A numerical study of Gibbs $u$-measures for partially hyperbolic diffeomorphisms on $\mathbb{T}^3$.

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Abstract. We consider a hyperbolic automorphism $A : \mathbb{T}^3 \to \mathbb{T}^3$ of the 3-torus whose 2-dimensional unstable distribution splits into weak and strong unstable subbundles. We unfold $A$ into two one-parameter families of Anosov diffeomorphisms — a conservative family and a dissipative one. For diffeomorphisms in these families we numerically calculate the strong unstable manifold of the fixed point. Our calculations strongly suggest that the strong unstable manifold is dense in $\mathbb{T}^3$. Further, we calculate push-forwards of the Lebesgue measure on a local strong unstable manifold. These numeric data indicate that the sequence of push-forwards converges to the SRB measure.

1. Introduction

1.1. The setting. Consider the 3-dimensional torus $\mathbb{T}^3 = \mathbb{R}^3 / \mathbb{Z}^3$ equipped with the standard $(x, y, z)$ coordinates and a hyperbolic automorphism $A : \mathbb{T}^3 \to \mathbb{T}^3$ induced by the following integral matrix with determinant 1

$$A = \begin{pmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 1 \end{pmatrix}$$

The eigenvalues of $A$ are real and approximately equal to 0.20, 1.55 and 3.25. We denote the largest eigenvalue by $\lambda$, $\lambda \approx 3.25$, and corresponding eigenvector by $v$, $Av = \lambda v$,

$$v \approx \begin{pmatrix} 0.80 \\ 1.00 \\ 0.45 \end{pmatrix}$$

We will view $A$ as a partially hyperbolic diffeomorphism whose center distribution is expanding. Further, we unfold $A$ into two families of partially hyperbolic diffeomorphisms: a dissipative family

$$f_{D, \epsilon}(x, y, z) = (2x + y + \epsilon \sin(2\pi x), x + 2y + z, y + z) \quad (1.1)$$

and a conservative family

$$f_{C, \epsilon}(x, y, z) = (2x + y + \epsilon \sin(2\pi x), x + 2y + z + \epsilon \sin(2\pi x), y + z) \quad (1.2)$$

It is well-known that for small values of $\epsilon > 0$ the diffeomorphisms $f_{*, \epsilon}$ (here $* = D, C$), remain Anosov, and also partially hyperbolic (with weakly expanding center distribution). Hence diffeomorphisms $f_{*, \epsilon}$ leave invariant a one-dimensional strongly expanding foliation $W_{f}^{uu}$ whose expansion rate is close to $\lambda$. Note that the point $p = (0, 0, 0)$ is fixed by all diffeomorphisms in the families.

1.2. Preview of the results and conjectures. We performed a very accurate (albeit non-rigorous) numerical calculations of the the finite-length strong unstable manifolds $W_{f, *, \epsilon}^{uu}(p, R)$ which pass through $p$, up to length $R \approx 1.3 \cdot 10^8$. These numerical calculations strongly support the following conjecture.

Conjecture 1.1. For all analytic diffeomorphisms $f$ in a sufficiently small neighborhood of $A$ the strong unstable foliation $W_f^{uu}$ is transitive, i.e., it has a dense leaf.

We actually expect the foliation $W_f^{uu}$ to be minimal. However we did not calculate strong unstable leaves through non-periodic points because it is a much harder task. Figures 1.1 and 1.2 give a preview of our numerics in support of the above conjecture. These panels display first $N$ intersection points of the strong unstable manifold $W_f^{uu}(p)$ with the 2-torus given by $y = 0$. We have calculated $10^8$ intersection points. In the figures we only show up to 200,000 points because of large file size and because past $10^6$ points one only sees Malevich’s black square.

Remark 1.2. In fact we show roughly $N/2$ points because we only display the points of intersection in the “one-half torus” given by $0 \leq z \leq 1/2$. We will display all our data on $[0, 1] \times [0, 1/2] \subset [0, 1] \times [0, 1] \simeq \mathbb{T}^2$ unless specified otherwise. Note that throughout the paper we maintain
the convention to indicate the total number of points $N$ in the captions to the figures. Hence, if readers counts the points on a figure then they would get approximately $N/2$ points. Displaying half of the torus helps to reduce file size. Also note that all diffeomorphisms $f_{\epsilon_0}$ commute with the involution $i: (x, y, z) \mapsto (-x, -y, -z)$. It follows that the measures which we are interested in are invariant under $i$ and all conditional measures on the $\mathbb{T}^2$ transversal are invariant under $(x, y) \mapsto (-x, -y)$.

In general, given a partially hyperbolic diffeomorphism $f: M \to M$, one reason to be interested in minimal sets of its strong unstable foliation $W^u_f$ is that minimal invariant sets support Gibbs $u$-measures associated to $W^u_f$ of Pesin and Sinai [PS83]. Gibbs $u$-measures are of great interest in partially hyperbolic dynamics because they govern statistical properties of the dynamical system [Dol01, Dol04a]. Of course, in our setting the dynamical system is a transitive Anosov diffeomorphism which admits a unique SRB-measure and, hence, statistical properties are very well understood. However, perturbations of linear partially hyperbolic automorphisms are nice model examples where $u$-measures are not fully understood. We elaborate on our motivation to carry out the numerical study at the end of the introduction.

We view diffeomorphisms given by (1.1) and (1.2) as partially hyperbolic diffeomorphisms with one-dimensional strong unstable subbundles. Recall that a Gibbs $u$-measure of a partially hyperbolic diffeomorphism $f: M \to M$ is an $f$-invariant measure $\mu$ whose conditional measures on strong unstable plaques are absolutely continuous with respect to the induced Riemannian volume on strong unstable plaques. Gibbs $u$-measures were introduced by Pesin
and Sinai [PS83] who also suggested a way to construct them as weak* partial limits of the sequence of averages

\[
\nu_{uu}^K \equiv \frac{\nu_{uu} + f_* (\nu_{uu}) + \ldots + f_*^{K-1} (\nu_{uu})}{K}, K \geq 1, \tag{1.3}
\]

where \(\nu_{uu}\) is a singular measure (on \(M\)) given by induced Riemannian volume on a strong unstable plaque. In our setting we can take \(\nu_{uu}\) to be the singular measure (on \(M\)) given by the Lebesgue measure on a small plaque of \(W_{uu}(p)\) with one end point being \(p\). Hence we amend our calculation of \(W_{uu}(p)\) with a numeric calculation of the strong unstable Jacobians of \(f\), \(i \leq K\), to obtain the averages numerically (more precisely, we look at the conditional measures of the averages on the 2-torus given by \(y = 0\)). Even though our evidence is not entirely conclusive we believe that the averages \(\nu_{uu}^K\) converge weakly. Further we calculate the SRB measure employing the zero-noise limit description of Young [You86]. The very different numeric procedures for calculating the \(u\)-measure and the SRB measure produce visually identical results for all values of \(\varepsilon\) as indicated on Figure 1.3. Hence we cautiously conjecture the following.

**Conjecture 1.3**. For all analytic diffeomorphisms \(f\) in a sufficiently small neighborhood of \(A\) there exists a unique Gibbs \(u\)-measure (an \(f\)-invariant measure with absolutely continuous conditionals on strong unstable leaves) which then, of course, coincides with the SRB measure.

This conjecture can be reformulated as follows: for any analytic diffeomorphisms \(f\) in a sufficiently small neighborhood of \(A\) any \(f\)-invariant measure with absolutely continuous conditional measures on one-dimensional strong unstable plaques, in fact, has absolutely continuous conditional measures on two-dimensional unstable plaques.

### 1.3. Motivation.

1.3.1. Our initial interest in transitivity (or minimality) question of the strong unstable foliation came from work on smooth conjugacy of higher dimensional Anosov diffeomorphisms [Gog08]. Transitivity of invariant expanding one-dimensional foliations (albeit not the strong unstable ones) played a key role in the arguments of [Gog08]. Families of diffeomorphisms in dimension three which we consider in this paper is the simplest setting where transitivity (minimality) is not understood.

We remark that minimality of the weak unstable foliation for \(f_{*,\varepsilon}\) follows easily from structural stability. Indeed

\[1\text{Pesin and Sinai used a stronger definition which is equivalent to the one we give here, see [BDV00, Chapter 11].}\]
technique is not applicable. Hence our setting can be considered as a complementary one to the setting of [BDU02].

Recall that homotopy class of $A: \mathbb{T}^3 \to \mathbb{T}^3$ contains the Mañé’s example. This is a robustly transitive diffeomorphism $f_M: \mathbb{T}^3 \to \mathbb{T}^3$ which is partially hyperbolic but not Anosov [Mañ78]. To the best of our knowledge minimality of strong unstable foliation of $f_M$ is also an open problem. Thus, understanding strong unstable foliation of perturbations of $A$ and Mañé’s example is a prerequisite for the following problem, which is a special case of Problem 1.6 in [BDU02].

**Problem 1.4.** Consider the space of robustly transitive partially hyperbolic diffeomorphisms $f: \mathbb{T}^3 \to \mathbb{T}^3$ which are homotopic to $A$, i.e., the induced map $f_*$ on first homology group is given by $A$. Is strong unstable foliation $W_u^f$ minimal? Or at least transitive? If not, then is minimality (transitivity) of $W_u^f$ a $C^1$-open and dense property in this space?

Related to this problem, Potrie asked whether transitivity (or chain-recurrence) follows from partial hyperbolicity of $f: \mathbb{T}^3 \to \mathbb{T}^3$ in the homotopy class of $A$ [Pot14]. Further, Potrie proved that there exists a unique quasi-attractor for each such $f$. Note that the attractor must be saturated by leaves of $W_u^f$. Hence, minimality of $W_u^f$ would imply that the attractor is whole $\mathbb{T}^3$.

We also remark that robust minimality of strong unstable foliation was established in [PS83] under so called SH-condition. This condition does not hold in our setting. Finally, minimal sets of strong unstable foliation can be analyzed better in $C^1$ generic setting, see [CP15, Section 5.3].

1.3.3. It is interesting to understand the space of Gibbs $u$-measure Gibbs$^u(f)$, its dependence on the diffeomorphism and what bifurcations can occur. Note that it is known that Gibbs$^u(f)$ depends continuously on $f$ in $C^1$ topology [Yan16] (see also [BDV00, Chapter 9]). Generalizing our numeric observation of uniqueness of the $u$ measure we ask the following question.

**Problem 1.5.** Let $f: M \to M$ be a partially hyperbolic diffeomorphism of a 3-manifold $M$. Assume that $f$ admits two distinct ergodic $u$-measures $\mu_1$ and $\mu_2$. Is it true that $\text{supp}(\mu_1) \neq \text{supp}(\mu_2)$? That is, do they necessarily have distinct supports? 3

1.4. Further discussion.

1.4.1. Our numerical evidence actually suggests that the push-forward measures $f^*\nu^u$ converge to the SRB measure as $n \to \infty$. This was easier to detect than convergence of averages (1.3), which clearly converge slower. Note that convergence of $f^*\nu^u$ to the unique Gibbs $u$-measure is a key assumption in the study of statistical properties of partially hyperbolic diffeomorphisms [Dol04]. This assumption is difficult to verify theoretically when dynamics along the center subbundle is non-linear 4.

1.4.2. Another observation is that our numerical experiments suggest that the $u$-measure coming from the strong unstable leaf through $p$ agrees with the SRB measure well beyond the range of small $\varepsilon$. (The splitting at $p$ survives for all $\varepsilon > 0$.) This is indicated on Figure 1.4. At $\varepsilon = \frac{1}{2\pi} \approx 0.159$ bifurcation from diffeomorphisms to non-invertible maps occurs and the “folding” which happens beyond this parameter value is clearly visible on Figure 1.4. Note that pictures of $u$ and SRB measures do not give any indication if the bifurcation from partially hyperbolic (or Anosov) world happens. Indeed, it is actually very plausible that prior to the critical value $\frac{1}{2\pi}$ no such bifurcations happen; that is, $f_{D,\varepsilon}$ stays Anosov with weak-strong unstable splitting for $\varepsilon < \frac{1}{2\pi}$.

To provide some support we numerically calculate points of period 3 and corresponding eigenvalues using the following procedure. We consider a dense $2,000 \times 2,000 \times 2,000$ mesh of points $(x_i, y_j, z_k)$ and apply dynamics 3 times to obtain the final point $f^3(x_i, y_j, z_k)$. If for some $(i, j, k)$ the starting and final points end up within $D = D(x_i, y_j, z_k) < 0.02$ we adjust the coordinates $(x_i, y_j, z_k)$ coordinates to minimize the Euclidean distance with a gradient descent method. The partial derivatives $(\frac{\partial D}{\partial x}, \frac{\partial D}{\partial y}, \frac{\partial D}{\partial z})$ at $(x_i, y_j, z_k)$ are calculated numerically. Once $D < 10^{-5}$, we find the cubic roots of eigenvalues of $D f^3(x_i, y_j, z_k)$.

This numerics gives 16 distinct eigenvalue graphs. And this is consistent with the Lefschetz formula which yields 91 points fixed by $f^3_{*,\varepsilon}$. One of these points is the fixed

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3It was suggested to us by Dmitry Dolgopyat that this question also makes sense in higher dimensions if one additionally assumes that $f$ is accessible (or considers $u$-measures supported on an accessibility class).

4However, for transitive Anosov diffeomorphisms mixing implies that $f^3\nu^u$ converges to the SRB measure, where $\nu^u$ is Lebesgue measure on an unstable plaque. This was explained to us by F. Rodriguez Hertz.
point $p$. The rest give 30 orbits of period 3 none of which is fixed by the involution $i$. Hence the involution breaks up these orbits into 15 pairs with identical eigenvalue data. Figure 1.5 displays dependence of the eigenvalue data on $\varepsilon$. We observe clear separation of the spectrum into three bands for both conservative and dissipative ($\varepsilon < \frac{1}{2\pi}$) families.

1.4.3. Ruelle provided a formula for the derivative of the SRB measure with respect to the diffeomorphism (Anosov, or more generally on a hyperbolic attractor) [Rue97], see also [Rue98, Jia12]. If we denote by $\mu_\varepsilon$ the SRB measure of $f_{D,\varepsilon}$ and expand with respect to $\varepsilon$

$$\mu_\varepsilon = \mu_0 + \varepsilon \delta \mu + \text{h.o.t.}$$

Then, according to Ruelle’s formula,

$$\delta \mu(\Phi) = \sum_{n \geq 0} \int (\nabla(\Phi \circ \Lambda^n, X)) d\mu$$

where $\Phi \in C^\infty(M, \mathbb{R})$ and $X$ is the vector field $\partial f_{D,\varepsilon}/\partial \varepsilon|_{\varepsilon=0}$. Similar formula for $u$-measure of a partially hyperbolic diffeomorphism (more specifically, an element of an Anosov action) was established by Dolgopyat [Dol04b]. Remarkably, the formula of Dolgopyat holds for families $f_\varepsilon$ even when the uniqueness of $u$-measure is not known for $\varepsilon > 0$. That is, all families of $u$-measures which start at $\mu_0$ have the same derivative. Thus, even though results of Dolgopyat are not directly applicable in our setting, we are less confident about Conjecture 1.3. One should exercise caution when perusing our numeric evidence for Conjecture 1.3 as it might be missing some higher order phenomena.

2. Background

In this section we briefly summarize the needed background. For in depth discussions of partial hyperbolicity, SRB measures and Gibbs $u$-measures we refer the reader to [Pes04, BDV00, You02, PS83, Dol01].

2.1. Anosov and partially hyperbolic diffeomorphisms. Recall that a self-diffeomorphism $f: M \to M$ of a compact Riemannian manifold is called Anosov if the tangent space $T_xM$ at every $x \in M$ is split into $Df$-invariant subbundles, $T_xM = E^s(x) \oplus E^u(x)$, and $Df|_{E^s}$ is uniformly expanding while $Df|_{E^u}$ is uniformly contracting.
An important generalization is the concept of partially hyperbolic diffeomorphism $f : M \to M$ which assumes existence of a $Df$-invariant splitting $T_x M = E^{ss}(x) \oplus E^c(x) \oplus E^{uu}(x)$, where $Df|_{E^{ss}}$ is uniformly expanding, $Df|_{E^{uu}}$ is uniformly contracting and $Df|_{E^c}$ has intermediate growth, that is,

$$
\|Df|_{E^c}\| \cdot \|(Df|_{E^{ss}})^{-1}\| < 1,
$$

and

$$
\|Df|_{E^{ss}}\| \cdot \|(Df|_{E^c})^{-1}\| < 1.
$$

It is well-known that $E^{ss}$ and $E^{uu}$ integrate to foliations which we denote by $W^{ss}$ and $W^{uu}$, respectively.

For sufficiently small $\varepsilon > 0$ the diffeomorphisms $f_{\varepsilon, \varepsilon}$ given by (1.1) and (1.2) are Anosov with 2-dimensional unstable distributions. However, because $A$ has real spectrum $\lambda_1 < 1 < \lambda_2 < \lambda_3$, the unstable distribution admits a finer invariant splitting $E^c \oplus E^{uu}$ and, hence, $f_{\varepsilon, \varepsilon}$ can also be viewed as a partially hyperbolic diffeomorphism.

### 2.2. SRB measures

Informally speaking, SRB measures are invariant measures which are most compatible with volume when volume itself is not invariant. More precisely, consider a self-diffeomorphism $f : M \to M$, then an invariant measure $\mu$ is called an SRB measure (or a physical measure) if its basin of attraction has positive volume; that is, the set of points $x \in M$ such that

$$
\forall \varphi \in C^0(M) \lim_{n \to \infty} \frac{1}{n} \sum_{i=0}^{n-1} \varphi(f^i x) = \int_M \varphi d\mu
$$

has positive volume. For transitive Anosov diffeomorphisms the SRB measure $\mu$ is unique and is well-understood by work of Sinai, Ruelle and Bowen. It can be characterized by the following equivalent conditions.

(C1) $\mu$ has absolutely continuous conditionals on unstable plaques;  
(C2) $\mu$ is the zero-noise limit of small random perturbations of $f$.

Our numeric calculations of the SRB measure for $f_{\varepsilon, \varepsilon}$ will rely on the second characterization which is due to L.-S. Young [You86]. We will elaborate on it later in Section 3.2.

### 2.3. Gibbs u-measures

The definition of Gibbs u-measures for partially hyperbolic diffeomorphisms comes from postulating characterization (C1) above. Given a partially hyperbolic diffeomorphism $f : M \to M$, an invariant measure $\mu$ is called a Gibbs u-measure if it has absolutely continuous conditionals on unstable plaques. Then the density of the conditional measure on a plaque $W^u_f(x, R)$ is given by

$$
\rho_x^{uu}(y) = \prod_{i \geq 0} \frac{\text{Jac}(f^{-i}|_{E^{uu}(f^{-i}(y))})}{\text{Jac}(f^{-i}|_{E^{uu}(f^{-i}(x))}), \ y \in W^u_f(x, R)} \quad (2.4)
$$

Note that in our setting the SRB measure is automatically a u-measure. In general, of course, the converse does not hold. Still, under additional assumptions this could be the case. For example, Bonatti and Viana showed that if $E^c$ is mostly contracting then there are finitely many ergodic Gibbs u-measures which are the SRB measures [BV00].

Dolgopyat, assuming uniqueness of the u-measure and that push-forwards $f^n_{\nu} u$, $n \geq 0$, converge to the u-measure, established various limit theorems previously known in the Anosov setting [Dol04a].

### 2.4. Numerics

We have chosen the C language to implement the numerical algorithms for computing orbits, u-measures, and SRB-measures in this study. Due to the high precision requirements in the calculation of u- and SRB-measures, we employed quadmath library available in gcc 4.4.6. The quadruple precision _float128 type provides machine epsilon $2^{-112} \approx 10^{-34}$. We relied on the Box-Muller transformation [BM58] of random numbers produced with a linear congruential generator [PM88] to introduce Gaussian noise in the SRB calculations. Generation of u-measures proved to be the most expensive part of this numerical study but the computational cost was fairly low, at about 1,000 CPU hours per $10^8$ points.

### 3. The results

#### 3.1. Numerics for the strong unstable manifold

Let $f : \mathbb{T}^3 \to \mathbb{T}^3$ be a partially hyperbolic diffeomorphism which belongs to the family (1.1) or the family (1.2) for small $\varepsilon > 0$. Here we explain numerics for the strong unstable manifold $W^u_f(p)$ and present the numerical evidence supporting Conjecture 1.1 using Figures 3.2 and 3.3.

Consider the universal cover $\mathbb{R}^3 \simeq \{(x, y, z)\}$. Denote by $\tilde{f} : \mathbb{R}^3 \to \mathbb{R}^3$ the lift of $f$ that fixes point $(0, 0, 0)$, which we still denote by $p$. Also denote by $\tilde{W}_f^{uu}(p)$ the connected component of the lift of $W^u_f(p)$ which contains $p$. Then the strong unstable manifold $\tilde{W}_f^{uu}(p)$ can be viewed as a graph of a function $\varphi^{uu}$ defined on the $y$-axis

$$
\varphi^{uu} : \mathbb{R} \to \mathbb{R}^2, \ y \mapsto (x_y, z_y) \overset{\text{def}}{=} \varphi^{uu}(y)
$$

as shown on Figure 3.1. For each integer $y_0$ the point $(x_{y_0}, z_{y_0})$ is the intersection point of the plane $\mathbb{R}^2 \simeq \{ y = \ldots}$

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5Sinai-Ruelle-Bowen.
shown on Figures 3.2 and 3.3 where as we “zoom in” at $T^{-1}$ any gaps in larger values of ages we obtain clearly indicate that points cluster more for $8$. We calculate up to $10$ to achieve convergence of $W_{p}^{u}$. Then we adjust the initial point to $y$ the end point of scaled eigenvector $p$. We calculate the point whose $y$-coordinate equals to $50$, its intersection point of the $2$-torus $T^{2}$, and let $v_{p}$ be a vector proportional to $v_{p}$ whose $y$-coordinate equals to $y_{0}$. Also let $\lambda_{p}$ be the corresponding eigenvalue, $D_{p}\tilde{f}v_{p} = \lambda_{p}v_{p}$. Then to calculate the point $q = (x_{0}, y_{0}, z_{0})$ we employ the following iterative algorithm. First we let $q_{0} = 0$ be the end point of scaled eigenvector $y_{0}$. We calculate the first approximation by using the dynamics $q^{z} = \tilde{f}^{50}(q_{0})$.

Then we look at the $y$-coordinate of $q^{x}$, compare it to $y_{0}$, use a linear mixing scheme with parameter $0.1$ to adjust the initial point $q_{0, 50}$ along $v_{p}$ to a new initial point $q_{2, 50}$, and repeat the procedure $n = 1, 000 - 2, 000$ times to achieve convergence of $q^{n}$ to the desired $q \in W_{f}^{u}(p)$ within $10^{-24}y_{0}$ from the plane $\{y = y_{0}\}$.

By repeating this iterative calculation we obtain the sequence of intersection points of $W_{f}^{u}(p)$ and $T^{2} \simeq \{y = 0\}$

\[
\{(x_{0}, y_{0}, z_{0}); y_{0} \geq 1\}
\]

We calculate up to $10^{8}$ points in this sequence. The images we obtain clearly indicate that points cluster more for larger values of $\epsilon$. Still the sequence does not seem to leave any gaps in $T^{2}$. This supports our density conjecture as shown on Figures 3.2 and 3.3 where as we “zoom in” at point $p$. Zooming in does not reveal any regions free of intersection points. For the conservative family, points tend to cluster much less and distribute more evenly. We include the figures for the dissipative family only since they are more interesting.

3.1.1. Reliability of numerics. The numerical data for the sequence of points $\{(x_{0}, y_{0}, z_{0}); y_{0} \geq 1\}$ is the key data providing support to our conjectures. Thus we briefly elaborate on the reliability of our calculations.

The iterative calculation of points $q = (x_{0}, y_{0}, z_{0})$ has two sources of numerical errors. The first one is algorithmic and is associated with the deviation of $q_{-K}$ ($K = 50$) from the strong unstable manifold $W_{f}^{u}(p)$ and can be estimated as follows. Recall that $q$ is obtained as $\tilde{f}^{K}(q_{-K})$, where $q_{-K}$ is first guessed as $\lambda_{p}^{-K}v_{p}$. Figure 3.4 shows

Figure 3.1. The lift of the strong unstable manifold and the sequence of points $\{(x_{0}, z_{0}); y_{0} \geq 1\}$. $y_{0}$ and $\tilde{W}_{f}^{u}(p)$ and the point

$$(x_{0}, z_{0}) \mod \mathbb{Z}^{2}$$

is an intersection point of the $2$-torus $T^{2} \simeq \{y = 0\}$ and $W_{f}^{u}(p)$.

To calculate $(x_{0}, z_{0})$ we carry out the following procedure. Denote by $v_{p}$ the vector tangent to $\tilde{W}_{f}^{u}(p)$ at $p$ (which is an eigenvector of $D_{p}\tilde{f}$). And let $cv_{p}$ be a vector proportional to $v_{p}$ whose $y$-coordinate equals to $y_{0}$. Also let $\lambda_{p}$ be the corresponding eigenvalue, $D_{p}\tilde{f}v_{p} = \lambda_{p}v_{p}$. Then to calculate the point $q = (x_{0}, y_{0}, z_{0}) \in W_{f}^{u}(p)$ we employ the following iterative algorithm. First we let $q_{1, 50}$ be the first approximation by using the dynamics

$q^{1} = \tilde{f}^{50}(q_{0}^{1, 50})$.

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Figure 3.2. Support for Conjecture 1.1. The intersection points are plotted for $f_{D}$ with $\varepsilon = 0.1$. Two lower panels are $\times 3$ and $\times 9$ zoom-ins. The number of points is increased proportionally to the area of the domain.
that, once converged, \( q - K = c v_p \) remains within a factor of 10 from the original guess for all \( 1 \leq y_0 \leq 10^8 \). The value of \( c \) does not exceed \( 10^{-17} \) for the considered \( \lambda_p \geq 3.25 \) corresponding to \( \varepsilon = 0.1 \). Since the strong unstable manifold and the line spanned by \( v_p \) have quadratic tangency at small values of \( c \), the transverse distance from \( q - K \) to \( \hat{W}^u_f(p) \) is of the order \( c^2 \). Over the \( K \) steps, the error in the determination of the intersection of the strong unstable manifold and \( T^2 \) transversal is amplified by a factor of \( \lambda_{\perp}^K \). According to Figure 1.5, \( \lambda_{\perp} \) along the weak unstable direction is below 2 for \( \varepsilon \leq 0.1 \). Hence, the resulting error does not exceed \( 10^{-34} \times 2^{50} < 10^{-18} \).

The second source of error is the accumulation of rounding errors due to machine precision. The resulting accuracy depends on the particular set of performed operations and is not easy to evaluate. In order to examine the sensitivity of our results to machine precision, we compare the errors in determining \( q \) for the full range of \( y_0 \) using single, double, and quadruple precisions. The machine epsilons corresponding to the float, double, and float128 data types are \( 2^{-23} \approx 10^{-7} \), \( 2^{-52} \approx 10^{-16} \), and \( 2^{-112} \approx 10^{-34} \), respectively. Figure 3.5(a) shows how far \( q \) points calculated with single or double precisions deviate from those calculated with the quadruple precision. One can see that the use of double instead of float reduces the error for all considered \( y_0 \) by about \( 10^{-9} \) which is consistent with the difference in the machine epsilons for the two data types. Based on this observation, we expect that the use of float128 instead of double reduces the error by an additional factor of about \( 10^{-18} \) and the rounding error should not exceed \( 10^{-25} \) for \( y_0 \) up to \( 10^8 \).

The relative magnitude of the two errors can be controlled by the choice of \( K \): the algorithmic error decreases with \( K \) roughly as \((\lambda_{\perp}^{\text{min}})^{-2K} \times (\lambda_{\perp}^{\text{max}})^K\) and the rounding error grows with \( K \). Our final test results plotted in Figure 3.5(b) illustrate that the chosen value of \( K = 50 \) ensures sufficient numerical accuracy to achieve iterative convergence of the \( q \) point positions in the intersecting plane within \( 10^{-24} y_0 \). Indeed, the distance between \( q \) calculated for \( K = 50 \) and those calculated at smaller \( K \) values drops below \( 10^{-22} \) by \( K = 35 \) for all considered \( y_0 \) values.
We conclude that both sources of numeric error are negligible ($10^{-18}$ and $10^{-25}$) than compared to our stopping criterion $10^{-24} y_0 \leq 10^{-16}$. Note that even though $10^{-24} y_0$ is deviation from the intersecting plane it also gives an error of the same magnitude within the plane because the strong unstable manifold intersects the plane with an angle, hence, the error projects to the plane. Therefore, the settings used in this computational study allow us to resolve the point positions better than $10^{-16}$ for $y_0$ up to $10^8$ and the procedure can be used (by further lowering the iteration stopping parameter below $10^{-24} y_0$) to generate points with much larger $y_0$ values.

### 3.2. Numerics for SRB measures

We recall the description of SRB measure as zero-noise limit by L.-S. Young [You86]. The idea is to approximate a diffeomorphism $f : M \to M$ by random Markov chains. To define the Markov chain consider Borel probability measures $p^\sigma(\cdot|x)$ for all $x \in M$. Given a Borel set $A \subset M$ one can think about $p(A|x)$ as the probability of sending $x$ to the set $A$. A measure $\mu$ on $M$ is stationary if

$$\mu(A) = \int_M p(A|x) d\mu(x)$$

for every Borel set $A$.

A small random perturbation of $f : M \to M$ is a one parameter family of Markov chains given by transition probabilities $p^\sigma(\cdot|x),\; x \in M$, which satisfy $p^\sigma(\cdot|x) \to \delta_{f(x)}$ as $\sigma \to 0$ uniformly in $x \in M$. (We will think of $\sigma$ as a discrete parameter.) The following properties were established in [You86].

1. If $x \mapsto p(\cdot|x)$ is continuous then a stationary measure exists;
2. If $p(\cdot|x)$ are absolutely continuous with respect to volume for all $x \in M$ then the stationary measure is also absolutely continuous;
3. If $\{p^\sigma(\cdot|x)\}$ is a small random perturbation of a diffeomorphism $f$, then all limit points of a sequence of stationary measures $\{\mu^\sigma\}$, as $\sigma \to 0$, are $f$-invariant.

Given a measure $\nu$ on a set of diffeomorphisms $\Omega \subset \text{Diff}(M)$ one can define the transition probabilities by

$$p(A|x) = \nu\{g : g(x) \in A\}. \quad (3.5)$$

Now, if $\nu^\sigma \to \delta f$ as $\sigma \to 0$ then corresponding Markov chains $\{p^\sigma(\cdot|x)\}$ yield a small random perturbation of $f$. Theoretical support for our computations of SRB measures which we are about to describe comes from the following theorem (which is a particular case of a more general result in [You86]).

**THEOREM 3.1** ([You86]). Let $f : M \to M$ be a transitive Anosov diffeomorphism. There exists a $C^1$ small and $C^2$-bounded neighborhood $\Omega \ni f$ such that if $\{\nu^\sigma\}$ are Borel probability measures on $\Omega$ with $\nu^\sigma \to \delta_f$, $\sigma \to 0$, and corresponding transition probabilities $\{p^\sigma(\cdot|x)\}$ given by (3.5) are absolutely continuous then (every) sequence of stationary measures $\mu^\sigma$ converges to the SRB measure.

Hence this theorem gives a lot of credibility to numerical calculations where one applies dynamics and small random noise at each step to obtain an approximation for the SRB measure. More precisely we consider a sequence of symmetric Gaussians $\xi^\sigma$ on $\mathbb{R}^3$ with zero mean and standard deviation $\sigma$. Let $\xi \to \delta_{(0,0,0)}$ as $\sigma \to 0$. We define

$$\nu^\sigma = f + (\xi^\sigma \mod \mathbb{Z}^3)$$

that is, we post-compose our dynamics with a small random translation on $\mathbb{T}^3$. Then, clearly, $\nu^\sigma \to \delta_f$ as $\sigma \to 0$ and one...
easily sees that the transition probabilities are absolutely continuous. Hence the theorem above applies. (Technically, we also need to truncate Gaussians to ensure $C^1$-smallness of the perturbation, but practically this makes no difference as we are interested in very small $\sigma$.)

The numeric scheme is as follows. We begin with a random point $q_0$ on $\mathbb{T}^3$ and generate a $\sigma$-approximation of the SRB measure by consecutive application of $f$ and addition of Gaussian noise $\xi^\sigma$. That is,

$$q_{i+1} = f(q_i) + \xi^\sigma$$

Note that we only work in the parameter range $\sigma \gg 10^{-32}$ so that the numeric error in calculation of $f$ is much smaller than the (small) random noise. Therefore, exponential accumulation of the numeric error is not of any concern. On Figure 3.6 we display several approximations for different values of $\sigma$. For all further SRB measures numerics, which we need for comparisons with $u$-measures numerics, we use $\sigma = 10^{-29}$.

3.3. Comparing Gibbs $u$ and SRB measures. Consider the averaged Dirac measures

$$\Sigma^u = \frac{1}{N} \sum_{y_0=1}^N \delta_{(x_{y_0}, z_{y_0})} \quad (3.6)$$

对应的 $u$-measure, and

$$\Sigma^{SRB} = \frac{1}{N} \sum_{q_i \in S} \delta_{q_i} \quad (3.7)$$

corresponding to the SRB measure, where $S$ is the slice \(\{(x, y, z) \in \mathbb{T}^3 : -0.005 \leq y \leq 0.005\}\) which contains $N$ points. For the dissipative family point distributions $\Sigma^u$ and $\Sigma^{SRB}$ visually coincide for all parameters in our range $\varepsilon \in [0, 0.25]$ (see Figure 1.3). On the other hand, for the conservative family, the SRB measure is the uniform Lebesgue measure as it supposed to be, while the $u$-measure appears to be an absolutely continuous measure with a non-constant density, as one can see on the top panel of Figure 3.7. The “non-uniformity” increases as we increase $\varepsilon$. The explanation for this discrepancy is that (3.7) gives the (approximation of) true conditional measure on $\mathbb{T}^2$ of the SRB measure, while (3.6) does not give (an approximation of) the conditional of $f_n \nu^u$. Hence we proceed with the numeric calculation of the true conditional of $f_n \nu^u$ on $\mathbb{T}^2$ and present the numeric evidence that the measures indeed coincide.

Remark 3.2. Note however that the conditional of $f_n \nu^u$ on $\mathbb{T}^2$ is absolutely continuous with respect to (3.6). Hence, if $\Sigma^u$ converges to an absolutely continuous measure (which we numerically verified by using histograms) then $f_n \nu^u$ converges to an absolutely continuous measure on $\mathbb{T}^3$ as $n \to \infty$. And, since this measure is invariant, it must be the volume. In view of this remark our further numeric verification of convergence of $f_n \nu^u$ to volume becomes somewhat redundant. However we still find it important to have direct numeric evidence.

Remark 3.3. By analyzing distribution functions of the Dirac averages (3.6) and (3.7) in the dissipative family we can also very clearly conclude that $\Sigma^u$ and $\Sigma^{SRB}$ do not converge to the same measure on $\mathbb{T}^2$. Hence, as to be expected, the above discussion also applies to the dissipative family. However visually the point distributions $\Sigma^u$ and $\Sigma^{SRB}$ are identical in this case. This happens because for...
singular measures, when looking at the pictures of approximating point distributions we can only see the measure class rather than the measure itself.

3.4. The conditional measure on $T^2 \simeq \{y = 0\}$ for Gibbs $u$-measure. Now we explain precisely our numerics for the conditional of the Gibbs $u$-measure. Let $r \in \hat{W}_f^{\nu_\alpha}(p)$ be a point very close to $p$. Consider the Lebesgue measure on $\hat{W}_f^{\nu_\alpha}(p)$ induced by the canonical flat Riemannian metric on $\mathbb{R}^3$. And denote by $\nu^{\alpha\nu}$ the normalized Lebesgue measure supported on the stable manifold $W^{\nu_\alpha}(p)$ at the intersection point $(x_{yo}, y_{yo}, z_{yo})$. Then, by using calculus, the density of $f_n^{\alpha\nu}$ with respect to the Lebesgue measure on $W_f^{\nu_\alpha}(p)$ is given by

$$\rho(q) = J ac(f^{-n}|E^{\nu_\alpha}(q)), \quad q \in [p, f^{n}(r)]^{\nu_\nu}$$

This Jacobian density can be easily evaluated numerically because, as we explained in Section 3.1, we can accurately calculate points $q$ on $\hat{W}_f^{\nu_\alpha}(p)$ together with their preimages under $f^{-n}$, $n \leq 50$. Hence to find $\rho(q)$ approximately we look at points $q - \Delta q, q + \Delta q$ on $\hat{W}_f^{\nu_\alpha}(p)$ and their preimages; and then evaluate the Jacobian numerically by taking the ratio.$^6$

Recall that we need to further take the conditional measure of $f_n^{\alpha\nu}$ on $T^2 \simeq \{y = 0\}$. Then, one can easily see (for example, by taking the limit as the width of the slice goes to zero) that the expression for the conditional measure at the intersection point $(x_{yo}, y_{yo}, z_{yo})$, $y_{yo} \in \mathbb{Z}^+$, depends on the angle between $W_f^{\nu_\alpha}(p)$ and $T^2$ at $(x_{yo}, y_{yo}, z_{yo})$. Namely, one has the following formula for the conditional of $f_n^{\alpha\nu}$ on $T^2 \simeq \{y = 0\}$

$$\Sigma_{\rho a}(p) = \frac{1}{W(N)} \sum_{y_{yo}=1}^{N} \rho(x_{yo}, y_{yo}, z_{yo}) a(x_{yo}, y_{yo}, z_{yo}) \delta(x_{yo}, z_{yo}),$$

where $\rho$ was defined above,

$$W(N) = \sum_{y_{yo}=1}^{N} \rho(x_{yo}, z_{yo}) a(x_{yo}, y_{0}, z_{yo});$$

and the “angle weight” is defined by

$$a(x_{yo}, z_{yo}) = \frac{1}{\langle v^\perp, \nu^{\alpha\nu} \rangle},$$

where $v^\perp$ is unit vector at $(x_{yo}, y_{yo}, z_{yo})$ perpendicular to $T^2$ and $\nu^{\alpha\nu}$ is the unit vector at $(x_{yo}, y_{0}, z_{yo})$ tangent to $\hat{W}_f^{\nu_\alpha}(p)$. Again, coefficient $a$ is easy to calculate since we can numerically calculate the tangent vectors $v^{\alpha\nu}$.

The density weights $\rho$ and $a$ have different genesis. Hence, for the purpose of analyzing $\Sigma_{\rho a}$ we also introduce the “component” Dirac averages

$$\Sigma_{\rho} = \frac{1}{W} \sum_{y_{yo}=1}^{N} \rho(x_{yo}, y_{yo}, z_{yo}) \delta(x_{yo}, z_{yo}),$$

and

$$\Sigma_a = \frac{1}{W} \sum_{y_{yo}=1}^{N} a(x_{yo}, z_{yo}) \delta(x_{yo}, z_{yo}),$$

which are normalized by the corresponding total weight $W$.

3.5. Comparing Gibbs $u$ and SRB measures numerically. Our calculations of $\rho$, $a$ and point distributions $\Sigma_{\rho}$, $\Sigma_{\nu}$, $\Sigma_a$ and $\Sigma_{\rho a}$ are summarized on Figures 3.7, 3.8 and 3.9.

On Figure 3.7 the values of weights are coded in color. The average value is normalized to equal 1. By the definition the “angle weight” $a = a(x_{yo}, z_{yo})$ is a continuous function on $T^2$. Notice that $a$ (middle panel) varies only slightly, within 2% of the average. On the other hand, on the bottom panel shows that $\rho$ varies a lot. Further, the graph of $\rho = \rho(x_{yo}, y_{0}, z_{yo}) = \rho(y_{yo})$ given on Figure 3.8 shows that $\rho$ is unbounded (that is, if normalize $\rho$ so that $\rho(0) = 1$ then $\rho$ is unbounded function of $y_{0}$) and have certain “self-similar” structure.

By examining Figure 3.7 one can see that $\rho$ smoothes out the point distribution $\Sigma_{\nu}$, that is, it makes $\Sigma_{\rho}$ more uniform than $\Sigma_{\nu}$. Curiously, and we have no good explanation for this, the “angle weight” $a$ makes point distribution less uniform, but, as we remarked before, $a$ has a very small effect on the distribution.

In order to quantify these observations coming from Figure 3.7 we use a $200 \times 200$ square grid to partition $T^2$ into 40,000 bins. For each bin $B$ we calculate its total weight

$$w_{\nu}(B) = \# \left\{ y_{yo} \leq N : (x_{yo}, z_{yo}) \in B \right\}$$

as well as the weight adjusted by $\rho$

$$w_{\rho}(B) = \sum_{(x_{yo}, z_{yo}) \in B} \rho(y_{yo}).$$

And weights $w_{SRB}(B)$, $w_{\nu}(B)$ and $w_{\rho}(B)$ are defined analogously. Further we calculate relative standard deviation

$^6$We use $\|\Delta\| = 10^{-7}$. With such step size, tests similar to ones in subsection 3.1.1 give an upper bound of $10^{-6}$ on the precision for weight values.
in order to have a single number which measures closeness to the uniform distribution

\[ RSD^u = \frac{1}{\tilde{w}^u} \left( \frac{1}{40,000} \sum_B (w^u(B) - \tilde{w}^u)^2 \right)^{\frac{1}{2}}, \]

where \( \tilde{w}^u \) is the average of the weights. Analogously we have relative standard deviations \( RSD_{SRB}^u, RSD_{\rho}^u, RSD_{\rho a}^u \) and \( RSD_{\rho a}^a \). The dependence of relative standard deviations on the number of points in the range \( N = 10^6, \ldots, 10^8 \) is shown on Figure 3.9. Indeed, we see that \( RSD_{SRB}^u, RSD_{\rho}^u \) and \( RSD_{\rho a}^u \) decay to zero roughly proportionally to \( \frac{1}{\sqrt{N}} \). Unfortunately we cannot differentiate between \( RSD_{\rho}^u \) and \( RSD_{\rho a}^u \).

If we denote by \( F_u^*: [0, 1]^2 \to [0, 1] \) the distribution function of \( \Sigma_u^* = \rho, \rho a \), given by

\[ F_u^*(c, d) = \Sigma_u^*(\{0, c\} \times \{0, d\}) \]

then weak* convergence of \( \Sigma_u^* \) to the Lebesgue measure is equivalent to convergence of \( F_u^*(c, d) \) to \( cd \) for all \((c, d) \in [0, 1]^2\) as \( N \to \infty \). We remark that convergence of \( RSD_{\rho a}^u \) to 0 is equivalent to

\[ F_u^*(c, d) \to cd, \ (c, d) = \left( \frac{i}{200}, \frac{j}{200} \right), \ i, j = 0, \ldots, 200. \]

Such convergence of distribution functions is known in statistics as Kolmogorov-Smirnov test [Smi39].

Finally let us mention that we have also performed similar numerics, such as the Kolmogorov-Smirnov test, for the dissipative family and the results are similar. It is more difficult to compare \( \Sigma_{SRB}^u \) and \( \Sigma_{\rho a}^u \) in this case because the SRB measure is not Lebesgue. The difficulty comes from the fact that \( \Sigma_{SRB}^u \) is defined by (3.7) using the slice \( S \) of thickness \( \Delta y = 10^{-2} \). In conservative case the value of \( \Delta y \) is irrelevant, but in the dissipative case the restriction of the SRB measure to the slice is no longer a product measure. Hence we also need to let \( \Delta y \to 0 \) in order to approximate the conditional measure on \( T^2 \). The additional parameter \( \Delta y \) makes numerics even more involved and we did not fully pursue it.
Acknowledgments. A.G. would like to thank Yakov Pesin who introduced him to questions in the spirit of our Conjecture 1.3 in his 2004 dynamics course. Also A.G. would like to thank Aleksey Gogolev who performed initial numerical experiments and created the first set of beautiful pictures back in 2008. During final stages of preparation of this paper discussions with Federico Rodriguez Hertz were very useful. We would like to acknowledge helpful feedback from Dmitry Dolgopyat, Yi Shi and Rafael Potrie. We also acknowledge helpful comments provided by the referee.

A.G. was partially supported by the NSF grant DMS-1204943. I.M. and A.N.K. gratefully acknowledge NSF support (Award No. DMR-1410514).

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