

Hamiltonian Chaos

Haris Skokos

**Department of Mathematics and Applied Mathematics
University of Cape Town, Cape Town, South Africa**

E-mail: haris.skokos@uct.ac.za

URL: http://math_research.uct.ac.za/~hskokos/

Outline

- **Hamiltonian systems**
(Example Hénon-Heiles system)
 - ✓ **Equations of motion**
 - ✓ **Chaos**
 - ✓ **Poincaré Surface of Section**
 - ✓ **Variational equations**
 - ✓ **Lyapunov exponents**

Autonomous Hamiltonian systems

Consider an **N degree of freedom** autonomous Hamiltonian system having a Hamiltonian function of the form:

$$H(\overbrace{q_1, q_2, \dots, q_N}^{\text{positions}}, \overbrace{p_1, p_2, \dots, p_N}^{\text{momenta}})$$

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The **time evolution** of an **orbit** (trajectory) with initial condition

$$P(0) = (q_1(0), q_2(0), \dots, q_N(0), p_1(0), p_2(0), \dots, p_N(0))$$

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Phase space: the $2N$ dimensional space defined by variables $q_1, q_2, \dots, q_N, p_1, p_2, \dots, p_N$

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$$\frac{dp_i}{dt} = -\frac{\partial H}{\partial q_i}, \quad \frac{dq_i}{dt} = \frac{\partial H}{\partial p_i} \Rightarrow \begin{cases} \dot{x} = p_x \\ \dot{y} = p_y \\ \dot{p}_x = -x - 2xy \\ \dot{p}_y = -y - x^2 + y^2 \end{cases}$$

Chaos

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Let V be a set and $f : V \rightarrow V$ a map on this set.

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2. f is **topologically transitive.**
3. **periodic points are dense** in V .

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There exist points arbitrarily close to $\mathbf{x}$ which eventually separate from $\mathbf{x}$ by at least δ under iterations of f .

Not all points near $\mathbf{x}$ need eventually move away from $\mathbf{x}$ under iteration, but there must be at least one such point in every neighborhood of $\mathbf{x}$.

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Consequently, the dynamical system cannot be decomposed into two disjoint invariant open sets.

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3. **An element of regularity** because it has periodic points which are dense.

Usually, in physics and applied sciences, people use the notion of chaos in relation to the sensitive dependence on initial conditions.

Regular vs Chaotic orbits

Hénon-Heiles system

$$H = \frac{1}{2}(p_x^2 + p_y^2) + \frac{1}{2}(x^2 + y^2) + x^2y - \frac{1}{3}y^3$$

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For $H=0.125$ we get a regular and a chaotic orbit with initial conditions (ICs):

$x=0, y=0.1, p_y=0$ and $x=0, y=-0.25, p_y=0$.

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Orbit

Perturbed

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t= 5000	x= 0.376999283889102310	0.376999283870156576
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t= 100	x= 0.090272817735167835	0.090272821355768668
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t= 200	x= 0.295031687482249283	0.295031884858625637
t= 300	x= 0.515226330109450181	0.515225440480693297

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t= 200	x= 0.295031687482249283	0.295031884858625637
t= 300	x= 0.515226330109450181	0.515225440480693297
t= 400	x= 0.063441889347425867	0.061359558551008345

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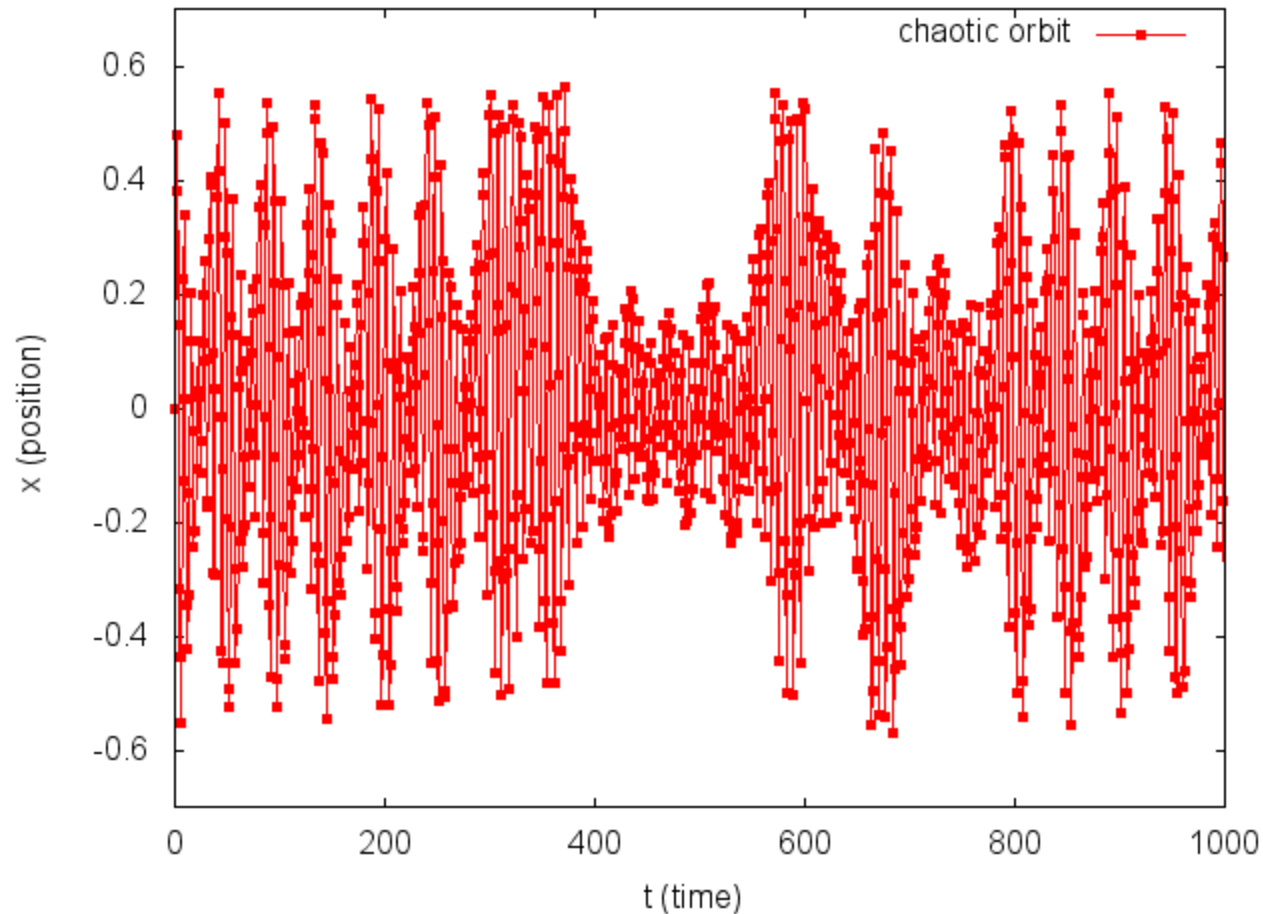
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t= 400	x= 0.063441889347425867	0.061359558551008345
t= 500	x= 0.078357719290523528	-0.270811022674341095

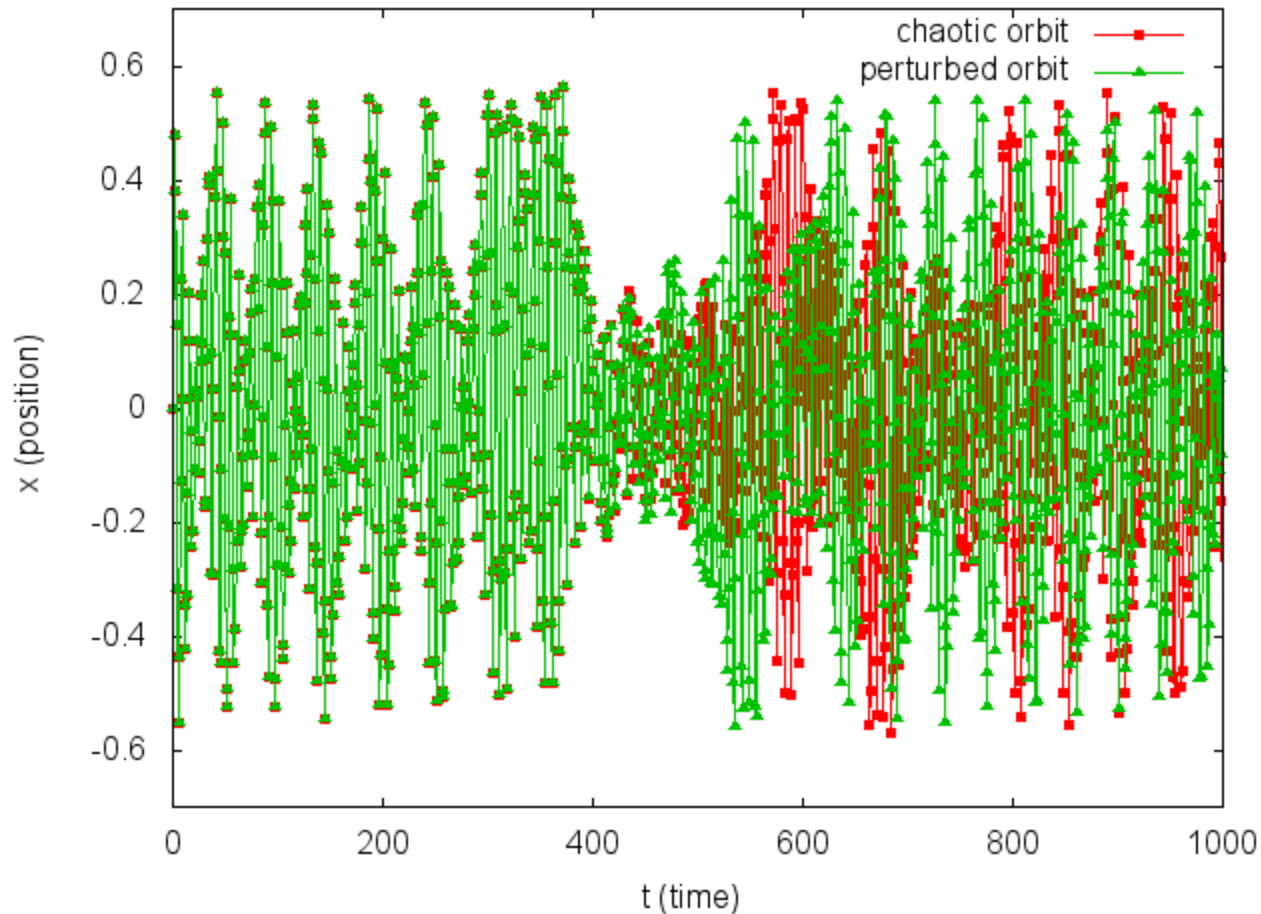
Regular vs Chaotic orbits

Chaotic orbit



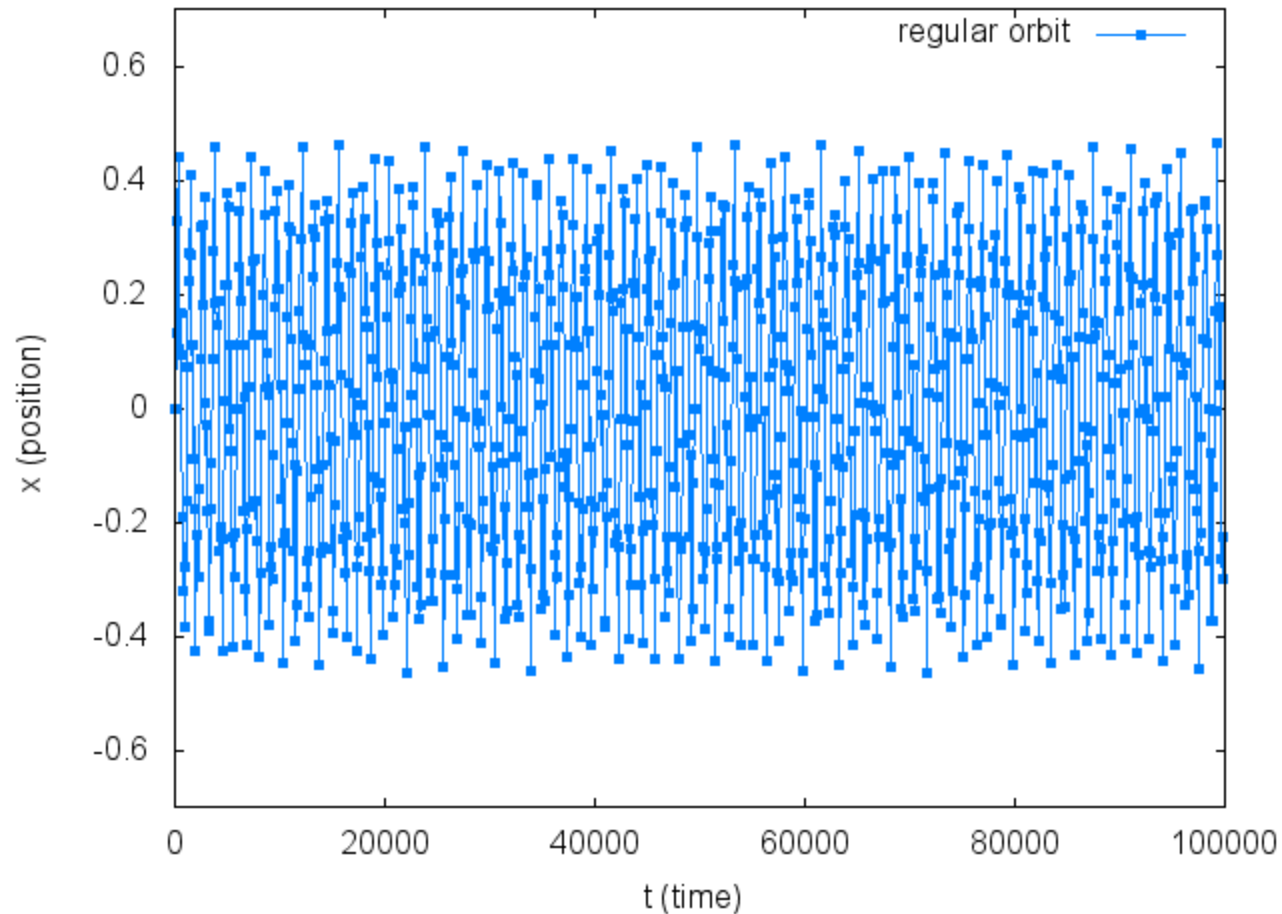
Regular vs Chaotic orbits

Chaotic orbit and its perturbation



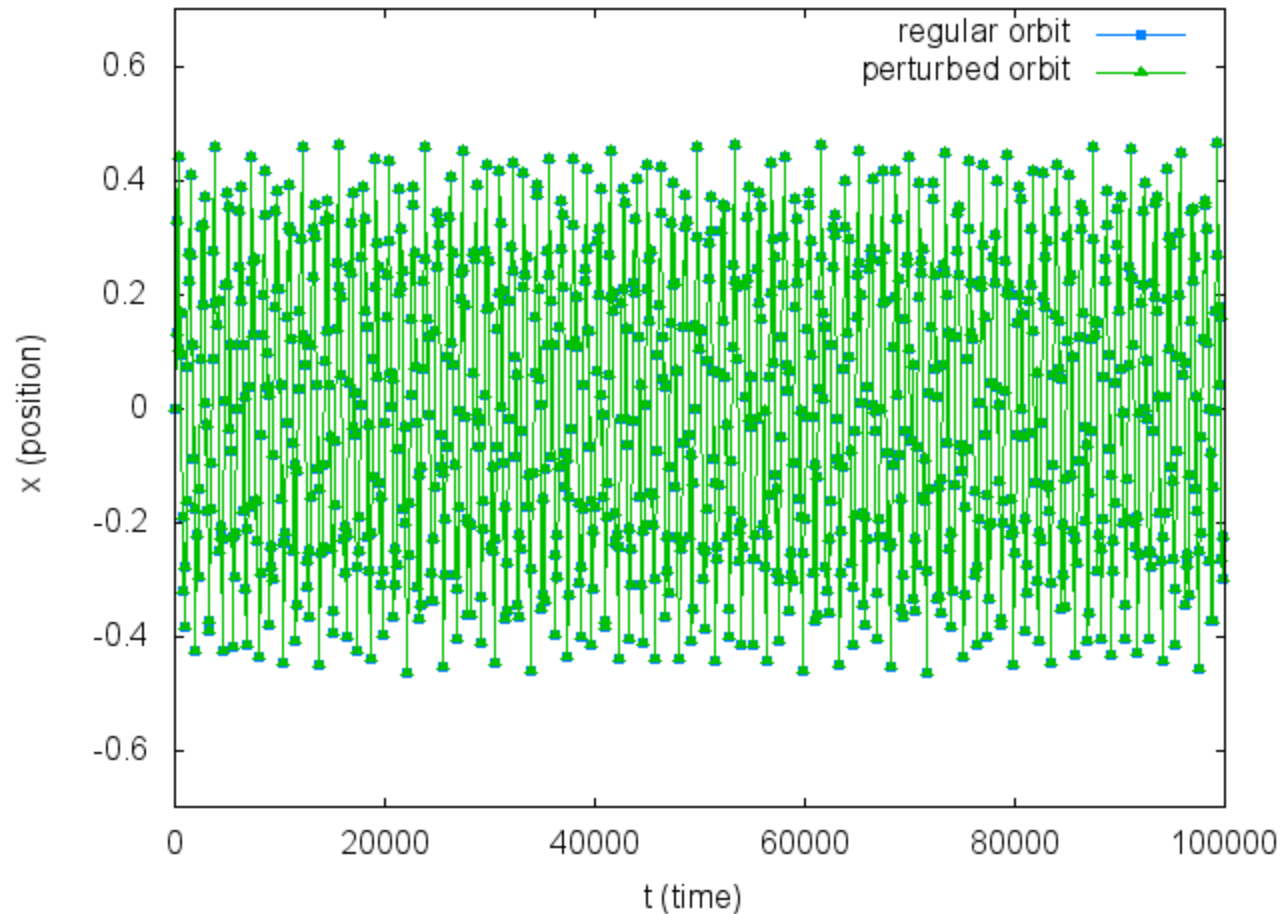
Regular vs Chaotic orbits

Regular orbit



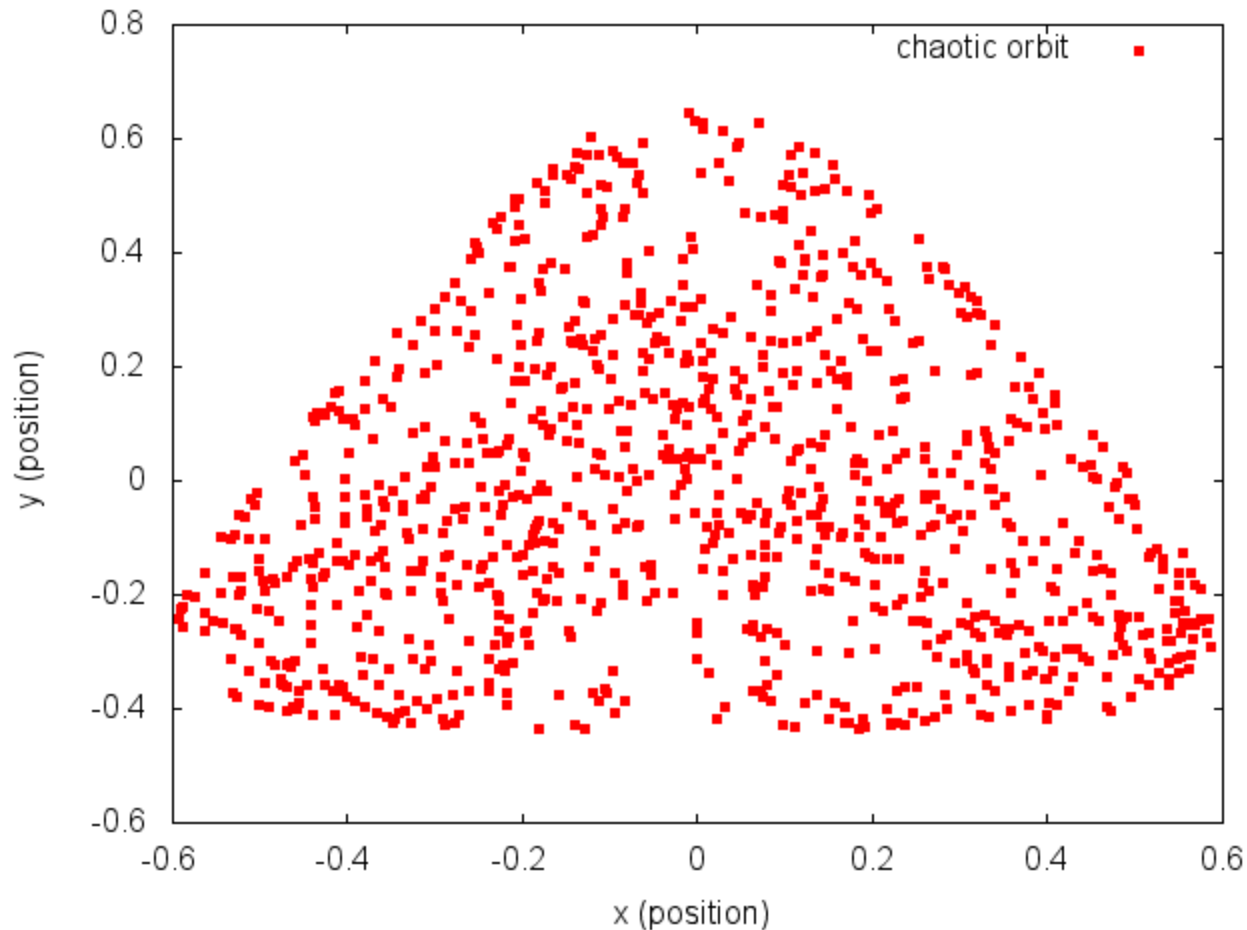
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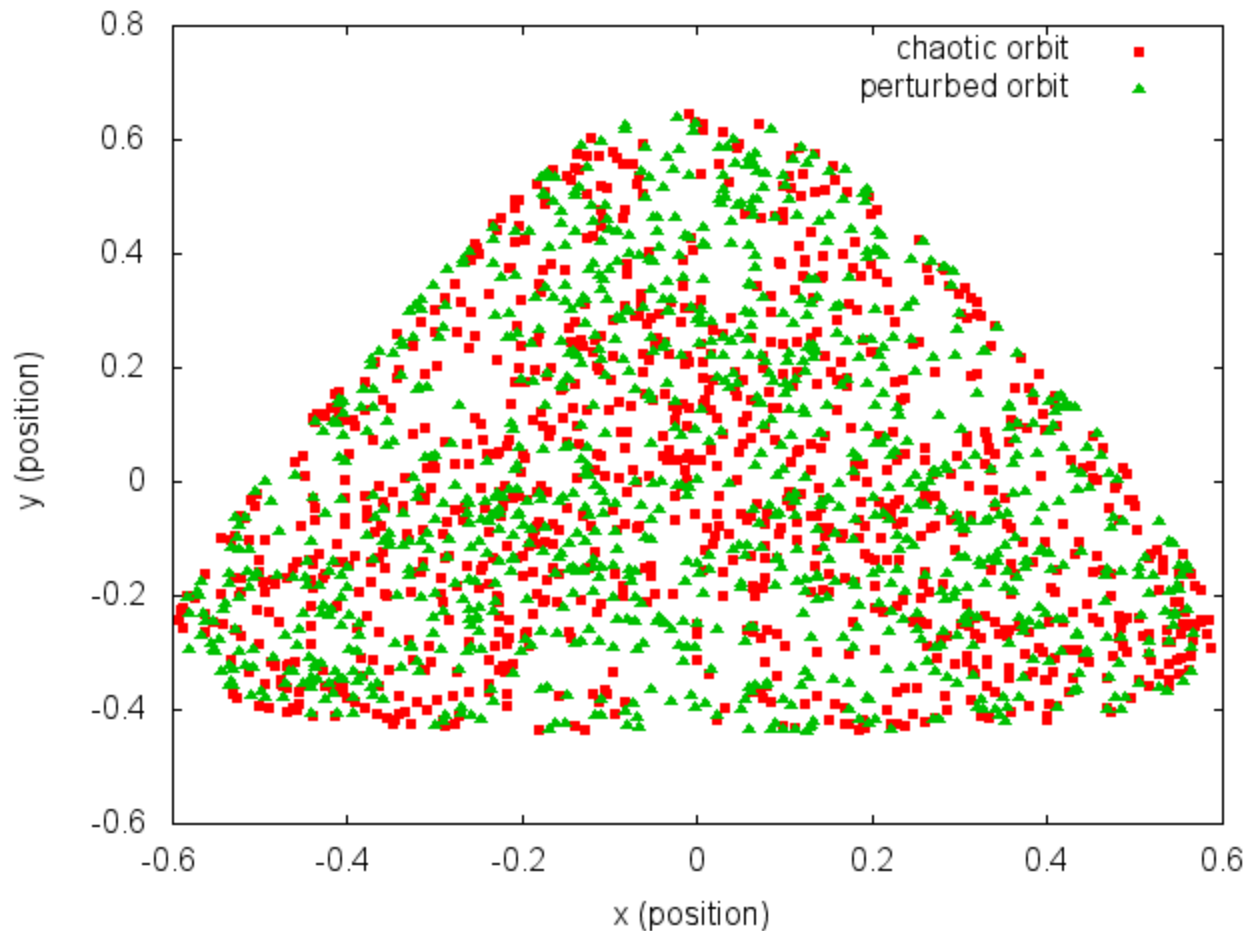
Chaotic orbit



Results for $0 \leq t \leq 10^5$

Regular vs Chaotic orbits

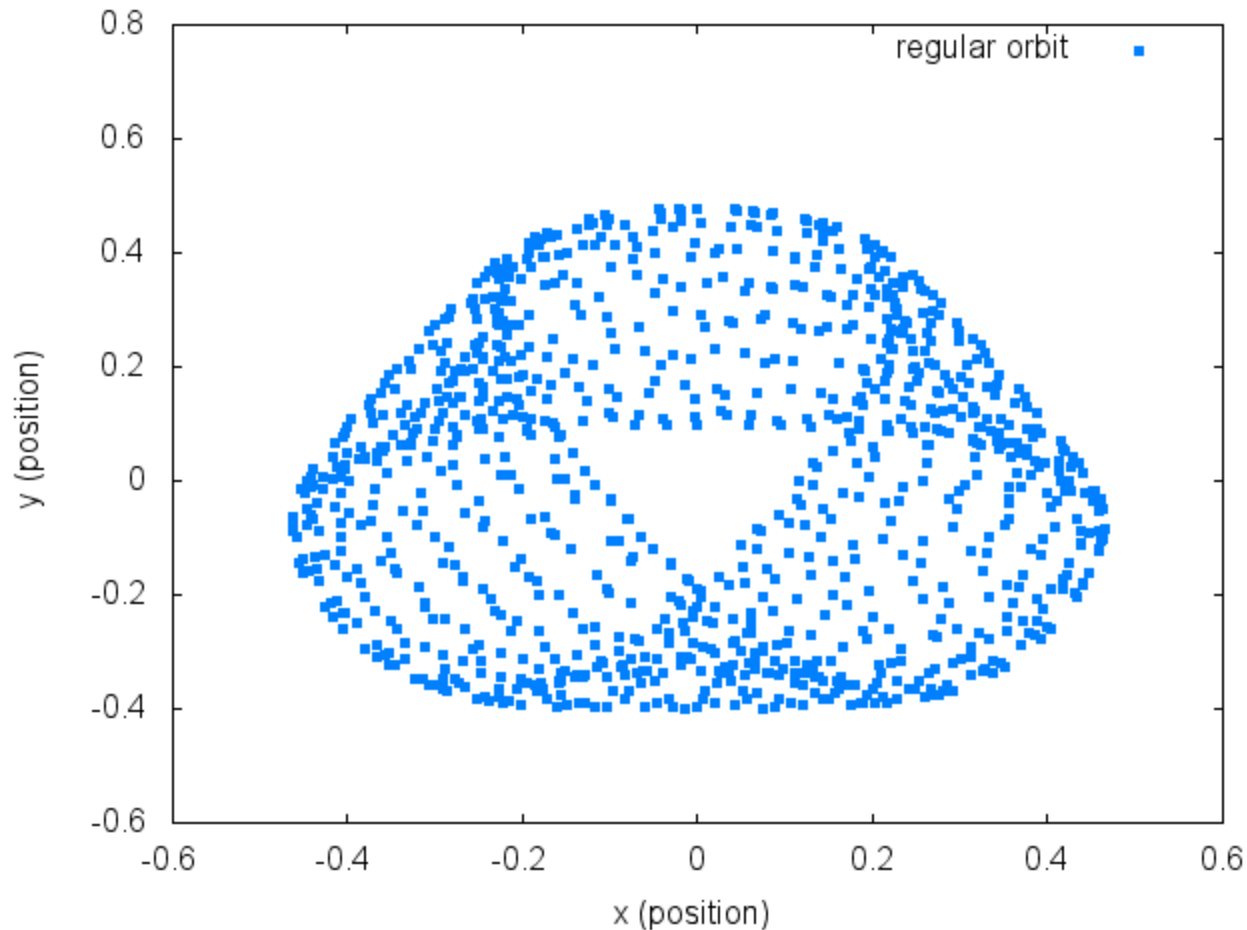
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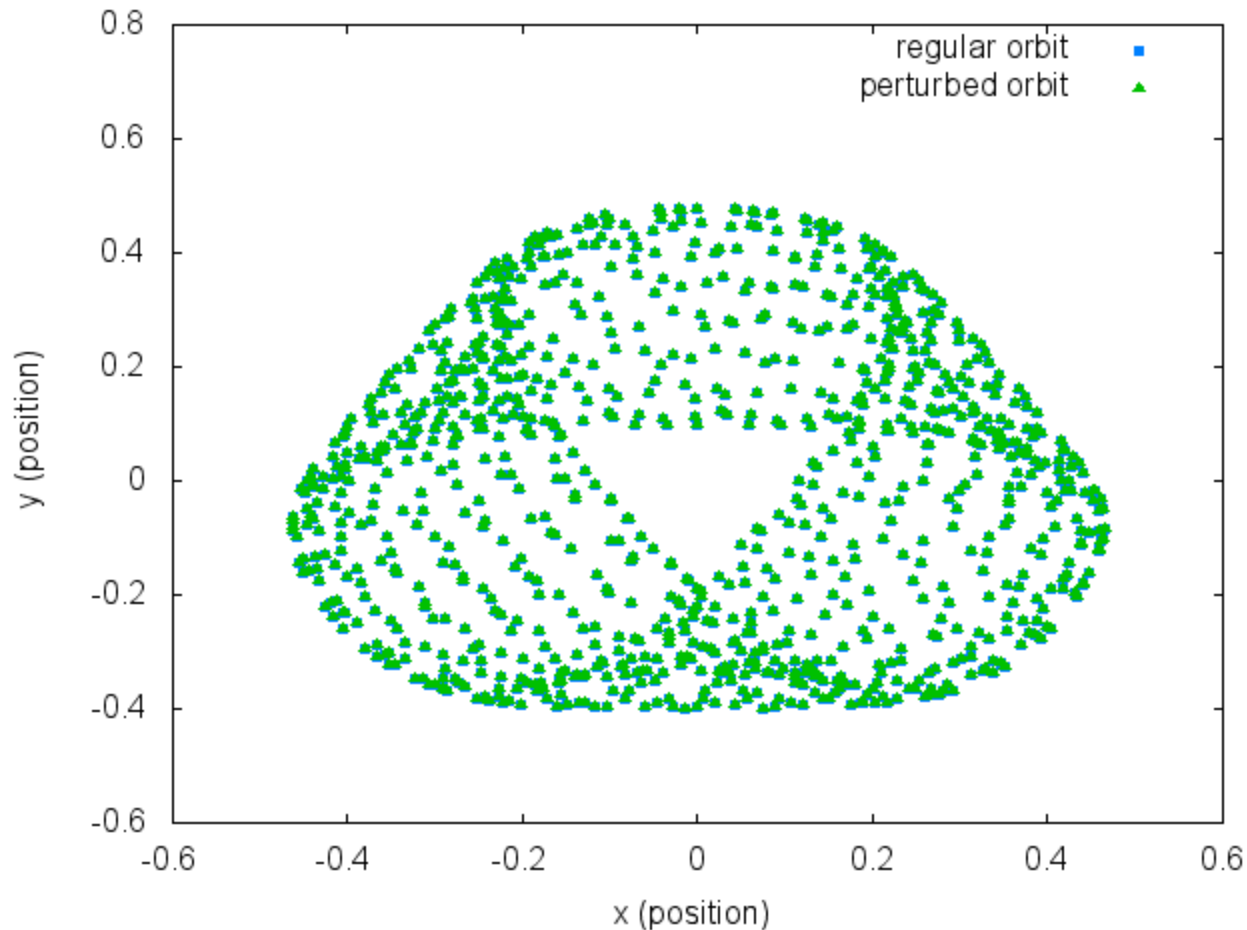
Regular orbit



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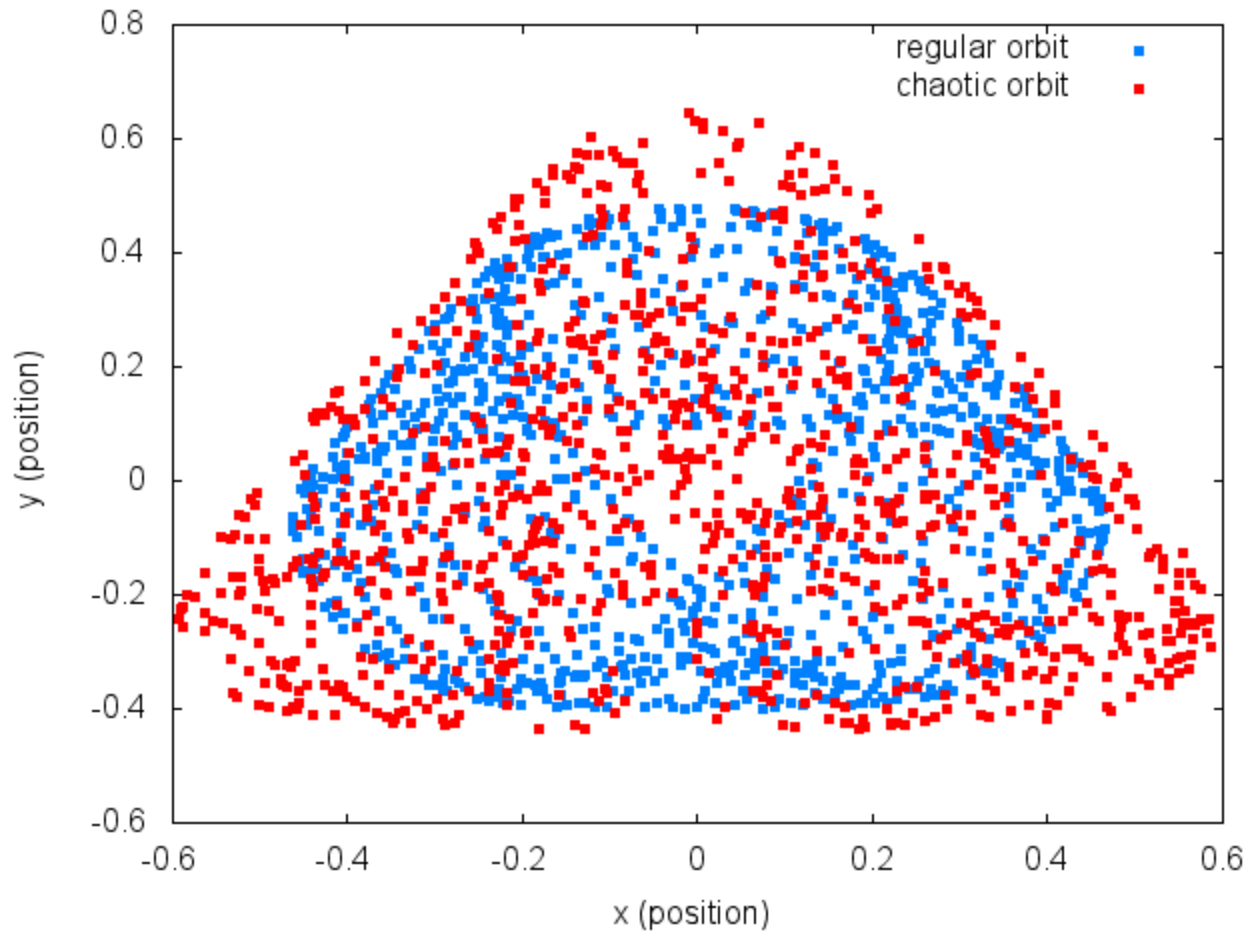
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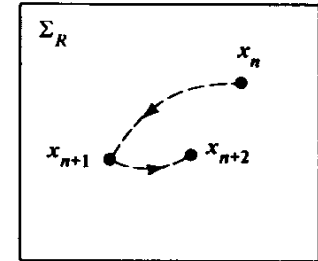
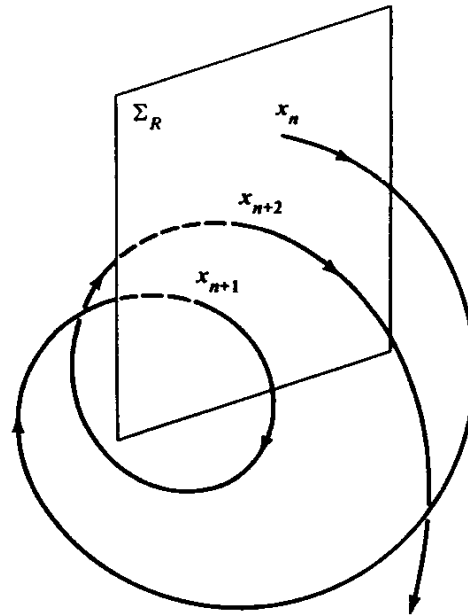
Regular vs Chaotic orbits



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Poincaré Surface of Section (PSS)

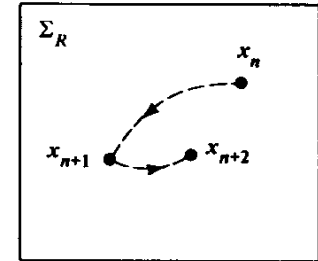
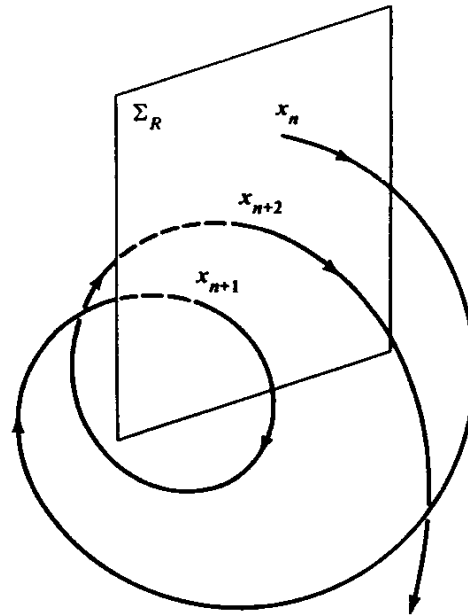
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Lieberman & Lichtenberg, 1992, *Regular and Chaotic Dynamics*, Springer.

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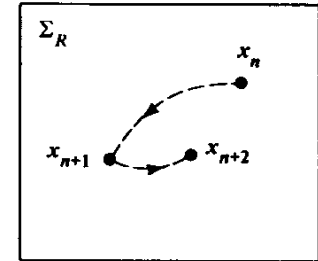
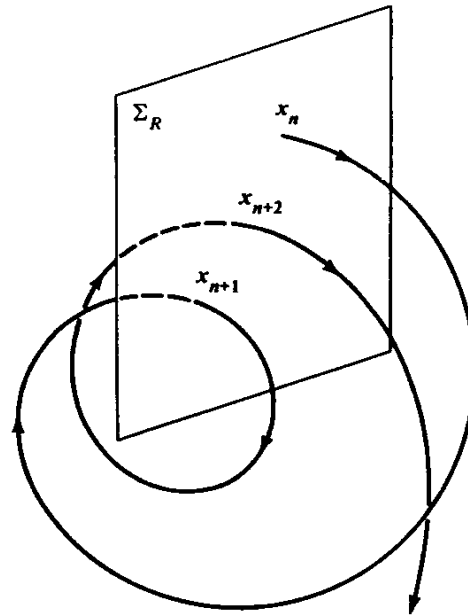


Lieberman & Lichtenberg, 1992, *Regular and Chaotic Dynamics*, Springer.

In general we can assume a PSS of the form **$q_{N+1} = \text{constant}$** . Then only variables $q_1, q_2, \dots, q_N, p_1, p_2, \dots, p_N$ are needed to describe the evolution of an orbit on the PSS, since p_{N+1} can be found from the Hamiltonian.

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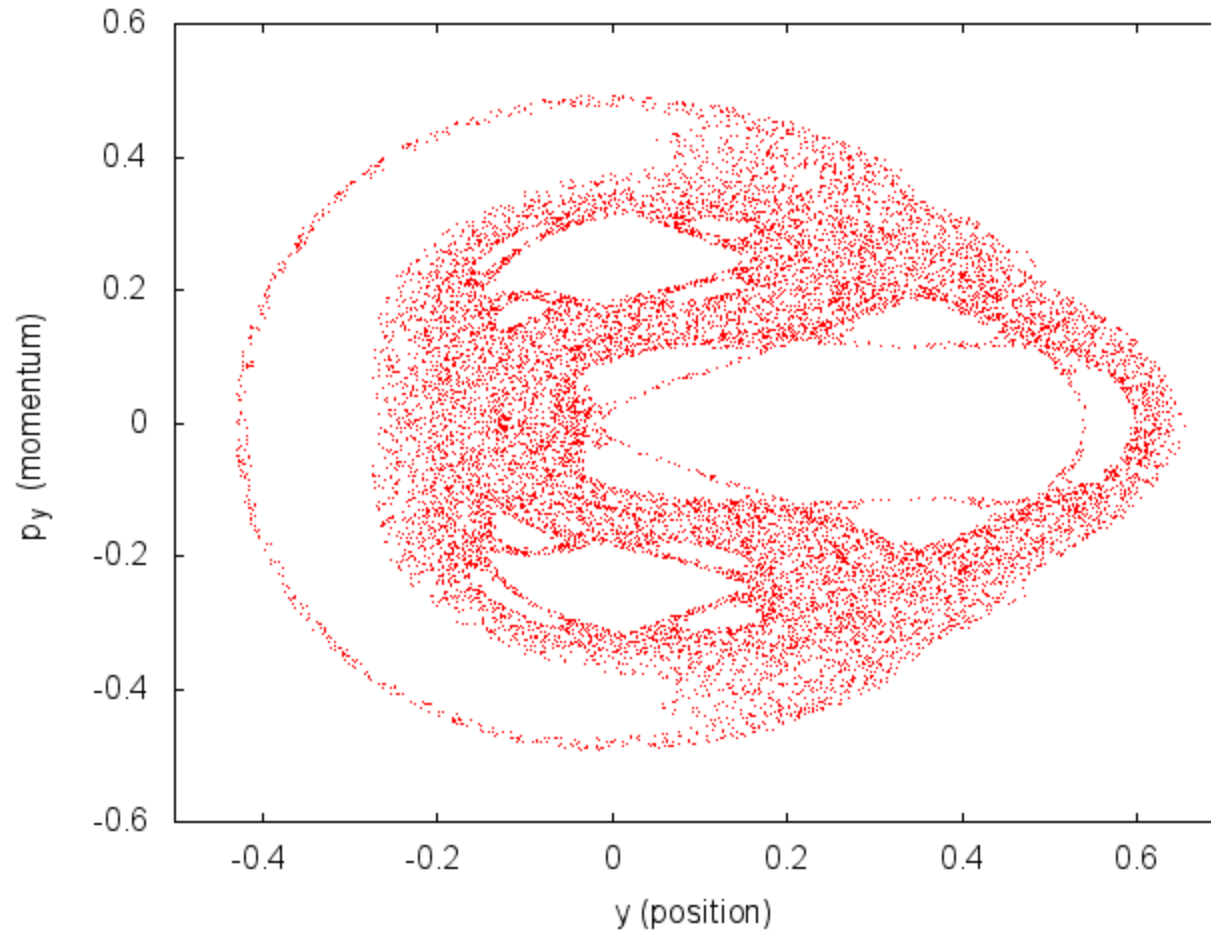


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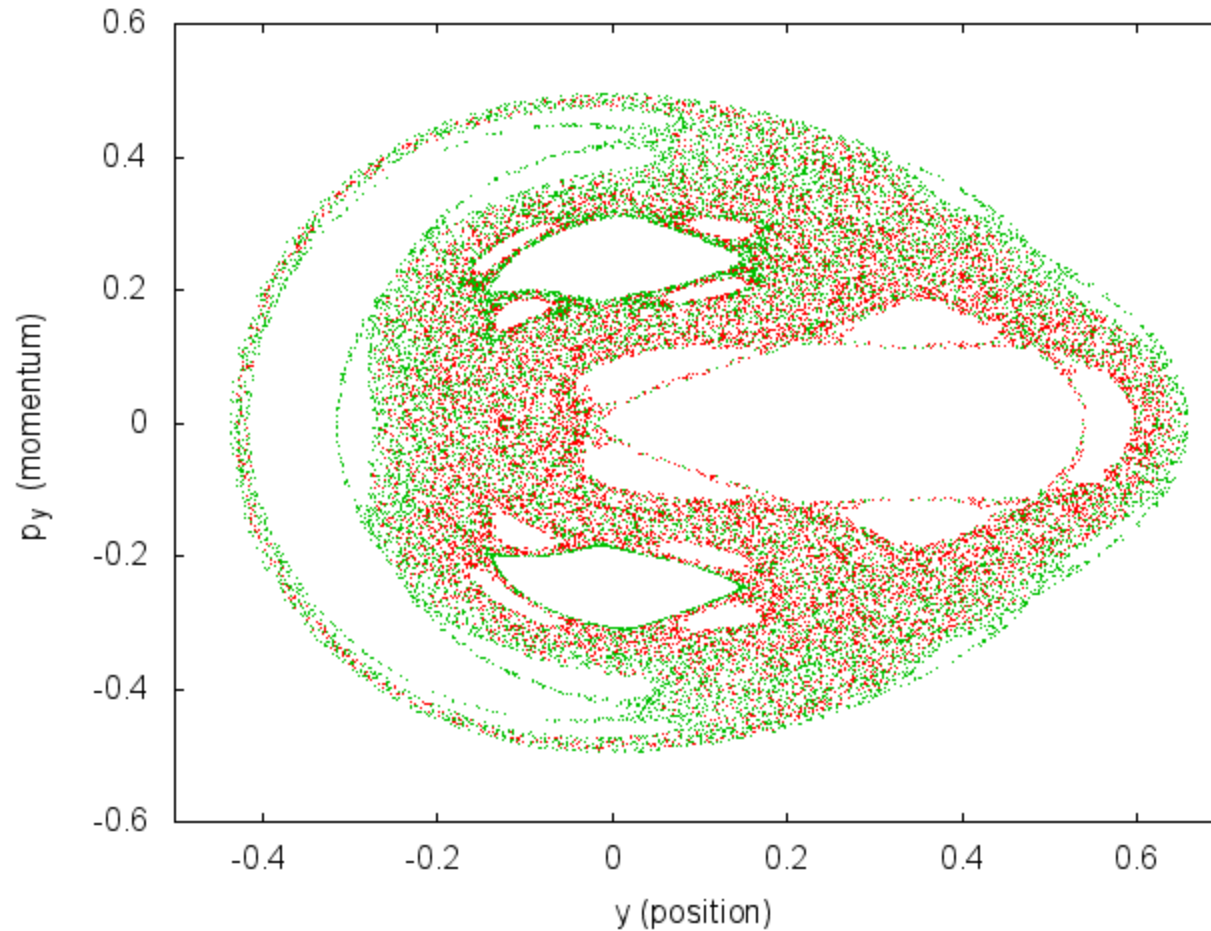
In this sense **an $N+1$ degree of freedom Hamiltonian system corresponds to a 2N-dimensional map.**

Hénon-Heiles system: PSS ($x=0$)



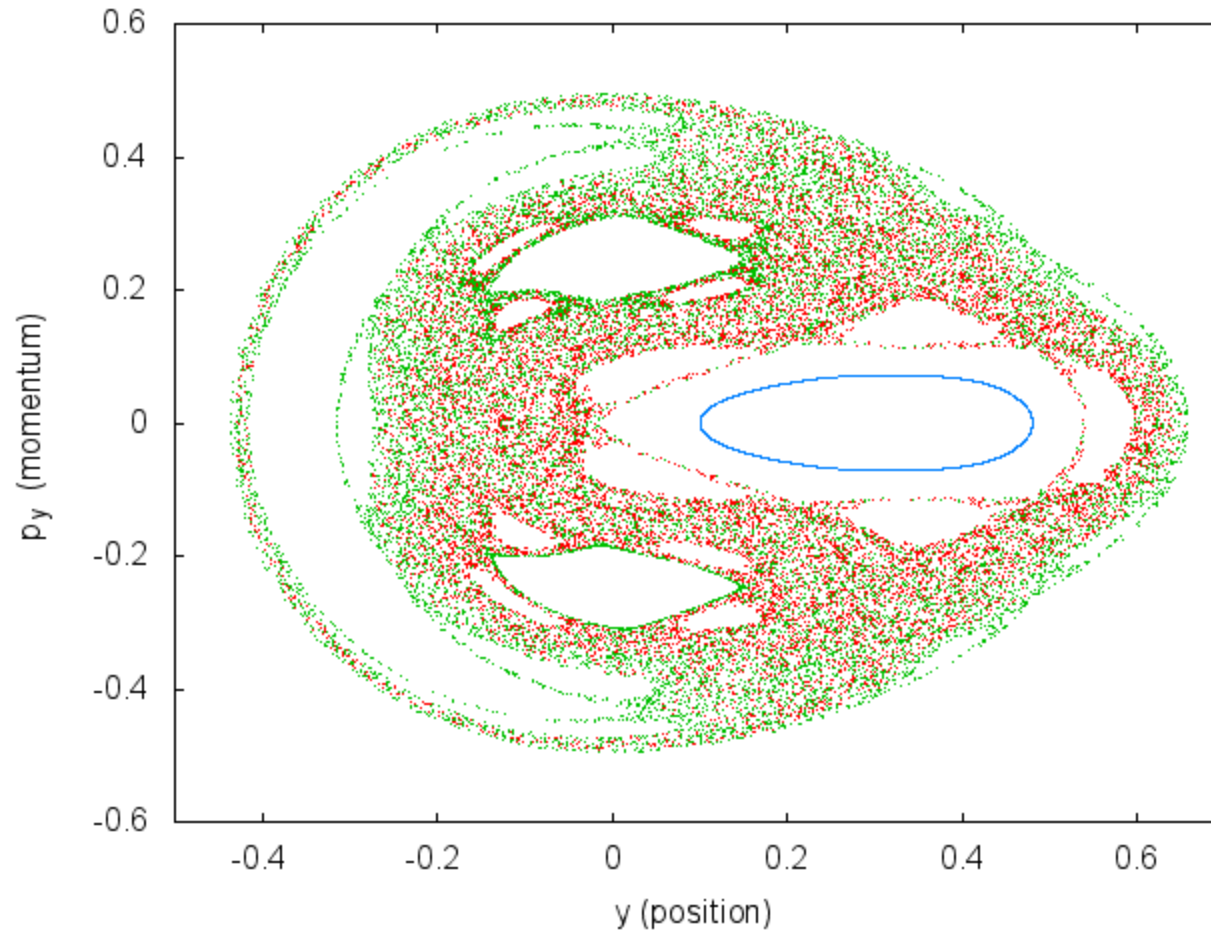
Chaotic orbit

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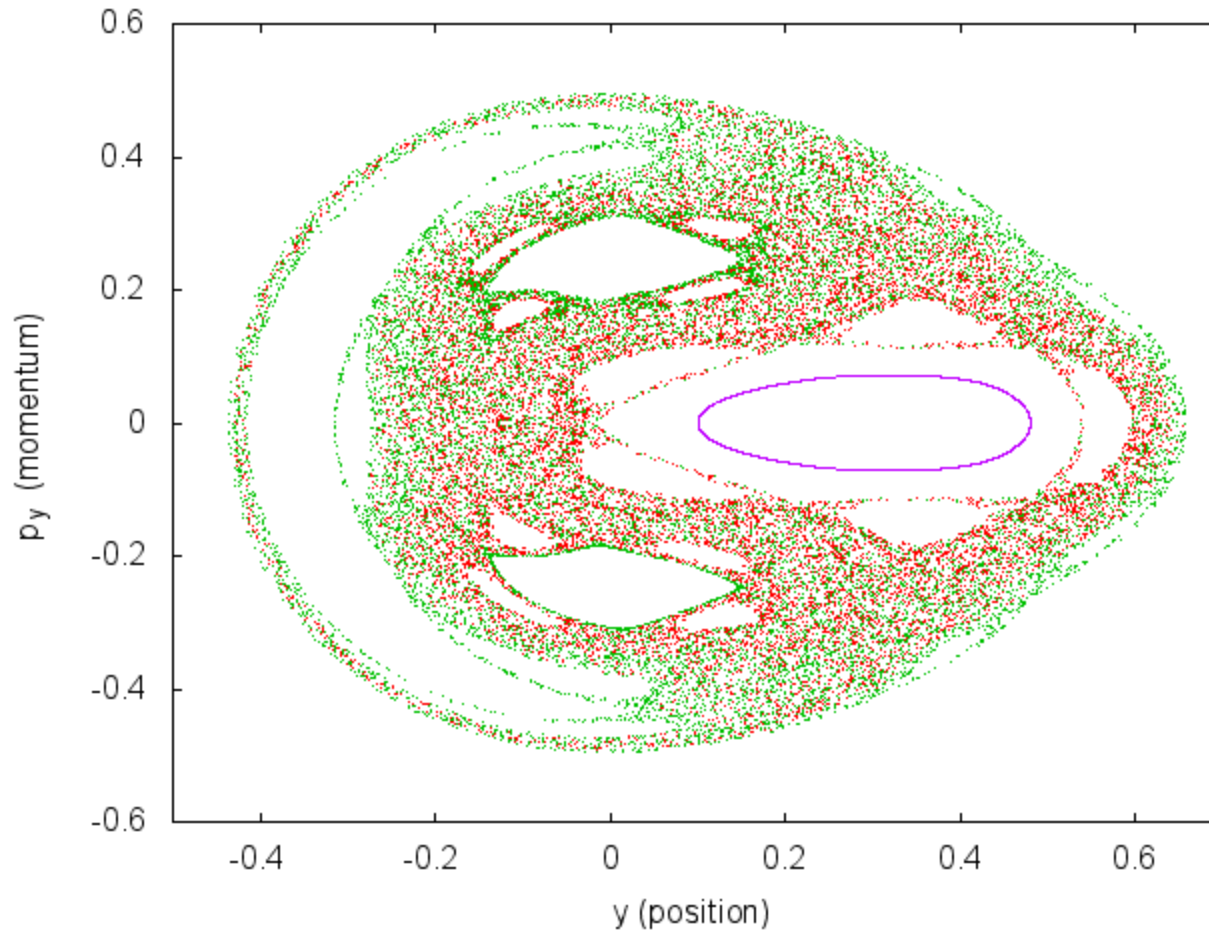
Chaotic orbit - **Perturbed chaotic orbit**

Hénon-Heiles system: PSS ($x=0$)



Chaotic orbit - **Perturbed chaotic orbit**
Regular orbit

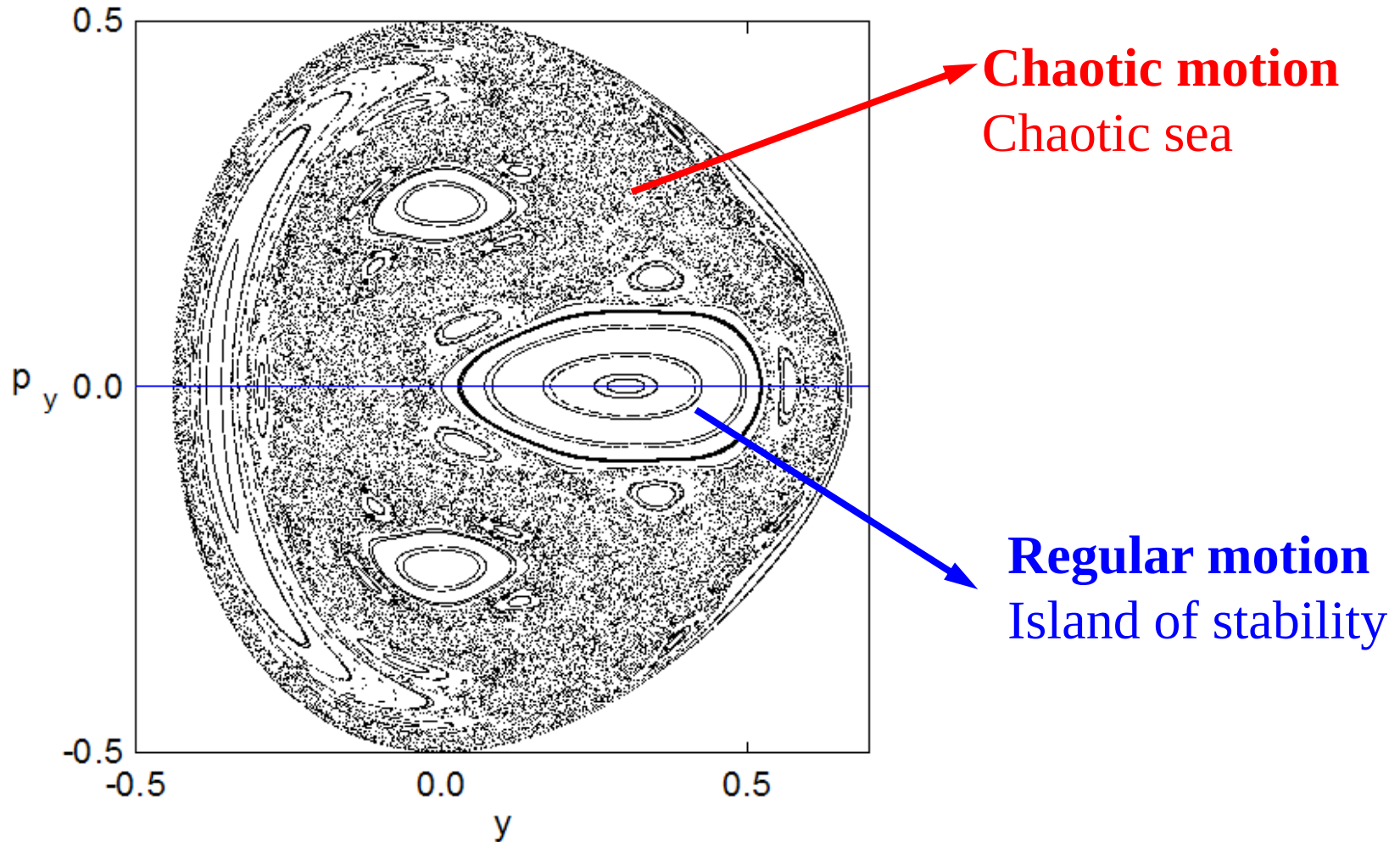
Hénon-Heiles system: PSS ($x=0$)



Chaotic orbit - **Perturbed chaotic orbit**

Regular orbit - **Perturbed regular orbit**

Hénon-Heiles system: PSS ($x=0$)



Variational Equations

We use the notation $\mathbf{x} = (q_1, q_2, \dots, q_N, p_1, p_2, \dots, p_N)^T$. The **deviation vector** from a given orbit is denoted by

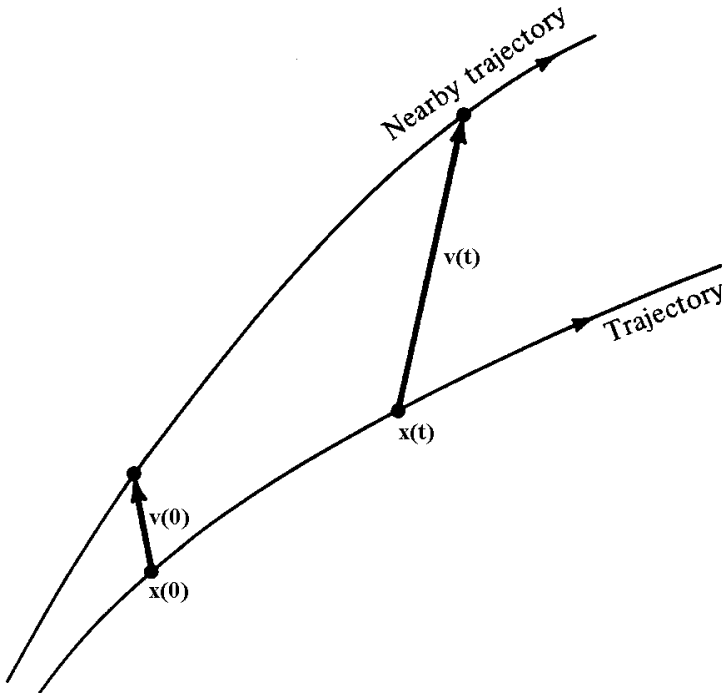
$$\mathbf{v} = (\delta x_1, \delta x_2, \dots, \delta x_n)^T, \text{ with } n=2N$$

The time evolution of $\mathbf{v}$ is given by the so-called **variational equations**:

$$\frac{d\mathbf{v}}{dt} = -\mathbf{J} \cdot \mathbf{P} \cdot \mathbf{v}$$

where

$$\mathbf{J} = \begin{pmatrix} \mathbf{0}_N & -\mathbf{I}_N \\ \mathbf{I}_N & \mathbf{0}_N \end{pmatrix}, \quad \mathbf{P}_{ij} = \frac{\partial^2 \mathbf{H}}{\partial \mathbf{x}_i \partial \mathbf{x}_j} \quad i, j = 1, 2, \dots, n$$



Example (Hénon-Heiles system)

$$H = \frac{1}{2}(p_x^2 + p_y^2) + \frac{1}{2}(x^2 + y^2) + x^2y - \frac{1}{3}y^3$$

Hamilton's equations of motion:

$$\frac{dp_i}{dt} = -\frac{\partial H}{\partial q_i}, \quad \frac{dq_i}{dt} = \frac{\partial H}{\partial p_i} \Rightarrow \begin{cases} \dot{x} = p_x \\ \dot{y} = p_y \\ \dot{p}_x = -x - 2xy \\ \dot{p}_y = -y - x^2 + y^2 \end{cases}$$

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In order to get the variational equations we **linearize** the above equations by substituting x, y, p_x, p_y with $x+v_1, y+v_2, p_x+v_3, p_y+v_4$ where $v=(v_1, v_2, v_3, v_4)$ is the deviation vector. So we get:

$$\begin{aligned} \dot{p}_x + \dot{v}_3 &= -x - v_1 - 2(x + v_1)(y + v_2) \Rightarrow \\ \dot{p}_x + \dot{v}_3 &= -x - v_1 - 2xy - 2xv_2 - 2yv_1 - 2v_1v_2 \Rightarrow \end{aligned}$$

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$$\dot{v}_4 = -v_2 - 2xv_1 + 2yv_2$$

$$\dot{p}_y = -y - x^2 + y^2$$

Complete set of equations

Lyapunov Exponents

Roughly speaking, the Lyapunov exponents of a given orbit characterize the **mean exponential rate of divergence** of trajectories surrounding it.

Consider an orbit in the $2N$ -dimensional phase space with **initial condition $\mathbf{x}(0)$** and an **initial deviation vector from it $\mathbf{v}(0)$** . Then the mean exponential rate of divergence is:

$$\sigma(\mathbf{x}(0), \mathbf{v}(0)) = \lim_{t \rightarrow \infty} \frac{1}{t} \ln \frac{\|\mathbf{v}(t)\|}{\|\mathbf{v}(0)\|}$$

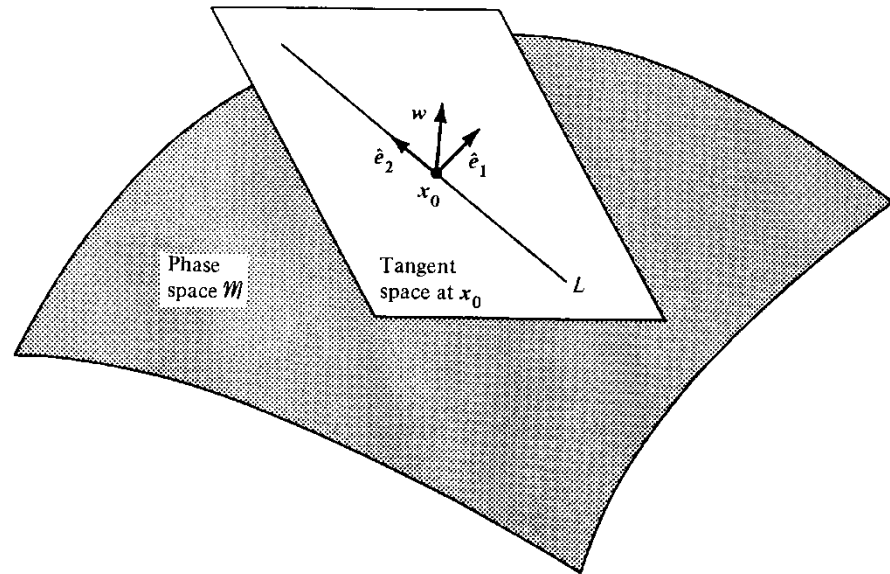
We commonly use the Euclidian norm and set $d(0) = \|\mathbf{v}(0)\| = 1$

Lyapunov Exponents

There exists an **M-dimensional basis** $\{\hat{e}_i\}$ of v such that for any v , σ takes one of the M (possibly nondistinct) values

$$\sigma_i(x(0)) = \sigma(x(0), \hat{e}_i)$$

which are the **Lyapunov exponents**.



Benettin & Galgani, 1979, in Laval and Gressillon (eds.), op cit, 93

In autonomous Hamiltonian systems the M exponents are ordered in **pairs of opposite sign numbers and two of them are 0**.

Computation of the Maximum Lyapunov Exponent

Due to the exponential growth of $v(t)$ (and of $d(t)=||v(t)||$) we **renormalize $v(t)$** from time to time.

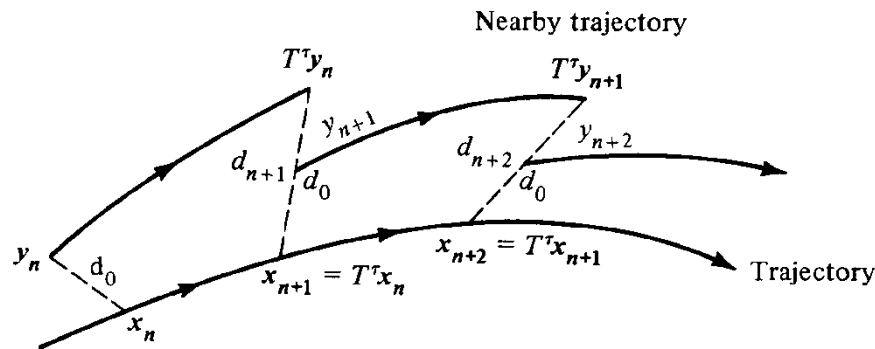


Figure 5.6. Numerical calculation of the maximal Liapunov characteristic exponent. Here $y = x + v$ and τ is a finite interval of time (after Benettin *et al.*, 1976).

Then the Maximum Lyapunov exponent is computed as

$$\sigma_1 = \lim_{n \rightarrow \infty} \frac{1}{n\tau} \sum_{i=1}^n \ln d_i$$

Maximum Lyapunov Exponent

$\sigma_1=0 \rightarrow$ Regular motion
 $\sigma_1 \neq 0 \rightarrow$ Chaotic motion

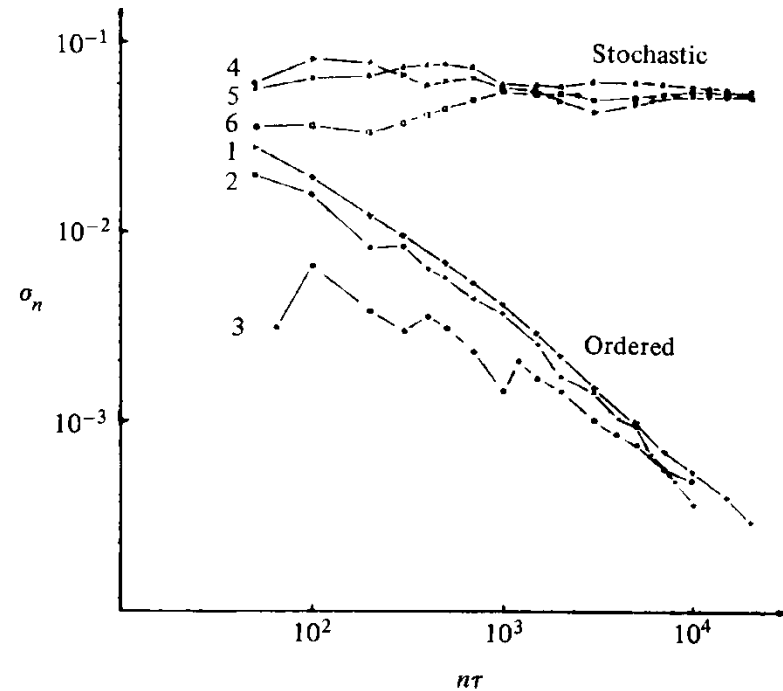
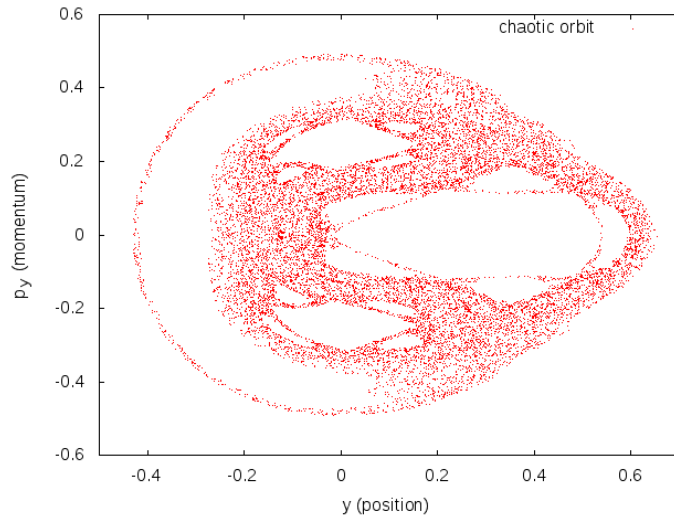


Figure 5.7. Behavior of σ_n at the intermediate energy $E = 0.125$ for initial points taken in the ordered (curves 1–3) or stochastic (curves 4–6) regions (after Benettin *et al.*, 1976).

If we start with more than one linearly independent deviation vectors they will **align to the direction defined by the largest Lyapunov exponent** for chaotic orbits.

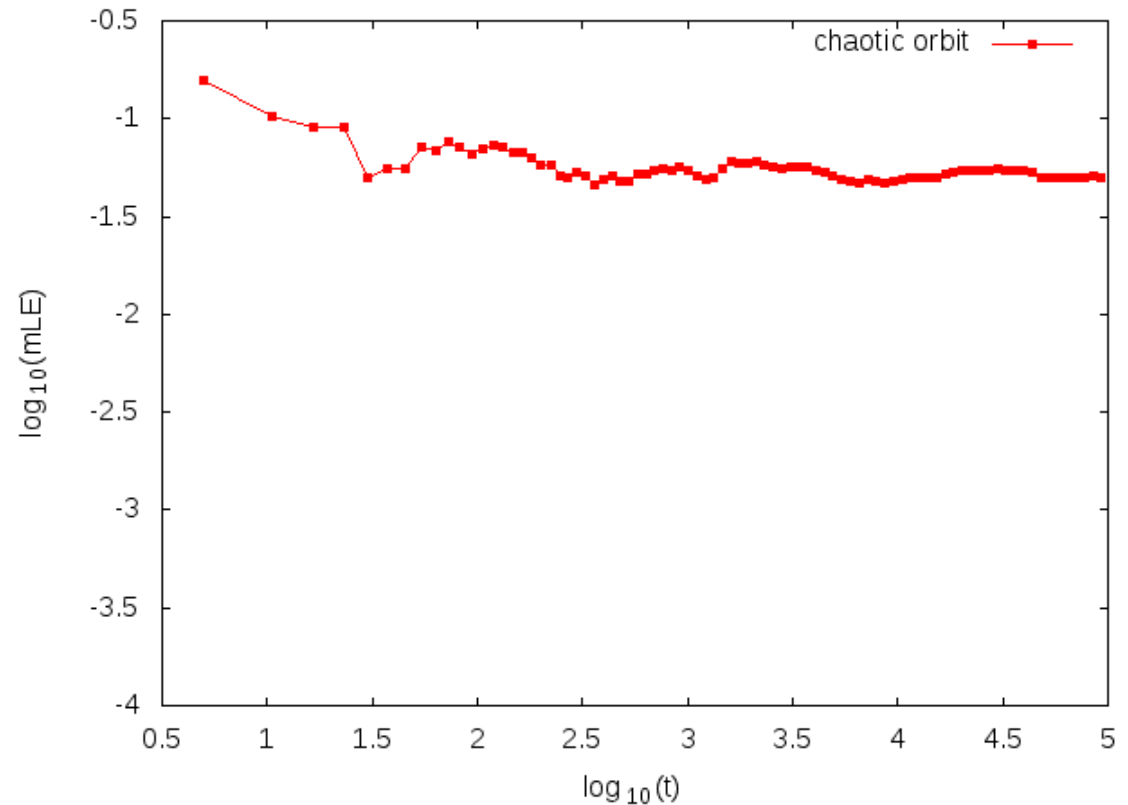
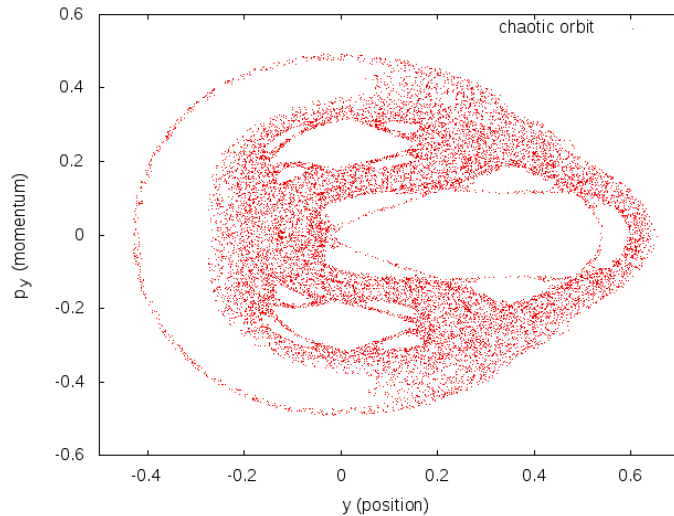
Maximum Lyapunov Exponent

Hénon-Heiles system: **Chaotic orbit**



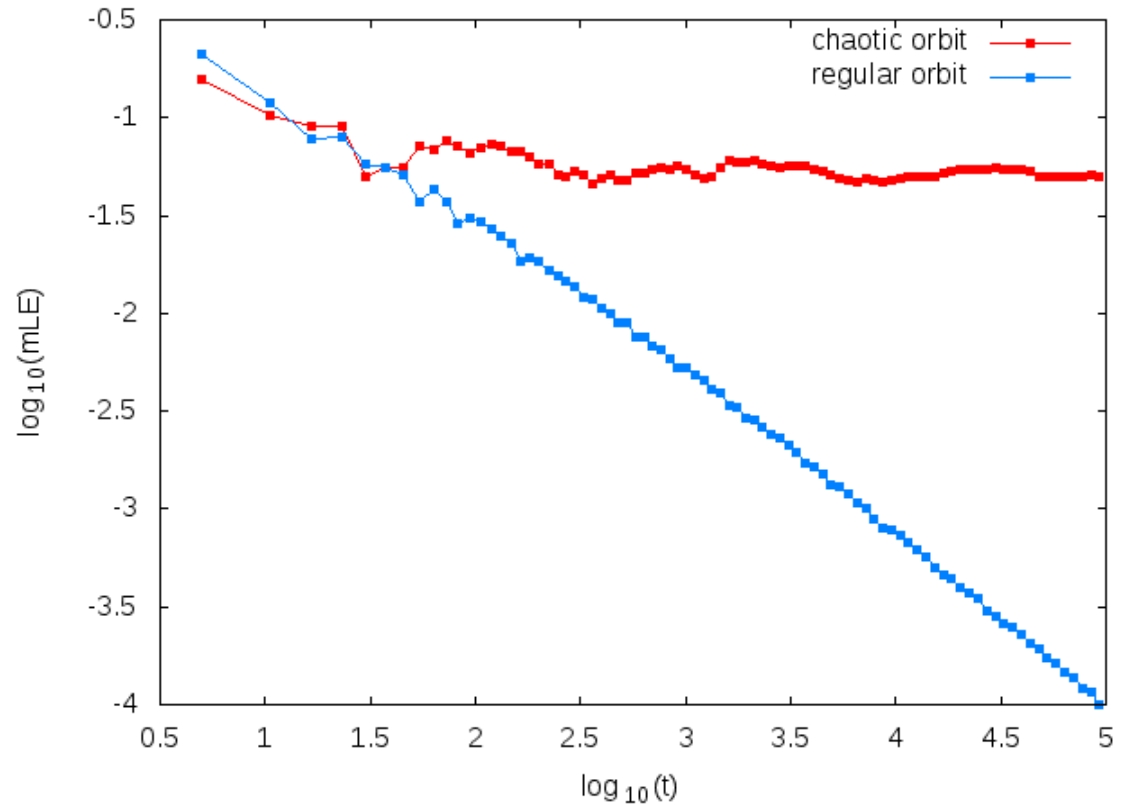
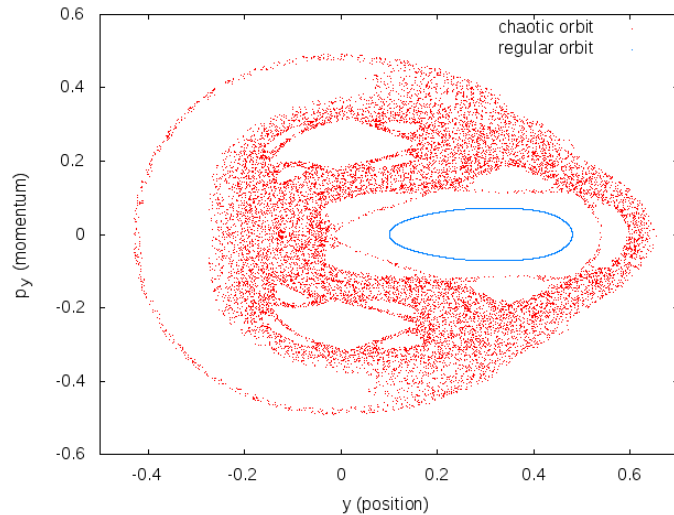
Maximum Lyapunov Exponent

Hénon-Heiles system: **Chaotic orbit**



Maximum Lyapunov Exponent

Hénon-Heiles system: **Chaotic orbit** and **Regular orbit**



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