with respect to Lebesgue measure. The operator P_T is quasi-compact on $BV(S)$ (see [1]).*

Proof. This follows by the standard technique (see [1]). ■

7. Example in \mathbb{R}^2 . In this section, we present an example of a random map which satisfies condition (C) of Theorem 6.3 and thus it preserves an absolutely continuous invariant measure.

EXAMPLE 7.1. Let T be a random map given by $\{\tau_1, \tau_2; p_1(x), p_2(x)\}$, where $\tau_1, \tau_2 : I^2 \rightarrow I^2$ are defined by

$$\tau_1(x_1, x_2) = \begin{cases} (3x_1, 2x_2) & \text{for } (x_1, x_2) \in S_1 = \{0 \leq x_1, x_2 \leq 1/3\}, \\ (3x_1 - 1, 2x_2) & \text{for } (x_1, x_2) \in S_2 = \{1/3 < x_1 \leq 2/3; 0 \leq x_2 \leq 1/3\}, \\ (3x_1 - 2, 2x_2) & \text{for } (x_1, x_2) \in S_3 = \{2/3 < x_1 \leq 1; 0 \leq x_2 \leq 1/3\}, \\ (3x_1, 3x_2 - 1) & \text{for } (x_1, x_2) \in S_4 = \{0 < x_1 \leq 1/3; 1/3 < x_2 \leq 2/3\}, \\ (3x_1 - 1, 3x_2 - 1) & \text{for } (x_1, x_2) \in S_5 = \{1/3 < x_1, x_2 \leq 2/3\}, \\ (3x_1 - 2, 3x_2 - 1) & \text{for } (x_1, x_2) \in S_6 = \{2/3 < x_1 \leq 1; 1/3 < x_2 \leq 2/3\}, \\ (3x_1, 3x_2 - 2) & \text{for } (x_1, x_2) \in S_7 = \{0 \leq x_1 \leq 1/3; 2/3 < x_2 \leq 1\}, \\ (3x_1 - 1, 3x_2 - 2) & \text{for } (x_1, x_2) \in S_8 = \{1/3 < x_1 \leq 2/3; 2/3 < x_2 \leq 1\}, \\ (3x_1 - 2, 3x_2 - 2) & \text{for } (x_1, x_2) \in S_9 = \{2/3 < x_1 \leq 1; 2/3 < x_2 \leq 1\}, \end{cases}$$

$$\tau_2(x_1, x_2) = \begin{cases} (3x_1, 3x_2) & \text{for } (x_1, x_2) \in S_1, \\ (2 - 3x_1, 3x_2) & \text{for } (x_1, x_2) \in S_2, \\ (3x_1 - 2, 3x_2) & \text{for } (x_1, x_2) \in S_3, \\ (3x_1, 3x_2 - 1) & \text{for } (x_1, x_2) \in S_4, \\ (2 - 3x_1, 3x_2 - 1) & \text{for } (x_1, x_2) \in S_5, \\ (3x_1 - 2, 3x_2 - 1) & \text{for } (x_1, x_2) \in S_6, \\ (3x_1, 3x_2 - 2) & \text{for } (x_1, x_2) \in S_7, \\ (2 - 3x_1, 3x_2 - 2) & \text{for } (x_1, x_2) \in S_8, \\ (3x_1 - 2, 3x_2 - 2) & \text{for } (x_1, x_2) \in S_9, \end{cases}$$

and

$$p_1(x) = \begin{cases} 0.215 & \text{for } (x_1, x_2) \in S_1, \\ 0.216 & \text{for } (x_1, x_2) \in S_2, \\ 0.216 & \text{for } (x_1, x_2) \in S_3, \\ 0.216 & \text{for } (x_1, x_2) \in S_4, \\ 0.215 & \text{for } (x_1, x_2) \in S_5, \\ 0.216 & \text{for } (x_1, x_2) \in S_6, \\ 0.216 & \text{for } (x_1, x_2) \in S_7, \\ 0.216 & \text{for } (x_1, x_2) \in S_8, \\ 0.215 & \text{for } (x_1, x_2) \in S_9, \end{cases} \quad p_2(x) = \begin{cases} 0.785 & \text{for } (x_1, x_2) \in S_1, \\ 0.784 & \text{for } (x_1, x_2) \in S_2, \\ 0.784 & \text{for } (x_1, x_2) \in S_3, \\ 0.784 & \text{for } (x_1, x_2) \in S_4, \\ 0.785 & \text{for } (x_1, x_2) \in S_5, \\ 0.784 & \text{for } (x_1, x_2) \in S_6, \\ 0.784 & \text{for } (x_1, x_2) \in S_7, \\ 0.784 & \text{for } (x_1, x_2) \in S_8, \\ 0.785 & \text{for } (x_1, x_2) \in S_9. \end{cases}$$

The derivative matrix of $\tau_{1,i}^{-1}$ is

$$\begin{pmatrix} 1/3 & 0 \\ 0 & 1/3 \end{pmatrix} \quad \text{or} \quad \begin{pmatrix} 1/3 & 0 \\ 0 & 1/2 \end{pmatrix},$$

$$\begin{aligned}
 & + p_2 \mathbf{Id}_9 \cdot \begin{pmatrix} 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 & 1/9 \end{pmatrix} \\
 & = \begin{pmatrix} a & a & a & a & a & a & b & b & b \\ c & c & c & c & c & c & d & d & d \\ c & c & c & c & c & c & d & d & d \\ e & e & e & e & e & e & e & e & e \\ e & e & e & e & e & e & e & e & e \\ e & e & e & e & e & e & e & e & e \\ e & e & e & e & e & e & e & e & e \\ e & e & e & e & e & e & e & e & e \\ e & e & e & e & e & e & e & e & e \end{pmatrix},
 \end{aligned}$$

where

$$\begin{aligned}
 p_1 &= (0.215, 0.216, 0.216, 0.216, 0.215, 0.216, 0.216, 0.216, 0.215), \\
 p_2 &= (0.785, 0.784, 0.784, 0.784, 0.785, 0.784, 0.784, 0.784, 0.785),
 \end{aligned}$$

\mathbf{Id}_9 is the 9×9 identity matrix and

$$\begin{aligned}
 a &= 0.12306, & b &= 0.087222, & c &= 0.12311, \\
 d &= 0.087111, & e &= 0.11111.
 \end{aligned}$$

The invariant density of T is

$$f = (f_1, f_2, f_3, f_4, f_5, f_6, f_7, f_8, f_9), \quad f_i = f|_{S_i}, \quad i = 1, \dots, 9,$$

normalized by

$$f_1 + f_2 + f_3 + f_4 + f_5 + f_6 + f_7 + f_8 + f_9 = 9,$$

and satisfying the equation $fM = f$. Then

$$f_1 = f_2 = f_3 = f_4 = f_5 = f_6 = \frac{9}{6.29739}, \quad f_7 = f_8 = f_9 = \frac{0.29739}{3} f_1.$$

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