

Chaos in Hamiltonian systems

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/

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)]

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

We say that f is chaotic on V if

Chaos

Definition [Devaney (1989)]

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

We say that f is **chaotic** on V if

1. f has **sensitive dependence on initial conditions.**

Chaos

Definition [Devaney (1989)]

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

We say that f is **chaotic** on V if

1. f has **sensitive dependence on initial conditions.**
2. f is **topologically transitive.**

Chaos

Definition [Devaney (1989)]

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

We say that f is **chaotic** on V if

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

Chaos

1. **f** has sensitive dependence on initial conditions.

Chaos

1. **f** has sensitive dependence on initial conditions.

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

Chaos

1. **f** has sensitive dependence on initial conditions.

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

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

Chaos

1. **f** has sensitive dependence on initial conditions.

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

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} .

Chaos

2. **f** is topologically transitive.

Chaos

2. **f** is topologically transitive.

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

Chaos

2. **f** is topologically transitive.

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

This implies the existence of points which eventually move under iteration from one arbitrarily small neighborhood to any other.

Chaos

2. **f** is topologically transitive.

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

This implies the existence of points which eventually move under iteration from one arbitrarily small neighborhood to any other.

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

Chaos

A chaotic system possesses three ingredients:

Chaos

A chaotic system possesses three ingredients:

- 1. Unpredictability** because of the sensitive dependence on initial conditions

Chaos

A chaotic system possesses three ingredients:

1. **Unpredictability** because of the sensitive dependence on initial conditions
2. **Indecomposability** because it cannot be decomposed into noninteracting subsystems due to topological transitivity

Chaos

A chaotic system possesses three ingredients:

1. **Unpredictability** because of the sensitive dependence on initial conditions
2. **Indecomposability** because it cannot be decomposed into noninteracting subsystems due to topological transitivity
3. **An element of regularity** because it has periodic points which are dense.

Chaos

A chaotic system possesses three ingredients:

1. **Unpredictability** because of the sensitive dependence on initial conditions
2. **Indecomposability** because it cannot be decomposed into noninteracting subsystems due to topological transitivity
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.

Autonomous Hamiltonian systems

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

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

Autonomous Hamiltonian systems

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

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

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**

$$\frac{d p_i}{d t} = - \frac{\partial H}{\partial q_i} , \quad \frac{d q_i}{d t} = \frac{\partial H}{\partial p_i}$$

Autonomous Hamiltonian systems

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

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

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**

$$\frac{d p_i}{d t} = - \frac{\partial H}{\partial q_i} , \quad \frac{d q_i}{d t} = \frac{\partial H}{\partial p_i}$$

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

Example (Hénon-Heiles system)

Example (Hénon-Heiles system)

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

Example (Hénon-Heiles system)

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

THE ASTRONOMICAL JOURNAL

VOLUME 69, NUMBER 1

FEBRUARY 1964

The Applicability of the Third Integral Of Motion: Some Numerical Experiments

MICHEL HÉNON* AND CARL HEILES

Princeton University Observatory, Princeton, New Jersey

(Received 7 August 1963)

The problem of the existence of a third isolating integral of motion in an axisymmetric potential is investigated by numerical experiments. It is found that the third integral exists for only a limited range of initial conditions.

Example (Hénon-Heiles system)

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

THE ASTRONOMICAL JOURNAL

VOLUME 69, NUMBER 1

FEBRUARY 1964

The Applicability of the Third Integral Of Motion: Some Numerical Experiments

MICHEL HÉNON* AND CARL HEILES

Princeton University Observatory, Princeton, New Jersey

(Received 7 August 1963)

The problem of the existence of a third isolating integral of motion in an axisymmetric potential is investigated by numerical experiments. It is found that the third integral exists for only a limited range of initial conditions.

Hamilton's equations of motion:

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

Example (Hénon-Heiles system)

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

THE ASTRONOMICAL JOURNAL

VOLUME 69, NUMBER 1

FEBRUARY 1964

The Applicability of the Third Integral Of Motion: Some Numerical Experiments

MICHEL HÉNON* AND CARL HEILES

Princeton University Observatory, Princeton, New Jersey

(Received 7 August 1963)

The problem of the existence of a third isolating integral of motion in an axisymmetric potential is investigated by numerical experiments. It is found that the third integral exists for only a limited range of initial conditions.

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

Hénon-Heiles system

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

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

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

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

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

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

Orbit

Perturbed

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

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

Orbit

Perturbed

t= 100 x= 0.132995718333307644 0.132995718337263064

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

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

Orbit

Perturbed

t= 100 x= 0.132995718333307644 0.132995718337263064

t= 5000 x= 0.376999283889102310 0.376999283870156576

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

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

Orbit

Perturbed

| | | |
|----------|-------------------------|-----------------------|
| t= 100 | x= 0.132995718333307644 | 0.132995718337263064 |
| t= 5000 | x= 0.376999283889102310 | 0.376999283870156576 |
| t= 10000 | x=-0.159094583356855224 | -0.159094583341260309 |

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

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

Orbit

Perturbed

| | | |
|----------|-------------------------|-----------------------|
| t= 100 | x= 0.132995718333307644 | 0.132995718337263064 |
| t= 5000 | x= 0.376999283889102310 | 0.376999283870156576 |
| t= 10000 | x=-0.159094583356855224 | -0.159094583341260309 |
| t= 50000 | x= 0.101992400739955760 | 0.101992400253961321 |

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

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

Orbit

Perturbed

| | | |
|----------|-------------------------|-----------------------|
| t= 100 | x= 0.132995718333307644 | 0.132995718337263064 |
| t= 5000 | x= 0.376999283889102310 | 0.376999283870156576 |
| t= 10000 | x=-0.159094583356855224 | -0.159094583341260309 |
| t= 50000 | x= 0.101992400739955760 | 0.101992400253961321 |
| t=100000 | x=-0.381120533746511780 | -0.381120533327258870 |

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

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

Orbit

Perturbed

| | | |
|----------|-------------------------|-----------------------|
| t= 100 | x= 0.132995718333307644 | 0.132995718337263064 |
| t= 5000 | x= 0.376999283889102310 | 0.376999283870156576 |
| t= 10000 | x=-0.159094583356855224 | -0.159094583341260309 |
| t= 50000 | x= 0.101992400739955760 | 0.101992400253961321 |
| t=100000 | x=-0.381120533746511780 | -0.381120533327258870 |
| t= 100 | x= 0.090272817735167835 | 0.090272821355768668 |

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

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

| | <u>Orbit</u> | <u>Perturbed</u> |
|----------|-------------------------|-----------------------|
| t= 100 | x= 0.132995718333307644 | 0.132995718337263064 |
| t= 5000 | x= 0.376999283889102310 | 0.376999283870156576 |
| t= 10000 | x=-0.159094583356855224 | -0.159094583341260309 |
| t= 50000 | x= 0.101992400739955760 | 0.101992400253961321 |
| t=100000 | x=-0.381120533746511780 | -0.381120533327258870 |
| t= 100 | x= 0.090272817735167835 | 0.090272821355768668 |
| t= 200 | x= 0.295031687482249283 | 0.295031884858625637 |

Regular vs Chaotic orbits

Hénon-Heiles system

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

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 **$x=0, y=-0.25, p_y=0$** .

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

| | <u>Orbit</u> | <u>Perturbed</u> |
|----------|-------------------------|-----------------------|
| t= 100 | x= 0.132995718333307644 | 0.132995718337263064 |
| t= 5000 | x= 0.376999283889102310 | 0.376999283870156576 |
| t= 10000 | x=-0.159094583356855224 | -0.159094583341260309 |
| t= 50000 | x= 0.101992400739955760 | 0.101992400253961321 |
| 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 |

Regular vs Chaotic orbits

Hénon-Heiles system

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

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 **$x=0, y=-0.25, p_y=0$** .

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

| | <u>Orbit</u> | <u>Perturbed</u> |
|----------|-------------------------|-----------------------|
| t= 100 | x= 0.132995718333307644 | 0.132995718337263064 |
| t= 5000 | x= 0.376999283889102310 | 0.376999283870156576 |
| t= 10000 | x=-0.159094583356855224 | -0.159094583341260309 |
| t= 50000 | x= 0.101992400739955760 | 0.101992400253961321 |
| 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 |

Regular vs Chaotic orbits

Hénon-Heiles system

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

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

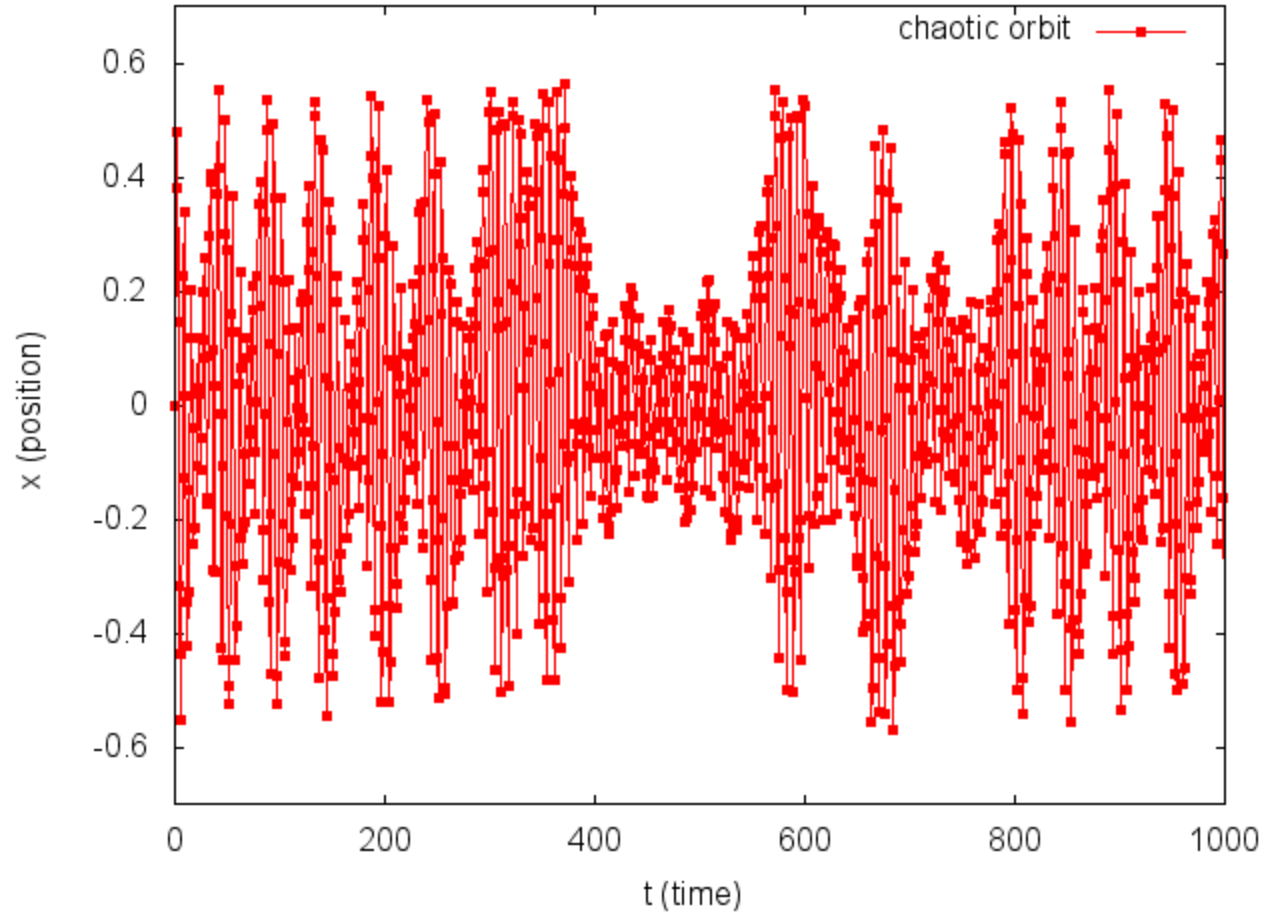
$x=0, y=0.1, p_y=0$ and **$x=0, y=-0.25, p_y=0$** .

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

| | <u>Orbit</u> | <u>Perturbed</u> |
|----------|-------------------------|-----------------------|
| t= 100 | x= 0.132995718333307644 | 0.132995718337263064 |
| t= 5000 | x= 0.376999283889102310 | 0.376999283870156576 |
| t= 10000 | x=-0.159094583356855224 | -0.159094583341260309 |
| t= 50000 | x= 0.101992400739955760 | 0.101992400253961321 |
| 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 |
| t= 500 | x= 0.078357719290523528 | -0.270811022674341095 |

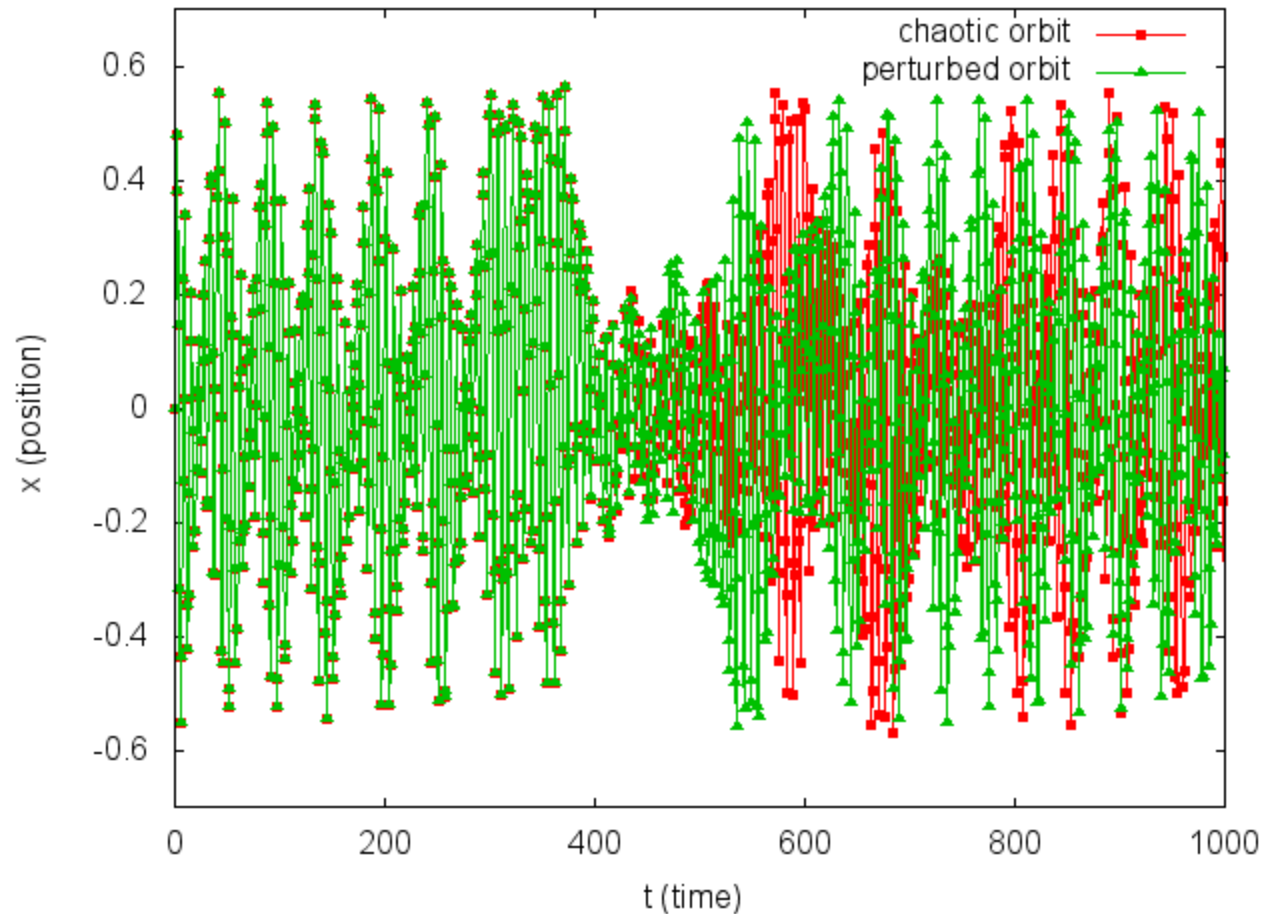
Regular vs Chaotic orbits

Chaotic orbit



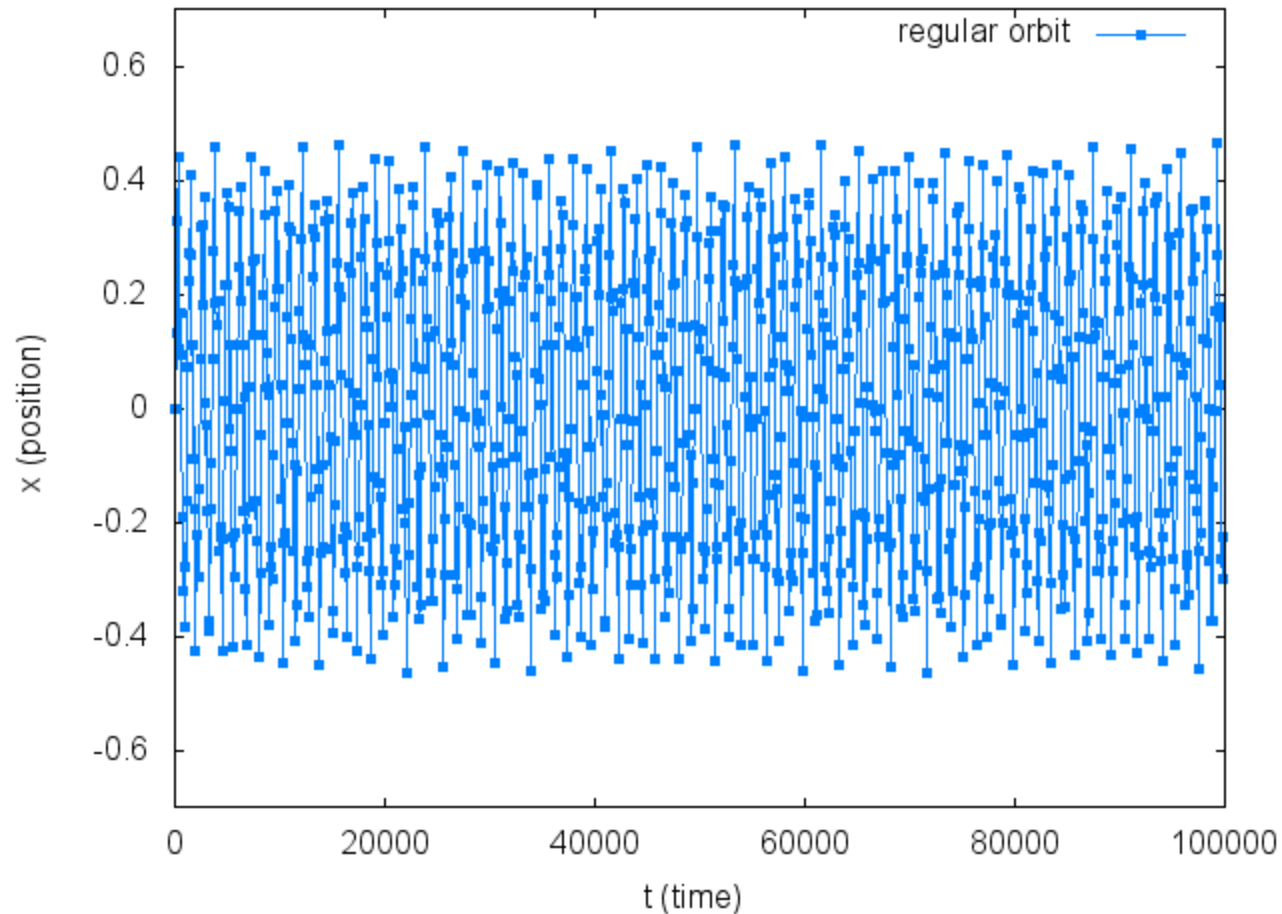
Regular vs Chaotic orbits

Chaotic orbit and its perturbation



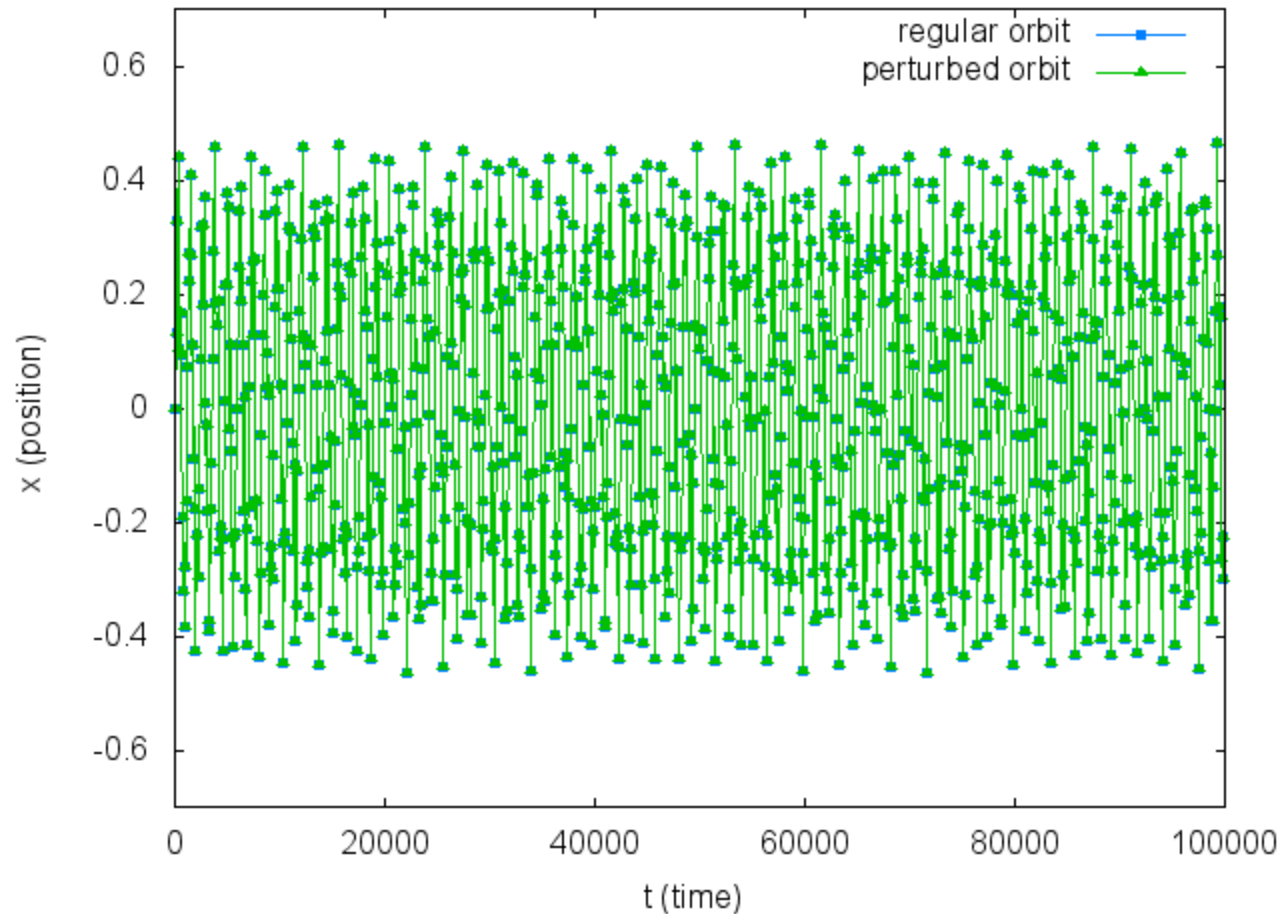
Regular vs Chaotic orbits

Regular orbit



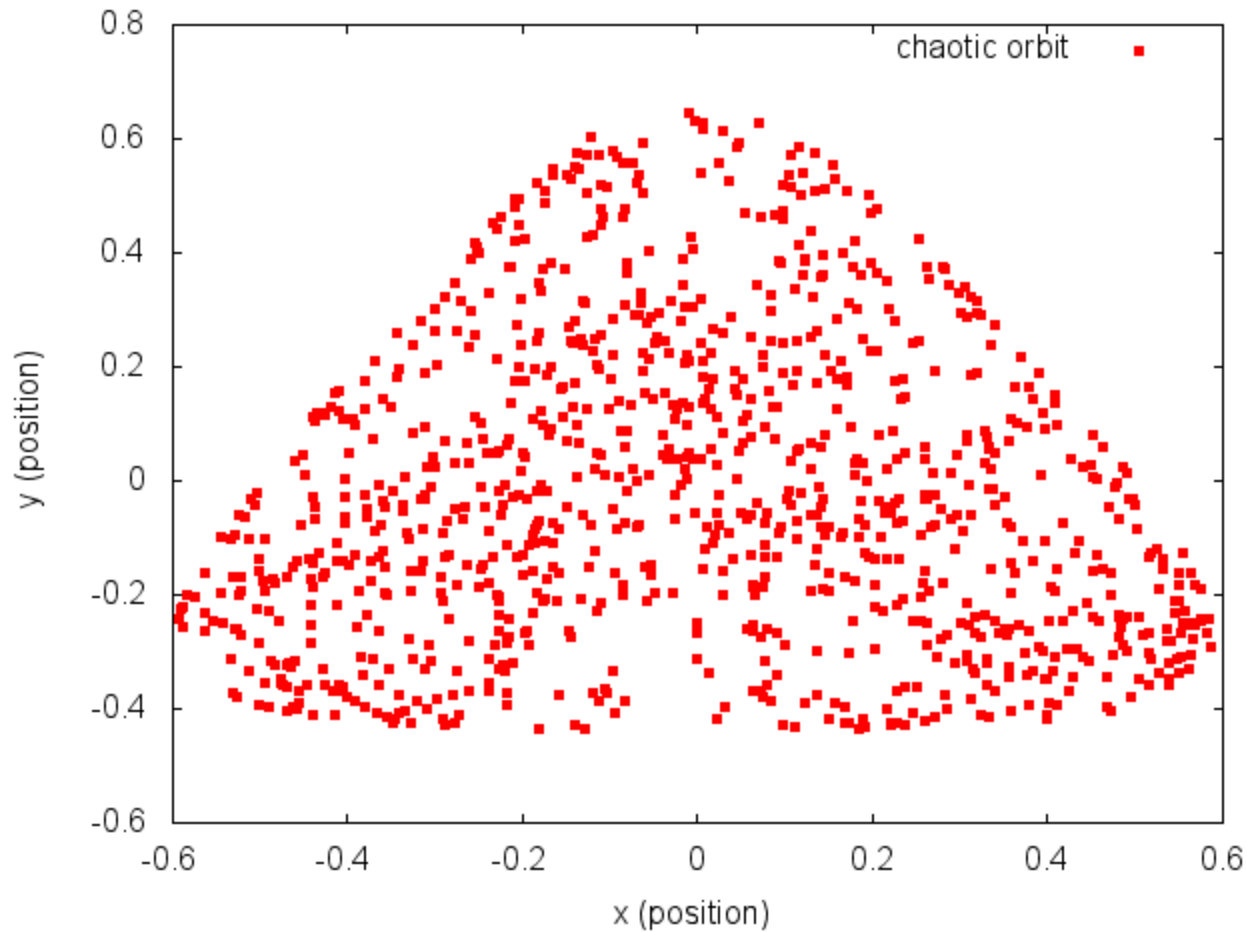
Regular vs Chaotic orbits

Regular orbit and its perturbation



Regular vs Chaotic orbits

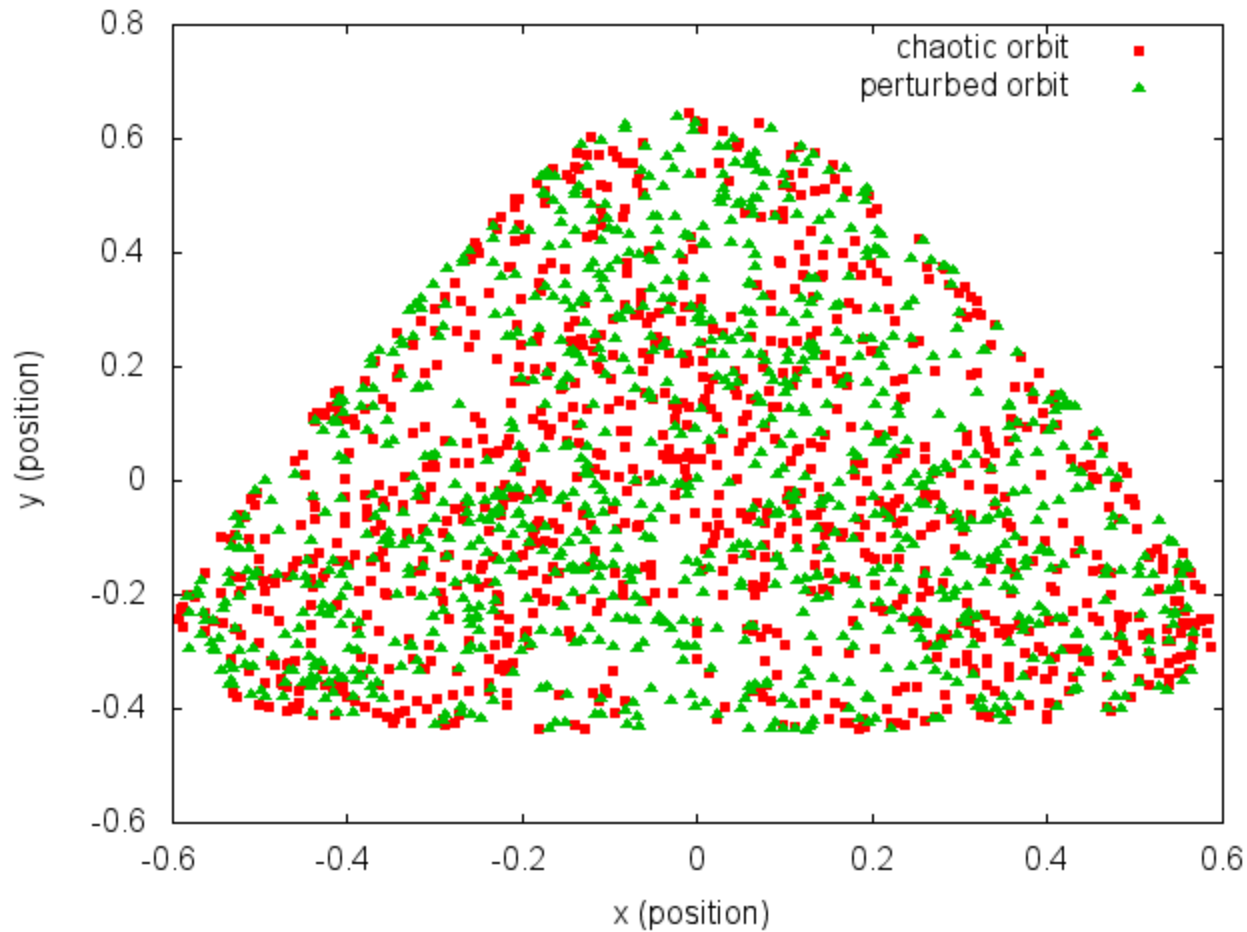
Chaotic orbit



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

Regular vs Chaotic orbits

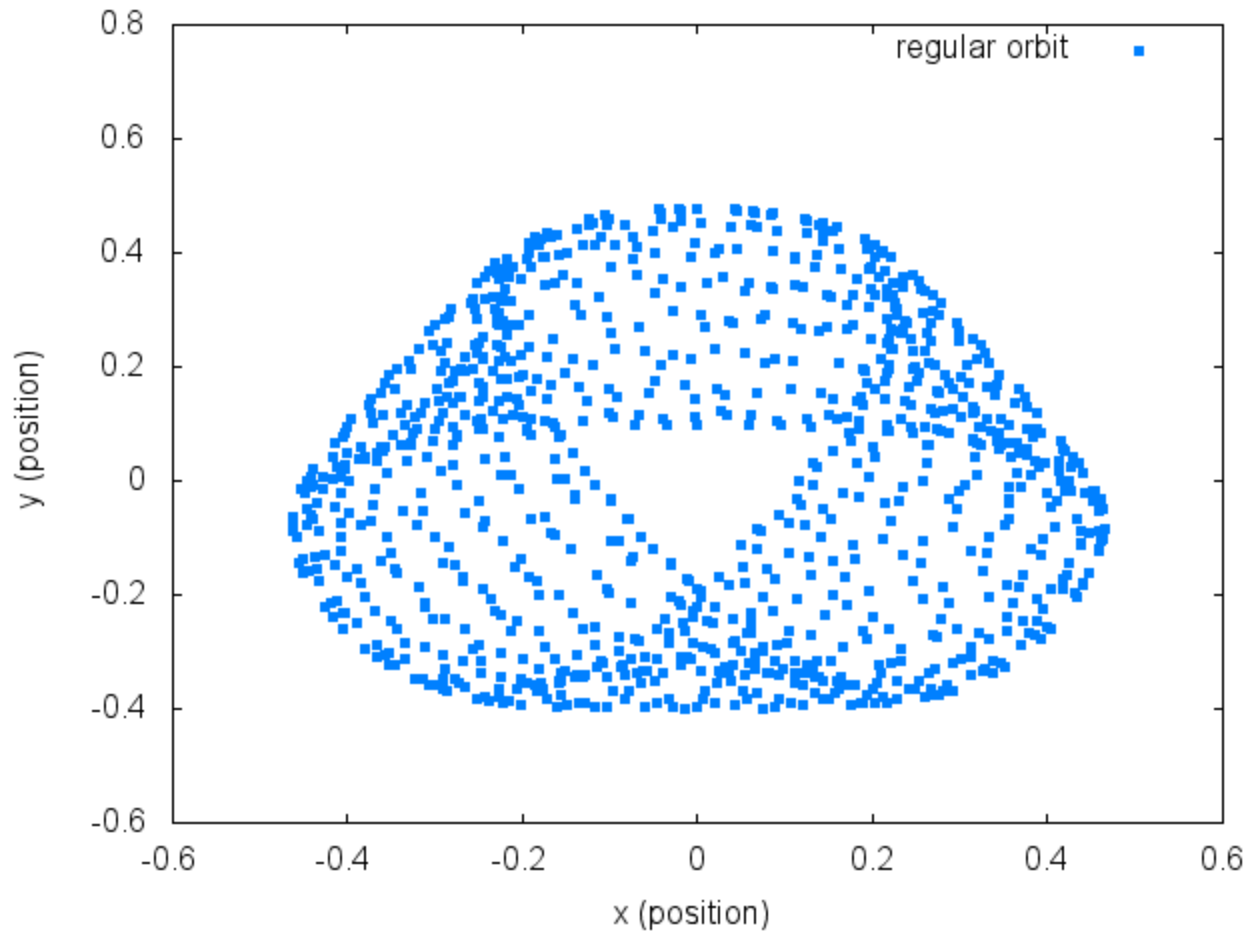
Chaotic orbit and its perturbation



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

Regular vs Chaotic orbits

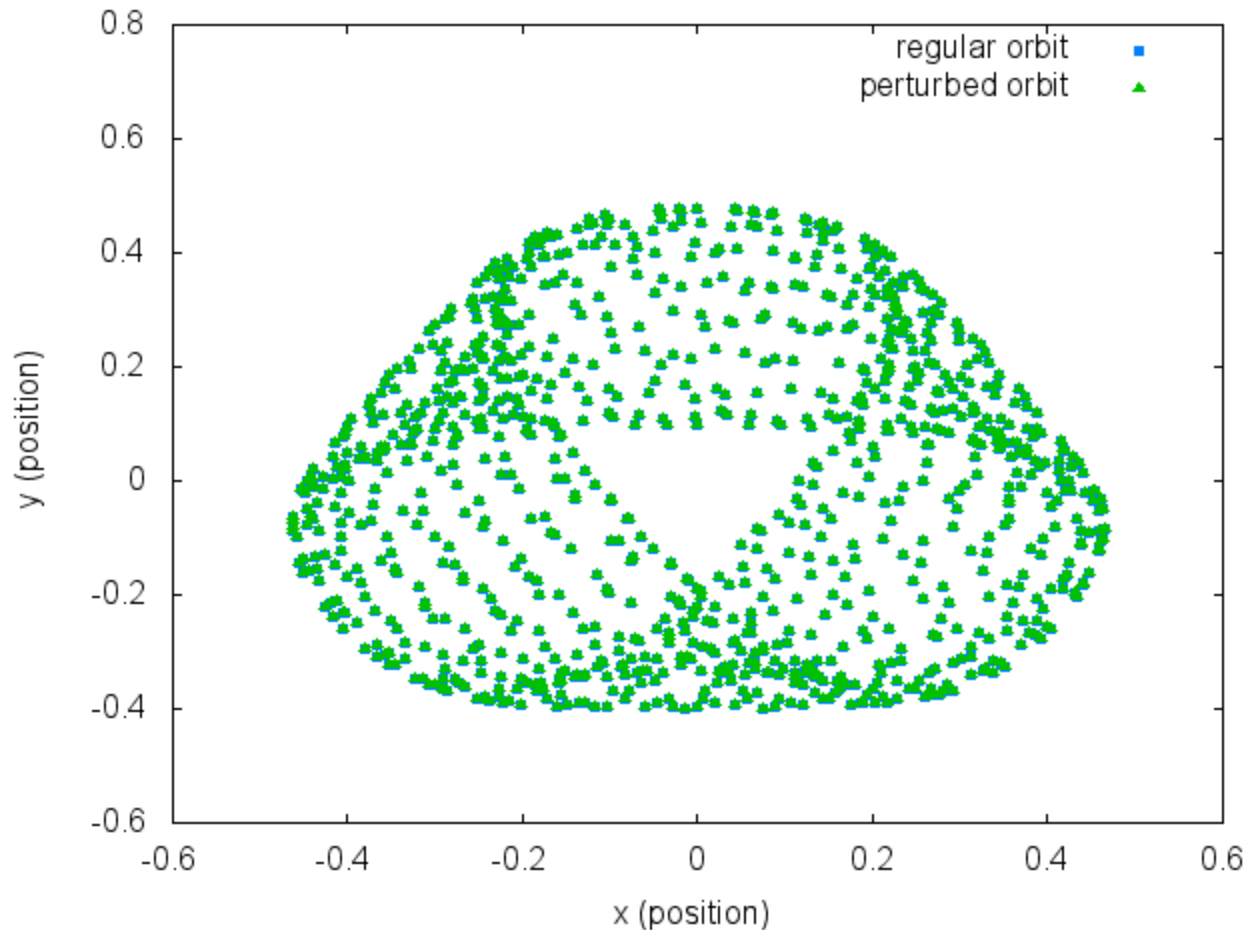
Regular orbit



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

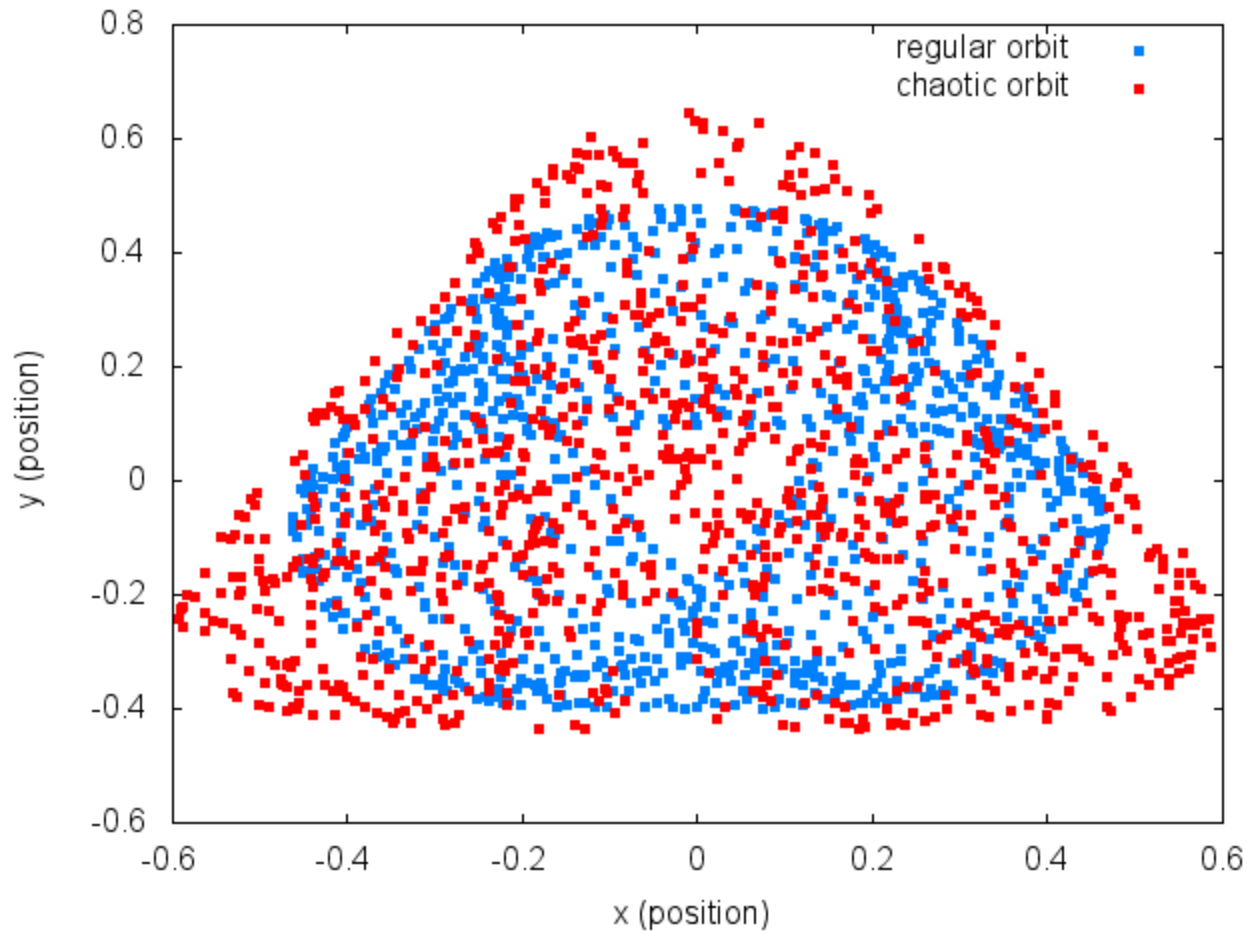
Regular vs Chaotic orbits

Regular orbit and its perturbation



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

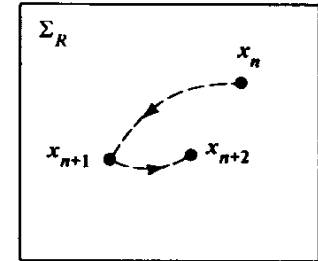
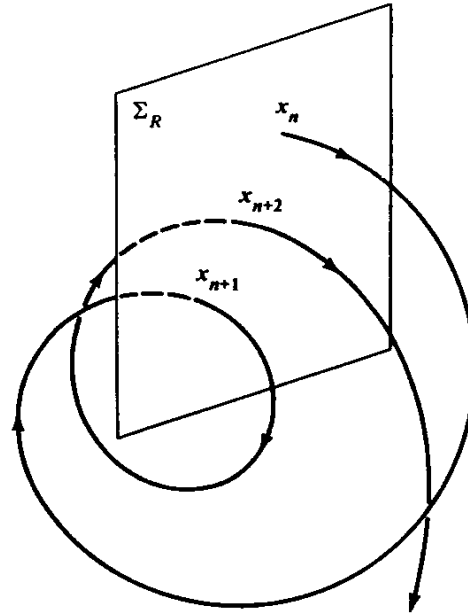
Regular vs Chaotic orbits



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

Poincaré Surface of Section (PSS)

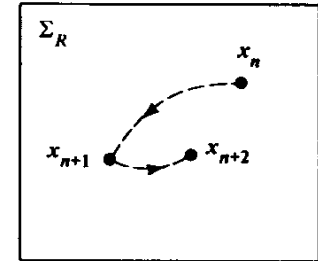
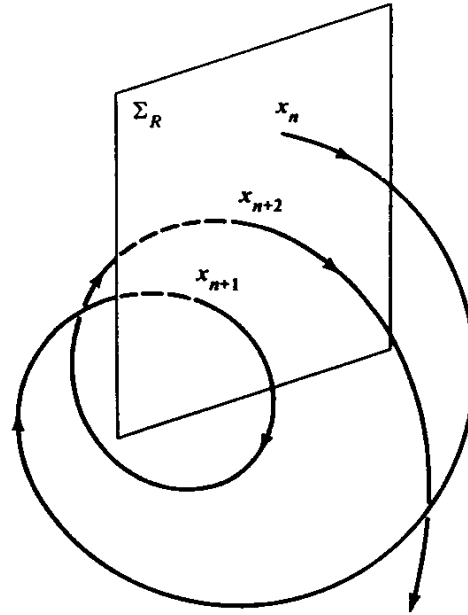
We can constrain the study of an $N+1$ degree of freedom Hamiltonian system to a **2N-dimensional subspace of the general phase space.**



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

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.**

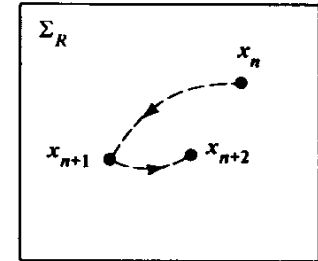
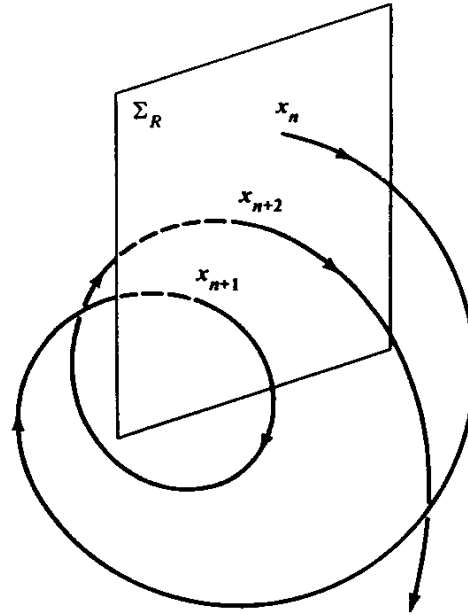


Lieberman & Lichtenberg, 1992, *Regular and Chaotic Dynamics*, Springer.

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.**

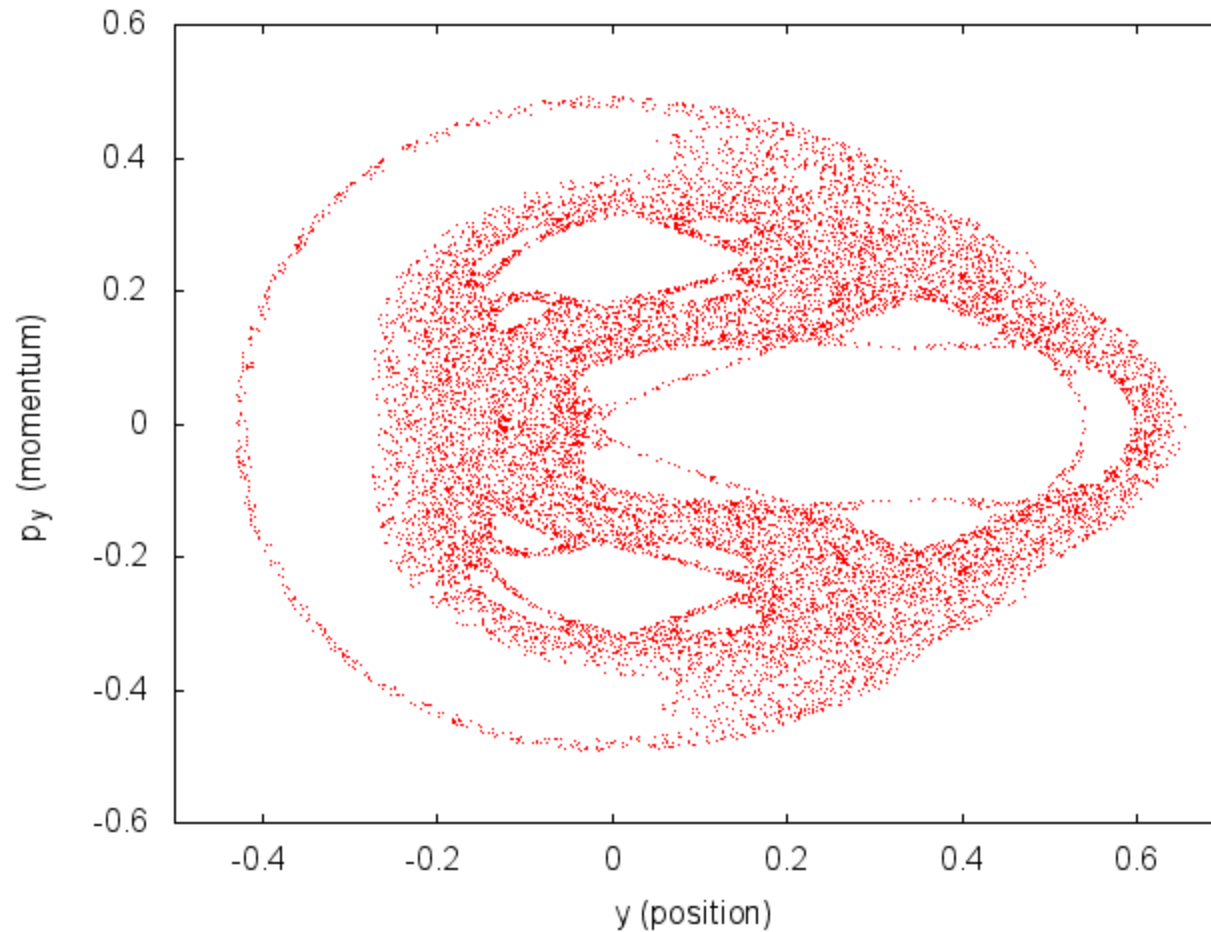


Lieberman & Lichtenberg, 1992, *Regular and Chaotic Dynamics*, Springer.

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.

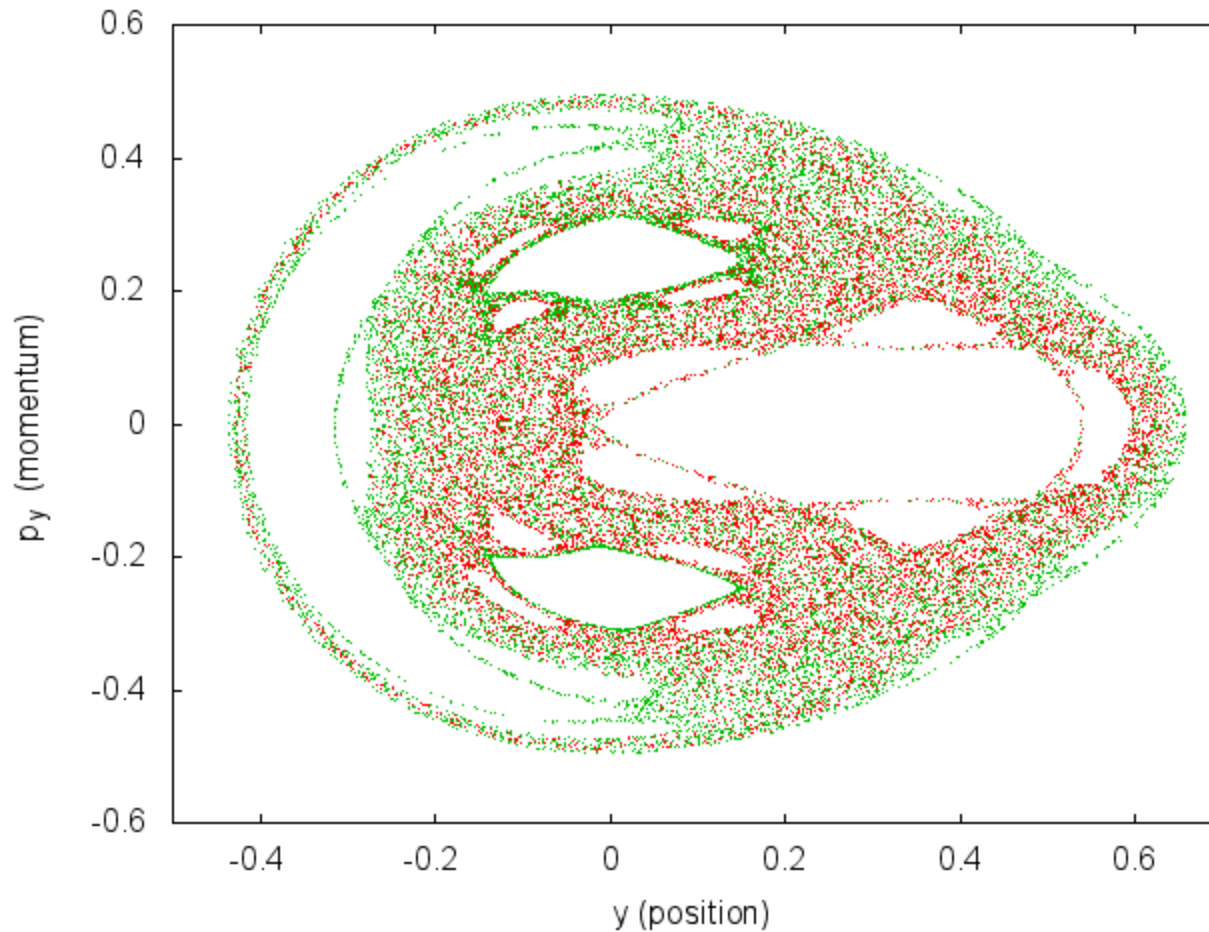
In this sense **an $N+1$ degree of freedom Hamiltonian system corresponds to a 2N-dimensional map.**

Hénon-Heiles system: PSS ($x=0$)



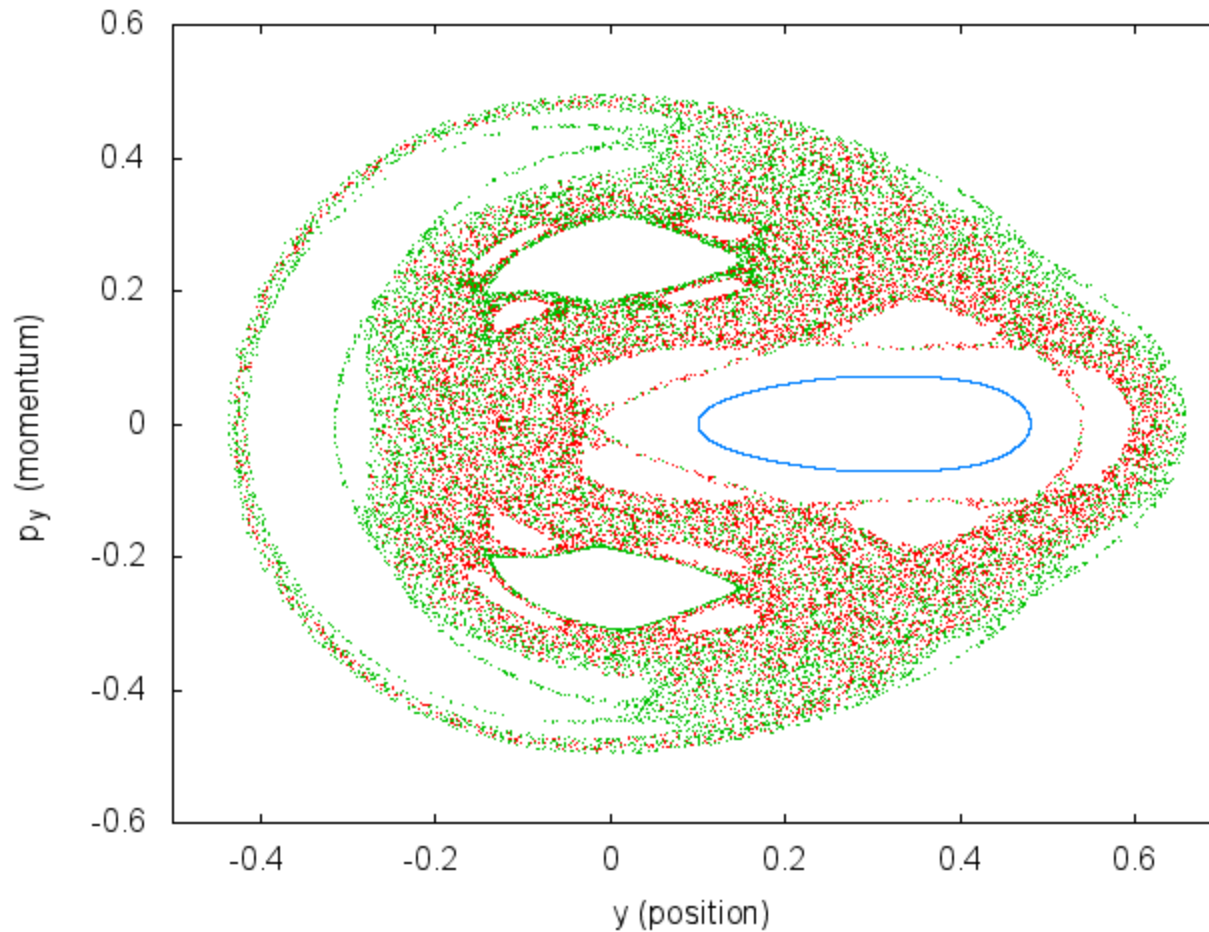
Chaotic orbit

Hénon-Heiles system: PSS ($x=0$)



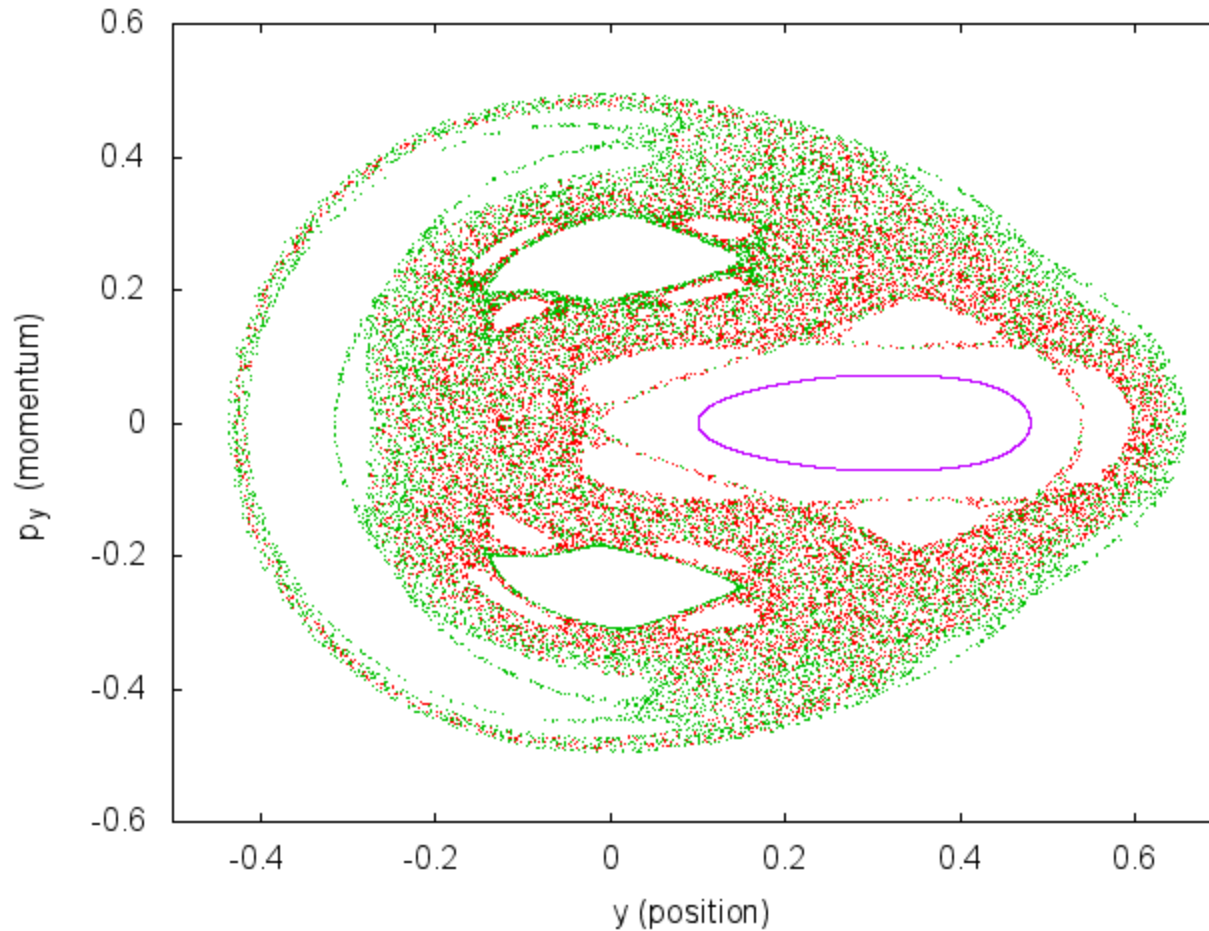
Chaotic orbit - **Perturbed chaotic orbit**

Hénon-Heiles system: PSS ($x=0$)



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

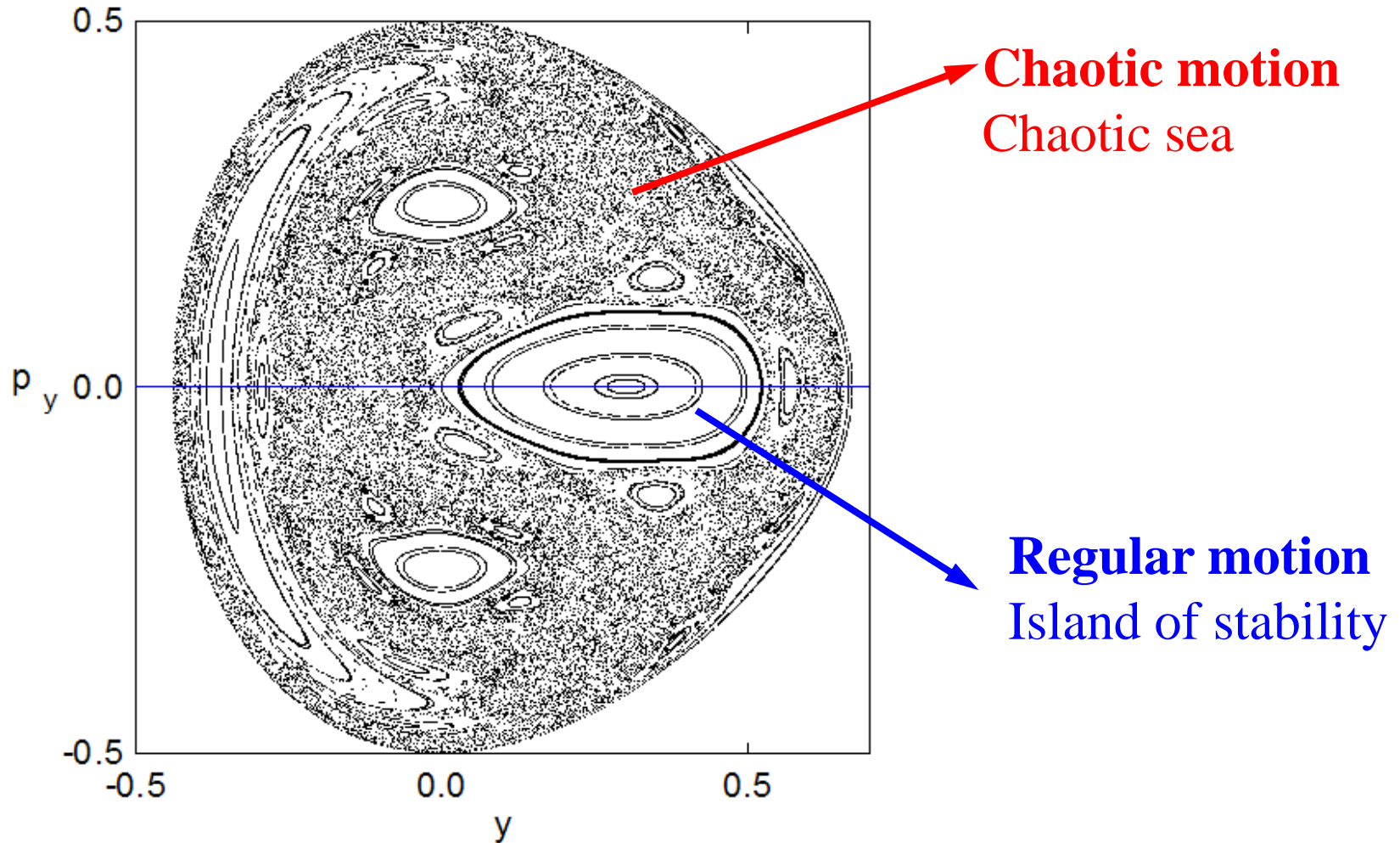
Hénon-Heiles system: PSS ($x=0$)



Chaotic orbit - **Perturbed chaotic orbit**

Regular orbit - **Perturbed regular orbit**

Hénon-Heiles system: PSS ($x=0$)



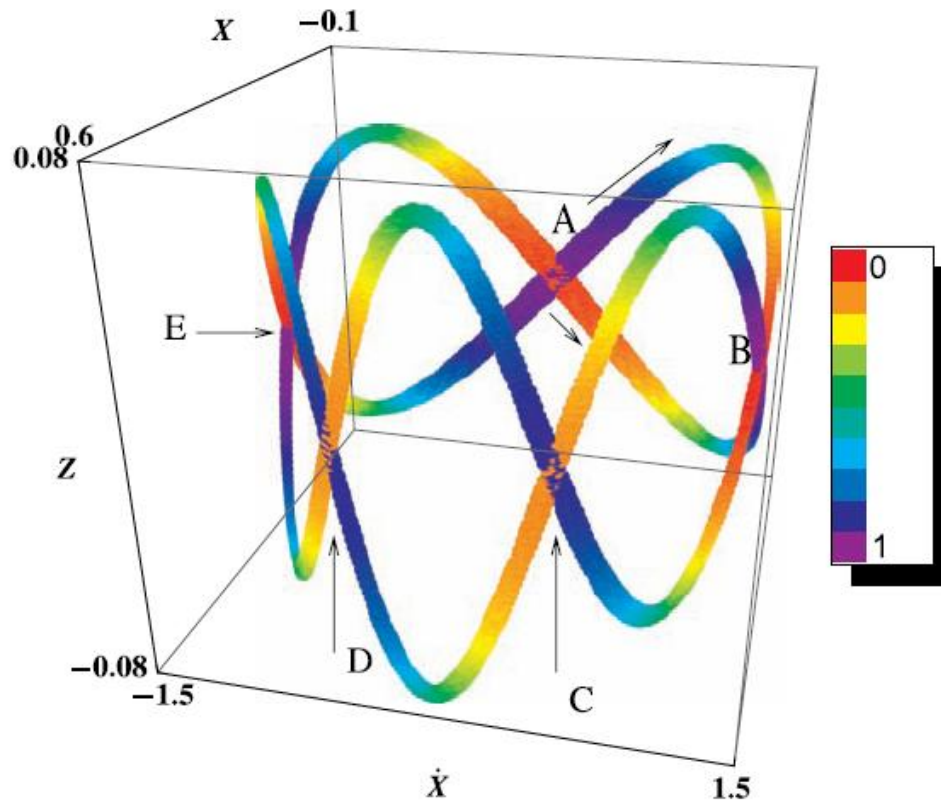
Chaos detection techniques

- **Based on the visualization of orbits**
 - ✓ **Poincaré Surface of Section (PSS)**
 - ✓ **the color and rotation (CR) method**
 - ✓ **the 3D phase space slices (3PSS) technique**

The color and rotation (CR) method

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

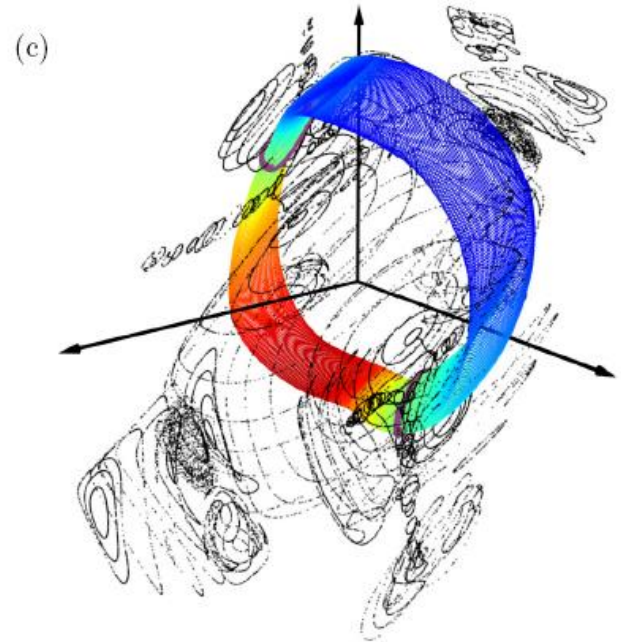
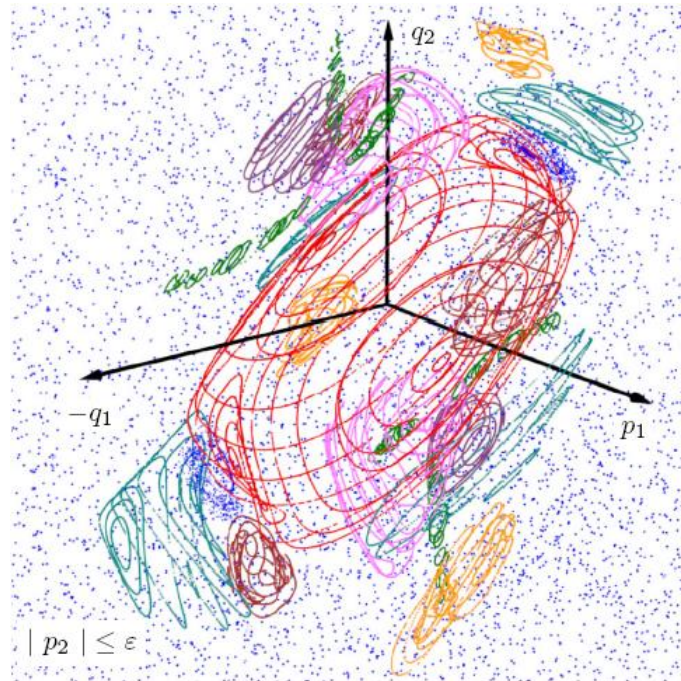
We consider the 3D projection of the PSS and use color to indicate the 4th dimension.



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. $|p_2| \leq \varepsilon$) and present intersections of orbits with these slices.



Chaos detection techniques

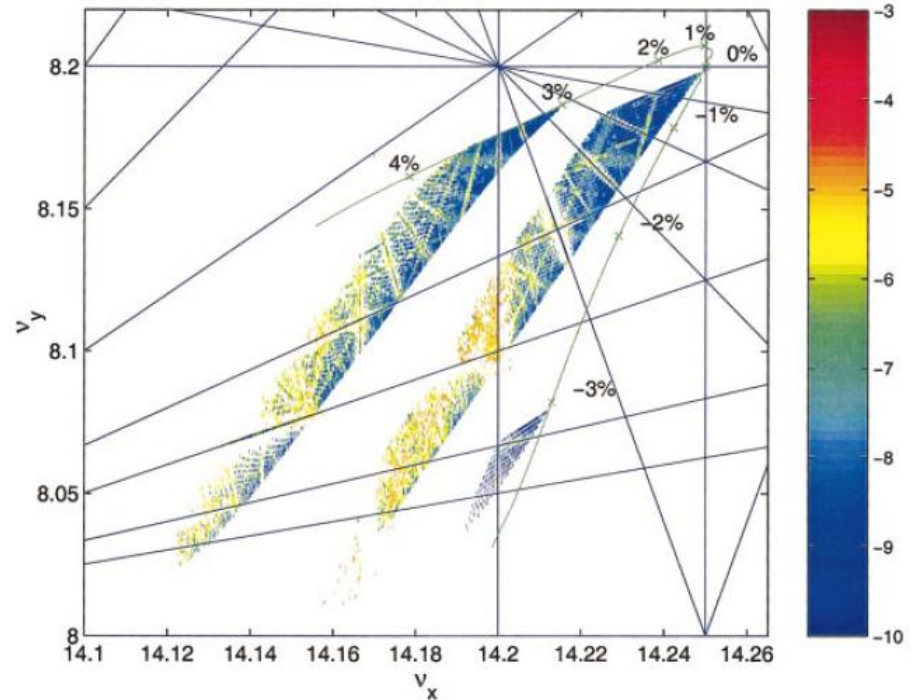
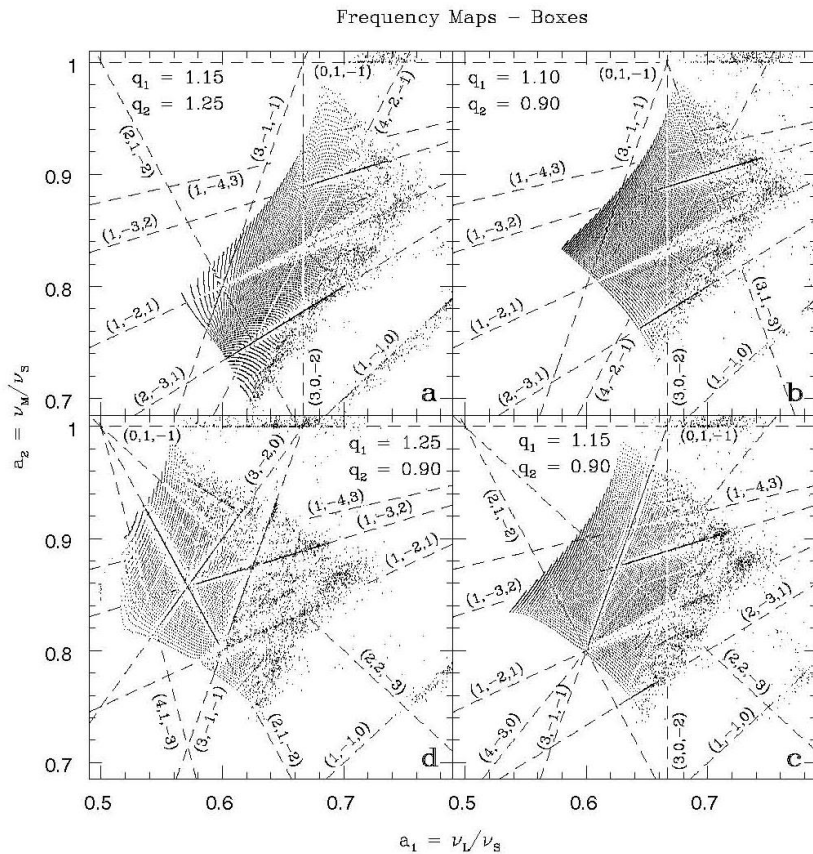
- **Based on the visualization of orbits**
 - ✓ **Poincaré Surface of Section (PSS)**
 - ✓ **the color and rotation (CR) method**
 - ✓ **the 3D phase space slices (3PSS) technique**
- **Based on the numerical analysis of orbits**
 - ✓ **Frequency Map Analysis**
 - ✓ **0-1 test**

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

Chaos detection techniques

- **Based on the visualization of orbits**
 - ✓ Poincaré Surface of Section (PSS)
 - ✓ the color and rotation (CR) method
 - ✓ the 3D phase space slices (3PSS) technique
- **Based on the numerical analysis of orbits**
 - ✓ Frequency Map Analysis
 - ✓ 0-1 test
- **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

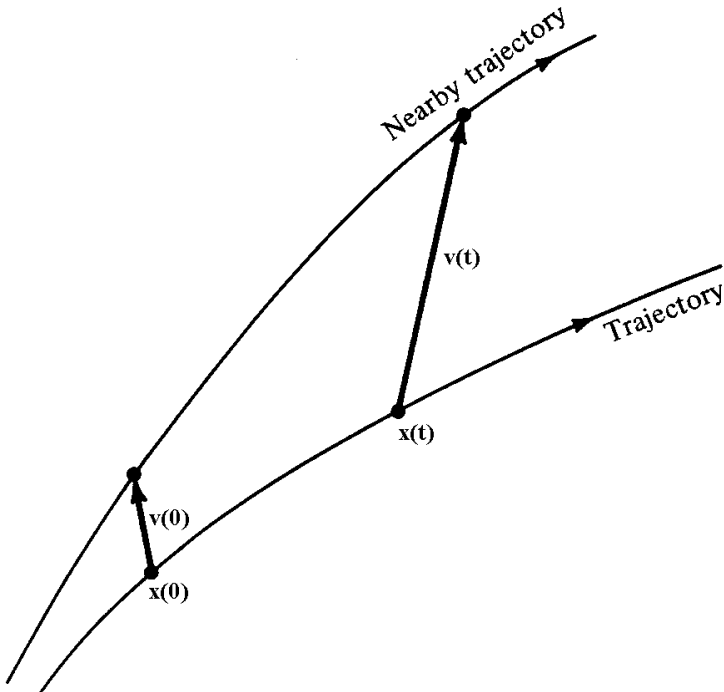
$$\mathbf{v} = (\delta x_1, \delta x_2, \dots, \delta x_n)^T, \text{ with } n=2N$$

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

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



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:

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

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:

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

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:

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

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:

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

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:

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

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:

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

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:

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

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:

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

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:

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

Example (Hénon-Heiles system)

Variational equations: $\frac{dv}{dt} = -J \cdot P \cdot v$

Example (Hénon-Heiles system)

Variational equations: $\frac{dv}{dt} = -J \cdot P \cdot v$

$$\begin{pmatrix} \dot{v}_1 \\ \dot{v}_2 \\ \dot{v}_3 \\ \dot{v}_4 \end{pmatrix}$$

Example (Hénon-Heiles system)

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

$$\begin{pmatrix} \dot{\mathbf{v}}_1 \\ \dot{\mathbf{v}}_2 \\ \dot{\mathbf{v}}_3 \\ \dot{\mathbf{v}}_4 \end{pmatrix} = - \begin{pmatrix} 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix}$$

Example (Hénon-Heiles system)

Variational equations: $\frac{dv}{dt} = -J \cdot P \cdot v$

$$\begin{pmatrix} \dot{v}_1 \\ \dot{v}_2 \\ \dot{v}_3 \\ \dot{v}_4 \end{pmatrix} = - \begin{pmatrix} 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1+2y & 2x & 0 & 0 \\ 2x & 1-2y & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

Example (Hénon-Heiles system)

Variational equations: $\frac{dv}{dt} = -J \cdot P \cdot v$

$$\begin{pmatrix} \dot{v}_1 \\ \dot{v}_2 \\ \dot{v}_3 \\ \dot{v}_4 \end{pmatrix} = - \begin{pmatrix} 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1+2y & 2x & 0 & 0 \\ 2x & 1-2y & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \end{pmatrix}$$

Example (Hénon-Heiles system)

Variational equations: $\frac{dv}{dt} = -J \cdot P \cdot v$

$$\begin{pmatrix} \dot{v}_1 \\ \dot{v}_2 \\ \dot{v}_3 \\ \dot{v}_4 \end{pmatrix} = - \begin{pmatrix} 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1+2y & 2x & 0 & 0 \\ 2x & 1-2y & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \end{pmatrix}$$

$$\begin{aligned} \dot{v}_1 &= v_3 \\ \dot{v}_2 &= v_4 \\ \dot{v}_3 &= -v_1 - 2xv_2 - 2yv_1 \\ \dot{v}_4 &= -v_2 - 2xv_1 + 2yv_2 \end{aligned}$$

Example (Hénon-Heiles system)

Variational equations: $\frac{dv}{dt} = -J \cdot P \cdot v$

$$\begin{pmatrix} \dot{v}_1 \\ \dot{v}_2 \\ \dot{v}_3 \\ \dot{v}_4 \end{pmatrix} = - \begin{pmatrix} 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1+2y & 2x & 0 & 0 \\ 2x & 1-2y & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \end{pmatrix}$$

$$\dot{v}_1 = v_3$$

$$\dot{v}_2 = v_4$$

$$\dot{v}_3 = -v_1 - 2xv_2 - 2yv_1$$

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

Example (Hénon-Heiles system)

Variational equations: $\frac{dv}{dt} = -J \cdot P \cdot v$

$$\begin{pmatrix} \dot{v}_1 \\ \dot{v}_2 \\ \dot{v}_3 \\ \dot{v}_4 \end{pmatrix} = - \begin{pmatrix} 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1+2y & 2x & 0 & 0 \\ 2x & 1-2y & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \end{pmatrix}$$

| | | |
|------------------------------------|---|------------------------------|
| $\dot{v}_1 = v_3$ | | $\dot{x} = p_x$ |
| $\dot{v}_2 = v_4$ | | $\dot{y} = p_y$ |
| $\dot{v}_3 = -v_1 - 2xv_2 - 2yv_1$ | + | $\dot{p}_x = -x - 2xy$ |
| $\dot{v}_4 = -v_2 - 2xv_1 + 2yv_2$ | | $\dot{p}_y = -y - x^2 + y^2$ |

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

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

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**

$$P(i+1)=T P(i) \text{ , } i=0,1,2,\dots$$

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**

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

Example – 2D map

Equations of the map:

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

Example – 2D map

Equations of the map:

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

Tangent map:

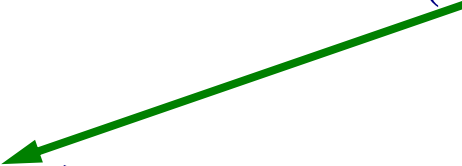
$$\mathbf{v}(\mathbf{i} + 1) = \left. \frac{\partial \mathbf{T}}{\partial \mathbf{P}} \right|_{\mathbf{i}} \cdot \mathbf{v}(\mathbf{i})$$

Example – 2D map

Equations of the map:

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

Tangent map:

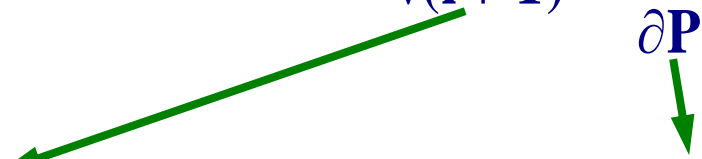
$$\mathbf{v}(\mathbf{i} + 1) = \left. \frac{\partial \mathbf{T}}{\partial \mathbf{P}} \right|_{\mathbf{i}} \cdot \mathbf{v}(\mathbf{i})$$

$$\begin{pmatrix} d\mathbf{x}'_1 \\ d\mathbf{x}'_2 \end{pmatrix}$$

Example – 2D map

Equations of the map:

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

Tangent map:

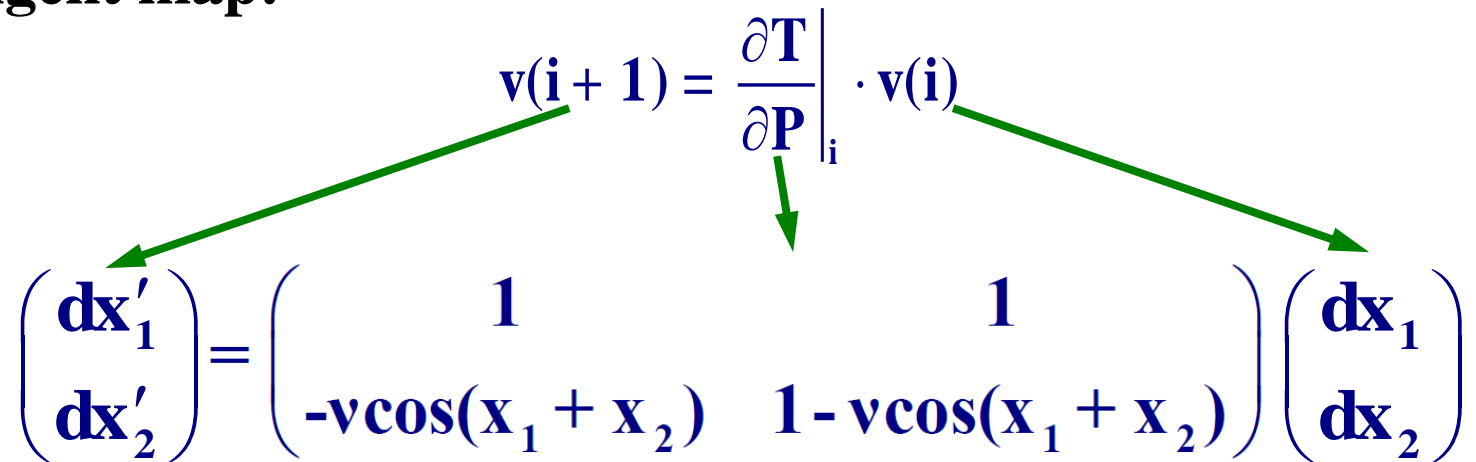
$$\mathbf{v}(i+1) = \left. \frac{\partial \mathbf{T}}{\partial \mathbf{P}} \right|_i \cdot \mathbf{v}(i)$$

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

Example – 2D map

Equations of the map:

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

Tangent map:

$$\mathbf{v}(i+1) = \left. \frac{\partial \mathbf{T}}{\partial \mathbf{P}} \right|_i \cdot \mathbf{v}(i)$$

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

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:

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

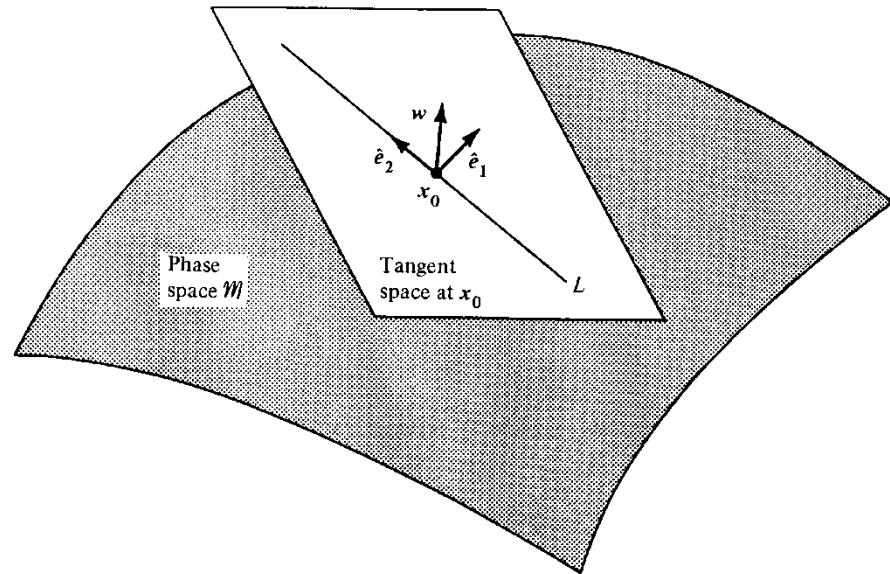
We commonly use the Euclidian norm and set $d(0)=\|\mathbf{v}(0)\|=1$

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 $\mathbf{v}(t)$ (and of $d(t)=\|\mathbf{v}(t)\|$) we **renormalize $\mathbf{v}(t)$** from time to time.

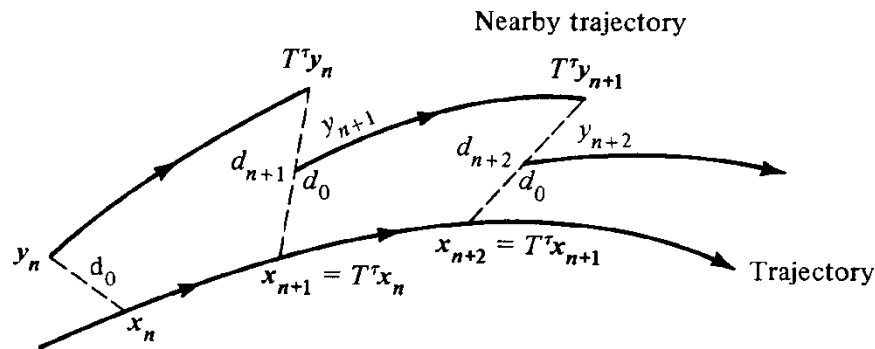


Figure 5.6. Numerical calculation of the maximal Liapunov characteristic exponent. Here $y = x + v$ and τ is a finite interval of time (after Benettin *et al.*, 1976).

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

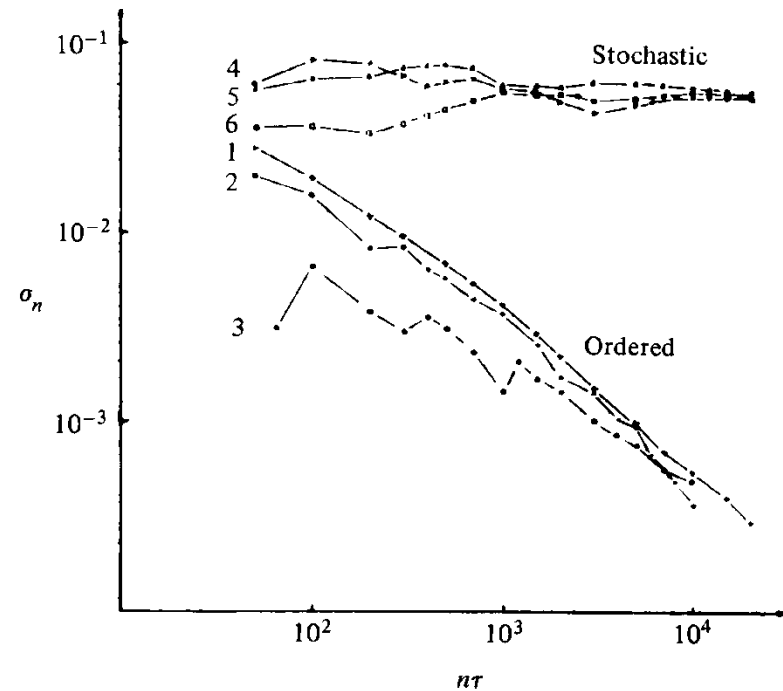
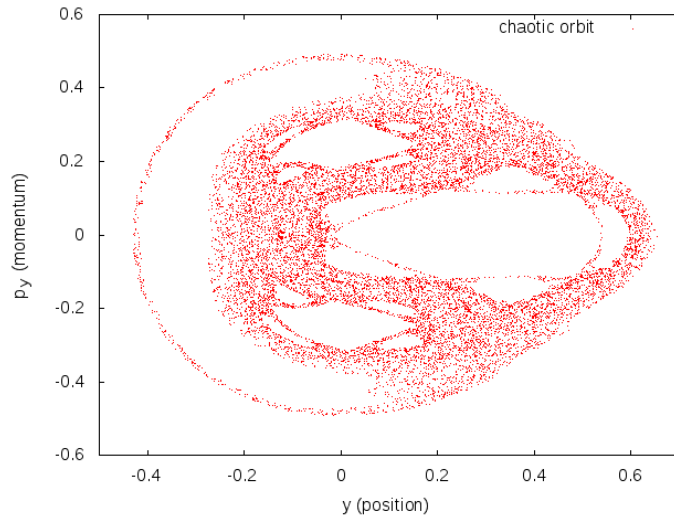


Figure 5.7. Behavior of σ_n at the intermediate energy $E = 0.125$ for initial points taken in the ordered (curves 1–3) or stochastic (curves 4–6) regions (after Benettin *et al.*, 1976).

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.

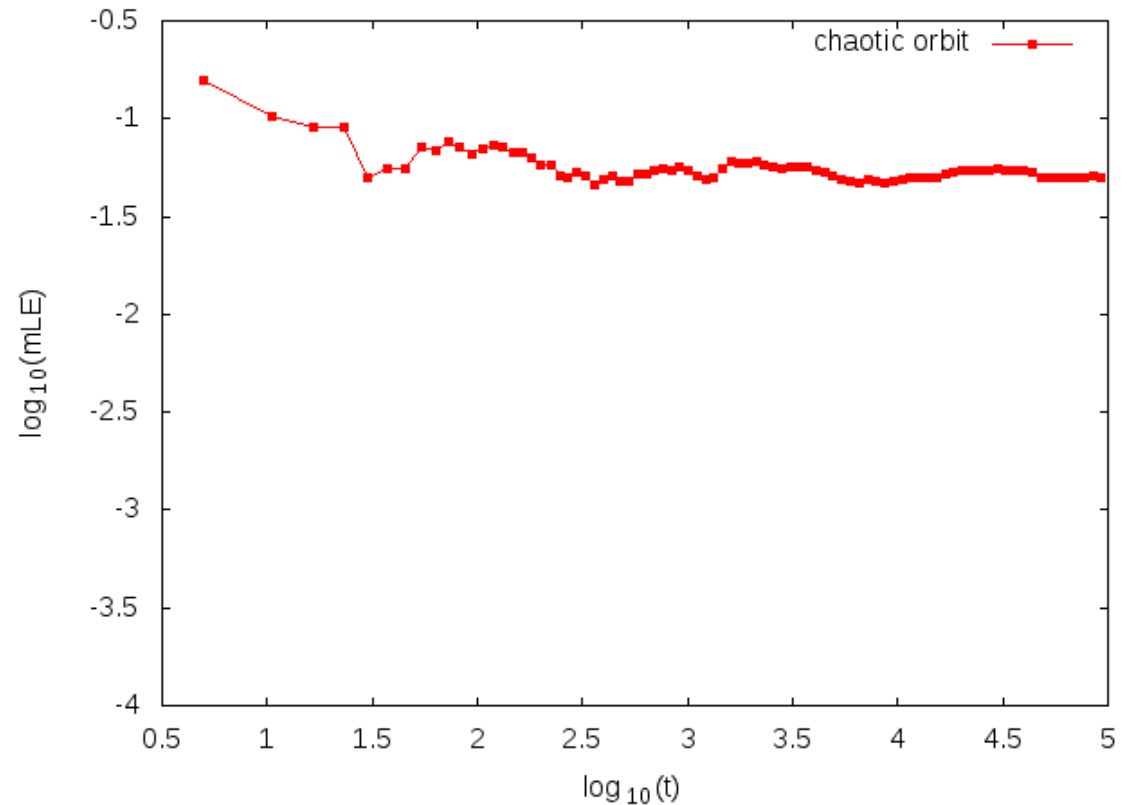
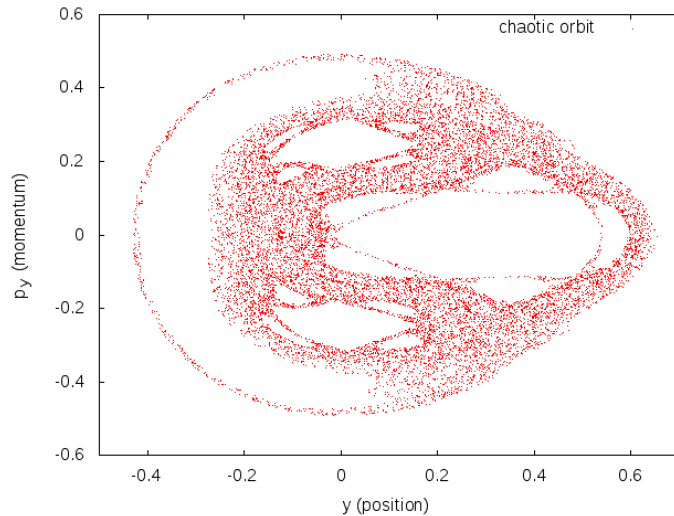
Maximum Lyapunov Exponent

Hénon-Heiles system: **Chaotic orbit**



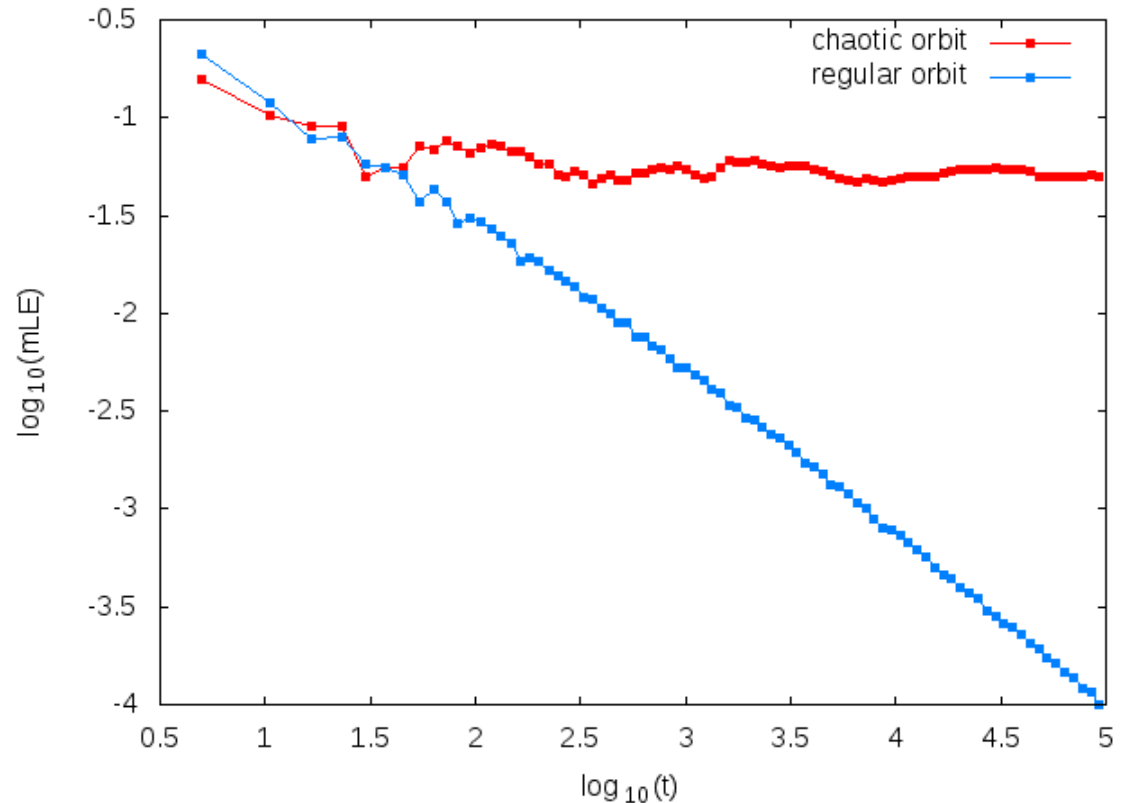
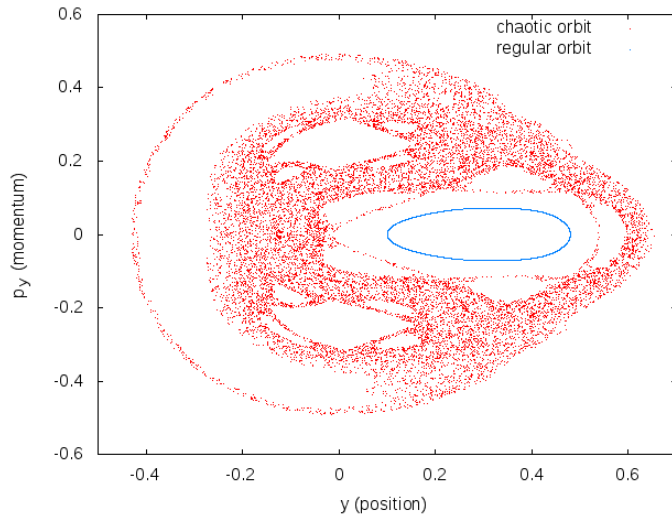
Maximum Lyapunov Exponent

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**

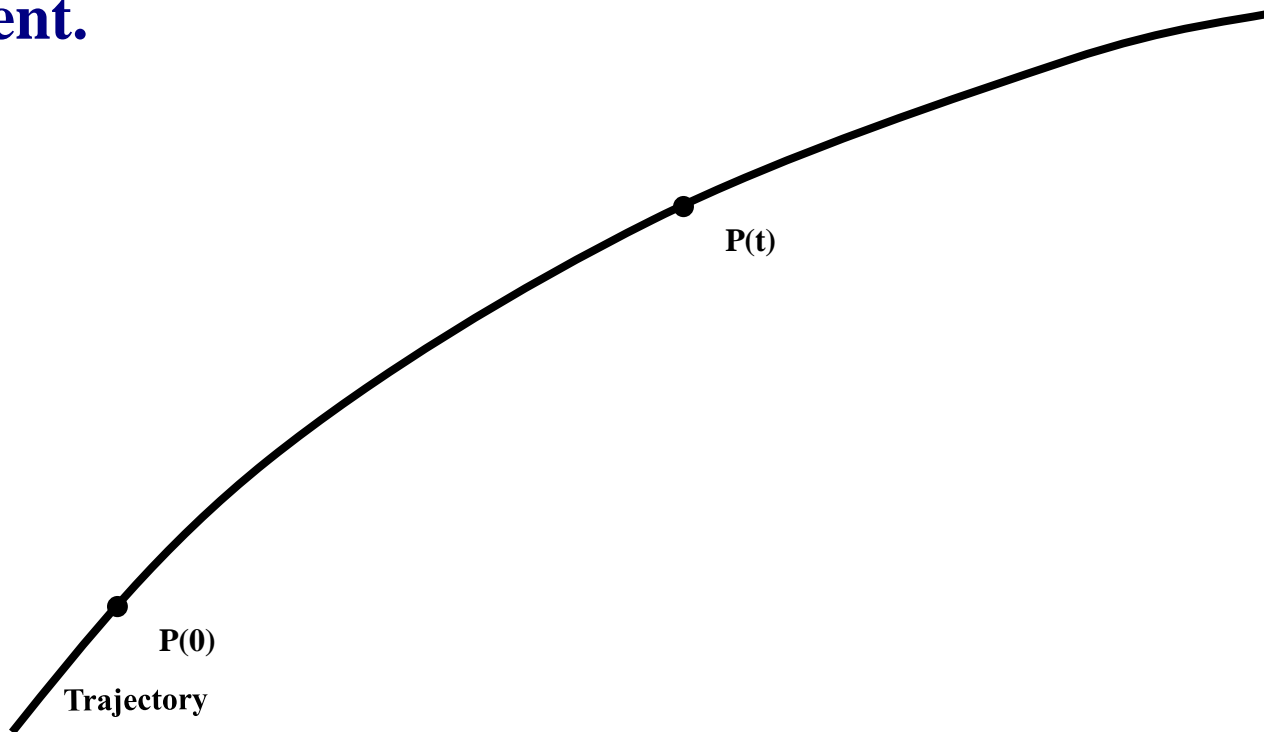
$$\text{SALI}(t) \rightarrow 0$$

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.

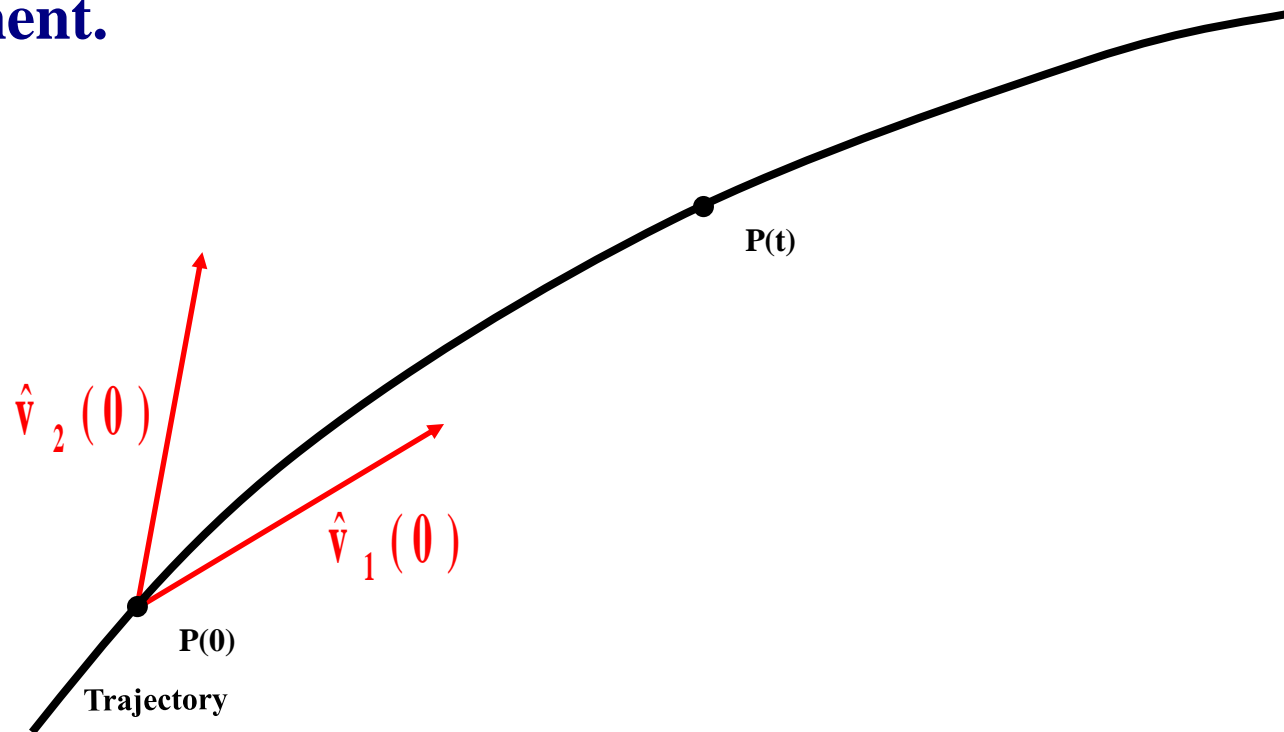
Behavior of the SALI for chaotic motion

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



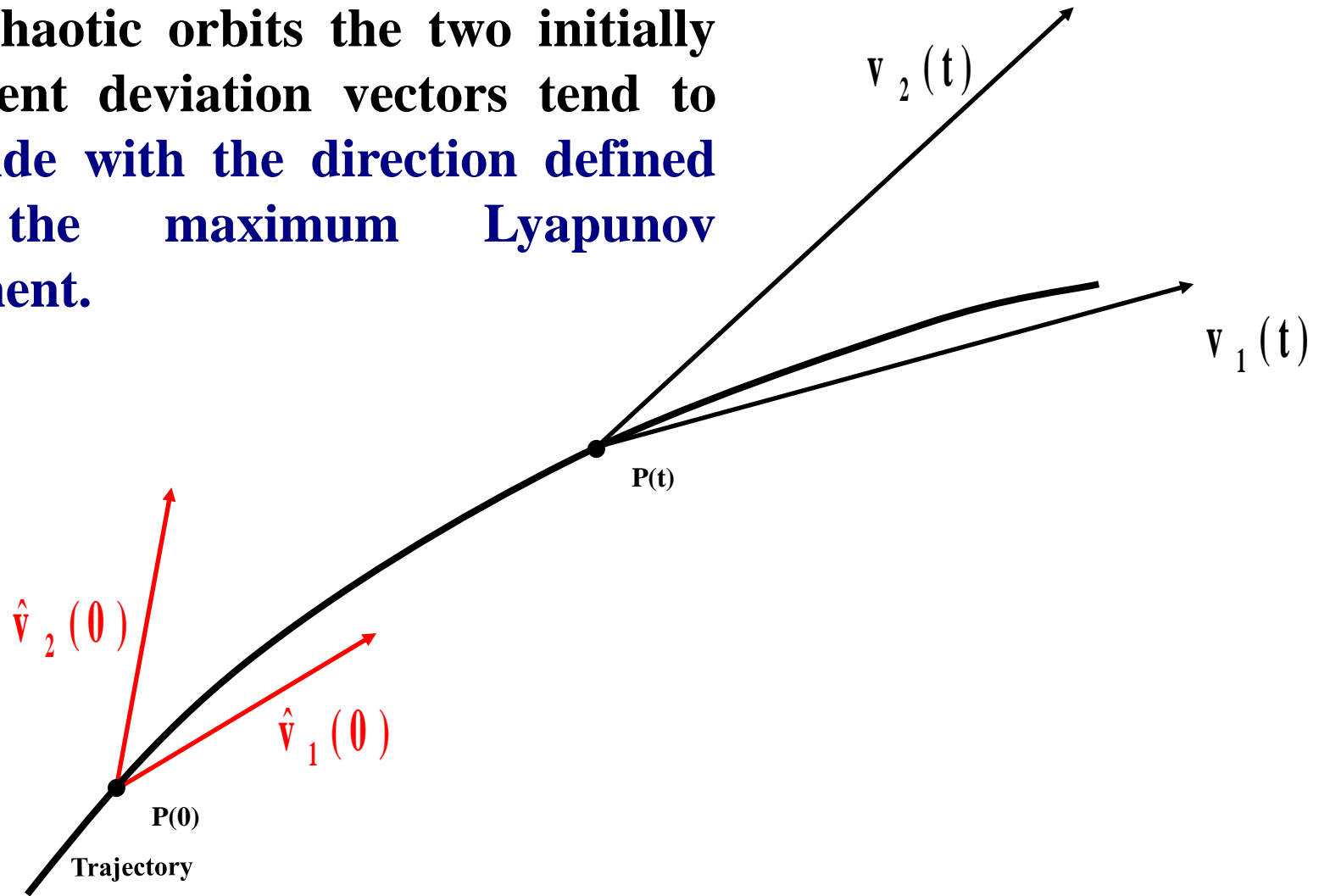
Behavior of the SALI for chaotic motion

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



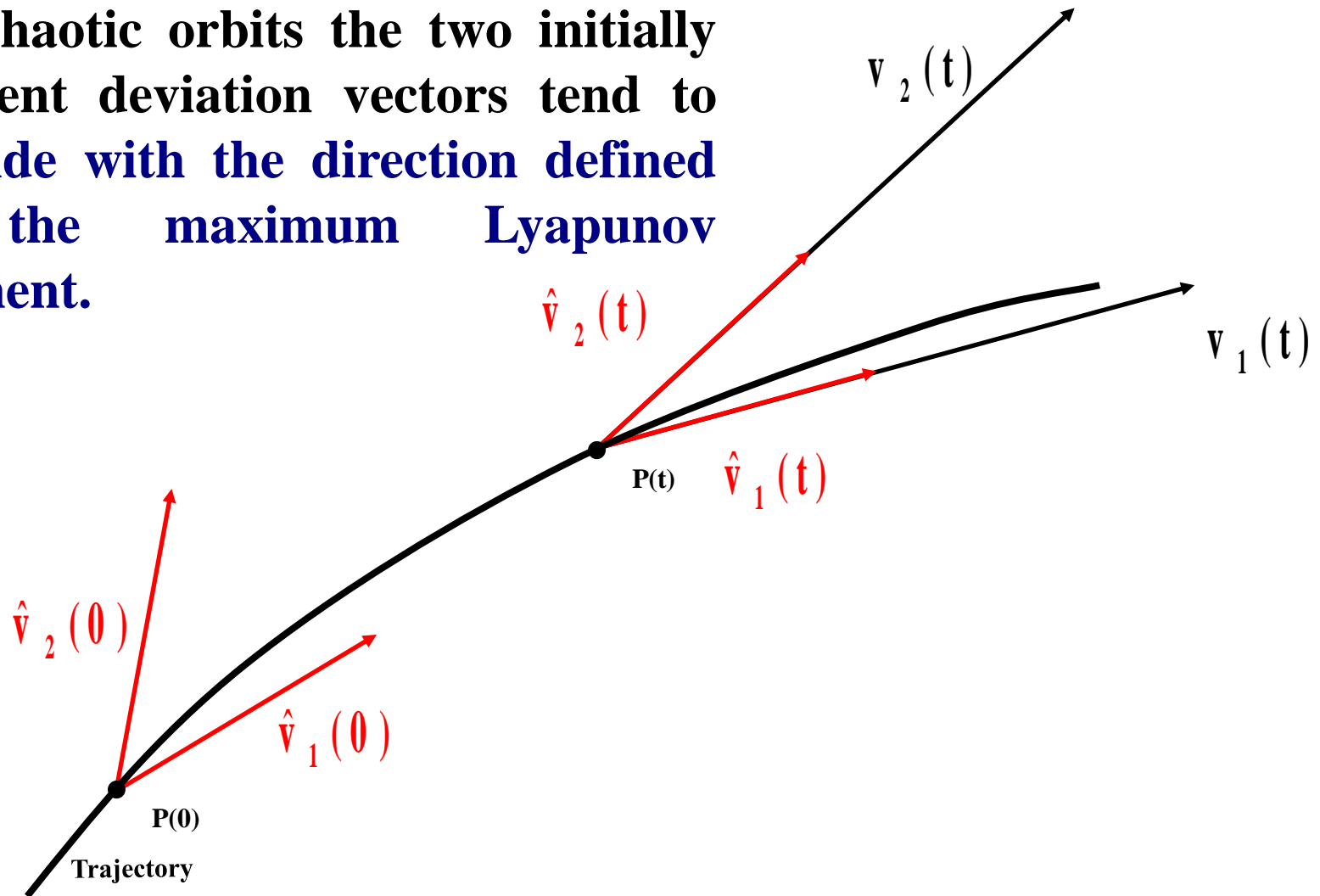
Behavior of the SALI for chaotic motion

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



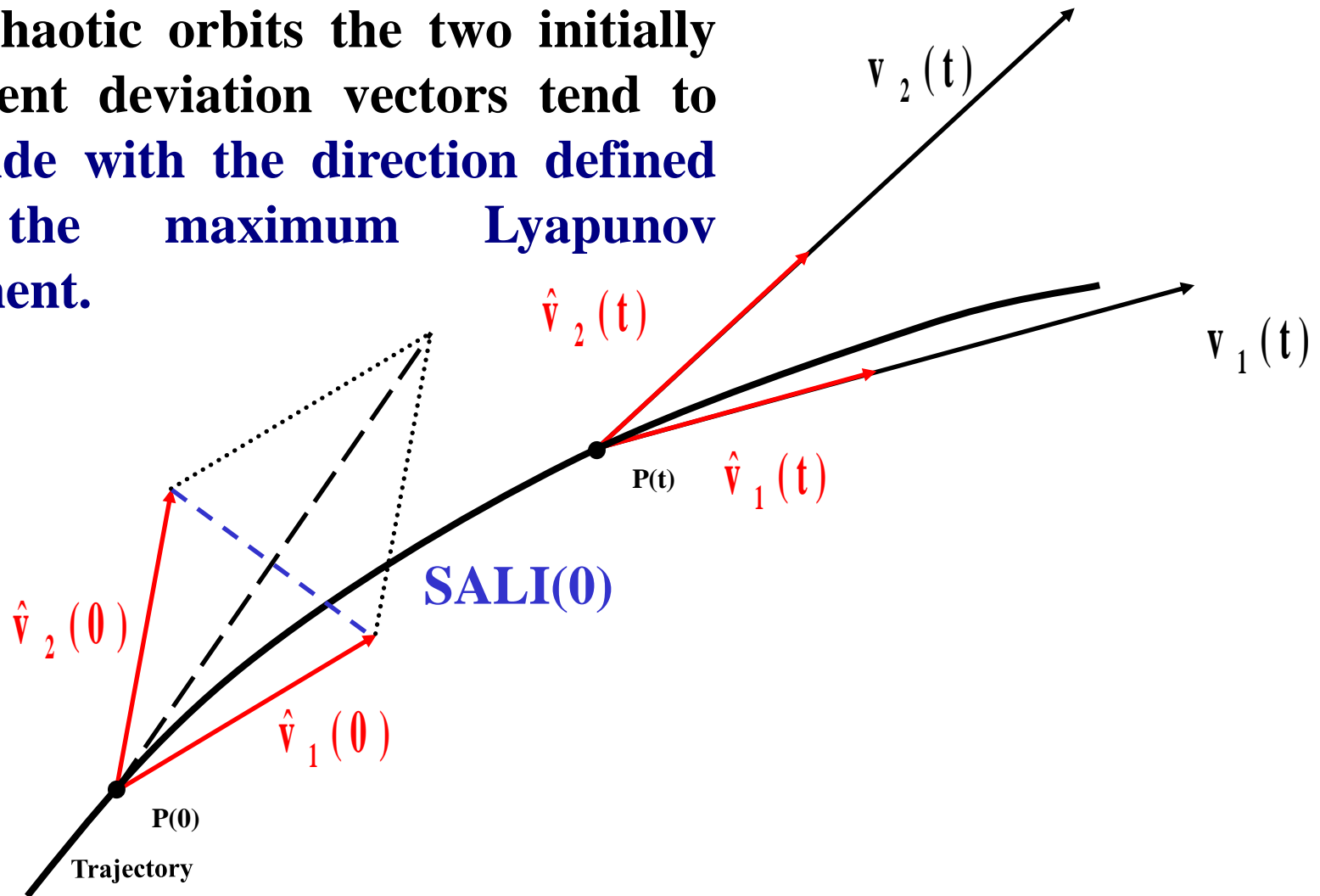
Behavior of the SALI for chaotic motion

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



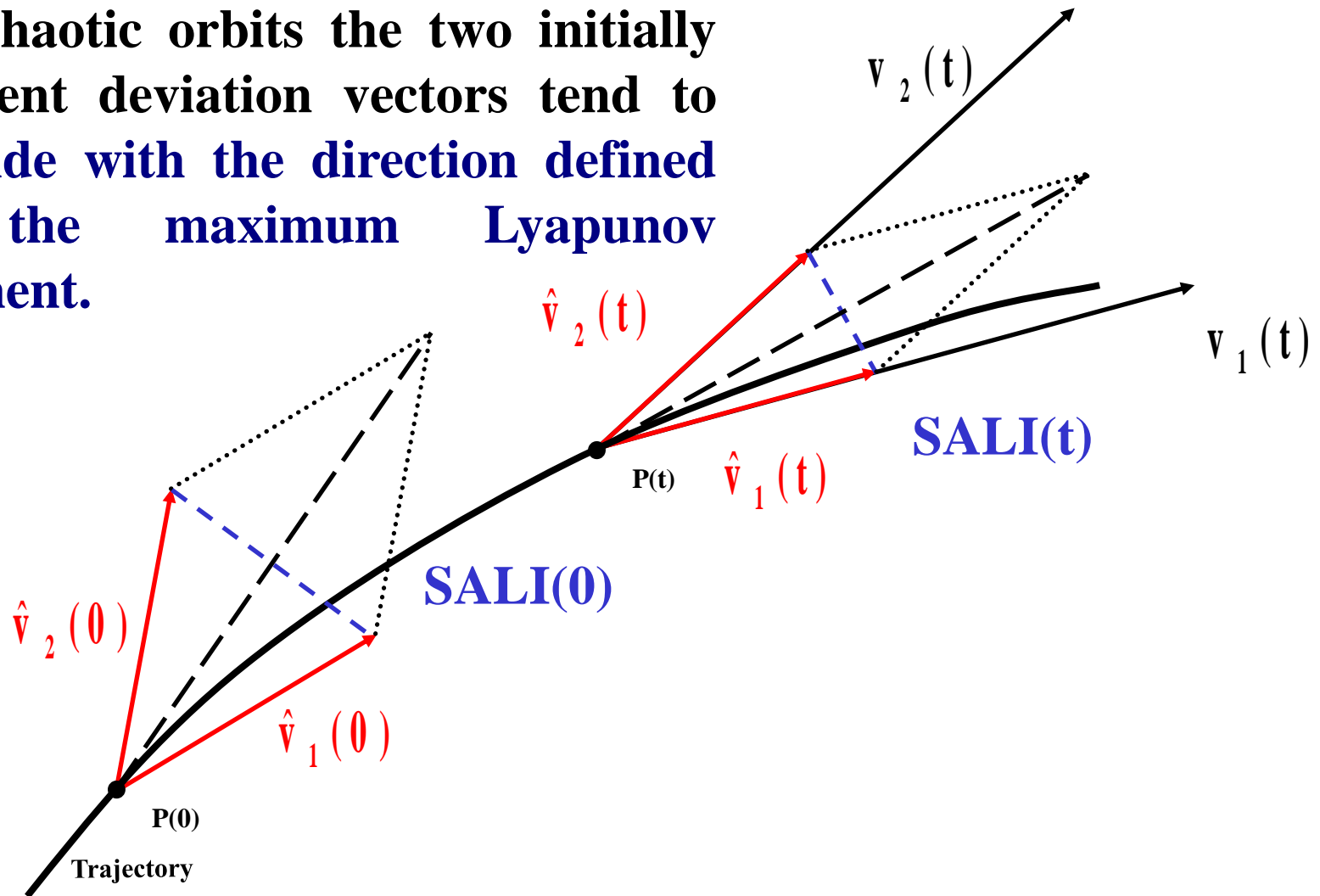
Behavior of the SALI for chaotic motion

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



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.

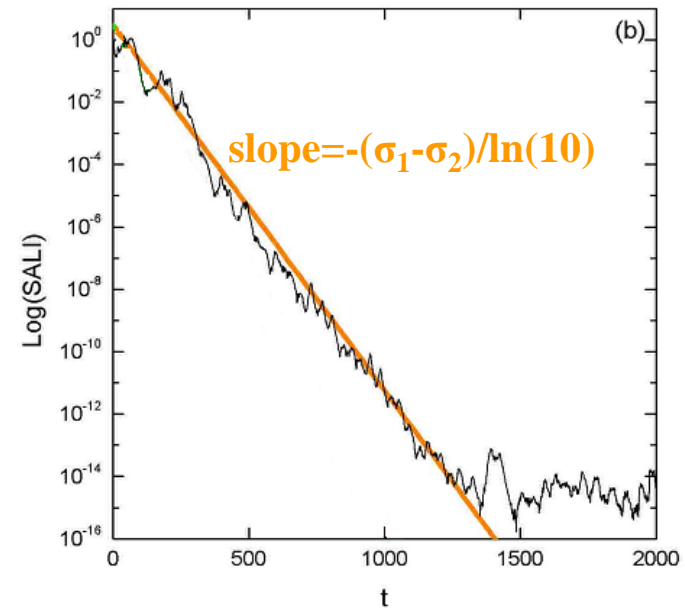
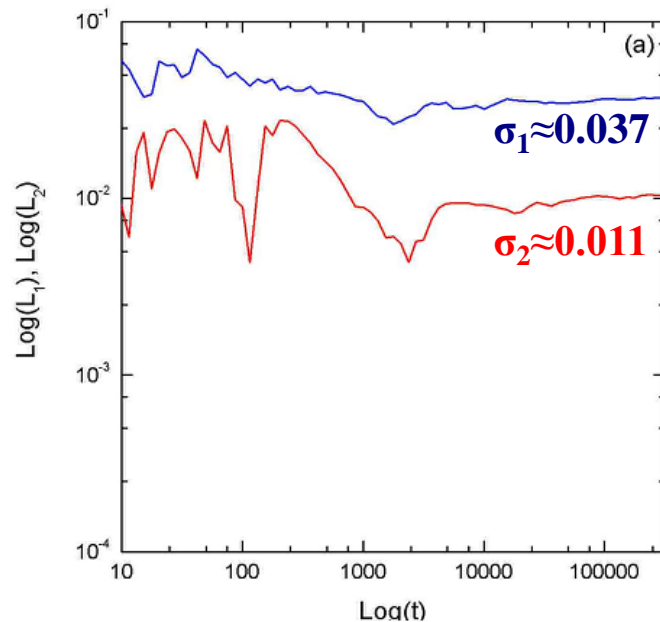


Behavior of the SALI for chaotic motion

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

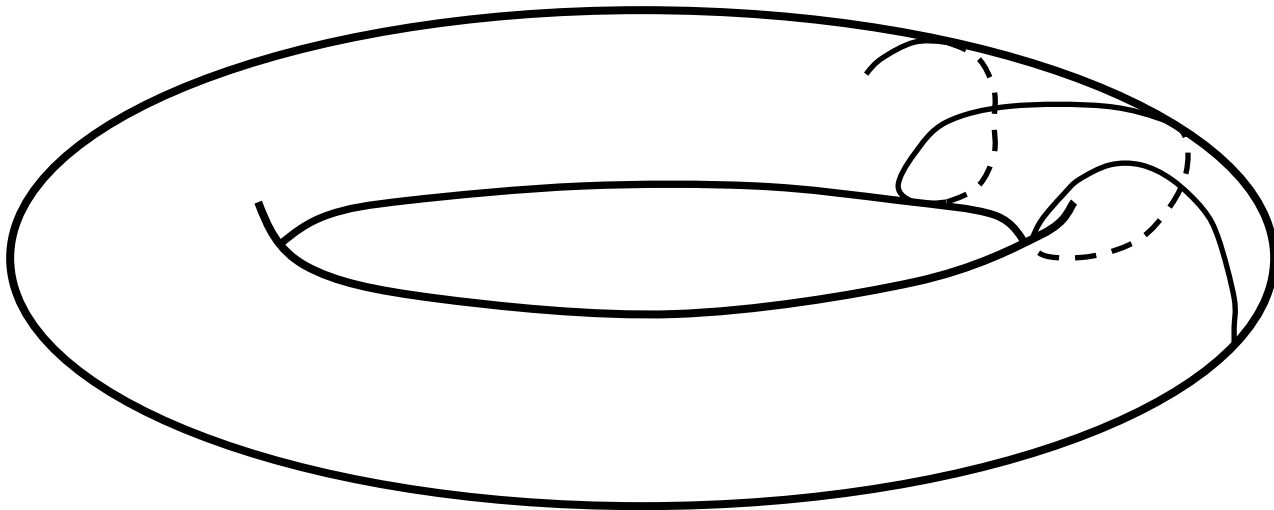


Behavior of the SALI for **regular motion**

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

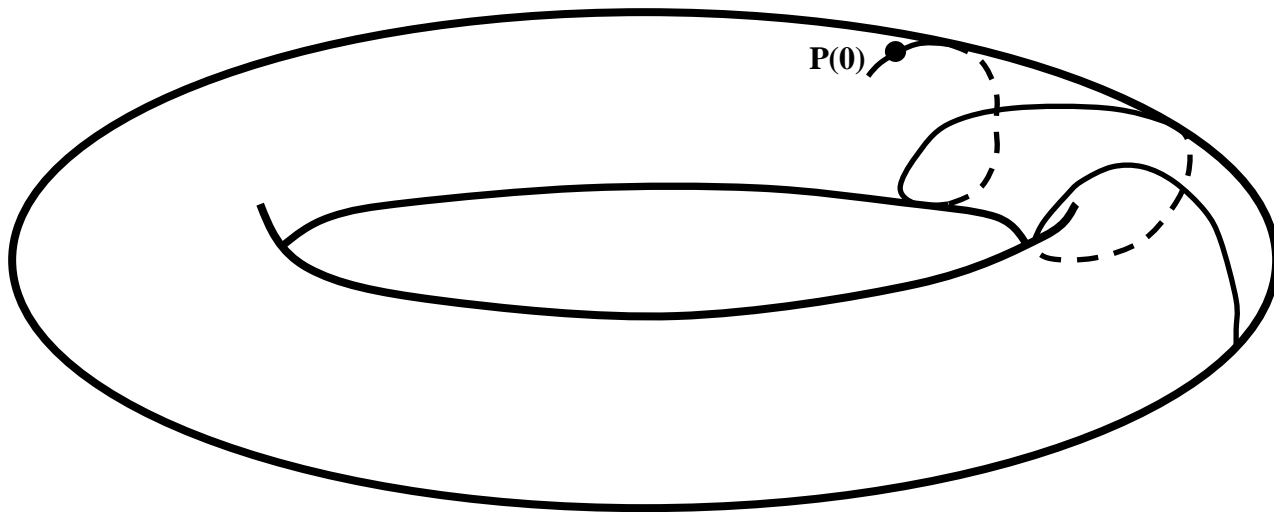
Behavior of the SALI for **regular motion**

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



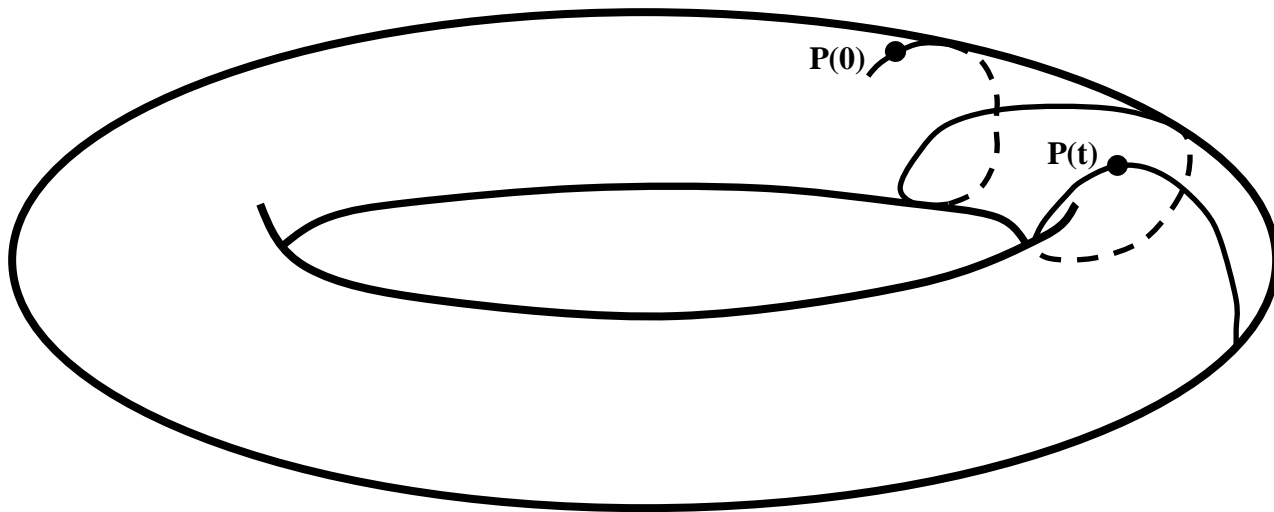
Behavior of the SALI for **regular motion**

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



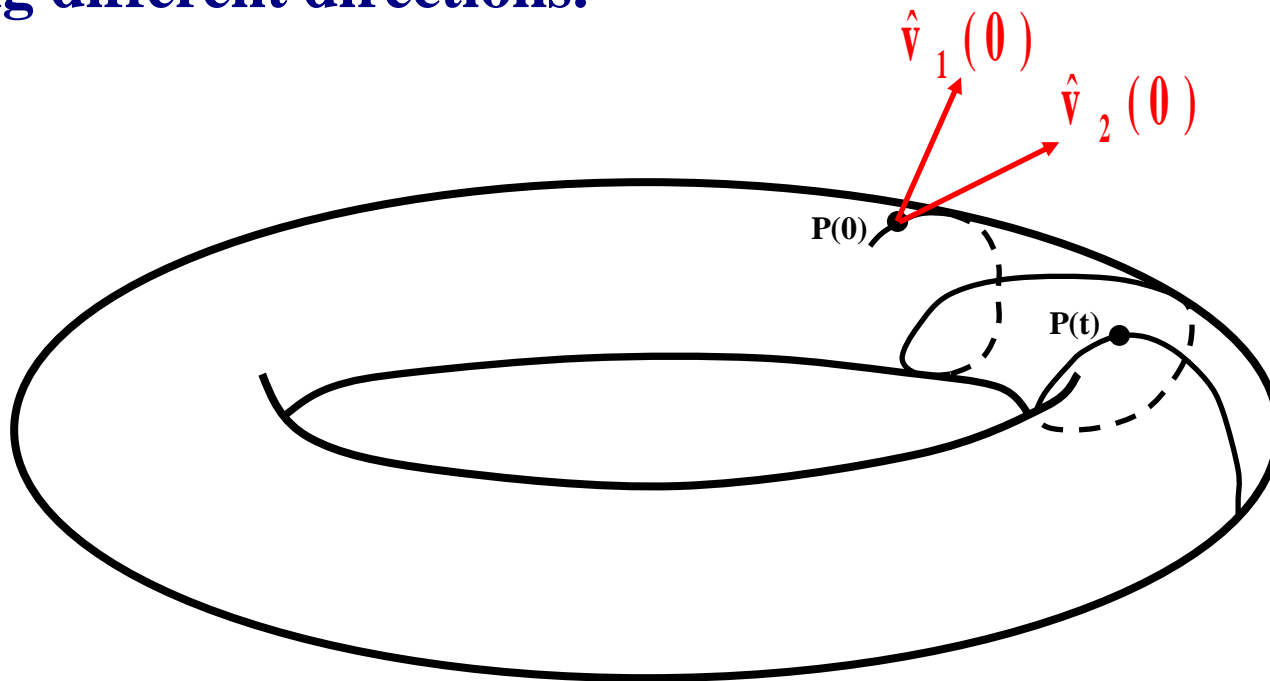
Behavior of the SALI for **regular motion**

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



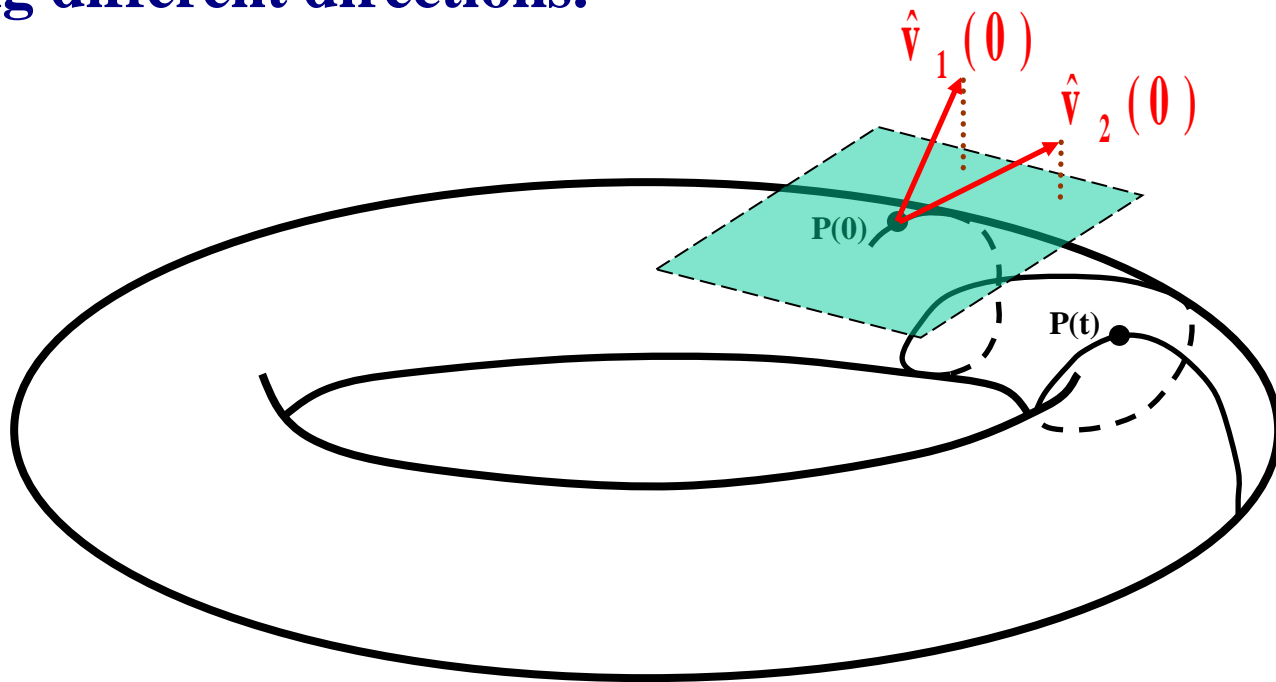
Behavior of the SALI for **regular motion**

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



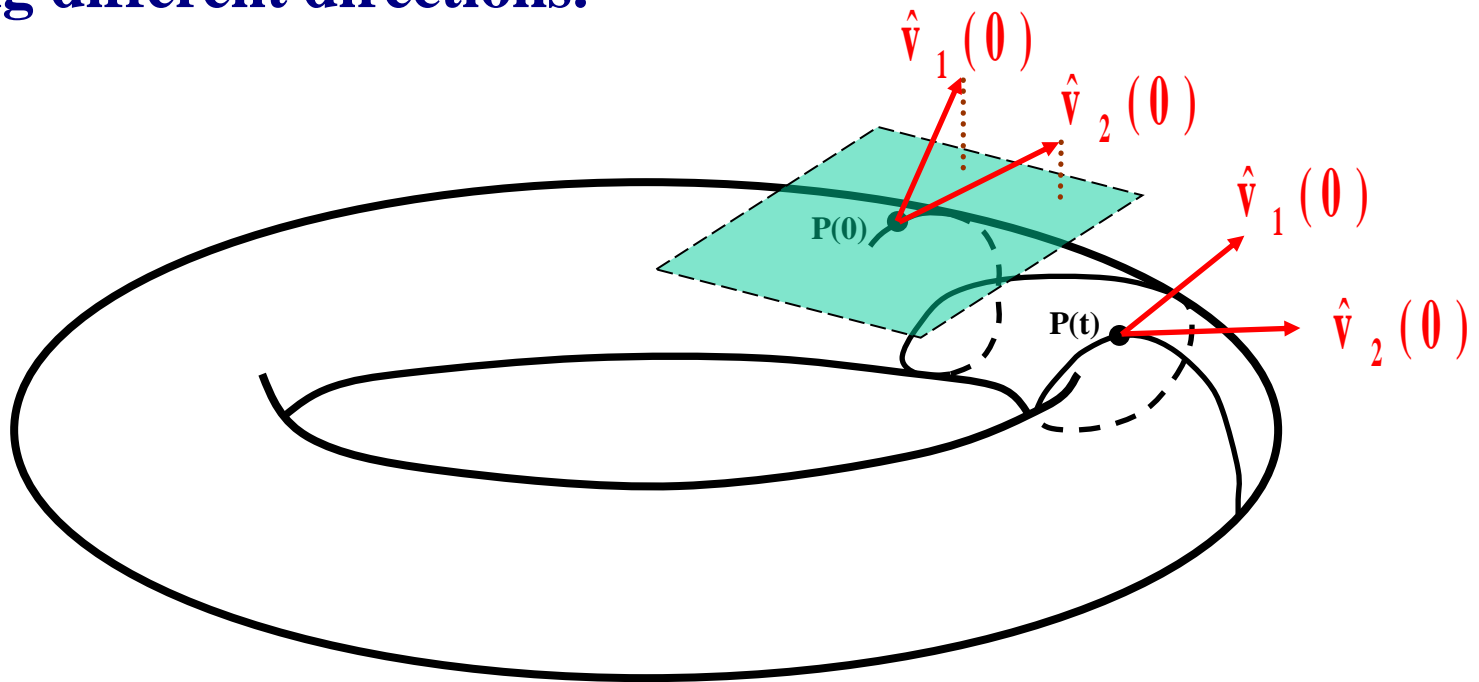
Behavior of the SALI for **regular motion**

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



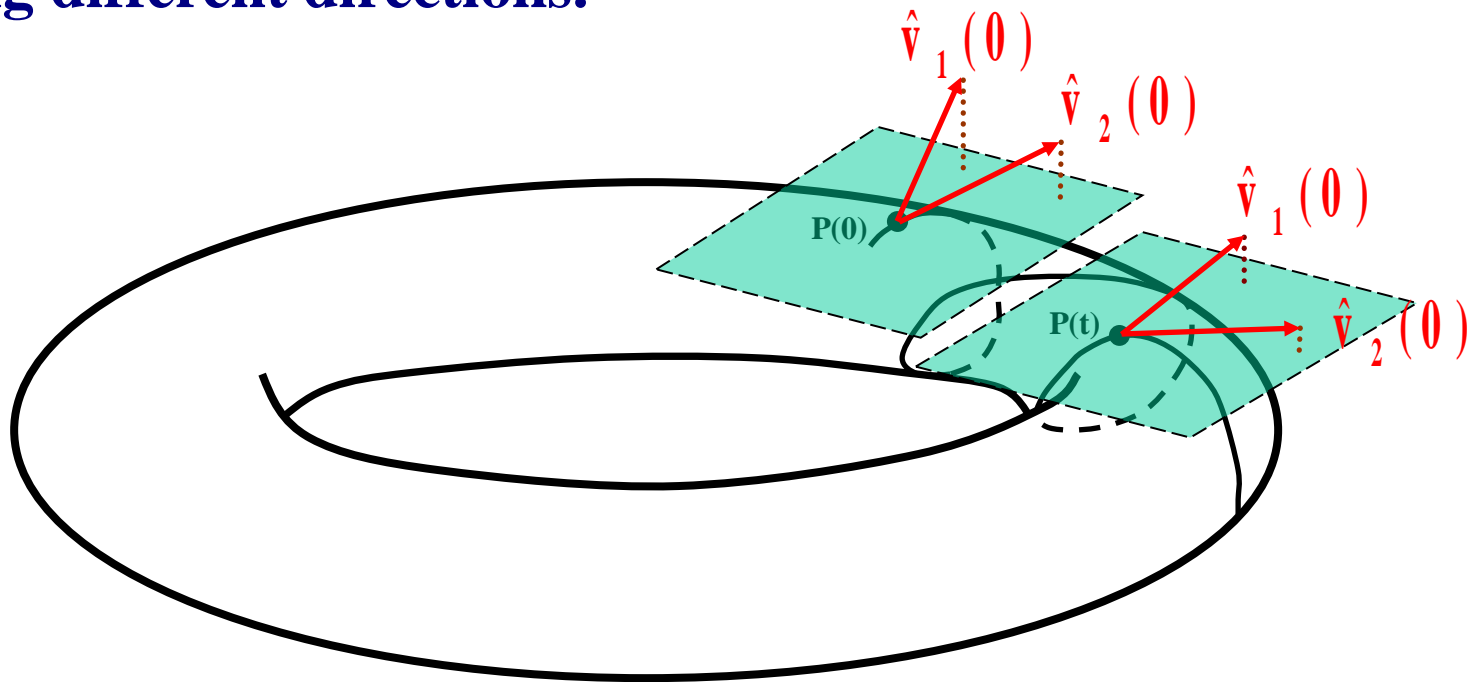
Behavior of the SALI for **regular motion**

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:

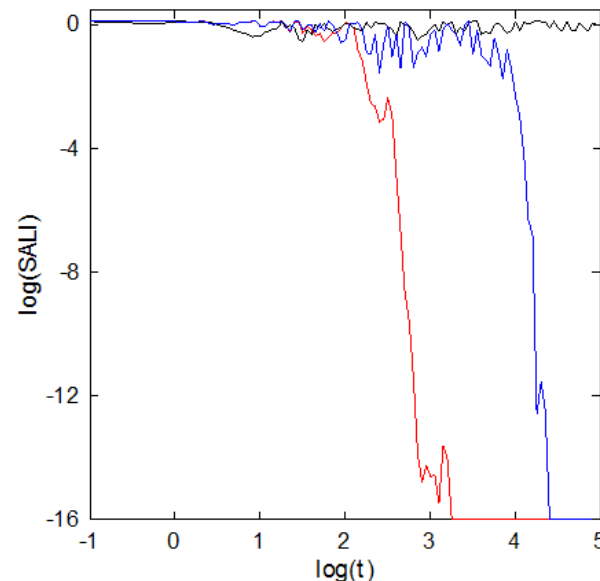
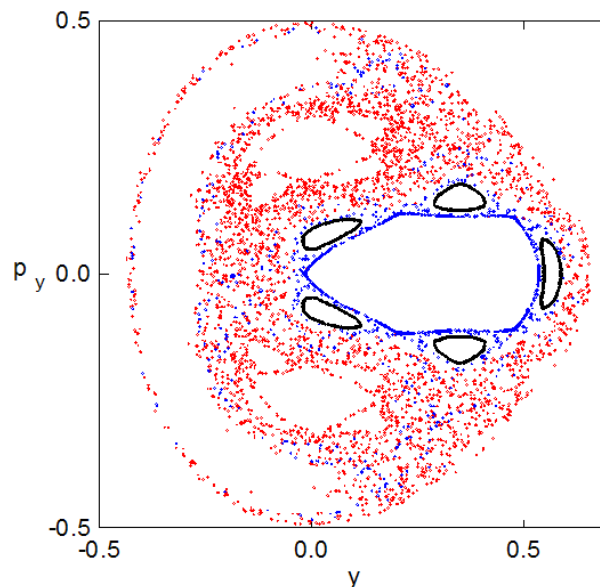
$$H_2 = \frac{1}{2}(p_x^2 + p_y^2) + \frac{1}{2}(x^2 + y^2) + x^2y - \frac{1}{3}y^3$$

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

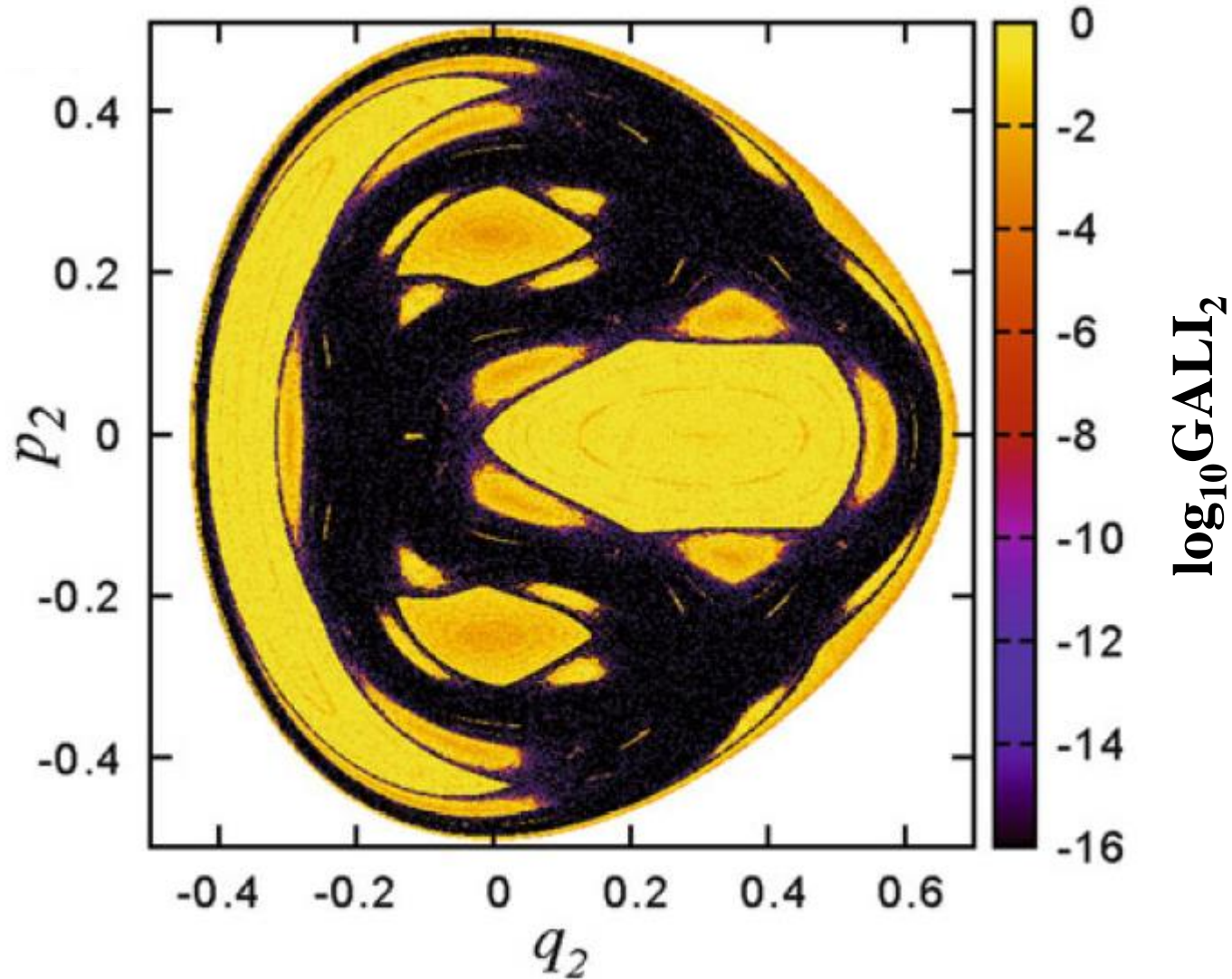
Regular orbit, $x=0$, $y=0.55$, $p_x=0.2417$, $p_y=0$

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



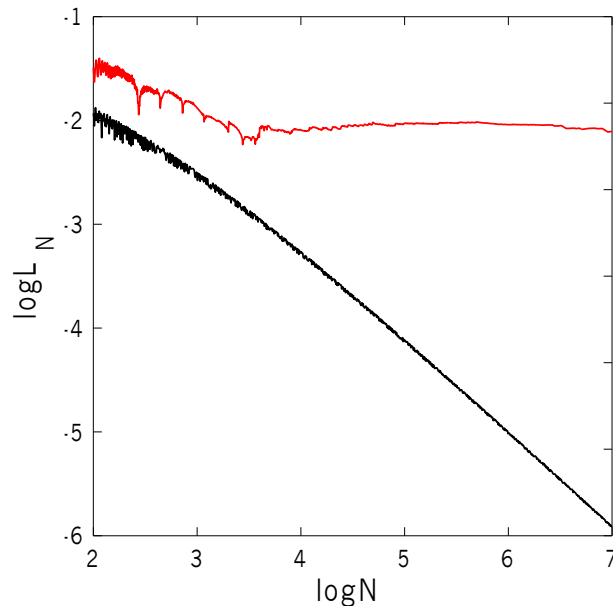
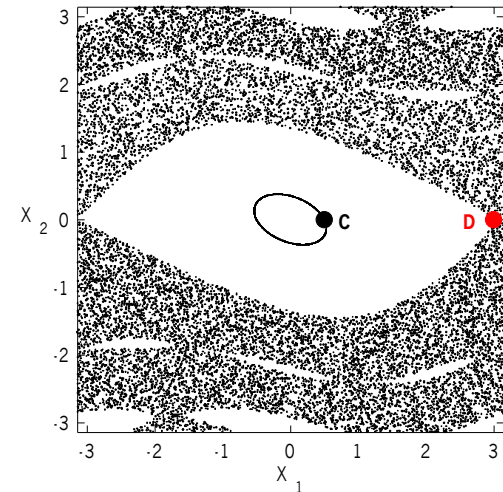
Applications – 4D map

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

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 $x_1=3$, $x_2=0$, $x_3=0.5$, $x_4=0$.



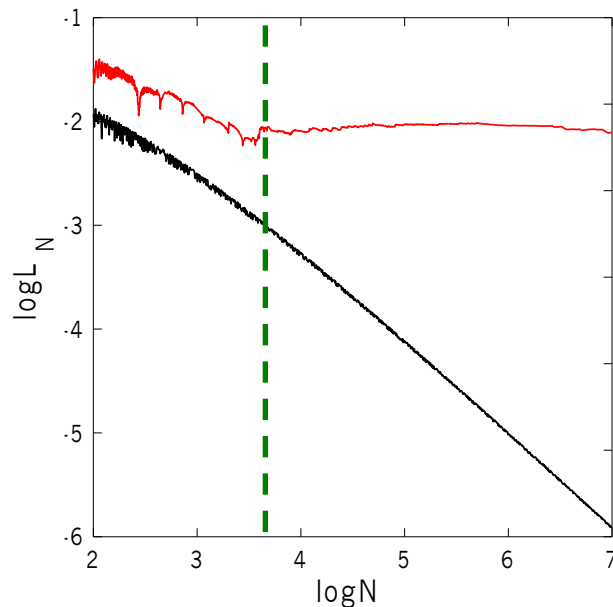
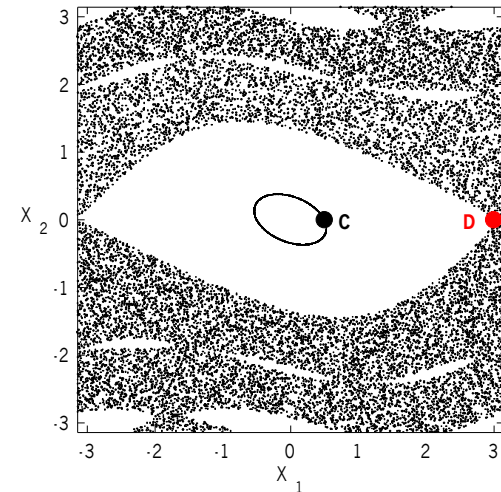
Applications – 4D map

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

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 $x_1=3$, $x_2=0$, $x_3=0.5$, $x_4=0$.



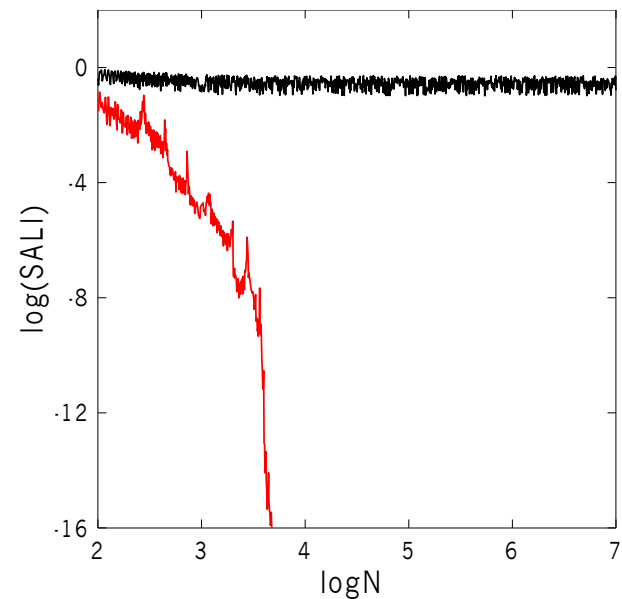
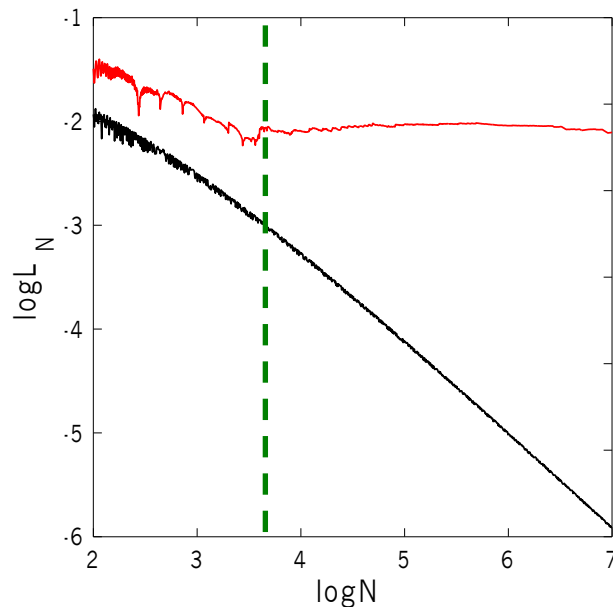
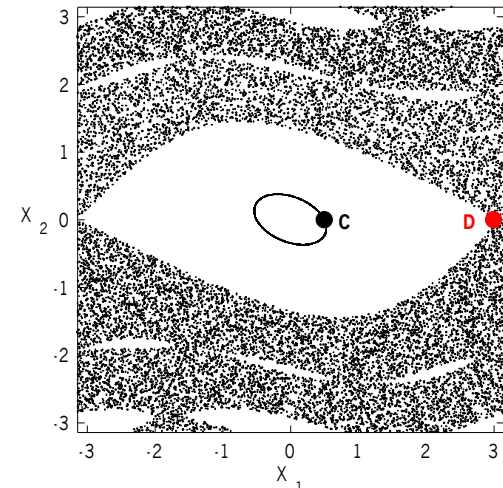
Applications – 4D map

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

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 $x_1=3$, $x_2=0$, $x_3=0.5$, $x_4=0$.



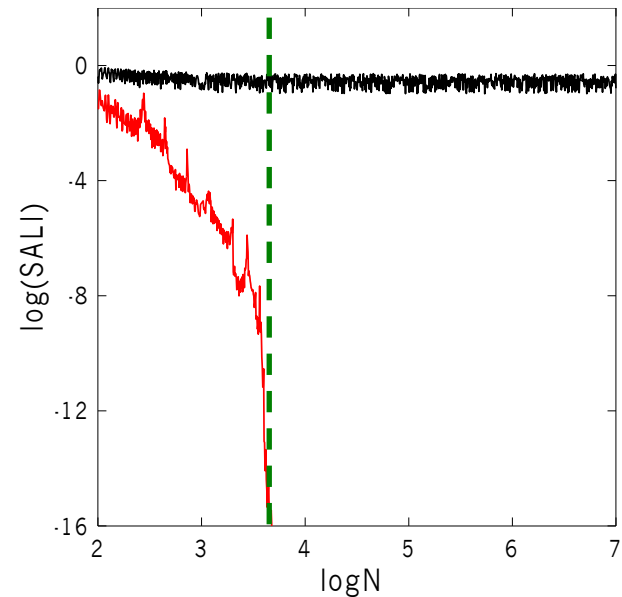
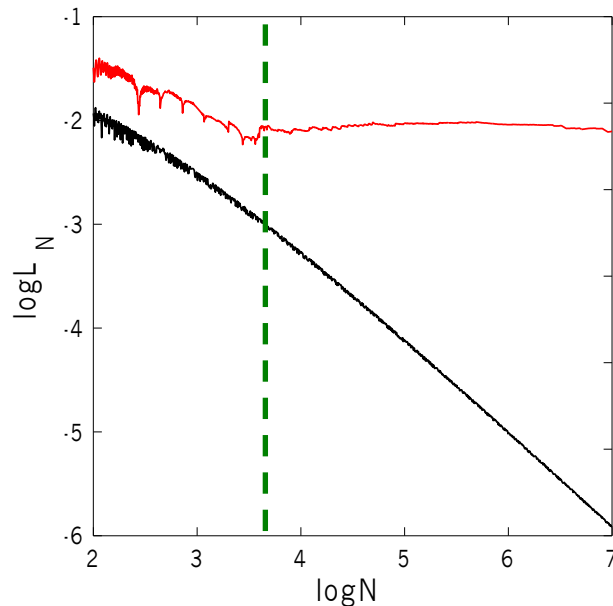
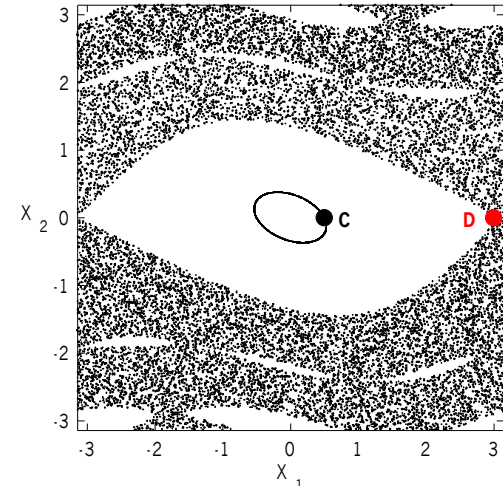
Applications – 4D map

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

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 $x_1=3$, $x_2=0$, $x_3=0.5$, $x_4=0$.



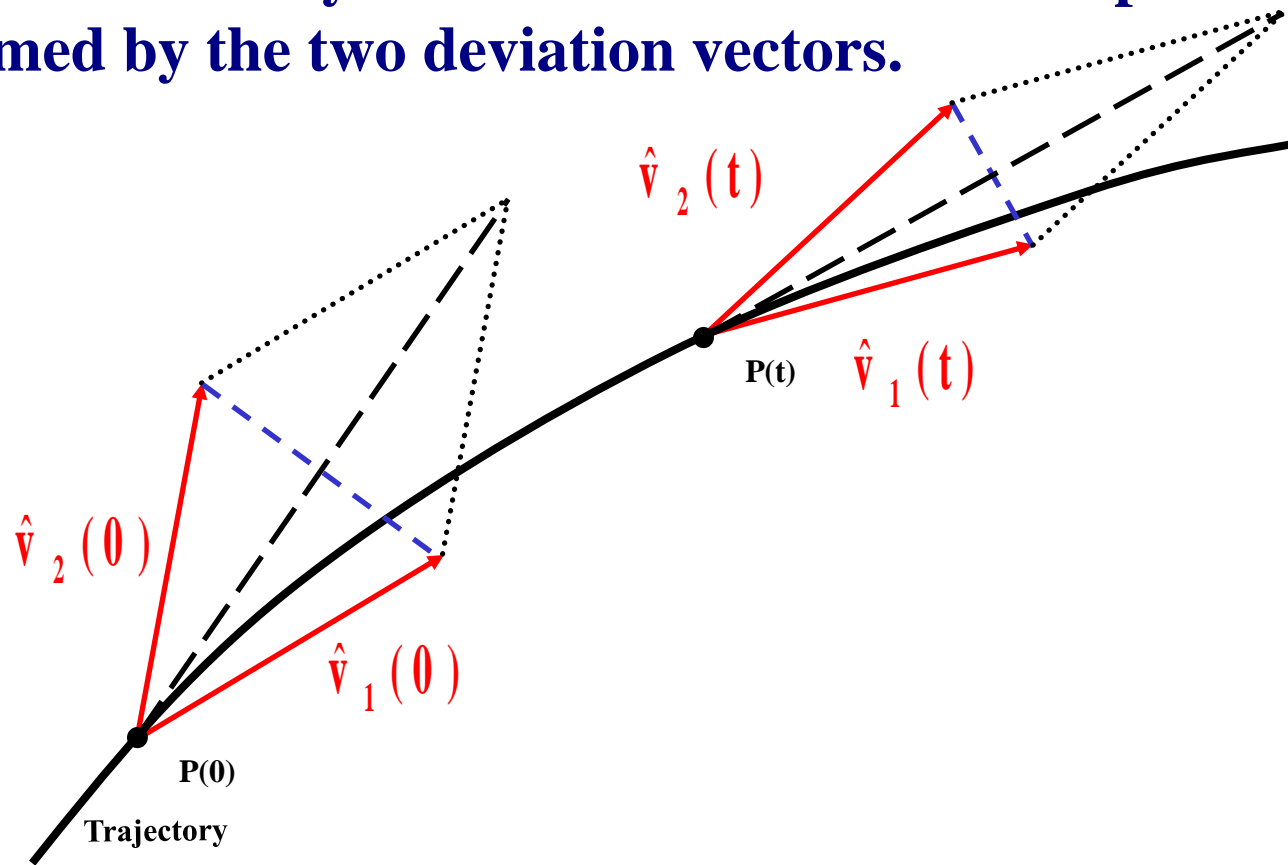
The Generalized ALignment Indices (GALIs) method

Definition of the Generalized Alignment Index (GALI)

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

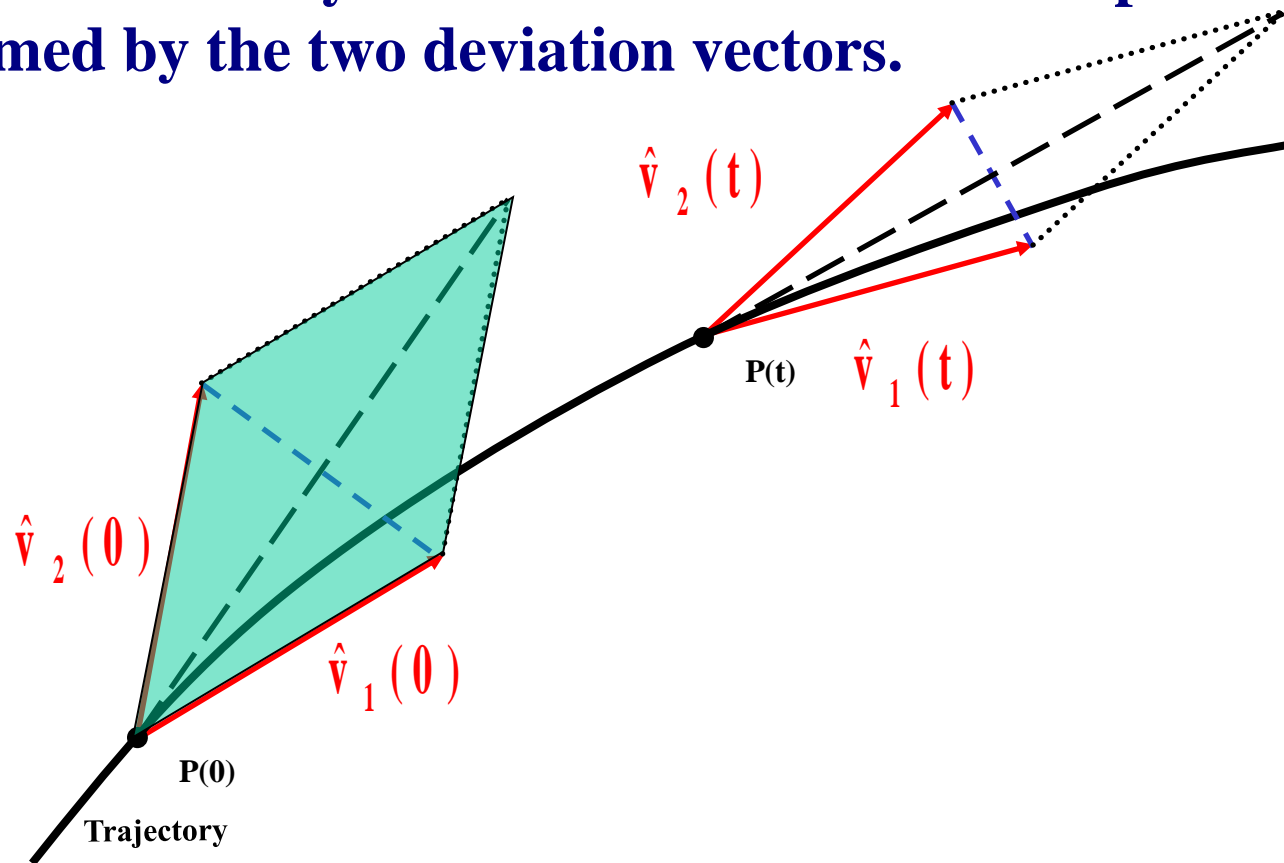
Definition of the Generalized Alignment Index (GALI)

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



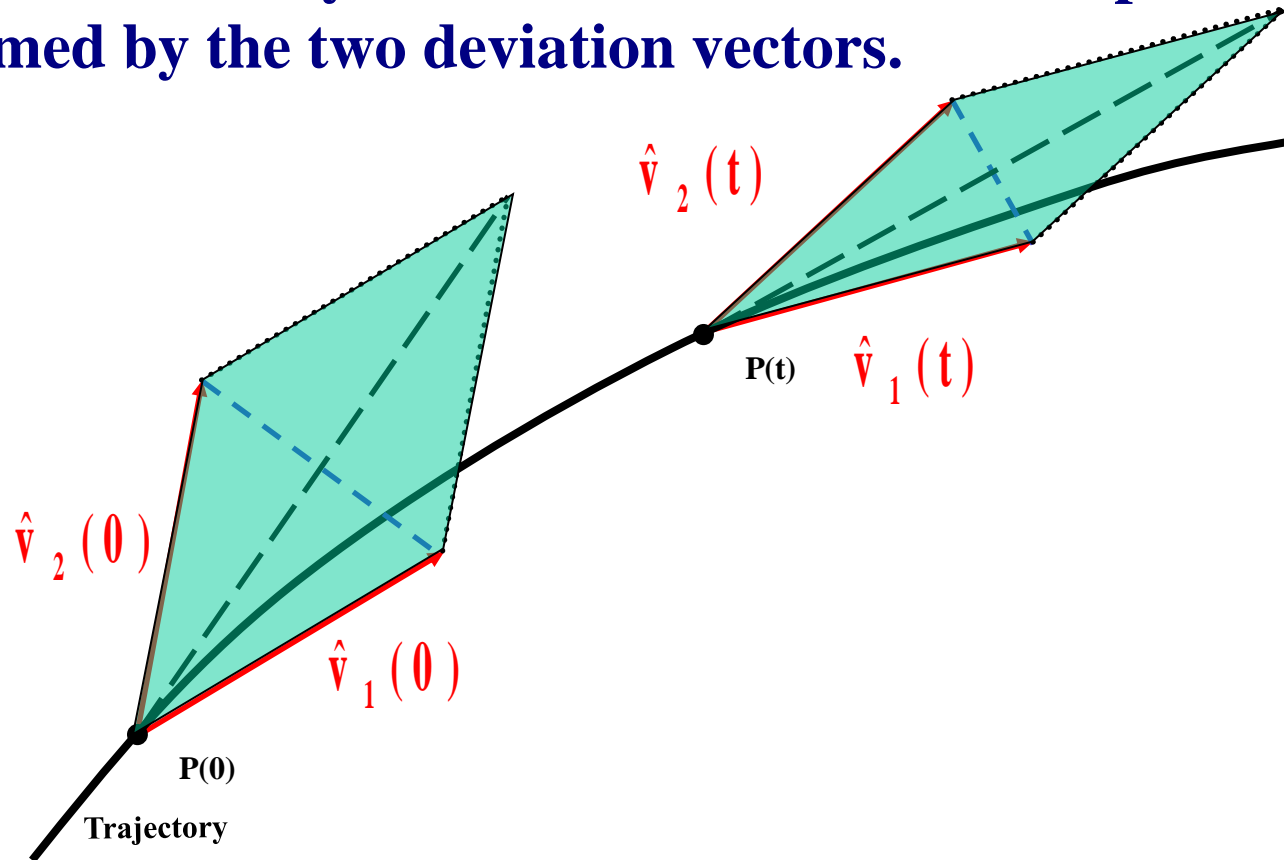
Definition of the Generalized Alignment Index (GALI)

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



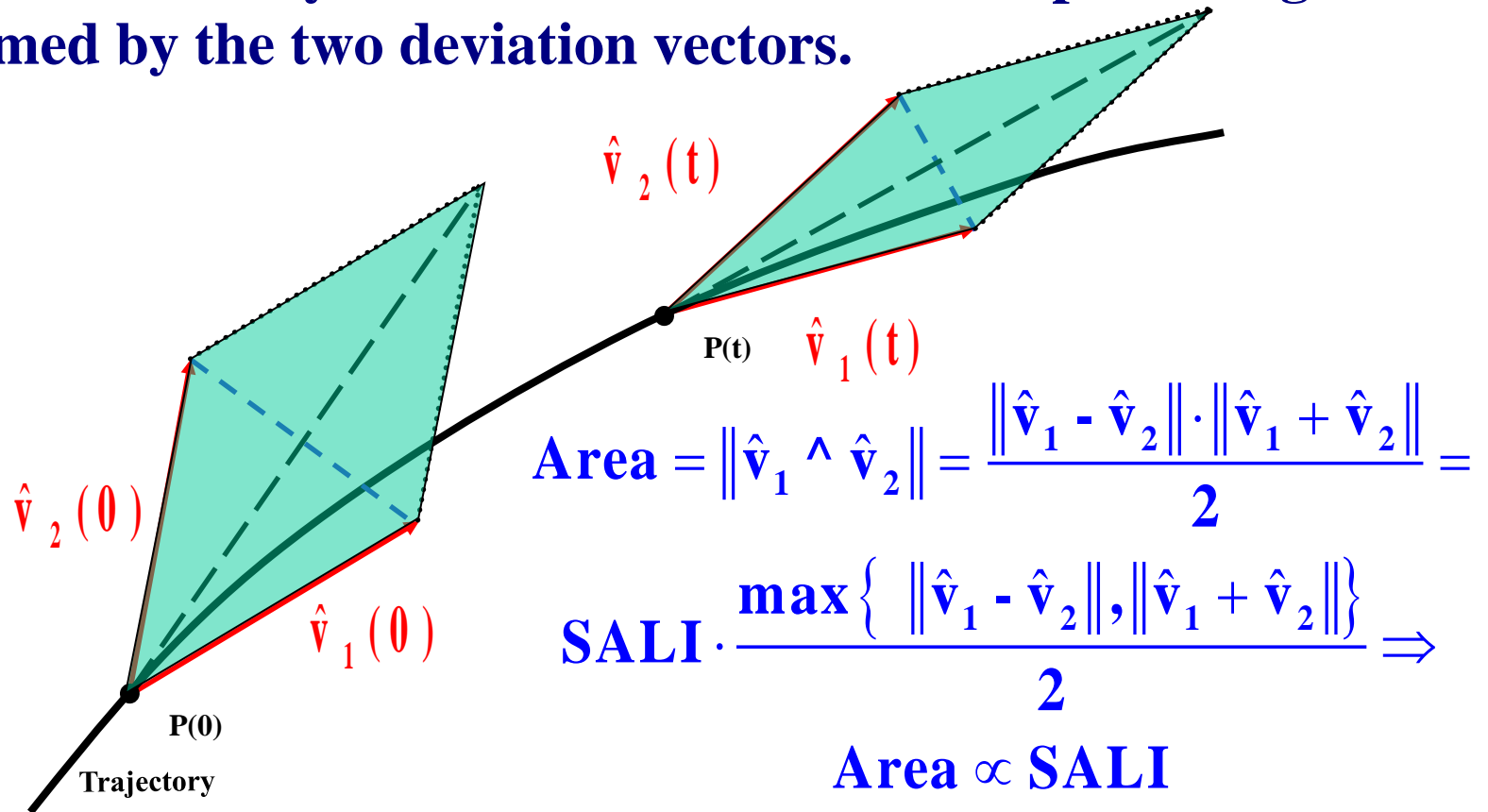
Definition of the Generalized Alignment Index (GALI)

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 :

$$GALI_k(t) = \left\| \hat{v}_1(t) \wedge \hat{v}_2(t) \wedge \dots \wedge \hat{v}_k(t) \right\|$$

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 $\sigma_1, \sigma_2, \dots, \sigma_k$:

$$GALI_k(t) \propto e^{-[(\sigma_1 - \sigma_2) + (\sigma_1 - \sigma_3) + \dots + (\sigma_1 - \sigma_k)]t}$$

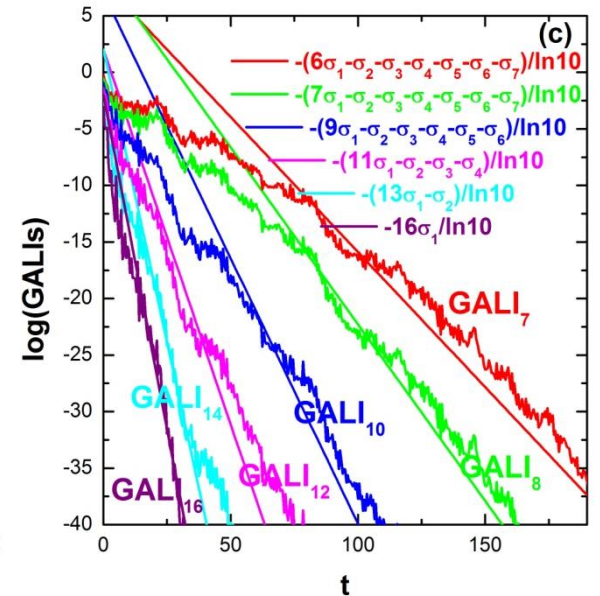
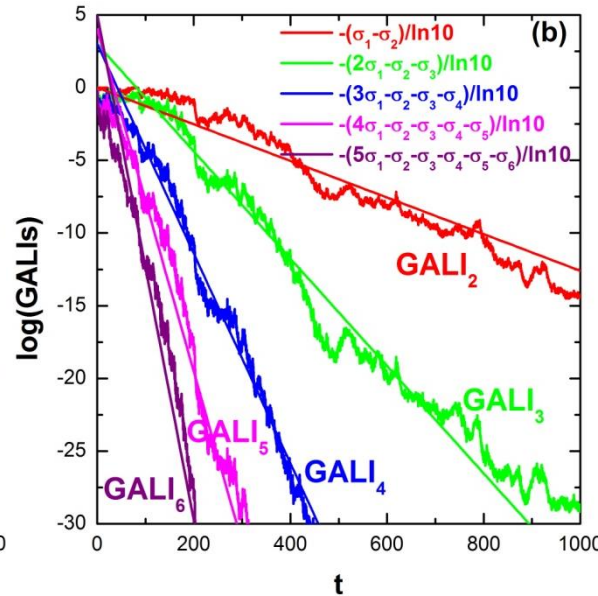
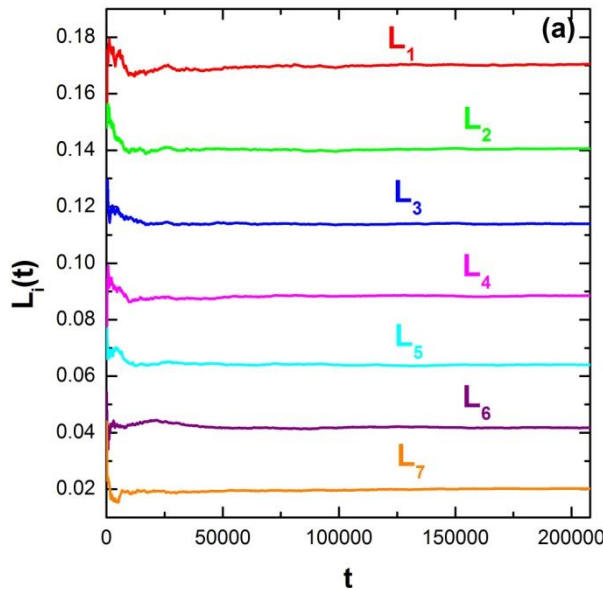
The above relation is valid even if some Lyapunov exponents are equal, or very close to each other.

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.):

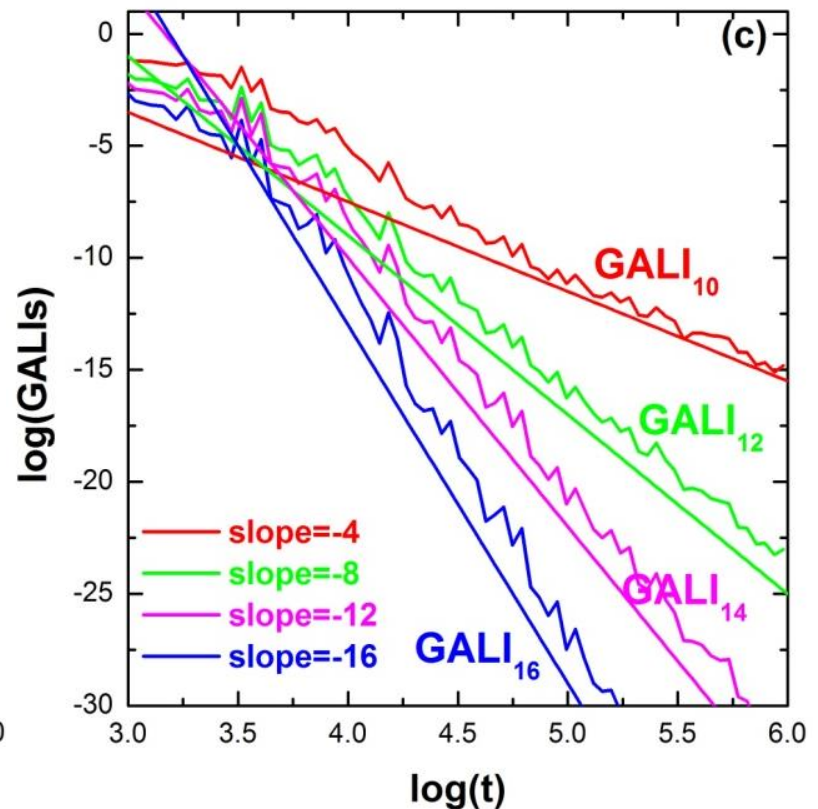
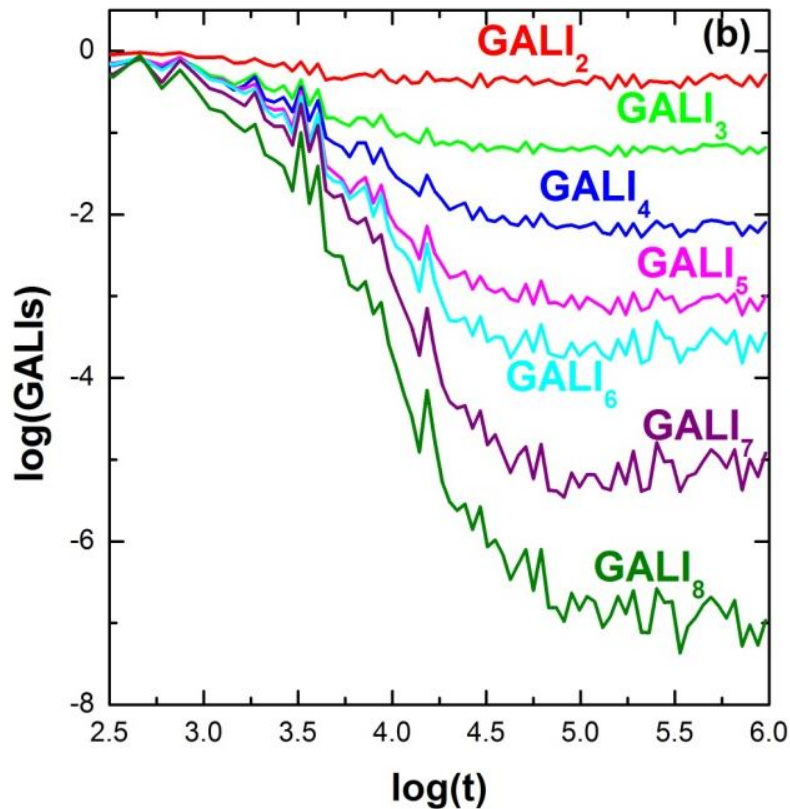
$$\text{GALI}_k(t) \propto \begin{cases} \text{constant} & \text{if } 2 \leq k \leq s \\ \frac{1}{t^{k-s}} & \text{if } s < k \leq 2N - s \\ \frac{1}{t^{2(k-N)}} & \text{if } 2N - s < k \leq 2N \end{cases}$$

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

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

Behavior of the $GALI_k$ for regular motion

N=8 FPU system



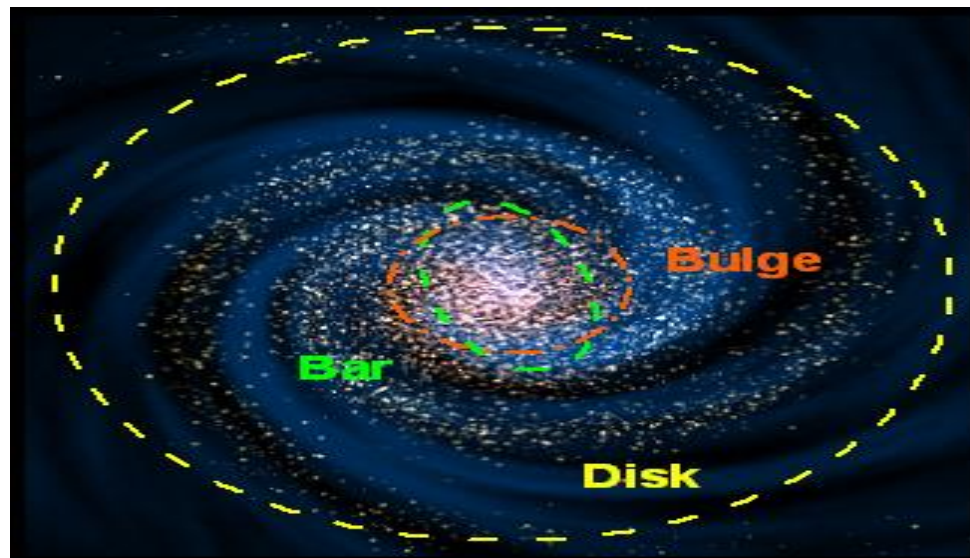
A time-dependent Hamiltonian system

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)

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

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

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

Its density is:

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

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:

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

ii) Miyamoto–Nagai disc:

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

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)

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

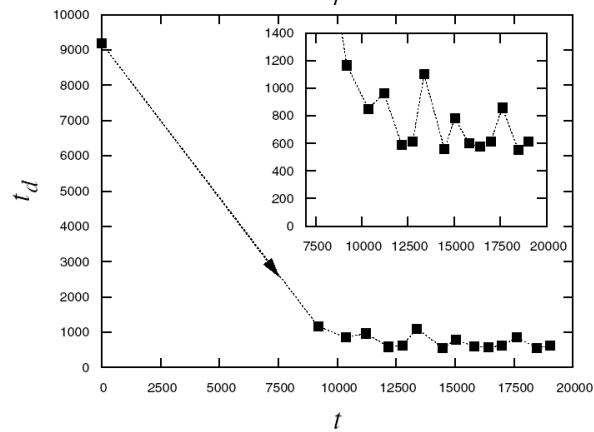
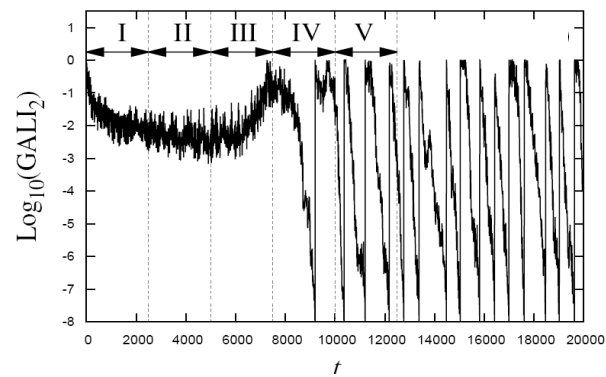
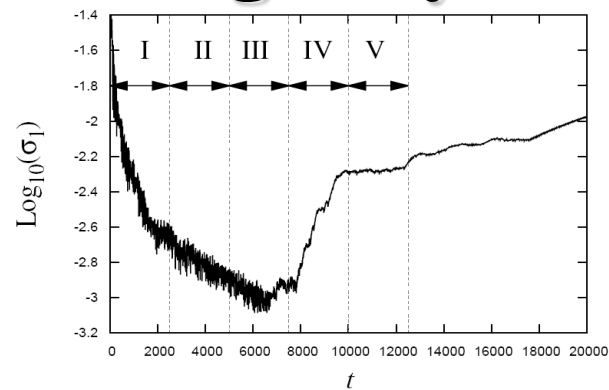
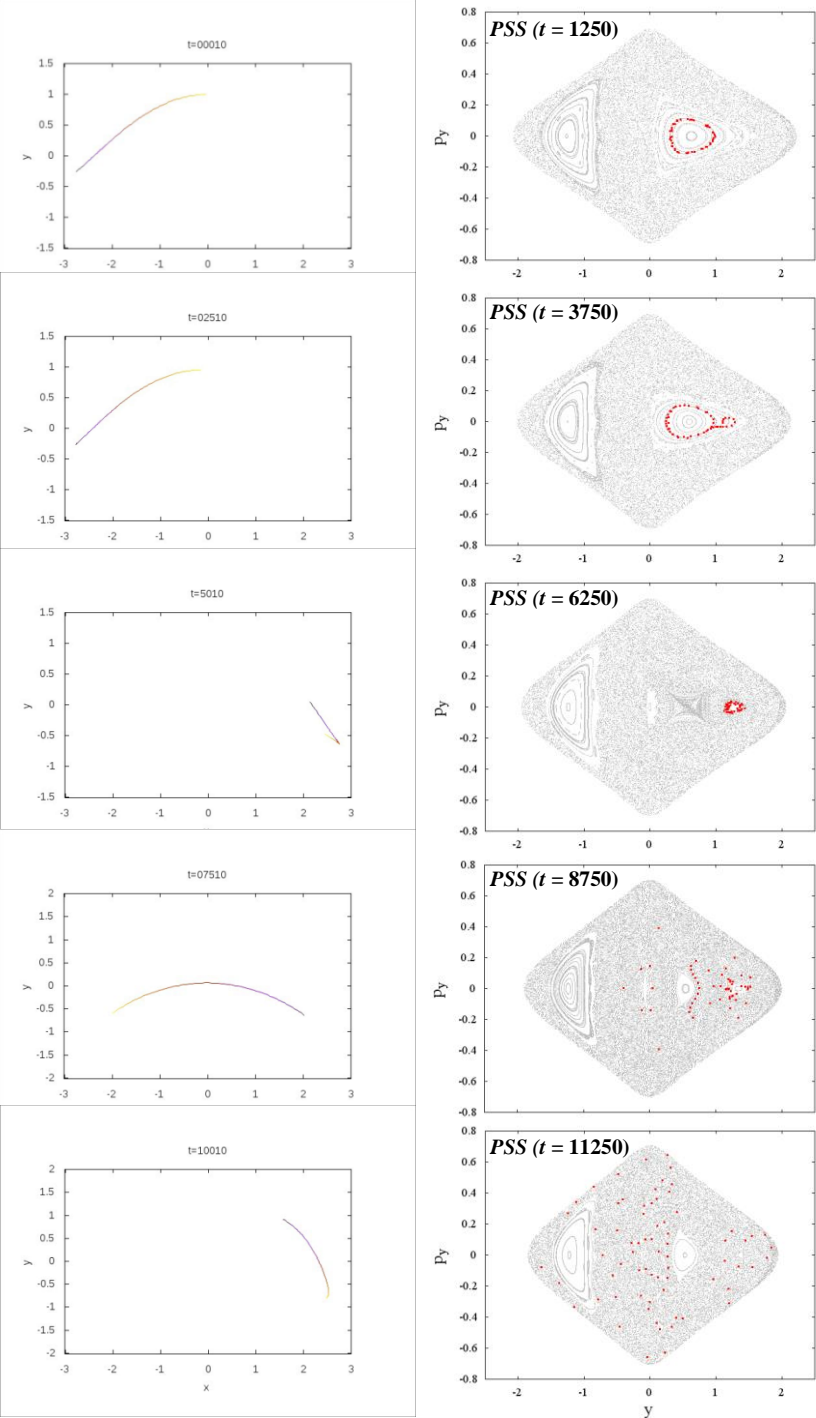
$$\text{where } m^2(u) = \frac{x^2}{a^2 + u} + \frac{y^2}{b^2 + u} + \frac{z^2}{c^2 + u}, \Delta^2(u) = (a^2 + u)(b^2 + u)(c^2 + u),$$

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

Its density is:

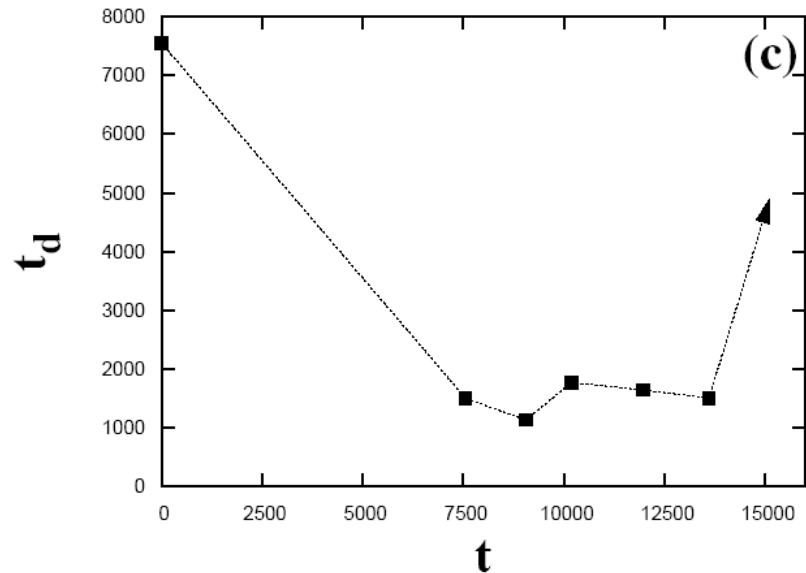
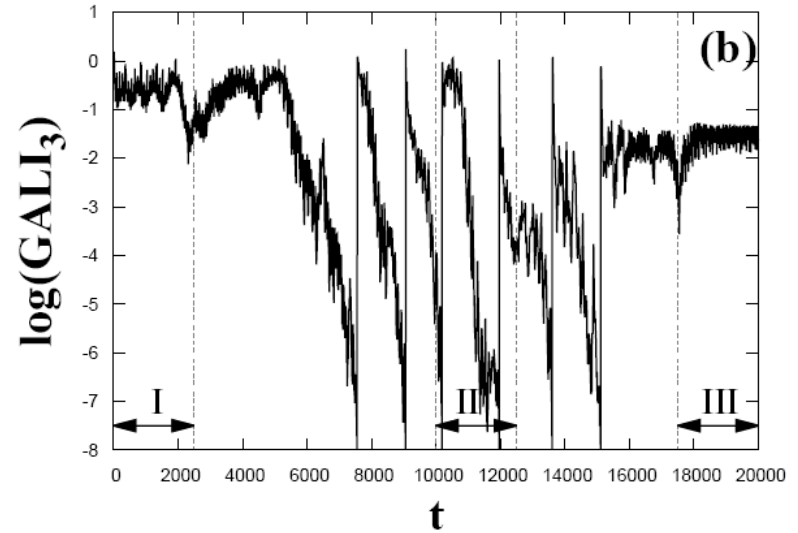
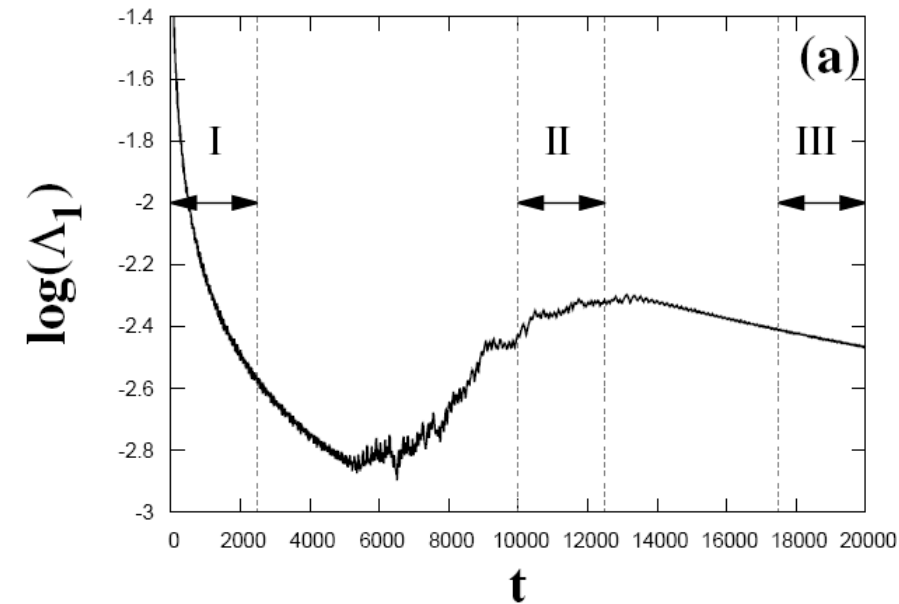
$$\rho = \begin{cases} \rho_c (1 - m^2)^n, & \text{for } m \leq 1 \\ 0, & \text{for } m > 1 \end{cases}, \text{ where } m^2 = \frac{x^2}{a^2} + \frac{y^2}{b^2} + \frac{z^2}{c^2}, a > b > c \text{ and } n = 2.$$

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)

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)
- The Smaller (SALI) and the Generalized (GALI) ALignment Index methods are **fast, efficient and easy to compute chaos indicator**.
- Behaviour of the Generalized ALignment Index of order k ($GALI_k$):
 - ✓ **Chaotic motion: it tends exponentially to zero**
 - ✓ **Regular motion: it fluctuates around non-zero values** (or goes to zero following power-laws)

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)
- The Smaller (SALI) and the Generalized (GALI) ALignment Index methods are **fast, efficient and easy to compute chaos indicator**.
- Behaviour of the Generalized ALignment Index of order k ($GALI_k$):
 - ✓ **Chaotic motion: it tends exponentially to zero**
 - ✓ **Regular motion: it fluctuates around non-zero values** (or goes to zero following power-laws)
- $GALI_k$ 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**

Main References I

- **The color and rotation (CR) method**
 - ✓ Patsis P. A. & Zachilas L (1994) Int. J. Bif. Chaos, 4, 1399
 - ✓ Katsanikas M & Patsis P A (2011) Int. J. Bif. Chaos, 21, 467
 - ✓ Katsanikas M, Patsis P A & Contopoulos G. (2011) Int. J. Bif. Chaos, 21, 2321
- **the 3D phase space slices (3PSS) technique**
 - ✓ Richter M, Lange S, Backer A & Ketzmerick R (2014), Phys. Rev. E 89, 022902
 - ✓ Lange S, Richter M, Onken F, Backer A & Ketzmerick R (2014), Chaos 24, 024409
 - ✓ Onken F, Lange S, Ketzmerick R & Backer A (2016), Chaos 26, 063124
- **Frequency Analysis**
 - ✓ Laskar J (1990) Icarus, 88, 266
 - ✓ Laskar J, Froeschle C & Celletti A (1992) Physica D, 56, 253
 - ✓ Laskar J (1993) Physica D, 67, 257
 - ✓ Bartolini R, Bazzani A, Giovannozzi M, Scandale W & Todesco E (1996) Part. Accel. 52, 147
 - ✓ Laskar J (1999) in Hamiltonian systems with three or more degrees of freedom (ed. Simo C / Plenum Press) p 134
- **Lyapunov exponents**
 - ✓ Oseledec V I (1968) Trans. Moscow Math. Soc., 19, 197
 - ✓ Benettin G, Galgani L, Giorgilli A & Strelcyn J-M (1980) Meccanica, March, 9
 - ✓ Benettin G, Galgani L, Giorgilli A & Strelcyn J-M (1980) Meccanica, March, 21
 - ✓ Wolf A, Swift J B, Swinney H L & Vastano J A (1985) Physica D, 16, 285
 - ✓ Ch.S. (2010) Lect. Notes Phys., 790, 63

Main References II

- **0-1 test**
 - ✓ Gottwald G A & Melbourne I (2004) Proc. R. Soc. A, 460, 603
 - ✓ Gottwald G A & Melbourne I (2005) Physica D, 212, 100
 - ✓ Gottwald G A & Melbourne I (2009) SIAM J. Appl. Dyn., 8, 129
 - ✓ Gottwald G A & Melbourne I (2016) Lect. Notes Phys., 915, 221
- **FLI – OFLI – OFLI2**
 - ✓ Froeschle C, Lega E & Gonczi R (1997) Celest. Mech. Dyn. Astron., 67, 41
 - ✓ Guzzo M, Lega E & Froeschle C (2002) Physica D, 163, 1
 - ✓ Fouchard M, Lega E, Froeschle C & Froeschle C (2002) Celest. Mech. Dyn. Astron., 83, 205
 - ✓ Barrio R (2005) Chaos Sol. Fract., 25, 71
 - ✓ Barrio R (2006) Int. J. Bif. Chaos, 16, 2777
 - ✓ Lega E, Guzzo M & Froeschle C (2016) Lect. Notes Phys., 915, 35
 - ✓ Barrio R (2016) Lect. Notes Phys., 915, 55
- **MEGNO**
 - ✓ Cincotta P M & Simo (2000) Astron. Astroph. Suppl. Ser., 147, 205
 - ✓ Cincotta P M, Giordano C M & Simo C (2003) Physica D, 182, 151
 - ✓ Cincotta P M, & Giordano C M (2016) Lect. Notes Phys., 915, 93
- **RLI**
 - ✓ Sandor Zs, Erdi B & Efthymiopoulos C (2000) Celest. Mech. Dyn. Astron., 78, 113
 - ✓ Sandor Zs, Erdi B, Szell A & Funk B (2004) Celest. Mech. Dyn. Astron., 90 127
 - ✓ Sandor Zs & Maffione N (2016) Lect. Notes Phys., 915, 183

Main References III

- **SALI**

- ✓ Ch.S. (2001) J. Phys. A, 34, 10029
- ✓ Ch.S., Antonopoulos Ch, Bountis T C & Vrahatis M N (2003) Prog. Theor. Phys. Supp., 150, 439
- ✓ Ch.S., Antonopoulos Ch, Bountis T C & Vrahatis M N (2004) J. Phys. A, 37, 6269
- ✓ Bountis T & Ch.S. (2006) Nucl. Inst Meth. Phys Res. A, 561, 173
- ✓ Boreaux J, Carletti T, Ch.S. & Vittot M (2012) Com. Nonlin. Sci. Num. Sim., 17, 1725
- ✓ Boreaux J, Carletti T, Ch.S., Papaphilippou Y & Vittot M (2012) Int. J. Bif. Chaos, 22, 1250219

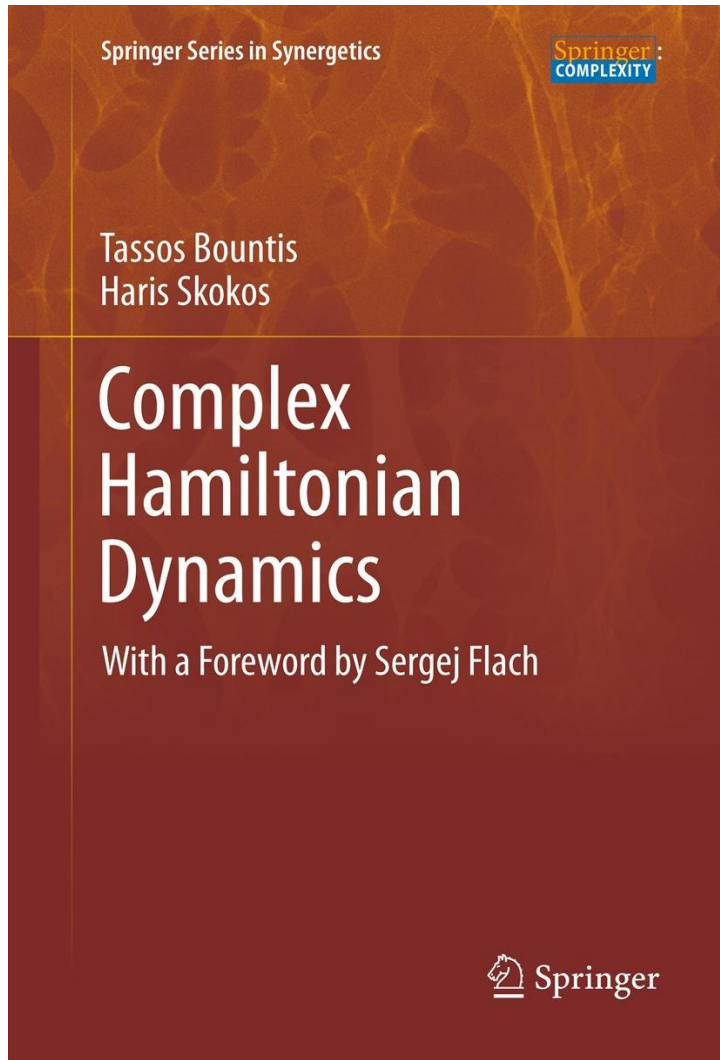
- **GALI**

- ✓ Ch.S., Bountis T C & Antonopoulos Ch (2007) Physica D, 231, 30
- ✓ Ch.S., Bountis T C & Antonopoulos Ch (2008) Eur. Phys. J. Sp. Top., 165, 5
- ✓ Gerlach E, Eggl S & Ch.S. (2012) Int. J. Bif. Chaos, 22, 1250216
- ✓ Manos T, Ch.S. & Antonopoulos Ch (2012) Int. J. Bif. Chaos, 22, 1250218
- ✓ Manos T, Bountis T & Ch.S. (2013) J. Phys. A, 46, 254017

- **Reviews on SALI and GALI**

- ✓ Bountis T C & Ch.S. (2012) ‘Complex Hamiltonian Dynamics’, Chapter 5, Springer Series in Synergetics
- ✓ Ch.S. & Manos T (2016) Lect. Notes Phys., 915, 129

A ...shameless promotion



2012, Springer Series in Synergetics

Contents

1. Introduction
2. Hamiltonian Systems of Few Degrees of Freedom
3. Local and Global Stability of Motion
4. Normal Modes, Symmetries and Stability
5. Efficient Indicators of Ordered and Chaotic Motion
6. FPU Recurrences and the Transition from Weak to Strong Chaos
7. Localization and Diffusion in Nonlinear One-Dimensional Lattices
8. The Statistical Mechanics of Quasi-stationary States
9. Conclusions, Open Problems and Future Outlook

Another ...shameless promotion

Contents

Lecture Notes in Physics 915

Charalampos (Haris) Skokos
Georg A. Gottwald
Jacques Laskar *Editors*

Chaos Detection and Predictability

 Springer

1. **Parlitz:** Estimating Lyapunov Exponents from Time Series
2. **Lega, Guzzo, Froeschlé:** Theory and Applications of the Fast Lyapunov Indicator (FLI) Method
3. **Barrio:** Theory and Applications of the Orthogonal Fast Lyapunov Indicator (OFLI and OFLI2) Methods
4. **Cincotta, Giordano:** Theory and Applications of the Mean Exponential Growth Factor of Nearby Orbits (MEGNO) Method
5. **Ch.S., Manos:** The Smaller (SALI) and the Generalized (GALI) Alignment Indices: Efficient Methods of Chaos Detection
6. **Sándor, Maffione:** The Relative Lyapunov Indicators: Theory and Application to Dynamical Astronomy
7. **Gottwald, Melbourne:** The 0-1 Test for Chaos: A Review
8. **Siebert, Kantz:** Prediction of Complex Dynamics: Who Cares About Chaos?