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Bifurcations of stationary measures of random diffeomorphisms

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Abstract. Random diffeomorphisms with bounded absolutely continuous noise are known to possess a finite number of stationary measures. We discuss the dependence of stationary measures on an auxiliary parameter, thus describing bifurcations of families of random diffeomorphisms. A bifurcation theory is developed under mild regularity assumptions on the diffeomorphisms and the noise distribution (e.g. smooth diffeomorphisms with uniformly distributed additive noise are included). We distinguish bifurcations where the density function of a stationary measure varies discontinuously or where the support of a stationary measure varies discontinuously. We establish that generic random diffeomorphisms are stable. The densities of stable stationary measures are shown to be smooth and to depend smoothly on an auxiliary parameter, except at bifurcation values. The bifurcation theory explains the occurrence of transients and intermittency as the main bifurcation phenomena in random diffeomorphisms. Quantitative descriptions by means of average escape times from sets as functions of the parameter are provided. Further quantitative properties are described through the speed of decay of correlations as a function of the parameter. Random differentiable maps which are not necessarily injective are studied in one dimension; we show that stable one-dimensional random maps occur open and dense and that in one-parameter families bifurcations are typically isolated. We classify codimension-one bifurcations for one-dimensional random maps; we distinguish three possible kinds, the random saddle node, the random homoclinic and the random boundary bifurcation. The theory is illustrated on families of random circle diffeomorphisms and random unimodal maps.

1. *Introduction*

To fix thoughts, consider a single map with an attracting fixed point P. Write $W^s(P)$ for its basin of attraction. Adding uniform noise of small amplitude gives a random map with a stationary density with support near P. Increasing the amplitude of the noise leads to a bifurcation when orbits can escape from $W^s(P)$. Two possibilities occur: escaping

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orbits can or cannot return near P. Escaping orbits lead to transient dynamics if orbits do not return or intermittent dynamics if orbits do return. How can such transitions occur? What are the quantitative characteristics? As a second issue, take a one-parameter family of maps that exhibits bifurcations and add small bounded noise to it. What happens to the bifurcation set and in what way do bifurcations in the randomly perturbed map manifest? More generally, one can consider maps and families of maps where noise is an intrinsic part of the description.

It is the purpose of this paper to work out bifurcation theory of random smooth diffeomorphisms from the perspective of stationary measures, providing answers to questions like the ones just stated. The context in which we perform this study is that of points being mapped into bounded domains according to a probability distribution, under some regularity conditions. One can think of points being mapped by a diffeomorphism, defined on some compact manifold $\mathcal{M}$, followed by a random perturbation. This defines a discrete Markov process on $\mathcal{M}$ given by transition functions $P(x, A)$ providing the chance that a point $x \in \mathcal{M}$ ends up in a Borel set $A \subset \mathcal{M}$. We will assume that the region into which x is mapped is a bounded domain U_x. We argue that from a modeling point of view there are clear and good reasons to consider bounded noise: in most physical systems random perturbations are limited in their effect.

We indicate the regularity conditions assumed in this paper. A first regularity assumption is smoothness of the density of the transition functions $P(x, A)$. It is not assumed that this density vanishes on the boundary of its support; densities that are positive on bounded domains to model uniform noise are incorporated. For $y \in \mathcal{M}$, write V_y for the set of points $x \in \mathcal{M}$ that are mapped to domains that include y:

$$V_y = \{x \in \mathcal{M} \mid y \in \text{support } P(x, \cdot)\}.$$

As a second regularity assumption, we suppose that the sets V_y are domains with piecewise smooth boundary varying smoothly with y. This assumption is natural in the context of diffeomorphisms followed by noise, but does not hold in the context of a map (possessing critical points) followed by noise. Precise formulations follow in §1.1 below.

An alternative description of the setup is by starting with a collection $\mathcal{F}$ of maps on $\mathcal{M}$ and a measure on $\mathcal{F}$. Maps are drawn randomly, and independently, from $\mathcal{F}$ according to the given measure. Similarly one can consider maps depending on parameters that are drawn randomly. The random parameters are drawn from a bounded domain according to a given distribution. Typical examples are given by smooth maps f with additive or parametric noise. It will turn out that there is no loss of generality, as far as statistical descriptions are concerned, when considering maps with finitely many random parameters. We address an appendix to the exploration of the range and connections of these definitions.

In the following we will mostly speak of random maps; maps depending on finitely many parameters that are drawn randomly. In the part of the paper developing the general theory, §§2–7, we assume the maps to be diffeomorphisms. This guarantees the regularity assumptions formulated above in the description as discrete Markov processes to hold. Random maps are studied in one dimension in §8.

Under the mild regularity assumptions, random diffeomorphisms possess finitely many stationary measures, whose support is the closure of an open set and whose density

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functions (stationary densities) are smooth on all of $\mathcal{M}$. The stationary densities are flat along the boundary of their support. This has immediate consequences for the statistical properties of orbits: orbits are very rarely found near the boundary of the support of the stationary densities. The main focus of this paper lies then in the description of the dependence of stationary densities on the random diffeomorphisms. This includes describing quantitative characteristics. Motivated by examples discussed below we call a random diffeomorphism *stable* if its stationary densities and their supports vary continuously with the random diffeomorphism (precise definitions follow shortly). Otherwise we speak of a *bifurcation*.

Below we introduce the precise setup and present our main results in a series of theorems. The presentation of the material is separated into a section treating random diffeomorphisms (§1.1) and a section treating families of random diffeomorphisms (§1.2). Various aspects of the bifurcation theory for random diffeomorphisms are presented in Theorems 1.3, 1.6, 1.7, 1.10, 1.13 and 1.15. Appendix A comments on the setup. The main body of the theory is developed in §§2–7. Sections 2, 3 and 4 contain respectively material on transfer operators, proofs of stability theorems, and discussions of parameter dependence. Sections 5 and 6 develop material on conditionally stationary measures and apply this to compute expected escape times. Section 7 treats the speed of decay of correlations depending on a parameter. Appendix B contains an implicit function theorem used to obtain regularity in the parameter of solutions of integral equations involving the transfer operator.

There is no easy analogous theory for random maps; stationary measures for random maps will in general be less regular resulting in different statistical properties. In §8 we present a satisfactory theory for one-dimensional random maps, including a classification of possible bifurcations. This extends the general theory in particular by classifying codimension-one bifurcations. Typically only finitely many bifurcations occur in one parameter families of random maps, in contrast to families of deterministic maps. The material on one-dimensional random maps is developed and presented in §8.

Section 9 contains two worked examples of random circle diffeomorphisms and random unimodal maps. In §9.1, we consider the standard circle diffeomorphism with small additive noise,

$$f_a(x;\omega) = x + a + \omega + \varepsilon \sin(2\pi x), \tag{1}$$

with x on the circle $\mathbb{R}/\mathbb{Z}$, for fixed $\varepsilon \in (0, 1)$ and a random parameter ω taken uniformly from a small interval. Section 9.2 discusses as a prototypical example of random maps on an interval, random logistic maps

$$f(x;\mu) = (\mu + \omega)x(1 - x), \tag{2}$$

with multiplicative noise obtained by varying ω with a uniform distribution in some interval.

There is a large body of literature on stochastic stability (e.g. [**2, 3, 7, 31, 47**]; see also [**10**]), considering bounded noise as a means to treat properties of single deterministic systems. This is done by letting the noise level decrease to zero. In contrast, we consider maps and families of maps where noise is an intrinsic part of the description.

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Previous attempts to study stochastic bifurcations fall into two categories. One way is to consider notions close to the traditional understanding of bifurcations for deterministic dynamics by embedding the random dynamical system into a skew product system. Another way studied frequently in the literature uses singularity theory to describe changes in the density of a stationary measure. This approach has been used to study systems with unbounded noise, often of a Gaussian nature, so that a unique smooth stationary density occurs. See [**5**] for further discussion. We consider the shape of the stationary densities, but the noise being bounded leaves the dynamics part of the picture.

Central in control theory are control sets, that is maximal sets of approximate controllability. The shape of the control sets varies with external parameters, where discontinuous changes are possible. Interpreting the control variable as noise, a relation with the present paper becomes apparent. Discontinuous changes of the control set, interpreted this way, are among the bifurcations identified in this paper. In the context of differential equations depending on a control variable, the study of control sets is taken up in [**14, 22**]. Bifurcation theory for such random differential equations in the spirit of this paper is considered in [**28**].

1.1. *Random diffeomorphisms.* The adjective *smooth* stands for C^∞. Let $\mathcal{M}$ be a smooth n-dimensional compact Riemannian manifold with measure m induced by the Riemannian structure. Let Δ be a closed domain in n-dimensional Euclidean space. Smoothness of a function g on Δ is to be understood in the sense that g can be extended to a smooth function on a neighborhood of Δ.

Definition 1.1. A *random differentiable map* is a smooth map $f : \mathcal{M} \times \Delta \to \mathcal{M}$, $x \mapsto f(x; \omega)$, depending on a random parameter $\omega \in \Delta$ drawn from a measure on Δ with smooth density function $g : \Delta \to \mathbb{R}$, $\omega \mapsto g(\omega)$. A *random diffeomorphism* is a random differentiable map so that $x \mapsto f(x; \omega)$ is a diffeomorphism for each ω.

Remark 1.2. Alternatively one can explicitly include the noise distribution g as part of the definition of random differentiable map (and speak of a pair (f, g)). For convenience we consider g given. The results in this paper have direct, easily obtained, analogs if g is allowed to vary.

Note that maps and diffeomorphisms are always assumed to be smooth. The basic setup we are treating is of points being mapped into bounded domains according to some probability. The following standing assumptions will be made with this setup in mind. The random parameters will be chosen from a region Δ such that:

(H1) Δ is a domain in $\mathbb{R}^n$ with a piecewise smooth boundary.

The number of random parameters is in particular equal to the dimension of the state space $\mathcal{M}$. The most important examples are where Δ is the unit ball $\Delta = \{x \in \mathbb{R}^n \mid \|x\| \leq 1\}$ or the unit box $\Delta = \{(x_1, \ldots, x_n) \in \mathbb{R}^n \mid |x_1|, \ldots, |x_n| \leq 1\}$. Throughout this paper we assume that:

(H2) $\omega \mapsto f(x; \omega)$ is an injective map for each x.

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Write ν for the measure on Δ with density function g. A smooth random map gives rise to a discrete Markov process through the transition functions

$$P(x, A) = \int_{\{\omega | f(x;\omega)\in A\}} d\nu(\omega) \tag{3}$$

for Borel sets A. With $h_x(\omega) = f(x; \omega)$, the measure $P(x, \cdot)$ equals $(h_x)_*\nu$ defined by $(h_x)_*\nu(A) = \nu(h_x^{-1}(A))$. *Vice versa*, a discrete Markov process with noise from a ball or a box such that its transition functions have smooth positive densities admits a representation by smooth random maps (depending injectively on a random parameter); see Appendix A. Some cases of parametric noise, where the maps do not depend injectively on the random parameter, do not fall into this setup. We refer to Appendix A for further discussion.

The general theory will be developed for random diffeomorphisms, instead of random differentiable maps. Non-invertible maps allow for pathological examples, for instance maps $f(x; \omega)$ that are constant in x. We will however discuss random differentiable maps in one-dimension (on a circle or a compact interval) in detail.

With a slight abuse of notation, iterates of $f(x; \omega)$ are given as

$$f^k(x; \omega_1, \ldots, \omega_k) = f(f^{k-1}(x; \omega_1, \ldots, \omega_{k-1}); \omega_k). \tag{4}$$

More generally, write $\Delta^{\mathbb{N}}$ for all infinite sequences $\boldsymbol{\omega} = \{\omega_i\}_{i\geq 1}$ with each $\omega_i \in \Delta$. Denote $f^k(x; \boldsymbol{\omega}) = f^k(x; \omega_1, \ldots, \omega_k)$. Let $\vartheta : \Delta^{\mathbb{N}} \to \Delta^{\mathbb{N}}$ be the left shift operator, $\vartheta\{\omega_i\}_{i\geq 1} = \{\omega_i\}_{i\geq 2}$. Consider the skew product system $S : \mathcal{M} \times \Delta^{\mathbb{N}} \to \mathcal{M} \times \Delta^{\mathbb{N}}$ given by

$$S(x, \boldsymbol{\omega}) = (f(x; \omega_1), \vartheta\boldsymbol{\omega}). \tag{5}$$

On $\Delta^{\mathbb{N}}$ one considers a measure ν^∞ which is the product of the measure ν over each Δ.

With these definitions in mind, we introduce the central notions of stationary measures and ergodic measures. A stationary measure μ for the smooth random map f is a probability measure on $\mathcal{M}$ with $\mu \times \nu^\infty$ S-invariant,

$$\mu \times \nu^\infty(S^{-1}(B)) = \mu \times \nu^\infty(B)$$

for Borel sets $B \subset \mathcal{M} \times \Delta^{\mathbb{N}}$. Equivalently, see [**4, 31**],

$$\mu(A) = \int_{\mathcal{M}} P(x, A)\, d\mu(x)$$

for Borel sets $A \subset \mathcal{M}$. We refer to a stationary density as the density of an absolutely continuous stationary measure.

A stationary measure μ is called *ergodic* if $\mu \times \nu^\infty$ is an ergodic measure for S in the usual sense that invariant subsets of $\mathcal{M} \times \Delta^{\mathbb{N}}$ for S have zero or full measure. See [**31**] for equivalent formulations. The Birkhoff ergodic theorem tells us that, for an ergodic stationary measure,

$$\lim_{n\to\infty} \frac{1}{n} \sum_{i=0}^{n-1} \phi(f^i(x; \boldsymbol{\omega})) = \int_{\mathcal{M}} \phi(x)\, d\mu(x) \tag{6}$$

for all integrable functions ϕ on $\mathcal{M}$ and $\mu \times \nu^\infty$ almost every point $(x, \boldsymbol{\omega})$. Taking $\phi = 1_A$, the characteristic function of a Borel set $A \subset \mathcal{M}$, it shows that the relative frequency with which typical orbits visit A is given by $\mu(A)$.

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Write $R^k(\mathcal{M})$ for the space of C^k random diffeomorphisms f on $\mathcal{M}$ (with $f(x;\omega)$ C^k jointly in $x \in \mathcal{M}$ and $\omega \in \Delta$), depending on a random parameter from Δ through a distribution with a C^k density function g.

Let a random diffeomorphism $f \in R^\infty(\mathcal{M})$ be given. The existence of finitely many ergodic stationary measures for f presented in the following theorem can be found in [**17**, Ch. 5] (valid under more general conditions). Similar results are contained in [**4**]. We add statements on the regularity of the stationary measures valid in our context. Differentiability of stationary densities is also discussed in [**8, 42**].

THEOREM 1.3. *The random diffeomorphism $f \in R^\infty(\mathcal{M})$ possesses a finite number of ergodic stationary measures $\mu_1, \ldots, \mu_m$ with mutually disjoint supports $E_i, \ldots, E_m$. All stationary measures are linear combinations of $\mu_1, \ldots, \mu_m$.*

The support E_i of μ_i consists of the closure of a finite number of connected open sets $C_i^1, \ldots, C_i^p$ that are moved cyclically by $f(\cdot;\Delta)$. The density ϕ_i of μ_i is a C^∞ function on $\mathcal{M}$.

Proof. See [**17**, Ch. 5] for the existence proof of cyclically permuted ergodic stationary measures.

Since a point x is mapped to a set $f(x;\Delta)$ diffeomorphic to Δ, the support E_i is the closure of an open set. We claim that the closures of the sets E_i, $1 \le i \le m$, are mutually disjoint. Suppose on the contrary that ∂E_i and ∂E_j with $j \neq i$ have a point z in common. Then z is mapped by $f(\cdot;\Delta)$ to the injective image of Δ. By the invariance of E_i and E_j under $f(\cdot;\Delta)$, $f(z;\Delta)$ is contained in both $\bar{E}_i$ and $\bar{E}_j$, which is not possible. Note that compactness of $\mathcal{M}$ implies that there are only finitely many stationary measures.

The regularity statements follow from Proposition 2.3 in §2. □

Densities of stationary measures for random differentiable maps are in general not smooth functions, but are less regular (see §8). Random differentiable maps without critical points, such as expanding maps, do possess finitely many smooth stationary densities. The proofs for random diffeomorphisms extend to cover such random differentiable maps. The regularity of a stationary density ϕ implies that ϕ is flat along the boundary of its support E. By the Birkhoff ergodic theorem (applied to the characteristic function of a neighborhood of ∂E), this means that typical orbits are very infrequently found near the boundary of E.

We introduce a topology on the space $R^k(\mathcal{M})$ of random diffeomorphisms in order to be able to compare the dynamics of nearby random diffeomorphisms. Natural topologies on $R^k(\mathcal{M})$ are the uniform C^k topologies on $C^k(\mathcal{M}\times\Delta, \mathcal{M})$. See e.g. [**25**] for generalities on these topologies. We will assume $R^k(\mathcal{M})$ to be equipped with this topology. Note that the alternative approach through discrete Markov processes suggests a topology using the densities of the transition functions.

Consider $f \in R^\infty(\mathcal{M})$. Write $\mu_1, \ldots, \mu_m$ for the stationary measures of $f \in R^\infty(\mathcal{M})$ given by Theorem 1.3.

Definition 1.4. A random differentiable map $f \in R^\infty(\mathcal{M})$ is *stable* if, for all $\tilde{f}$ sufficiently close to f, the following two properties are satisfied.

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(i) For each i, $1 \le i \le m$, the random differentiable map $\tilde{f}$ has a stationary measure $\tilde{\mu}_i$ whose density is C^0 close to that of μ_i.
(ii) The supports of $\tilde{\mu}_i$ and μ_i are close in the Hausdorff metric.

We speak of a *bifurcation*, or a *bifurcating random differentiable map*, if at least one of these properties is violated.

Definition 1.5. An ergodic stationary measure of $f \in R^\infty(\mathcal{M})$ is called *isolated* or *attracting* if there exists an open set W (an isolating neighborhood) containing the support E of μ, so that $\overline{f(W;\Delta)} \subset W$ and μ is the only ergodic stationary measure of f with support in W.

For each $\tilde{f}$ close to f the closure of $\tilde{f}(W;\Delta)$ is contained in W. The following stability result shows that nearby random diffeomorphisms have indeed a unique stationary measure with support in W. For bifurcations where stationary measures vary discontinuously the condition of being isolated must therefore be violated. The proof of the following theorem is found in §3.

THEOREM 1.6. *Let μ be an isolated ergodic stationary measure of $f \in R^\infty(\mathcal{M})$ with density ϕ with isolating neighborhood W. Then each $\tilde{f} \in R^\infty(\mathcal{M})$ sufficiently close to f possesses a unique ergodic stationary measure $\tilde{\mu}$ with support in W. The density $\tilde{\phi}$ of $\tilde{\mu}$ is C^∞ close to ϕ.*

Note though that the above theorem leaves open the possibility that the supports $\tilde{E}$ of $\tilde{\mu}$ and E of μ are not close in the Hausdorff metric. An illustrative example of this phenomenon is described in §9.1.

The following theorem establishes that stable random diffeomorphisms are generic. Its proof is in §3. The argument also shows that random diffeomorphisms with a locally constant number of smoothly varying stationary densities (ignoring variations in their support) form an open and dense subset of $R^\infty(\mathcal{M})$.

THEOREM 1.7. *The set of stable random diffeomorphisms in $R^\infty(\mathcal{M})$ contains a countable intersection of open and dense sets.*

A thorough description of the dynamics of random circle diffeomorphisms is given in §9.1. Random diffeomorphisms, and even random differentiable maps, on the circle are shown to form an open and dense set; see Theorem 8.3.

1.2. *Families of random diffeomorphisms.* Bifurcations are best studied in families depending on finitely many parameters. We will consider families of random diffeomorphisms depending on a single real parameter, where we have the goal to focus on bifurcations that typically occur by varying one parameter.

Definition 1.8. A *smooth family of random differentiable maps* is a family of random differentiable maps $\{f_a\}$ depending on parameters a, so that $f_a(x;\omega)$ depends smoothly on (x, ω, a). A *smooth family of random diffeomorphisms* is a smooth family of random differentiable maps where each map $f_a(\cdot;\omega)$ is a diffeomorphism.

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Remark 1.9. Alternatively, one can explicitly include noise densities g_a in the definition (considering pairs (f_a, g_a)) with $g_a(\omega)$ varying smoothly with (ω, a); compare Remark 1.2. For convenience we consider fixed noise densities, but completely analogous results hold if the noise densities are allowed to vary with a.

Consider a smooth one-parameter family $\{f_a\}$ of random diffeomorphisms, with a from an interval I. Consider a parameter value a_0 and an ergodic stationary measure μ_{a_0} with support E_{a_0}. The following result extends Theorem 1.6, providing an analogous statement in the context of families. If μ_{a_0} is an isolated ergodic stationary measure then there are ergodic invariant measures μ_a for a near a_0 with nearby densities.

THEOREM 1.10. *Suppose μ_{a_0} is an isolated ergodic stationary measure. Then the stationary density $x \mapsto \phi_a(x)$ of μ_a depends C^∞ on (x, a).*

See §4 for the proof. We stress again that the support E_a of μ_a can still vary discontinuously in the Hausdorff metric with a. The number of components of the support of the stationary measure can also change, while the stationary density varies smoothly.

Consider a smooth function ϕ with support on an isolating neighborhood W for μ_{a_0} and compute averages of ϕ along orbits $f_a^k(x; \boldsymbol{\omega})$. By the Birkhoff ergodic theorem, for typical initial points $x \in W$ and noise sequences $\boldsymbol{\omega}$, the averages lie on a smooth function of a for a near a_0.

The two types of bifurcation distinguished in Definition 1.4 give rise to a particular dynamical phenomenon associated to either intermittency or transients. Consider a family $\{f_a\}$ of random diffeomorphisms in $R^\infty(\mathcal{M})$, with a from an open interval I. Suppose that $a_0 \in I$ is a bifurcation value for $\{f_a\}$ involving a stationary measure μ. Write ϕ for the density of μ. Analogies with deterministic dynamics suggest the following two definitions. In §8 we will see that in typical one parameter families of random interval or circle differentiable maps bifurcations are isolated and of these two types.

Definition 1.11. The bifurcation at a_0 is called an *intermittency bifurcation* if there is a stationary density ϕ_a for $\{f_a\}$ with $\phi_{a_0} = \phi$ and depending continuously on a, so that the support E_a of ϕ_a varies with a, for a near a_0, as follows.

(i) The support E_a varies continuously for a from one side of a_0. Without loss of generality, we assume this to be the case for $a < a_0$.

(ii) The support E_a is discontinuous at $a = a_0$ and E_a contains an open set disjoint from E_{a_0} for $a > a_0$.

An orbit piece outside a small neighborhood W of E_{a_0} is called a *burst*. Out of the substantial literature on intermittency in dynamics, we point to references [**18, 23, 24, 26, 27, 40**].

Definition 1.12. The bifurcation at a_0 is called a *transient bifurcation* if there is a stationary density ϕ_a for $\{f_a\}$ with $\phi_{a_0} = \phi$ for a close to a_0 from one side of a_0 (without loss of generality, we assume this to be the case for $a < a_0$), so that:

(i) ϕ_a and its support E_a vary continuously with a, for $a \leq a_0$; and

(ii) there is no stationary density near ϕ_{a_0} for a close to a_0 and $a > a_0$.

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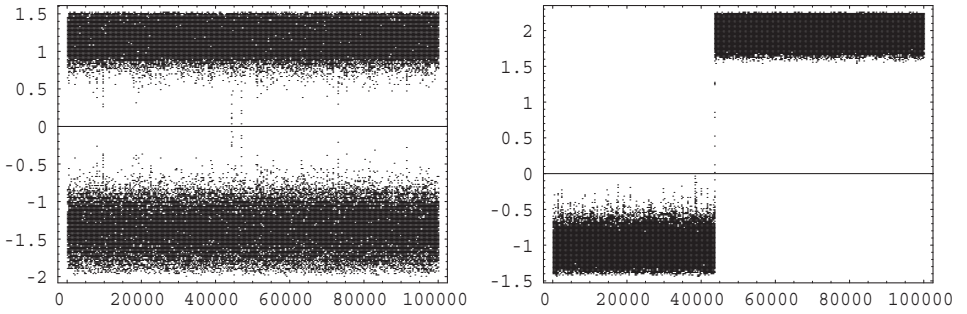


FIGURE 1. Typical times series for intermittent dynamics on the left and transient dynamics on the right. The time series are for bifurcation values after the bifurcation has taken place. The intermittency bifurcation involves interval diffeomorphisms with a single stationary measure; the support consisting of two intervals for a period-two cycle bifurcates to form a single interval. In the transient bifurcation one stationary measure out of two stationary measures existing prior to the bifurcation disappears.

We end this section with a quantitative description of time series near intermittent or transient random bifurcations (see Figure 1). We do this through the estimation of the expected escape time from a small neighborhood of the support of the bifurcation stationary measure. Estimating the speed of decay of correlations as a function of a parameter gives further details of changes through bifurcations.

As before, let $\{f_a\}$ be a family of random diffeomorphisms in $R^\infty(\mathcal{M})$, with a from an open interval I. Let μ_{a_0} be a stationary measure of f_{a_0} for some $a_0 \in I$ and let W be an open neighborhood of the support E_{a_0} of μ_{a_0} such that no other stationary measure has support intersecting $\overline{W}$. If f_{a_0} is stable, then there is a unique stationary measure μ_a that is the continuation of μ_{a_0}. The stationary measure μ_a has its support in W for a near a_0. If a_0 is a bifurcation value for $\{f_a\}$, iterates $f_a^k(x;\boldsymbol{\omega})$, for certain $x \in W$ and a near a_0, may leave W.

For $x \in W$ and $\boldsymbol{\omega} \in \Delta^{\mathbb{N}}$, define

$$\chi_a(x,\boldsymbol{\omega}) = \min\{k \mid f_a^k(x;\boldsymbol{\omega}) \notin W\}. \tag{7}$$

The following result shows how the average escape time from a neighborhood of the support of a bifurcating stationary measure is more than polynomially large in an unfolding parameter. This makes it difficult to establish accurately the bifurcation parameter value using finite data, even in numerical simulations. It explains the occurrence of very long transients near a transient bifurcation and the very irregular occurrence of bursts in intermittent time series. The proof, in §6, relies on the construction of conditionally stationary measures in §5.

THEOREM 1.13. *Let $\{f_a\}$ be a family of random diffeomorphisms in $R^\infty(\mathcal{M})$, with the parameter a from an open interval I. Let μ_{a_0} be a stationary measure of f_{a_0} for some $a_0 \in I$ and let W be an open neighborhood of the support E_{a_0} of μ_{a_0} such that no other stationary measure has support intersecting $\overline{W}$.*

For each $k > 0$ there is a constant $C_k > 0$ so that

$$\int_W \int_{\Delta^{\mathbb{N}}} \chi_a(x,\boldsymbol{\omega})\, d\nu^\infty(\boldsymbol{\omega})\, dm(x) \geq C_k\, |a - a_0|^{-k}\,.$$

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A single random map with an isolated measure supported on a single component has exponential decay of correlations as the following proposition makes precise. The interest from our perspective in computing the speed of decay of correlations lies in the study of bifurcations where the support of a stationary measure has several components merging. This will be discussed below. Proofs of the following statements are in §7. The reader can consult [**6, 45**] for background on decay of correlations. Write

$$U^n\psi(x) = \int_{\Delta^n} \psi \circ f^n(x; \omega_1, \ldots, \omega_n)\, d\nu(\omega_1) \cdots d\nu(\omega_n).$$

PROPOSITION 1.14. *Let f be a random map with an isolated stationary measure μ with connected support. Let W be an isolating neighborhood for μ. Take $\varphi, \psi \in \mathcal{L}^2(W)$. Then*

$$\left| \int_{\mathcal{M}} \varphi(x) U^n \psi(x)\, dm(x) - \int_{\mathcal{M}} \varphi(x)\, dm(x) \int_{\mathcal{M}} \psi(x)\, d\mu(x) \right| \leq C\eta^n$$

for some $C > 0$, $0 < \eta < 1$.

Note that the exponential decay of correlations holds for observables $\varphi, \psi \in \mathcal{L}^2(W)$, thus including characteristic functions of open sets.

We return to a family $\{f_a\}$, $a \in I$, of random maps. Assume that f_a has an isolated measure μ_a for all $a \in I$ with an isolating neighborhood W. Suppose $a_0 \in I$ is a bifurcation value for an intermittency bifurcation so that:

(a) the support of μ_a consists of k components for $a \leq a_0$; and
(b) the support of μ_a consists of a single component for $a > a_0$.

We incorporate the dependence of U^n on a into the notation by writing U_a^n.

THEOREM 1.15. *Let $\{f_a\}$ be as above. Take $\varphi, \psi \in \mathcal{L}^2(W)$. There are a constant $C > 0$ and a smooth function $a \mapsto \eta_a$ with $\eta_a = 1$ for $a < a_0$ and $\eta_a < 1$ for $a > a_0$, so that*

$$\left| \int_{\mathcal{M}} \varphi(x) U_a^n \psi(x)\, dm(x) - \int_{\mathcal{M}} \varphi(x)\, dm(x) \int_{\mathcal{M}} \psi(x)\, d\mu_a(x) \right| \leq C\eta_a^n$$

for $a > a_0$.

The smoothness properties of η_a imply that $\eta_a - 1$ is a flat function of a at $a = a_0$. As in the discussion of escape times, that shows how slowly the bifurcation manifests itself in time series when moving the parameter a.

2. *Transfer operators*

Let f be a random diffeomorphism on the manifold $\mathcal{M}$. Associated to f is the stochastic transition function,

$$P(x, A) = \int_{\{\omega \mid f(x;\omega) \in A\}} d\nu(\omega) = \int_{\Delta} 1_A(f(x; \omega))\, d\nu(\omega),$$

for Borel sets $A \subset \mathcal{M}$. Write $h_x(\omega) = f(x; \omega)$, and note that h maps Δ injectively onto $U_x = f(x; \Delta)$. The density

$$k(x, y) = d\,(h_x)_*\, \nu/dm \tag{8}$$

of $P(x, \cdot)$ vanishes for y outside U_x and is a smooth function on its support.

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Write $\mathcal{L}^1(\mathcal{M})$ for the space of integrable functions on $\mathcal{M}$. Define the transfer operator L acting on $\mathcal{L}^1(\mathcal{M})$ by

$$\int_A L\phi(x)\,dm(x) = \int_{\mathcal{M}} P(x, A)\phi(x)\,dm(x). \tag{9}$$

The transfer operator is a positive linear operator. A stationary density is a fixed point of L.

Define

$$V_x = \{z \in \mathcal{M} \mid x \in f(z; \Delta)\}, \tag{10}$$

which is the set of points in $\mathcal{M}$ that are mapped to x by some random map. The assumptions on the random diffeomorphism f imply that the equation $f(x; \omega) = y$ can be solved for x as a diffeomorphic map of ω. Therefore V_x is diffeomorphic to Δ and thus a domain with piecewise smooth boundary, depending smoothly on x.

Write $P_{f(\cdot;\omega)}$ for the Perron–Frobenius operator, defined by

$$\int_A P_{f(\cdot;\omega)}\phi(x)\,dm(x) = \int_{f^{-1}(A;\omega)} \phi(x)\,dm(x) \tag{11}$$

for Borel sets A. That is, for a measure μ with density $\phi = d\mu/dm$, $f_*\mu$ has density $P_f\phi = df_*\mu/dm$ (see e.g. [**33**]). The following lemma gives the transfer operator as an average over the random parameters ω of the Perron–Frobenius operators for $f(\cdot; \omega)$ and gives an equivalent formulation as an integral over the state space $\mathcal{M}$.

LEMMA 2.1. *The transfer operator is given by*

$$L\phi(x) = \int_\Delta P_{f(\cdot;\omega)}\phi(x)\,d\nu(\omega) \tag{12}$$

or

$$L\phi(x) = \int_{V_x} k(y, x)\phi(y)\,dm(y). \tag{13}$$

Proof. By (9), for a continuous function ψ,

$$\int_{\mathcal{M}} \psi(x)L\phi(x)\,dm(x) = \int_{\mathcal{M}}\int_\Delta \psi(f(x; \omega))\phi(x)\,d\nu(\omega)\,dm(x).$$

Calculate

$$\begin{aligned}\int_{\mathcal{M}}\int_\Delta \psi(f(x; \omega))\phi(x)\,d\nu(\omega)\,dm(x) &= \int_\Delta\int_{\mathcal{M}} \psi(f(x; \omega))\phi(x)\,dm(x)\,d\nu(\omega)\\ &= \int_\Delta\int_{\mathcal{M}} \psi(x)P_{f(\cdot;\omega)}\phi(x)\,dm(x)\,d\nu(\omega).\end{aligned}$$

This implies (12). Alternatively,

$$\begin{aligned}\int_{\mathcal{M}}\int_\Delta \psi(f(x; \omega))\phi(x)\,d\nu(\omega)\,dm(x) &= \int_{\mathcal{M}}\int_{U_x} \psi(y)\phi(x)k(y, x)\,dm(y)\,dm(x)\\ &= \int_{\mathcal{M}}\int_{V_y} \psi(y)\phi(x)k(y, x)\,dm(x)\,dm(y)\end{aligned}$$

which proves (13). □

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Remark 2.2. The transfer operator L preserves integrals, as the following computation shows:

$$\begin{aligned}\int_{\mathcal{M}} L\phi(y)\, dm(y) &= \int_{\mathcal{M}} \int_{V_y} k(x, y)\phi(x)\, dm(x)\, dm(y)\\ &= \int_{\mathcal{M}} \int_{U_x} k(x, y)\phi(x)\, dm(y)\, dm(x)\\ &= \int_{\mathcal{M}} \phi(x)\, dm(x),\end{aligned}$$

because $\int_{U_x} k(x, y)\, dm(x) = 1$.

Iterating L gives

$$\begin{aligned}L^2\phi(x) &= \int_{V_x} k(z, x)L\phi(z)\, dz\\ &= \int_{V_x} \int_{V_z} k(z, x)k(y, z)\phi(y)\, dy\, dz\\ &= \int_{f^{-1}(V_x,\Delta)} \int_{V_x} k(z, x)k(y, z)\, dz\, \phi(y)\, dy,\end{aligned}$$

which is of a similar form, namely

$$\int_{f^{-1}(V_x,\Delta)} k_2(y, x)\phi(y)\, dy$$

with

$$k_2(y, x) = \int_{V_x} k(z, x)k(y, z)\, dz,$$

as $L\phi(x)$. Inductively similar expressions are derived for higher iterates of L.

Denote by

$$\mathcal{D}(\mathcal{M}) = \left\{\phi \in \mathcal{L}^1(\mathcal{M}) \mid \phi \geq 0, \int_{\mathcal{M}} \phi(x)\, dm(x) = 1\right\}$$

the space of densities on $\mathcal{M}$. The above remark shows that L maps $\mathcal{D}$ into itself. Smoothness of stationary densities is obtained by showing that L maps a space of smooth densities into itself.

PROPOSITION 2.3. *The transfer operator L maps $C^k(\mathcal{M})$ into itself and is a compact operator on $C^k(\mathcal{M})$.*

The number 1 is an eigenvalue of L with equal algebraic and geometric multiplicity $m \geq 1$. The densities $\phi_1, \ldots, \phi_m$ provide a basis of eigenfunctions with mutually disjoint support. Each eigenfunction ϕ_i is C^∞ and its support consists of a finite number c_i of connected components.

Proof. Theorem 1.3 gives the m invariant densities $\phi_1, \ldots, \phi_m$. The geometric multiplicity of the eigenvalues 1 is equal to m. Since L preserves the $\mathcal{L}^1$ norm, the algebraic and geometric multiplicities of 1 are equal. To see this, suppose on the contrary that there is a non-trivial vector in $\ker(L - I)^2 \backslash \ker(L - I)$. Elementary linear algebra gives the

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existence of a sequence of vectors ψ_n inside $\ker(L-I)^2$ converging to an eigenvector ϕ, such that $\lim_{n\to\infty} L^n\psi_n = 2\phi$. Indeed, take $\psi \in \ker(L-I)^2$ with $L\psi = \phi + \psi$ and let $\psi_n = (1/n)\psi$. From $L^n(\psi) = n\phi + \psi$ it follows that $L^n(\phi + \psi_n) = 2\phi + \psi_n$, which converges to 2ϕ if $n \to \infty$. This contradicts the preservation of the $\mathcal{L}^1$ norm by L.

There can be no additional eigenvectors of L that do not correspond to linear combinations of densities. Namely, suppose that ϕ is an eigenvector taking both positive and negative values. Write $\phi = \phi_+ - \phi_-$ for non-negative functions $\phi_+ = \max\{\phi, 0\}$, $\phi_- = \max\{-\phi, 0\}$. If ϕ is not a linear combination of densities, the supports of ϕ_+ and ϕ_- cannot be invariant. Since L is positive and preserves the $\mathcal{L}^1$ norm, $(L\phi)_+ < L\phi_+$ so that ϕ cannot be an eigenvector.

We will show that L maps $\mathcal{L}^1(\mathcal{M})$ into $C^0(\mathcal{M})$ and $C^i(\mathcal{M})$ into $C^{i+1}(\mathcal{M})$. From this it follows that $\phi_i \in C^\infty(\mathcal{M})$. Take $\psi \in \mathcal{L}^1(\mathcal{M})$. Use a chart to identify a neighborhood of $x \in \mathcal{M}$ with an open set in $\mathbb{R}^n$. With h a small vector in $\mathbb{R}^n$, consider

$$\begin{aligned} L\psi(x+h) - L\psi(x) &= \int_{V_{x+h}} k(y,x)\psi(y)\,dm(y) - \int_{V_x} k(y,x)\psi(y)\,dm(y) \\ &= \int_{V_{x+h}\cap V_x} (k(y,x+h) - k(y,x))\psi(y)\,dm(y) \\ &\quad + \int_{V_{x+h}\setminus(V_{x+h}\cap V_x)} k(y,x+h)\psi(y)\,dm(y) \\ &\quad - \int_{V_x\setminus(V_{x+h}\cap V_x)} k(y,x)\psi(y)\,dm(y). \end{aligned} \tag{14}$$

The first term on the right-hand side is small for h small by the continuity of k and integrability of ψ. The other two terms are small for h small by the continuous dependence of V_x on x. The continuity of $L\psi$ follows. Suppose next that $\psi \in C^0(\mathcal{M})$ and consider $(1/|h|)\,(L\psi(x+h) - L\psi(x))$. This equals the right-hand side of (14) divided by $|h|$. Note that

$$\lim_{h\to 0} \frac{1}{|h|} \int_{V_{x+h}\cap V_x} (k(y,x+h) - k(y,x))\psi(y)\,dm(y) = \int_{V_x} \frac{\partial}{\partial x} k(y,x) \frac{h}{|h|} \psi(y)\,dm(y)$$

is a continuous function of x. To check the continuity of the remaining two terms it suffices to do a local calculation by covering the boundary of V_x by finitely many balls and using a partition of unity. Without loss of generality we may assume that, near a smooth part of the boundary of V_x, V_x is bounded from below by the graph of a continuously differentiable function $H_x : [0,1]^{n-1} \to \mathbb{R}$. We may also assume that m equals Lebesgue measure on $\mathbb{R}^n$. Then

$$\begin{aligned} &\lim_{h\to 0} \frac{1}{|h|} \int_{V_x\setminus(V_{x+h}\cap V_x)} k(y,x)\psi(y)\,dy \\ &\quad = \lim_{h\to 0} \frac{1}{|h|} \int_{[0,1]^{n-1}} \int_{H_x(y_1)}^{H_{x+h}(y_1)} k(y_1, y_2, x)\psi(y)\,dy_2\,dy_1 \\ &\quad = \int_{[0,1]^{n-1}} \frac{\partial}{\partial x} H_x(y_1) \frac{h}{|h|} k(y_1, H_x(y_1), x)\psi(y_1, H_x(y_1))\,dy_1 \end{aligned}$$

is a continuous function of x. The contribution near the finitely many points where V_x is not smooth vanishes in the limit $h \to 0$. Summarizing, $D(L\psi)$ has an expression of the

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form

$$D(L\psi)(x) = \int_{V_x} \frac{\partial}{\partial x} k(y, x)\psi(y)\, dm(y) + \int_{\partial V_x} n(y, x)k(y, x)\psi(y)\, dS(y), \qquad (15)$$

where $n(y, x)$ measures the change of ∂V_x in the direction of the unit normal vector to V_x and dS is the volume on ∂V_x. We remark that the formula is a variant of the transport theorem 7.1.12 and the Gauss theorem 7.2.9 in [**1**]. It follows that $L\psi$ is continuously differentiable if ψ is continuous. Higher-order derivatives are computed inductively. This gives that $L\psi \in C^{k+1}(\mathcal{M})$ for $\psi \in C^k(\mathcal{M})$.

We prove compactness on $C^k(\mathcal{M})$ by modifying the argument in [**35**]. Let $B^k(\mathcal{M})$ be the unit sphere in $C^k(\mathcal{M})$. Consider first $k = 0$. By the Arzela–Ascoli theorem, compactness of L on $C^0(\mathcal{M})$ follows from the following two properties (compare [**48**]):

(a) for all $x \in \mathcal{M}$, $\{|L\psi(x)| \mid \psi \in B^0(\mathcal{M})\}$ is bounded; and

(b) LF is equicontinuous.

For $\psi \in B^0(\mathcal{M})$, $|L\psi(x)| \leq \int_{\mathcal{M}} k(y, x)\, dm(y)$. This is a continuous function of x and hence bounded. This proves the first item. The above computations showing that $L\psi$ is continuously differentiable also show that $\|D(L\psi)(x)\|$ is uniformly bounded on $B^0(\mathcal{M})$. This proves that LF is equicontinuous. Compactness in $C^k(\mathcal{M})$ follows similarly by noting that:

- for all $x \in \mathcal{M}$, $i \leq k$, $\{\|D^i(L\psi)(x)\| \mid \psi \in B^k(\mathcal{M})\}$ is bounded; and
- $\|D^{k+1}(L\psi)(x)\|$ is uniformly bounded on $B^k(\mathcal{M})$.

This completes the proof of Proposition 2.3. □

Remark 2.4. The transfer operator L is compact on the space $\mathcal{L}^2(\mathcal{M})$ of quadratic integrable functions on $\mathcal{M}$ [**46**, §X.2] and [**15**]. The proof of Proposition 2.3, demonstrating that the transfer operator increases regularity of functions, implies that the spectrum of L on $\mathcal{L}^2(\mathcal{M})$ equals that of L on $C^k(\mathcal{M})$.

Remark 2.5. The spectral radius of L is 1, and the eigenvalue 1 occurs with multiplicity m equal to the number of stationary measures. The peripheral spectrum on the unit circle consists of eigenvalues $e^{2\pi i/p}$, $0 \leq i < p$, for each p occurring as the number of connected components of a stationary measure. See [**21**] and [**44**, Theorem V.4.9] for a proof.

PROPOSITION 2.6. *The transfer operator L as a linear map on $C^k(\mathcal{M})$ or $\mathcal{L}^2(\mathcal{M})$ depends continuously on $f \in R^k(\mathcal{M})$.*

Proof. Consider $\tilde{f}$ near f. Write $\tilde{L}$ and L for the corresponding transfer operators. We need to prove that $\tilde{L} - L$ has small norm. Consider the transfer operators operating on $C^k(\mathcal{M})$ (continuity on $\mathcal{L}^2(\mathcal{M})$ is treated analogously). The transfer operator $\tilde{L}$ is given as $\tilde{L}\phi(x) = \int_{\tilde{V}_x} \tilde{k}(y, x)\phi(y)\, dm(y)$. For $\phi \in B^k(\mathcal{M})$, the unit sphere in $C^k(\mathcal{M})$,

$$\tilde{L}\phi(x) - L\phi(x) = \int_{\tilde{V}_x} \tilde{k}(y, x)\phi(y)\, dm(y) - \int_{V_x} k(y, x)\phi(y)\, dm(y)$$

is small, uniformly in x, since $\tilde{k}(y, x)$ is close to $k(y, x)$ on $\tilde{V}_x \cap V_x$ and $\tilde{V}_x$ is close to V_x. The derivative $D(L\phi)$ is given by (15). An analogous formula holds for $D(\tilde{L}\phi)$. Since the functions and sets involved in the two formulas for $D(\tilde{L}\phi)$ and $D(L\phi)$ are close,

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$D(\tilde{L}\phi)(x)$ is uniformly close to $D(L\phi)(x)$. Closeness of higher-order derivatives, up to order k, is treated analogously. Continuity on $\mathcal{L}^2(\mathcal{M})$ is proved analogously. □

Remark 2.7. Consider two nearby random diffeomorphisms f and $\tilde{f}$ from $R^k(\mathcal{M})$. Write L and $\tilde{L}$ for the corresponding transfer operators on $C^k(\mathcal{M})$. Let $\lambda_1, \ldots, \lambda_l$ be a finite set of eigenvalues for L and denote by F the sum of the corresponding generalized eigenspaces. Then $\tilde{L}$ possesses a nearby set of eigenvalues $\tilde{\lambda}_1, \ldots, \tilde{\lambda}_l$. The sum $\tilde{F}$ of the corresponding generalized eigenspaces is a small perturbation of F (in the sense that F and $\tilde{F}$ have nearby bases). See [**29**, Theorem IV.3.16].

3. *Stable random diffeomorphisms*

This section contains the proofs of Theorem 1.6 on the stability of isolated stationary measures and Theorem 1.7 establishing the generic stability of random diffeomorphisms.

For the restriction of random maps to an isolating neighborhood W we consider the transfer operator acting on functions vanishing outside W and at the boundary of W. We write

$$C_0^k(W) = \{f \in C^k(\mathcal{M}) \mid \text{the support of } f \text{ is contained in } \overline{W}\}.$$

Then L acting on $C_0^k(W)$ is well defined. The results in the previous section hold for L acting on $C_0^k(W)$.

Proof of Theorem 1.6. Recall that the closure of $f(W; \Delta)$ is contained in the isolating neighborhood W. This property extends to random diffeomorphisms sufficiently close to f. Restrict the map $x \mapsto f(x; \omega)$ to W and consider the transfer operator L acting on $C_0^k(W)$. Then L has a single eigenvalue 1. Since the spectrum of the transfer operator varies continuously with the random diffeomorphism at f, the transfer operator corresponding to each nearby random diffeomorphism possesses a single eigenvalue 1. The corresponding eigenvector is near ϕ. □

LEMMA 3.1. *Write L_f for the transfer operator on $C^k(\mathcal{M})$ for $f \in R^\infty(\mathcal{M})$. The densities of stationary measures vary continuously with $f \in R^\infty(\mathcal{M})$ at a random diffeomorphism $\bar{f}$ precisely if the multiplicity of the eigenvalue* 1 *for L_f is locally constant in f for f near $\bar{f}$.*

Proof. Consider $\bar{f} \in R^\infty(\mathcal{M})$ with an eigenvalue 1 of multiplicity m. Let $\bar{\mu}_1, \ldots, \bar{\mu}_m$ be the ergodic stationary measures with densities $\bar{\phi}_1, \ldots, \bar{\phi}_m$. Write $F_{\bar{f}}$ as the direct sum of the lines spanned by $\bar{\phi}_1, \ldots, \bar{\phi}_m$. By Remark 2.7, the transfer operator for any $f \in R^\infty(\mathcal{M})$ sufficiently close to $\bar{f}$ possesses an m-dimensional invariant linear space F_f that is the continuation of $F_{\bar{f}}$.

The spectrum of L_f restricted to F_f is in general close to 1. Suppose now that all eigenvalues equal 1. Then $\phi \in F_f$ implies that $L_f\phi = \phi$. Write $\phi = \phi^+ - \phi^-$ with $\phi^+ = \max\{0, \phi\}$ and $\phi^- = \max\{0, -\phi\}$ the positive and negative parts of ϕ. Because $L_f \geq 0$ and L_f preserves the $\mathcal{L}^1$ norm, $(L_f\phi)^+ < L_f\phi^+$ precisely if ϕ^- and ϕ^+ are not invariant. Thus ϕ^+ and ϕ^- are necessarily invariant. It follows that invariant densities are obtained by taking positive parts of invariant eigenfunctions. In this way m invariant densities for f near those of $\tilde{f}$ can be obtained, proving the lemma. □

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Proof of Theorem 1.7. Consider diffeomorphisms on an open neighborhood U of a random diffeomorphism $\bar{f} \in R^\infty(\mathcal{M})$. Write m for the multiplicity of the eigenvalue 1 for the transfer operator corresponding to $\bar{f}$. There is a neighborhood D of 1 in the complex plane, so that, for U small enough, each $f \in U$ has m eigenvalues counting multiplicity in D. Let F denote the m-dimensional invariant linear space corresponding to these eigenvalues. Consider the map that assigns to $f \in U$ the union of the support of all functions in F. By the continuous dependence of F on f, this is a lower semicontinuous set valued mapping and therefore continuous on a set $B_2 \subset U$ of Baire second category [**20**].

Consider the map that assigns to random diffeomorphisms $f \in R^\infty(\mathcal{M})$ the multiplicity $m(f)$ of the eigenvalue 1 for the corresponding transfer map. By the continuous dependence of eigenvalues of the transfer map on f, the map m is upper semicontinuous. Since m takes on finitely many values, it is continuous on an open and dense subset of $R^\infty(\mathcal{M})$. Indeed, consider $A_n = \{f \in R^\infty(\mathcal{M}) \mid m(f) < n\}$. The set of points of continuity of m, in the vicinity of some map in $R^\infty(\mathcal{M})$, equals the intersection of a finite collection of open and dense sets $A_n \cup \left(R^\infty(\mathcal{M}) \backslash \overline{A_n}\right)$, namely with n ranging over a finite set of positive integers.

With reference to Lemma 3.1, the two above items combined prove the theorem. □

4. *Auxiliary parameters*

Consider a smooth one-parameter family of random diffeomorphisms $x \mapsto f_a(x; \omega)$ depending on a from an open interval I in $\mathbb{R}$. The transition map P and its density k depend on a; we write P_a and k_a. The support $U_{x,a}$ of k_a is assumed to vary smoothly with x and a. The density $k_a(x, y)$ is a smooth function of $(a, x, y) \in \bigcup_{a,x} \{a\} \times \{x\} \times U_{x,a}$ in the sense that it can be extended to a smooth function on an open neighborhood. Let L_a denote the transfer operator for f_a, given by

$$L_a\phi(x) = \int_{V_{x,a}} k_a(y, x)\phi(y)\, dm(y). \tag{16}$$

The domain of integration $V_{x,a} = \{y \in \mathcal{M} \mid y \in f_a(x; \Delta)\}$ depends smoothly on (x, a).

PROPOSITION 4.1. *For $r \geq 0$, the transfer operator $\phi \mapsto L_a\phi$ as a map from $C^{k+r}(\mathcal{M})$ into $C^k(\mathcal{M})$ is a C^{r+1} map of a and ϕ.*

Proof. By Proposition 2.6, L_a depends continuously on a. For the derivative of $L_a\phi$ with respect to a we find an expression similar to (15),

$$\frac{\partial}{\partial a} L_a\phi(x) = \int_{V_{x,a}} \frac{\partial}{\partial a} k_a(y, x)\phi(y)\, dm(y) + \int_{\partial V_{x,a}} s_a(y, x)k_a(y, x)\phi(y)\, dS(y) \tag{17}$$

for some smooth function s_a. It follows that, for $\phi \in C^k(\mathcal{M})$, $(\partial/\partial a)L_a\phi \in C^k(\mathcal{M})$. This implies the differentiability of $(\phi, a) \mapsto L_a\phi$ for $\phi \in C^k(\mathcal{M})$. Higher differentiability is treated similarly. □

The operator $L_a : C^k(\mathcal{M}) \to C^k(\mathcal{M})$ does not depend C^2 on a, since $(\partial^2/\partial a^2)L_a\phi$ may not exist if $\phi \in C^0(\mathcal{M})$ and $(\partial^2/\partial a^2)L_a\phi$ is a C^{k-1} function if $\phi \in C^k(\mathcal{M})$. What does hold is that $(x, a) \mapsto L_a\phi_a(x)$ is C^{k+1} if $(x, a) \mapsto \phi_a(x)$ is C^k.

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Proof of Theorem 1.10. Let W be the isolating neighborhood for μ_{a_0}. For a near a_0, $f_a(W;\Delta)$ is strictly contained in W and f_a has a unique stationary measure with support in W. Restrict f_a to W for such values of a_0.

Consider the transfer operator L_a for f_a acting on $C_0^k(W)$. Write F for the line in $C_0^k(W)$ spanned by ϕ_{a_0}. Then $C_0^k(W) = F \oplus H_0^k(W)$ with $H_0^k(W)$ consisting of C^k functions with vanishing integral,

$$H_0^k(W) = \left\{ \phi \in C_0^k(W) \;\middle|\; \int_{\mathcal{M}} \phi(x)\, dm(x) = 0 \right\}.$$

Write ϕ_a for the eigenvectors of L_a continuing ϕ_{a_0} provided by Theorem 1.6. Decompose $\phi_a = \phi_{a_0} + \psi_a$ with $\psi_a \in H_0^k(W)$. Then ψ_a is a solution of $L_a\psi_a = \psi_a + \phi_{a_0} - L_a\phi_{a_0}$. Note that $(x, a) \mapsto \phi_{a_0}(x) - L_a\phi_{a_0}(x)$ is C^∞. The spectrum of $L_{a_0}|_{H_0^k(W)}$ is away from 1. Proposition B.3 in Appendix B implies the result. □

5. *Conditionally stationary measures*

To study average escape times from open sets we make use of conditionally stationary measures, which are measures for which on average a fixed percentage of mass escapes under an iterate. We recall the notion of conditionally invariant measure; see [**12, 16, 26, 38, 39**] for its use in deterministic dynamics. Let a map $f : \mathcal{M} \to \mathcal{M}$ be given and restrict f to a domain $W \subset \mathcal{M}$. Let $V \subset W$ be the set of points in W that are mapped into W; points in the complement of V in W are mapped outside W. Consider $f : V \to W$. A conditionally invariant measure for f on W is a measure μ on $\mathcal{M}$ so that $\mu(A) = \mu(f^{-1}(A))/\mu(f^{-1}(W))$ for Borel sets $A \subset W$.

Definition 5.1. Let $f \in R^\infty(\mathcal{M})$. Let W be an open domain in $\mathcal{M}$. A measure $\bar{\mu}$ on $\overline{W}$ is a *conditionally stationary measure* if

$$\bar{\mu}(A) = \int_W P(x, A)\, d\bar{\mu}(x) \Big/ \int_W P(x, W)\, d\bar{\mu}(x)$$

for Borel sets $A \subset W$.

See [**19, 34**] where this notion is called a quasistationary measure. Note that a conditionally stationary measure is a stationary measure if $\int_W P(x, W)\, d\bar{\mu}(x) = 1$, that is, if the support of the conditionally stationary measure lies inside $\overline{W}$.

LEMMA 5.2. *A measure $\bar{\mu}$ on $\overline{W}$ is a conditionally stationary measure for f if and only if $\bar{\mu} \times \nu^\infty$ is a conditionally invariant measure for (f, ϑ) on $\overline{W} \times \Delta^{\mathbb{N}}$.*

Proof. Write $D^1(x) = \{\boldsymbol{\omega} \in \Delta^{\mathbb{N}} \mid f(x;\boldsymbol{\omega}) \in W\}$. Consider $S : \bigcup_{x\in W}(\{x\} \times D^1(x)) \to W \times \Delta^{\mathbb{N}}$, $S(x, \boldsymbol{\omega}) = (f(x;\omega), \vartheta\boldsymbol{\omega})$. We must show that the following two statements are equivalent:

(i) $\bar{\mu} \times \nu^\infty(S^{-1}(A)) \,/\, \bar{\mu} \times \nu^\infty(S^{-1}(W \times \Delta^{\mathbb{N}})) = \bar{\mu} \times \nu^\infty(A)$ for Borel sets $A \subset W \times \Delta^{\mathbb{N}}$;

(ii) $$\frac{\int_W \int_\Delta 1_U(f(x;\omega))\, d\bar{\mu}(x)\, d\nu(\omega)}{\int_W \int_\Delta 1_W(f(x;\omega))\, d\bar{\mu}(x)\, d\nu(\omega)} = \int_W 1_U(x)\, d\bar{\mu}(x) \text{ for Borel sets } U \subset W.$$

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Take a Borel set $U \times V$ with $U \subset W$ and $V \subset \Delta^{\mathbb{N}}$ and compute

$$\begin{aligned}\bar{\mu} \times \nu^{\infty}(S^{-1}(U \times V)) &= \bar{\mu} \times \nu^{\infty}\left(\bigcup_{\omega \in \Delta} f^{-1}(U; \omega) \times \{\omega\} \times V\right) \\ &= \bar{\mu} \times \nu\left(\bigcup_{\omega \in \Delta} f^{-1}(U; \omega) \times \{\omega\}\right) \nu^{\infty}(V) \\ &= \int_W \int_{\Delta} 1_U(f(x; \omega))\, d\bar{\mu}(x)\, d\nu(\omega)\, \nu^{\infty}(V). \qquad (18)\end{aligned}$$

Further

$$\bar{\mu} \times \nu^{\infty}(U \times V) = \int_W 1_U(x)\, d\bar{\mu}(x)\, \nu^{\infty}(V). \qquad (19)$$

Equations (18) and (19) contain the implication (i) $\Rightarrow$ (ii) when applied for $U \times \Delta^{\mathbb{N}}$ for the enumerator and for $W \times \Delta^{\mathbb{N}}$ for the denominator.

To show that (ii) implies (i), note that (18) and (19) show that (i) holds for Borel sets A of the form $U \times V$ if (ii) is assumed. Therefore it holds for all Borel sets in $W \times \Delta^{\mathbb{N}}$. □

We continue with the introduction of transfer operators whose fixed points are the densities of conditionally stationary measures. The transfer operator $\bar{L}$, defined for functions in $\mathcal{L}^1(\overline{W})$ with integral 1, is given by

$$\bar{L}(\phi) = 1_{\overline{W}} L\phi \Big/ \int_W L\phi(x)\, dm(x). \qquad (20)$$

We write $\tilde{L}\phi = 1_{\overline{W}} L\phi$.

PROPOSITION 5.3. *The transfer operator $\tilde{L}$ maps $C^0(\overline{W})$ into itself and is a compact operator on it.*

Proof. If $k(x, y)$ denotes the density of the stochastic transition function $P(x, \cdot)$, then

$$\tilde{L}\phi(x) = \int_{\overline{W} \cap V_x} k(y, x)\phi(y)\, dm(y).$$

Note that $\overline{W} \cap V_x$ depends continuously on x. Recall from the proof of Proposition 2.3 that we must show that:

- for all $x \in \overline{W}$, $\{|\tilde{L}\psi(x)| \mid \psi \in B^0(\overline{W})\}$ is bounded; and
- $\tilde{L}F$ is equicontinuous.

Here $B^0(\overline{W})$ is the unit sphere in $C^0(\overline{W})$. The first item follows as before: $|\tilde{L}\psi(x)| \leq \int_W k(y, x)\, dm(y)$ is bounded by a continuous function and thus bounded. For the second item we must show that for each $\epsilon > 0$ there is $\delta > 0$ so that, for all $x \in \overline{W}$, $\psi \in B^0(\overline{W})$, we have $|\tilde{L}\psi(x + h) - \tilde{L}\psi(x)| < \epsilon$ if $|h| < \delta$. Recall, see (14), that

$$\begin{aligned}\tilde{L}\psi(x + h) - \tilde{L}\psi(x) = &\int_{W \cap V_{x+h} \cap V_x} (k(y, x + h) - k(y, x))\psi(y)\, dm(y) \\ &+ \int_{W \cap V_{x+h} \setminus (W \cap V_{x+h} \cap V_x)} k(y, x + h)\psi(y)\, dm(y) \\ &- \int_{W \cap V_x \setminus (W \cap V_{x+h} \cap V_x)} k(y, x)\psi(y)\, dm(y).\end{aligned}$$

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Now

$$\left|\int_{W\cap V_{x+h}\cap V_x}(k(y,x+h)-k(y,x))\psi(y)\,dm(y)\right| \le \int_{W\cap V_{x+h}\cap V_x}|k(y,x+h)-k(y,x)|\,dm(y),$$

which is small for $|h|$ small by uniform continuity of k. In addition,

$$\left|\int_{W\cap V_{x+h}\setminus(W\cap V_{x+h}\cap V_x)}k(y,x+h)\psi(y)\,dm(y)\right| \le \int_{W\cap V_{x+h}\setminus(W\cap V_{x+h}\cap V_x)}|k(y,x+h)|\,dm(y)$$

is small for $|h|$ small by the boundedness of k and uniform continuity of the volume of V_x in x. Similarly for the third term. This proves equicontinuity. □

Recall from Proposition 2.6 that L depends continuously on the random diffeomorphism. The same argument shows that $\tilde{L}$ depends continuously on the random diffeomorphism.

PROPOSITION 5.4. *The transfer operator $\tilde{L}$ as a linear map on $C^0(\mathcal{M})$ depends continuously on $f \in R^k(\mathcal{M})$.*

We obtain conditionally stationary measures by a perturbation argument, perturbing from an invariant measure. We do not develop general existence results for conditionally stationary measures, as such general results are not needed for our purposes. Let $\{f_a\}$ be a family of random diffeomorphisms depending on a real parameter $a \in I$. Consider, for $a_0 \in I$, a stationary density ϕ_{a_0} with support E_{a_0}. Let W be a neighborhood of E_{a_0} disjoint from the supports of possible other stationary densities of f_{a_0}.

PROPOSITION 5.5. *For a close to a_0, f_a possesses a conditionally stationary density $\bar{\phi}_a$ on W, with $\bar{\phi}_{a_0} = \phi_{a_0}$ and $(x,a) \mapsto \bar{\phi}_a(x)$ continuous in (x,a). One has*

$$\tilde{L}\bar{\phi}_a = \alpha(a)\bar{\phi}_a,$$

where $\alpha(a) = \int_W L_a\bar{\phi}_a(x)\,dm(x)$ depends continuously on a.

Proof. The operator $\tilde{L}_a$ varies continuously with a and therefore possesses a single eigenvalue close to 1 for a close to a_0. The function $\bar{\phi}_a$ is the corresponding eigenfunction. □

Recall the definition of the escape time $\chi_a(x,\boldsymbol{\omega})$ for $x \in W$ and $\boldsymbol{\omega} \in \Delta^{\mathbb{N}}$:

$$\chi_a(x,\boldsymbol{\omega}) = \min\{k \mid f_a^k(x;\boldsymbol{\omega}) \notin W\}.$$

LEMMA 5.6. *We have*

$$\int_W\int_{\Delta^{\mathbb{N}}}\chi_a(x;\boldsymbol{\omega})\,dv^\infty(\omega)\,\bar{\phi}_a(x)\,dm(x) = \frac{1}{1-\alpha(a)}.$$

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Proof. Let $S_a(x, \boldsymbol{\omega}) = (f_a(x; \omega), \vartheta\boldsymbol{\omega})$. Write

$$D_a^n = \{(x, \boldsymbol{\omega}) \mid f_a^n(x; \omega_1, \ldots, \omega_n) \in W\}$$

for the set of points in $W \times \Delta^{\mathbb{N}}$ that remain in $W \times \Delta^{\mathbb{N}}$ for n iterates of S_a. The exit set E_a^n of points that leave $W \times \Delta^{\mathbb{N}}$ in n iterates equals $D_a^{n-1} \setminus D_a^n$. Thus $\chi_a(x, \boldsymbol{\omega}) = n$ on E_a^n. Write $\bar{\mu}_a$ for the conditionally stationary measure with density $\bar{\phi}_a$. From (18) with $A = W \times \Delta^{\mathbb{N}}$ we get

$$\begin{aligned}\alpha(a) &= \int_W P(x, W)\, d\bar{\mu}_a(x) \\ &= \int_W \int_\Delta 1_W(f(x; \omega))\, d\nu(\omega)\, d\bar{\mu}_a(x) \\ &= \bar{\mu}_a \times \nu^\infty(S_a^{-1}(W \times \Delta^{\mathbb{N}})).\end{aligned}$$

It follows that $\bar{\mu}_a \times \nu^\infty(E_a^k) = \alpha^{k-1} - \alpha^k = \alpha^{k-1}(1 - \alpha)$. Calculate

$$\begin{aligned}\int_W \int_{\Delta^{\mathbb{N}}} \chi_a(x; \boldsymbol{\omega})\, d\nu^\infty(\omega)\, d\bar{\mu}_a(x) &= \sum_{k=1}^{\infty} \bar{\mu}_a \times \nu^\infty(E_a^n) \\ &= \sum_{k=1}^{\infty} k\alpha^{k-1}(1 - \alpha) \\ &= \frac{1}{1 - \alpha}.\end{aligned}$$

□

As a corollary we obtain that the average escape time from W goes to infinity as $a \to a_0$. More precise estimates are derived in the following section.

PROPOSITION 5.7. *We have that* $\int_W \int_{\Delta^{\mathbb{N}}} \chi_a(x, \boldsymbol{\omega})\, d\nu^\infty(\boldsymbol{\omega})\, dm(x)$ *converges to* ∞ *as* $a \to a_0$.

Proof. Let E_a be the interior of the support of $\bar{\phi}_a$:

$$\begin{aligned}\int_W \int_{\Delta^{\mathbb{N}}} \chi_a(x, \boldsymbol{\omega})\, d\nu^\infty(\boldsymbol{\omega})\, dm(x) &\geq \int_{E_a} \int_{\Delta^{\mathbb{N}}} \frac{\chi_a(x, \boldsymbol{\omega})}{\bar{\phi}_a(x)}\, d\nu^\infty(\boldsymbol{\omega}) \bar{\phi}_a(x)\, dm(x) \\ &\geq C \int_{E_a} \int_{\Delta^{\mathbb{N}}} \chi_a(x, \boldsymbol{\omega})\, d\nu^\infty(\boldsymbol{\omega}) \bar{\phi}_a(x)\, dm(x) \\ &= \frac{C}{1 - \alpha(a)},\end{aligned}$$

for $C = 1/\max\{x \in W \mid \bar{\phi}_a(x)\}$. For x from the support E_{a_0}, the image $f_{a_0}(x; \Delta)$ is contained in E_{a_0} by the invariance of E_{a_0}. By the continuity of f_a, $f_a(x; \Delta) \subset W$ for a close enough to a_0. Since $\bar{\phi}_a$ depends continuously on a, $1 - \alpha(a) \leq \int_{W \setminus E_{a_0}} \bar{\phi}_a(x)\, dm(x)$ converges to 0 as $a \to a_0$. The proposition follows. □

6. *Escape times*

In this section estimates for the average escape time from small neighborhoods of the support of a stationary measure that undergoes a bifurcation are derived, as a function of the unfolding parameter.

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Let $\{(f_a, g_a)\}$, $a \in I$, be a smooth one-parameter family of random diffeomorphisms on $\mathcal{M}$. Suppose that $a_0 \in I$ is a bifurcation value for $\{(f_a, g_a)\}$. Let μ be a stationary measure for f_{a_0} involved in a bifurcation. Let W be a small neighborhood of E. By Proposition 5.5, for W sufficiently close to E and a near a_0, $\{(f_a, g_a)\}$ possesses a unique conditionally stationary measure $\bar{\mu}_a$ with support in W. Write $\bar{\phi}_a$ for the density of $\bar{\mu}_a$. The transfer operator $\bar{L}_a$ acting on $C^0(\overline{W})$ has $\bar{\phi}_a$ as a unique fixed point.

Let X^0 be the set of points in $\overline{W}$ with $\partial V_{x,a} \cap \partial W \neq \emptyset$ for $x \in X^0$. For $i \geq 0$, define $X^{i+1} = f(X^i; \partial\Delta)$. We suppress the dependence of X^i on a from the notation.

LEMMA 6.1. *If $x_i \in X^i$, then for $x_{i+1} \in f(x_i; \partial\Delta)$ one has $x_i \in \partial V_{x_{i+1},a}$.*

Proof. This is clear from the definition. □

At $x \in X^0$, the boundary of $V_{x,a} \cap W$ varies continuously but not smoothly with (x, a). It follows that $\bar{L}_a(\phi)$ cannot be expected to be more than continuous on X^0 even for smooth ϕ.

LEMMA 6.2. *Suppose that $(x, a) \mapsto \phi_a(x)$ is C^k outside $X^0 \cup \cdots \cup X^{k-1}$, such that derivatives up to order k are bounded and their restrictions to a component of $W \backslash (X^0 \cup \cdots \cup X^{k-1})$ extend continuously to the boundary of the component. Then $(x, a) \mapsto \bar{L}_a\phi_a(x)$ is C^{k+1} outside $X^0 \cup \cdots \cup X^k$. Likewise, derivatives up to order $k+1$ are bounded and their restrictions to a component of $W \backslash (X^0 \cup \cdots \cup X^k)$ extend continuously to the boundary of the component.*

Proof. For $x \notin X^0$, the derivative of $\tilde{L}_a\phi$ is of the form

$$D(\tilde{L}_a\phi)(x) = \int_{V_{x,a}\cap W} \frac{\partial}{\partial x} k_a(y, x)\phi(y)\, dm(y) + \int_{\partial(V_{x,a}\cap W)} n_a(y, x)k_a(y, x)\phi(y)\, dS(y), \tag{21}$$

for a smooth function k_a and a piecewise smooth function n_a (smooth outside the intersection of ∂W with $\partial V_{x,a}$). This identity shows that $\tilde{L}_a\phi$ is C^1 with bounded derivatives outside X^0, for any $\phi \in C^0(\overline{W})$. The same holds for $\bar{L}_a(\phi)$. Higher-order derivatives are treated inductively. Similar to (21) one has

$$\frac{\partial}{\partial a}\tilde{L}_a\phi(x) = \int_{V_{x,a}\cap W} \frac{\partial}{\partial a} k_a(y, x)\phi(y)\, dm(y) + \int_{\partial(V_{x,a}\cap W)} s_a(y, x)k_a(y, x)\phi(y)\, dS(y) \tag{22}$$

for some piecewise smooth function s_a. This shows that $(x, a) \mapsto \tilde{L}_a\phi_a(x)$ is C^1 outside X^0 for continuous functions $(x, a) \mapsto \phi_a(x)$. The derivatives of $(x, a) \mapsto \tilde{L}_a\phi_a(x)$ on $W \backslash X^0$ are bounded; moreover the derivatives on a component of $W \backslash X^0$ extend continuously to the boundary of the component.

Higher-order derivatives are treated inductively. Suppose that $(x, a) \mapsto \phi_a(x)$ is C^k outside $X^0 \cup \cdots \cup X^{k-1}$, such that derivatives up to order k are bounded and their restrictions to a component of $W \backslash (X^0 \cup \cdots \cup X^{k-1})$ extend continuously to the boundary of the component. Then $(x, a) \mapsto \tilde{L}_a\phi_a(x)$ is C^{k+1} outside $X^0 \cup \cdots \cup X^k$. Likewise, derivatives up to order $k+1$ are bounded and their restrictions to a component of $W \backslash (X^0 \cup \cdots \cup X^k)$ extend continuously to the boundary of the component.

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The transfer operator $\bar{L}_a$ is the composition of the linear map $\tilde{L}_a$ and the projection

$$\Pi(\phi) = \phi \Big/ \int_W \phi(x)\, dm(x).$$

The projection Π is a smooth map which is well defined near ϕ_{a_0} in $C^0(\overline{W})$; a direct computation shows that

$$D\Pi(\phi)h = \frac{1}{\int_W \phi(x)\, dm(x)} h - \frac{\phi}{(\int_W \phi(x)\, dm(x))^2} \int_W h(x)\, dm(x).$$

Also,

$$\frac{\partial^i}{\partial a^i} \int_W \phi_a(x)\, dm(x) = \int_W \frac{\partial^i}{\partial a^i} \phi_a(x)\, dm(x).$$

This implies that also $(x, a) \mapsto \bar{L}_a(\phi_a)(x)$ is C^{k+1} outside $X^0 \cup \cdots \cup X^k$ and has bounded derivatives. □

PROPOSITION 6.3. *For each $k \geq 1$, $\bar{\phi}_a$ is C^k outside $X^0 \cup \cdots \cup X^{k-1}$ jointly in (x, a), and the derivatives up to order k are uniformly bounded.*

Proof. Proposition 5.5 gives that $\bar{\phi}_a(x)$ is continuous in (x, a). Recall that the support E of μ consists of finitely many, say k, connected components, permuted cyclically by the random diffeomorphism. An iterate of f_{a_0} thus maps each component into itself. The restriction of the transfer operator $L^k_{a_0}$ to a small neighborhood of E has a single eigenvalue 1 and a remaining spectrum strictly inside the unit circle [**21**]. There is therefore no loss in generality to assume that the support of μ consists of a single connected component E, which we will assume for the remainder of the proof.

We write $H^k(\overline{W}) = \{\psi \in C^k(\overline{W}) \mid \int_W \psi = 0\}$. Define the operator $T_a : H^0(\overline{W}) \to H^0(\overline{W})$ by

$$T_a(\psi) = \bar{L}_a(\phi_{a_0} + \psi) - \phi_{a_0}. \tag{23}$$

Decompose $\bar{\phi}_a = \bar{\phi}_{a_0} + \bar{\psi}_a$, so that $T_a(\bar{\psi}_a) = \bar{\psi}_a$. From the proof of Lemma 6.2, we get that T_a is a smooth map on $H^0(\overline{W})$:

$$DT_a(\psi) = D\Pi(\tilde{L}_a(\phi_{a_0} + \psi))\tilde{L}_a.$$

However, $\tilde{L}_a$ maps continuously differentiable functions to continuous functions, so that T_a does not define a map from $H^k(\overline{W})$, $k \geq 1$, to itself. As a consequence we cannot obtain smoothness properties of $\bar{\phi}_a$ by applying the implicit function theorem. To get the smooth dependence of $\bar{\phi}_a$ outside sets X^i we reason as follows. We derive equations that the derivatives of $\bar{\phi}_a$ must satisfy, establish that the equations can be solved, and show that the solutions are the derivatives of $\bar{\phi}_a$. The reasoning follows the lines of the proof of Proposition B.3 in Appendix B.

To prove that $\bar{\psi}_a$ varies C^1 with a in points outside X^0, note that $(\partial/\partial a)\bar{\psi}_a(x)$ should be a solution $M_a(x)$ to

$$\frac{\partial}{\partial a} T_a(\bar{\psi}_a)(x) + DT_a(\bar{\psi}_a)M_a(x) = M_a(x). \tag{24}$$

We claim that this equation is uniquely solvable. The spectral radius of $DT_{a_0}(0) = L_{a_0}$ is smaller than 1. As a consequence of the continuous dependence of $\tilde{L}_a$ on a (see

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Proposition 5.4), also $DT_a(\psi_a)$ varies continuously with a. For a sufficiently close to a_0, the spectral radius of $DT_a(\bar{\psi}_a)$ is therefore also smaller than 1. Hence

$$(I - DT_a(\bar{\psi}_a))^{-1} = I + \sum_{i=1}^{\infty} (DT_a(\bar{\psi}_a))^i; \tag{25}$$

see [**29**]. This formula can be applied for $DT_a(\bar{\psi}_a)$ acting on $\mathcal{L}^2$ functions. Indeed, DT_a is compact on $\mathcal{L}^2(\overline{W}) \subset \mathcal{L}^1(\overline{W})$, see Remark 2.4, and has spectrum strictly inside the unit circle in $\mathbb{C}$. From

$$\begin{aligned} M_a(x) &= (I - DT_a(\bar{\psi}_a))^{-1} \frac{\partial}{\partial a} T_a(\bar{\psi}_a)(x) \\ &= (I + DT_a(\bar{\psi}_a) + (DT_a(\bar{\psi}_a))^2 + (DT_a(\bar{\psi}_a))^3 + \cdots) \frac{\partial}{\partial a} T_a(\bar{\psi}_a)(x), \end{aligned} \tag{26}$$

we get that M_a is continuous outside X^0 since it equals the sum of $(\partial/\partial a)T_a(\bar{\psi}_a)$ and a uniform limit of continuous functions (compare Lemma 6.2). In particular M_a is uniformly bounded and has continuous extensions to the closure of components of $W \backslash X^0$. We must show that

$$\left|\bar{\psi}_{a+h}(x) - \bar{\psi}_a(x) - M_a(x)h\right| = o(|h|)$$

for $x \notin X^0$, as $h \to 0$. Consider $\gamma_a(x) = \bar{\psi}_{a+h}(x) - \bar{\psi}_a(x)$ for $x \notin X^0$. Then

$$\begin{aligned} \gamma_a(x) &= T_{a+h}(\bar{\psi}_a + \gamma_a)(x) - T_a(\bar{\psi}_a)(x) \\ &= DT_a(\bar{\psi}_a)\gamma_a(x) + \frac{\partial}{\partial a} T_a(\bar{\psi}_a)(x)h + R(x), \end{aligned} \tag{27}$$

where

$$R(x) = T_{a+h}(\bar{\psi}_a + \gamma_a)(x) - T_a(\bar{\psi}_a)(x) - DT_a(\bar{\psi}_a)\gamma_a(x) - \frac{\partial}{\partial a} T_a(\bar{\psi}_a)(x)h.$$

We claim that for any $\epsilon > 0$ there is $\delta > 0$ so that $|R| < \epsilon(|\gamma_a| + |h|)$, if $|h|$ and $|\gamma_a|$ are smaller than δ. Since $\gamma_a(x)$ is continuous in h we may further restrict δ in this estimate so that $|R| < \epsilon(|\gamma_a| + |h|)$ holds for $|h|$ smaller than δ. Further,

$$(I - DT_a(\bar{\psi}_a))\gamma_a(x) = \frac{\partial}{\partial a} T_a(\bar{\psi}_a)(x)h + R(x).$$

Using (25) and the bound on $|R|$ gives $|\gamma_a| \le k|h|$ for some k if $|h| < \delta$. Therefore $|R| < \epsilon(1 + k)|h|$ for some $k > 0$, if $|h| < \delta$. Now (24) and (27) give

$$\gamma_a(x) - M_a(x)h = (I - DT_a(\bar{\psi}_a))^{-1} R(x).$$

Using (25) it follows that $|\gamma_a - M_a h| = o(|h|)$, $h \to 0$. This proves that M_a equals the partial derivative $(\partial/\partial a)\bar{\psi}_a$.

Higher-orders of differentiability are proved by induction. Assume that $(x, a) \mapsto \bar{\psi}_a(x)$ has been shown to be C^k outside $X^0 \cup \cdots \cup X^{k-1}$. Recall from Lemma 6.2 that, for C^k maps $(x, a) \mapsto \bar{\psi}_a(x)$, $(\partial/\partial a)\bar{L}_a(\bar{\psi}_a)$ is C^k outside $X^0 \cup \cdots \cup X^k$. The right-hand side of (26) is therefore C^k outside $X^0 \cup \cdots \cup X^k$. The above reasoning shows that $M_a = (\partial/\partial a)\bar{\psi}_a$ outside $X^0 \cup \cdots \cup X^k$. Therefore $(\partial/\partial a)\bar{\psi}_a$ is C^k outside $X^0 \cup \cdots \cup X^k$. Also $D\bar{\psi}_a = D(T_a(\bar{\psi}_a))$ is C^k outside $X^0 \cup \cdots \cup X^k$, so that $(x, a) \mapsto \bar{\psi}_a(x)$ is C^{k+1} outside $X^0 \cup \cdots \cup X^k$. The same clearly holds for $(x, a) \mapsto \bar{\phi}_a(x)$. □

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Proof of Theorem 1.13. We repeat the computation in the proof of Proposition 5.7. Let E_a be the interior of the support of $\bar{\phi}_a$. Applying Lemma 5.6,

$$\begin{aligned}\int_W \int_{\Delta^{\mathbb{N}}} \chi_a(x,\boldsymbol{\omega})\, d\nu^{\infty}(\boldsymbol{\omega})\, dm(x) &\geq \int_{E_a}\int_{\Delta^{\mathbb{N}}} \frac{\chi_a(x,\boldsymbol{\omega})}{\bar{\phi}_a(x)}\, d\nu^{\infty}(\boldsymbol{\omega})\, \bar{\phi}_a(x)\, dm(x)\\ &\geq C\int_{E_a}\int_{\Delta^{\mathbb{N}}} \chi_a(x,\boldsymbol{\omega})\, d\nu^{\infty}(\boldsymbol{\omega})\, \bar{\phi}_a(x)\, dm(x)\\ &= \frac{C}{1-\alpha(a)},\end{aligned}$$

for $C = 1/\max\{x \in \overline{W} \mid \bar{\phi}_a(x)\}$. By Proposition 6.3, $(x,a) \mapsto \bar{\phi}_a(x)$ is C^k almost everywhere and has uniformly bounded derivatives. For each integer k there is a constant C with $|\bar{\phi}_a| \leq C|a-a_0|^k$ on $\overline{W \setminus E_{a_0}}$. As in the proof of Proposition 5.7 we get that for each k there is a constant $C_k > 0$, so that $1-\alpha(a) \leq C_k|a-a_0|^k$. □

7. *Decay of correlations*

Consider a random family $\{f_a\}$ restricted to an isolating neighborhood W of a stationary measure μ_a, for all values of a from an interval I. The transfer operator L_a on $C_0^k(W)$ possesses a single eigenvalue at 1. If the support of μ_a consists of r components, L_a has eigenvalues $e^{2\pi i/j}$, $0 \leq j < r$, on the unit circle in the complex plane. These eigenvalues make up the peripheral spectrum of L_a; see Remark 2.5. In this section we consider bifurcations in which the number of components of the support of μ_a changes. We will see how the rate of decay of correlations varies with the parameter a, providing a proof of Theorem 1.15.

PROPOSITION 7.1. *Let $\{f_a\}$, $a \in I$, be a family of random diffeomorphisms with an isolating neighborhood W. The eigenvalues and eigenvectors of the peripheral spectrum of L_a on $C_0^k(W)$ vary smoothly with a.*

Proof. Let λ_a be an eigenvalue that depends continuously on a and lies on the unit circle for $a = a_0$. Since μ_{a_0} is an isolated stationary measure, λ_{a_0} is a simple eigenvalue (see Remark 2.5). Proposition B.3 in Appendix B implies the result. □

Recall (11) and Lemma 2.1. Write

$$\begin{aligned}L_a^n\varphi(x) &= \int_{\Delta^n} P_{f_a(x;\omega_1,\ldots,\omega_n)}\varphi(x)\, d\nu(\omega_1)\cdots d\nu(\omega_n),\\ U_a^n\psi(x) &= \int_{\Delta^n} \psi\circ f_a(x;\omega_1,\ldots,\omega_n)\, d\nu(\omega_1)\cdots d\nu(\omega_n).\end{aligned}$$

As in the computation for Lemma 2.1,

$$\int_{\mathcal{M}} L_a^n\varphi(x)\psi(x)\, dm(x) = \int_{\mathcal{M}} \varphi(x) U_a^n\psi(x)\, dm(x). \tag{28}$$

After these preparations we now prove the statements on the speed of decay of correlations. First consider a single random map f.

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Proof of Proposition 1.14. Let ϕ be the stationary density. Write

$$L^n\varphi = \left(\int_{\mathcal{M}} \varphi(y)\, dm(y)\right)\phi + R^n\varphi.$$

Compute

$$\begin{aligned}\int_{\mathcal{M}} \varphi(x)U^n\psi(x)\, dm(x) &= \int_{\mathcal{M}} L^n\varphi(x)\psi(x)\, dm(x)\\ &= \int_{\mathcal{M}}\left[\left(\int_{\mathcal{M}} \varphi(y)\, dm(y)\right)\phi(x) + R^n\varphi(x)\right]\psi(x)\, dm(x),\end{aligned}$$

so that

$$\begin{aligned}&\left|\int_{\mathcal{M}} \varphi(x)U^n\psi(x)\, dm(x) - \int_{\mathcal{M}} \varphi(x)\, dm(x)\int_{\mathcal{M}} \psi(x)\phi(x)\, dm(x)\right|\\ &\quad= \left|\int_{\mathcal{M}} R^n\varphi(x)\psi(x)\, dm(x)\right|.\end{aligned}$$

Note that the spectral radius of R is smaller than 1. By the continuity of R, there is $N > 0$ so that $\|R^N\| < 1$ for all a near a_0. Hence for $n \in \mathbb{N}$, one has $\|R^n\| < C\eta^n$ for some $C > 0$ and $\eta < 1$. The proposition follows from

$$\left|\int_{\mathcal{M}} R^n\varphi(x)\psi(x)\, dm(x)\right| \le \|R^n\varphi\|_{\mathcal{L}^2(\mathcal{M})}\|\psi\|_{\mathcal{L}^2(\mathcal{M})} \le C\eta^n\|\varphi\|_{\mathcal{L}^2(\mathcal{M})}\|\psi\|_{\mathcal{L}^2(\mathcal{M})}. \quad \square$$

Proof of Theorem 1.15. This is proved by following the computation in the proof of Proposition 1.14 above and noting that L_a has, for $a > a_0$, a single eigenvalue 1 and $k-1$ eigenvalues that have moved smoothly into the unit circle. Write η_a for the largest radius of the eigenvalues of L_a that lie inside the unit circle. As a consequence of the smooth dependence of the eigenvalues near the unit circle, see Proposition 7.1, η_a is a smooth function of a.

We claim that there exists $C > 0$ so that, for all a near a_0, $\|R^n\| \le C\eta_a^n$. For $a = a_0$, let E be the union of the eigenspaces for the eigenvalues in the peripheral spectrum. For a near a_0, let E_a be the continuation of $E_{a_0} = E$. As E_a is finite dimensional and has a basis depending smoothly on a, it is clear that there exists $C > 0$ so that, for $a > a_0$, $n \in \mathbb{N}$,

$$\|R^n\big|_{E_a}\| \le C\eta_a^n. \tag{29}$$

Let F be a subspace of $\mathcal{L}^2(W)$ complementary to E. Write P_a for the projection to E_a along F. Then $R = P_aR + (I - P_a)R$. By the continuity of R and P_a, there is $N > 0$ so that $\|\ ((I - P_a)R)^N\ \| < 1$ for all a near a_0. Hence for $n \in \mathbb{N}$,

$$\|\ ((I - P_a)R)^n\ \| < C\nu^n \tag{30}$$

for some $C > 0$ and $\nu < 1$. Now (29) and (30) prove the claim. As before, the proposition follows from

$$\left|\int_{\mathcal{M}} R^n\varphi(x)\psi(x)\, dm(x)\right| \le \|R^n\varphi\|_{\mathcal{L}^2(\mathcal{M})}\|\psi\|_{\mathcal{L}^2(\mathcal{M})} \le C\eta_a^n\|\varphi\|_{\mathcal{L}^2(\mathcal{M})}\|\psi\|_{\mathcal{L}^2(\mathcal{M})}. \quad \square$$

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8. *One-dimensional random maps*

The most complete description of bifurcations in smooth random maps is derived for random maps in one dimension. Consider a random differentiable map $f(x;\omega)$ on the circle $\mathbb{S}^1$. The random parameter ω is drawn from $\Delta=[-1,1]$. Recall that we assume (H2), an injective dependence of $f(\cdot;\omega)$ on ω, for the implementation of the noise. What is proved below for random differentiable maps on the circle holds with obvious modifications for random differentiable maps on a compact interval that is mapped inside itself by all differentiable maps.

A pathological example occurs if $f(x;\omega)$ is constant in x; the (unique) stationary measure is then a push forward of the measure on Δ. To avoid pathologies we assume the open and dense condition that:

(H3) the critical points of each map $x\mapsto f(x;\omega)$ have finite order.

Also under this condition one finds that the regularity of stationary measures for random differentiable maps is substantially less than for random diffeomorphisms; their densities are only continuous.

THEOREM 8.1. *The random differentiable map $f\in R^\infty(\mathbb{S}^1)$ possesses a finite number of ergodic stationary measures $\mu_1,\dots,\mu_m$ with mutually disjoint supports $E_i,\dots,E_m$. All stationary measures are linear combinations of $\mu_1,\dots,\mu_m$.*

The support E_i of μ_i consists of the closure of a finite number of connected open sets $C_i^1,\dots,C_i^p$ that are moved cyclically by $f(\cdot;\Delta)$. The density ϕ_i of μ_i is a C^0 function on $\mathbb{S}^1$.

Proof. The condition on the critical points of $x\mapsto f(x;\omega)$ implies that V_x varies continuously with x. The reasoning used to prove Proposition 2.3 shows that the transfer operator L maps $\mathcal{L}^1(\mathbb{S}^1)$ into $C^0(\mathbb{S}^1)$. This implies continuity of invariant densities. □

THEOREM 8.2. *Let μ be an isolated ergodic stationary measure of $f\in R^\infty(\mathbb{S}^1)$ with density ϕ. Then each $\tilde f\in R^\infty(\mathbb{S}^1)$ sufficiently close to f possesses a unique ergodic stationary measure $\tilde\mu$ with support in V. The density $\tilde\phi$ of $\tilde\mu$ is C^0 close to ϕ.*

Proof. This is similar to the proof of Theorem 1.6. We repeat the remark that V_x varies continuously with x due to the condition on the critical points of $x\mapsto f(x;\omega)$. Arguments like in the proofs of Propositions 2.3 and 5.3 show that L is a compact operator on $C^0(\mathcal{M})$. □

Recall that iterates of a random map f are defined through (4). A periodic point $\bar x$ of period k is a point satisfying $f^k(\bar x;\omega_1,\dots,\omega_k)=\bar x$ for some $\omega_1,\dots,\omega_k\in\Delta$. It is hyperbolic if $(d/dx)f^k(x;\omega_1,\dots,\omega_k)$ at $x=\bar x$ differs from $0,1,-1$. By the implicit function theorem, a family $\{f_a\}$ of random differentiable maps with $f_{a_0}=f$ possesses a hyperbolic periodic point $\bar x_a$, $\bar x_{a_0}=\bar x$, for a near a_0 and for the same values of $\omega_1,\dots,\omega_k$, depending smoothly on a.

THEOREM 8.3. *The set of stable random differentiable maps in $R^\infty(\mathbb{S}^1)$ is open and dense.*

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Proof. Take $f \in R^\infty(\mathbb{S}^1)$. If the entire circle is the support of a stationary measure of f, then f is stable by Theorem 8.2. Suppose that μ is a stationary measure whose support E is a union $\bigcup_{i=0}^{k-1} C_i$ of intervals C_i mapped cyclically by $f(\cdot;\Delta)$: $f(C_i;\Delta) = C_{i+1}$ (the indices are taken modulo k). If μ is an isolated measure, then f restricted to an isolating neighborhood of E is stable.

The measure μ is certainly isolated if, for each boundary point $x \in E$, either

$$f^k(x;\Delta^{\mathbb{N}}) \subset \text{interior } E$$

or

$$f^k(x;\omega_1,\dots,\omega_k) = x, \quad f^j(x;\omega_1,\dots,\omega_j) \in \text{interior } E$$

for some $\omega_1,\dots,\omega_k \in \Delta$, $j < k$. Indeed, invariance of E shows that in both cases $f^k(y;\Delta^{\mathbb{N}}) \in E$ for any y near x.

If not all boundary points are as above, then there is a boundary point $x \in \partial E$ so that $f^l(x;\omega_1,\dots,\omega_l) = x$ for $l = k$ or $l = 2k$ (l minimal) and $f^j(x;\omega_1,\dots,\omega_j) \in \partial E$ for $0 < j < l$. Write $x_0 = x$ and $x_j = f^j(x;\omega_1,\dots,\omega_j)$ for $j > 0$. From $x_j \in \partial E$, $x_{j+1} = f(x_j;\omega_{j+1}) \in \partial E$ and $(\partial/\partial\omega) f(\cdot;\omega) \neq 0$, we see that $\omega_{j+1} \in \partial\Delta$. Thus $\omega_1,\dots,\omega_l$ are all contained in $\partial\Delta$. Note that $(d/dx) f^l(x;\omega_1,\dots,\omega_l) \geq 0$ since otherwise x is an interior point of E.

For $f \in R^\infty(\mathbb{S}^1)$, there are a neighborhood U of f and an integer N so that, for each $\tilde{f} \in U$, the support of the union of its stationary measures has at most N connected components. A random periodic orbit in the boundary of the support of a stationary measure of $\tilde{f} \in U$ therefore has its period bounded by $2N$.

By transversality techniques a number of arbitrary small perturbations of f are carried through. The perturbations affect $f(\cdot;\omega)$ for $\omega \in \partial\Delta$ and can be extended to other values of ω using test functions. We will not present the detailed perturbations, but refer to [**36**, §III.2] for a description of the techniques. By a small perturbation of f we may assume that the graph of each map $f^i(\cdot;(\partial\Delta)^i)$, $1 < i \leq 2N$, intersects the diagonal in $\mathbb{S}^1 \times \mathbb{S}^1$ transversally. That is,

(H4) the periodic orbits of period $i \leq 2N$ for $f(\cdot;\partial\Delta)$ are hyperbolic.

There is then a bounded number of random periodic orbits with period bounded by $2N$. A further small perturbation ensures that:

(H5) each periodic point x of period $i \leq 2N$ is periodic for only one sequence $\omega_1,\dots,\omega_i \in \partial\Delta$.

Write $\mathcal{P}$ for the points in these periodic orbits. Recall that the number of critical points of $f(\cdot;\partial\Delta)$ is finite. A final small perturbation ensures that:

(H6) the critical values of $f(\cdot;\partial\Delta)$ are disjunct from $\mathcal{P}$.

Conditions (H4), (H5), (H6) are clearly open and thus describe an open and dense subset of $R^\infty(\mathbb{S}^1)$.

Consider f from this open and dense set. Let μ be a stationary measure of f with support E. Let x be a boundary point of E belonging to a periodic orbit in ∂E. By (H4), x belongs to a hyperbolic periodic orbit. By (H5), there is a unique graph $f^l(\cdot,\omega_1,\dots,\omega_l)$

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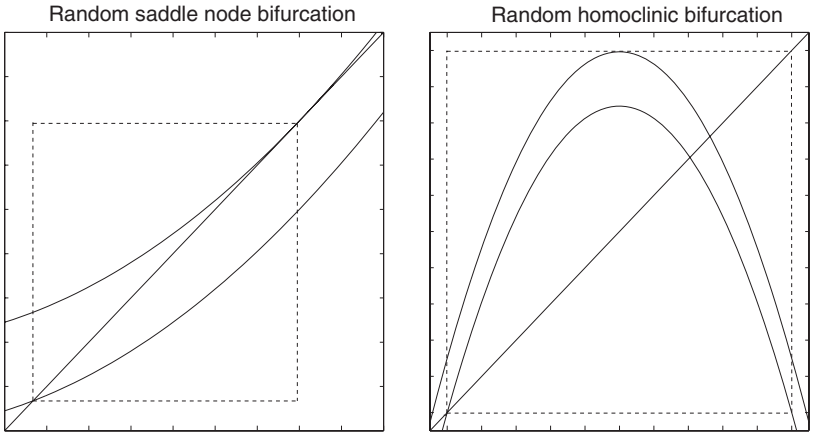


FIGURE 2. Consider a random map f_{a_0} for which points are mapped randomly into the region bounded by the two graphs. Depicted on the left are the graphs of a random map $f(\cdot;\omega)$, $\omega \in \partial\Delta$, with a random saddle node bifurcation. The support of the stationary density is the interval between the hyperbolic fixed point of the lower map and the saddle node fixed point of the upper map. The right picture shows graphs of a random map with a random homoclinic bifurcation. Here the support of the stationary density stretches from the left hyperbolic fixed point of the lower map to the critical value of the upper map.

with $\omega_1, \ldots, \omega_l \in \partial\Delta$ through x. It is not possible that $(d/dx)f^l(x;\omega_1,\ldots,\omega_l) > 1$, since other orbits would then be repelled and x would not be in the boundary of E. Hence $0 < (d/dx)f^l(x;\omega_1,\ldots,\omega_l) < 1$: the random periodic orbit through x is an attracting periodic orbit for $f^l(\cdot;\omega_1,\ldots,\omega_l)$. By (H6), there are no interior points in E being mapped onto x under iterates of f. As a consequence, μ is isolated. Therefore f is stable. □

As a next step we consider one parameter families of random maps and show that bifurcations typically occur at isolated parameter families. Theorem 8.7 below moreover describes the possible codimension-one bifurcations. The space of smooth families of smooth random maps $x \mapsto f_a(x;\omega)$, $a \in I$, will be given the uniform C^k topology as maps $(x,\omega,a) \mapsto f_a(x;\omega)$ on $\mathbb{S}^1 \times \Delta \times I$.

We start with a description of three types of bifurcations (see Figures 2 and 3) caused by violation of one of the conditions (H4), (H5), (H6). These are proved to be the only codimension-one bifurcations.

Definition 8.4. The smooth one-parameter family of random differentiable maps f_a on the circle undergoes a *random saddle node bifurcation* at $a = a_0$, if there exists $\bar{x}$ in the boundary of the support of a stationary measure such that

$$f^k_{a_0}(\bar{x};\omega_1,\ldots,\omega_k) = \bar{x}, \quad \frac{d}{dx} f^k_{a_0}(\bar{x};\omega_1,\ldots,\omega_k) = 1 \tag{31}$$

for some $\omega_1, \ldots, \omega_k \in \partial\Delta$. The random saddle node bifurcation is said to *unfold generically* if

$$\left(\frac{d}{dx}\right)^2 f^k_{a_0}(\bar{x};\omega_1,\ldots,\omega_k) \neq 0, \quad \frac{\partial}{\partial a} f^k_a(\bar{x};\omega_1,\ldots,\omega_k) \neq 0 \tag{32}$$

at $a = a_0$.

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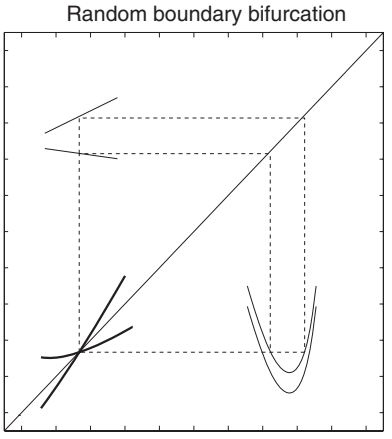


FIGURE 3. Depicted are parts of the graphs of a random map $f(\cdot;\omega)$, $\omega \in \partial\Delta$. The solid curves lie on two of the graphs of $f(f(\cdot;\omega_1);\omega_2)$, $\omega_1, \omega_2 \in \partial\Delta$, intersecting in a point that lies on two hyperbolic period-two orbits (one stable, one unstable) distinguished by different ω values. A random boundary bifurcation results if this point lies on the boundary of the support of a stationary measure.

Definition 8.5. The smooth one-parameter family of random differentiable maps f_a on the circle undergoes a *random homoclinic bifurcation* at $a = a_0$, if there exists

(i) a stationary measure μ with support E with a hyperbolic periodic point $\bar{x}_a$ in the boundary of E for all a near a_0, and

(ii) a critical point $\bar{y}_a$ for $f_a(\cdot;\omega_1)$, $\omega_1 \in \partial\Delta$, in the interior of E,

such that

$$f^l_{a_0}(\bar{y}_{a_0};\omega_1,\ldots,\omega_l) = \bar{x}_{a_0} \tag{33}$$

for some $\omega_2,\ldots,\omega_l \in \partial\Delta$. The random homoclinic bifurcation *unfolds generically* if

$$\frac{\partial}{\partial a}(f^l_a(\bar{y}_a;\omega_1,\ldots,\omega_l) - \bar{x}_a) \neq 0 \tag{34}$$

at $a = a_0$.

Definition 8.6. The smooth one-parameter family of random differentiable maps f_a on the circle undergoes a *random boundary bifurcation* at $a = a_0$, if there exists $\bar{x}$ in the boundary of the support of a stationary measure and $(\omega_1,\ldots,\omega_k) \neq (\tilde{\omega}_1,\ldots,\tilde{\omega}_k) \in (\partial\Delta)^k$, such that

$$f^k_{a_0}(\bar{x};\omega_1,\ldots,\omega_k) = \bar{x}, \quad \frac{d}{dx} f^k_{a_0}(\bar{x};\omega_1,\ldots,\omega_k) \in (0,1) \tag{35}$$

and

$$f^k_{a_0}(\bar{x};\tilde{\omega}_1,\ldots,\tilde{\omega}_k) = \bar{x}, \quad \frac{d}{dx} f^k_{a_0}(\bar{x};\tilde{\omega}_1,\ldots,\tilde{\omega}_k) \in (1,\infty). \tag{36}$$

Write $\bar{x}_a(\omega_1,\ldots,\omega_k)$ and $\bar{x}_a(\tilde{\omega}_1,\ldots,\tilde{\omega}_k)$ for the continuations of the hyperbolic periodic points. The random boundary bifurcation is said to *unfold generically* if

$$\frac{\partial}{\partial a} f^k_a(\bar{x}_a(\omega_1,\ldots,\omega_k);\omega_1,\ldots,\omega_k) \neq \frac{\partial}{\partial a} f^k_a(\bar{x}_a(\tilde{\omega}_1,\ldots,\tilde{\omega}_k;)\tilde{\omega}_1,\ldots,\tilde{\omega}_k) \tag{37}$$

at $a = a_0$.

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For an open interval I, write $R^k(I, \mathcal{M})$ for the space of C^k families of random maps in $R^k(\mathcal{M})$ depending on a parameter in I. Equip the space $R^k(I, \mathcal{M})$ with the C^k topology.

THEOREM 8.7. *For $\{f_a\}$ from an open and dense subset of $R^\infty(I, \mathbb{S}^1)$, f_a has only finitely many bifurcations. A bifurcation point is a random saddle node bifurcation, a random homoclinic bifurcation, or a random boundary bifurcation and is generically unfolding. If the number of stationary measures is locally constant at a bifurcation point, the bifurcation is an intermittency bifurcation. Otherwise the bifurcation is a transient bifurcation.*

Proof. For a bifurcation value for a family in $R^\infty(I, \mathbb{S}^1)$, either (H4), (H5) or (H6) is violated.

Similar transversality arguments as in the proof of Theorem 8.3 show the following. For an open and dense subset of $R^\infty(I, \mathbb{S}^1)$, at most one of these conditions is violated at a bifurcation value and the resulting bifurcation unfolds generically as stated in Definitions 8.4, 8.5 or 8.6. Since the random bifurcations are unfolding generically, they occur isolated. □

9. *Case studies*

In this section we illustrate the general theory on two examples: a randomized version of standard circle diffeomorphisms and a randomized version of logistic maps on the interval. We explain how random saddle node bifurcations occur in both examples and random homoclinic bifurcations in random logistic maps. For the random circle diffeomorphisms we consider rotation numbers and study their dependence on parameters. The reader is referred to [**36**] for the theory of deterministic circle and interval maps.

9.1. *Random circle diffeomorphisms.*

The standard circle map acting on $x \in \mathbb{R}/\mathbb{Z}$ and depending on parameters a and ε is given by

$$f_a(x) = x + a + \frac{\varepsilon}{2\pi} \sin(2\pi x) \mod 1.$$

Consider f_a for a fixed value of $\varepsilon \in (0, 1)$ for which f_a is a diffeomorphism. Introduce the lift $F_a : \mathbb{R} \mapsto \mathbb{R}$,

$$F_a(x) = x + a + \frac{\varepsilon}{2\pi} \sin(2\pi x).$$

It is well known that the rotation number ρ_a of f_a,

$$\rho_a = \lim_{k\to\infty} \frac{F_a^k(x) - x}{k}, \tag{38}$$

is well defined and independent of x. The rotation number depends continuously on a. The rotation number is rational precisely if f_a possesses periodic orbits. For a fixed rational number r, the rotation number of f_a equals r for an interval of a values. In the interior of such an interval, f_a has exactly one hyperbolic periodic attractor and one hyperbolic periodic repeller; see [**37**].

In the following we consider standard circle diffeomorphisms with a random parameter,

$$f_a(x; \omega) = x + \frac{\varepsilon}{2\pi} \sin(2\pi x) + a + \sigma\omega, \tag{39}$$

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for $x \in \mathbb{R}/\mathbb{Z}$ and a random parameter ω chosen from a uniform distribution on $\Delta = [-1, 1]$. The value of σ determines the amplitude of the noise; we assume it has a fixed value. We consider fixed $\varepsilon \in (0, 1)$ for which $x \mapsto f_a(x; \omega)$ is a diffeomorphism. Write

$$F_a(x; \omega) = x + a + \sigma\omega + \frac{\varepsilon}{2\pi} \sin(2\pi x) \tag{40}$$

for the lift of $f_a(x; \omega)$. Note that $F_a(x; \omega) - x$ is periodic in x with period one.

PROPOSITION 9.1. *For each parameter value a, the random standard circle family* $\{f_a\}$ *has a unique stationary measure* μ_a. *The density* ϕ_a *of* μ_a *is smooth and depends smoothly on a. The support of* μ_a *is either the entire circle or finitely many intervals strictly contained in the circle. The latter possibility is only possible if* ρ_b *is rational for each* $b \in [a - \sigma, a + \sigma]$. *Bifurcations where the support of* μ_a *changes discontinuously are generic saddle node bifurcations. There are finitely many such bifurcations.*

Remark 9.2. Observe that f_a has a hyperbolic fixed point for $a \in (-\varepsilon/2\pi, \varepsilon/2\pi)$. Hence, f_a has a stationary measure supported on a single interval precisely if both $a - \sigma > -\varepsilon/2\pi$ and $a + \sigma < \varepsilon/2\pi$ (see Figure 4). This occurs for a non-empty interval of a values if $\sigma < \varepsilon/2\pi$.

Proof of Proposition 9.1. It is well known that a circle diffeomorphism with irrational rotation number has its orbits lying dense in $\mathbb{R}/\mathbb{Z}$. It follows that if the family of circle maps $f_a(x; \omega)$ for varying $\omega \in \Delta$ contains a member with irrational rotation number, there is a (necessarily unique) stationary measure supported on all of $\mathbb{R}/\mathbb{Z}$.

Suppose now that $f_a(x; \omega)$ for each $\omega \in \Delta$ has rational rotation number $\rho_{a+\sigma\omega} = p/q$. Write x_ω for a periodic point from a periodic attractor of $x \mapsto f_a(x; \omega)$ depending continuously on ω. Recall that $x \mapsto f_a(x; \omega)$ has a unique periodic attractor. Let $V_a = \bigcup_{\omega\in\Delta} x_\omega = [x_{-1}, x_1]$. The random standard family is increasing in x and in ω, so that for all $x \in V_a$ and all $\boldsymbol{\omega} \in \Delta^{\mathbb{N}}$ we have

$$x_{-1} = F_a^q(x_{-1}; -1) \le F_a^q(x_{-1}; \boldsymbol{\omega}) \le F_a^q(x; \boldsymbol{\omega}) \le F_a^q(x_1; \boldsymbol{\omega}) \le F_a^q(x_1; 1) = x_1.$$

It follows that the orbit of V_a is invariant. For a fixed $\omega \in \Delta$, all points outside the unique periodic repeller of $f_a(\cdot; \omega)$ are attracted to its periodic attractor. This implies that there is a unique stationary measure supported on the orbit of V_a.

Compute

$$\frac{\partial}{\partial a} f_a^k(x; \boldsymbol{\omega}) = \sum_{i=0}^{k} \frac{\partial}{\partial a} f(f^i(x; \boldsymbol{\omega}; \boldsymbol{\omega})) \frac{d}{dx} f_a^i(f_a^{k-i}(x; \boldsymbol{\omega}; \boldsymbol{\omega})).$$

As all terms in the sum are positive, a random saddle node bifurcation occurs isolated. The random family $\{f_a\}$ therefore has only a finite number of random saddle node bifurcations. □

We define the rotation number for the random standard circle map, when its exist, by

$$\rho_a(x; \boldsymbol{\omega}) = \lim_{k\to\infty} \frac{F_a^k(x; \boldsymbol{\omega}) - x}{k}. \tag{41}$$

The rotation number measures the average rotation per iterate of f_a. Note that ρ_a is a random variable, depending also on the starting point x.

A simple but useful lemma shows that ρ_a is independent of the initial condition x.

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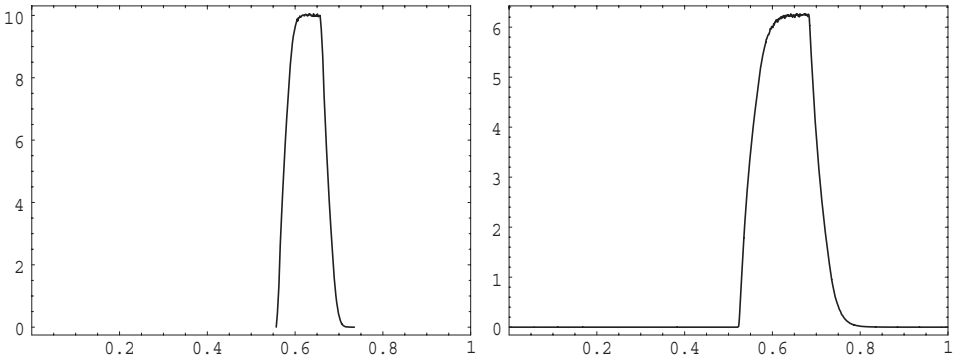


FIGURE 4. Numerically computed stationary densities of the random standard circle map, on the left for $|a| + \sigma < \varepsilon/2\pi$, and on the right for $|a| + \sigma > \varepsilon/2\pi$. The explosion of the support of the stationary density follows a random saddle node bifurcation.

LEMMA 9.3. *If $\rho_a(x; \boldsymbol{\omega})$ exists for some $x \in \mathbb{R}/\mathbb{Z}$, $\boldsymbol{\omega} \in \Delta^{\mathbb{N}}$, then $x \mapsto \rho_a(x; \boldsymbol{\omega})$ exists for all $x \in \mathbb{R}/\mathbb{Z}$ and is constant in x.*

Proof. Observe that $F_a^k(\cdot; \omega)$ is a lift of $f_a^k(\cdot; \boldsymbol{\omega})$, so that $F_a^k(x; \omega) - x$ is periodic in x with period one. Thus

$$\max_{x\in\mathbb{R}}\{F_a^k(x; \omega) - x\} - \min_{x\in\mathbb{R}}\{F_a^k(x; \omega) - x\} < 1.$$

Compute

$$\begin{aligned} |F_a^k(x; \boldsymbol{\omega}) - F_a^k(y; \boldsymbol{\omega})| &\le |(F_a^k(x; \boldsymbol{\omega}) - x) - (F_a^k(y; \boldsymbol{\omega}) - y)| + |x - y| \\ &\le 1 + |x - y|, \end{aligned}$$

so that

$$\lim_{k\to\infty}\left(\frac{F_a^k(x; \boldsymbol{\omega}) - x}{k} - \frac{F_a^k(y; \boldsymbol{\omega}) - y}{k}\right) = 0.$$

It follows that the limit

$$\lim_{k\to\infty}\frac{F_a^k(x; \omega) - x}{k},$$

if it exists, is independent of x. □

Write

$$F_a(x; \boldsymbol{\omega}) = x + \delta_a(x; \boldsymbol{\omega}),$$

where the function $\delta_a(x; \boldsymbol{\omega})$ is periodic with period one in the variable x. We can consider δ_a as a function defined on $\mathbb{R}/\mathbb{Z}$. A simple induction argument gives, for each $k \in \mathbb{N}$,

$$f_a^k(x; \boldsymbol{\omega}) = x + \sum_{i=0}^{k-1} \delta_a \circ S^i(x; \boldsymbol{\omega}), \tag{42}$$

where S is the skew product system (see equation (5)) on $\mathbb{R}/\mathbb{Z} \times \Delta^{\mathbb{N}}$. Recall that $\mu_a \times \nu^\infty$ is an S-invariant measure.

Rotation numbers for random circle maps are treated by Ruffino [**43**]; this work includes a proof of the existence of rotation numbers, also part of the following proposition. Figure 5 gives an impression of the rotation number ρ_a as a function of a.

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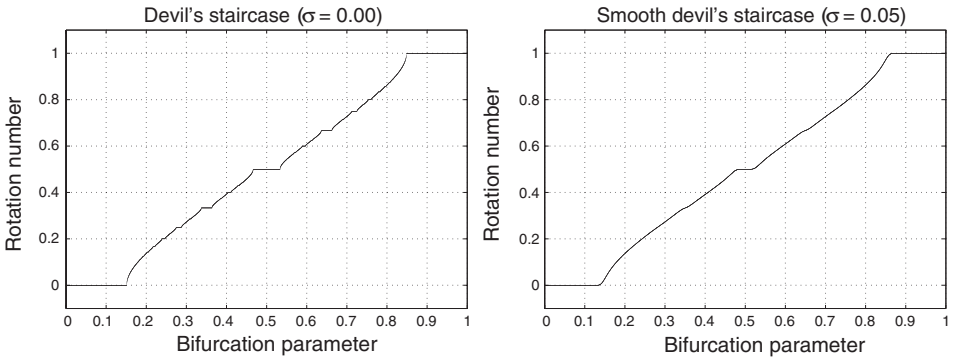


FIGURE 5. The function $a \mapsto \rho_a$. On the left, the devil's staircase; the rotation number of the deterministic standard family. On the right, the rotation number of the random standard family.

PROPOSITION 9.4. *We have*

$$\rho_a = \int_{\mathbb{R}/\mathbb{Z}} \mathbb{E}(\delta_a(x;\omega))\, d\mu_a(s) \qquad \nu^\infty\text{-a.s.}, \tag{43}$$

where $\mathbb{E}$ is the expectation operator. The right-hand side of (43) is independent of x and is a smooth and non-decreasing function of a.

Proof. Consider the Birkhoff sum in equation (42),

$$\frac{f_a^k(x;\omega) - x}{k} = \frac{1}{k}\sum_{i=0}^{k-1} \delta_a \circ S^i(s;\omega).$$

By Birkhoff's ergodic theorem,

$$\begin{aligned}\lim_{k\to\infty} \frac{f_a^k(x;\omega) - x}{k} &= \int_{\mathbb{R}/\mathbb{Z}} \int_\Omega \beta_a(s;\omega)\mu_a(ds)\nu^\infty(d\omega) \qquad \mu_a \times \nu^\infty\text{-a.s.} \\ &= \int_{\mathbb{R}/\mathbb{Z}} \mathbb{E}(\beta_a(s;\omega))\mu_a(ds) \qquad \mu_a \times \nu^\infty\text{-a.s.} \end{aligned} \tag{44}$$

The fact that the rotation number when it exists is independent of the initial point, see Lemma 9.3, implies that this equality holds for all x and ν^∞-a.s.

Write $a \mapsto h(a)$ for the right-hand side of (43). Smoothness of h follows from smoothness of the stationary density ϕ_a and (44). Write the standard family as $R_a \circ f(.;\omega)$ where $f(.;\omega)$ is the random map

$$f(x;\omega) = x + \frac{\varepsilon}{2\pi}\sin(2\pi x) + \sigma\xi(\omega)$$

and R_a is the translation with coefficient a. Then for $a_1 < a_2$ and $k \geq 1$, $(R_{a_1} \circ f(.;\omega))^k < (R_{a_2} \circ f(.;\omega))^k$ and thus

$$\rho_{a_1} = \lim_{k\to\infty} \frac{(R_{a_1} \circ f(.;\omega))^k - \mathrm{Id}}{k} \leq \lim_{k\to\infty} \frac{(R_{a_2} \circ f(.;\omega))^k - \mathrm{Id}}{k} = \rho_{a_2}. \qquad \square$$

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9.2. *Random unimodal maps.* This section is devoted to the investigation of the randomized version of the logistic family

$$f_a(x;\omega) = (a+\sigma\omega)x(1-x) \tag{45}$$

on $[0, 1]$. The random parameter ω is chosen from a uniform distribution on $\Delta = [-1, 1]$. Note that f_a is not a random diffeomorphism and also that $x = 0, 1$ are fixed points for all values of ω. Throughout this section we will assume that

$$a + \sigma\omega \in (1, 4), \tag{46}$$

for all $\omega \in \Delta$. As a consequence, the interval $[0, 1]$ is mapped into itself by each map $x \mapsto f_a(x;\omega)$ and the fixed point at the origin is repelling. Any stationary measure will therefore have support contained in $(0, 1)$. We will demonstrate that there is only one stationary measure.

PROPOSITION 9.5. *The random logistic map f_a has a unique stationary measure.*

Proof. We collect some facts from unimodal dynamics needed in the sequel of the proof. The following facts hold for unimodal maps with negative Schwarzian derivative such as the logistic map $x \mapsto ax(1-x)$. By Guckenheimer's theorem (see [**36**, Theorem III.4.1]) $x \mapsto ax(1-x)$ possesses a unique attractor Λ_a. The attractor Λ_a is either a periodic attractor, a solenoidal attractor, or a finite union of intervals on which the map acts transitively. In all cases, the omega-limit set of the critical point c (with $c = \frac{1}{2}$ for the logistic map) is contained in Λ_a. In fact, if Λ_a is not a periodic attractor, then c is contained in Λ_a. It follows from a result of Misiurewicz (see [**36**, Theorem III.3.2]) that the basin of attraction of Λ_a is an open and dense subset of $(0, 1)$.

We will distinguish the following two cases:
Case (i) there exists $\omega \in \Delta$, so that $c \in \Lambda_{a+\sigma\omega}$;
Case (ii) otherwise.
The two cases are treated separately.

Case (i). Write

$$W = \bigcap_{n\geq 0} \overline{\bigcup_{i\geq n} f_a^i(c;\, \Delta^{\mathbb{N}})}$$

for the omega-limit set of c under all possible random iterations. Observe that W is an invariant set. From the properties of the noise, W consists of a finite union of intervals. Note that $c \in W$, so that W equals the closure of the positive orbit $\overline{\bigcup_{i\geq 0} f_a^i(c;\, \Delta^{\mathbb{N}})}$ of c under all possible random iterations. We will prove that for each $x \in (0, 1)$, $y \in W$ and $\varepsilon > 0$, there exist $n > 0$ and $\bar{\boldsymbol{\omega}} \in \Delta^{\mathbb{N}}$ with the property that

$$|f_a^n(x;\bar{\boldsymbol{\omega}}) - y| < \varepsilon. \tag{47}$$

This implies that W is the unique minimal invariant set, which in turn implies the theorem in the first case.

Fix $x \in (0, 1)$, $y \in W$, $\varepsilon > 0$. From the construction of W, there exist $\boldsymbol{\omega}_1 \in \Delta^{\mathbb{N}}$, $i > 0$, so that $|f_a^i(c;\boldsymbol{\omega}_1) - y| < \varepsilon$. By the continuity of $x \mapsto f_a^i(x;\boldsymbol{\omega}_1)$, the same holds with c replaced by a point from a δ neighborhood of c for some $\delta > 0$. We need to establish

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the existence of $\hat{\boldsymbol{\omega}} \in \Delta^{\mathbb{N}}$ and $j > 0$ so that $|f_a^j(x; \hat{\boldsymbol{\omega}}) - c| < \delta$. Let $\omega_2 \in \Delta$ be such that $c \in \Lambda_{a+\sigma\omega_2}$. Since the basin of attraction of $\Lambda_{a+\sigma\omega_2}$ is open and dense, there exists $\omega_3 \in \Delta$ with $x_1 = f_a(x; \omega_3)$ contained in the basin of attraction of $\Lambda_{a+\sigma\omega_2}$. For i large, $x_i = f_a^{i-1}(x_1; \omega_2, \omega_2, \ldots)$ is as close as desired to $\Lambda_{a+\sigma\omega_2}$. If $\Lambda_{a+\sigma\omega_2}$ is a finite union of intervals, we get that x_i is contained in $\Lambda_{a+\sigma\omega_2}$ for large enough i. As inverse images of c for $x \mapsto f_a(x; \omega_2)$ are dense in $\Lambda_{a+\sigma\omega_2}$, one deduces that there exist $\boldsymbol{\omega}_4 \in \Delta^{\mathbb{N}}$ and $k > 0$ so that $f_a^k(x_{i+1}; \boldsymbol{\omega}_4)$ lies in a δ neighborhood of c. Indeed, if $\Lambda_{a+\sigma\omega_2}$ is a finite union of intervals, then we find ω_5 with $x_{i+1} = f(x_i; \omega_5)$ equal to an inverse image of c. If $\Lambda_{a+\sigma\omega_2}$ is a solenoidal attractor, then $x \mapsto f_a(x; \omega_2)$ is infinitely renormalizable. In this case one can use $\boldsymbol{\omega}_4 = (\omega_2, \omega_2, \ldots)$. Also for a periodic attractor $\Lambda_{a+\sigma\omega_2}$ containing c one uses $\boldsymbol{\omega}_4 = (\omega_2, \omega_2, \ldots)$.

Case (ii). By Guckenheimer's theorem, $x \mapsto f_a(x; \omega)$ possesses a unique periodic attractor for each $\omega \in \Delta$. Write V for the union of $\Lambda_{a+\sigma\omega}$ over $\omega \in \Delta$. Define

$$W = \overline{\bigcup_{i \geq 0} f_a^i(V; \Delta^{\mathbb{N}})}.$$

This is clearly an invariant set. Note that we do not claim that c is outside of W. Arguments as before prove (47) with this definition of W: for suitable noise one finds an orbit starting at x that approaches a point in V and then with further iterates approaches $y \in W$. □

Remark 9.6. The above proof applies to show that a random unimodal map $g(x; \omega)$ with negative Schwarzian derivative for each ω (the invoked theorem by Guckenheimer is true for these maps) has a unique stationary measure.

Perturbing away from the deterministic logistic family one sees that both random saddle node bifurcations and random homoclinic bifurcations occur in the random logistic family $\{f_a\}$ for small noise levels. Typically one can expect the following scenario. We start by recalling some facts concerning the dynamics of the deterministic map $f_a(\cdot; 0)$. The map $f_a(\cdot; 0)$ is called *renormalizable* if there exists an interval I and a positive integer q, so that $f_a^q(I; 0) \subset I$. Let $[a_-, a_+]$ be a maximal interval so that $f_a(\cdot; 0)$ is renormalizable for $a \in [a_-, a_+]$ with q constant. Then f_a undergoes a saddle node bifurcation at $a = a_-$ involving a periodic orbit of period q. At $a = a_+$, f_a undergoes a homoclinic bifurcation, where an iterate of f_a maps the critical point onto a periodic orbit of period q. For small noise levels (i.e. σ small) one expects a random saddle node bifurcation near $a = a_-$ and a random homoclinic bifurcation near $a = a_+$. Figure 6 illustrates this by computing Birkhoff averages for the logistic family and the random logistic family. See [**26**] for explanations of the computations for the logistic family.

A. *Appendix. Representations of discrete Markov processes*

In this appendix we explore the relation between random maps and discrete Markov processes given by stochastic transition functions. The random maps considered in this paper depend on random parameters, where the number of random parameters equals the dimension n of the state space $\mathcal{M}$. Proposition A.1 gives a wide class of Markov processes that can be represented by random maps by n random parameters. The Markov process

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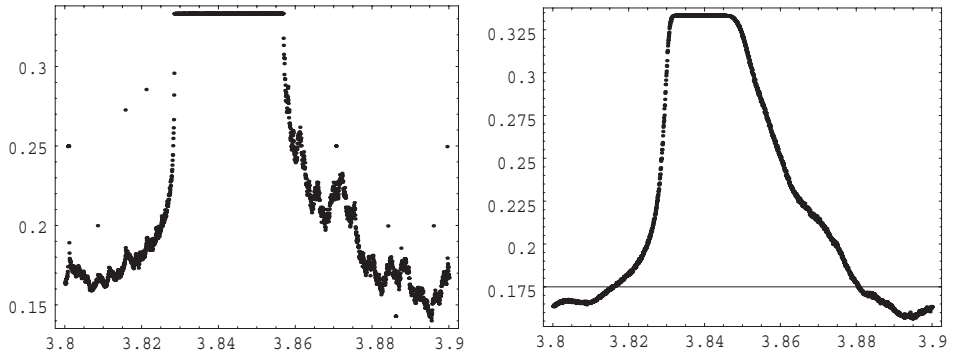


FIGURE 6. Numerically computed Birkhoff averages $\lim_{n\to\infty}(1/n)\sum_{i=0}^{n-1}\phi(f_a^i(x;\boldsymbol{\omega}))$ of $\phi(x)=1_{[0.4,0.6]}$ for the logistic family (left picture) and the random logistic family with $\sigma=0.005$ (right picture), for parameters a ranging from 3.8 to 3.9. The flat part in the left picture, where the average equals $1/3$, runs from a saddle node bifurcation to a homoclinic bifurcation. The numerical computations show that for the random logistic family these are replaced by their random versions; for parameters from the flat part the stationary measure is supported on three disjoint intervals cycled by the random map.

given by random maps depending on a larger number of random parameters (or even given by some measure on the space of maps) can be represented by random maps with n random parameters.

Iterating a random map involves more random parameters obtained by independent draws at each iterate. By means of an example we explain how random maps with a smaller number of random parameters may be brought into the context of this paper. Consider the delayed logistic map $x_{n+1}=\mu x_n(1-x_{n-1})$. Let $y_{n+1}=x_n$. This defines a dynamical system $(x_{n+1}, y_{n+1})=(\mu x_n(1-y_n), x_n)$. Assume now that μ is a random parameter varying in some interval with some distribution. This yields a random map

$$f(x, y; \mu)=(\mu x(1-y), x).$$

The derivative Df is singular along $x=0$. As μ is a single random parameter, this random diffeomorphism does not fit into the context considered in this paper. Considering two iterates gives two independent draws (μ, ν) of the random parameter (that is, random parameters taken from a square) and yields the random map

$$f^2(x, y; \mu, \nu)=(\mu x(1-y), \nu\mu x(1-x)(1-y)).$$

If x and y stay away from 0 and 1, the map and the dependence of (μ, ν) are injective. The second iterate of the delayed logistic map with bounded parametric noise fulfills the assumptions in this paper.

There are other examples of maps with parametric noise that cannot be made to fulfill the assumptions used in this paper. For instance, random maps $f(x;\omega)=x+(x-\omega)^2$ with random ω from an interval fail to satisfy the injectivity assumption of $\omega\mapsto f(x;\omega)$. If ω is chosen from a uniform distribution, then the density of the transition function will not be bounded. Figure 7 indicates a random boundary bifurcation for a similar random map.

Consider discrete Markov processes given by transition functions $P(x,\cdot)$. The following properties hold.

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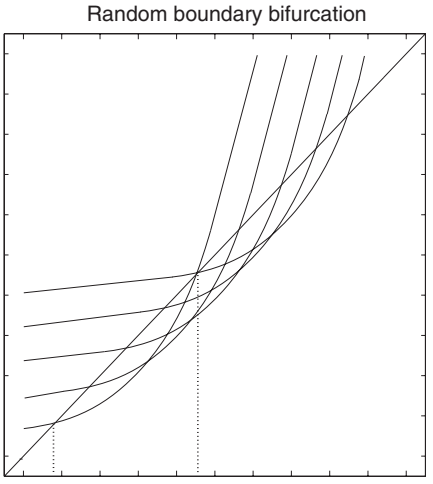


FIGURE 7. A random map $f(x; \omega) = f(x - \omega; 0) + \omega$ with a random boundary bifurcation of the stationary measure with support between the ordinates indicated by dotted lines. If ω is chosen from a uniform distribution, then the density of the transition function will not be bounded.

(i) For fixed $A \in \mathcal{B}$, $x \mapsto P(x, A)$ is measurable.
(ii) For fixed $x \in \mathbb{R}^n$, $P(x, \cdot)$ is a probability measure.

Denote by $y \mapsto k(x, y)$ the density of $P(x, \cdot)$. Write U_x for the support of $k(x, \cdot)$ and let $U = \bigcup_x (\{x\} \times U_x)$. Assume that U_x is diffeomorphic to the closed unit ball Δ in $\mathbb{R}^n$ and varies smoothly with x. We will assume that $y \mapsto k(x, y)$ depends smoothly on $(x, y) \in U$, meaning that k can be extended to a smooth function defined on an open neighborhood of U. Under these conditions we will construct a representation by a finitely parameterized family of differentiable maps. That is, we will construct a family of differentiable maps $\{f_\mu\}$ on $\mathbb{R}^n$, with parameters μ from an n-dimensional ball, and a measure ν on the parameter space so that $P(x, A)$ equals $\nu\{\mu \in \Delta \mid f_\mu(x) \in A\}$. A corresponding result holds for discrete Markov processes with noise from an n-dimensional box; see [**10**, Appendix D]. See [**30**] for a discussion of the existence of representations by sets of measurable or continuous maps. The paper [**41**] contains a result on representations by differentiable maps, under the assumption of unbounded noise.

PROPOSITION A.1. *There is a family of differentiable maps f_μ, $\mu \in \Delta$, and a measure ν on Δ with smooth strictly positive density, so that:*

(1) *$(x, \mu) \mapsto f_\mu(x)$ is smooth;*
(2) *for each $x \in \mathcal{M}$, $\mu \mapsto f_\mu(x)$ is injective; and*
(3) $P(x, A) = \nu(\mu \in \Delta \mid f_\mu(x) \in A)$.

Proof. We follow the arguments in [**10**], combined with the use of polar coordinates to map the unit ball Δ to $[0, 1]^n$. Let $\psi_x : V_x \to \Delta$ be a diffeomorphism, depending smoothly on x, from the support V_x of $y \mapsto k(x, y)$ to the unit ball. Consider polar coordinates

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$\chi : [0, 1]^n \to \Delta$ on the unit ball,

$$\chi(\xi_1, \ldots, \xi_n) = \xi_1 \begin{pmatrix} \cos(\pi\xi_2) \\ \sin(\pi\xi_2)\cos(\pi\xi_3) \\ \vdots \\ \sin(\pi\xi_2)\cdots\sin(\pi\xi_{n-1})\cos(2\pi\xi_n) \\ \sin(\pi\xi_2)\cdots\sin(\pi\xi_{n-1})\sin(2\pi\xi_n) \end{pmatrix}.$$

For $\xi \in [0, 1]^n$, define sets $B_i(\xi)$, $0 \le i \le n$, by

$$B_i(\xi) = \prod_{j=1}^{i} [0, \xi_j] \times \prod_{j=i+1}^{n} [0, 1].$$

Write $C_i(\xi) = \psi_x^{-1}\chi^{-1}(B_i(\xi))$ and let $\omega = (\omega_1, \ldots, \omega_n)$ be given by

$$\omega_i = \int_{C_i(\xi)} k(x, y)\, dm(y) \Big/ \int_{C_{i-1}(\xi)} k(x, y)\, dm(y).$$

Since $k > 0$, $\omega = \Theta(\xi)$ gives a one-to-one correspondence. Let $\eta_i = \int_{C_i(\xi)} dm(y) / \int_{C_{i-1}(\xi)} dm(y)$. Here $\eta = \Psi(\xi)$ is a one-to-one correspondence. The correspondence $\omega \to \eta$ is a smooth diffeomorphism as k is smooth and strictly positive. Then

$$f_\mu(x) = \psi_x \chi \Psi^{-1} \Theta \chi^{-1}(\mu)$$

gives the required smooth random maps. □

For discrete Markov processes on a circle there is an easy necessary and sufficient condition on the transition maps for a representation by random diffeomorphisms.

PROPOSITION A.2. *Let $\mathcal{M}$ be the circle endowed with Lebesgue measure. Write $V_x = [l_-(x), l_+(x)]$. There is a representation by random smooth diffeomorphisms if and only if*

$$-k(x, l_-(x))l'_-(x) + \int_{l_-(x)}^{z} \frac{\partial}{\partial x} k(x, y)\, dy \neq 0$$

for $z \in V_x$.

Proof. The construction of the representation by random smooth maps proceeds as follows. For $\xi \in [0, 1]$, write $C(\xi) = [l_-(x), l_-(x) + \xi(l_+(x) - l_-(x))]$ and let

$$\omega = \int_{C(\xi)} k(x, y)\, dy.$$

Since $k > 0$, the map Θ, $\Theta(\xi) = \omega$, is a diffeomorphism. The representation by random diffeomorphisms is given through

$$f_\omega(x) = l_-(x) + \Theta^{-1}(\omega)(l_+(x) - l_-(x)).$$

Note that, for fixed ω,

$$\frac{d}{dx} \int_{[l_-(x), f_\omega(x)]} k(x, y)\, dy = 0. \tag{48}$$

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With, say, $l_- < f_\omega$, (48) yields

$$-k(x, l_-(x))l'_-(x) + k(x, f_\omega(x))f'_\omega(x) + \int_{l_-(x)}^{f_\omega(x)} \frac{\partial}{\partial x}k(x, y)\,dy = 0.$$

Hence $f'_\omega(x) = 0$ precisely if

$$-k(x, l_-(x))l'_-(x) + \int_{l_-(x)}^{f_\omega(x)} \frac{\partial}{\partial x}k(x, y)\,dy = 0. \qquad \square$$

B. *Appendix. Regularity of solutions of integral equations*

In a number of places in this paper eigenvalue equations $L_a\phi = \lambda\phi$ for the transfer operator L_a arise. As L_a depends only C^1 on a, a direct application of the implicit function theorem as found, for example, in [**9, 13**] yields only weak regularity properties of the solutions.

The following remark is a variant of Proposition 4.1. It allows an application of [**11**, Proposition 3.6.1] to show that eigenvectors and eigenvalues of L_a, in the case of simple eigenvalues, vary smoothly with a.

Remark B.1. For $a \in I$, $a \mapsto L_a$ is a C^{r+1} map from I into $\mathcal{L}(C^{k+r}(\mathcal{M}), C^k(\mathcal{M}))$, the space of bounded linear maps from $C^{k+r}(\mathcal{M})$ into $C^k(\mathcal{M})$.

We include in this appendix an alternative route to obtain such smoothness, as an introduction to the more involved reasoning in §6.

Given is $L_{a_0}\phi_{a_0} = \lambda_{a_0}\phi_{a_0}$ with L_a acting for a near a_0 on $C^k(\mathcal{M})$ (here we are considering complex valued functions). Similarly we can consider L_a acting on $C^k_0(W)$ for an isolating neighborhood W. Assume that λ_{a_0} is an isolated eigenvalue of L_{a_0}. Denote by E the span of ϕ_{a_0} and let $F^k(W)$ be a complement of E in $C^k(\mathcal{M})$. Consider functions $\phi_a = \phi_{a_0} + \psi_a$ with $\psi_a \in F^k(\mathcal{M})$. We wish to solve $L_a\phi_a = \lambda_a\phi_a$. Let P be the projection to E along $F^k(\mathcal{M})$. Considering a second parameter λ, $L_a\phi_a = \lambda\phi_a$ decomposes as

$$\begin{cases} (I - P)L_a(\phi_{a_0} + \psi_a) = \lambda\psi_a, \\ PL_a(\phi_{a_0} + \psi_a) \qquad\quad = \lambda\phi_{a_0}. \end{cases}$$

The top equation can be solved for ψ_a as a function of λ and a for a near a_0 and λ near λ_{a_0}. In fact, by the Fredholm alternative, $\psi_a = ((I - P)L_a - \lambda I)^{-1}\left((I - P)L_a\phi_{a_0}\right)$. Putting this into the bottom equation yields a single equation for λ. Note that for stationary measures, $\lambda = 1$ automatically solves this equation; compare the proof of Theorem 1.10 in §4.

Write the top equation as a fixed point equation $T_\alpha\psi_\alpha = \psi_\alpha$ with parameters α. The map T_α is a compact linear map mapping $F^k(\mathcal{M})$ into $F^{k+1}(\mathcal{M})$; compare the proof of Proposition 2.3.

LEMMA B.2. *If $(x, a) \mapsto \psi_a(x)$ is C^k, then $(x, a) \mapsto T_a\psi_a(x)$ is C^{k+1}.*

Proof. See §4. $\square$

PROPOSITION B.3. *Consider the integral equation $T_\alpha\psi_\alpha = \psi_\alpha$ with T_α as above. The fixed point $x \mapsto \psi_\alpha(x)$ is smooth jointly in x, α.*

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Proof. Given is a unique fixed point ψ_α depending continuously on α. Formally differentiating $T_\alpha \psi_\alpha = \psi_\alpha$ with respect to α gives

$$\frac{\partial}{\partial \alpha}\,(T_\alpha \psi_\alpha(x)) = \frac{\partial}{\partial \alpha} T_\alpha \psi_\alpha(x) + T_\alpha \frac{\partial}{\partial \alpha}\psi_\alpha(x) = \frac{\partial}{\partial \alpha}\psi_\alpha(x).$$

So $(\partial/\partial\alpha)\psi_\alpha$ should be the solution M_α of $(\partial/\partial\alpha)T_\alpha\psi_\alpha(x) + T_\alpha M_\alpha(x) = M_\alpha(x)$. That is,

$$M_\alpha = (I - T_\alpha)^{-1}\frac{\partial}{\partial \alpha} T_\alpha \psi_\alpha. \tag{49}$$

By the Fredholm alternative [**32**], $I - T_\alpha$ is invertible on $F^k(\mathcal{M})$. The right-hand side of (49) is therefore a continuous function. To establish that M_α is the derivative $(\partial/\partial\alpha)\psi_\alpha$, we must show that $|\psi_{\alpha+h}(x) - \psi_\alpha(x) - M_\alpha(x)h| = o(|h|)$ as $h \to 0$ (compare the proof of the implicit function theorem in e.g. [**9**] or [**13**]). Write $\gamma_\alpha(x) = \psi_{\alpha+h}(x) - \psi_\alpha(x)$. Now

$$\begin{aligned}\gamma_\alpha(x) &= T_{\alpha+h}(\psi_\alpha + \gamma_\alpha)(x) - T_\alpha(\psi_\alpha)(x)\\ &= T_\alpha\gamma_\alpha(x) + \frac{\partial}{\partial\alpha}T_\alpha\psi_\alpha(x)h + R(x),\end{aligned} \tag{50}$$

where

$$R(x) = T_{\alpha+h}(\psi_\alpha + \gamma_\alpha)(x) - T_\alpha(\psi_\alpha)(x) - T_\alpha\gamma_\alpha(x) - \frac{\partial}{\partial\alpha}T_\alpha\psi_\alpha(x)h.$$

Since $(\psi, \alpha) \mapsto T_\alpha(\psi)$ is differentiable, for any $\epsilon > 0$ there is $\delta > 0$ with $|R| < \epsilon(|\gamma_\alpha| + |h|)$ if $|\gamma_\alpha|, |h| < \delta$. Since γ_α is continuous in h, we may further restrict δ so that this estimate on $|R|$ holds for $|h| < \delta$. From (50) we get

$$\gamma_\alpha = (I - T_\alpha)^{-1}\left(\frac{\partial}{\partial\alpha}T_\alpha\psi_\alpha h + R\right)$$

so that $|\gamma_\alpha| < C|h|$ for some C, if $|h| < \delta$. This implies that $|R| < \epsilon(1 + C)|h|$ for $|h| < \delta$. As

$$(I - T_\alpha)(\gamma_\alpha - M_\alpha h) = R$$

(from (49) we get $(I - T_\alpha)M_\alpha = (\partial/\partial\alpha)T_\alpha\psi_\alpha$), we derive $|\gamma_\alpha(x) - M_\alpha(x)h| < K\epsilon|h|$ for some $K > 0$, if $|h| < \delta$. This proves that M_α equals the partial derivative $(\partial/\partial\alpha)\psi_\alpha$.

Higher-order derivatives are treated by induction. Assume that $(x, \alpha) \mapsto \psi_\alpha(x)$ has been shown to be C^j. By Lemma B.2, $(x, \alpha) \mapsto T_\alpha\psi_\alpha(x)$ is C^{j+1}. So $D\psi_\alpha = D(T_\alpha\psi_\alpha)$ is C^j. As $(I - T_\alpha)^{-1}$ maps $F^j(\mathcal{M})$ to $F^j(\mathcal{M})$, the right-hand side of (49) is a C^j function. The above reasoning shows that $M_\alpha = (\partial/\partial\alpha)\psi_\alpha$. Therefore $(\partial/\partial\alpha)\psi_\alpha$ is C^j, so that $(x, \alpha) \mapsto \psi_\alpha(x)$ is C^{j+1}. □

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