

Chaos in Hamiltonian systems

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28th Summer School - Conference 'Dynamical Systems and Complexity', Greece, 19 July 2022

Outline

- **Chaos**
- **Autonomous Hamiltonian systems. Example: Hénon-Heiles system**
- **Regular vs Chaotic motion**
- **Visualization of chaos: Poincaré Surface of Section (PSS)**
- **Chaos Indicators**
 - ✓ **Variational equations and Tangent map**
 - ✓ **Lyapunov exponents**
 - ✓ **Smaller ALignment Index – SALI**
 - ✓ **Generalized ALignment Index – GALI**

Chaos

Definition [Devaney (1989)]

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1. f has **sensitive dependence on initial conditions.**
2. f is **topologically transitive.**
3. **periodic points are dense in V .**

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

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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 \mathbf{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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$\mathbf{f} : V \rightarrow V$ 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.

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

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

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

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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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Hamilton's equations of motion:

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

Regular vs Chaotic orbits

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

$$\mathbf{x}=0, \mathbf{y}=0.1, \mathbf{p}_y=0 \text{ and } \mathbf{x}=0, \mathbf{y}=-0.25, \mathbf{p}_y=0.$$

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We perturb both ICs by $\delta\mathbf{p}_y=10^{-11}$ (!) and check the evolution of x

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Perturbed

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Orbit

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t= 100 x= 0.132995718333307644 0.132995718337263064

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

$$t= 5000 \quad x= 0.376999283889102310 \quad 0.376999283870156576$$

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| 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= 300 | x= 0.515226330109450181 | 0.515225440480693297 |

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| t=100000 | x=-0.381120533746511780 | -0.381120533327258870 |
| t= 100 | x= 0.090272817735167835 | 0.090272821355768668 |
| 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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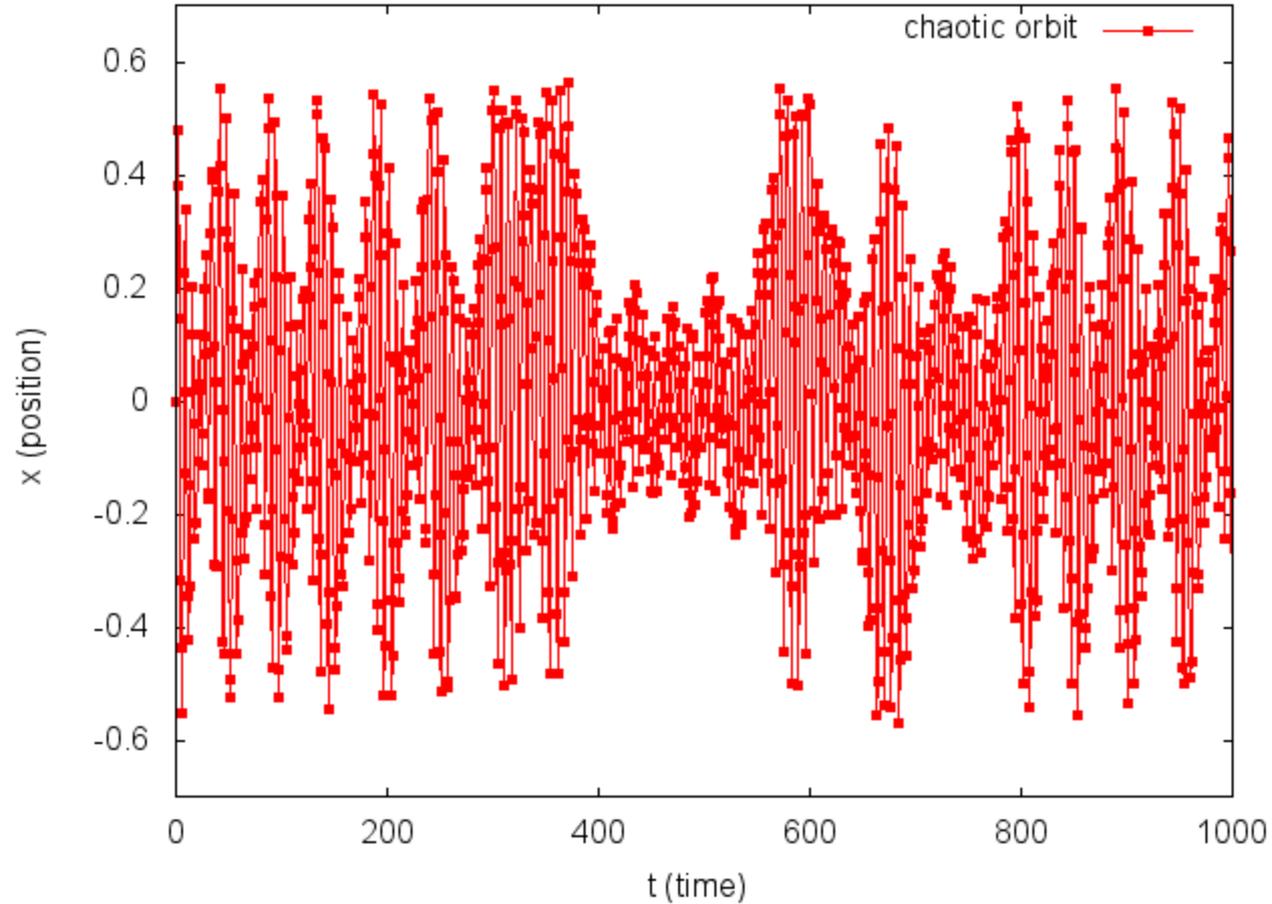
$x=0, y=0.1, p_y=0$ and **$x=0, y=-0.25, p_y=0$** .

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

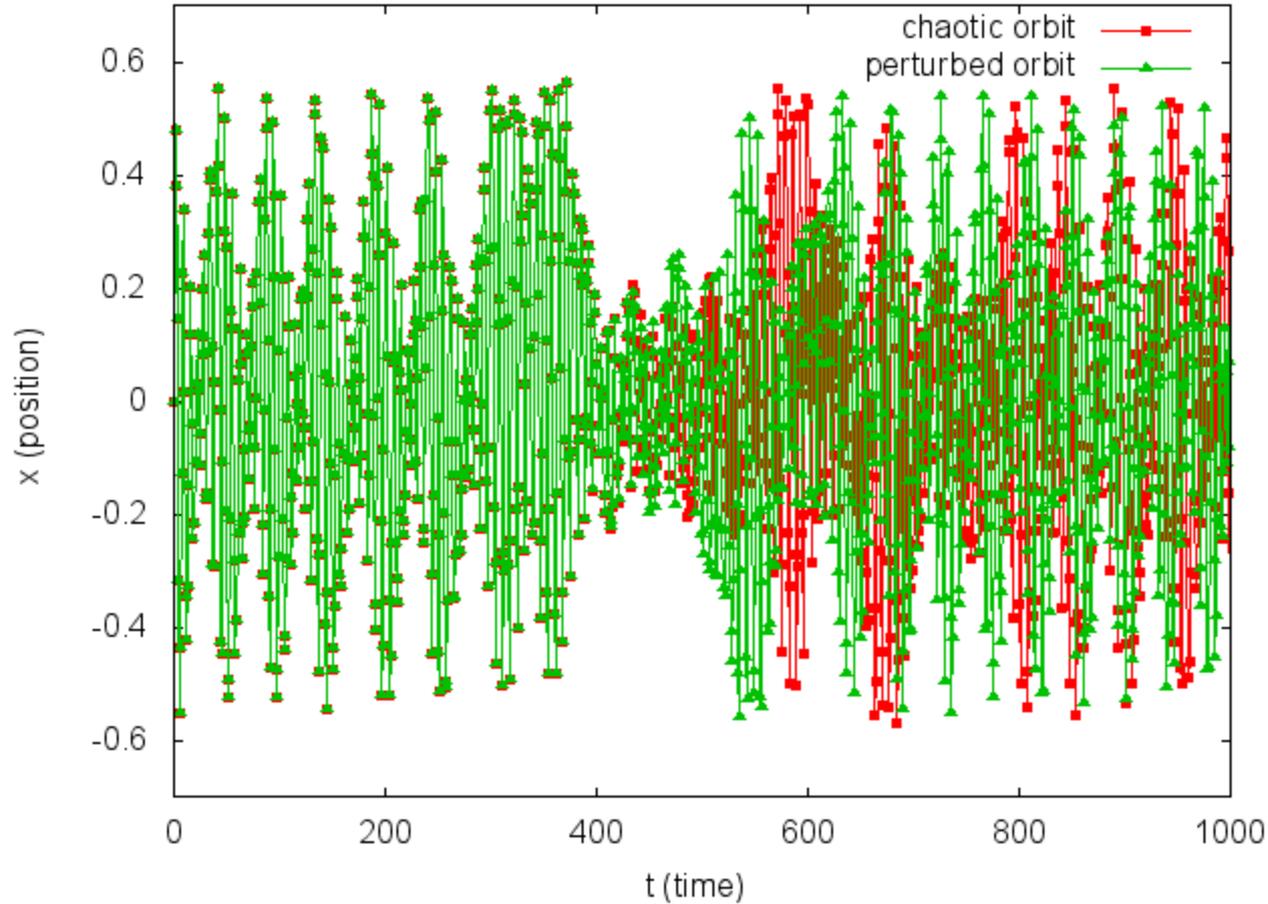
Regular vs Chaotic orbits

Chaotic orbit



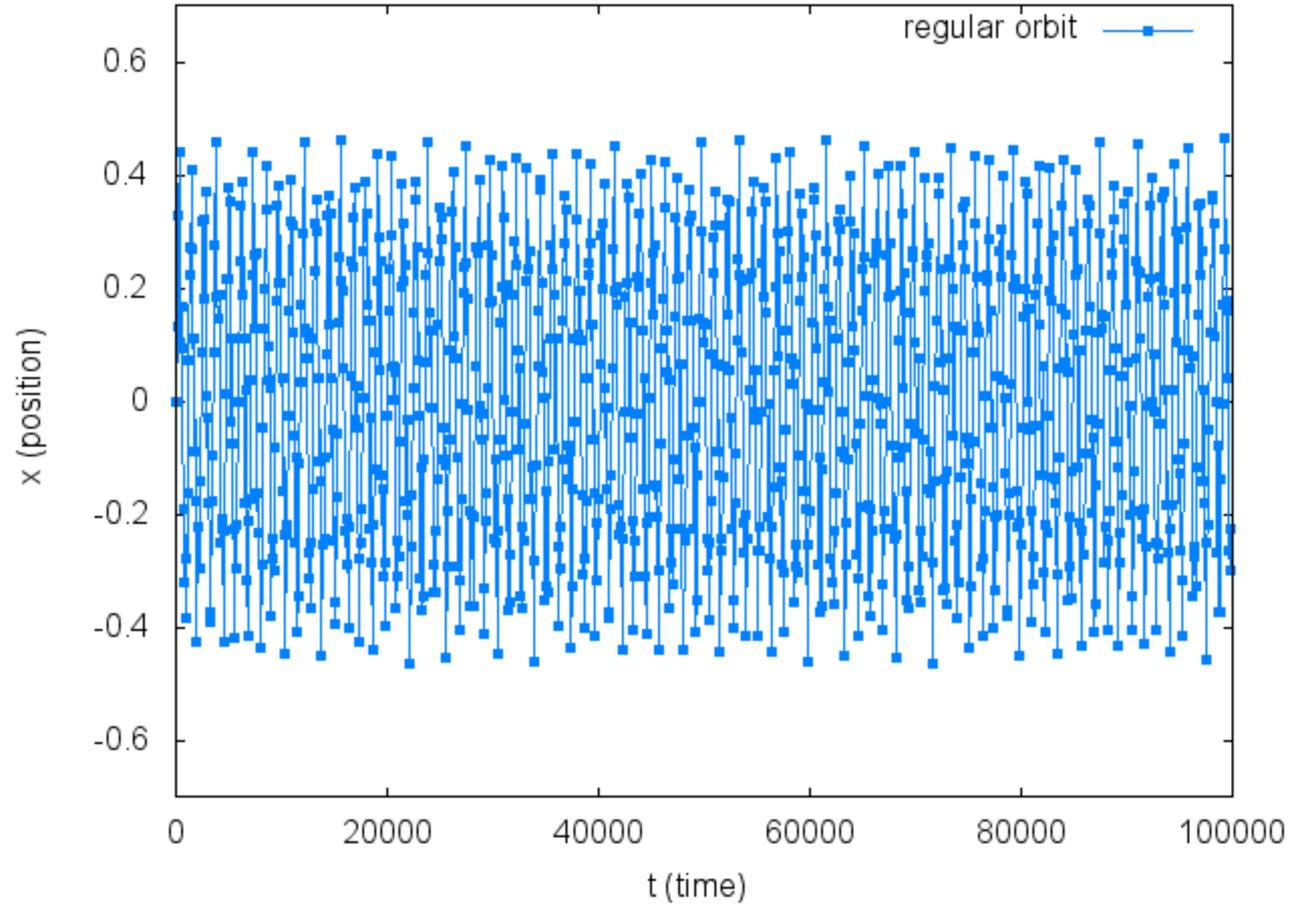
Regular vs Chaotic orbits

Chaotic orbit and its perturbation



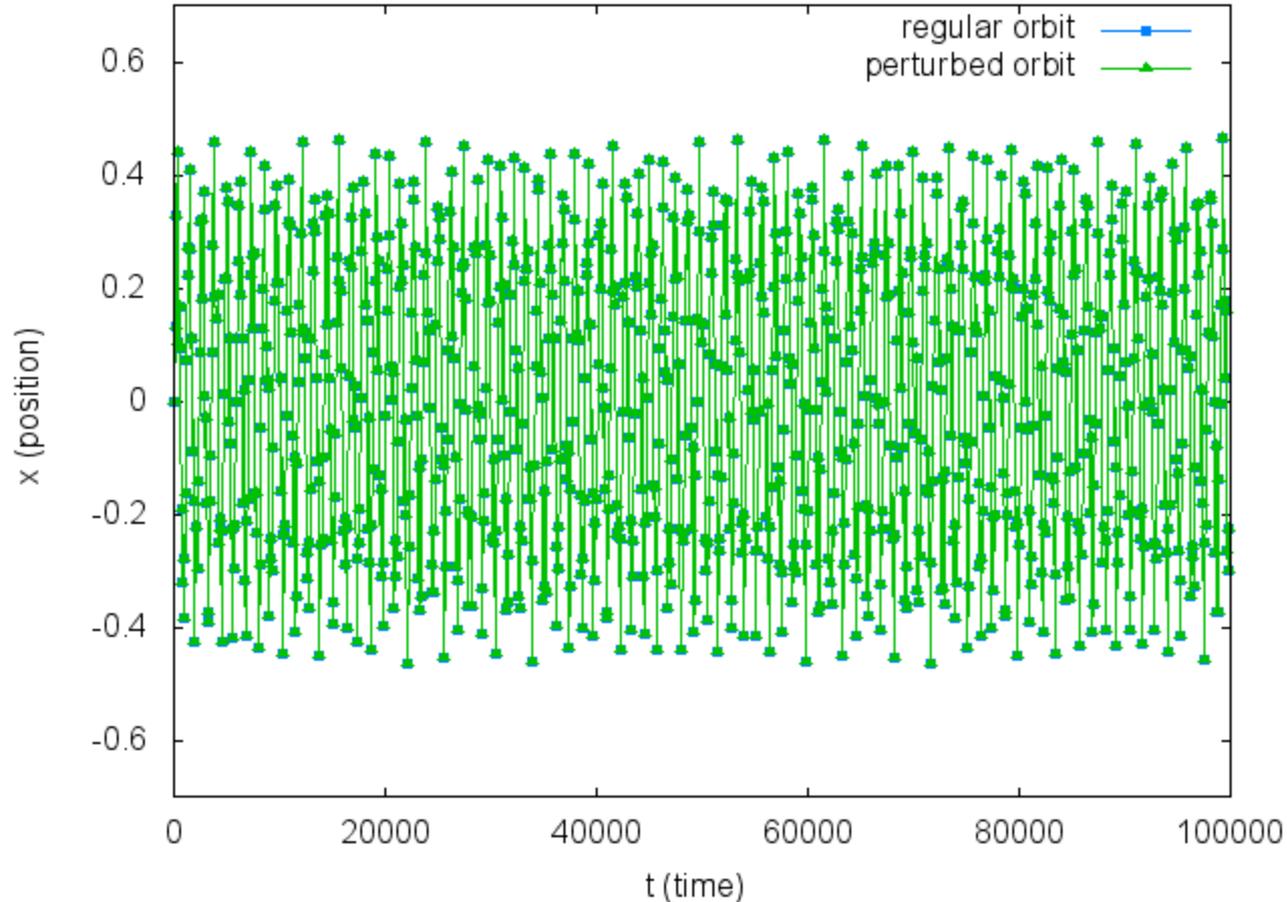
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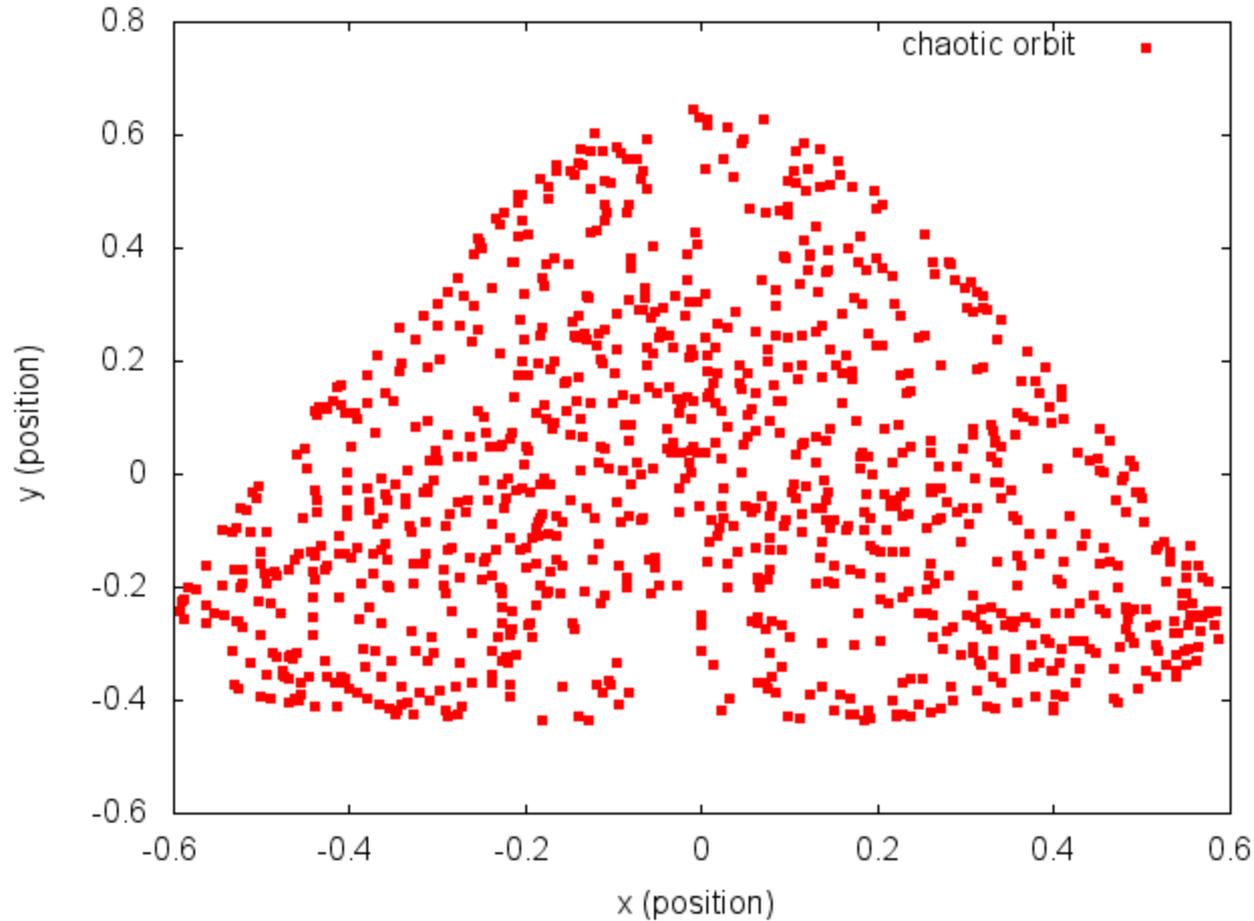
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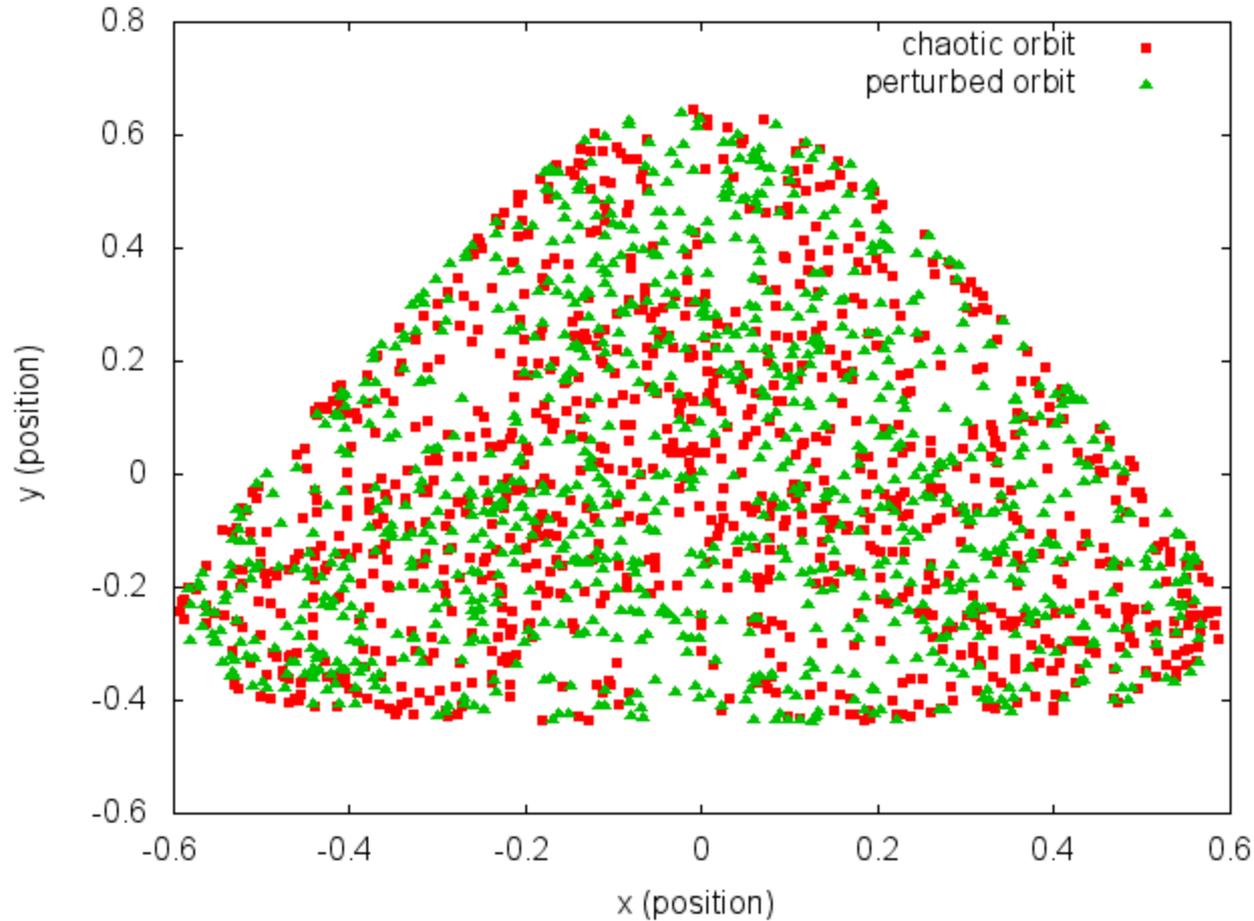
Chaotic orbit



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

Regular vs Chaotic orbits

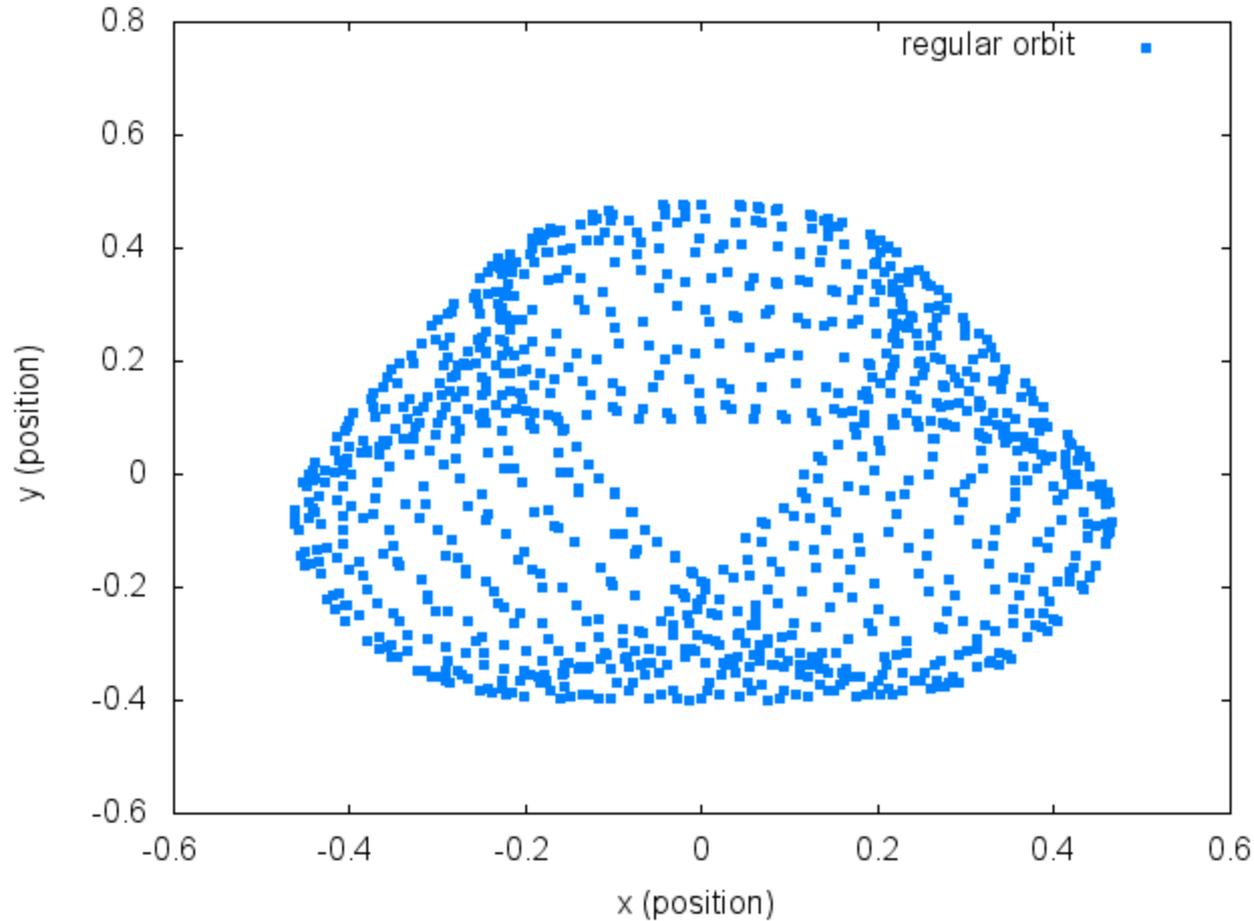
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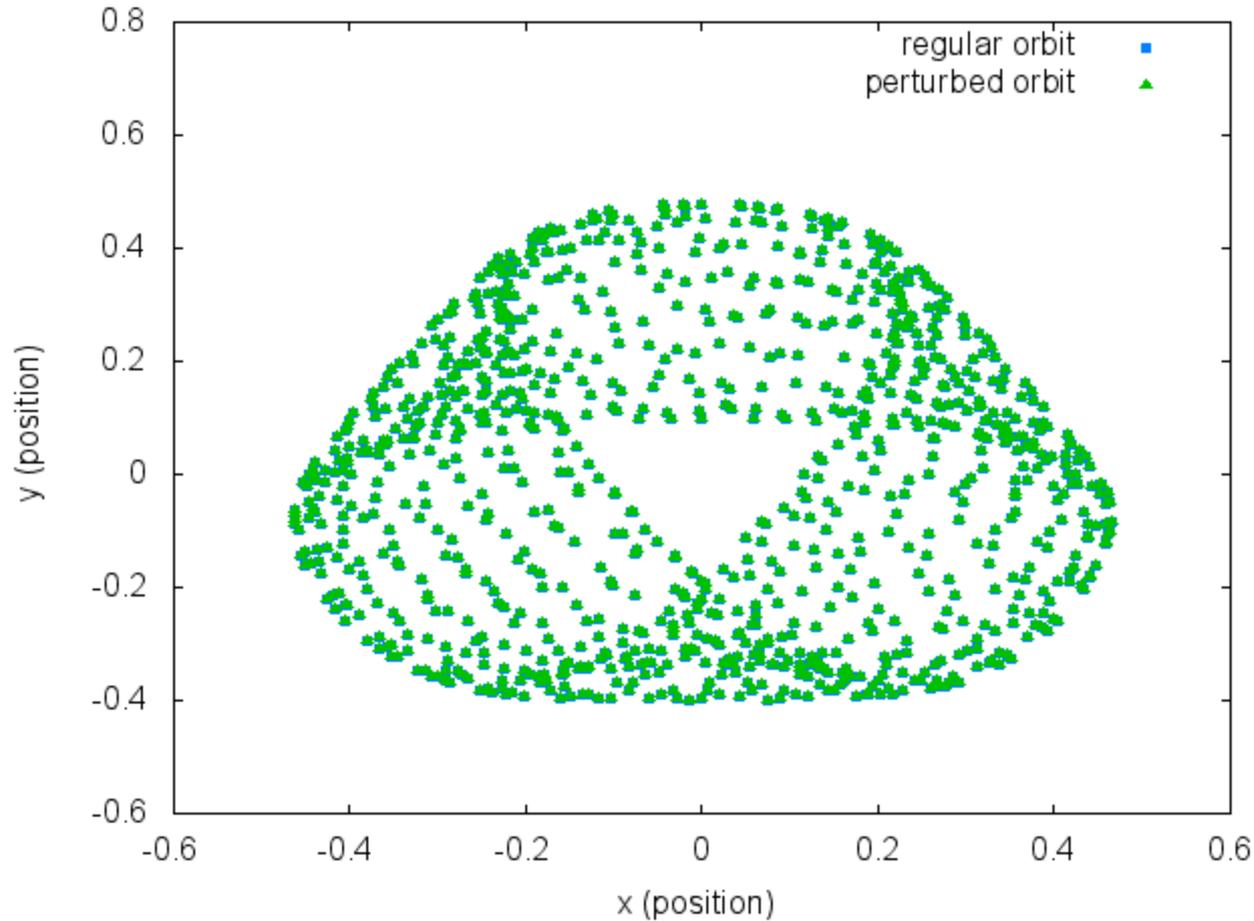
Regular orbit



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

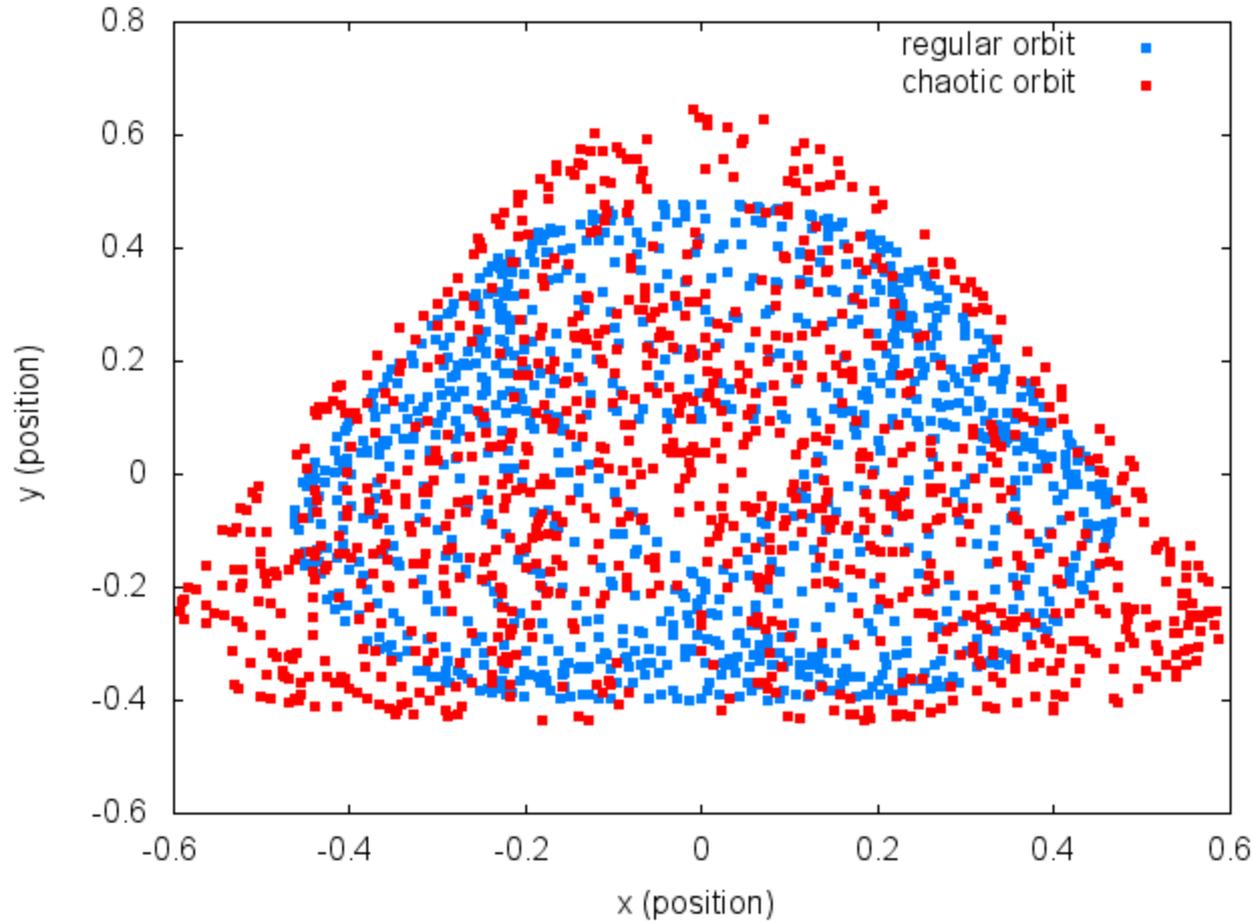
Regular vs Chaotic orbits

Regular orbit and its perturbation



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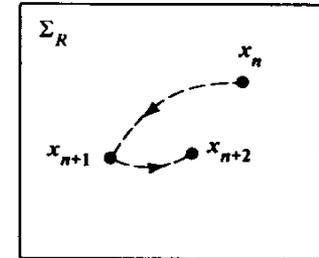
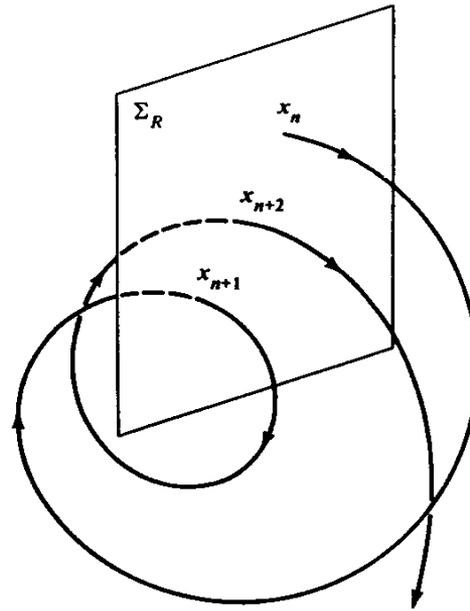
Regular vs Chaotic orbits



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Poincaré Surface of Section (PSS)

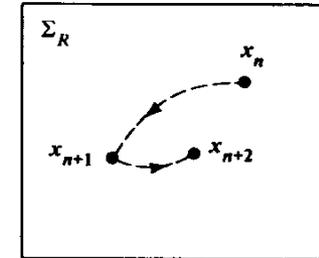
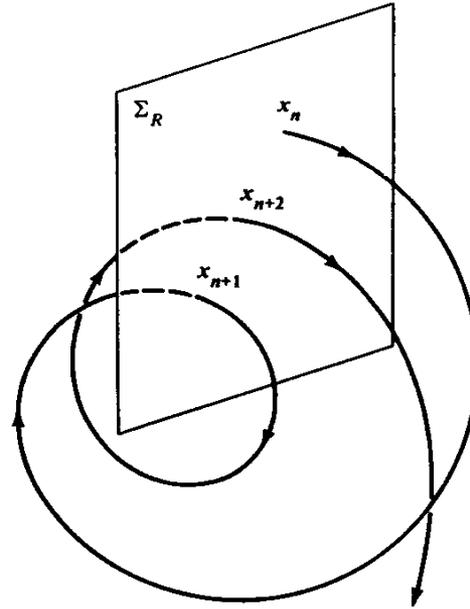
We can constrain the study of an $N+1$ degree of freedom Hamiltonian system to a **2N-dimensional subspace of the general phase space.**



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

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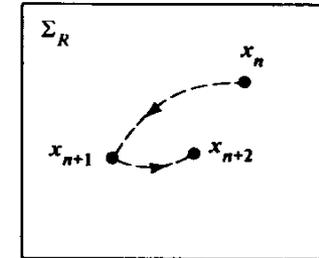
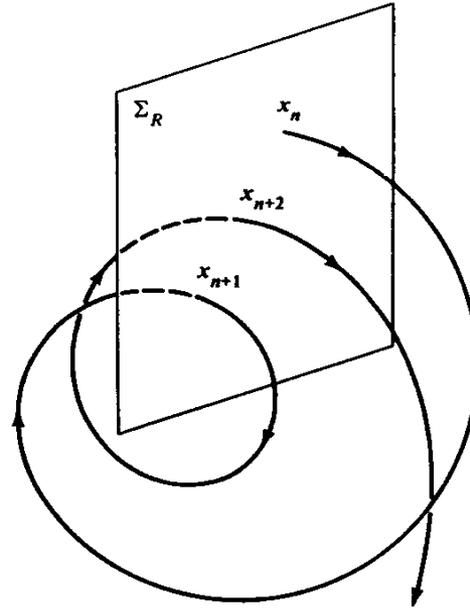


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In general we can assume a PSS of the form **$q_{N+1} = \text{constant}$** . Then only variables $q_1, q_2, \dots, q_N, p_1, p_2, \dots, p_N$ are needed to describe the evolution of an orbit on the PSS, since p_{N+1} can be found from the Hamiltonian.

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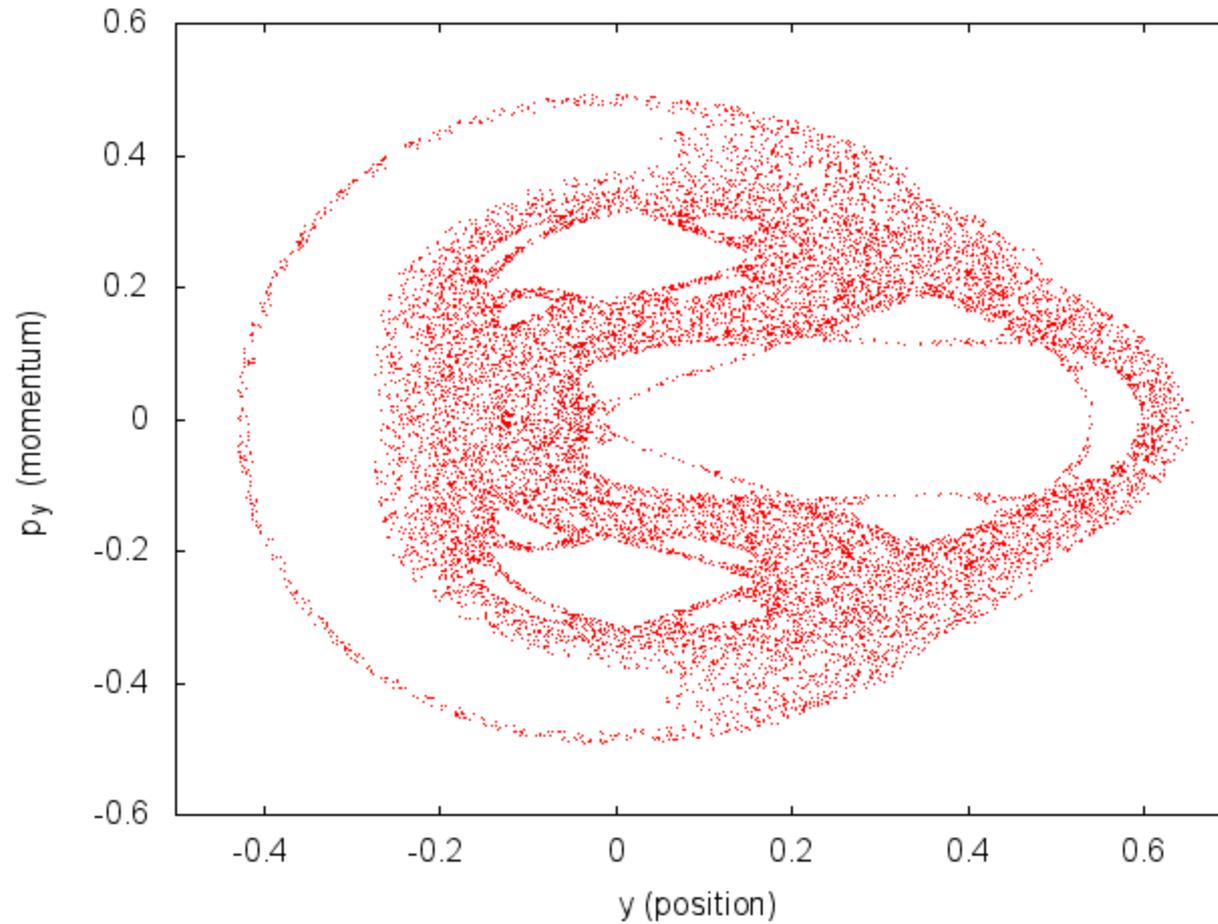


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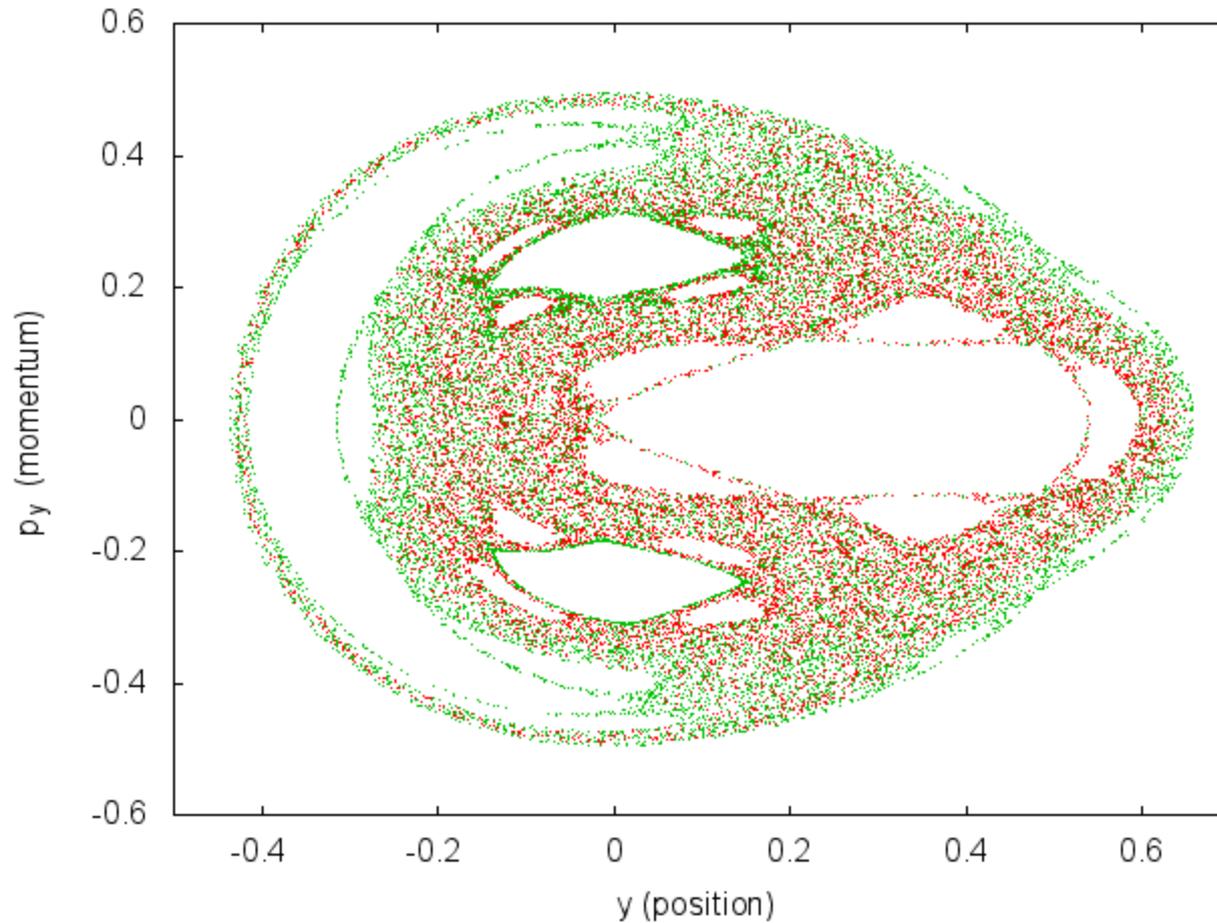
In this sense **an $N+1$ degree of freedom Hamiltonian system corresponds to a 2N-dimensional map.**

Hénon-Heiles system: PSS ($x=0$)



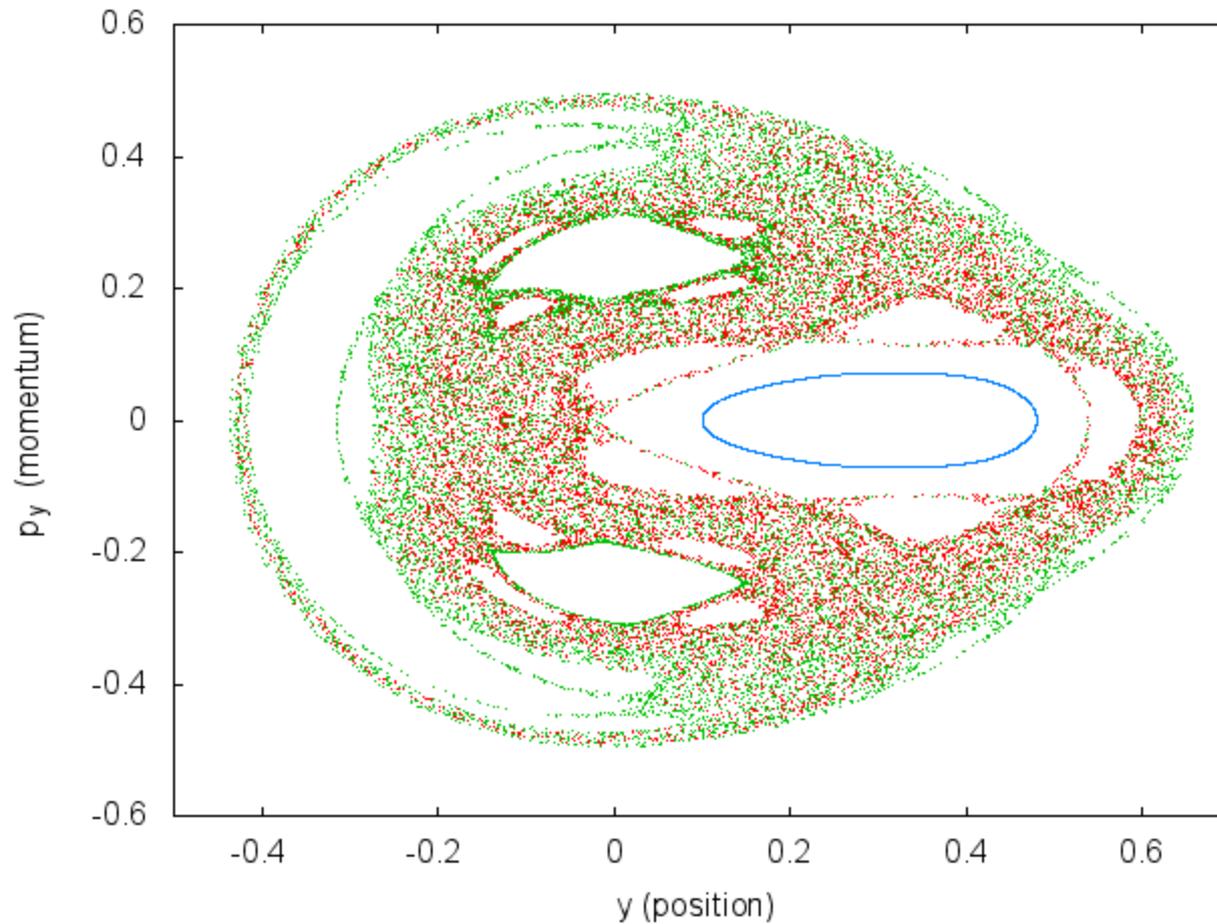
Chaotic orbit

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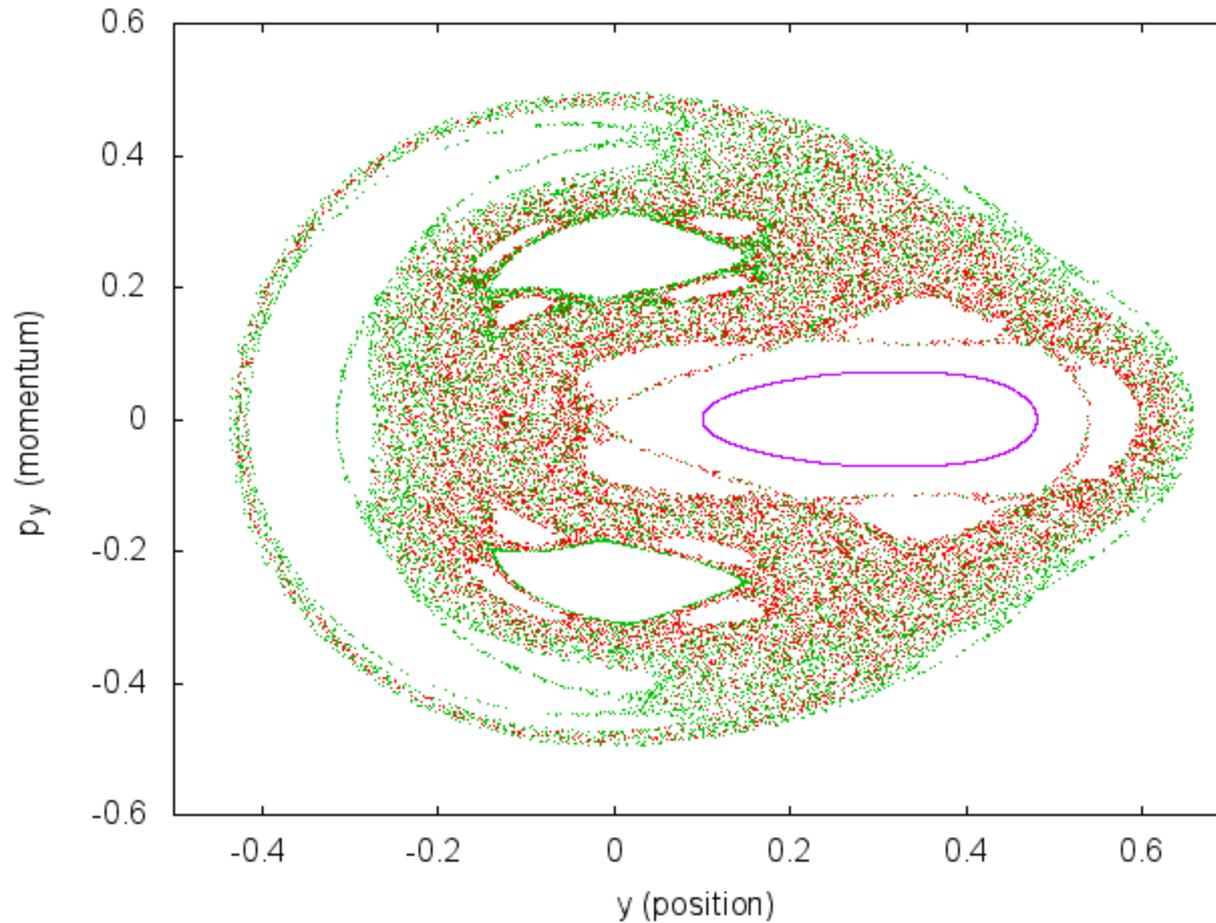
Chaotic orbit - **Perturbed chaotic orbit**

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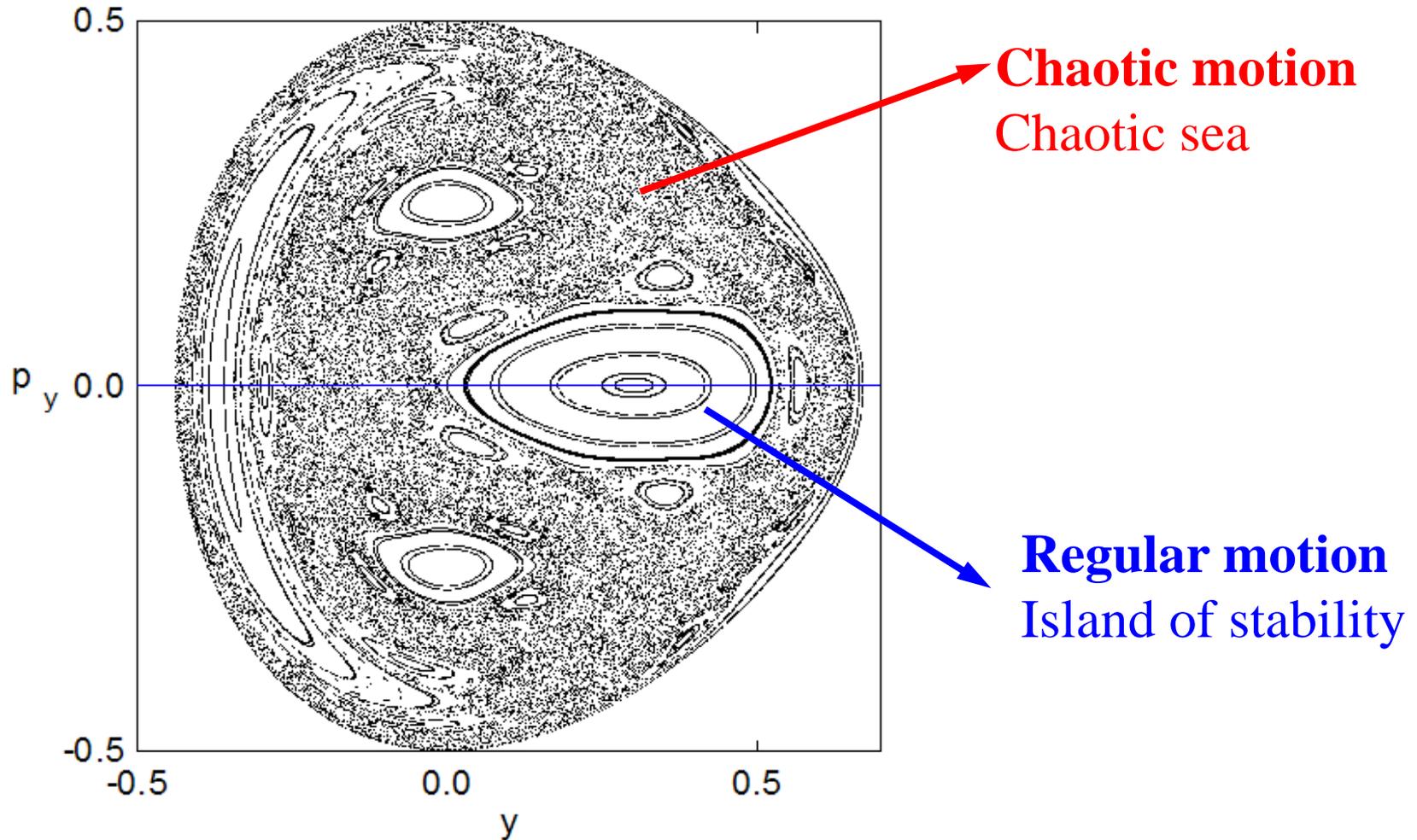
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Chaotic orbit - **Perturbed chaotic orbit**
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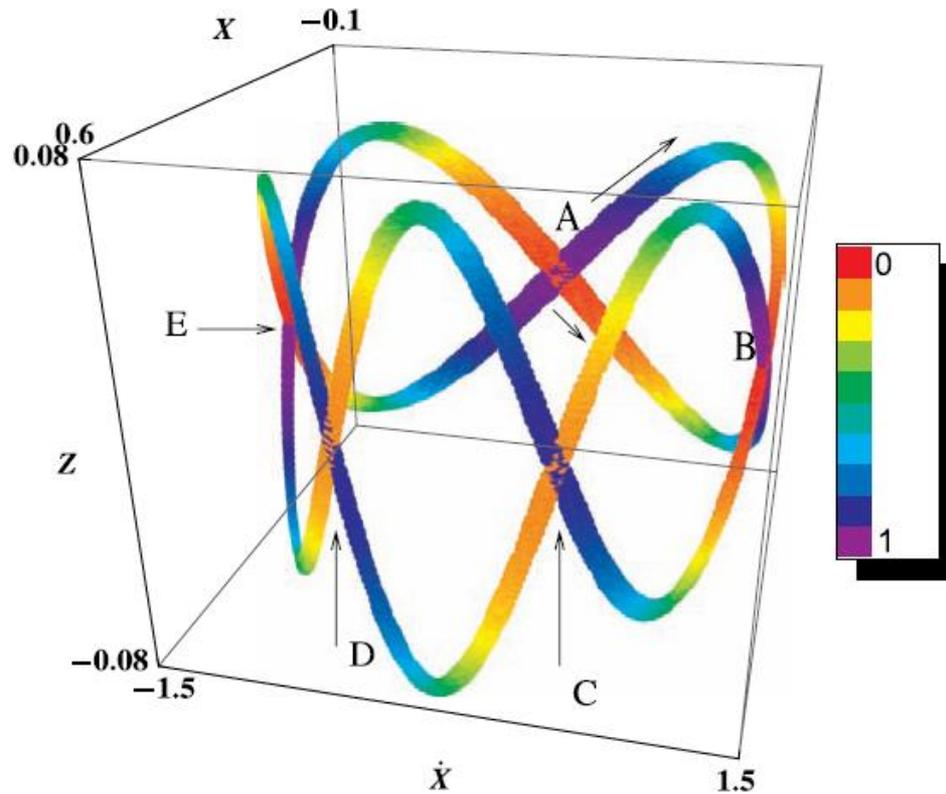
Chaos detection techniques

- **Based on the visualization of orbits**
 - ✓ **Poincaré Surface of Section (PSS)**
 - ✓ **the color and rotation (CR) method**
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The color and rotation (CR) method

For 3 degree of freedom Hamiltonian systems and 4 dimensional symplectic maps:

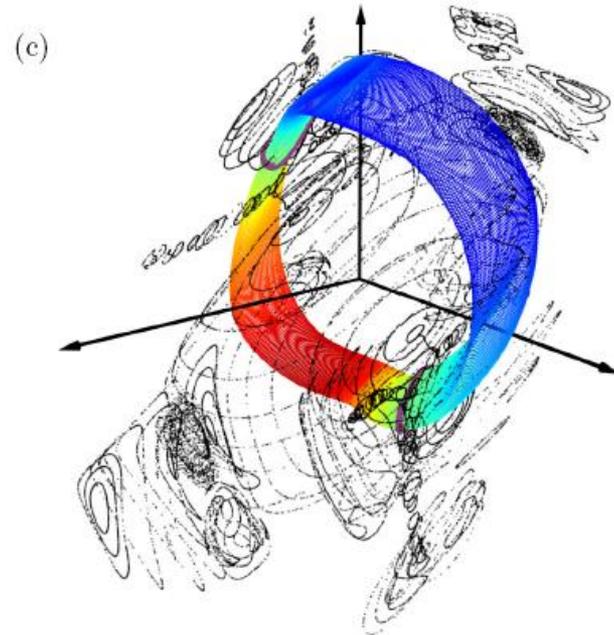
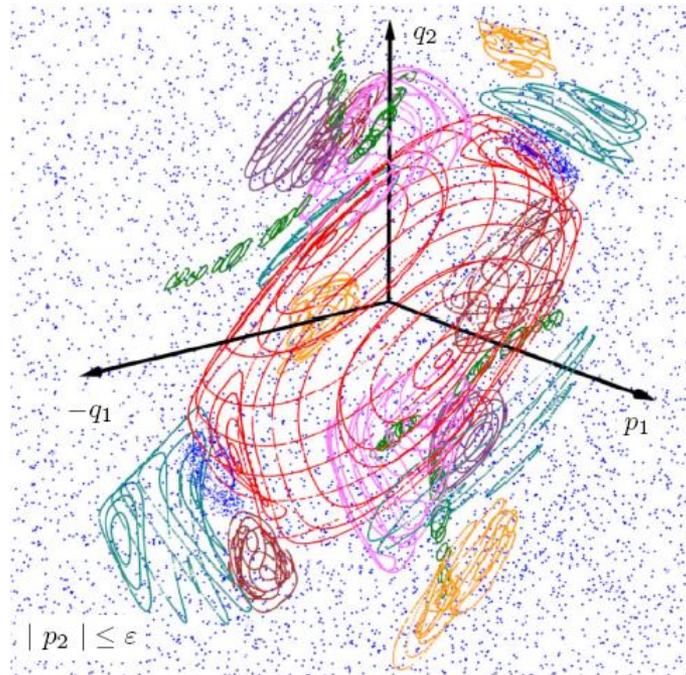
We consider the 3D projection of the PSS and use color to indicate the 4th dimension.



The 3D phase space slices (3PSS) technique

For 3 degree of freedom Hamiltonian systems and 4 dimensional symplectic maps:

We consider thin 3D phase space slices of the 4D phase space (e.g. $|p_2| \leq \epsilon$) and present intersections of orbits with these slices.



Chaos detection techniques

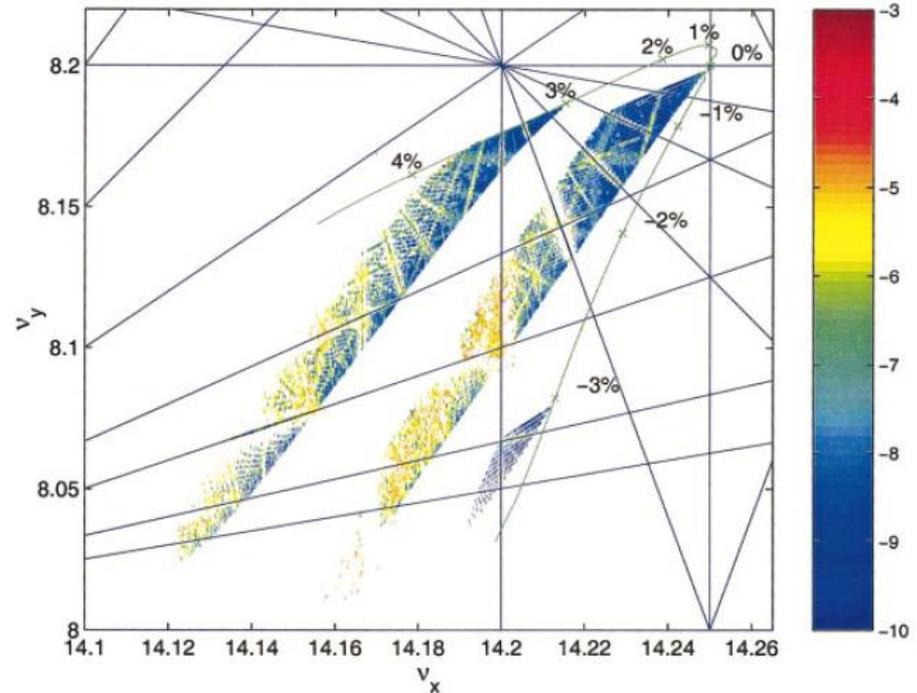
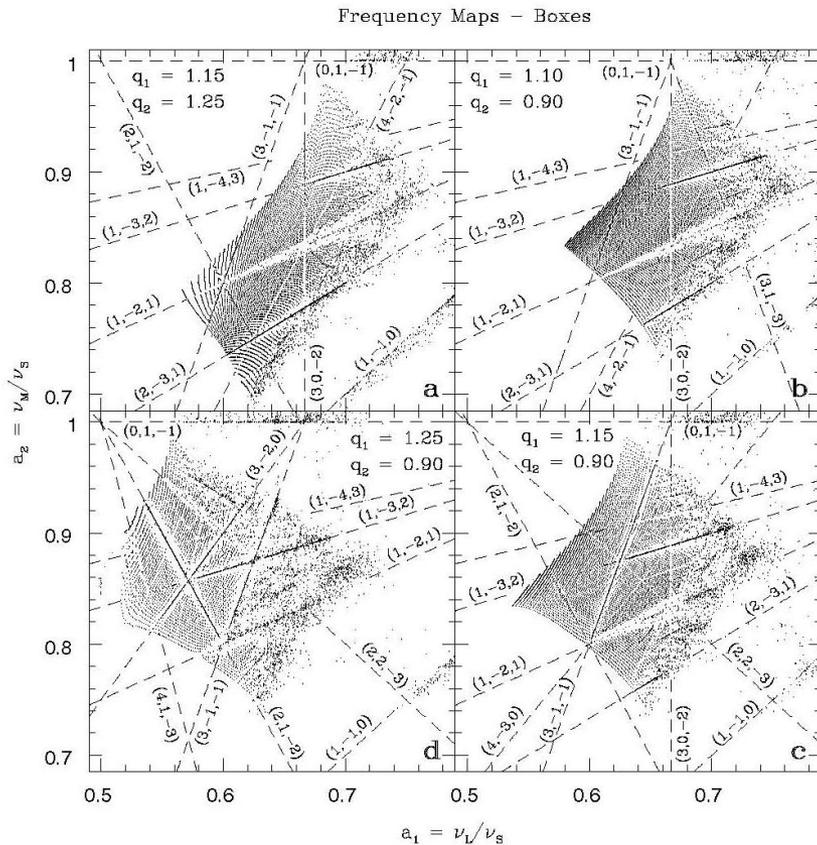
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 - ✓ **Frequency Map Analysis**
 - ✓ **0-1 test**

Frequency Map Analysis

Create **Frequency Maps** by computing the fundamental frequencies of orbits.

Regular motion: The computed frequencies do not vary in time

Chaotic motion: The computed frequencies vary in time



Steier C et al. 2002 Phys. Rev. E 65 056506

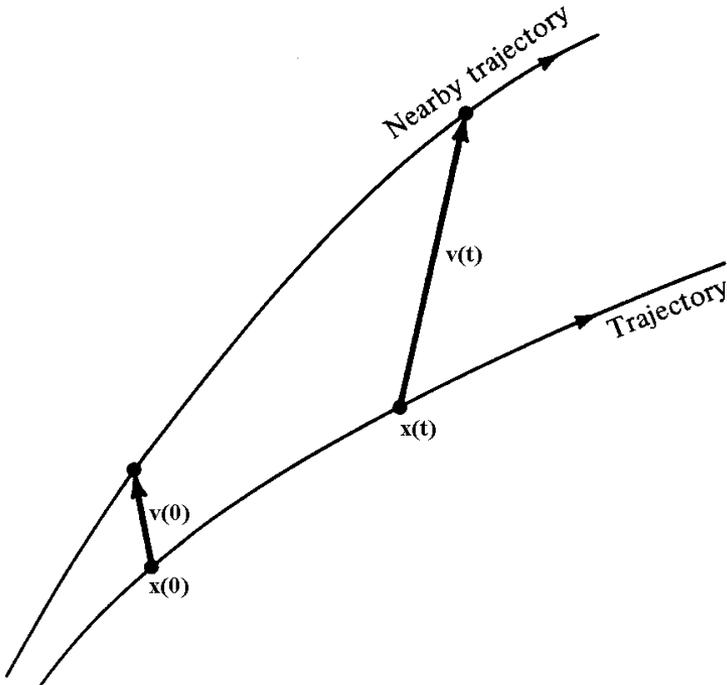
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- **Chaos indicators based on the evolution of deviation vectors from a given orbit**
 - ✓ **Maximum Lyapunov Exponent**
 - ✓ Fast Lyapunov Indicator (FLI) and Orthogonal Fast Lyapunov Indicators (OFLI and OFLI2)
 - ✓ Mean Exponential Growth Factor of Nearby Orbits (MEGNO)
 - ✓ Relative Lyapunov Indicator (RLI)
 - ✓ **Smaller ALignment Index – SALI**
 - ✓ **Generalized ALignment Index – GALI**

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

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

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:

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

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Complete set of equations

Symplectic Maps

Consider an **2N-dimensional symplectic map T**. In this case we have **discrete time**.

This is an area-preserving map whose Jacobian matrix

$$\mathbf{M} = \frac{\partial \mathbf{T}}{\partial \mathbf{x}} = \begin{bmatrix} \frac{\partial \mathbf{T}_1}{\partial \mathbf{x}_1} & \frac{\partial \mathbf{T}_1}{\partial \mathbf{x}_2} & \dots & \frac{\partial \mathbf{T}_1}{\partial \mathbf{x}_{2N}} \\ \frac{\partial \mathbf{T}_2}{\partial \mathbf{x}_1} & \frac{\partial \mathbf{T}_2}{\partial \mathbf{x}_2} & \dots & \frac{\partial \mathbf{T}_2}{\partial \mathbf{x}_{2N}} \\ \vdots & \vdots & & \vdots \\ \frac{\partial \mathbf{T}_{2N}}{\partial \mathbf{x}_1} & \frac{\partial \mathbf{T}_{2N}}{\partial \mathbf{x}_2} & \dots & \frac{\partial \mathbf{T}_{2N}}{\partial \mathbf{x}_{2N}} \end{bmatrix}$$

satisfies

$$\mathbf{M}^T \cdot \mathbf{J}_{2N} \cdot \mathbf{M} = \mathbf{J}_{2N}$$

Symplectic Maps

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The evolution of an **orbit** with initial condition

$$P(0) = (x_1(0), x_2(0), \dots, x_{2N}(0))$$

is governed by the **equations of map T**

$$P(i+1) = T P(i) \quad , \quad i = 0, 1, 2, \dots$$

The evolution of an initial **deviation vector**

$$v(0) = (\delta x_1(0), \delta x_2(0), \dots, \delta x_{2N}(0))$$

is given by the corresponding **tangent map**

$$v(i+1) = \left. \frac{\partial T}{\partial P} \right|_i \cdot v(i) \quad , \quad i = 0, 1, 2, \dots$$

Example – 2D map

Equations of the map:

$$\begin{pmatrix} \mathbf{x}'_1 \\ \mathbf{x}'_2 \end{pmatrix} = \mathbf{T} \begin{pmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{pmatrix} \Rightarrow \begin{aligned} \mathbf{x}'_1 &= \mathbf{x}_1 + \mathbf{x}_2 \\ \mathbf{x}'_2 &= \mathbf{x}_2 - \mathbf{v} \sin(\mathbf{x}_1 + \mathbf{x}_2) \end{aligned} \quad (\text{mod } 2\pi)$$

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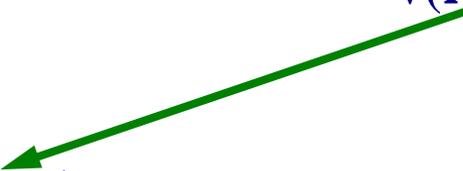
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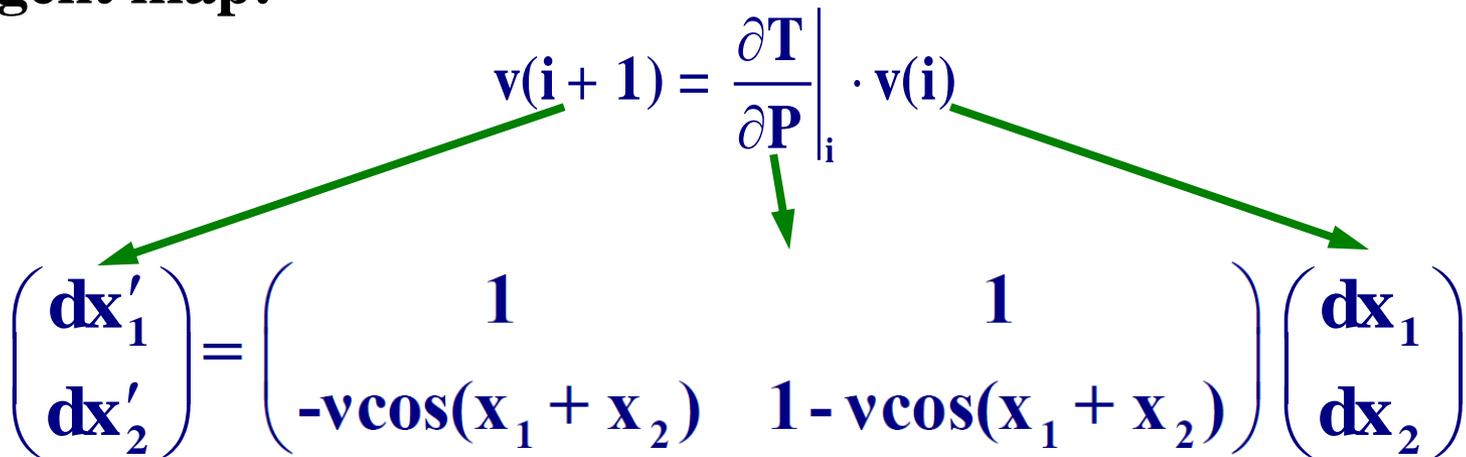
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$$\begin{pmatrix} d\mathbf{x}'_1 \\ d\mathbf{x}'_2 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ -v \cos(\mathbf{x}_1 + \mathbf{x}_2) & 1 - v \cos(\mathbf{x}_1 + \mathbf{x}_2) \end{pmatrix} \begin{pmatrix} d\mathbf{x}_1 \\ d\mathbf{x}_2 \end{pmatrix}$$

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:

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

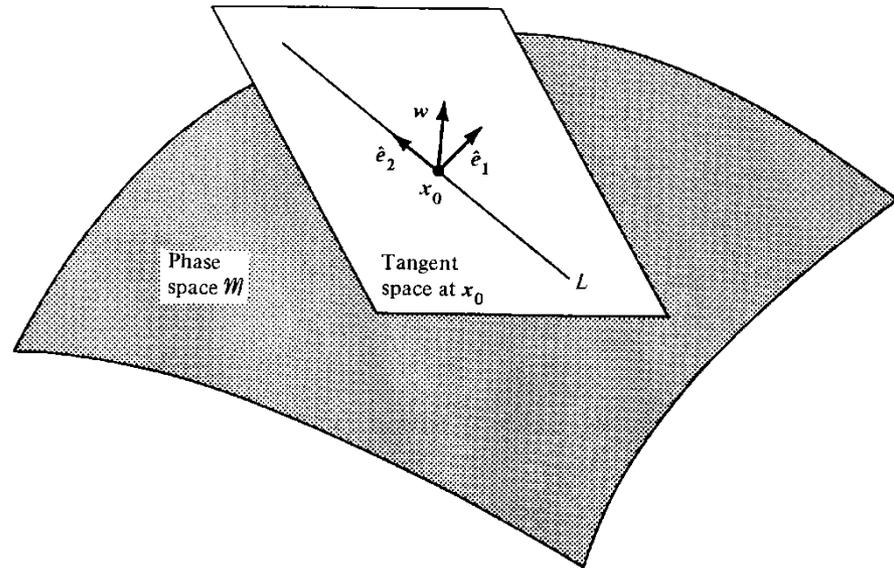
We commonly use the Euclidian norm and set $\|\mathbf{v}(0)\|=1$

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.

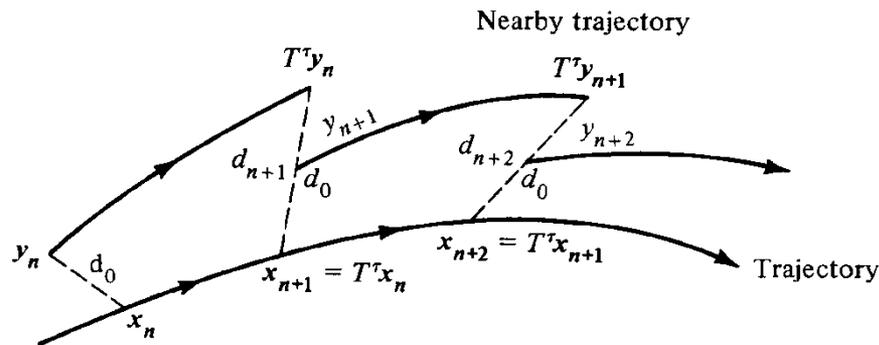


Figure 5.6. Numerical calculation of the maximal Liapunov characteristic exponent. Here $y = x + v$ and τ is a finite interval of time (after Benettin *et al.*, 1976).

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

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

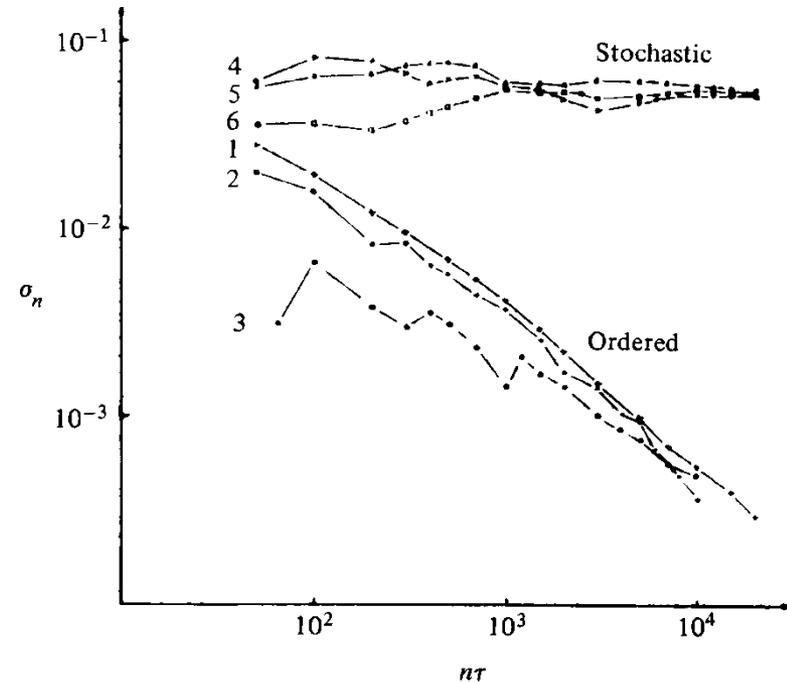
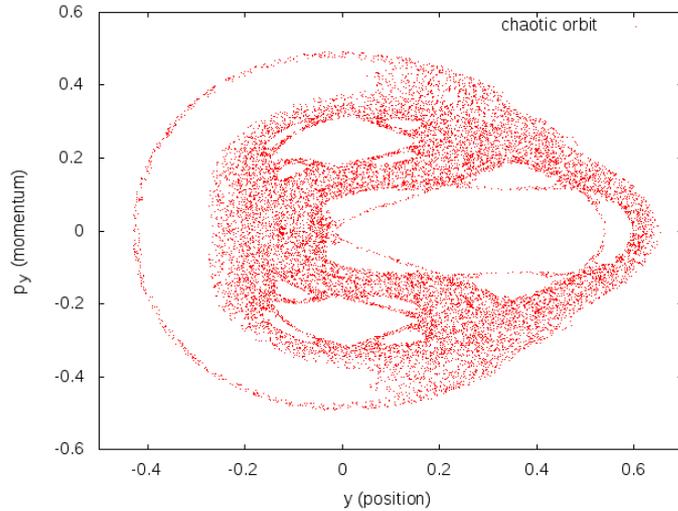


Figure 5.7. Behavior of σ_n at the intermediate energy $E = 0.125$ for initial points taken in the ordered (curves 1–3) or stochastic (curves 4–6) regions (after Benettin *et al.*, 1976).

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.

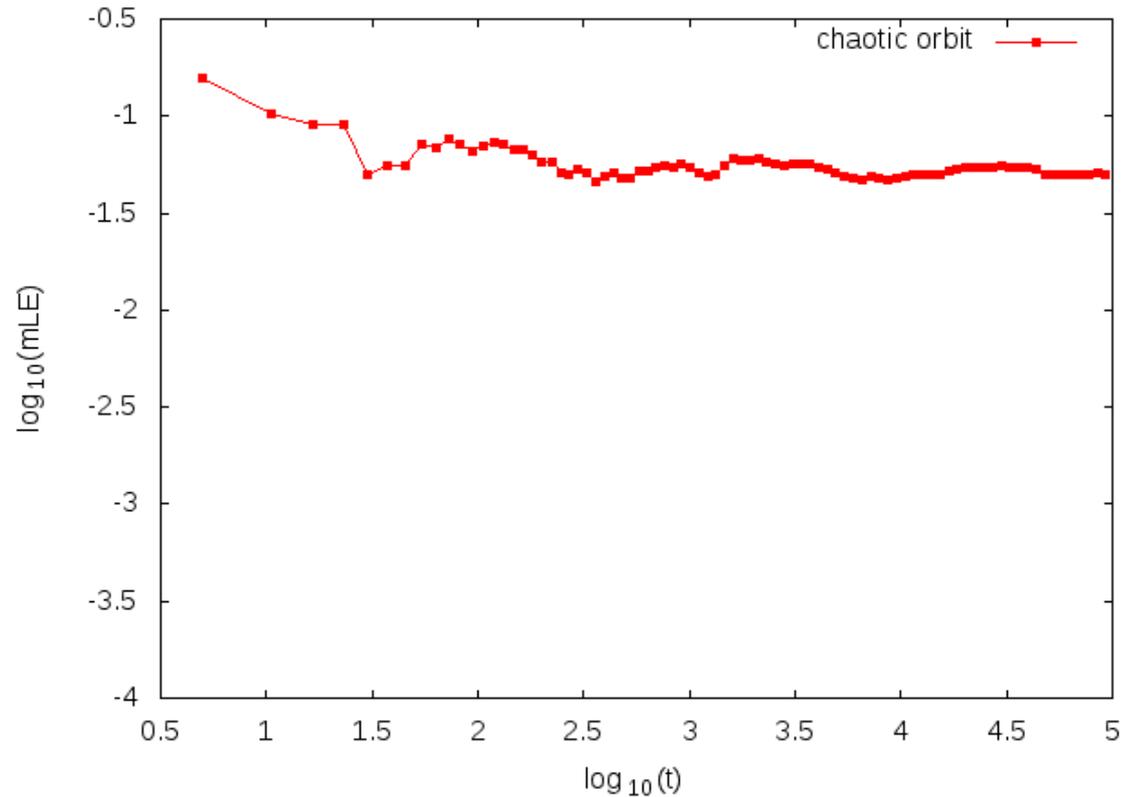
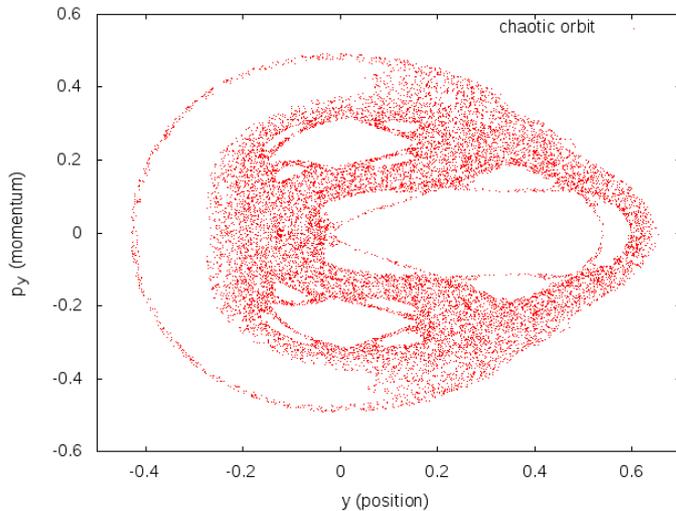
Maximum Lyapunov Exponent

Hénon-Heiles system: **Chaotic orbit**



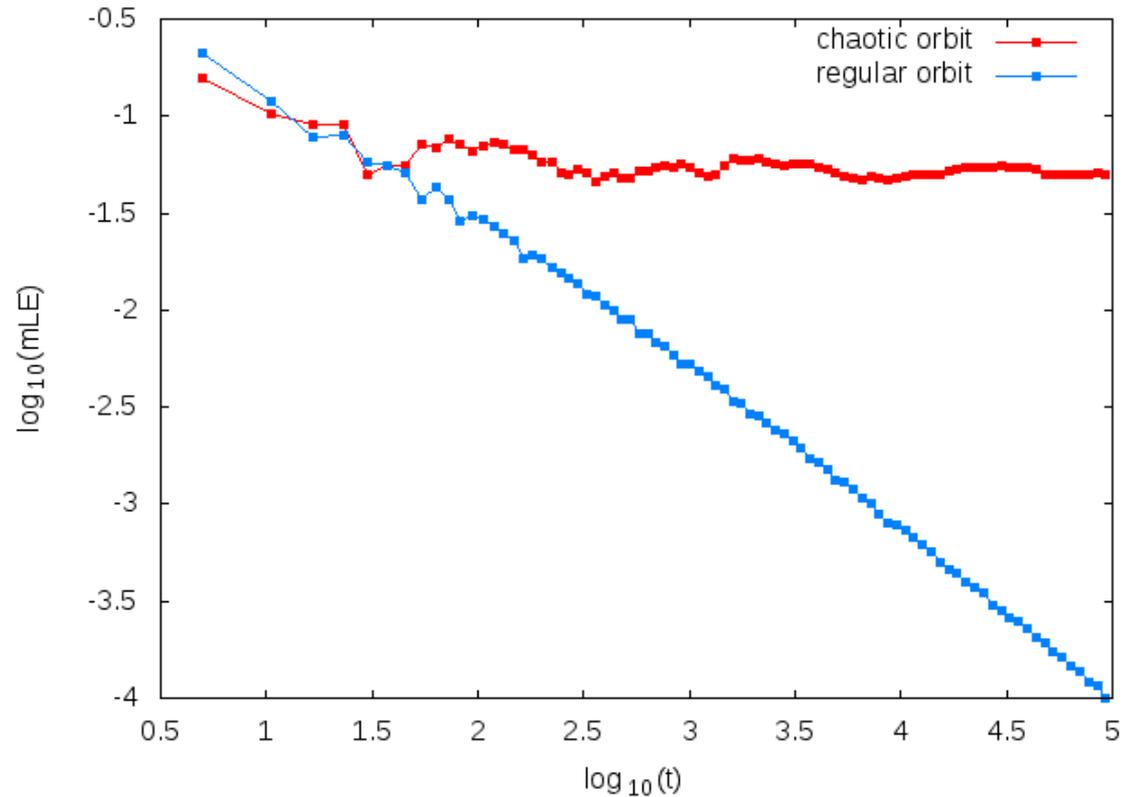
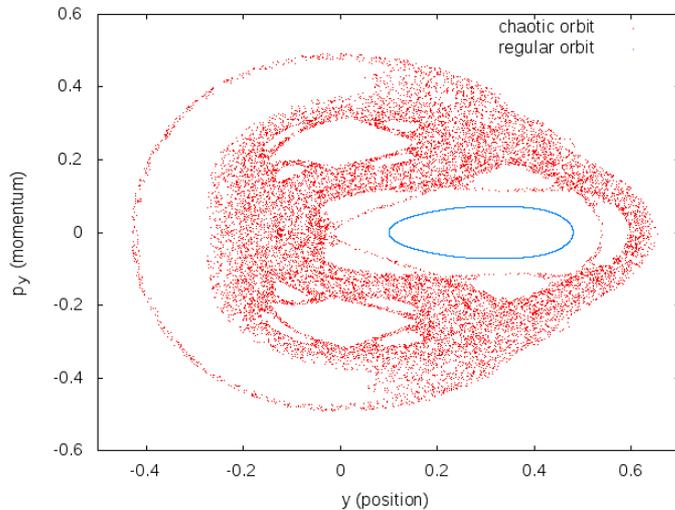
Maximum Lyapunov Exponent

Hénon-Heiles system: **Chaotic orbit**



Maximum Lyapunov Exponent

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



**The
Smaller ALignment Index
(SALI)
method**

Definition of the SALI

We follow the evolution in time of two different initial deviation vectors ($\mathbf{v}_1(0)$, $\mathbf{v}_2(0)$), and define the SALI (**Ch.S. 2001, J. Phys. A**) as:

$$\text{S A L I}(t) = \min \left\{ \left\| \hat{\mathbf{v}}_1(t) + \hat{\mathbf{v}}_2(t) \right\|, \left\| \hat{\mathbf{v}}_1(t) - \hat{\mathbf{v}}_2(t) \right\| \right\}$$

where

$$\hat{\mathbf{v}}_1(t) = \frac{\mathbf{v}_1(t)}{\|\mathbf{v}_1(t)\|}$$

When the two vectors become **collinear**

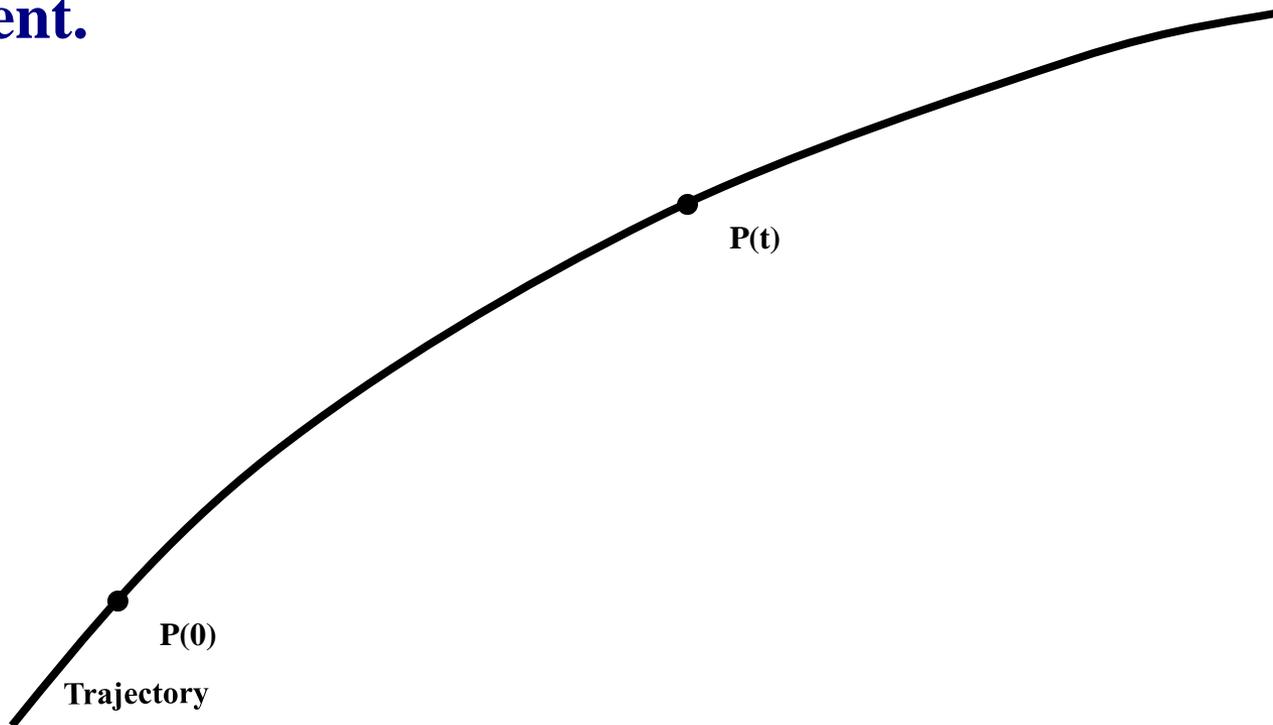
$$\text{SALI}(t) \rightarrow 0$$

Behavior of the SALI for chaotic motion

For chaotic orbits the two initially different deviation vectors tend to coincide with the direction defined by the maximum Lyapunov exponent.

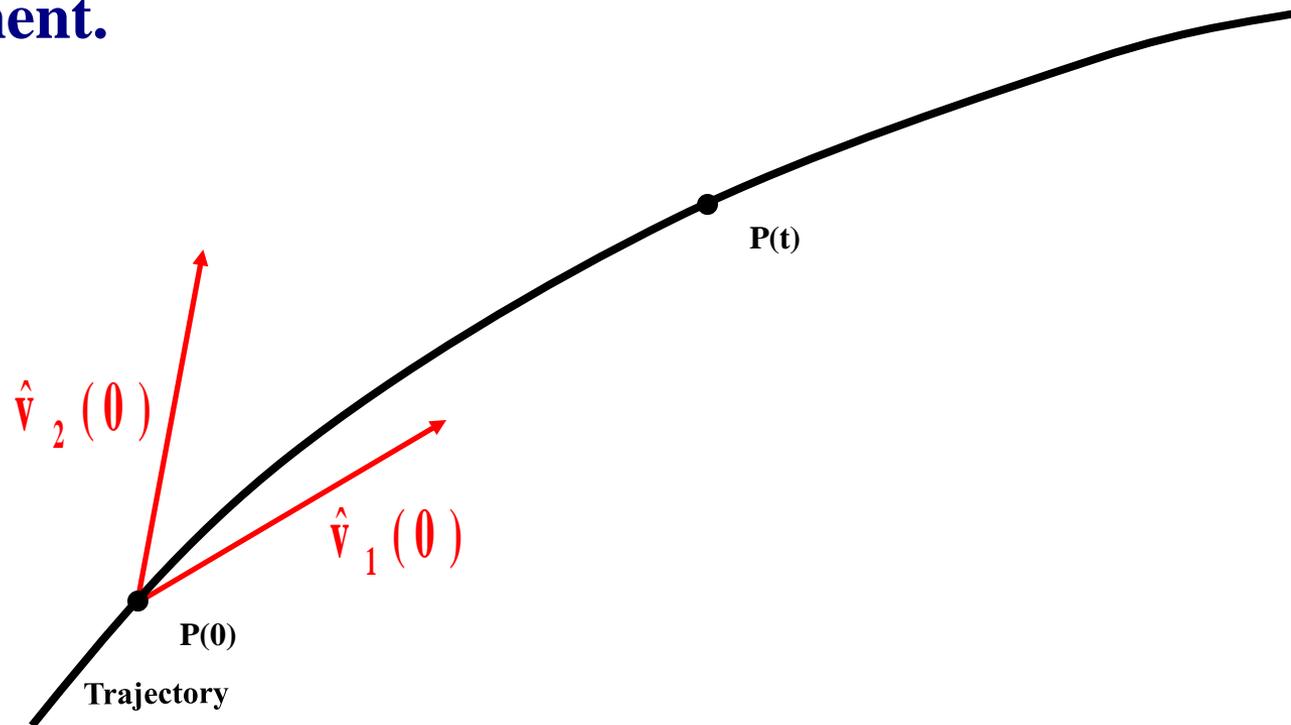
Behavior of the SALI for chaotic motion

For chaotic orbits the two initially different deviation vectors tend to coincide with the direction defined by the maximum Lyapunov exponent.



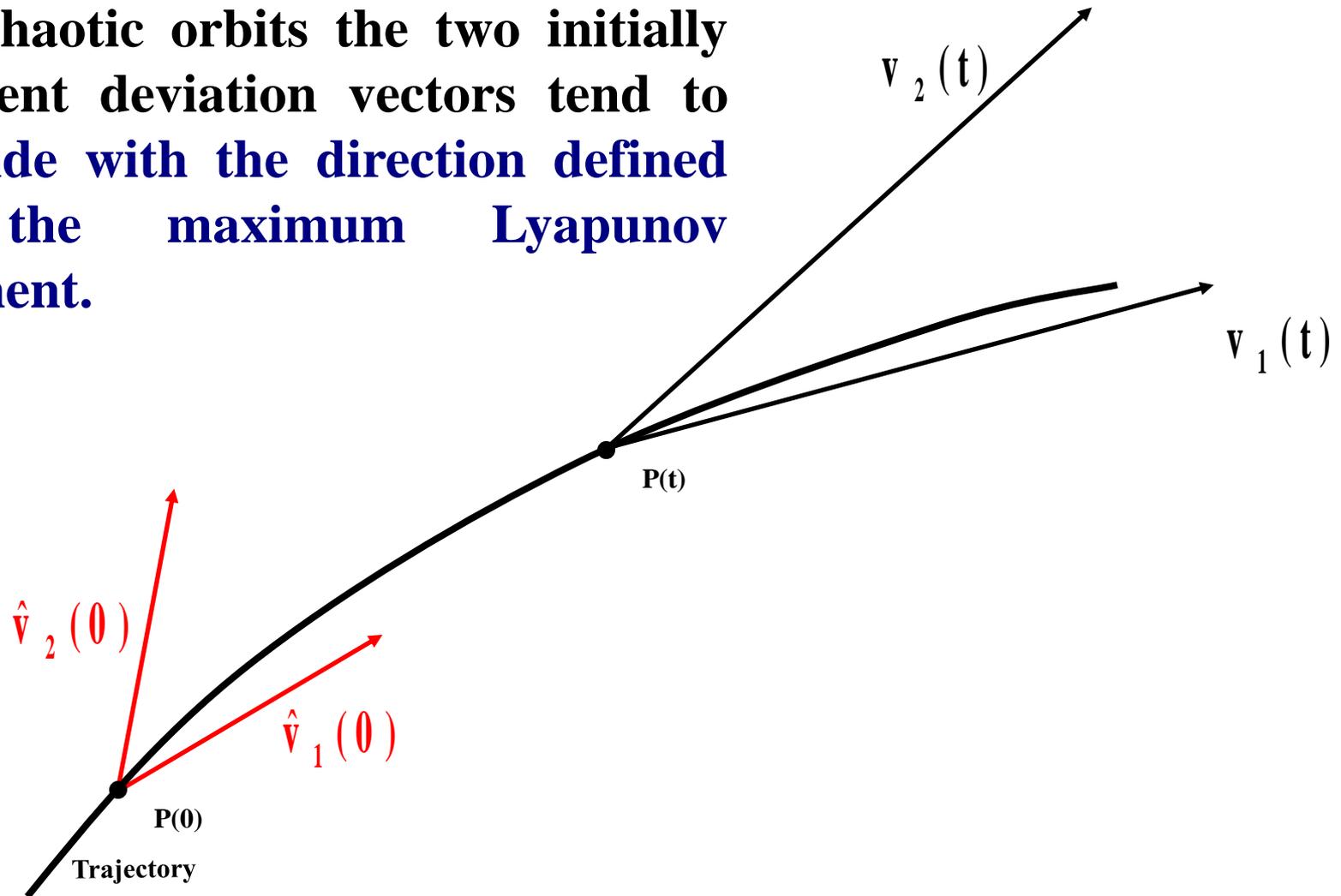
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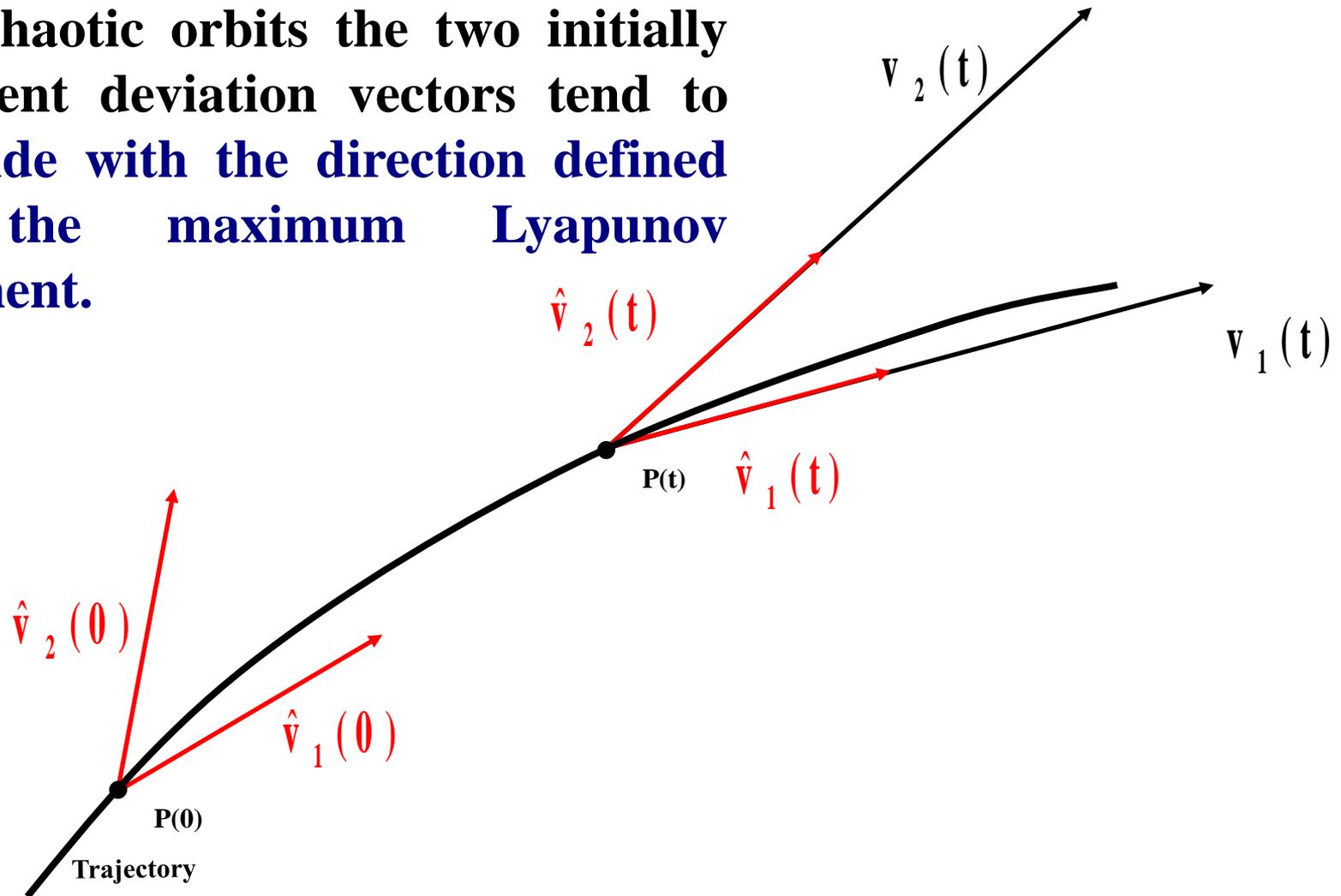
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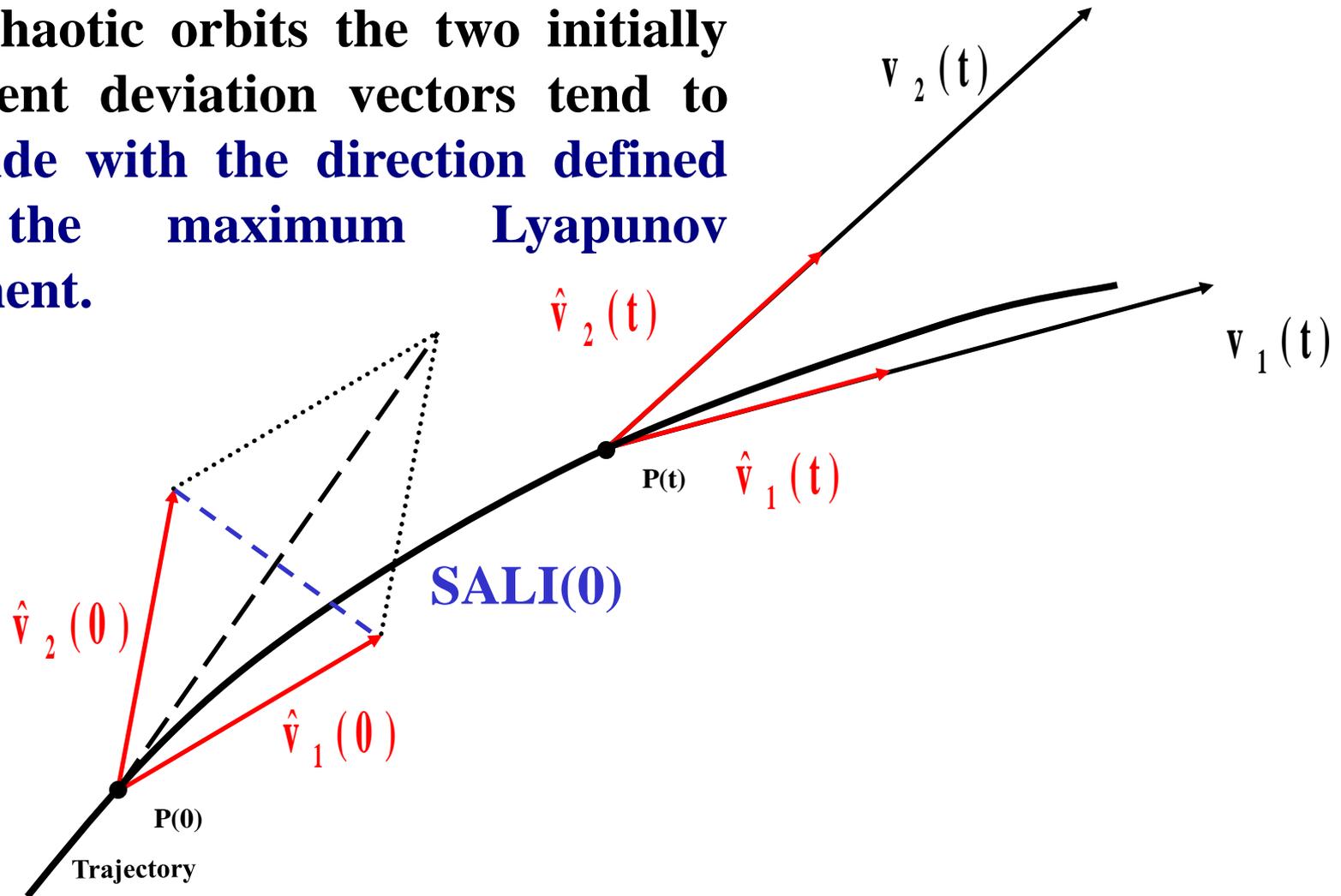
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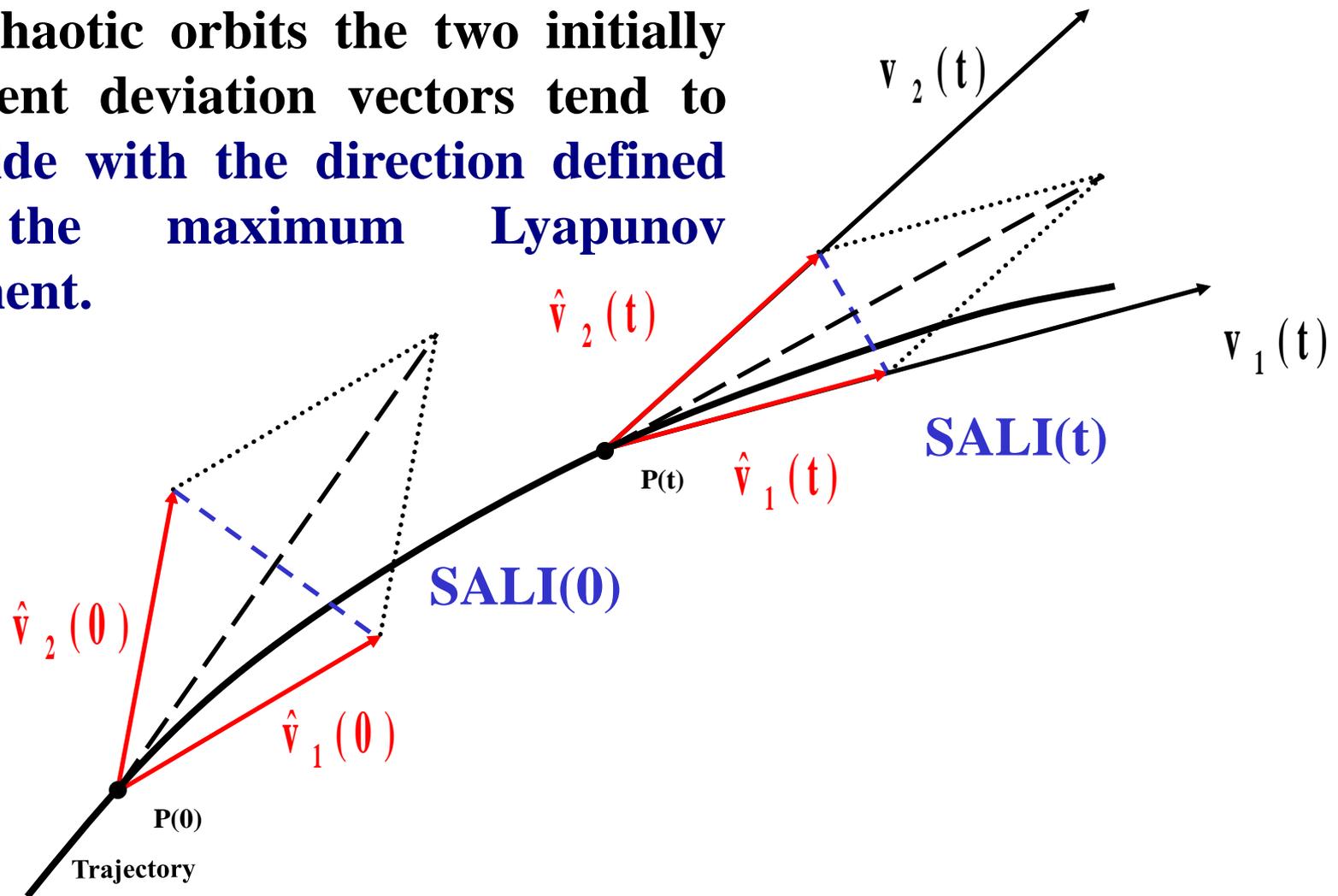
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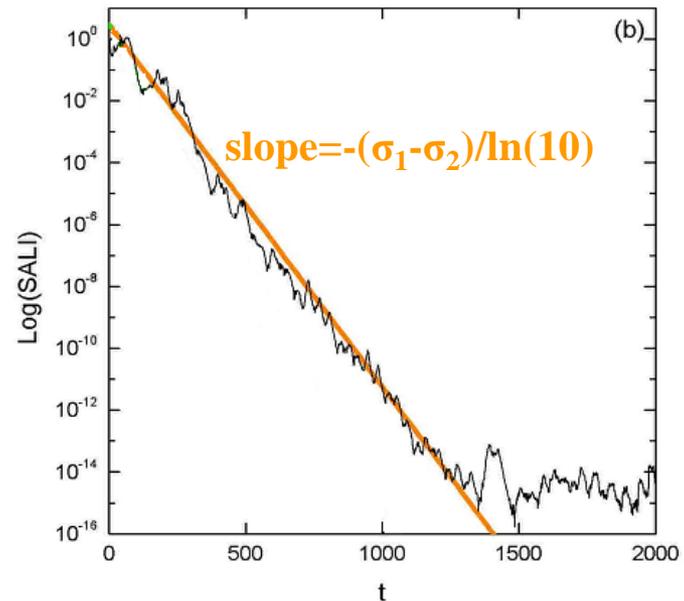
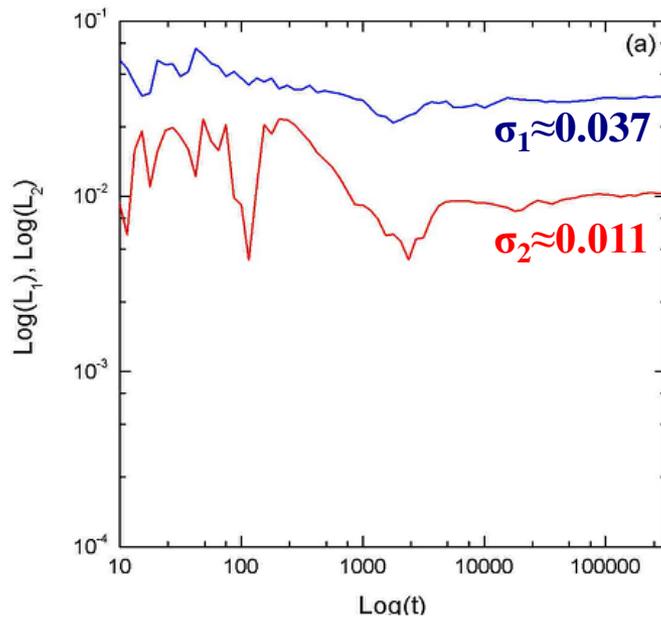


Behavior of the SALI for chaotic motion

We test the validity of the approximation $\text{SALI} \sim e^{-(\sigma_1 - \sigma_2)t}$ (Ch.S., Antonopoulos, Bountis, Vrahatis, 2004, J. Phys. A) for a chaotic orbit of the 3D Hamiltonian

$$H = \sum_{i=1}^3 \frac{\omega_i}{2} (q_i^2 + p_i^2) + q_1^2 q_2 + q_1^2 q_3$$

with $\omega_1=1$, $\omega_2=1.4142$, $\omega_3=1.7321$, $H=0.09$

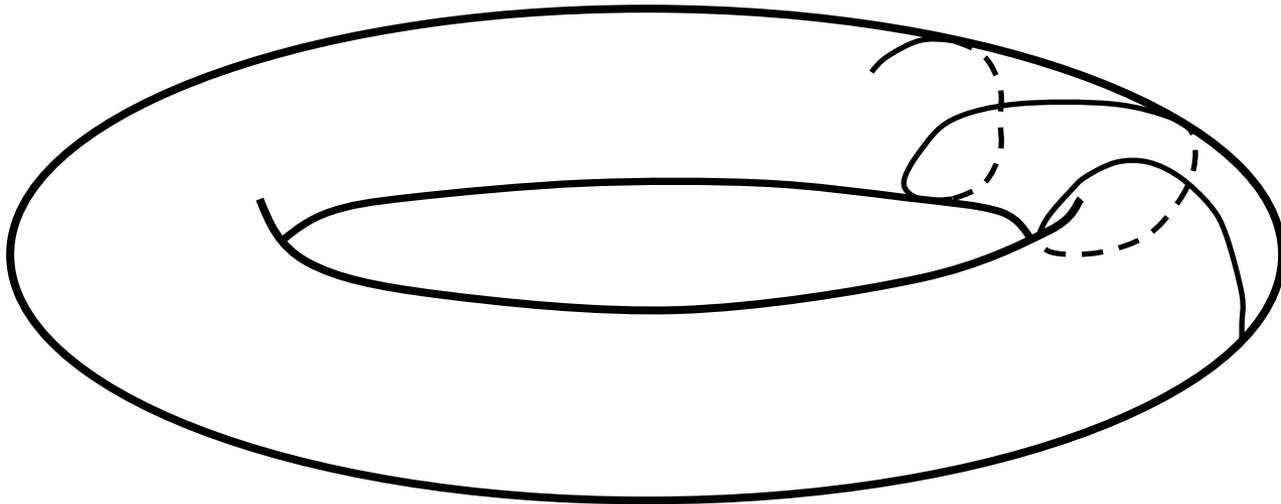


Behavior of the SALI for **regular motion**

Regular motion occurs on a torus and two different initial deviation vectors **become tangent to the torus, generally having different directions.**

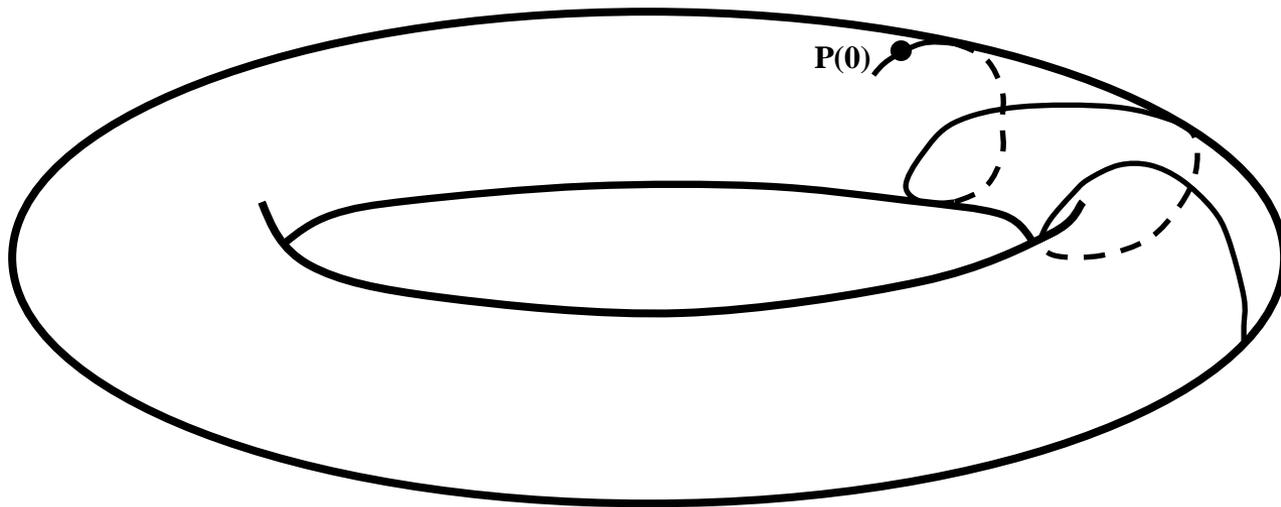
Behavior of the SALI for **regular motion**

Regular motion occurs on a torus and two different initial deviation vectors **become tangent to the torus, generally having different directions.**



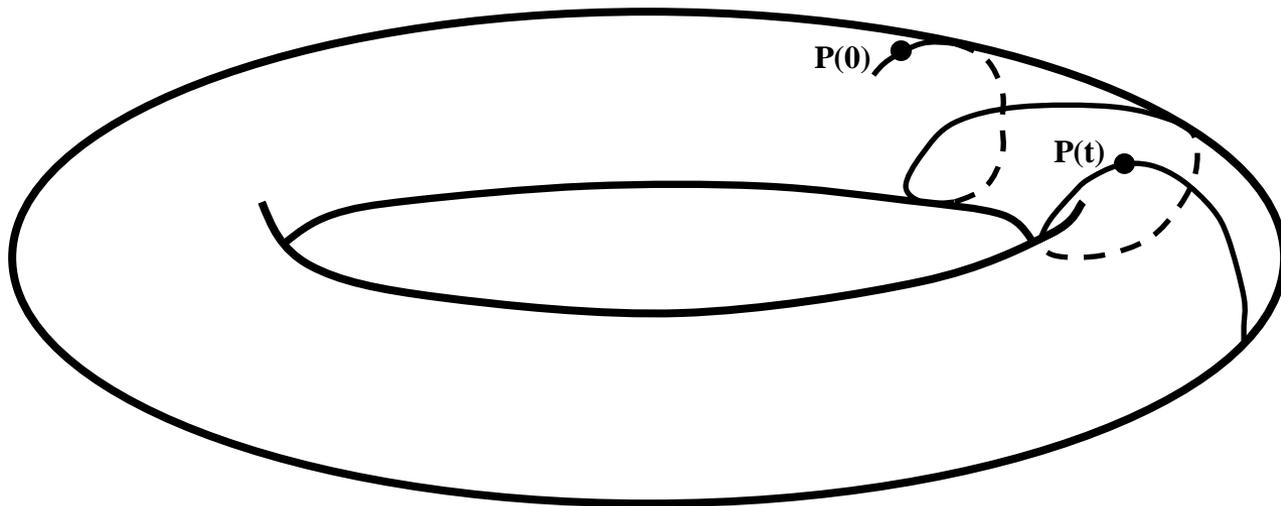
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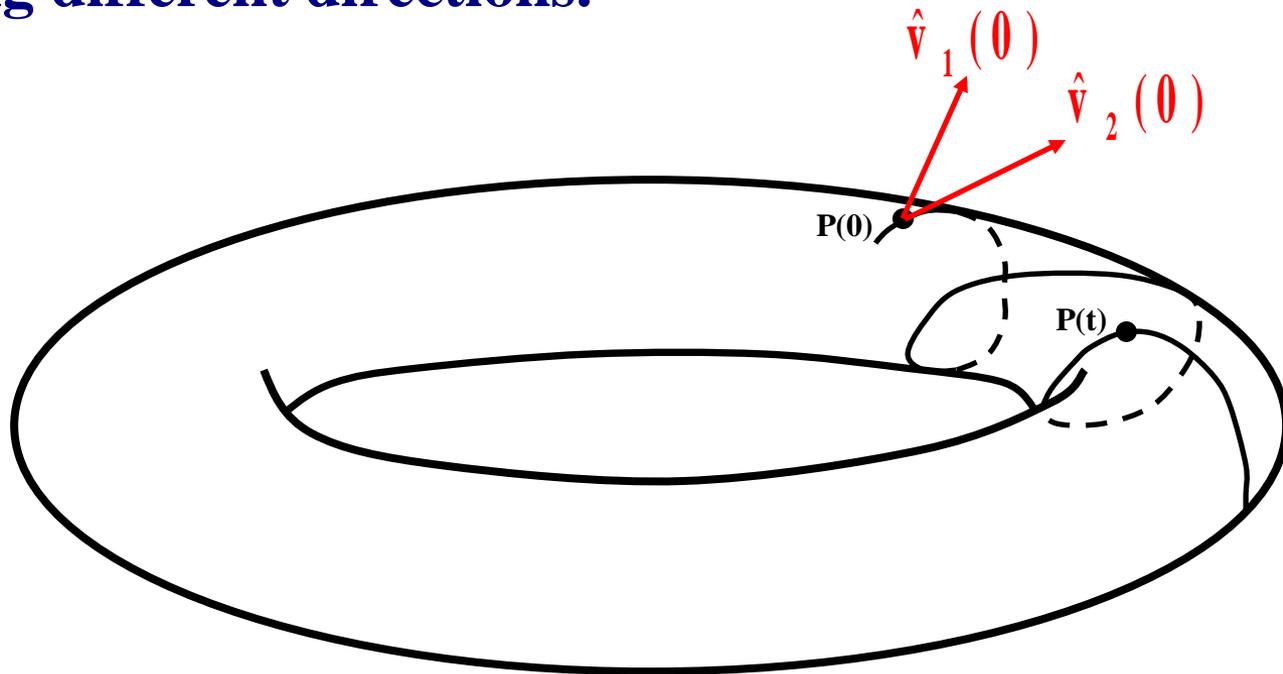
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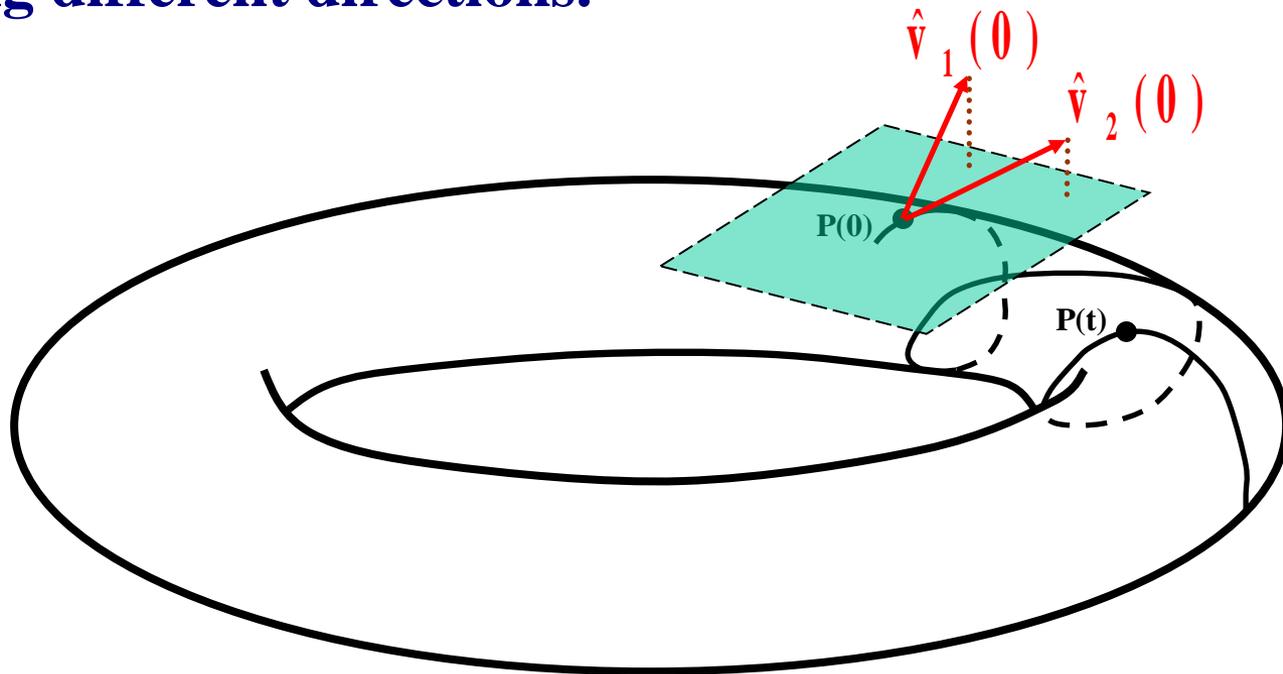
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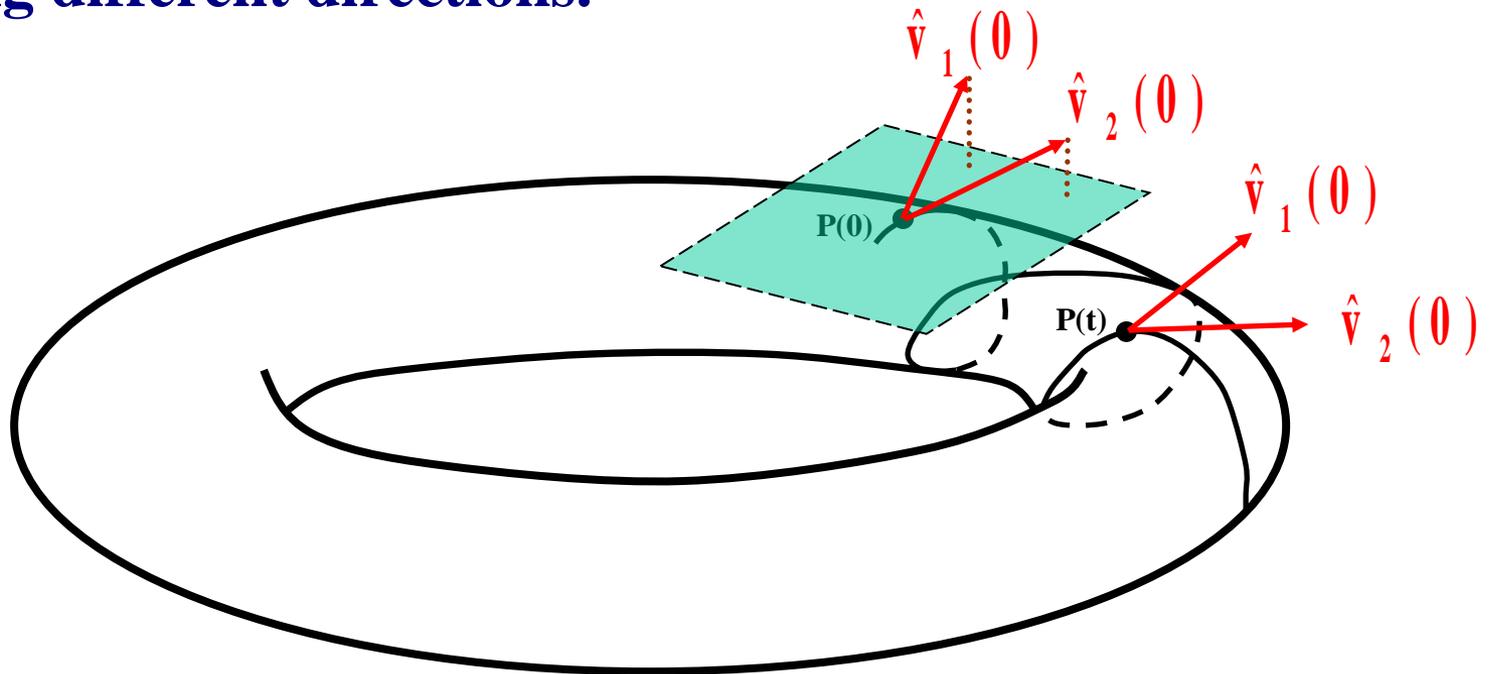
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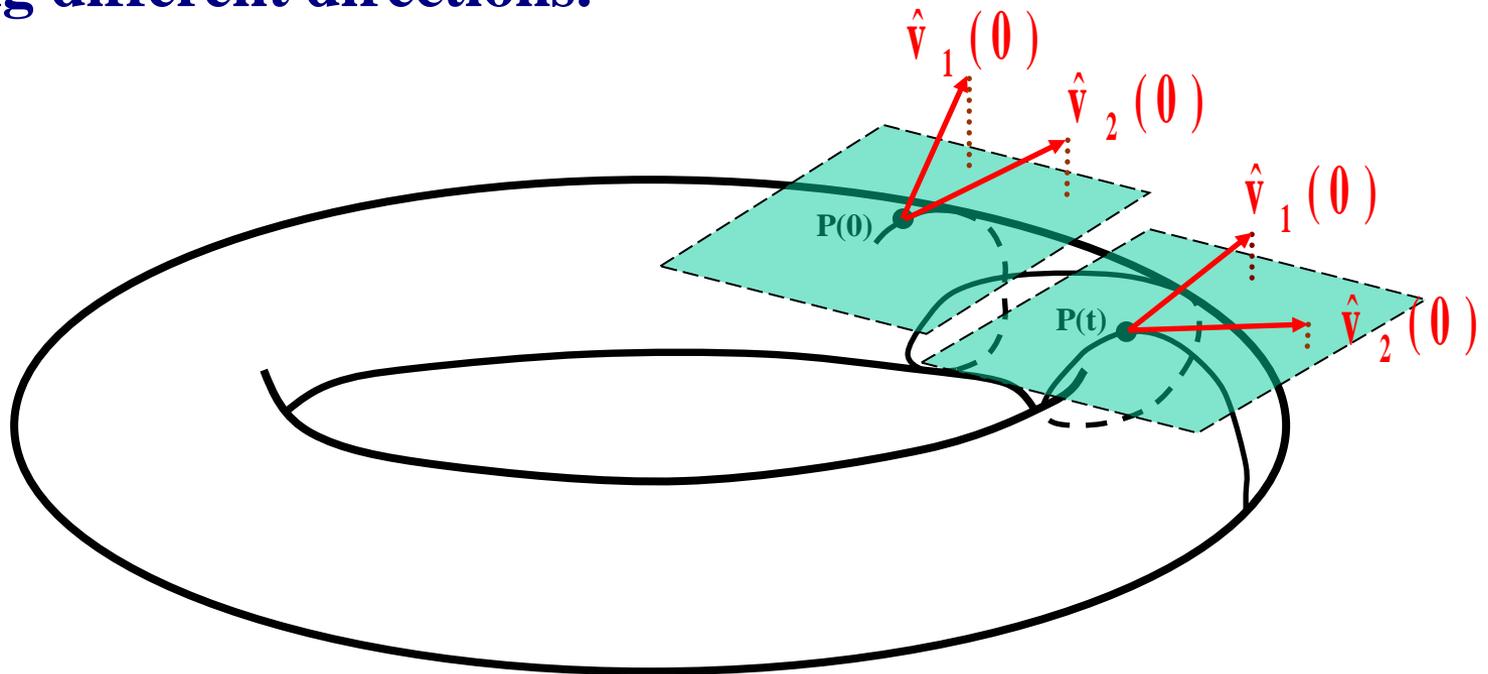
Behavior of the SALI for **regular motion**

Regular motion occurs on a torus and two different initial deviation vectors **become tangent to the torus**, generally having different directions.



Behavior of the SALI for regular motion

Regular motion occurs on a torus and two different initial deviation vectors become tangent to the torus, generally having different directions.



Applications – Hénon-Heiles system

As an example, we consider the 2D Hénon-Heiles system:

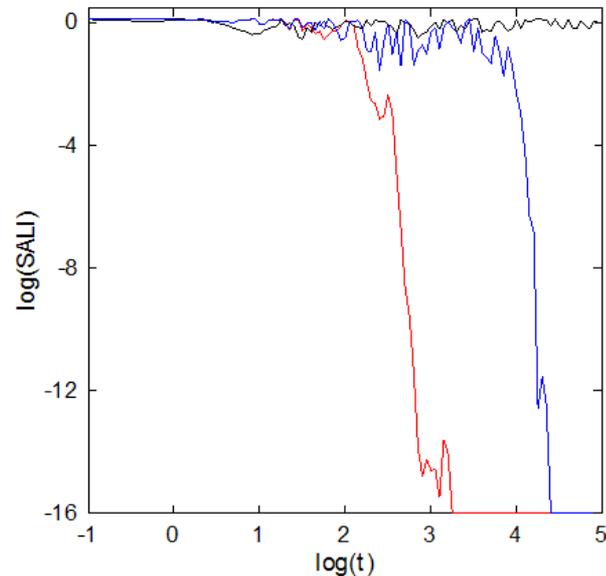
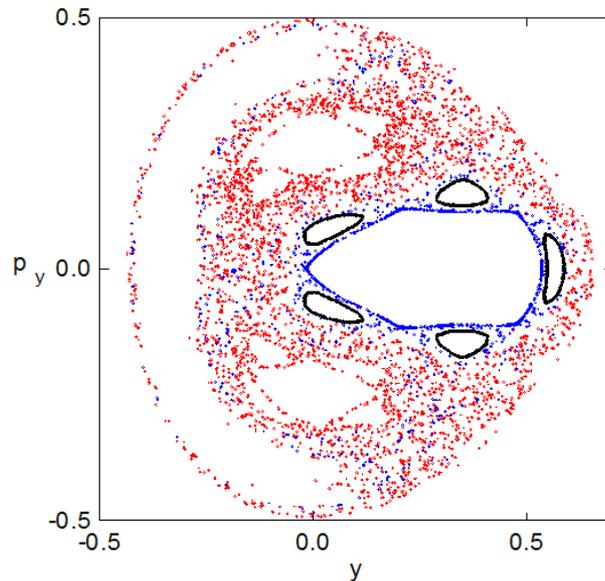
$$H_2 = \frac{1}{2}(p_x^2 + p_y^2) + \frac{1}{2}(x^2 + y^2) + x^2y - \frac{1}{3}y^3$$

For $E=1/8$ we consider the orbits with initial conditions:

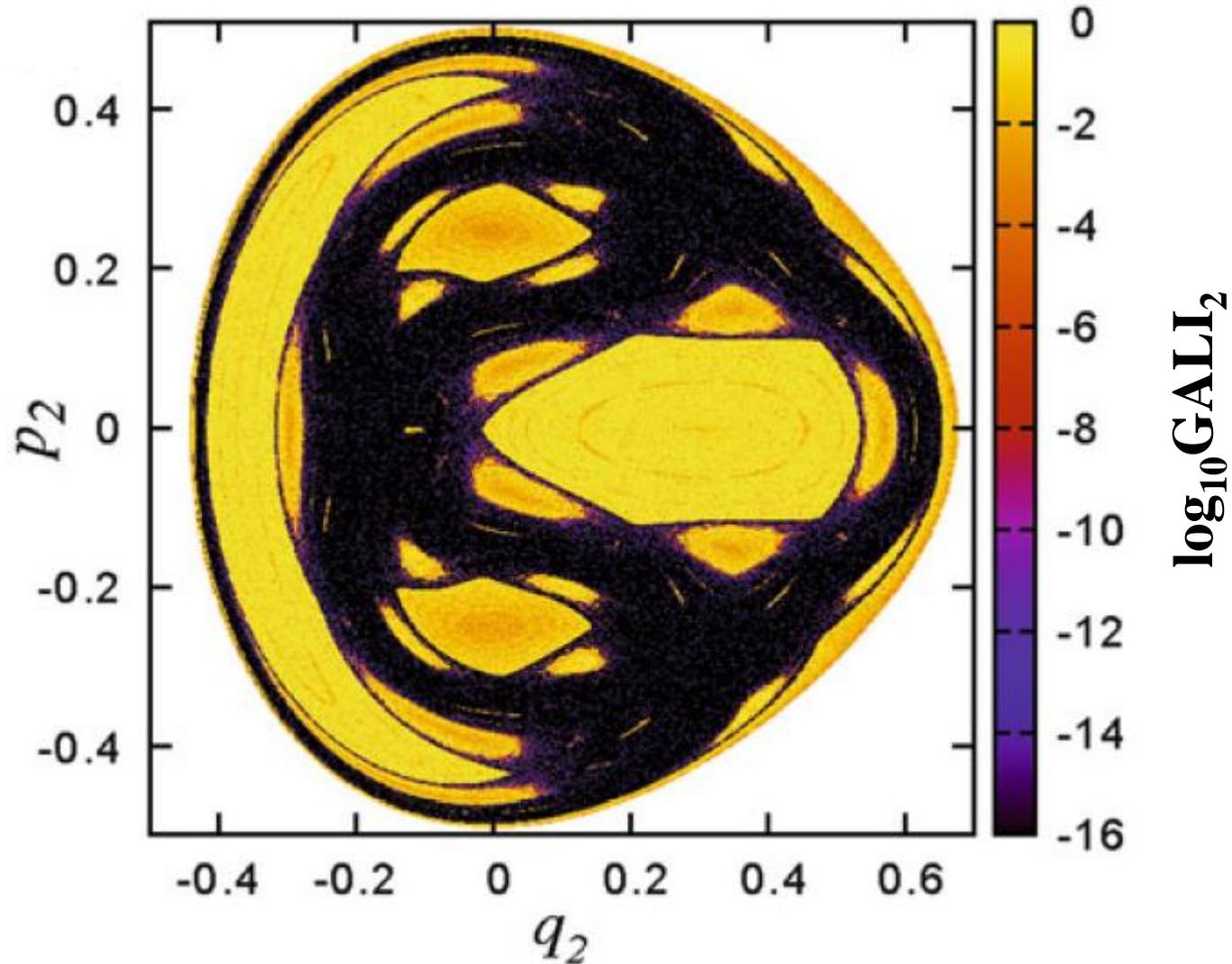
Regular orbit, $x=0$, $y=0.55$, $p_x=0.2417$, $p_y=0$

Chaotic orbit, $x=0$, $y=-0.016$, $p_x=0.49974$, $p_y=0$

Chaotic orbit, $x=0$, $y=-0.01344$, $p_x=0.49982$, $p_y=0$



Applications – Hénon-Heiles system



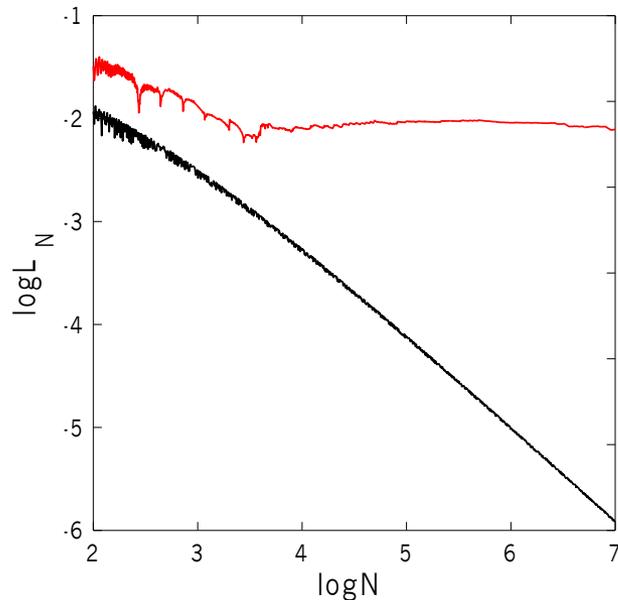
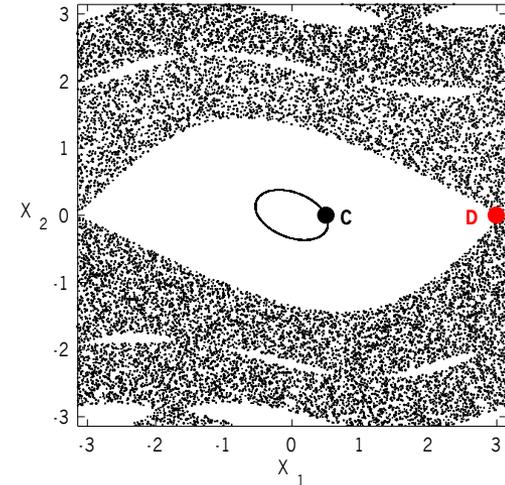
Applications – 4D map

$$\begin{aligned} \mathbf{x}'_1 &= \mathbf{x}_1 + \mathbf{x}_2 \\ \mathbf{x}'_2 &= \mathbf{x}_2 - \nu \sin(\mathbf{x}_1 + \mathbf{x}_2) - \mu [1 - \cos(\mathbf{x}_1 + \mathbf{x}_2 + \mathbf{x}_3 + \mathbf{x}_4)] \\ \mathbf{x}'_3 &= \mathbf{x}_3 + \mathbf{x}_4 \\ \mathbf{x}'_4 &= \mathbf{x}_4 - \kappa \sin(\mathbf{x}_3 + \mathbf{x}_4) - \mu [1 - \cos(\mathbf{x}_1 + \mathbf{x}_2 + \mathbf{x}_3 + \mathbf{x}_4)] \end{aligned} \quad (\text{mod } 2\pi)$$

For $\nu=0.5$, $\kappa=0.1$, $\mu=0.1$ we consider the orbits:

regular orbit C with initial conditions $x_1=0.5$, $x_2=0$, $x_3=0.5$, $x_4=0$.

chaotic orbit D with initial conditions $x_1=3$, $x_2=0$, $x_3=0.5$, $x_4=0$.



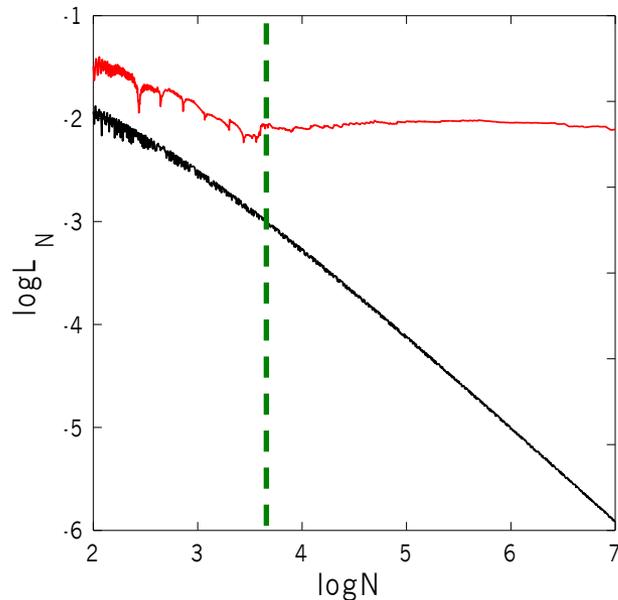
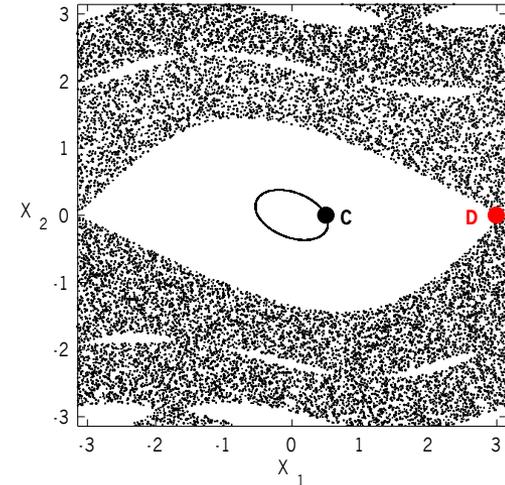
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 \mathbf{x}'_3 &= \mathbf{x}_3 + \mathbf{x}_4 \\
 \mathbf{x}'_4 &= \mathbf{x}_4 - \kappa \sin(\mathbf{x}_3 + \mathbf{x}_4) - \mu [1 - \cos(\mathbf{x}_1 + \mathbf{x}_2 + \mathbf{x}_3 + \mathbf{x}_4)]
 \end{aligned} \pmod{2\pi}$$

For $\nu=0.5$, $\kappa=0.1$, $\mu=0.1$ we consider the orbits:

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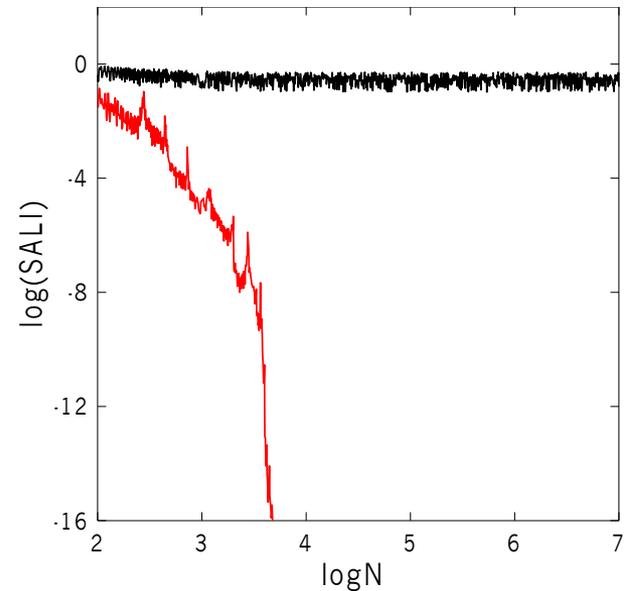
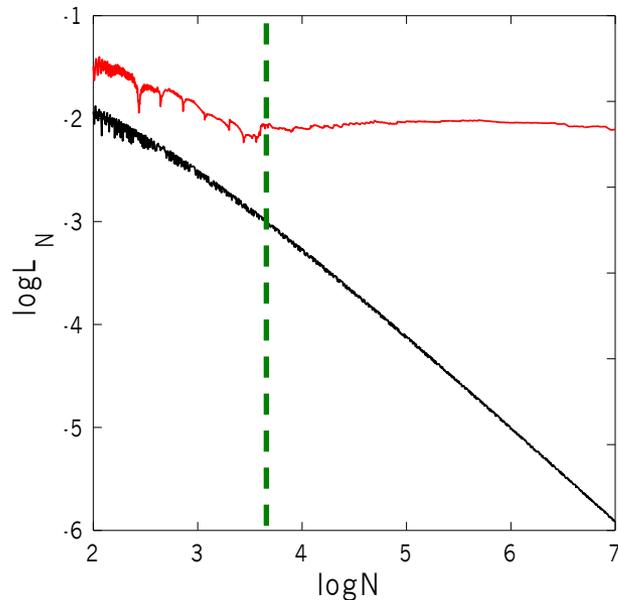
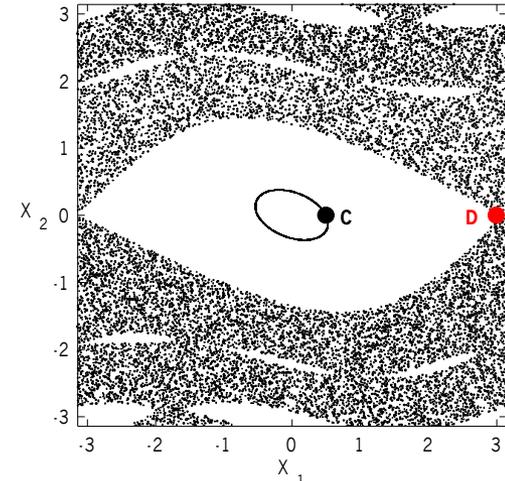
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 \mathbf{x}'_3 &= \mathbf{x}_3 + \mathbf{x}_4 \\
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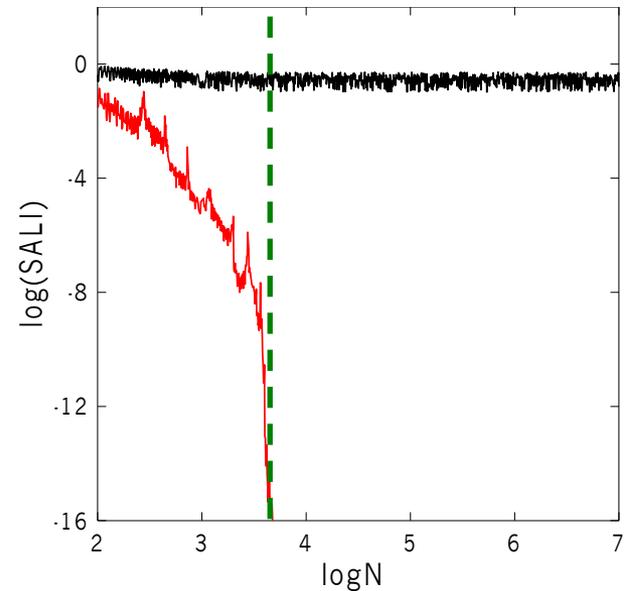
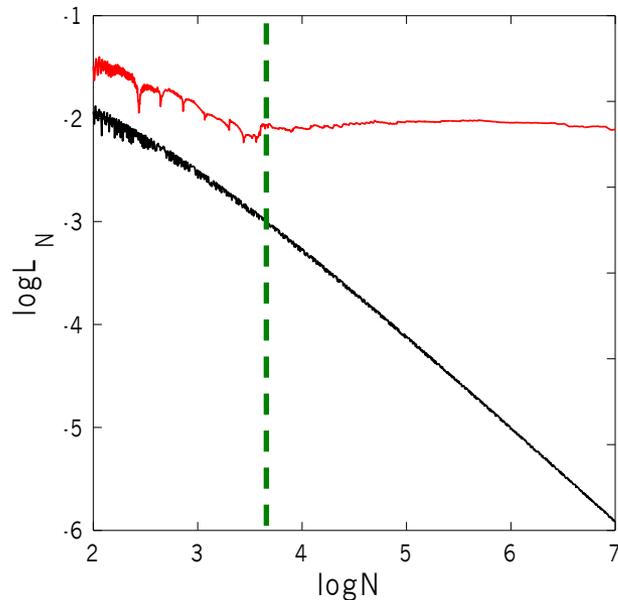
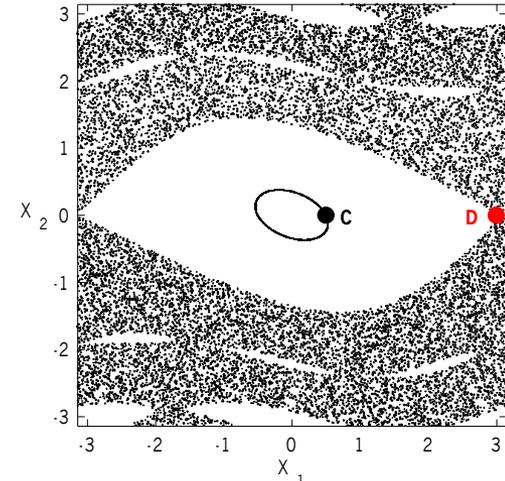
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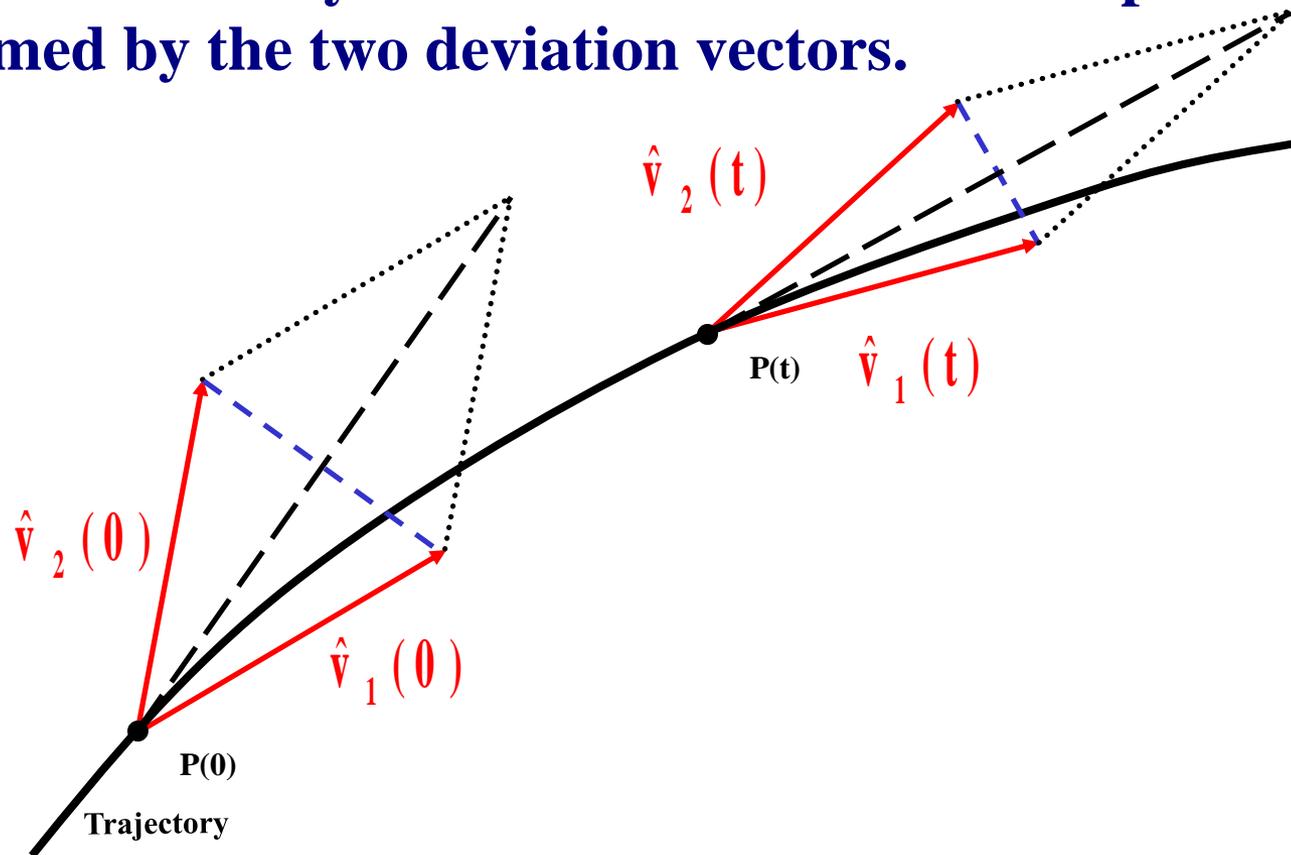
**The
Generalized ALignment Indices
(GALIs)
method**

Definition of the Generalized Alignment Index (GALI)

SALI effectively measures the 'area' of the parallelogram formed by the two deviation vectors.

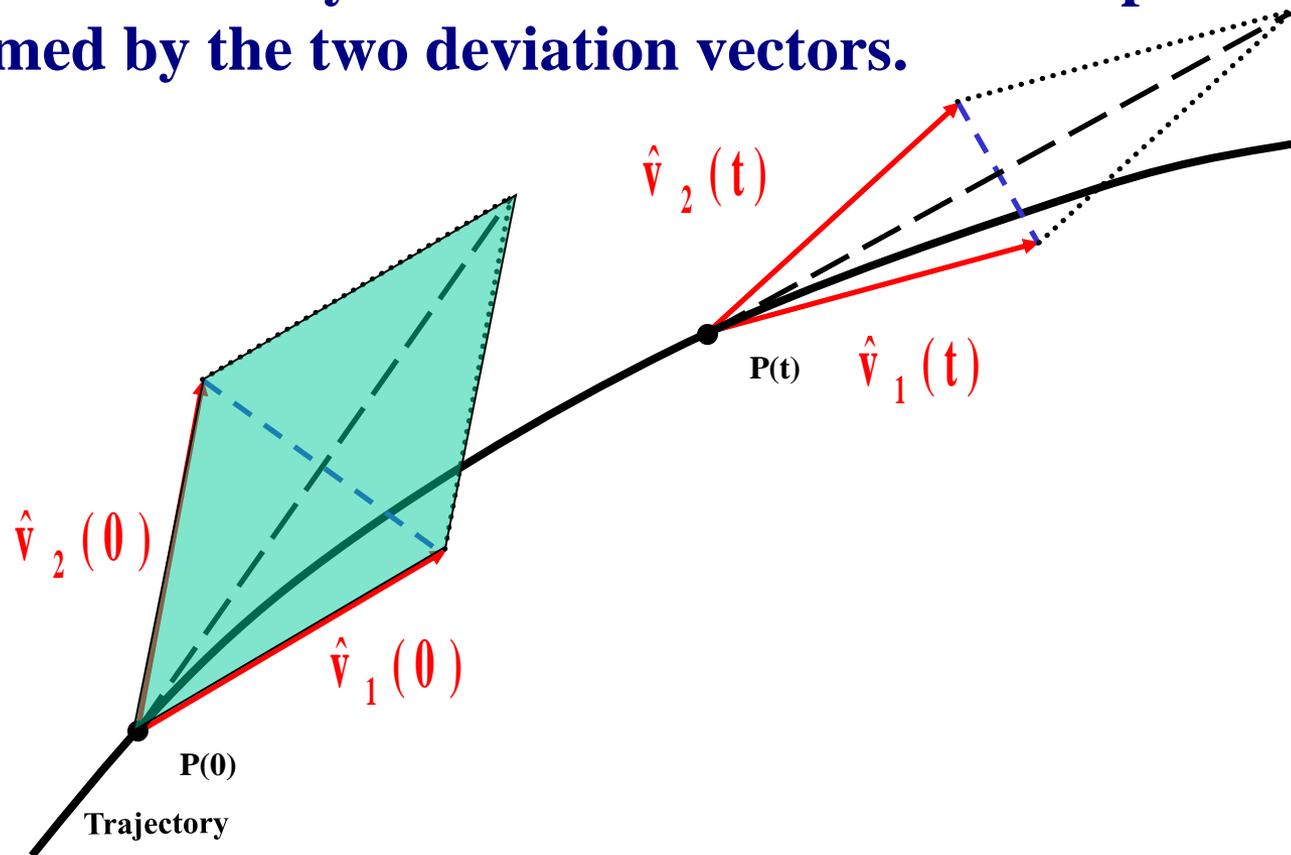
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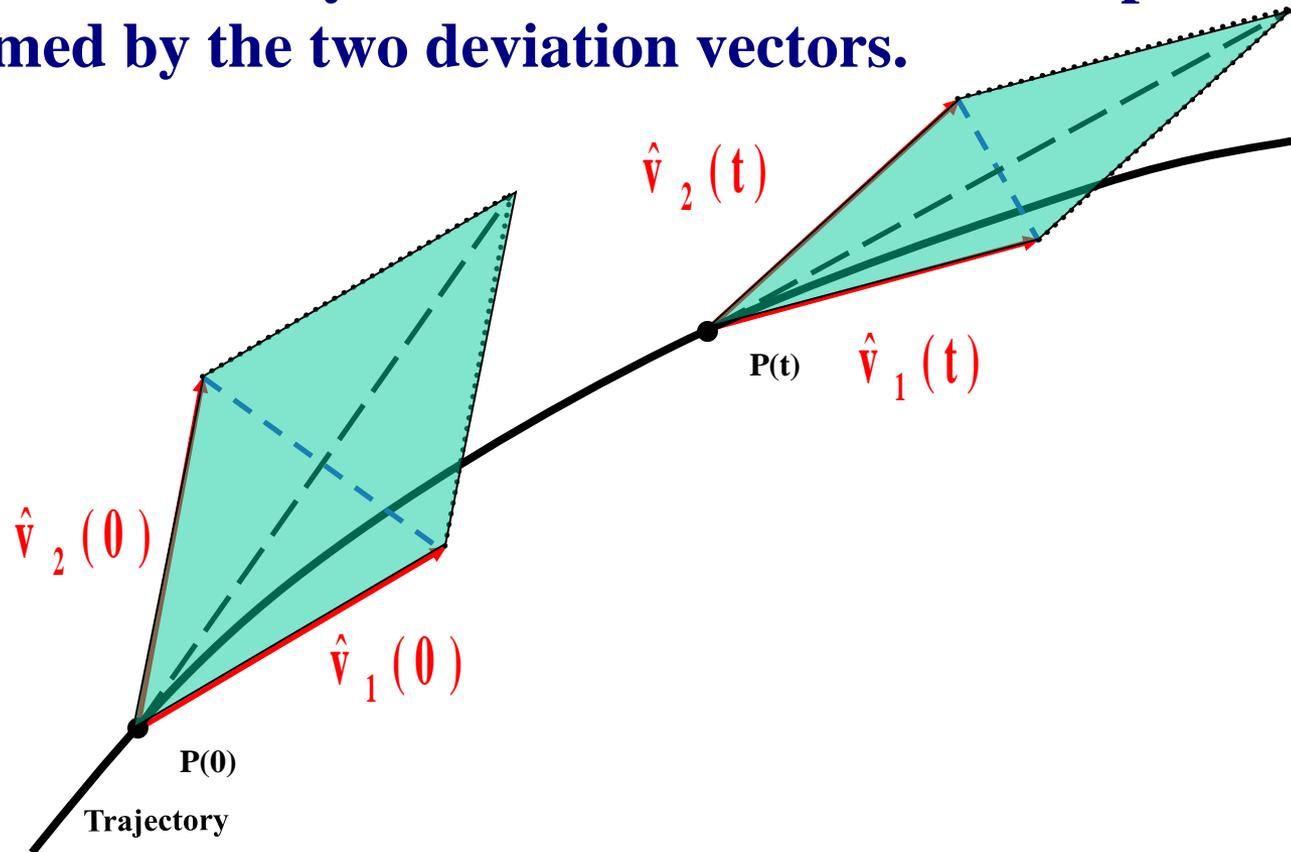
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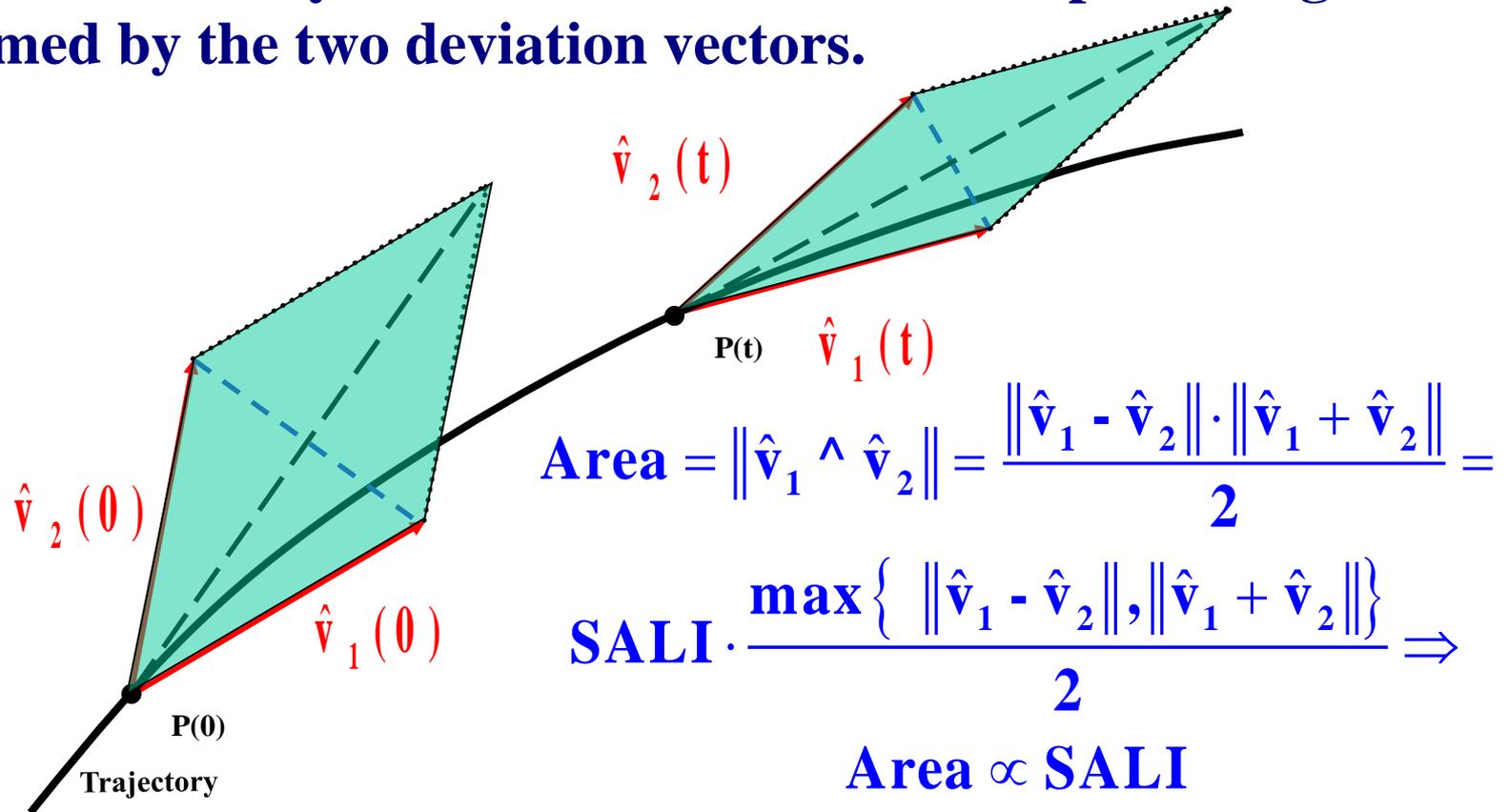
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SALI effectively measures the 'area' of the parallelogram formed by the two deviation vectors.



Definition of the Generalized Alignment Index (GALI)

SALI effectively measures the 'area' of the parallelogram formed by the two deviation vectors.



Definition of the GALI

In the case of an N degree of freedom Hamiltonian system or a $2N$ symplectic map we follow the evolution of

k deviation vectors with $2 \leq k \leq 2N$,

and define (Ch.S., Bountis, Antonopoulos, 2007, Physica D) the Generalized Alignment Index (GALI) of order k :

$$\text{G A L I}_k(t) = \left\| \hat{v}_1(t) \wedge \hat{v}_2(t) \wedge \dots \wedge \hat{v}_k(t) \right\|$$

where

$$\hat{v}_1(t) = \frac{\mathbf{v}_1(t)}{\|\mathbf{v}_1(t)\|}$$

Behavior of the $GALI_k$ for chaotic motion

$GALI_k$ ($2 \leq k \leq 2N$) tends exponentially to zero with exponents that involve the values of the first k largest Lyapunov exponents $\sigma_1, \sigma_2, \dots, \sigma_k$:

$$G A L I_k (t) \propto e^{-[(\sigma_1 - \sigma_2) + (\sigma_1 - \sigma_3) + \dots + (\sigma_1 - \sigma_k)]t}$$

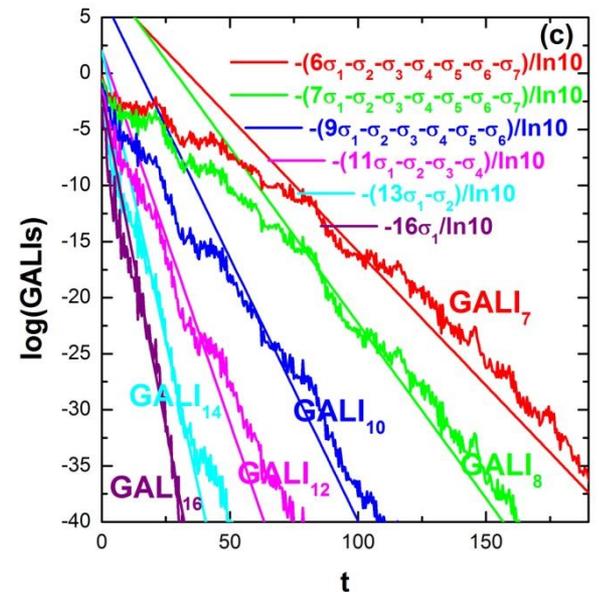
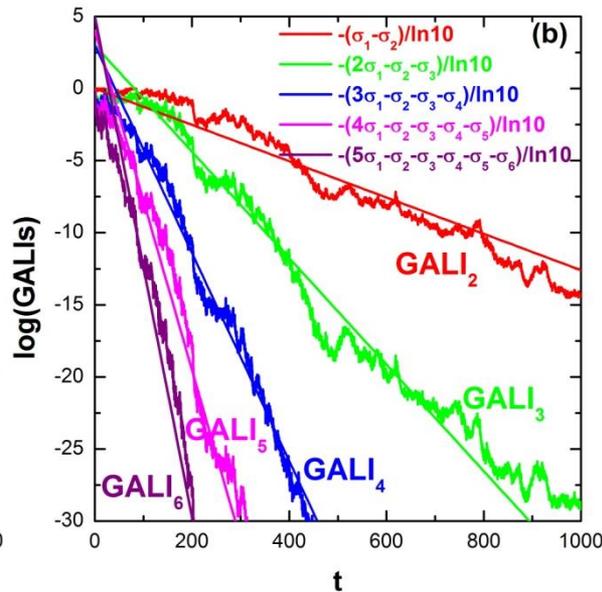
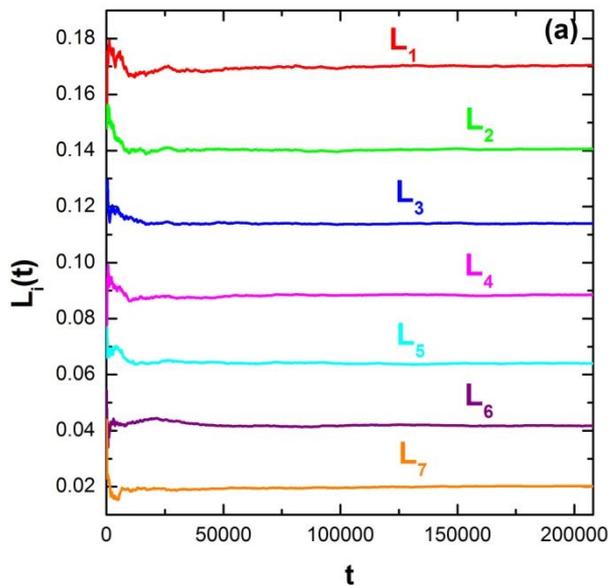
The above relation is valid even if some Lyapunov exponents are equal, or very close to each other.

Behavior of the $GALI_k$ for chaotic motion

N particles Fermi-Pasta-Ulam (FPU) system:

$$H = \frac{1}{2} \sum_{i=1}^N p_i^2 + \sum_{i=0}^N \left[\frac{1}{2} (q_{i+1} - q_i)^2 + \frac{\beta}{4} (q_{i+1} - q_i)^4 \right]$$

with fixed boundary conditions, $N=8$ and $\beta=1.5$.



Behavior of the $GALI_k$ for regular motion

If the motion occurs on an s -dimensional torus with $s \leq N$ then the behavior of $GALI_k$ is given by (Ch.S., Bountis, Antonopoulos, 2008, Eur. Phys. J. Sp. Top.):

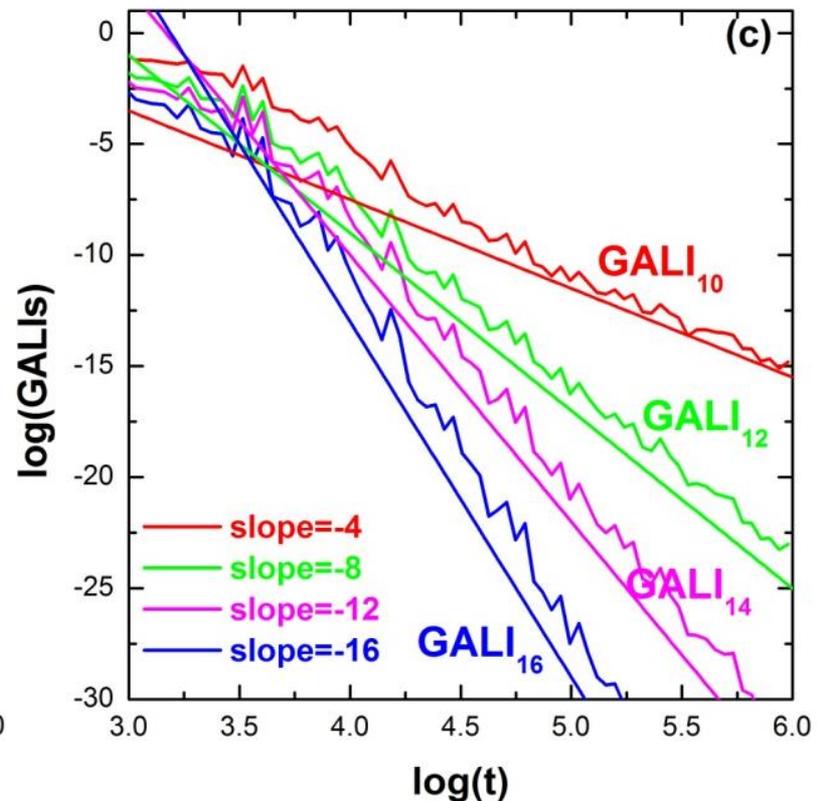
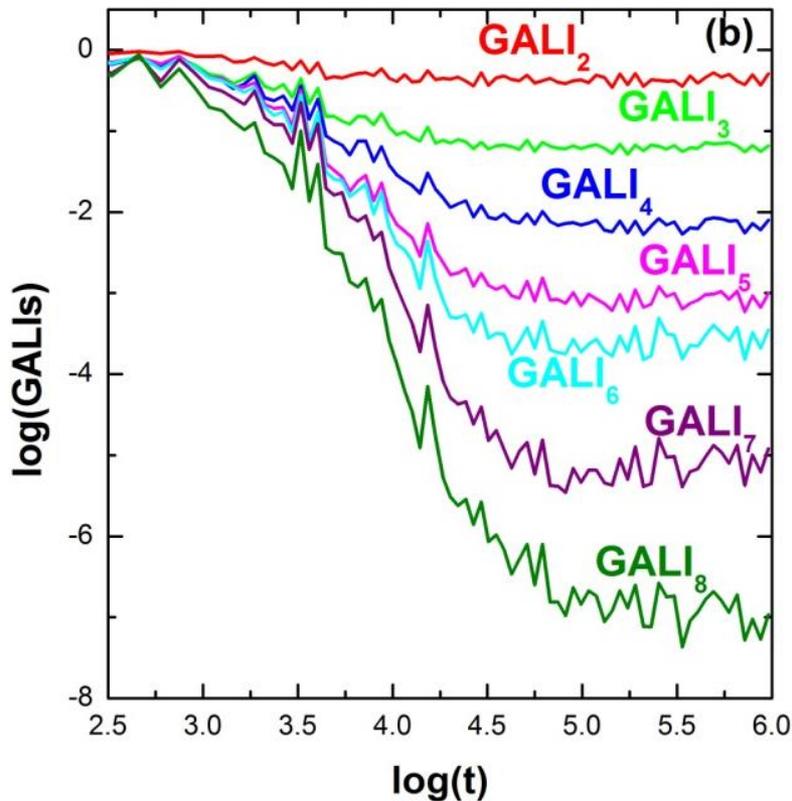
$$GALI_k(t) \propto \begin{cases} \text{constant} & \text{if } 2 \leq k \leq s \\ \frac{1}{t^{k-s}} & \text{if } s < k \leq 2N - s \\ \frac{1}{t^{2(k-N)}} & \text{if } 2N - s < k \leq 2N \end{cases}$$

while in the common case with $s=N$ we have :

$$GALI_k(t) \propto \begin{cases} \text{constant} & \text{if } 2 \leq k \leq N \\ \frac{1}{t^{2(k-N)}} & \text{if } N < k \leq 2N \end{cases}$$

Behavior of the $GALI_k$ for regular motion

N=8 FPU system



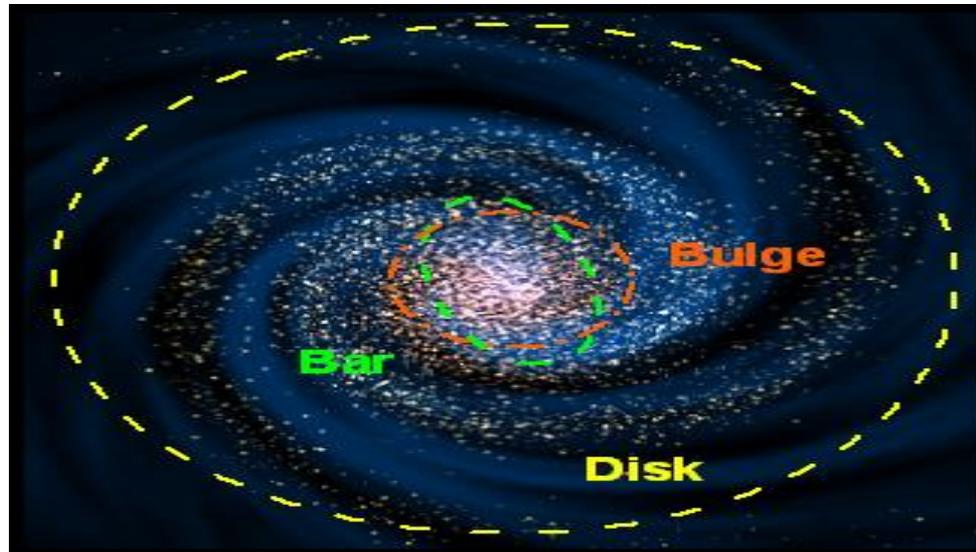
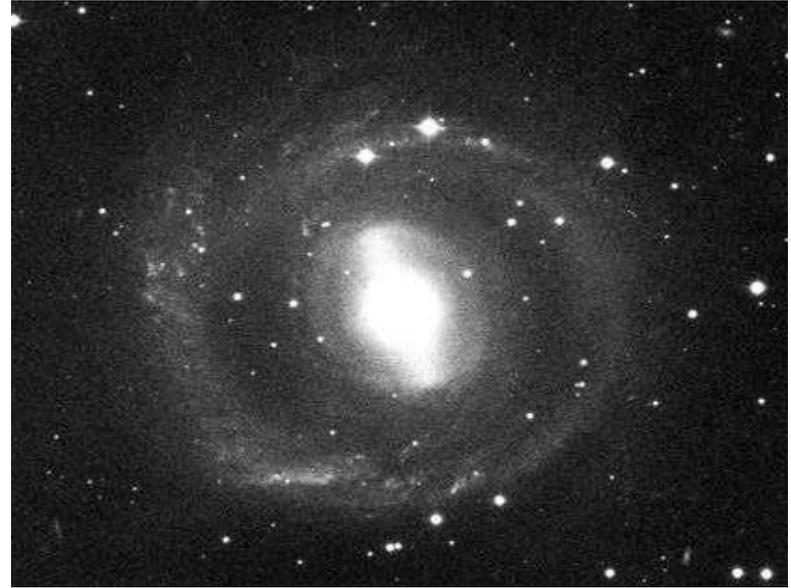
A time-dependent Hamiltonian system

Barred galaxies

NGC 1433



NGC 2217



Barred galaxy model

The 3D bar rotates around its short z -axis (x : long axis and y : intermediate). The Hamiltonian that describes the motion for this model is:

$$H = \frac{1}{2}(p_x^2 + p_y^2 + p_z^2) + V(x, y, z) - \Omega_b(xp_y - yp_x) \equiv \text{Energy}$$

This model consists of the superposition of potentials describing an **axisymmetric** part and a **bar** component of the galaxy (**Manos, Bountis, Ch.S., 2013, J. Phys. A**).

a) Axisymmetric component:

i) Plummer sphere:

$$V_{\text{sphere}}(x, y, z) = -\frac{GM_S}{\sqrt{x^2 + y^2 + z^2 + \epsilon_s^2}}$$

ii) Miyamoto–Nagai disc:

$$V_{\text{disc}}(x, y, z) = -\frac{GM_D}{\sqrt{x^2 + y^2 + (A + \sqrt{B^2 + z^2})^2}}$$

b) Bar component: $V_{\text{bar}}(x, y, z) = -\pi Gabc \frac{\rho_c}{n+1} \int_{\lambda}^{\infty} \frac{du}{\Delta(u)} (1 - m^2(u))^{n+1}$,

(Ferrers bar)

$$\rho_c = \frac{105}{32\pi} \frac{GM_B}{abc}$$

where $m^2(u) = \frac{x^2}{a^2 + u} + \frac{y^2}{b^2 + u} + \frac{z^2}{c^2 + u}$, $\Delta^2(u) = (a^2 + u)(b^2 + u)(c^2 + u)$,

n : positive integer ($n = 2$ for our model), λ : the unique positive solution of $m^2(\lambda) = 1$

Its density is:

$$\rho = \begin{cases} \rho_c (1 - m^2)^n, & \text{for } m \leq 1 \\ 0, & \text{for } m > 1 \end{cases}, \text{ where } m^2 = \frac{x^2}{a^2} + \frac{y^2}{b^2} + \frac{z^2}{c^2}, \text{ } a > b > c \text{ and } n = 2.$$

Time-dependent barred galaxy model

The 3D bar rotates around its short z -axis (x : long axis and y : intermediate). The Hamiltonian that describes the motion for this model is:

$$H = \frac{1}{2}(p_x^2 + p_y^2 + p_z^2) + V(x, y, z, t) - \Omega_b(xp_y - yp_x) \equiv \text{Energy}$$

This model consists of the superposition of potentials describing an **axisymmetric** part and a **bar** component of the galaxy (Manos, Bountis, Ch.S., 2013, J. Phys. A).

a) Axisymmetric component:

$$M_S + M_B(t) + M_D(t) = 1, \text{ with } M_B(t) = M_B(0) + \alpha t$$

i) Plummer sphere:

$$V_{sphere}(x, y, z) = -\frac{GM_S}{\sqrt{x^2 + y^2 + z^2 + \epsilon_s^2}}$$

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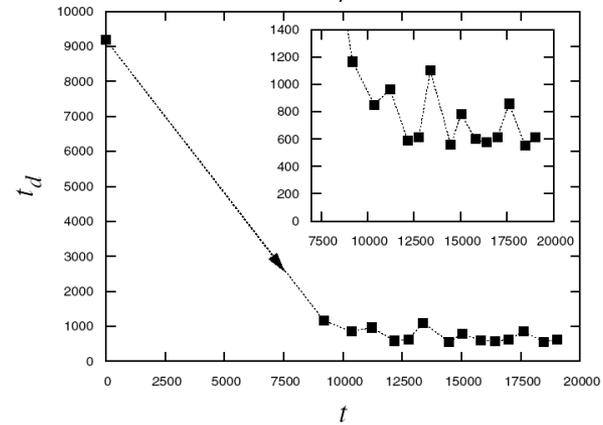
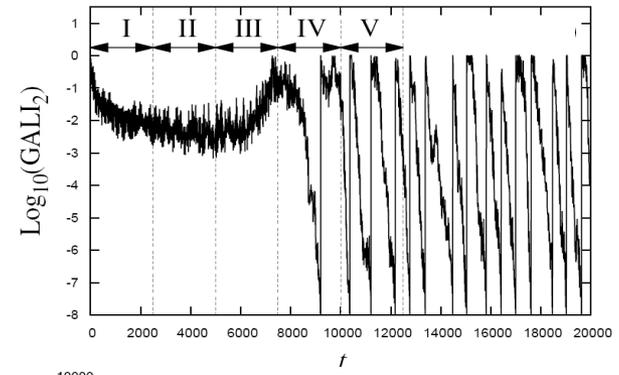
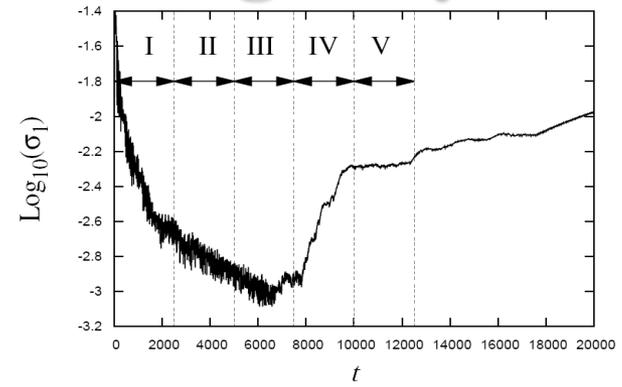
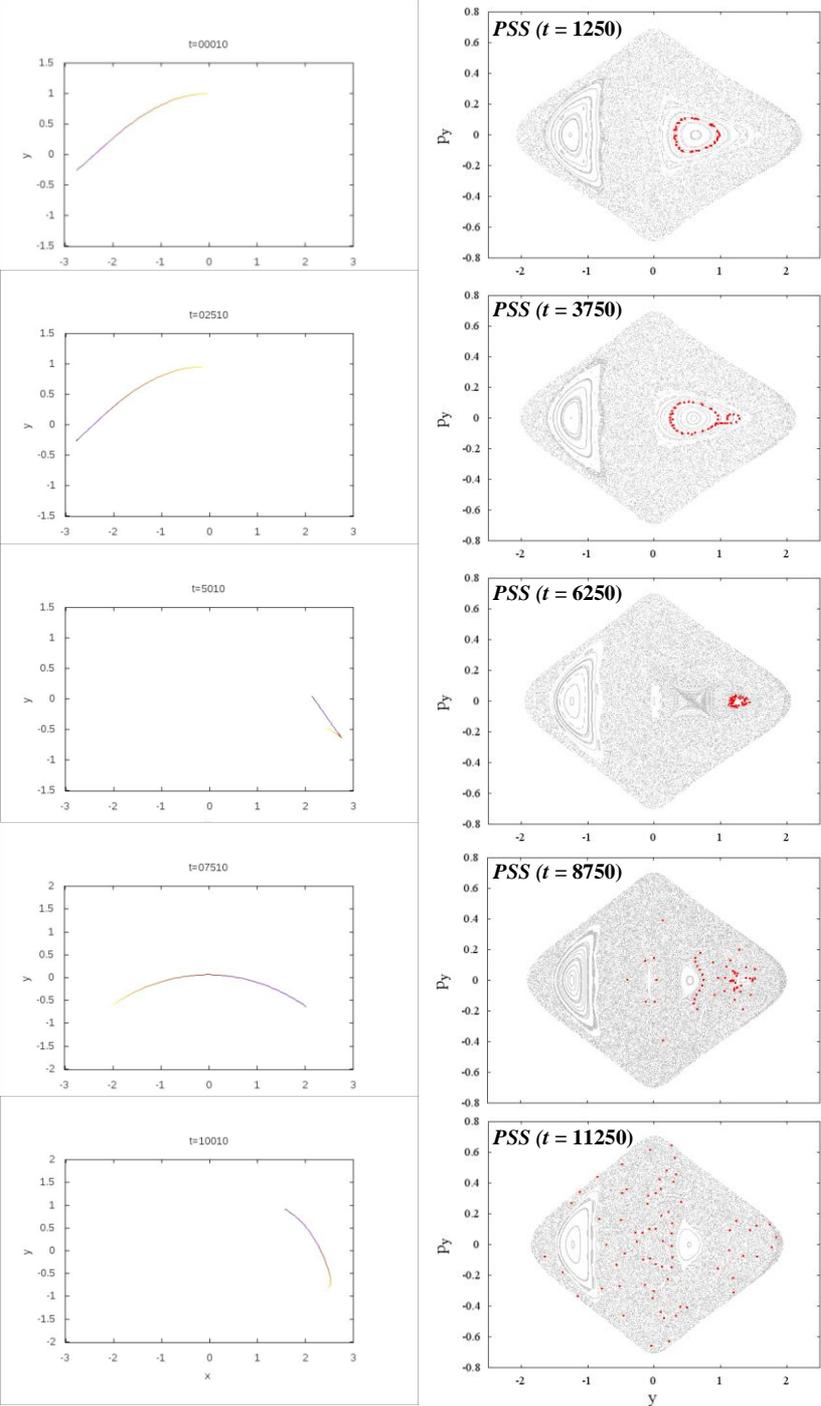
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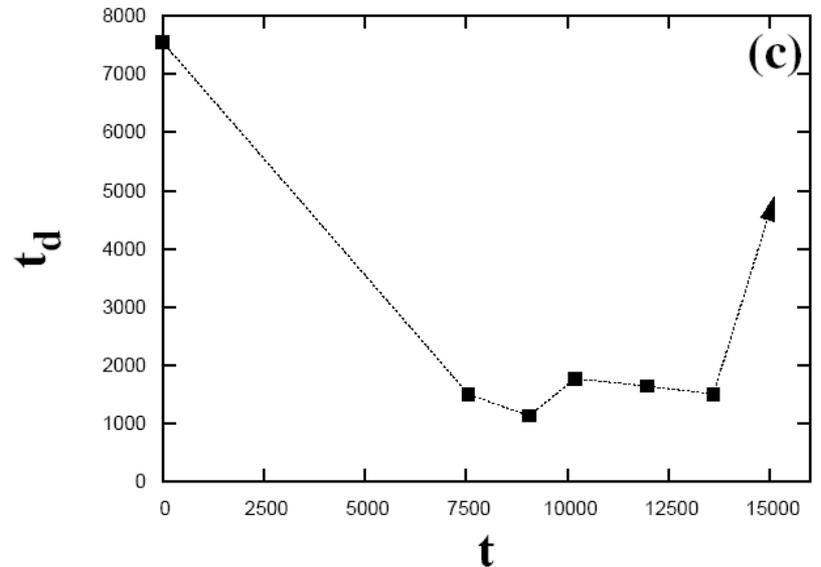
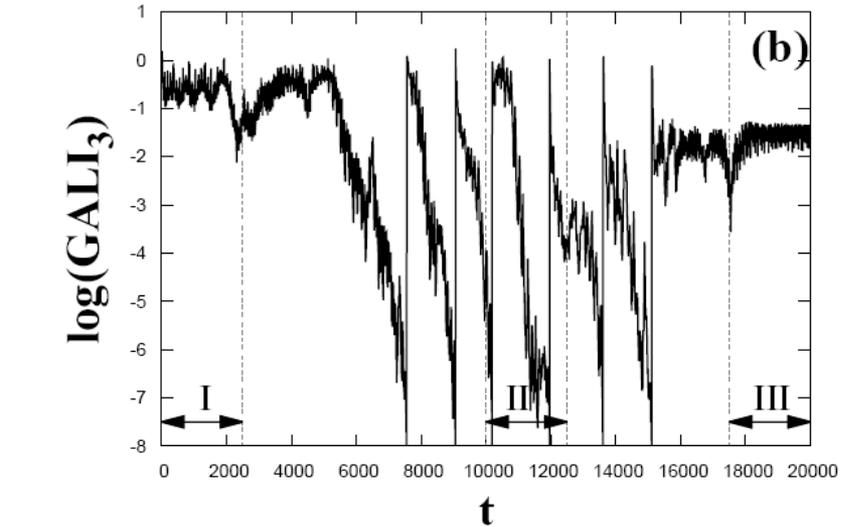
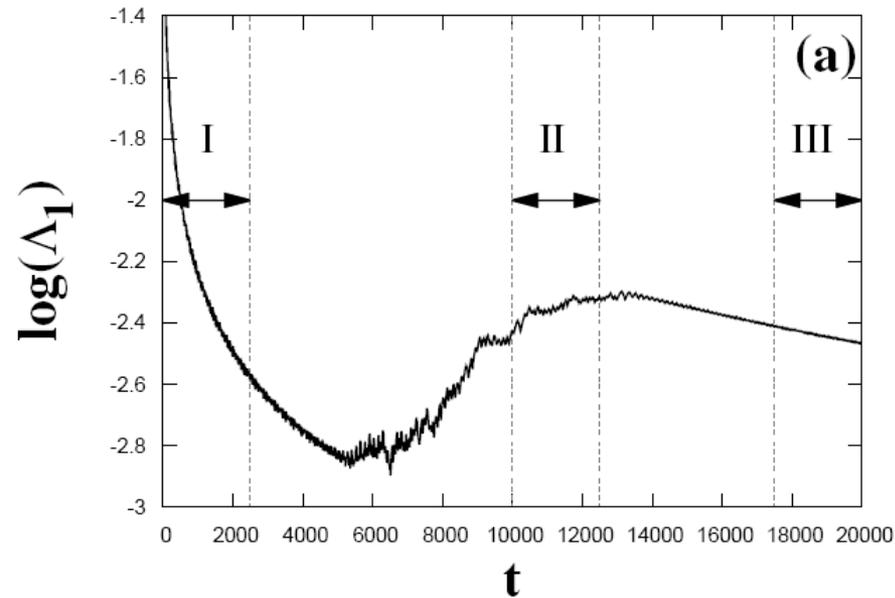
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Time-dependent 2D barred galaxy model



Time-dependent 3D barred galaxy model

Interplay between chaotic and regular motion



Summary

- We discussed methods of chaos detection based on
 - ✓ the visualization of orbits
 - ✓ the numerical analysis of orbits
 - ✓ the evolution of deviation vectors (variational equations – tangent map)

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- $GALI_k$ indices :
 - ✓ can **distinguish rapidly and with certainty between regular and chaotic motion**
 - ✓ can be used to characterize **individual orbits as well as "chart" chaotic and regular domains** in phase space
 - ✓ are perfectly suited for **studying the global dynamics of multidimensional systems, as well as of time-dependent models**

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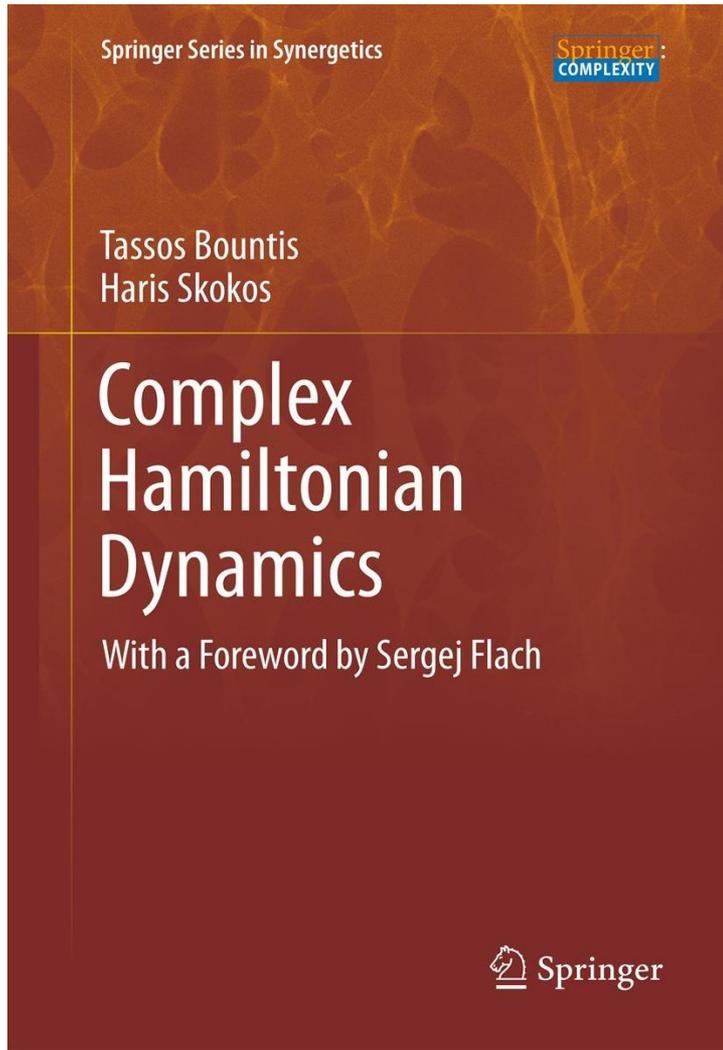
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