

# Complexity of Magnetic Biliards in a Torus

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# 1 Introduction

Mathematical billiards is a broad topic in dynamical systems which studies the long term motion of a particle in the presence of an obstruction. The obstruction can be a boundary upon which the particle collides and then is elastically reflected, or there can be a vector field that deflects the particle, for example, a magnetic field. In this paper we consider the latter. We take the work [KS17] and [Gas21] as inspiration.

Let  $H : \mathbb{R}^3 \times \mathbb{R}^3 \rightarrow \mathbb{R}$  be a magnetic Hamiltonian function:

$$H(q, p) = \frac{1}{2} \|p - A(q)\|^2 \quad (1)$$

where  $q = (q_1, q_2, q_3)$  is position,  $p = (p_1, p_2, p_3)$  is momentum,  $A : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  is a magnetic vector field defined as follows:

$$A(q) = (-b(q_2 \bmod 1), 0, 0) \mathbb{1}_S(q \bmod 1) \quad (2)$$

$$S = \{x \in \mathbb{R}^2 : \|x - 1/2\| \leq R\}, \quad (3)$$

with magnetic field strength  $b \in (0, \infty)$ , and  $R \in (0, 1/2)$ , the radius of the disk  $S$  centered at  $(1/2, 1/2)$  in the plane. Lastly,  $\mathbb{1}_S$  is an indicator function on  $S$ , that is  $\mathbb{1}_S(x) = 1$  if  $x \in S$  and  $\mathbb{1}_S(x) = 0$  otherwise.

We note that  $A$  is independent of  $q_3$ , so for  $q_3 = p_3 = 0$ , a solution of the Hamiltonian equations of  $H$  is contained within the  $q_1 q_2$ -plane. So, we ignore the  $q_3$  variable completely. Furthermore,  $A$  is 1-periodic in both  $q_1$  and  $q_2$ , so we can consider  $H$  on the torus  $\mathbb{T} = \mathbb{R}^2 / \mathbb{Z}^2$  via the quotient. We will refer to  $H$  on either surfaces interchangeably. Likewise, the periodicity of  $A$  gives rise to a lattice of disks centered at the points  $N + 1/2$  for  $N \in \mathbb{Z}^2$ , we use the variable  $S$  to refer to either the disc in  $[0, 1]^2$  or to the lattice, and make it clear which we are referring to when necessary.

The motion of a particle under  $H$  in the interior and the exterior of  $S$  is well understood. In the exterior of  $S$  there is no magnetic field, so the particle experiences free motion and travels in a straight line. In the interior of  $S$ , we know that the particle travels along arcs of a *Larmor* circle with *Larmor* radius  $\|p\|/b$ . Since eq. (1) is discontinuous along the boundary  $\partial S$ , we have an issue when considering particles that enter  $\partial S$  tangentially, in which case the solution is not necessarily unique. In such cases, we assume the particle is in free motion.

We see an example trajectory in fig. 1 which demonstrates the types of motion exhibited by the system. In blue we indicate free motion, and in red we have motion in the magnetic field. The red segments should instead be circular arcs, we reserve some artistic liberty in this choice. We also only draw the boundary of disks that are hit by the particle, and skip the rest to improve legibility. Focusing now on the trajectory itself we see:

- erratic behavior, that is, the trajectory seems to bounce around in a chaotic manner;
- evidence for quasi-periodic motion, specifically referring to the spot where the particle is trapped between four discs before eventually escaping;
- two distinct scales, the intervals of magnetic motion serve as a perturbation or deflection and are rather local, while the intervals of free motion

can be long and in fact can be arbitrarily long provided that the particle exits a disc at a shallow enough angle.

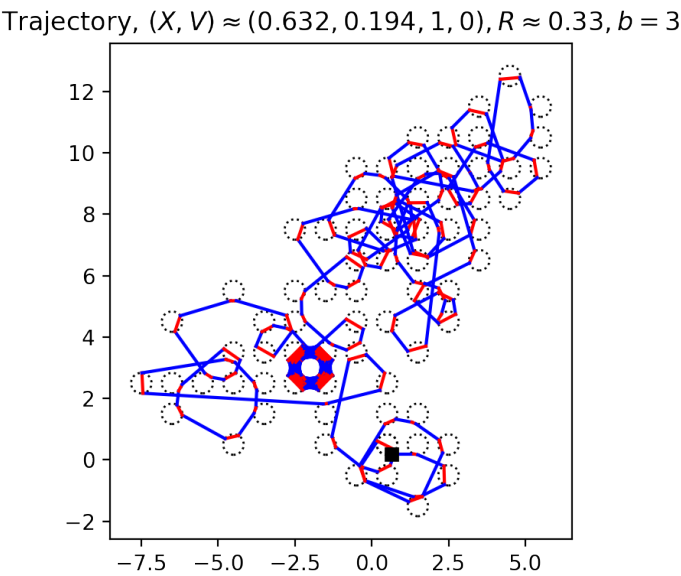


Figure 1: An example trajectory illustrating various types of motion.

So far, we see that the motion is not trivial, and warrants study. What is not yet evident from fig. 1 is the influence of the parameters  $R$  and  $b$  on the general behavior of the system. This is what we aim to better understand by the end of the paper.

We outline how we study this system. We focus on varying  $b$ , and identify three modes for some  $b_1, b_2 \in \mathbb{R}$  with  $b_1 < b_2$ . **Continue this later.**

	$b < b_1$	$b_1 < b < b_2$	$b_2 < b$
Uniform perturbation in a disk	Usual	test cases (difficult in general)	Perturbation of Sinai Billiards (maybe)
Arbitrary perturbations		use paper by Donnay-Liverani	Hard?

Table 1: What it is that we want to do

## 2 Weak magnetic fields and KAM theory

We begin by recalling Kolmogorov-Arnold-Moser (KAM) theory, state one of the main KAM theorems, and briefly outline the main points of the theory before delving into its application. We refer the reader to [Kna18] and [Ser22] for a more detailed account. We strongly recommend [Pö82] for reference, as it is the version of KAM we use here. What we will see is that for “small” values of  $b$ , our system (1) can be viewed as a perturbation of at least two different Hamiltonian systems, which can explain certain behavior we find. We corroborate our reasoning with many numerical simulations.

### 2.1 Summary of KAM theory

KAM theory is a method for studying perturbations of integrable Hamiltonian systems. Its origins lie in Celestial and Hamiltonian mechanics, where it was used to study the orbits of planets. Hamiltonian mechanics is a strong tool for modeling and studying systems, however it is strongest for conservative systems. Naturally, we find in practice many non-conservative systems, or conservative ones that are too complicated in full, in which case a smaller subsystem is modeled and the rest is viewed as a perturbation. We are interested in the second scenario, we denote by  $H^0(q, p)$  an integrable Hamiltonian and by  $H^1(q, p, \varepsilon)$  a perturbation.

Focusing on the integrable case, it is known by the Liouville-Arnold theorem that there exist *action-angle* coordinates so that  $H^0 := H^0(p)$  can be expressed in terms of the action variable only. The equations of motion in action-angle coordinates are given by:

$$\dot{q} = \omega, \quad \dot{p} = 0,$$

where  $\omega = \partial_p H^0(p)$  and  $\partial_p H^0 : I \rightarrow \Omega$  is the so-called *frequency map*. In these action-angle coordinates, the phase space becomes  $\mathbb{T}^n \times I$  where  $I \subseteq \mathbb{R}^n$  and the dynamics of the system are completely expressed as rotations on the torus. Specifically, phase space is foliated into a family of invariant tori  $\mathbb{T}^n \times \{p\}$  for each  $p \in I \subseteq \mathbb{R}^n$ . We consider only integrable Hamiltonians with a *non-degenerate* frequency map, that is,  $\det \partial_p^2 H^0 \neq 0$ . Now, KAM deals with Hamiltonians of the form

$$H(q, p) = H^0(p) + \varepsilon H^1(q, p),$$

where  $1 \gg \varepsilon > 0$  is considered small,  $H^0$  is the integrable part and  $H^1$  is the perturbation. We assume that  $H$  is  $2\pi$ -periodic in each component of  $q$ . What KAM theory ensures is that under the correct conditions, a “large” subset  $\Omega_{\gamma, \tau} \subseteq \Omega$ ,  $\gamma, \tau > 0$  of invariant tori of  $H^0(p)$  are preserved, though possibly deformed, under the perturbation  $H^1$ . The set  $\Omega_{\gamma, \tau}$  is given by:

$$\Omega_{\gamma, \tau} = \bigcap_{\substack{k \in \mathbb{Z} \\ k \neq 0}} \{ \omega \in \Omega : |\omega \cdot k| \geq \gamma |k|^{-\tau} \}. \quad (4)$$

The condition for  $\Omega_{\gamma, \tau}$  is called the *small divisor condition*. It can be shown for  $\tau > n - 1$  that for almost all  $x \in \mathbb{R}^n$  there exists a  $\gamma > 0$  such that  $x \in \Omega_{\gamma, \tau}$ , so

in particular, we can find  $\gamma > 0$  so that a point  $\omega \in \Omega$  satisfies  $\omega \in \Omega_{\gamma,\tau}$ . We finally consider the *Cantor set*

$$\hat{\Omega}_{\gamma,\tau} = \Omega_{\gamma,\tau} \cap \{\omega \in \Omega : d(\omega, \partial\Omega) \geq \gamma\},$$

that is, we remove points in  $\Omega_{\gamma,\tau}$  that have distance less than  $\gamma$  from the boundary of  $\Omega$ . It can be shown  $\Omega \setminus \bigcup_{\gamma>0} \hat{\Omega}_{\gamma,\tau}$  is a set of measure zero, so the measure of  $\hat{\Omega}_{\gamma,\tau}$  becomes large for small  $\gamma$ , justifying the term “large”. We can now give the KAM theorem as stated in [Pö82].

**Theorem 1.** Let the integrable Hamiltonian  $H^0 : \mathbb{T}^n \times I \rightarrow \mathbb{R}$  be real analytic and non-degenerate, such that the frequency map  $\partial_p H^0 : I \rightarrow \Omega$  is a diffeomorphism and let the perturbed Hamiltonian  $H = H^0 + \varepsilon H^1$  be of class  $C^{\alpha\lambda+\lambda+\tau}$  with  $\lambda > \tau + 1 > n$  and  $\alpha > 1$ . Then there exists a positive  $\gamma$ -independent  $\delta$  such that for  $|\varepsilon| < \gamma^2 \delta$  with  $\gamma$  sufficiently small, there exists a diffeomorphism

$$\mathcal{T} : \mathbb{T}^n \times \Omega \rightarrow \mathbb{T}^n \times I,$$

which on  $\mathbb{T}^n \times \hat{\Omega}_{\gamma,\tau}$  transforms the equations of motion of  $H$  into

$$\dot{\theta} = \omega, \quad \dot{\omega} = 0.$$

The map  $\mathcal{T}$  is of class  $C^\alpha$  for non-integer  $\alpha$  and close to the inverse of the frequency map; its Jacobian determinant is uniformly bounded from above and below. In addition, if  $H$  is of class  $C^{\beta\lambda+\lambda+\tau}$  with  $\alpha \leq \beta \leq \infty$ , then one can modify  $\mathcal{T}$  outside  $\mathbb{T}^n \times \hat{\Omega}_{\gamma,\tau}$  so that  $\mathcal{T}$  is of class  $C^\beta$  for noninteger  $\beta$ .

So, for  $\omega \in \hat{\Omega}_{\gamma,\tau}$ , we parametrize an invariant torus via the map  $\theta \mapsto \mathcal{T}(\theta, \omega)$ . There are a few theorems in use now that are titled the *KAM theorem*, and they differ mainly whether they discuss analytic or smooth perturbations. It is easier to find sources discussing the analytic versions, since they provide stronger results about the invariant torii. Having said this, we use the  $C^r$  version because it is easier to construct smooth approximations of discontinuous functions as opposed to analytically approximating them. We bring smooth approximations into the mix, since (1) alone is discontinuous.

## 2.2 Approximating locally $L^1$ functions

The Hamiltonian (1) we wish to study is discontinuous, which by itself is not suitable for the KAM theorem. We can, however, smoothly approximate the Hamiltonian by using *mollifiers*. The KAM theorem then can be applied to the smoothed Hamiltonian, which of course means we are not directly studying (1) but instead gaining an intuition for the true behavior.

The *standard mollifier*  $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}$  is the following function:

$$\varphi(x) = \begin{cases} c \exp\left(\frac{1}{|x|^2-1}\right), & |x| < 1 \\ 0, & |x| \geq 1, \end{cases}$$

where  $c > 0$  is a scaling factor chosen so that the integral of  $\varphi$  over  $\mathbb{R}^n$  is 1. Also,  $\varphi$  is commonly called a *bump* function, since its support is compact. For  $\varepsilon > 0$ , let

$$\varphi_\varepsilon(x) = \frac{1}{\varepsilon^n} \varphi\left(\frac{x}{\varepsilon}\right),$$

this function has the following properties:

$$\begin{aligned} \varphi_\varepsilon &\in C_c^\infty(\mathbb{R}^n), \quad \varphi \geq 0, \\ \int_{\mathbb{R}^n} \varphi_\varepsilon &= 1, \quad \text{supp}(\varphi_\varepsilon) \subset B_\varepsilon(0) = \{x \in \mathbb{R}^n : |x| < \varepsilon\}, \end{aligned}$$

that is, the function  $\varphi_\varepsilon$  is smooth in  $\mathbb{R}^n$  with compact support, it is positive, its integral is 1, and the support of  $\varphi_\varepsilon$  is fully contained in the unit ball of radius  $\varepsilon > 0$  centered at the origin.

Let  $f \in L_{\text{loc}}^1(X)$  be a locally integrable function in  $X \subseteq \mathbb{R}^n$ . The *mollification* of  $f$  is defined as the convolution of  $f$  with  $\varphi_\varepsilon$ , that is,  $\varphi_\varepsilon * f : X_\varepsilon \rightarrow \mathbb{R}$  where  $X_\varepsilon = \{x \in X : d(x, \partial X) > \varepsilon\}$ . Explicitly,

$$\begin{aligned} f_\varepsilon(x) &= (\varphi_\varepsilon * f)(x) = \int_X \varphi_\varepsilon(x - y) f(y) dy \\ &= \int_{B_\varepsilon(0)} \varphi_\varepsilon(y) f(x - y) dy, \quad x \in X_\varepsilon \end{aligned}$$

Some properties that the mollification  $f_\varepsilon$  has are summarized here:

**Theorem 2.** Let  $f \in L_{\text{loc}}^1(X)$ . Then the mollification  $f_\varepsilon$  has the following properties:

1.  $f_\varepsilon \in C^\infty(X_\varepsilon)$ ,
2.  $f_\varepsilon \rightarrow f$  almost everywhere as  $\varepsilon \rightarrow 0$ ,
3. if  $f$  is continuous on  $X$ , then  $f_\varepsilon \rightarrow f$  as  $\varepsilon \rightarrow 0$  uniformly on compact subsets of  $X$ ,
4. if  $1 \leq p < \infty$  and  $f \in L_{\text{loc}}^p(X)$ , then  $f_\varepsilon \rightarrow f$  as  $\varepsilon \rightarrow 0$  in  $L_{\text{loc}}^p(X)$

*Proof.* The proof of this theorem can be found in Appendix C of [Eva98] ■

What we gain from Theorem 2 is not only a smooth approximation of our discontinuous Hamiltonian but an approximation that can be made arbitrarily precise almost everywhere. Of course, the points which cannot be approximated accurately are concentrated at the boundary of each disk, where the discontinuities lie. Despite this, it is reasonable to assume that for sufficiently small values of  $\varepsilon$ , the flow of the equations of motion provided by the mollified Hamiltonian approximate the flow of the discontinuous one very well.

### 2.3 Investigating small magnetic field strengths

In fig. 2 we numerically solve the system and observe some interesting relations. Each plot shows 5 trajectories varying  $b = 10^{-1}, \dots, 10^{-5}$ . In fig. 2a and fig. 2b the init. cond. and params. are the same  $X \approx (0.38, 0.81)$  and  $V \approx (0, 1)$ , and  $R = 1/3$ , only the duration of the simulation is longer in fig. 2b. In fig. 2c,  $V$  is the same,  $X \approx (0.44, 0.65)$  and  $R = 1/6$ .

First, we see in each figure the trajectory for  $b = 0.1$  appears as a smear near the origin, this is likely because  $b$  is large enough for the deflection in each disk to be significant. We note, however, in later sections we also find circle-like quasiperiodic trajectories for significantly larger values of  $b$ . The difference here

is that these circular arcs seem to be generic, while for larger  $b$  the circle-like trajectories are more scarce and each has a smaller region of stability.

For the other values of  $b$ , we see the trajectories follow a circular arc, and for smaller  $b$  the radius of the respective circular arc is larger. In the limit  $b \rightarrow 0$ , the trajectories will straighten out, since the magnetic field vanishes. These two observations motivates the use of KAM, specifically, considering (1) as a perturbation of either free motion or of a uniform magnetic field in the plane.

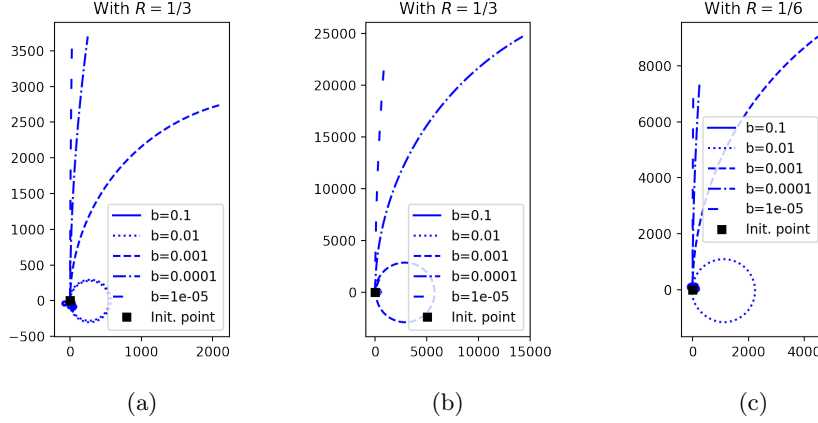


Figure 2: Trajectories for various choices of  $R$  and small  $b$ .

We note that the free motion perturbation idea seems to be less robust than the uniform field idea. For any small  $b > 0$ , we can expect that after a sufficiently long time  $t$  the accumulated deflection is substantial, for example, compare the trajectory for  $b = 0.0001$  in fig. 2a and fig. 2b. Having said that, it is not visible in the figure but what appear as circular trajectories in fact do not close either, the displacement is simply very small, but this is precisely what motivated the previous comment: the circles stay circular for longer.

## 2.4 Perturbations of linear motion

Recall, that a particle in the plane moves along Larmor circles when in the presence of a uniform magnetic field, with strength  $b$ , orthogonal to the plane of motion. The Larmor radius  $R$  depends inversely with respect to the field strength,  $Rb \propto 1$ , that is, a weaker the field strength corresponds to a larger Larmor radius. So, as  $b \rightarrow 0$ , we expect  $R \rightarrow \infty$ , which means that locally the trajectory approaches linear motion in the plane. This intuition is corroborated when considering the Hamiltonian (1) and setting  $b = 0$ . Hence, it is natural to consider the Hamiltonian  $H_0(q, p) = \|p\|^2/2$  and perturb it.

We see  $H_0$  is real analytic, it is also in action-angle coordinates from which we can deduce it is non-degenerate  $\det \partial_p^2 H_0 = 1 \neq 0$ , and the frequency map  $\partial_p H_0(p) = p$  is a diffeomorphism. Now, we attempt to isolate  $H_0$  in (1) from



the perturbation term. Writing  $A = (A_1, A_2, A_3)$ , we see:

$$\begin{aligned} H(q, p) &= \frac{1}{2} \|p - A(q)\|^2 = \frac{1}{2} (p_1 + A_1(q))^2 + \frac{1}{2} p_2^2 \\ &= \frac{\|p\|^2}{2} + \frac{1}{2} (2p_1 A_1(q) + A_1^2(q)) \\ &\stackrel{(\star)}{=} H_0(p) + bH_1(q, p, b) \end{aligned}$$

where in  $(\star)$  we used that we can factor out  $b$  from  $A_1(q)$ . As previously discussed,  $H_1$  is discontinuous, so to apply KAM, we need to mollify  $H_1$ . Hence, consider the mollified perturbation  $\tilde{H}_1 = \varphi_\varepsilon * H_1$ , where  $\varepsilon > 0$  is a parameter independent of  $b$ . Now, by Theorem 1, for sufficiently small  $b > 0$  there are tori of  $H_0$  that are preserved under the perturbation  $H_1$ .

## 2.5 Perturbation of motion in a constant field

We first reason heuristically to see that this idea is valid. Comparing the trajectories for  $b = 0.01$  and  $0.001$  in fig. 2a and 2b, we see that the radius of the trajectory is  $\approx 250$  and  $\approx 2500$ , respectively. If we assume this is the value of the Larmor radius  $\hat{L}$  in each case, then we see  $\hat{L} \propto 1/b$  or  $\hat{b} \propto b$ . Now, comparing the trajectories for  $b = 0.01$  in fig. 2a and 2c, we see halving the radius  $R$  of the magnetic bumps roughly quadruples the radius  $\hat{L}$  from 250 to 1000, that is,  $\hat{L} \propto 1/R^2$ , and the strength then relates as  $\hat{b} \propto R^2 b$ . The task then is to determine  $C \in \mathbb{R}$  such that  $\hat{b} = CR^2 b$ .

We can reason another way. Focusing on  $[0, 1]^2$ , we would like the uniform field  $\hat{\mathbf{B}} = \nabla \times \hat{A} = (-\hat{b}q_2, 0, 0)$  to deflect trajectories as would  $\mathbf{B} = \nabla \times A$ . How much a trajectory is deflected depends on the flux of the field, since the flux measures the “flow” of the field through a surface. Equating the fluxes  $\Phi_{\hat{\mathbf{B}}} = \Phi_{\mathbf{B}}$ , we can compute the required strength  $\hat{b}$  for  $\hat{B}$ . The flux of  $\mathbf{B}$  through a bounded region in the plane is given by:

$$\Phi_{\mathbf{B}} = \iint_S \mathbf{B} \cdot d\mathbf{A} = \iint_S \nabla \times A \cdot d\mathbf{A} = \iint_S (0, 0, b) \cdot (0, 0, 1) dA = bA(S),$$

where  $A(S)$  denotes the area of the surface. In our case,  $\Phi_{\hat{\mathbf{B}}} = \Phi_{\mathbf{B}}$  implies  $\hat{b} = \pi R^2 b$ , which is about what we expected. Now, we should numerically test this hypothesis. To test the validity of the relation, we give two tests, the results of which can be found in fig. 3. The first test:

1. For  $1 \leq i \leq 50$ , sample  $(R_i, b_i)$  uniformly from  $[0.25, 0.45] \times [10^{-10}, 10^{-6}]$
2. For  $R_i, b_i$  uniformly sample initial conditions  $X_{ij}, V_{ij}$  with  $1 \leq j \leq 20$ .
3. Using the method in [Coo93], fit a circle to the trajectory of each  $X_{ij}, V_{ij}$ , the radius of which is  $\hat{L}_{ij}$ . We take the average  $\hat{L}_i = \sum_{j=1}^{20} \hat{L}_{ij} / 20$ .
4. Via a least squares method, we fit a general cubic:

$$\begin{aligned} a_0 + a_1 R_i + a_2 b_i + a_3 R_i^2 \\ + a_4 R_i b_i + a_5 b_i^2 + a_6 R_i^3 + a_7 R_i^2 b_i + a_8 R_i b_i^2 + a_9 b_i^3 = 1/\hat{L}_i. \end{aligned}$$

The second test is similar, we fix  $R_i = 1/3$ , and fit a line  $a_0 + a_1 R_i^2 b_i = 1/\hat{L}_i$ . We opt to fit a general cubic function to avoid any bias in reasoning. We should expect after fitting that only  $a_7$  contributes significantly.

The coefficients of the fitted cubic surface in fig. 3a come out to

$$\begin{aligned} a_0 &\approx 6.7 \cdot 10^{-8}, & a_1 &\approx -6.2 \cdot 10^{-7}, & a_2 &\approx 6.7 \cdot 10^{-3}, \\ a_3 &\approx 1.9 \cdot 10^{-6}, & a_4 &\approx -5.3 \cdot 10^{-2}, & a_5 &\approx -1.9 \cdot 10^{+2}, \\ a_6 &\approx -1.8 \cdot 10^{-6}, & a_7 &\approx 3.2, & a_8 &\approx -6.6 \cdot 10^{+1}, \\ a_9 &\approx -3.0 \cdot 10^{-4}, \end{aligned}$$

the values of  $a_0$  to  $a_4$ ,  $a_6$  and  $a_9$  are negligible as expected, likewise  $a_7 \approx \pi$ . We notice that  $a_5$  and  $a_8$  are quite large but we reason that the contribution of their respective monomial term is still small, since both contain a factor of  $b^2$  which has an order of magnitude at most  $10^{-6}$ . The coefficients of the fitted line in fig. 3b are  $a_0 \approx 4.4 \cdot 10^{-10}$  and  $a_1 \approx 3.12$ , which can be explained in the same way. So, the relation  $\hat{b} = \pi R^2 b$  seems valid, and this motivates perturbing a uniform magnetic field into the bump field.

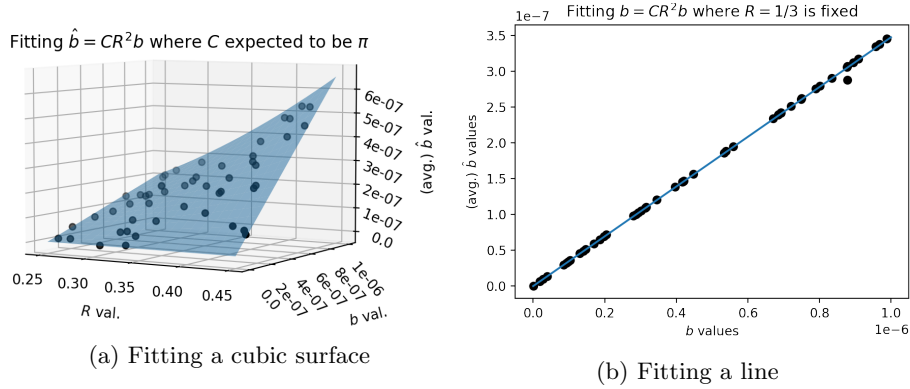


Figure 3: Plots of data and fitted polynomial models.

Overall, the results show that our assumptions are plausible, we approximately see  $\pi$  in the coefficient. The results are not as precise as desired but that can be due to randomly choosing the initial conditions for the trajectories. The circular trajectories correspond to invariant tori, and since not all tori are preserved under the perturbation, we expect that, chosen at random, some trajectories will not follow closely a circular path. Similarly, the chosen range for sampling  $R$  and  $b$  could be too large, though in tests we made that we omit here, we noticed that the relation holds more or less for a wider range of  $[0.1, 0.45] \times [10^{-16}, 10^{-2}]$ .

### 3 Complexity and symbolic dynamics

In this section we consider the dynamics of (1) for  $b \gg 0$ , that is, in the case where KAM and perturbative methods are not readily applicable. With no clear battle plan, we approach the question in an exploratory way: we look for (quasi)-periodic orbits, consider their stability, and see where stability is missing. To this end, we prove the existence of a Poincaré section, later we induce “coarse” symbolic dynamics and apply the Lempel-Ziv complexity to make sense of the dynamics. What we find is rich dynamics and a visual method of analysis well suited for similar problems.

The phase space of our system is  $\mathbb{R}^2 \times \mathbb{R}^2$ , however we can reduce the phase space to a Poincaré section as follows:

**Proposition 1.** Let  $S$  be the union of discs of radius  $R$  centered at  $\mathbb{Z}^2 + 1/2$ . The sets  $S_{\text{in}}$  and  $S_{\text{out}}$  defined as:

$$S_{\text{in}} = \bigcup_{x \in \partial S} \{x\} \times \{v \in \mathbb{R}^2 : v \cdot (x - 1/2) < 0\}, \quad (5)$$

$$S_{\text{out}} = \bigcup_{x \in \partial S} \{x\} \times \{v \in \mathbb{R}^2 : v \cdot (x - 1/2) > 0\}, \quad (6)$$

are Poincaré sections for the system (1).

We prove this in a later subsection. This greatly helps visualize long term behavior, since the magnitude of the velocity of a trajectory is constant. Furthermore, we can pass the system to the torus, which reduces  $S$  to one disc. So, when plotting we only need two dimensions: one to parametrize  $\partial S$  and another for the velocity.

#### 3.1 First impressions and lots of quasiperiodicity

For the remainder of this section we fix  $R = 1/3$  and  $\|p\| = 1$ . In 4

#### 3.2 Constructing a Poincaré section

**Lemma 1.** The flow of (1) induces a well-defined map  $P_{\text{io}} : S_{\text{in}} \rightarrow S_{\text{out}}$ .

*Proof.* Let  $(x, v) \in S_{\text{in}}$ , under the flow of (1) we know the Larmor circle  $C$  with initial conditions  $(x, v)$  intersects  $\partial S$  in  $x$  and that at  $x$  the circles are not tangential. Two circles intersecting at a point non-tangentially must intersect at a second point, so there exists  $x \neq y \in \partial S \cap C$ . By the flow of (1) there must also exist a velocity  $w \in \mathbb{R}^2$  such that  $w \cdot (x - 1/2) > 0$ , that is  $(y, w) \in S_{\text{out}}$ . Hence we have  $P_{\text{io}}(x, v) = (y, w)$ . ■

We can say more, there exists a line  $\ell$  through the center  $a$  of the Larmor circle  $C$  and the center  $(1/2, 1/2)$  of  $\partial S$ . The two circles are symmetric with respect to the reflection  $T$  with respect to  $\ell$ . This means  $Tx = y$ , and  $Tv = -w$ . The following statement is more interesting.

**Lemma 2.** The flow of (1) induces a well-defined map  $P_{\text{oi}} : S_{\text{out}} \rightarrow S_{\text{in}}$ .

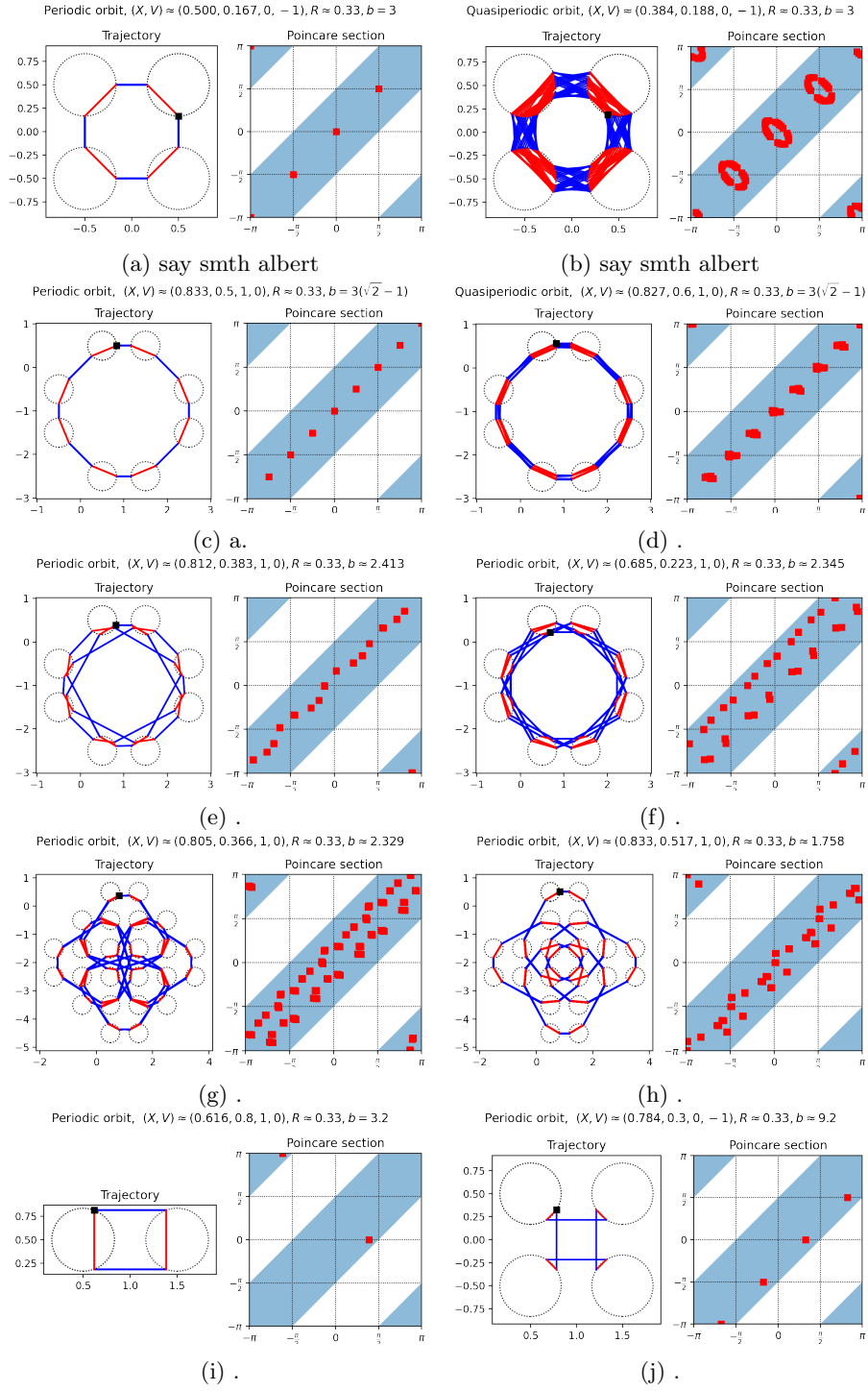


Figure 4: Examples of (quasi)-periodic for varying choices of parameters.

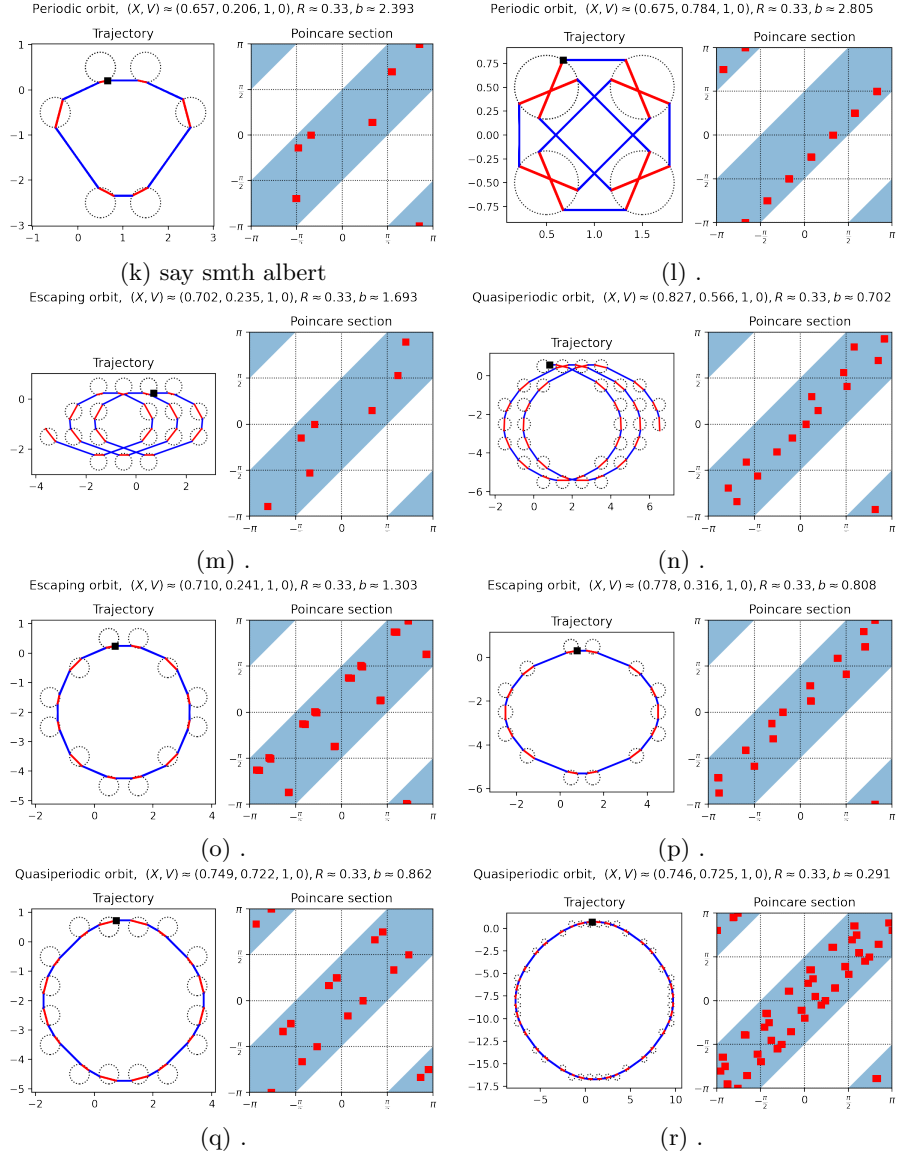


Figure 4: Examples of (quasi)-periodic for varying choices of parameters.

*Proof.* Let  $(x, v) \in S_{\text{out}}$ , then under the flow of (1) we see linear motion. Ignoring the presence of  $S$ , we see that depending on  $v$ , the trajectory is either periodic or is dense in  $\mathbb{T}$ .

Focusing on the first case, in finite time we return to the point  $x$ . Since the trajectory is a line  $\ell$  intersecting  $\partial S$  transversally, there must be another point of intersection  $x \neq y \in \partial S \cap \ell$ . The velocity at  $y$  must also be  $v$ , so due to the geometry of a circle,  $(y, v) \in S_{\text{in}}$ . For the second case, we can pick  $\varepsilon > 0$  such that all points in  $U = (\partial S \cap B_\varepsilon(x)) \times \{v\} \subset S_{\text{out}}$  are transversal to  $\partial S$ . Let  $(y, v) \in U$ , since the flow is dense for the velocity  $v$ , ignoring  $S$ , we know that in finite time the point  $(x, v)$  comes to  $(y, v)$ . The rest of this case follows the

same way as previously.

In either case, the  $(y, v)$  we find is not necessarily the first time we enter  $S_{\text{in}}$  but since the number of such points is countable, we can pick the first one  $(y_0, v)$ , hence,  $P_{\text{oi}}(x, v) = (y_0, v)$  is a well-defined map. ■

Now, we can finalize the Poincaré map.

Notice that the reasoning is independent of the radius of  $S$ , nor does it depend on the exact position of  $S$ . Pulling back to the plane with an infinite number of disks, we can then generalize the setup. Consider a fundamental finite set  $A$  of disks where each disk in  $A$  can have a different magnetic strength and radius and the disks need not be regularly placed. The above result holds for configurations of disks which can be decomposed as a tiling of such a set  $A$ .

In [KS17] a similar result is proved for a configuration of finitely many bumps. In that case a different method was used that did not rely on an infinite number of bumps, in ours the reasoning was simplified due to this.

With the Poincaré map we can study the stability of (quasi)-periodic trajectories of the system for a choice of parameters. In fig. 4 we provide some examples we constructed using analytic geometry. In fig. 4a we see a periodic trajectory and its corresponding orbit in the Poincaré section. In fig. 4b we shift the initial condition slightly and see that the orbit in fig. 4a displays stable behavior. We provide a proof using symbolic algebra in [HERE](#). The same can be said comparing fig. 4c and fig. 4d. On the other hand, in [add plots with unstable periodic orbit](#) we have a periodic trajectory that does not seem to be stable.

Constructing periodic trajectories and judging their stability by checking numerically is a reliable approach, however it is limited, since we need a good guess to start with. Uncovering more elaborate (quasi)-periodic behavior in this manner becomes impractical, so we need a tool that can help us single out good candidates ahead of time. We discuss this in the next section.

### 3.3 Difficulties with symbolic dynamics

In the previous section we discretized the continuous time system via a Poincaré map, and studied some examples of (quasi)-periodic orbits directly. To study the discrete system further, a typical approach taken is to construct a Markov partition, and to pass to a symbol space with a shift map. Providing a Markov partition in our case proved more complicated than expected. We need a better understanding of the (un)stable manifold of a point  $x \in S_{\text{out}}$ . Furthermore, it is not exactly clear how to “carve” up  $S_{\text{out}}$  in the first place.

To avoid this, we decided to introduce the Lempel-Ziv complexity as described in [LZ76]. The Lempel-Ziv complexity operates on finite length sequences of symbols by applying a compression algorithm, the complexity of the original sequence is then quantified by the result of the compression. Partitioning  $S_{\text{out}}$  and prescribing an alphabet is still an issue but we first discuss Lempel-Ziv.

### 3.4 The Lempel-Ziv compression algorithm

To explain the algorithm, we use a series of examples, as in the paper [KS87]. We are not interested in the technicalities of the original paper by Lempel and

Ziv, so we only cite results when needed.

Consider a sequence  $s = s_1 s_2 \dots s_n$  of length  $n \in \mathbb{N}$ , the algorithm decides what is the smallest number of “words” in the sequence necessary for reconstruction. Suppose we have reconstructed  $s$  up to the index  $k < n$  and the word counter is  $c$ , that is, we have  $s_1 \dots s_k$ . For the next iteration, the algorithm decides what is the largest  $k < \ell \leq n$  such that the subsequence  $s_{k+1} \dots s_\ell$  appears at some index in  $s_1 \dots s_{\ell-1}$ . Once this  $\ell$  is found, the word counter is increased to  $c+1$ . If  $\ell = n$ , then the algorithm terminates, otherwise the process is repeated with  $s_1 \dots s_{\ell+1}$ . We provide an example below. The  $\cdot$  is used as a delimiter between words, the top line indicates the longest subsequence found at each iteration, and the bottom line shows where they can be found.

$$\begin{aligned}
\bar{0}1011010001101110010 &\xrightarrow{(1)} 0 \cdot \bar{0}1011010001101110010 \\
&\xrightarrow{(2)} \underline{0 \cdot 1} \cdot \bar{0}11010001101110010 \\
&\xrightarrow{(3)} \underline{0 \cdot 1 \cdot 011} \cdot \bar{0}1\bar{0}001101110010 \\
&\xrightarrow{(4)} 0 \cdot 1 \cdot \underline{011 \cdot 0100} \cdot \bar{0}1\bar{0}1\bar{1}110010 \\
&\xrightarrow{(5)} 0 \cdot 1 \cdot 011 \cdot \underline{0100} \cdot 011011 \cdot \bar{0}010 \\
&\xrightarrow{(6)} \underline{0 \cdot 1 \cdot 011 \cdot 0100} \cdot 011011 \cdot 1001 \cdot \bar{0} \\
&\xrightarrow{(7)} 0 \cdot 1 \cdot 011 \cdot 0100 \cdot 011011 \cdot 1001 \cdot 0
\end{aligned}$$

At the start, we scan from the left, and notice we have never encountered a 0, so for step (1) we add 0 on its own. Next, we encounter 1 for the first time, and add it in step (2). Next, we see that we have encountered 01, so we add 011. Notice, how we ignore the delimiter between words. We indicate the rest of the steps without commentary. Once the algorithm terminates, we see that the number of words is 7, so the complexity of this sequence is 7.

The complexity is not normalized to  $[0, 1]$ , this is not needed since we only ever consider finite length sequences and those clearly have bounded LZ complexity. Comparing two sequences of varying length, the sequences can have the same complexity, though it is expected that the longer sequence will have a higher complexity. Now, that we have an idea how the Lempel-Ziv (LZ) complexity is computed, we can discuss how LZ captures the regularity (or lack of it) in symbolic dynamics.

Suppose for now, we have a map taking orbits of a system to elements of a symbol space, then we can distinguish three levels for LZ complexity: high, low, and intermediate. Since LZ is not normalized, and varies depending on the length of the sequences we consider, it is best to work with sequences of the same length. Furthermore, provided we fix the length, the boundary between cases is not well defined. A rule of thumb is to judge the level by comparison, for each system there will be an average complexity for a sampling of orbits, so low complexity will be well below the average and high will be well above. The average can be skewed depending on the sampling but this should be discussed on a case by case basis.

We can characterize the orbits by their complexity as follows:

- If a sequence has high complexity that means we cannot reconstruct the sequence from a small number of words. A word corresponds to some substructure in the orbit, and if there are many unique substructures, we should expect the orbit to be disorganised or chaotic.
- If the complexity is low, meaning very few words are required to reconstruct the sequence, then the corresponding orbit should be regular and repetitive. So, we would expect a (quasi)-periodic orbit to have a low complexity.
- For intermediate complexity the scenario can be mixed. The orbit can appear mostly random with intermittent sections of structure. The orbit can also be (quasi)-periodic but with a significantly longer period for recurrence than an orbit with low complexity.

Of course, this is not formal reasoning and just an intuitive approach. An example where our reasoning breaks down is if a periodic orbit requires  $n$  symbols to describe one period and we only sampled  $n$  symbols. In this case the complexity can look high, instead of low as one would expect from a periodic orbit. In practice, we find it easy to work around these limitations as will be shown in the next section.

### 3.5 Partitioning the Poincaré section and compression

Now, to utilize LZ we need a map from orbits to symbol sequences. We have many options here from the partitioning of the Poincaré section to the size of the alphabet, and we do not necessarily need to use a Markov partition for LZ to give us good results. We decided to use a relaxed approach.

Consider the system in the plane, now via the pullback from the torus,  $S_{\text{out}}$  becomes the outward-pointing boundary of circles centered at lattice points. We know by lemma 2 that a point  $x \in S_{\text{out}}$  is mapped to another point on  $\partial S$ . We disregard the exact point and velocity of  $P_{\text{oi}}(x)$  and only record the relative position of the new circle we intersected. For example, if we have initially  $x, v = (1/2, 5/6, 0, 1)$ , so we are on the circle centered at  $(1/2, 1/2)$ , then  $P_{\text{oi}}(x, v) = (1/2, 7/6, 0, 1)$  which is on the circle centered at  $(1/2, 3/2)$ , the symbol we record here is  $(1/2, 3/2) - (1/2, 1/2) = (0, 1)$ . We see immediately there are infinitely many such pairs  $(n, m) \in \mathbb{Z}^2$  that we can attain. We argue that this is not a problem for computing LZ, since there can only be finitely many of these pairs in a finite sequence anyway, i.e, the complexity of a sequence is capped by its length. Another way to look at it: the pairs  $(n, m) \in \mathbb{Z}^2$  are words constructed using a finite alphabet  $i \in \mathbb{Z}$  with  $|i| \leq 9$ , so we can unpair the coordinates and work with a stream of integers.

An example to get the idea. Consider the orbit in fig. 4a, the orbit goes down, left, up, right, and repeats, so in symbols that is

$$((0, -1), (-1, 0), (0, 1), (1, 0), \dots),$$

and this pattern repeats indefinitely. To compute LZ, we take a finite slice, for example the first 20 symbols. If computed correctly, LZ should be 5, we record the first 4 symbols as words, and the 5th word is the rest of the 16 symbols, since they just repeat the first 4. The complexity in fig. 4b is also 5, though in



fig. 4c and fig. 4d the complexity is 9. We notice also that if we took the first 100 symbols of these orbits, the complexity would still be 5 and 9, respectively, provided they are stable (quasi)-periodic orbits.

### 3.6 Examples and thoughts

We now look at some examples. We find LZ performs well when we compute it for a range of initial conditions and parameters.

Consider again the examples from fig. 4a and fig. 4b, the second one was obtained by slightly perturbing the initial position and this hints that the periodic trajectory is stable. We ask now how much can we perturb the initial conditions before we leave the region of stability for this periodic orbit. We take  $R = 1/3, b = 3$  and compute LZ for the family of initial conditions:

$$x = \left( \frac{1}{2} + \sqrt{R^2 - y^2}, \delta \right), \quad v = (1, 0), \quad \text{for } y \in [-0.32, 0.32] + \frac{1}{2},$$

that is, we keep the velocity the same, and we sweep the right half-circle centered at  $(1/2, 1/2)$ . We report the results in fig. 5a. We see that the region spans values of  $y$  roughly in  $[0.346, 0.653]$ . We also notice there are some dips outside the region of stability. Out of curiosity we plot the orbit of the far left dip in fig. 1, interestingly, the orbit at first is scattered and then gets trapped between four bumps. Later, we computed more iterations of the orbit and noticed it eventually leaves the four bumps, so it seems the orbit came close to the stable region but did not enter it.

In fig. 5b we instead fix the initial conditions  $x, v = (5/6, 1/2, 1, 0)$  and vary  $b \in (0.001, 10)$ . We see that near  $b = 4$ , LZ is consistently high, while for  $b < 4$  there are some dips: the dips that contain  $b = 3$  and  $b = 3(\sqrt{2} - 1) \approx 1.25$  should correspond to the regions of stability for the orbit in fig. 4a, and fig. 4c, respectively. There are other dips in the plot, however we ignore them, they could either be comparatively small stable regions or one of the few “false positives” that we discussed before.

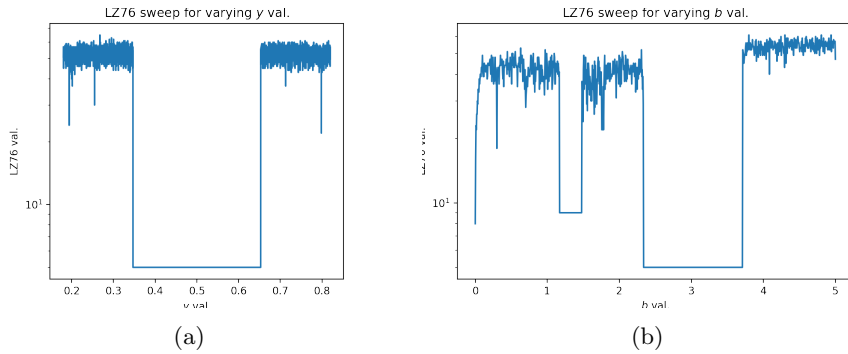
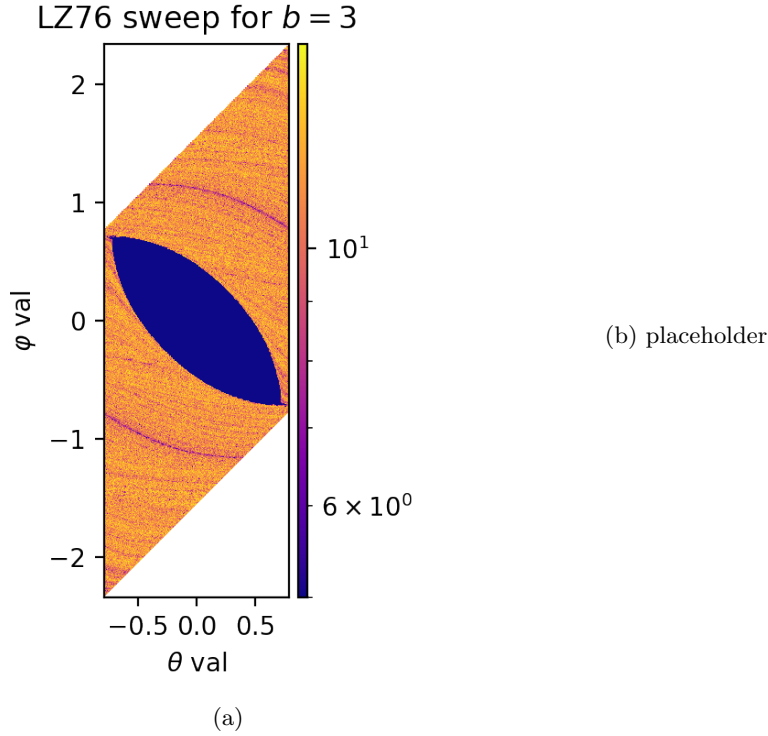


Figure 5

If instead we sample a grid of points on the Poincaré section  $S_{\text{out}}$ , we can compute LZ as in fig. 6a. We only compute LZ for a strip of the Poincaré section, specifically, for  $\theta \in [-\pi/4, \pi/4]$ , since the rest of the picture can be reconstructed via translation. This corresponds to the 4-fold rotational symmetry of a square

lattice, that is, if we rotate  $\mathbb{R}^2$  at the origin by an angle of  $n\pi/2$  for  $n = 0, \dots, 3$ , the lattice of circles stays the same, and the resulting dynamics are the same as well.

In fig. 6a we compute LZ for  $R = 1/3$  and  $b = 3$ . We see a large stable “eye” at  $(\theta, \varphi) = (0, 0)$ . This is not unexpected, since we’ve seen evidence for a large stable region in fig. 5a. Besides the eye, we see some lines that have intermediate complexity, and on those either lie unstable periodic orbits or special occurrences like fig. 1. We include some close ups of the eye: the top left corner of the eye where we notice some obvious fractal behavior, and another spot along the boundary of the eye. We must note again that the emerging features in the images are sensitive to varying the depth of simulation. When attempting to plot the close ups, for example, the second one, we noticed that more yellow strips would appear and their width would increase the longer we simulated. The deeper we simulate, the more the boundary moves, which is not unreasonable since we capture more information.



(b) placeholder

Figure 6

In fig. 7a we plot the same data except  $b \approx 3(\sqrt{2} - 1)$ , so we plot the Poincaré section corresponding to fig. 4c. We notice there are several very stable regions, however by comparing with fig. 4c they should correspond to the same family of (quasi)-periodic orbits. The shape of the very stable regions is quite different, they are twisted with fractal “digits”. Besides the large regions there are intermediate complexity regions scattered about, due to their globular shapes we suspect they are also relatively stable.

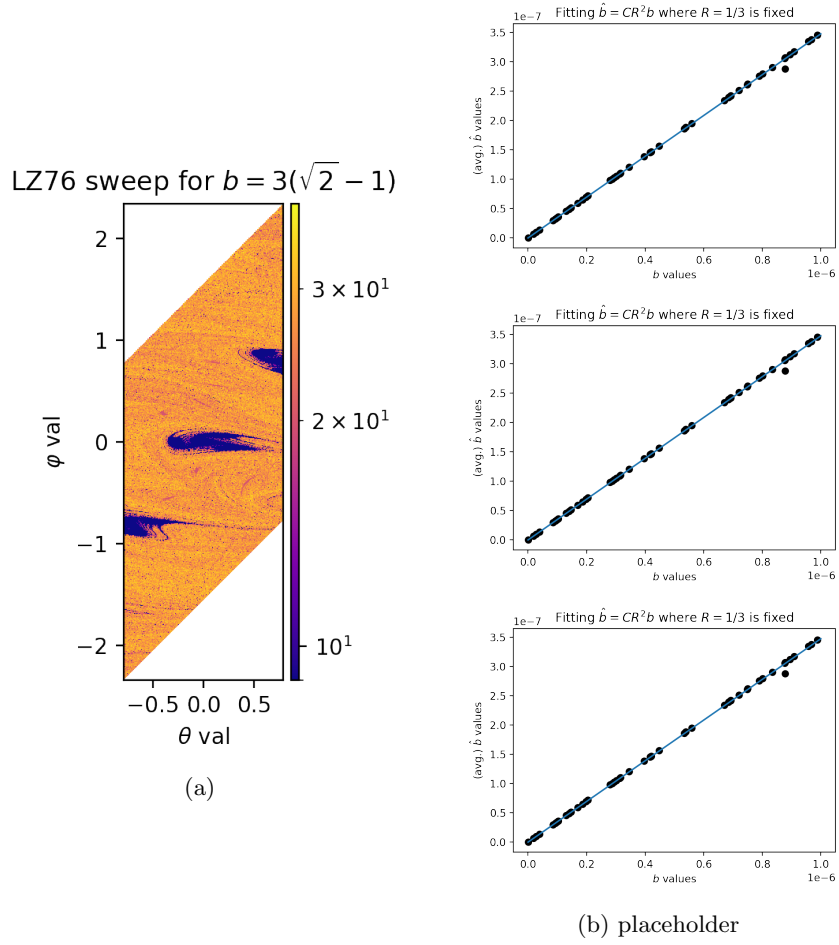
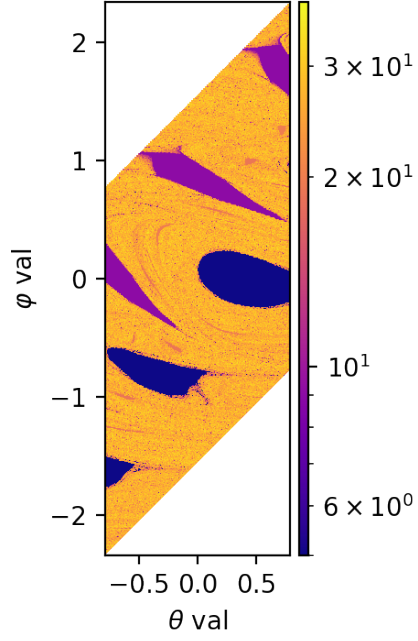


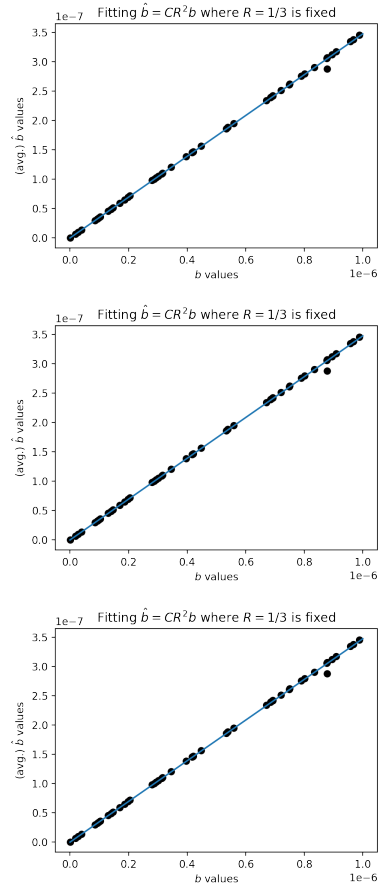
Figure 7

In fig. 8a, we pick a value of  $b = 2.32$  and find a range of stable regions. Generally, the shapes are elongated, exaggerated compared to the other figures and it is not clear why this is the case. It can be noticed that there are 3 different colors for the stable regions, indicating there could be 3 different stable orbits here.

LZ76 sweep for  $b = 2.32$



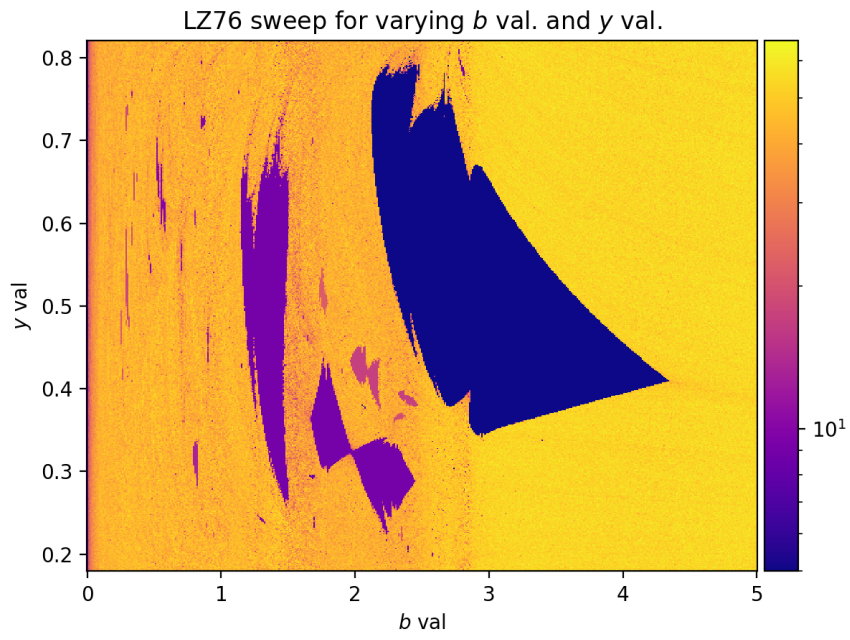
(a)



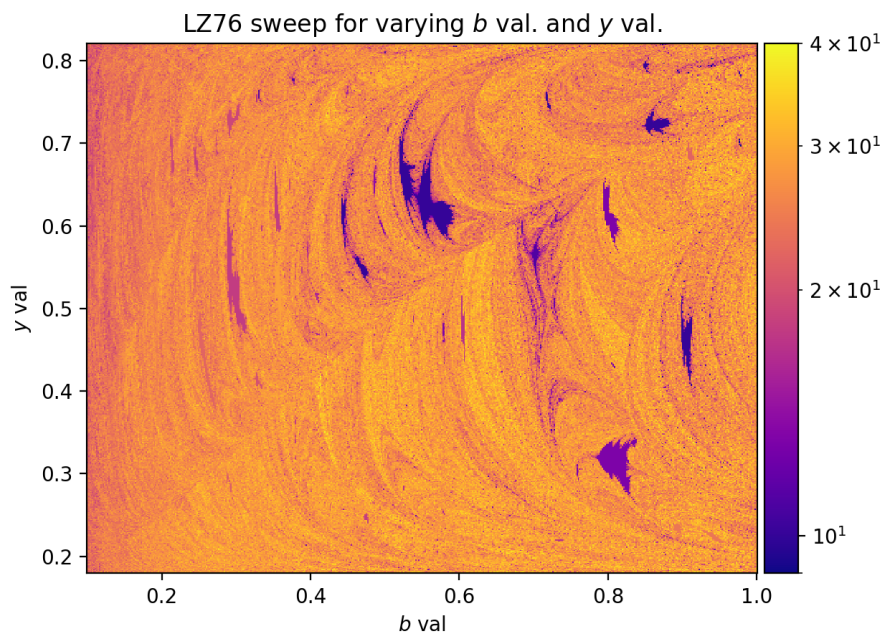
(b) placeholder

Figure 8

## 4 Levy Flights



(a)



(b) placeholder

Figure 9

## 5 Conclusion

## References

- [Coo93] I. D. Coope. Circle fitting by linear and nonlinear least squares. *Journal of Optimization Theory and Applications*, 76:381–388, Feb 1993. <https://doi.org/10.1007/BF00939613>.
- [Eva98] Lawrence C. Evans. *Partial Differential Equations*. American Mathematics Society, 1998. Providence, RI.
- [Gas21] Sean Gasiorsek. On the dynamics of inverse magnetic billiards. *Nonlinearity*, 34(3):1503–1524, mar 2021. <https://doi.org/10.1088/2F1361-6544/2Fabe2f1>.
- [Kna18] Andreas Knauf. *Mathematical Physics: Classical Mechanics*. Springer Berlin, Heidelberg, mar 2018. <https://doi.org/10.1007/978-3-662-55774-7>.
- [KS87] Kaspar and Schuster. Easily calculable measure for the complexity of spatiotemporal patterns. *Physical review. A, General physics*, 36 2:842–848, 1987.
- [KS17] Andreas Knauf and Marcello Seri. Symbolic dynamics of magnetic bumps. *Regular and Chaotic Dynamics*, 22(4):448–454, jul 2017. <https://doi.org/10.1134/2Fs1560354717040074>.
- [LZ76] A. Lempel and J. Ziv. On the complexity of finite sequences. *IEEE Transactions on Information Theory*, 22(1):75–81, January 1976.
- [Pö82] Jürgen Pöschel. Integrability of hamiltonian systems on cantor sets. *Communications on Pure and Applied Mathematics*, 35(5):653–696, 1982. <https://onlinelibrary.wiley.com/doi/abs/10.1002/cpa.3160350504>.
- [Ser22] Marcello Seri. *Hamiltonian Mechanics*. AMS Open Math Notes, mar 2022. <https://www.ams.org/open-math-notes/omn-view-listing?listingId=110861>.