

Nonlinear physics: 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:

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

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

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

is governed by the **Hamilton's equations of motion**

$$\frac{dp_i}{dt} = -\frac{\partial H}{\partial q_i}, \quad \frac{dq_i}{dt} = \frac{\partial H}{\partial p_i}$$

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

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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

Definition [Devaney (1989)]

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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$f : V \rightarrow V$ has *sensitive dependence on initial conditions* if there exists $\delta > 0$ such that, for any $x \in V$ and any neighborhood \mathcal{A} of x , there exist $y \in \mathcal{A}$ and $n \geq 0$, such that $|f^n(x) - f^n(y)| > \delta$, where f^n denotes n successive applications of f .

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

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

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f : $V \rightarrow V$ is said to be *topologically transitive* if for any pair of open sets $U, W \subset V$ there exists $n > 0$ such that $\mathbf{f}^n(U) \cap W \neq \emptyset$.

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

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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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$$t= 100 \quad x= 0.132995718333307644 \quad 0.132995718337263064$$

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$t=100$	$x=0.132995718333307644$	0.132995718337263064
$t=5000$	$x=0.376999283889102310$	0.376999283870156576

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t= 100	x= 0.090272817735167835	0.090272821355768668
t= 200	x= 0.295031687482249283	0.295031884858625637

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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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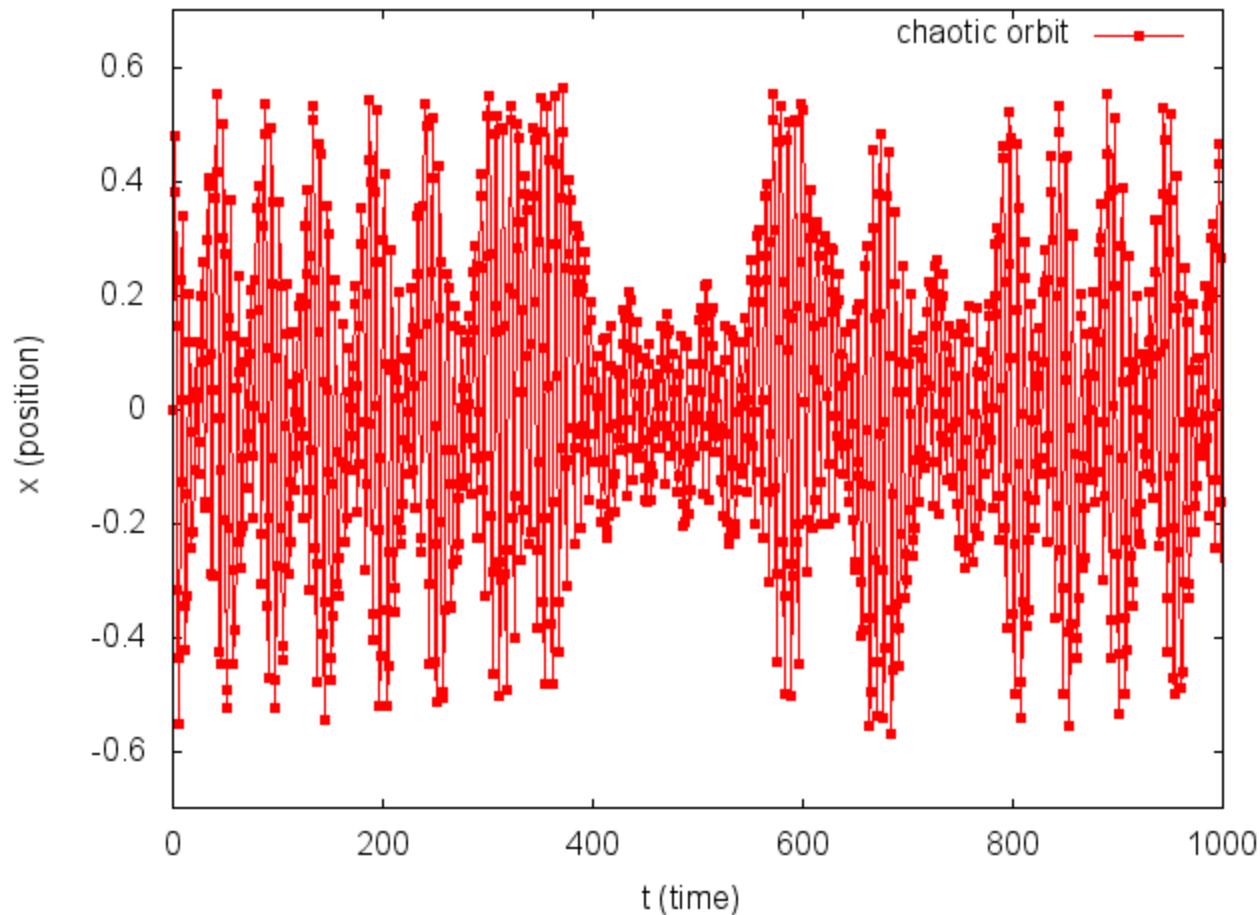
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t= 500	x= 0.078357719290523528	-0.270811022674341095

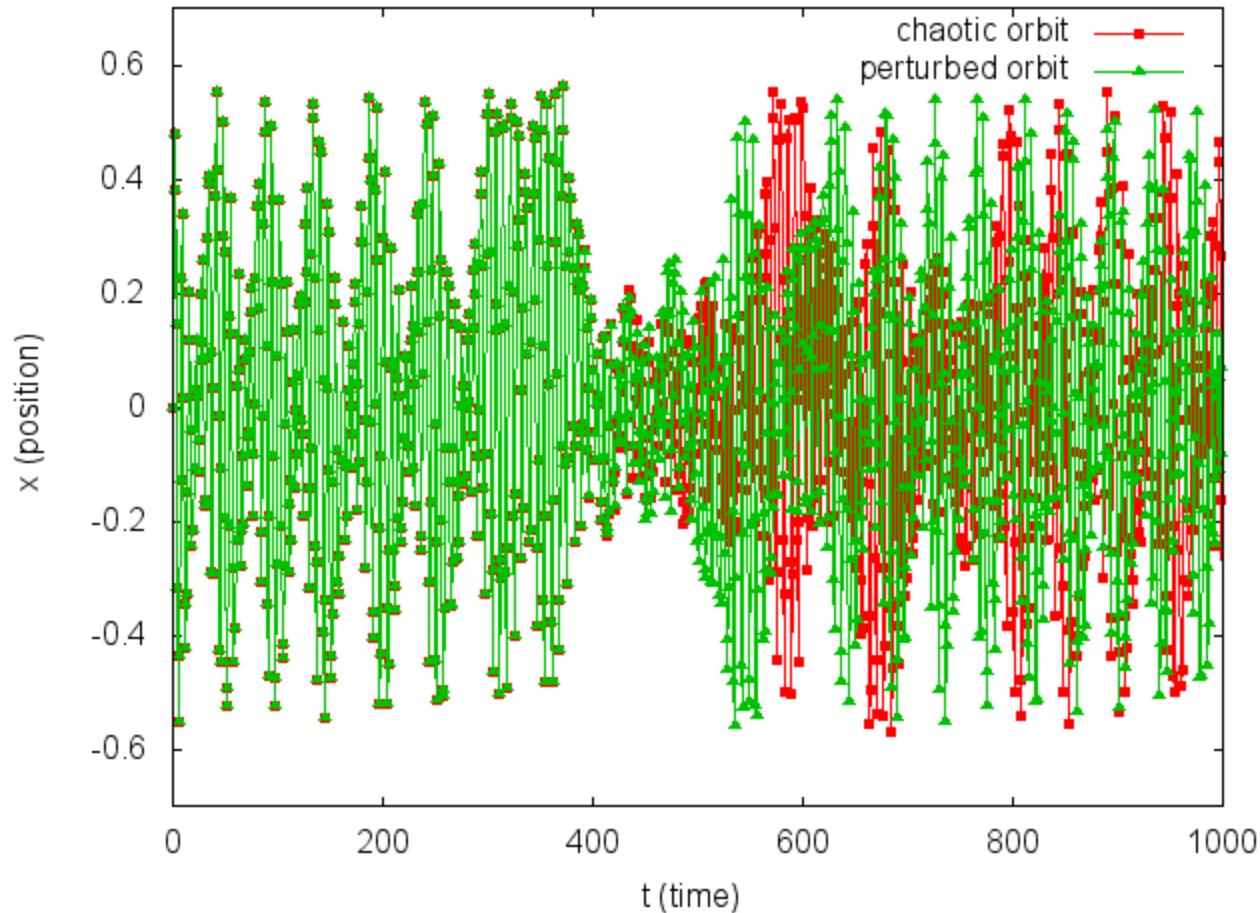
Regular vs Chaotic orbits

Chaotic orbit



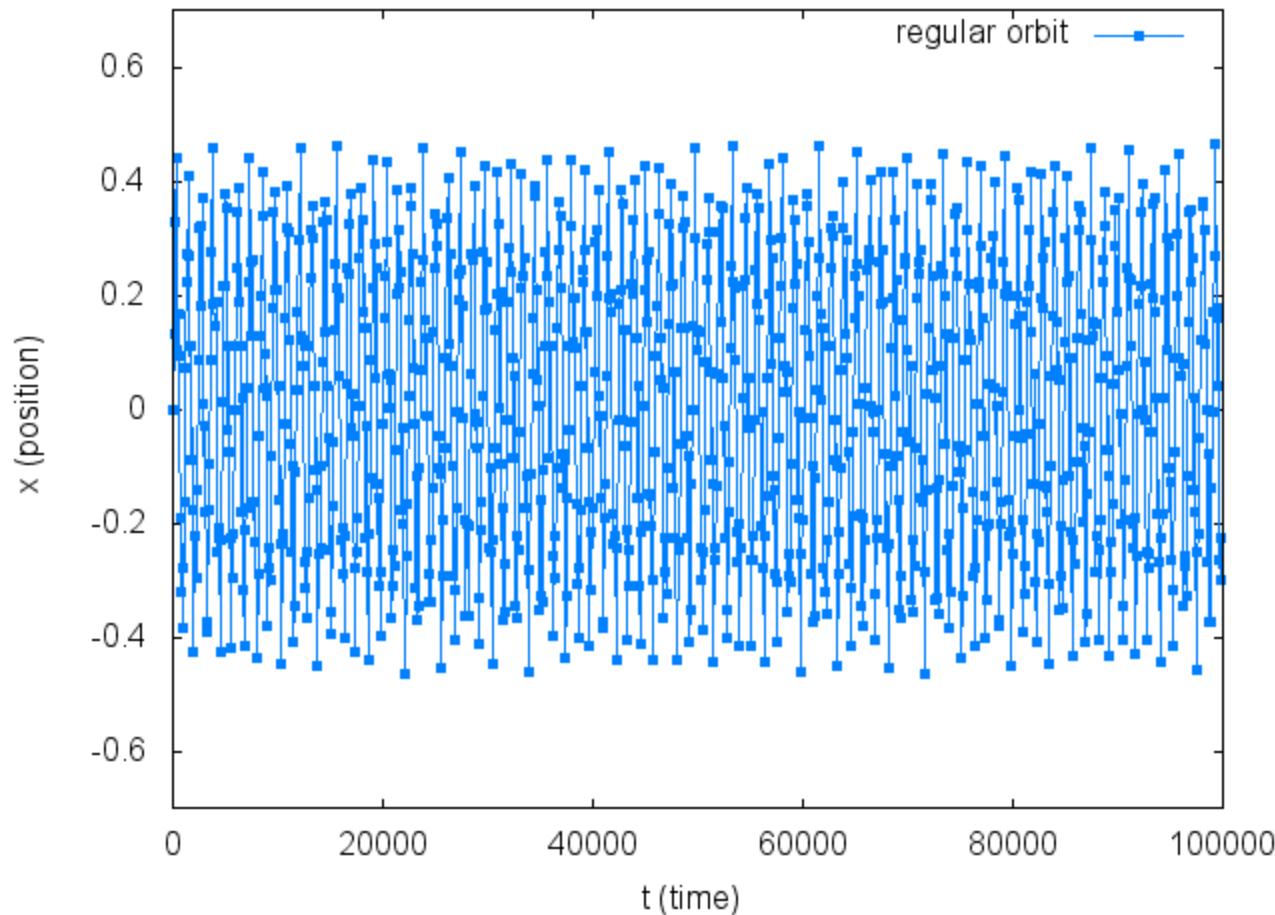
Regular vs Chaotic orbits

Chaotic orbit and its perturbation



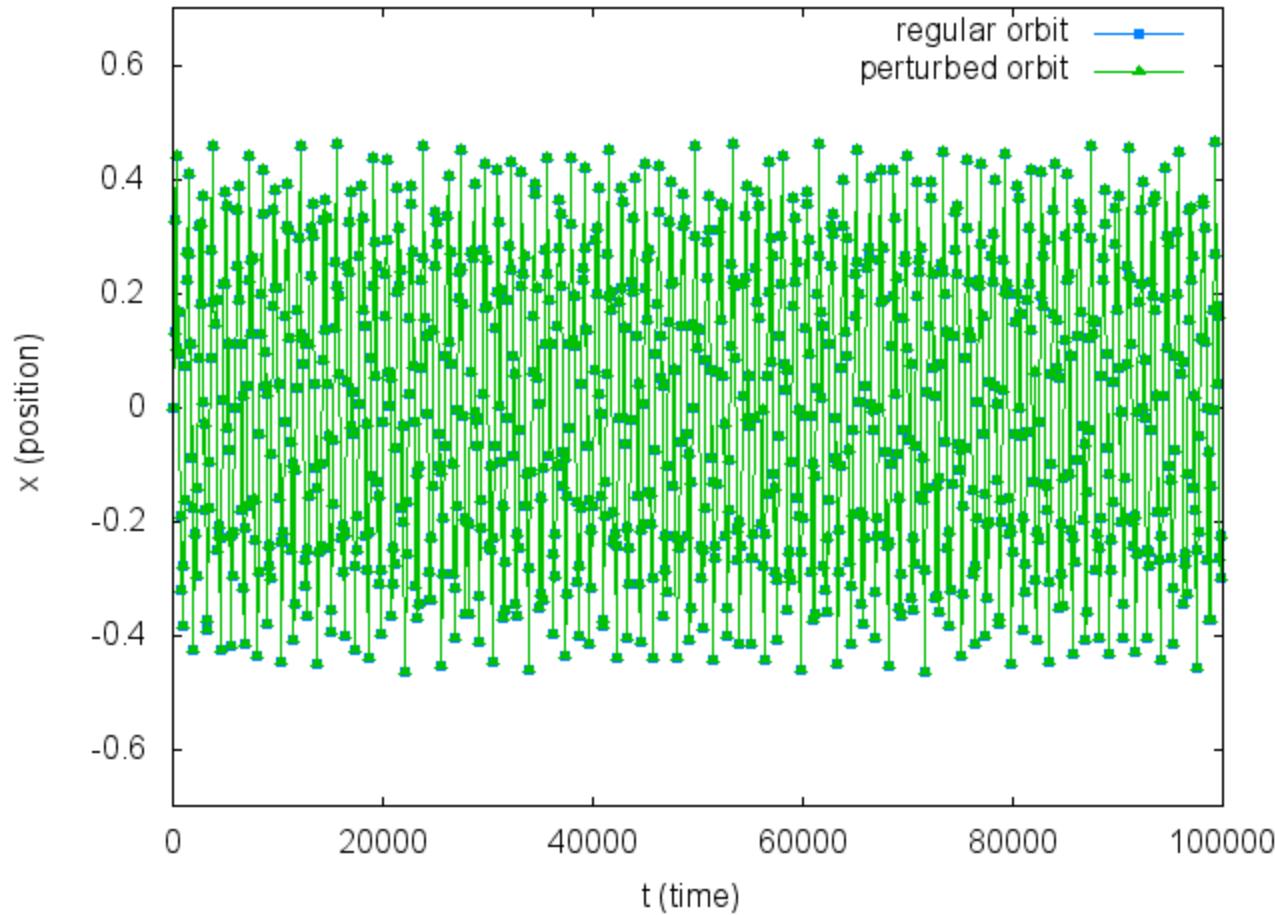
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Regular orbit



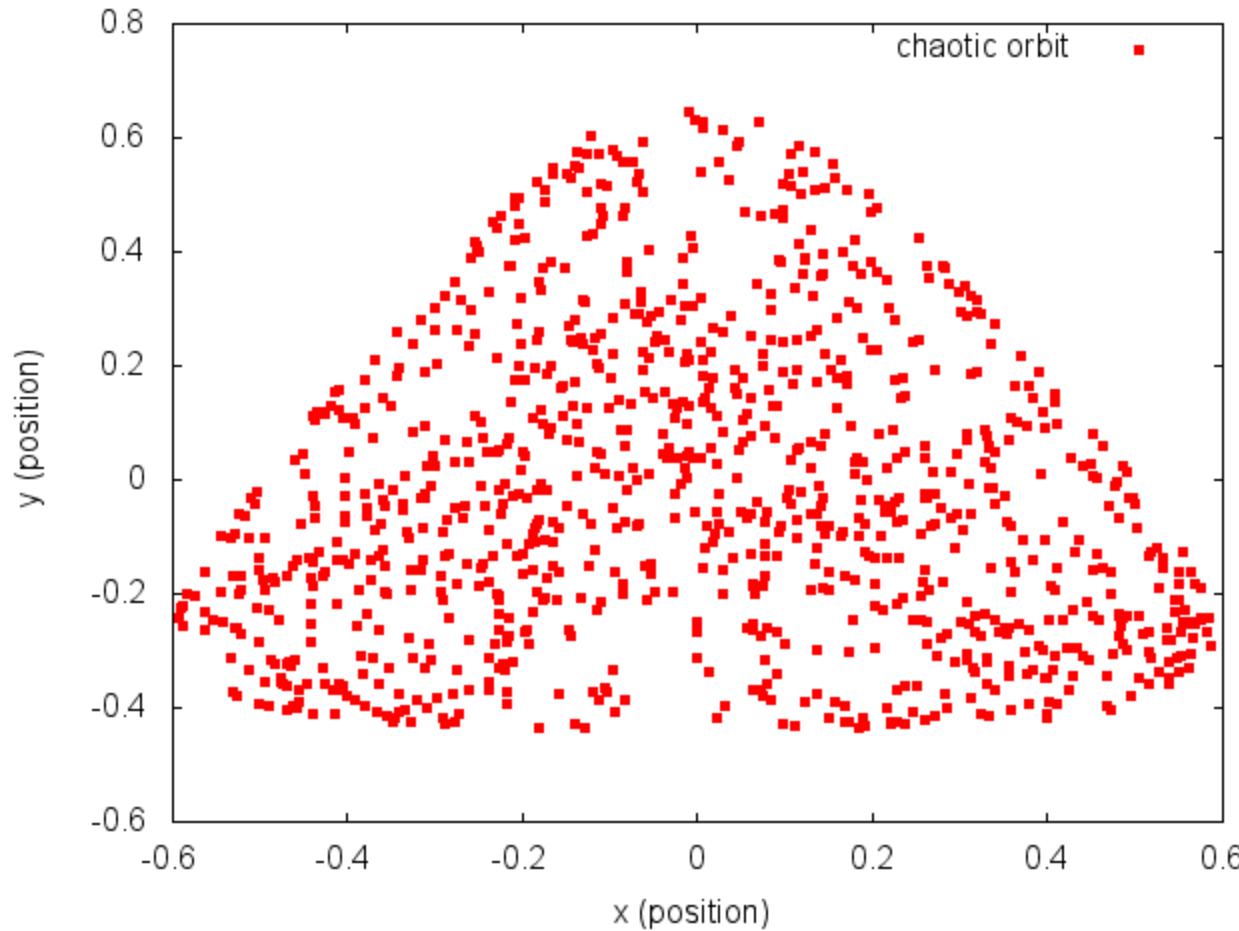
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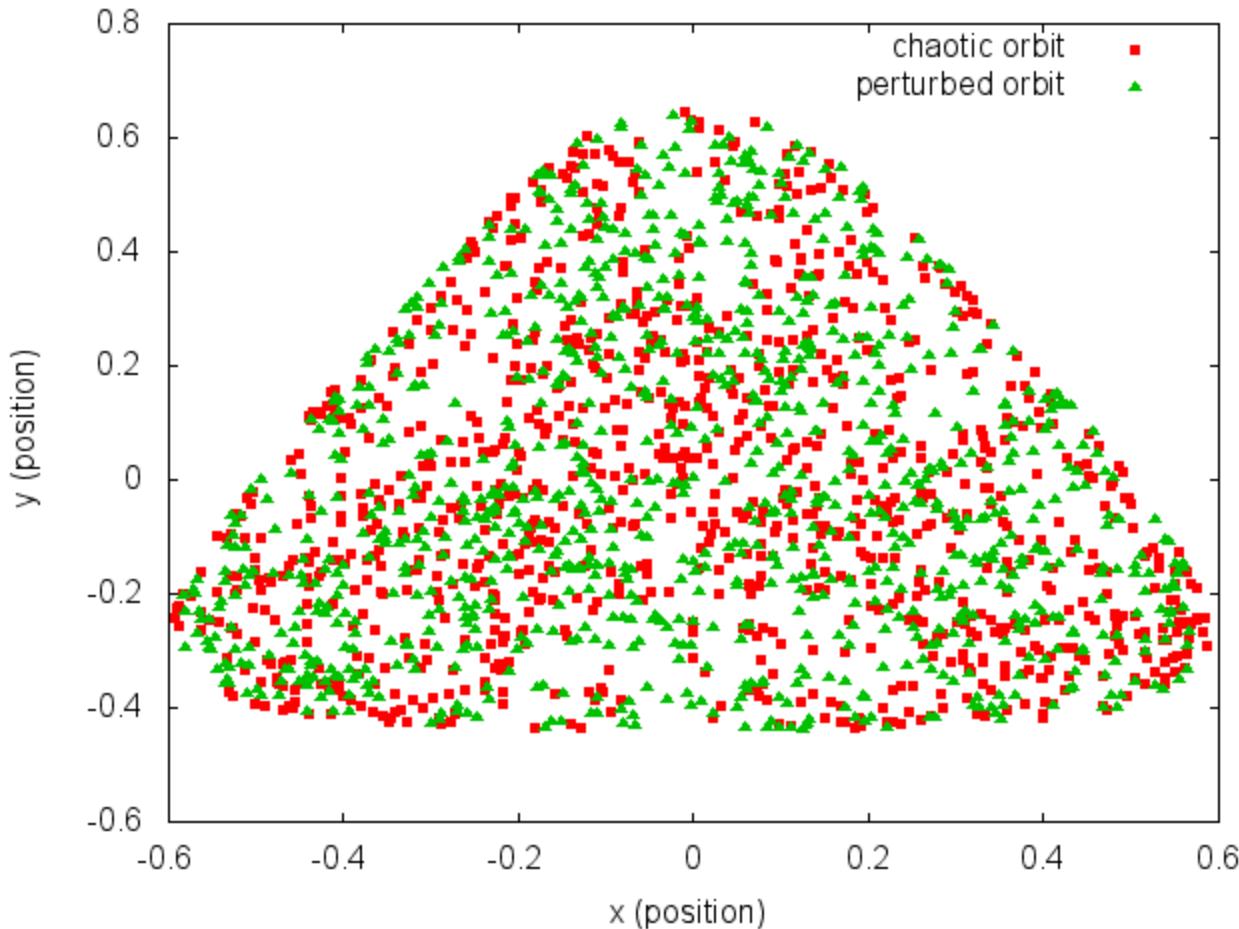
Chaotic orbit



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

Regular vs Chaotic orbits

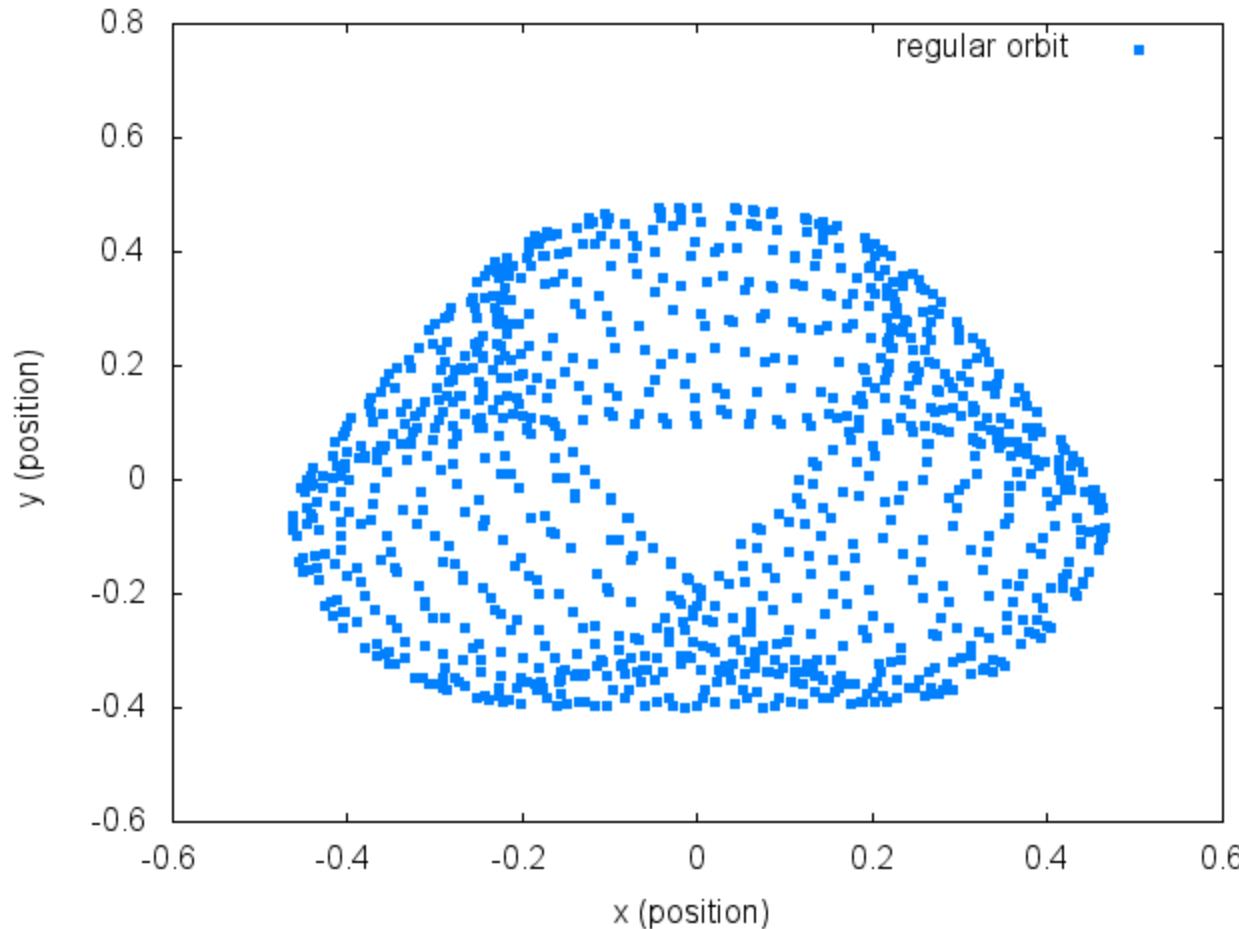
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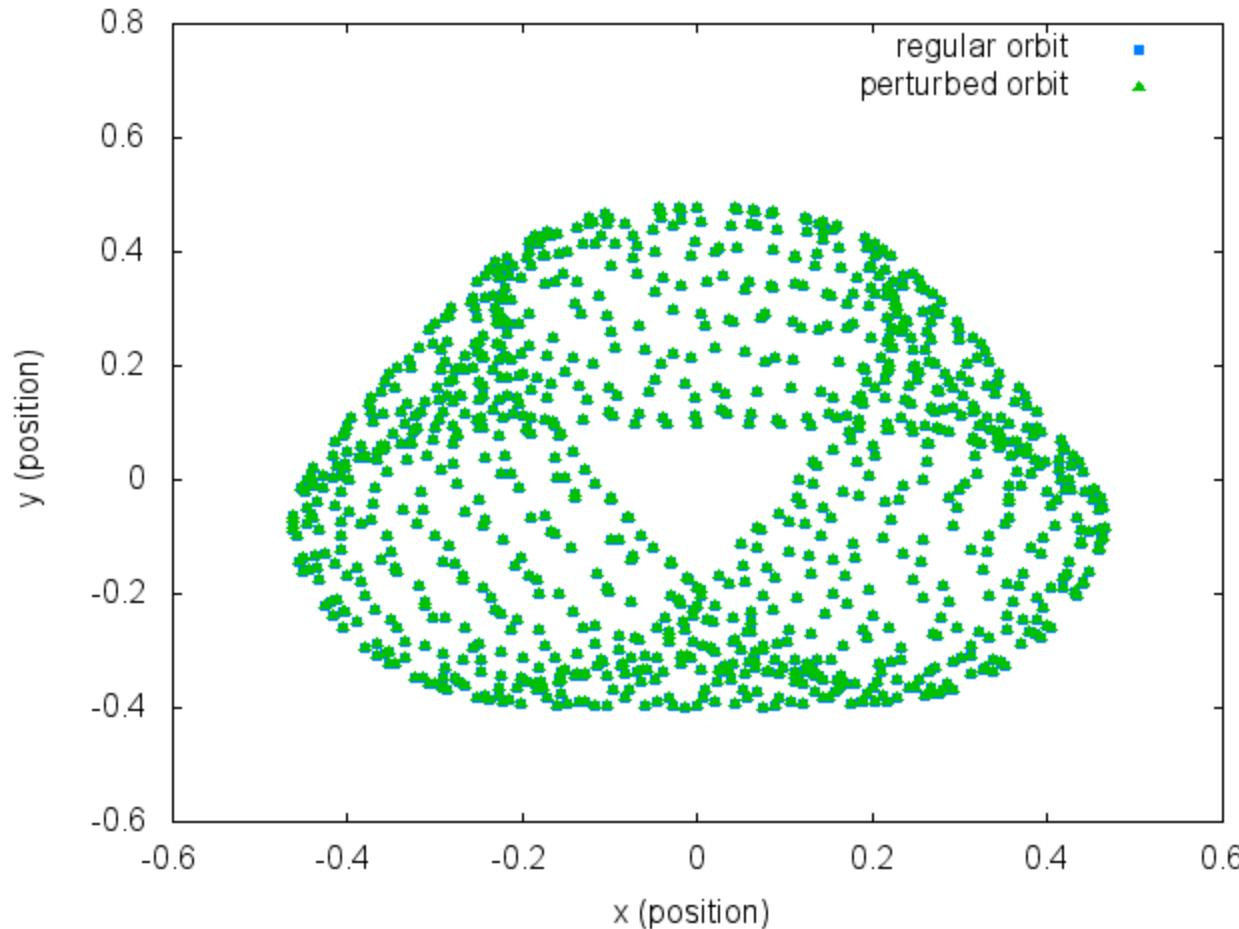
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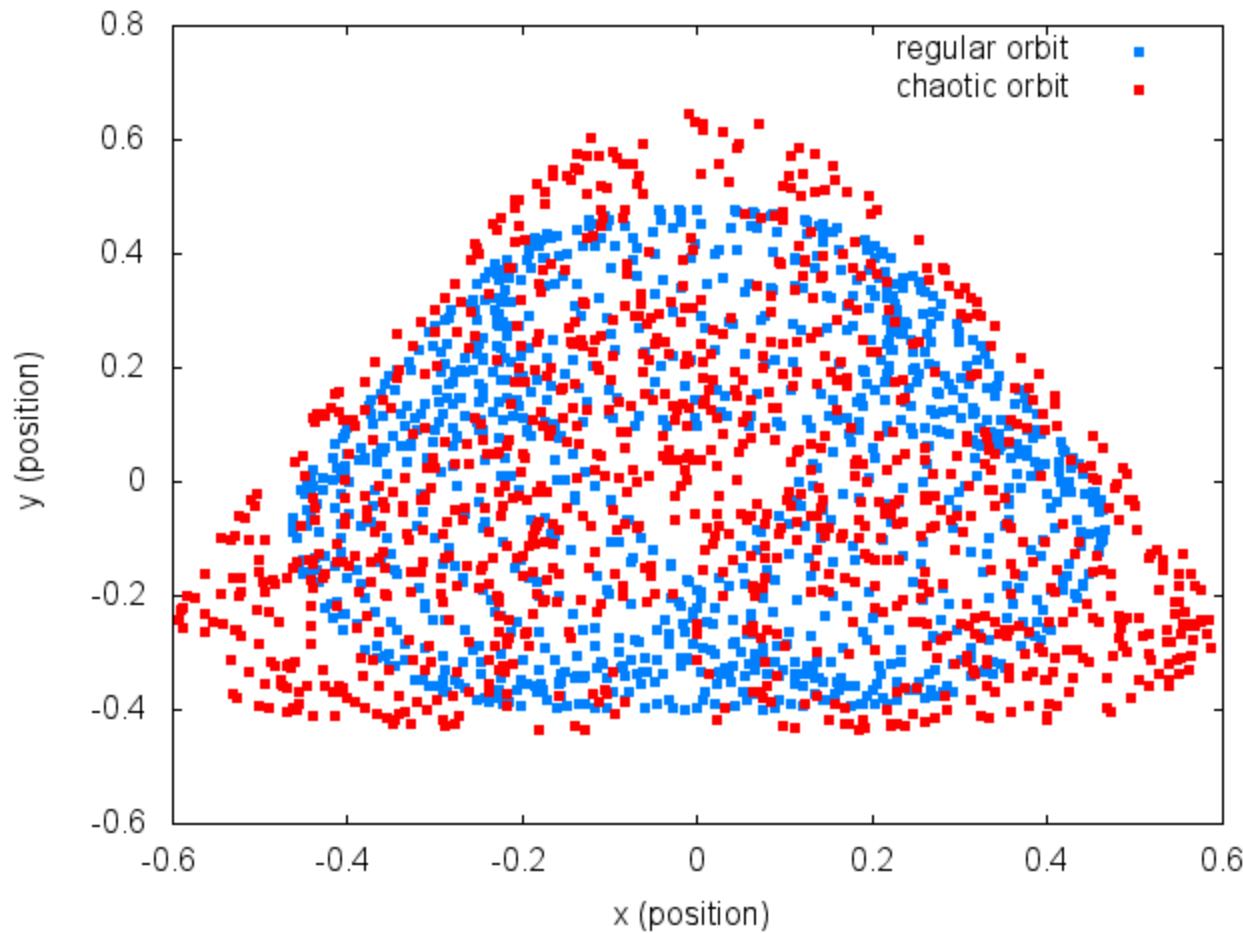
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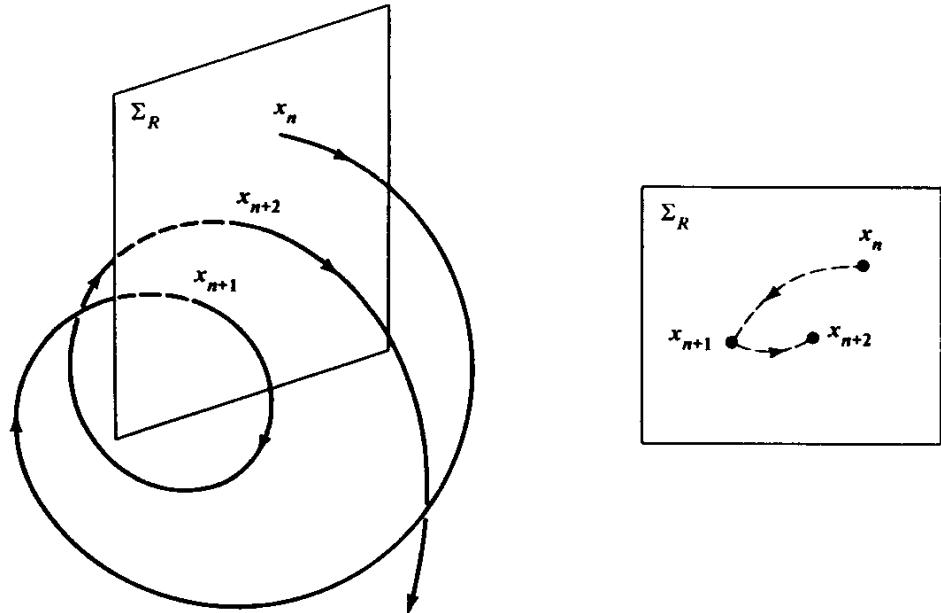
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Poincaré Surface of Section (PSS)

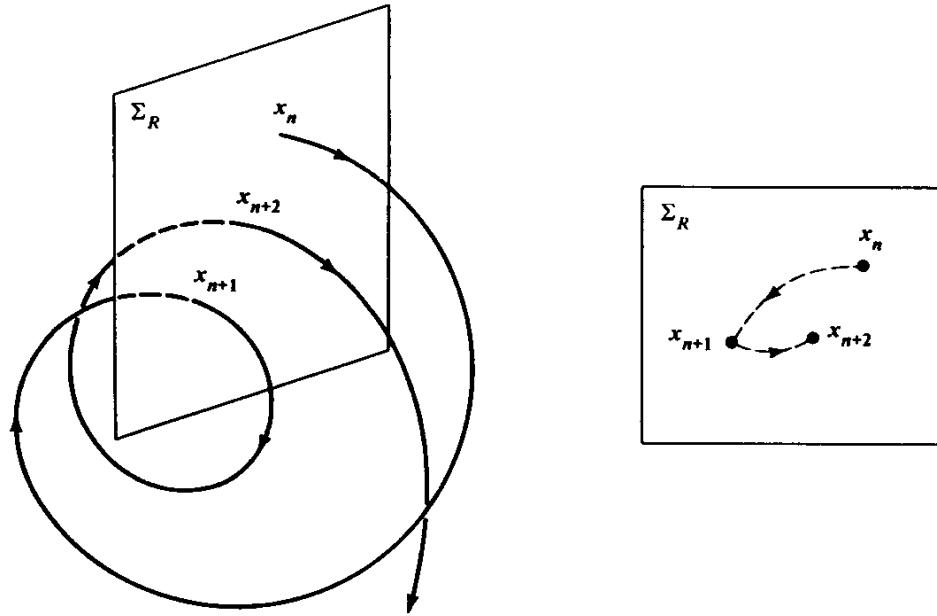
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Lieberman & Lichtenberg, 1992, *Regular and Chaotic Dynamics*, Springer.

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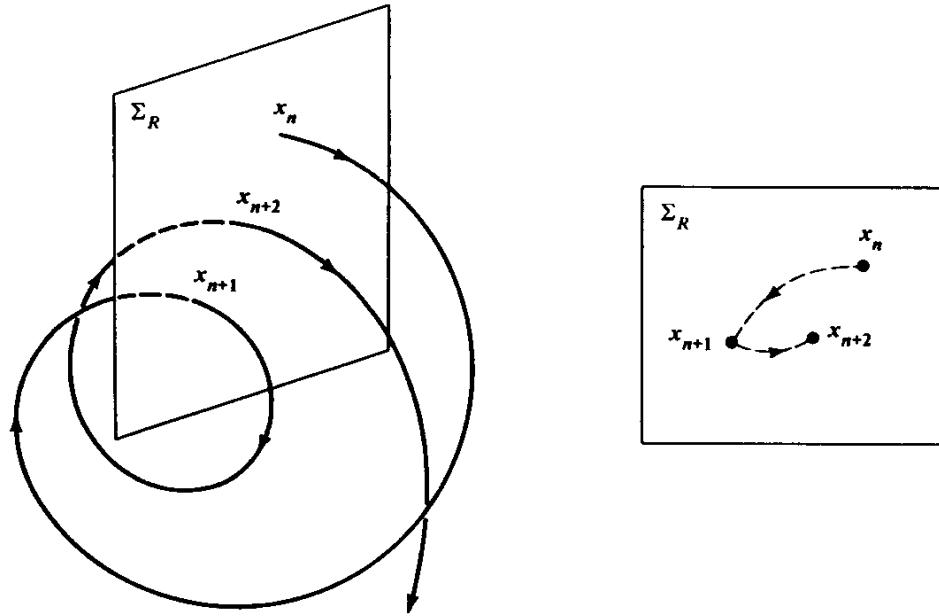


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In general we can assume a PSS of the form $q_{N+1} = \text{constant}$. Then only variables $q_1, q_2, \dots, q_N, p_1, p_2, \dots, p_N$ are needed to describe the evolution of an orbit on the PSS, since p_{N+1} can be found from the Hamiltonian.

Poincaré Surface of Section (PSS)

We can constrain the study of an $N+1$ degree of freedom Hamiltonian system to a **2N-dimensional subspace** of the general phase space.

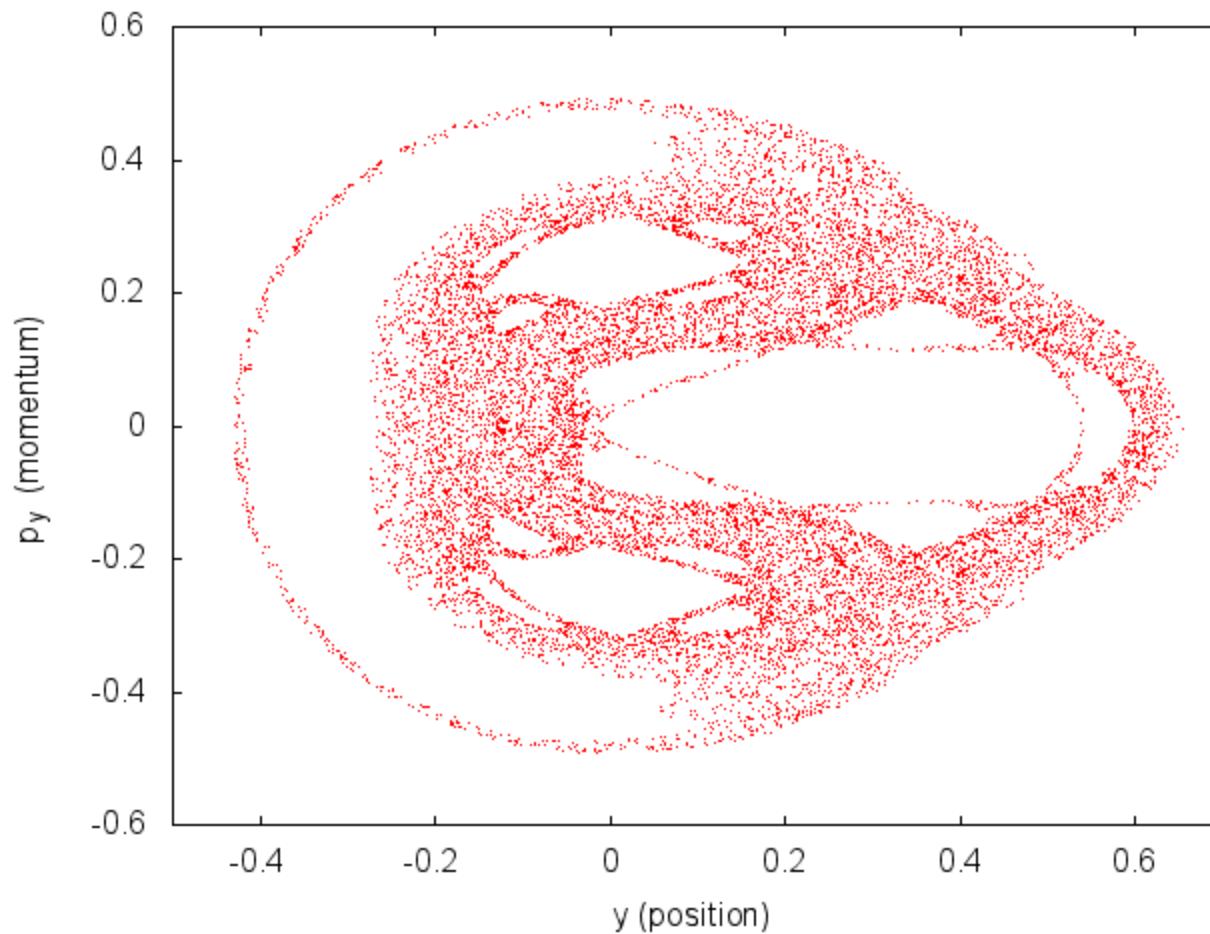


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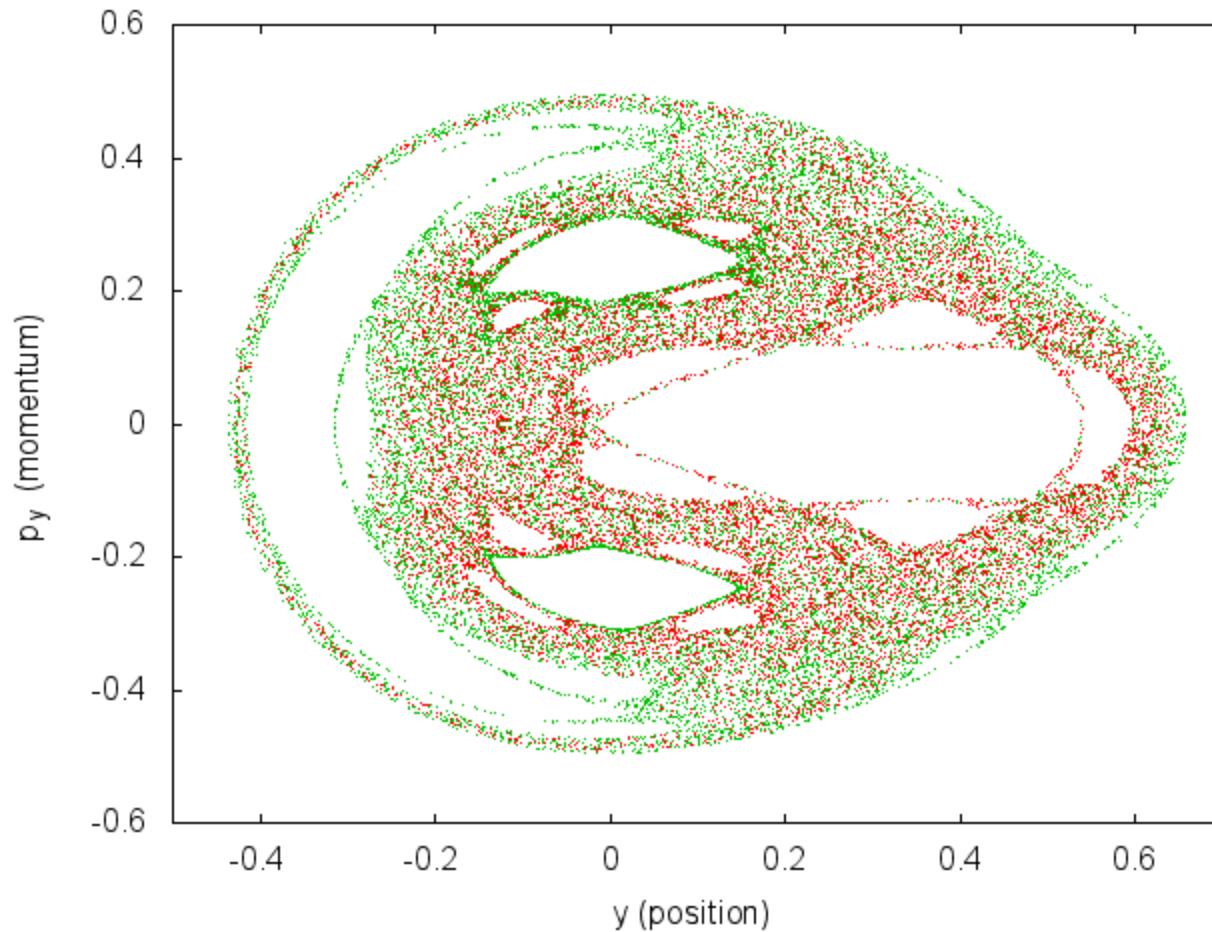
In this sense an $N+1$ degree of freedom Hamiltonian system corresponds to a **2N-dimensional map**.

Hénon-Heiles system: PSS ($x=0$)



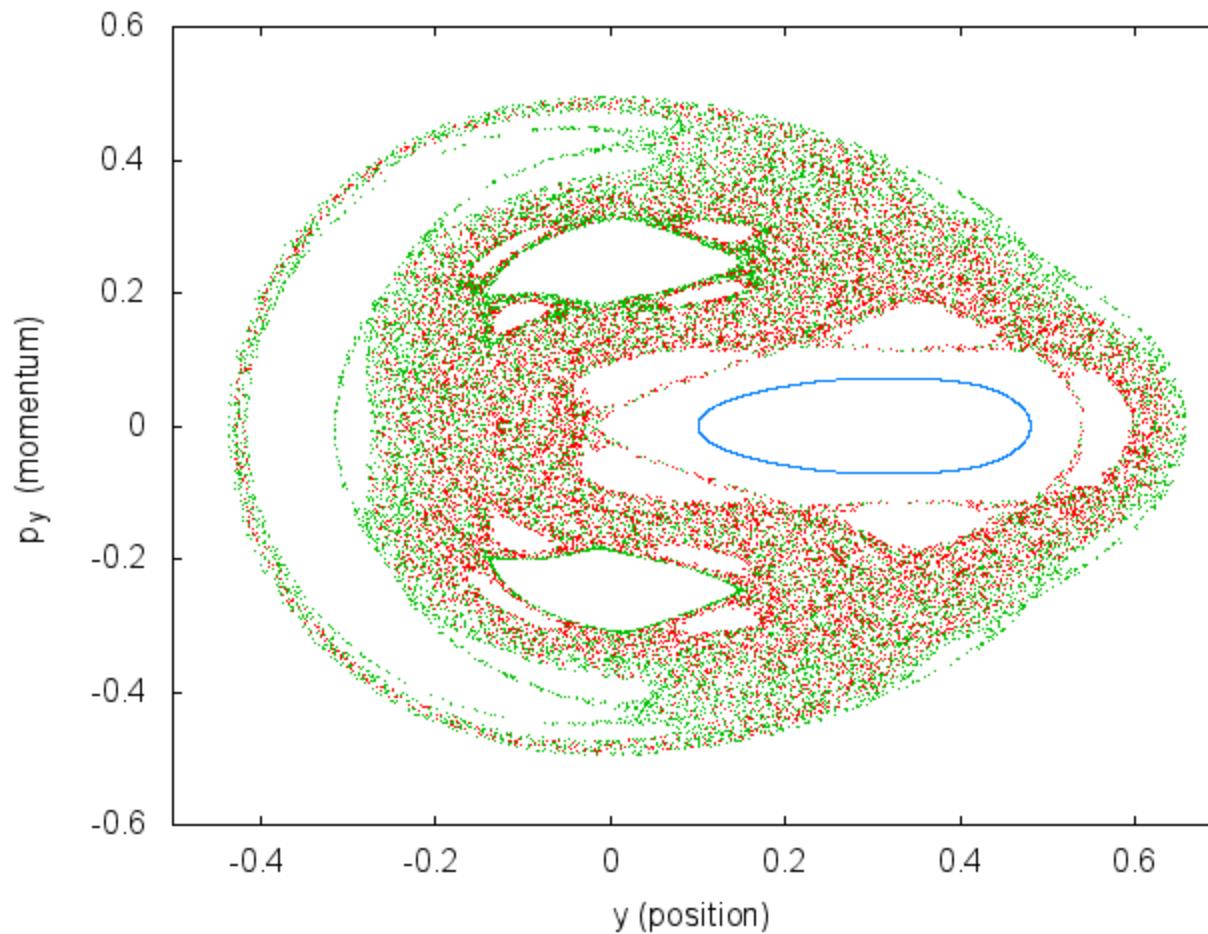
Chaotic orbit

Hénon-Heiles system: PSS ($x=0$)



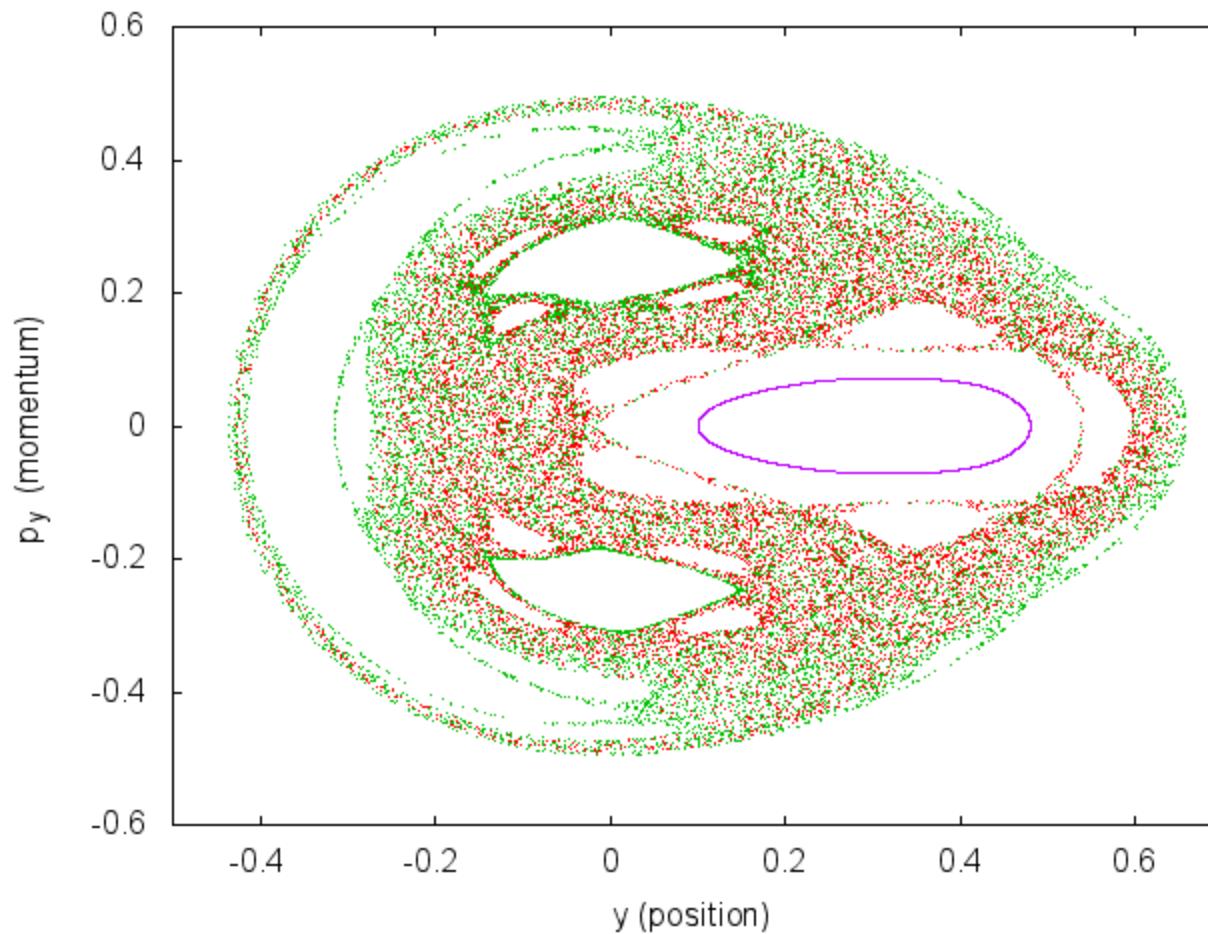
Chaotic orbit - Perturbed chaotic orbit

Hénon-Heiles system: PSS ($x=0$)



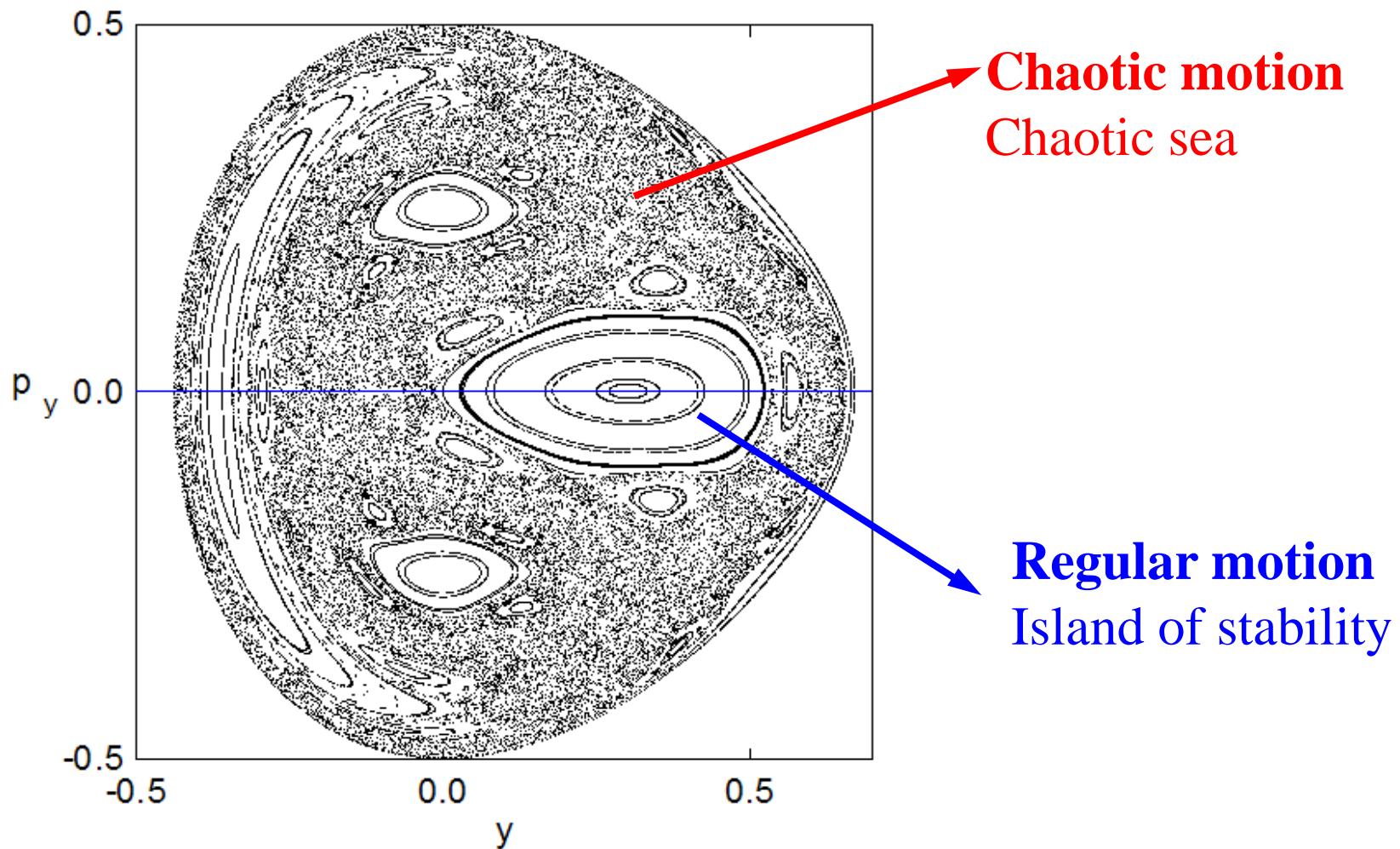
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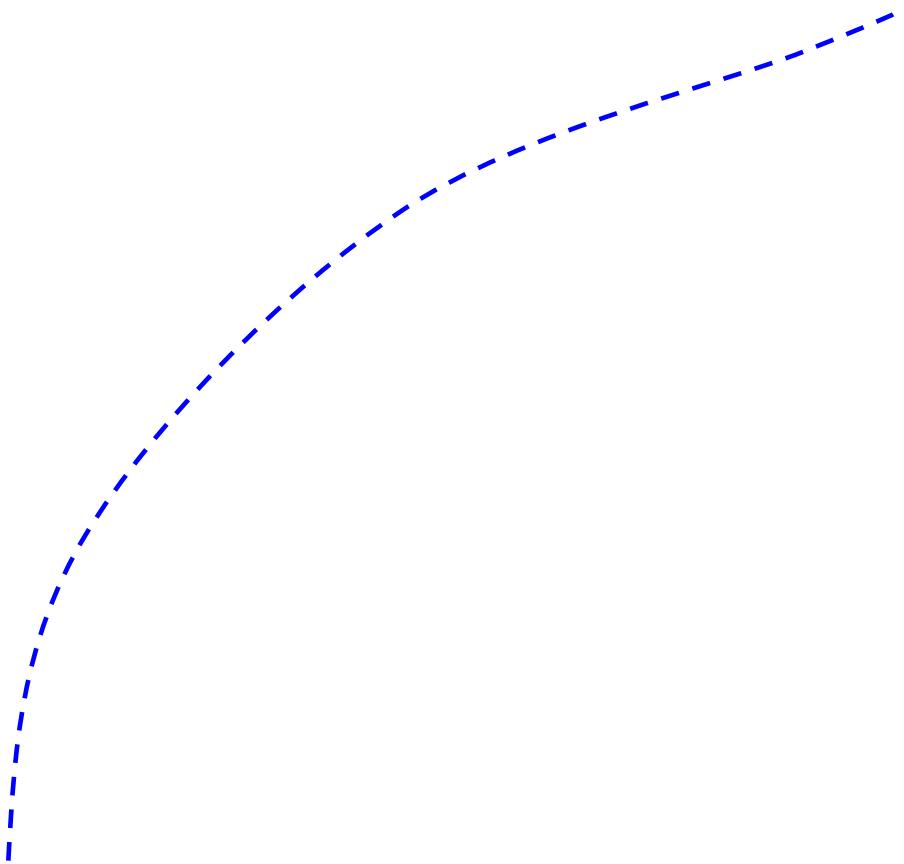


Chaotic orbit - Perturbed chaotic orbit
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Hénon-Heiles system: PSS ($x=0$)



Computation of the PSS



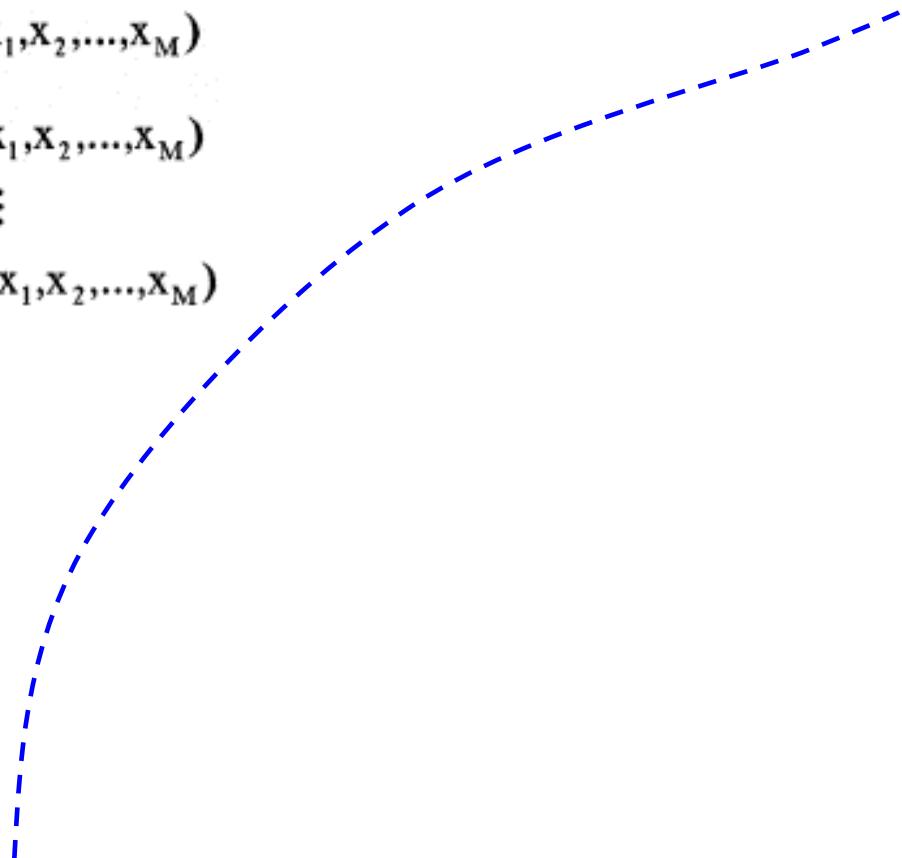
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$$\frac{dx_M}{dt} = f_M(x_1, x_2, \dots, x_M)$$



$t_n + \tau, x_M(t_n + \tau) - A > 0$



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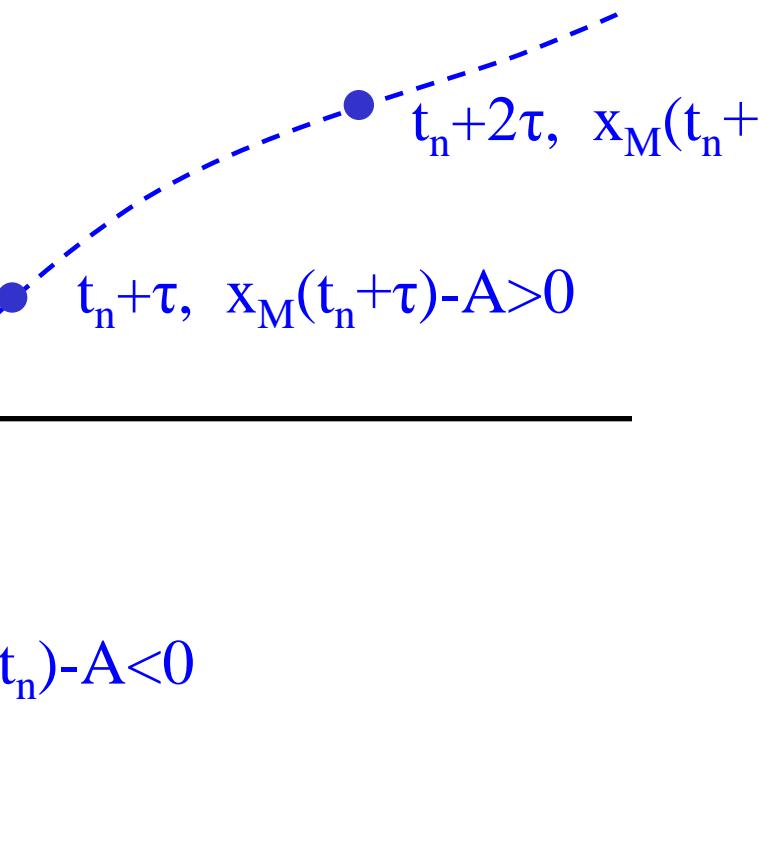
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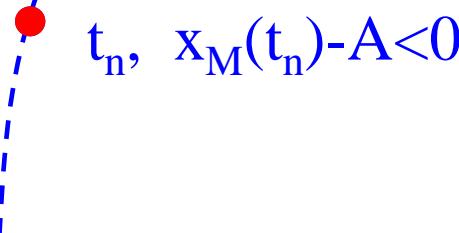
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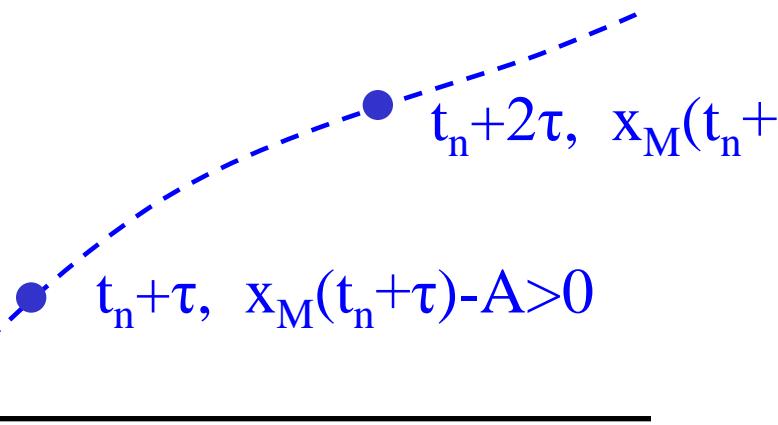
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$t=t_c, x_M(t_c)-A=0 !$

$t_n, x_M(t_n)-A < 0$

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PSS: $x_M-A=0$

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$$\vdots$$

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$$\frac{dt}{d\tau} = K$$

$$\begin{matrix} \tau=t, \\ K=1 \end{matrix}$$

$$\begin{aligned} \frac{dx_1}{dt} &= f_1(x_1, x_2, \dots, x_M) \\ \frac{dx_2}{dt} &= f_2(x_1, x_2, \dots, x_M) \\ &\vdots \\ \frac{dx_M}{dt} &= f_M(x_1, x_2, \dots, x_M) \end{aligned}$$

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$$\frac{dx_M}{d\tau} = K f_M(x_1, x_2, \dots, x_M)$$

$$\frac{dt}{d\tau} = K$$

$$\begin{aligned} & \tau = x_M, \\ & K = 1/f_M(x_1, x_2, \dots, x_M) \end{aligned}$$

$$\begin{array}{l} \tau = t, \\ K = 1 \end{array}$$

$$\frac{dx_1}{dt} = f_1(x_1, x_2, \dots, x_M)$$

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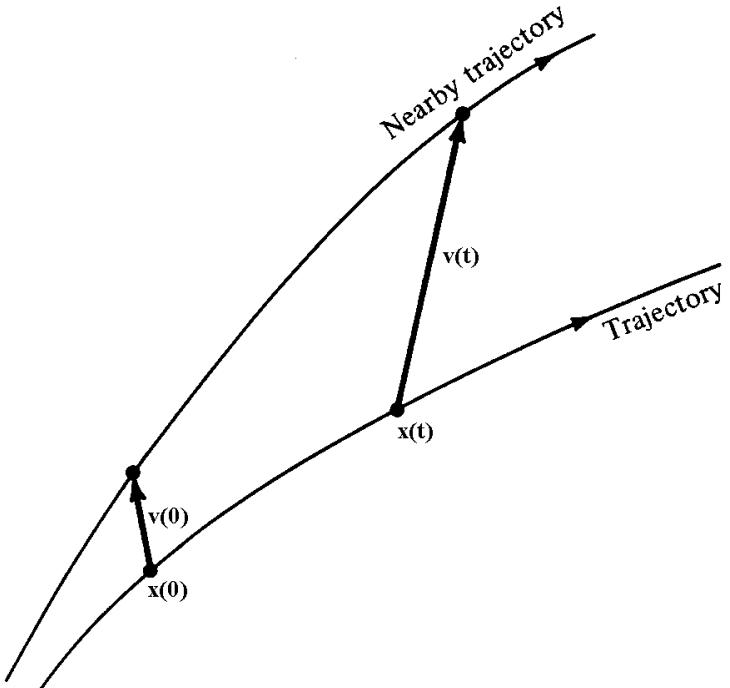
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Variational Equations

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

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

Example (Hénon-Heiles system)

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

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, px, py 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:

$$\dot{p}_x + \dot{v}_3 = -x - v_1 - 2(x + v_1)(y + v_2) \Rightarrow$$

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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 $x(0)$** and an **initial deviation vector from it $v(0)$** . Then the mean exponential rate of divergence is:

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

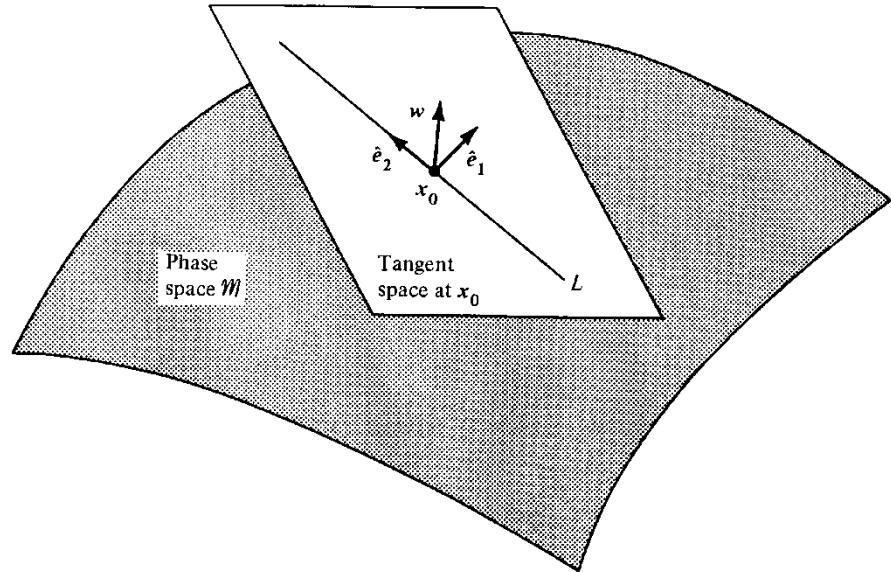
We commonly use the Euclidian norm and set $d(0) = \|v(0)\| = 1$

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.

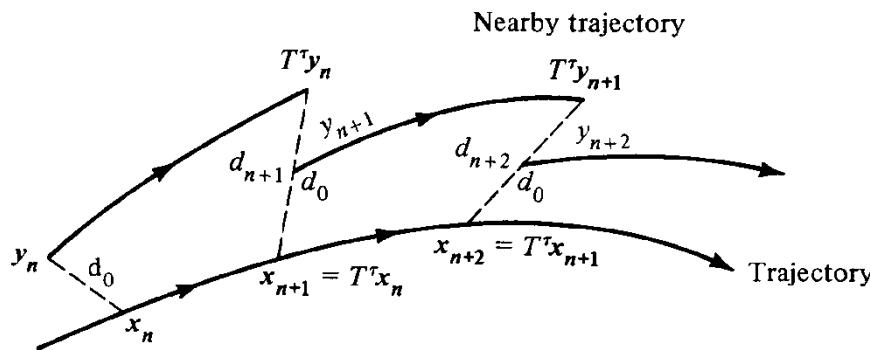


Figure 5.6. Numerical calculation of the maximal Liapunov characteristic exponent. Here $y = x + v$ and τ is a finite interval of time (after Benettin *et al.*, 1976).

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

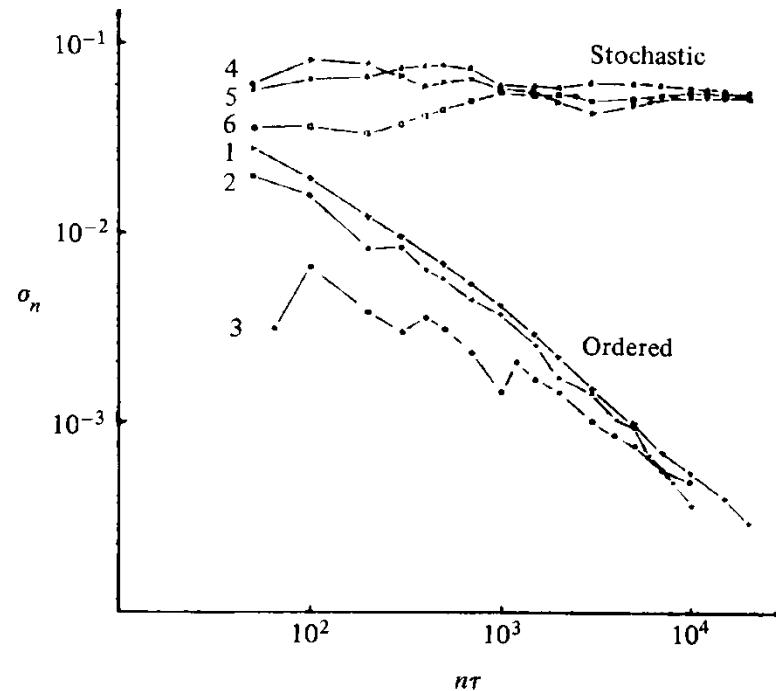
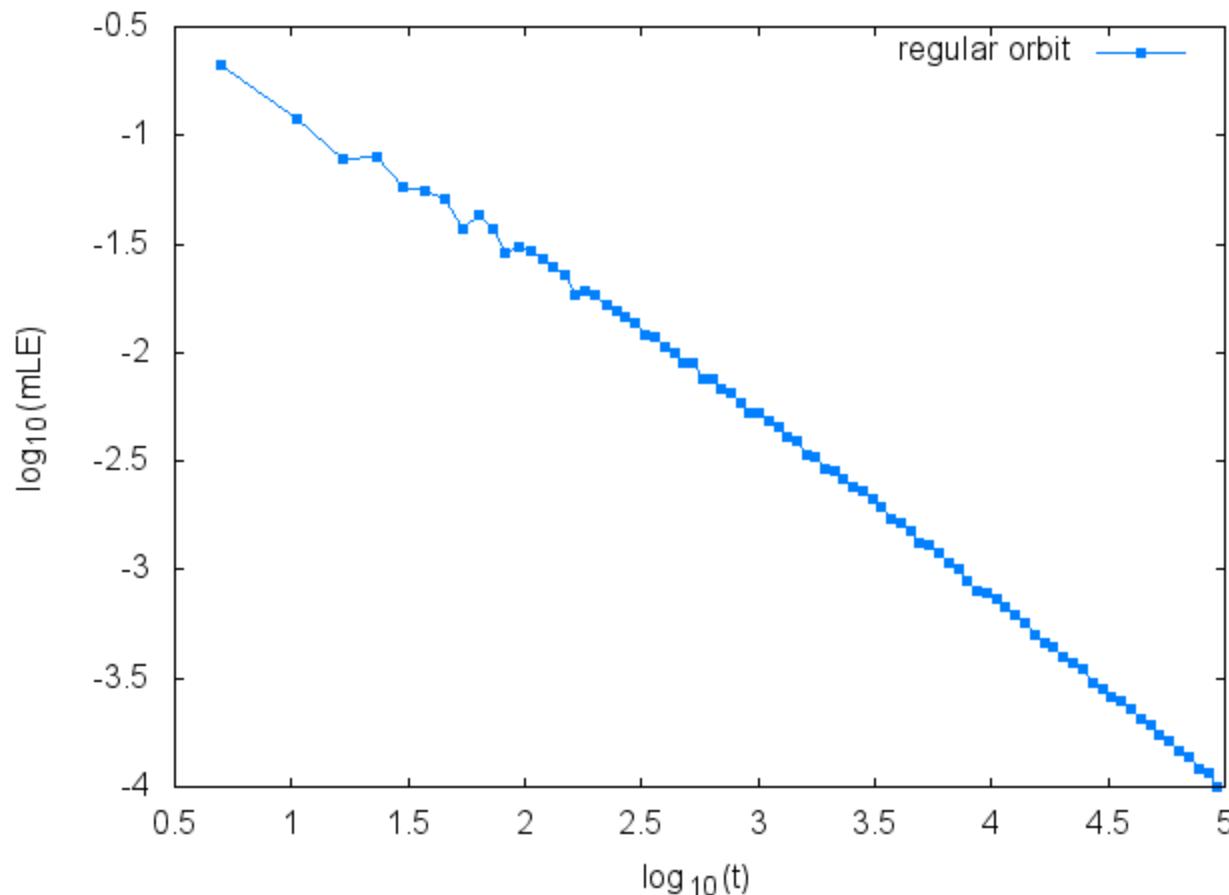


Figure 5.7. Behavior of σ_n at the intermediate energy $E = 0.125$ for initial points taken in the ordered (curves 1–3) or stochastic (curves 4–6) regions (after Benettin *et al.*, 1976).

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.

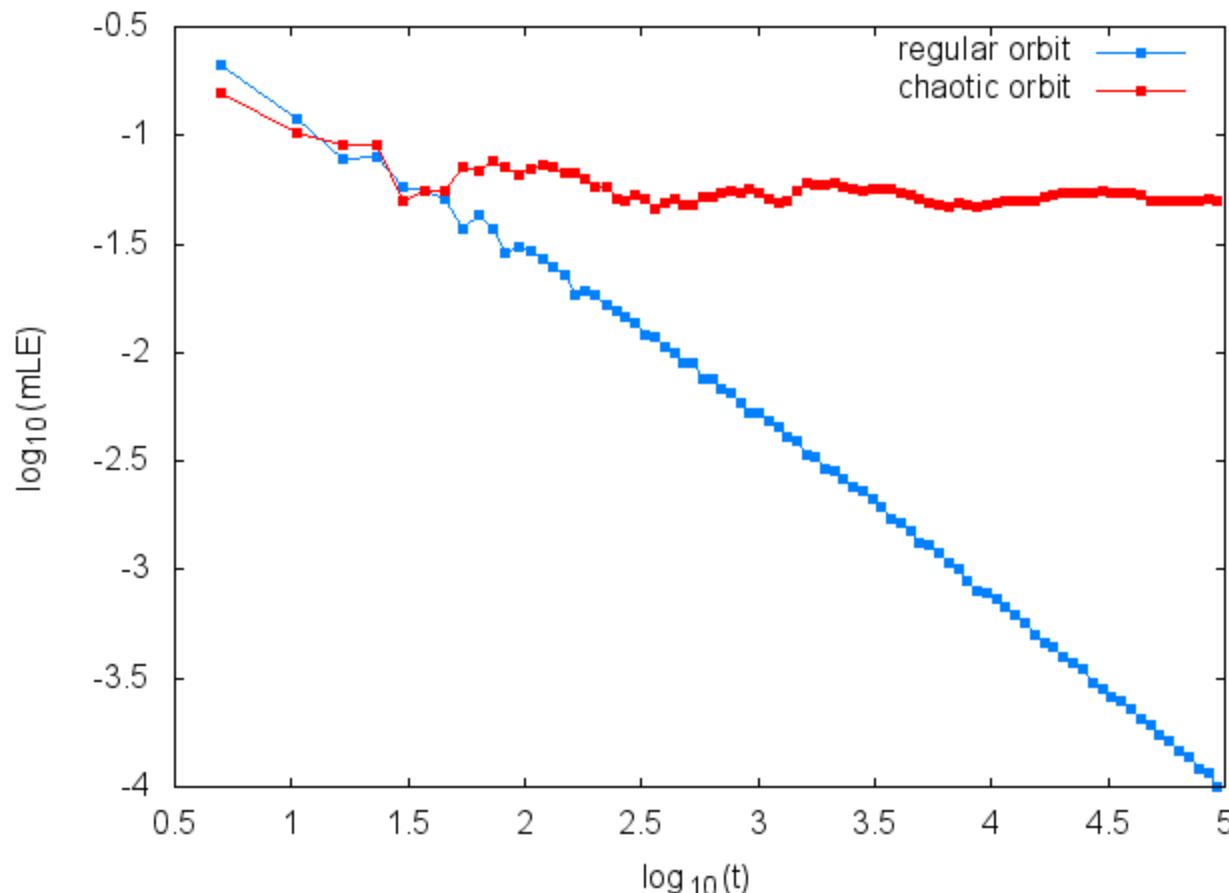
Maximum Lyapunov Exponent

Hénon-Heiles system: **Regular orbit**



Maximum Lyapunov Exponent

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



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