



The global modeling technique

Background & Applications

Sylvain MANGIAROTTI
Researcher at 



Theory of nonlinear dynamical systems

- Henri Poincaré



probably the first one to understand
how a system can be both

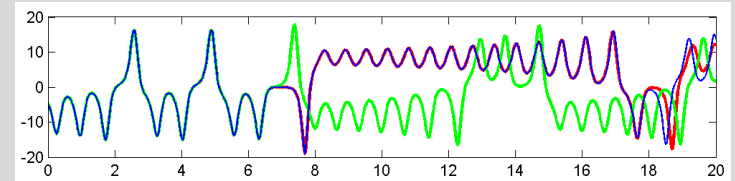
Deterministic & unpredictable at long term



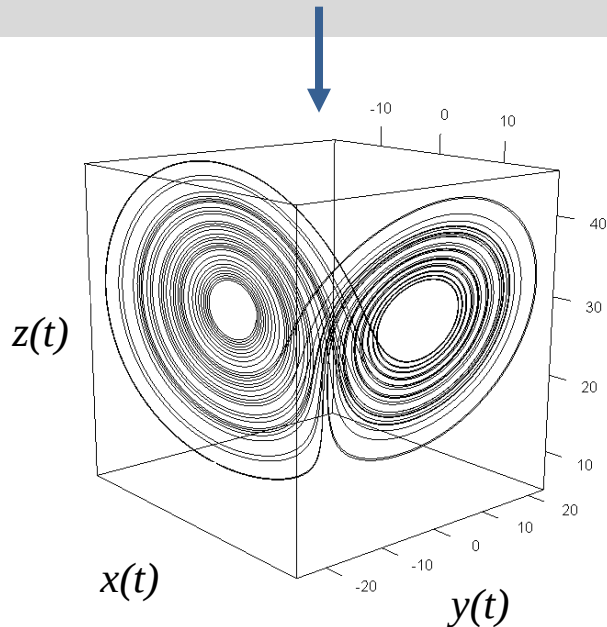
Lorenz system (1963)

$$\begin{cases} \dot{x} = \sigma(y - x) \\ \dot{y} = -xz + \rho x - y \\ \dot{z} = xy - \beta z \end{cases}$$

$x(t)$



$x(t), y(t), z(t)$



The *Phase Space* (or *State space*) :
a space that provides the complete solutions

Theory of nonlinear dynamical systems

- Henri Poincaré



- E.N. Lorenz-1963, 1st chaotic system

Simple systems, unpredictable at long term

Followed by Rössler-76, Rössler-79 .. Malasoma-2000



E. Lorenz

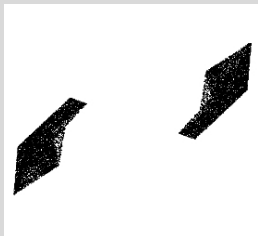


O. Rössler



J-M Malasoma

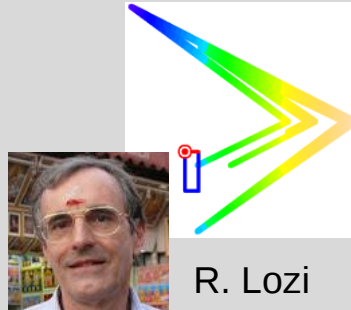
Discrete systems: Mira-69, Hénon-76, Lozi-78, Rössler-79



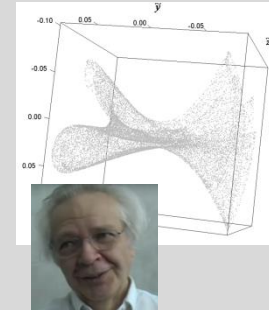
C. Mira



M. Hénon



R. Lozi



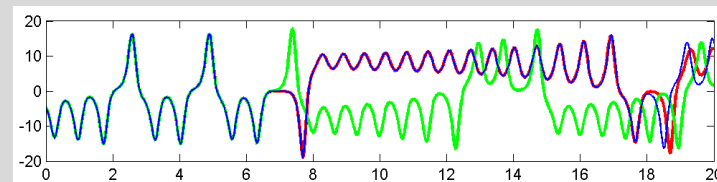


F. Takens

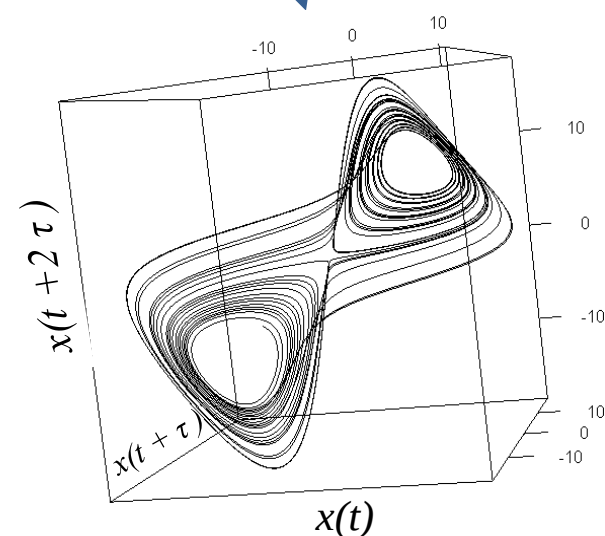
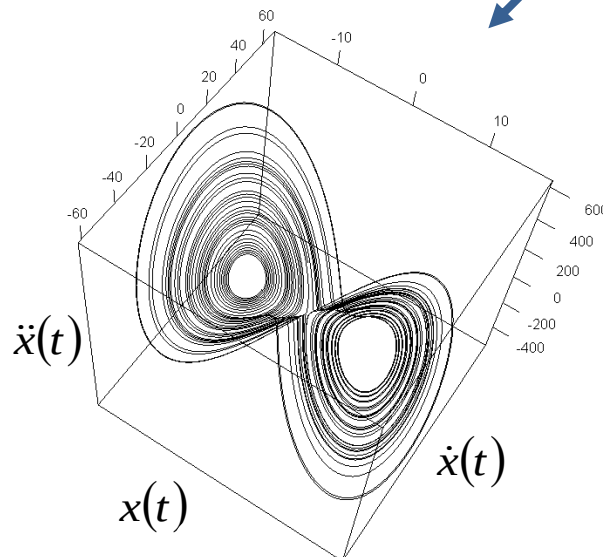
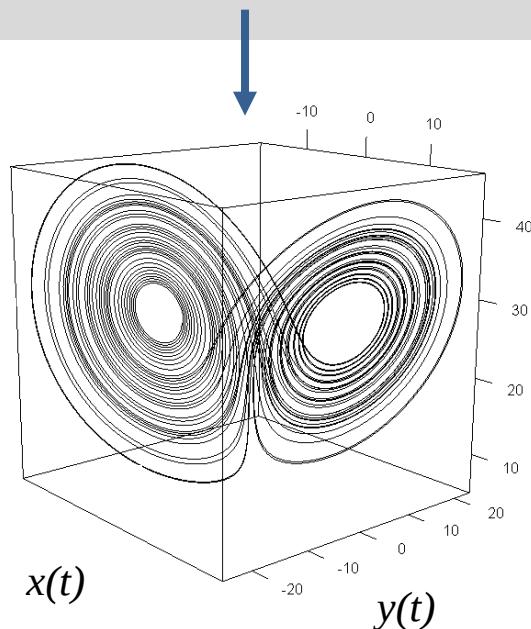
Takens Theorem (1981)

$$\begin{cases} \dot{x} = \sigma(y - x) \\ \dot{y} = -xz + \rho x - y \\ \dot{z} = xy - \beta z \end{cases}$$

$x(t)$



$x(t), y(t), z(t)$



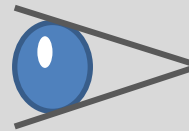
Nonlinear invariants are conserved

Global modeling

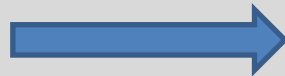
- multivariate

$$\begin{cases} \dot{x}_1 = f_1(x_1, x_2, \dots, x_n) \\ \dot{x}_2 = f_2(x_1, x_2, \dots, x_n) \\ \dots \\ \dot{x}_n = f_n(x_1, x_2, \dots, x_n) \end{cases}$$

x_i is observed



?



Equations from one
single observed variable?

J. Crutchfield



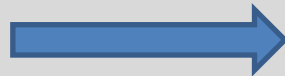
Crutchfield & McNamara (1987)

Global modeling

- multivariate

$$\begin{cases} \dot{x}_1 = f_1(x_1, x_2, \dots, x_n) \\ \dot{x}_2 = f_2(x_1, x_2, \dots, x_n) \\ \dots \\ \dot{x}_n = f_n(x_1, x_2, \dots, x_n) \end{cases}$$

x_i is observed



- univariate

$$\begin{cases} \dot{x}_i = X_2 \\ \dot{X}_2 = X_3 \\ \dots \\ \dot{X}_n = F(x_i, X_2, \dots, X_n) \end{cases}$$



G. Gouesbet

Gouesbet (1991)

Global modeling

- multivariate

$$\begin{cases} \dot{x}_1 = f_1(x_1, x_2, \dots, x_n) \\ \dot{x}_2 = f_2(x_1, x_2, \dots, x_n) \\ \dots \\ \dot{x}_n = f_n(x_1, x_2, \dots, x_n) \end{cases}$$

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- univariate

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G. Gouesbet

C. Letellier



Gouesbet & Letellier (1994)

Polynomial
approximation



$$F(x_i, X_2, \dots, X_n) = P(x_i, X_2, \dots, X_n)$$

Good results for some variables of some systems / but not all ...

Global modeling

C. Letellier



L. Aguirre



« Problem »
of
observability

- multivariate

$$\begin{cases} \dot{x}_1 = f_1(x_1, x_2, \dots, x_n) \\ \dot{x}_2 = f_2(x_1, x_2, \dots, x_n) \\ \dots \\ \dot{x}_n = f_n(x_1, x_2, \dots, x_n) \end{cases}$$

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Letellier & Aguirre (2001)

Global modeling

C. Letellier



L. Aguirre



- multivariate

$$\begin{cases} \dot{x}_1 = f_1(x_1, x_2, \dots, x_n) \\ \dot{x}_2 = f_2(x_1, x_2, \dots, x_n) \\ \dots \\ \dot{x}_n = f_n(x_1, x_2, \dots, x_n) \end{cases}$$

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- univariate

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Letellier & Aguirre (2001)

Lie derivatives

$$\mathcal{L}_f f_i(x) = \frac{\partial f_i(x)}{\partial x} f(x) = \sum_{k=1}^m \frac{\partial f_i(x)}{\partial x} f_k$$

Sophus Lie



Global modeling technique

- Empirical approach

Few *a priori* knowledge required

Directly applies to time series

- Well adapted to low-dimensional systems

Can bring strong argument for determinism

All the conditions and properties of chaos in a consistent fashion

- Strong and rich theoretical background

Global solutions

Theory of nonlinear dynamical systems

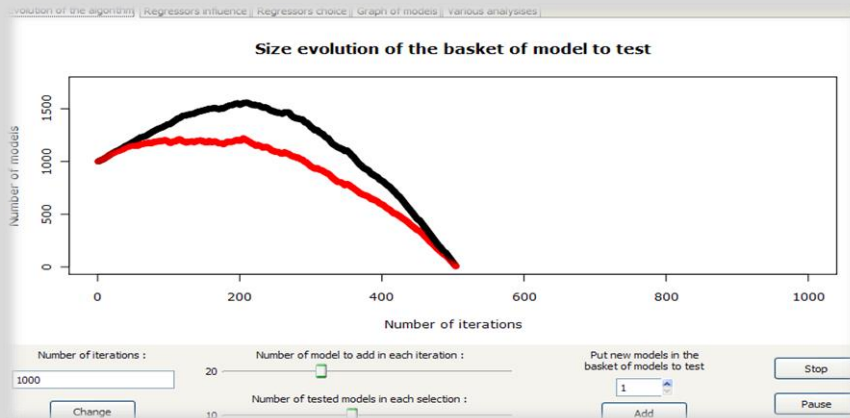
(Poincaré's work, Takens' Theorem, topology of chaos, etc.)



H. Poincaré

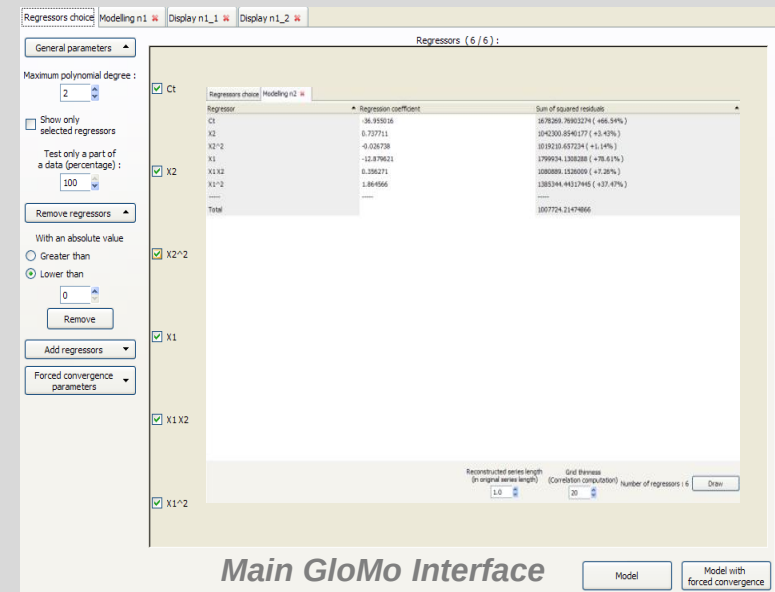
Algorithmic tools *R*-packages

PoMoS



Models selection algorithm interaction

GloMo



L. Drapeau
Ing. IRD



M. Huc
Ing. CNRS



R. Coudret



F. Le Jean



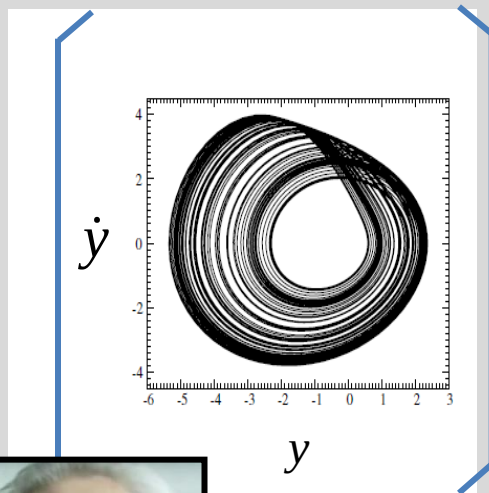
M. Chassan

Mangiarotti et al.
Phys. Rev. E (2012)

GPoM platform, the potential of the approach

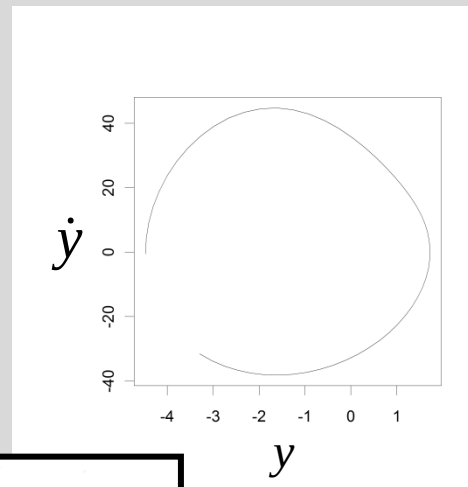
Rössler-y

Original system



O. Rössler

Unstable periodic orbits (e.g. period 1)

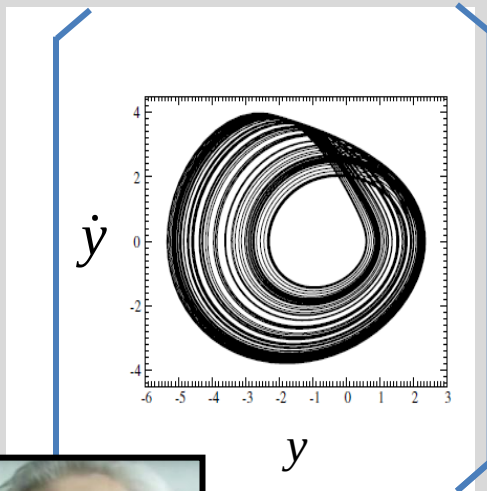


Letellier, Mangiarotti & Aguirre (under revision)
Letellier et al., Entropie 1997

GPoM platform, the potential of the approach

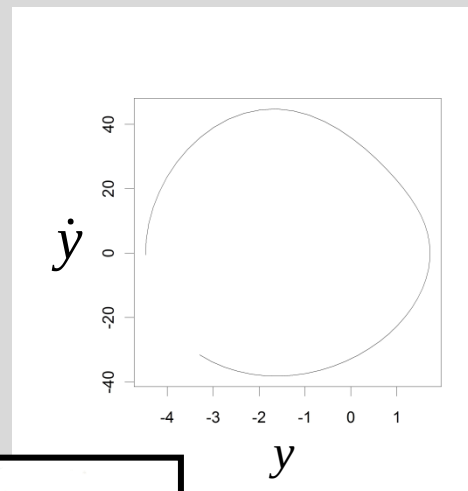
Rössler-y

Original system

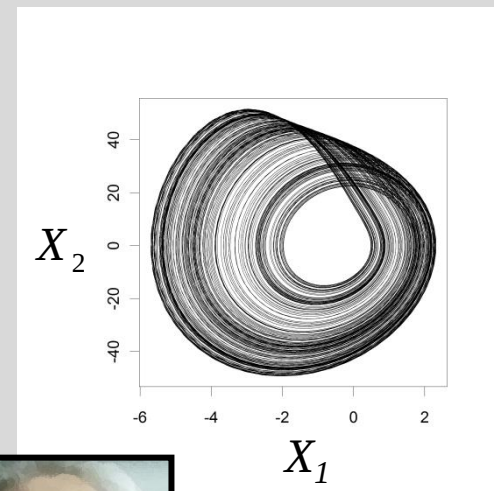


O. Rössler

Unstable periodic orbits (e.g. period 1)



Global model



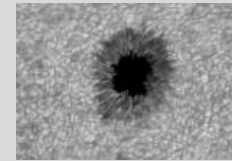
Letellier, Mangiarotti & Aguirre (under revision)
Letellier et al., Entropie 1997

Global modeling applied to real observations univariate

- sun

• sunspot cycles

NARMA

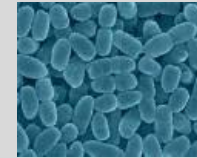


Aguirre et al. 2008

- epidemiology

• whooping cough

NARMA



Boudjema et Cazelles 2001

- ecological

• Canadian lynx

ODE



Maquet et al. 2007

- agricultural

• cereal crops

ODE



Mangiarotti et al.
2011 RNL; 2014 *Chaos*

- climatic

• snow cycles

ODE



Mangiarotti 2014 *HDR*

.. systems

Normalized Difference Vegetation Index

$$NDVI = \frac{\rho_{PIR} - \rho_R}{\rho_{PIR} + \rho_R}$$

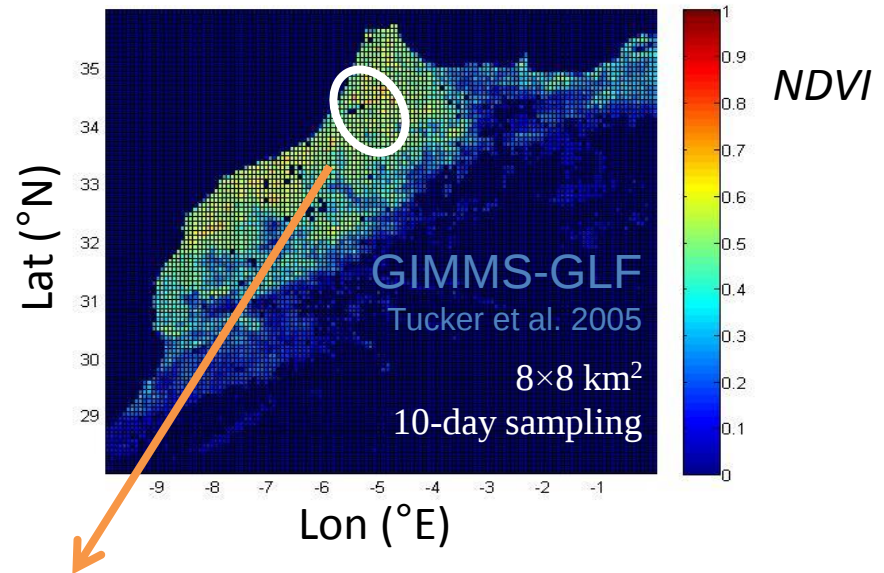


0.1

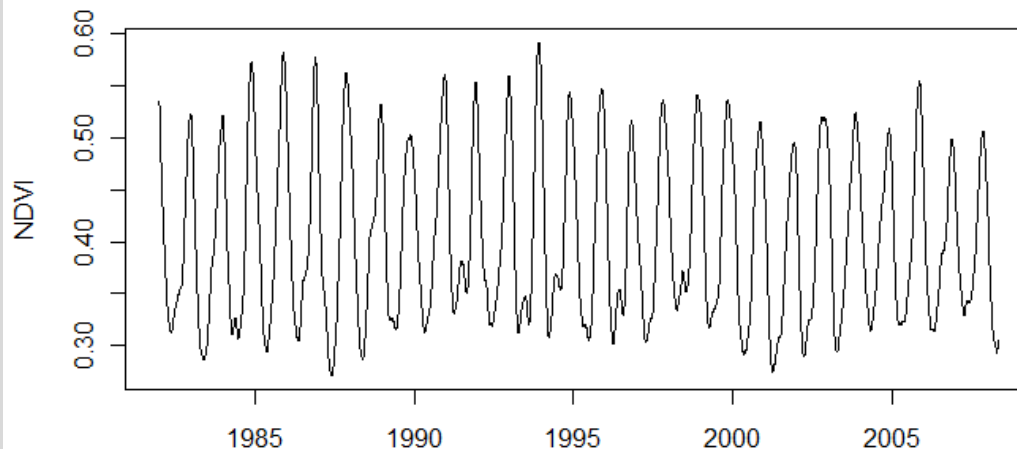
0.7



NDVI map (Morocco, February 2000)

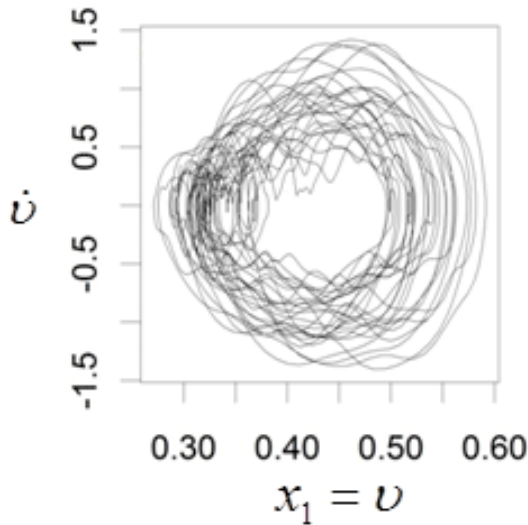


Time series (North Morocco)

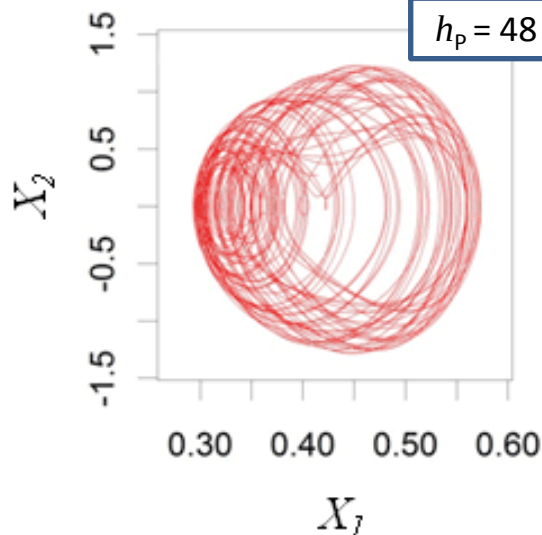


A global model for *cereal crops*

data

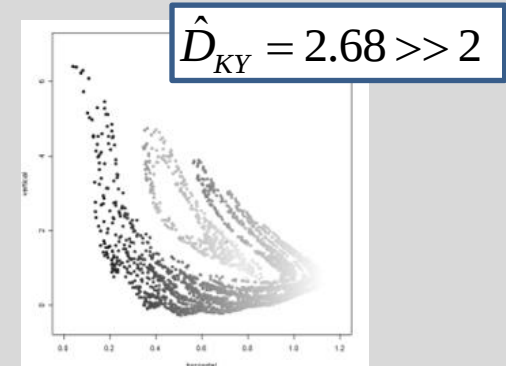


model

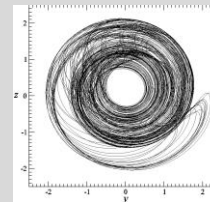


- Complex structure
- Very few previous cases
- Never directly obtained from data

*Poincaré
section*

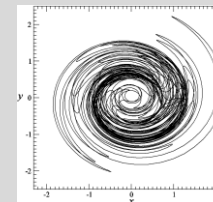


$$\hat{D}_{KY} = 2.39$$



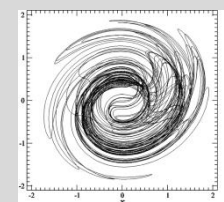
Lorenz-84

$$\hat{D}_{KY} = 2.76$$



Wieczorek 1999

$$\hat{D}_{KY} = 2.54$$



Chlouverakis 2002

Snow surface cycles

Snow surface estimated from satellite (NDSI)

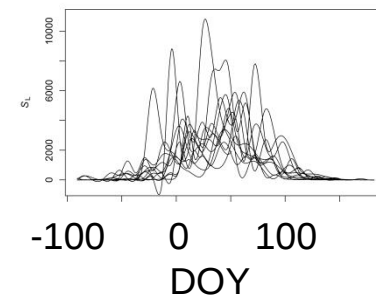
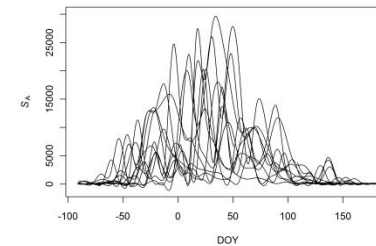
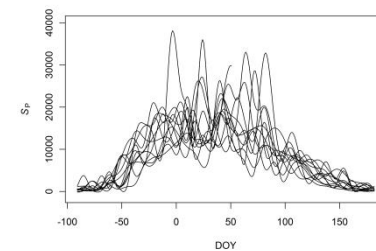
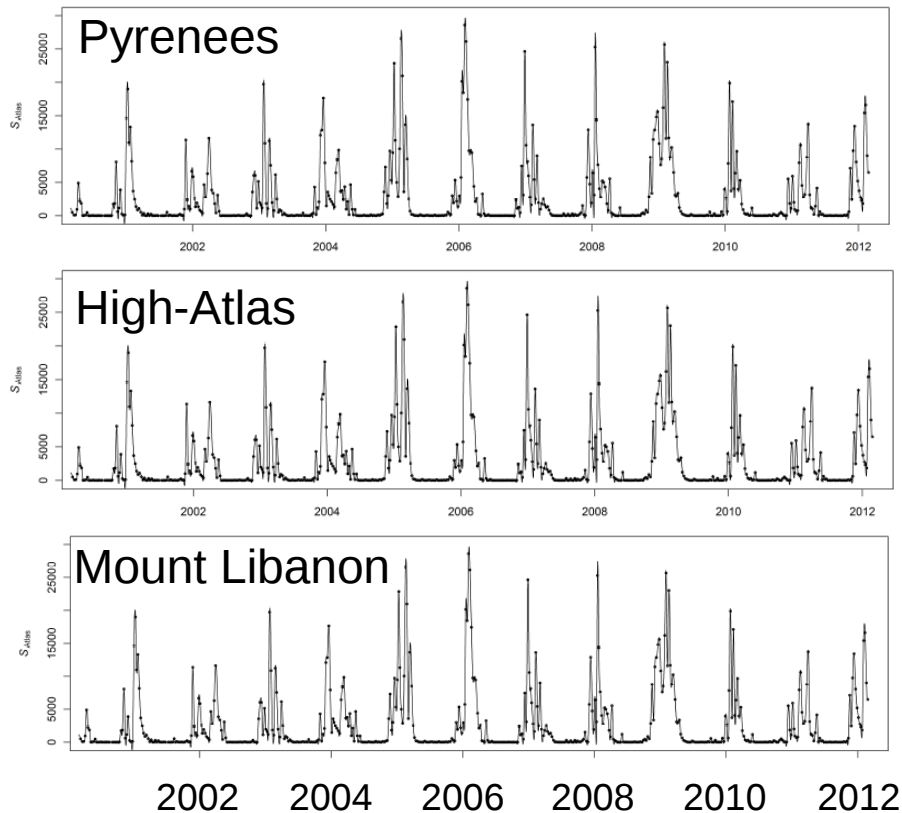
- daily MODIS product used: MOD10A1
- missing alues are interpolated (in time and space)

(Gascoin et al. *HESS* 2015)

S. Gascoin



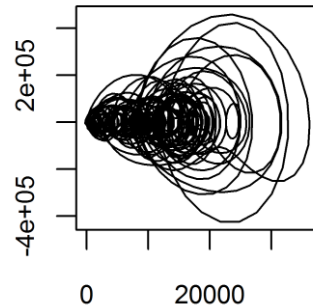
L. Drapeau



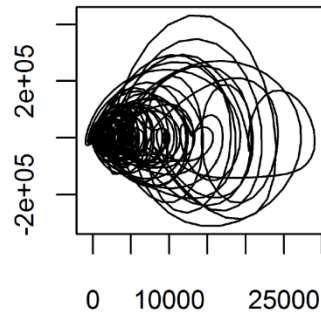
Snow surface cycles

Phase portraits

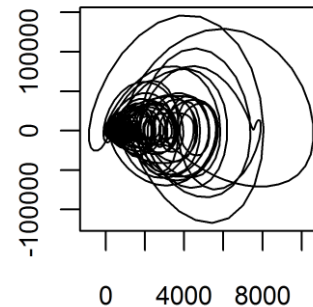
Pyrenees



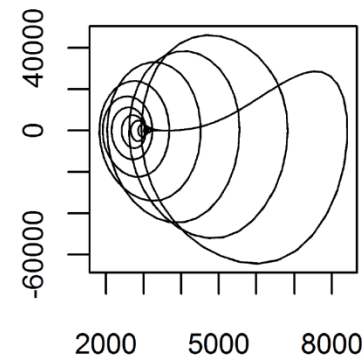
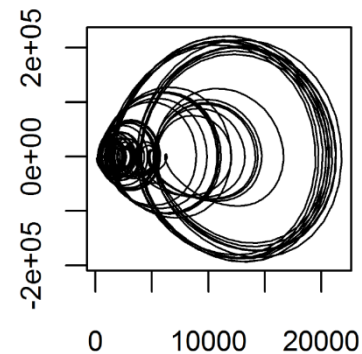
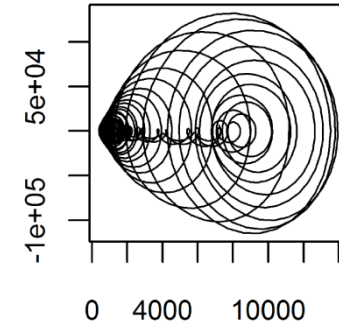
High-Atlas



Mount Libanon



Global Models

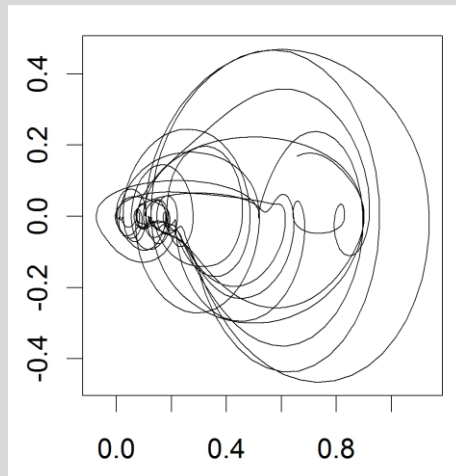


Autres analyses univariées

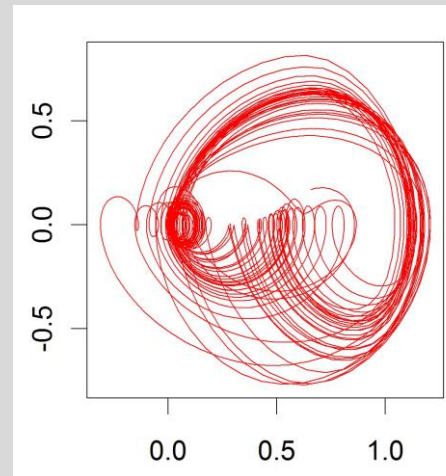
- Soil Evaporative Efficiency
(thèse Vivien Stephan, Dir. O. Merlin)



Original data
(from TEC model)



4D Global Model



Autres analyses univariées

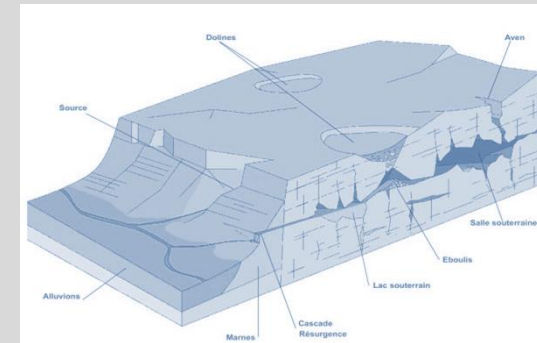
le Doubs



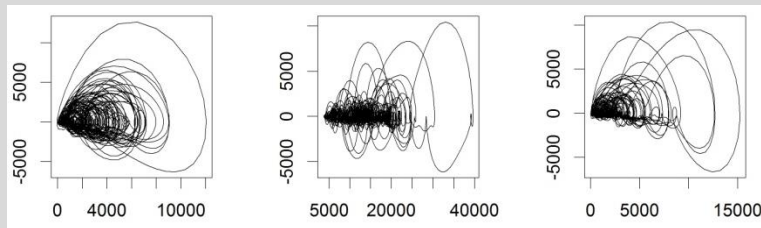
la Touvre



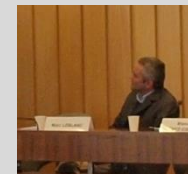
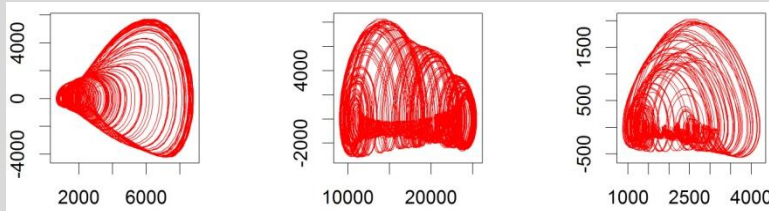
le Lez



Original data



Global models



M. Leblanc



Y. Zhang

- The discharge of karstic springs (in France)

Global modeling applied to real observations ^{red}multivariate

- biomedical

- eco-epidemiology

- human breath

- bubonic plague (1896-1911)

- Ebola Virus Disease (2014-2016)

NARMA

ODE

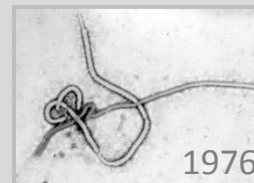
ODE



Letellier et al. 2013



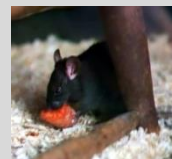
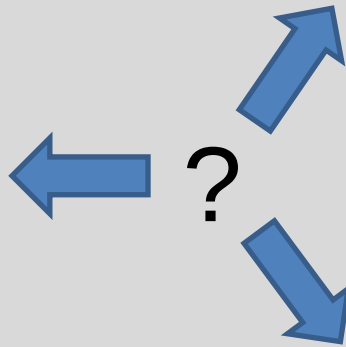
Mangiarotti 2015



Huc & Mangiarotti 2016

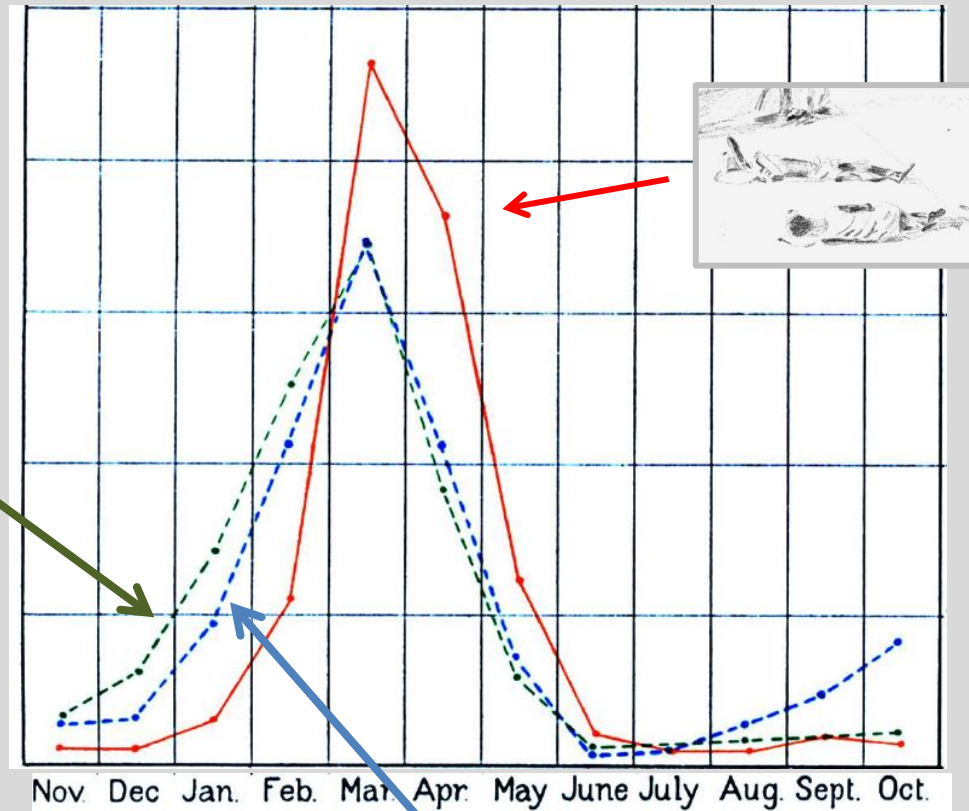
Projet MoMu
2015-2016
(AO LEFE-Manu)

Multivariate global modeling



$$\begin{cases} \dot{x}_1 = P_1(x_1, x_2, x_3) \\ \dot{x}_2 = P_2(x_1, x_2, x_3) \\ \dot{x}_3 = P_3(x_1, x_2, x_3) \end{cases}$$

Prevalence of plague (in percent) 1906-07

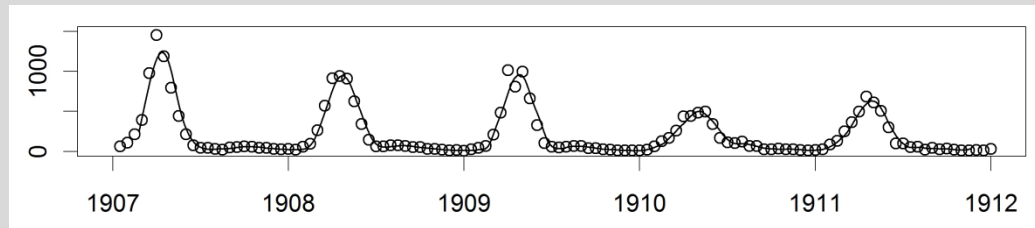


- Human plague
- - - Plague in *M. decumanus*
- - - Plague in *M. rattus*

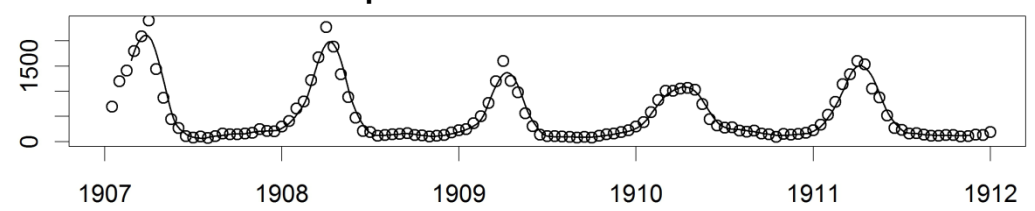


Reproduced from
« Report on Plague investigations in India »
J. of Hygiene 8(2) 1908

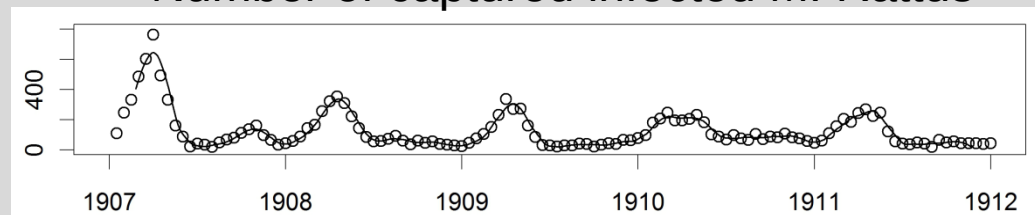
Plague death per half-month



Number of captured infected *M. Decumanus*

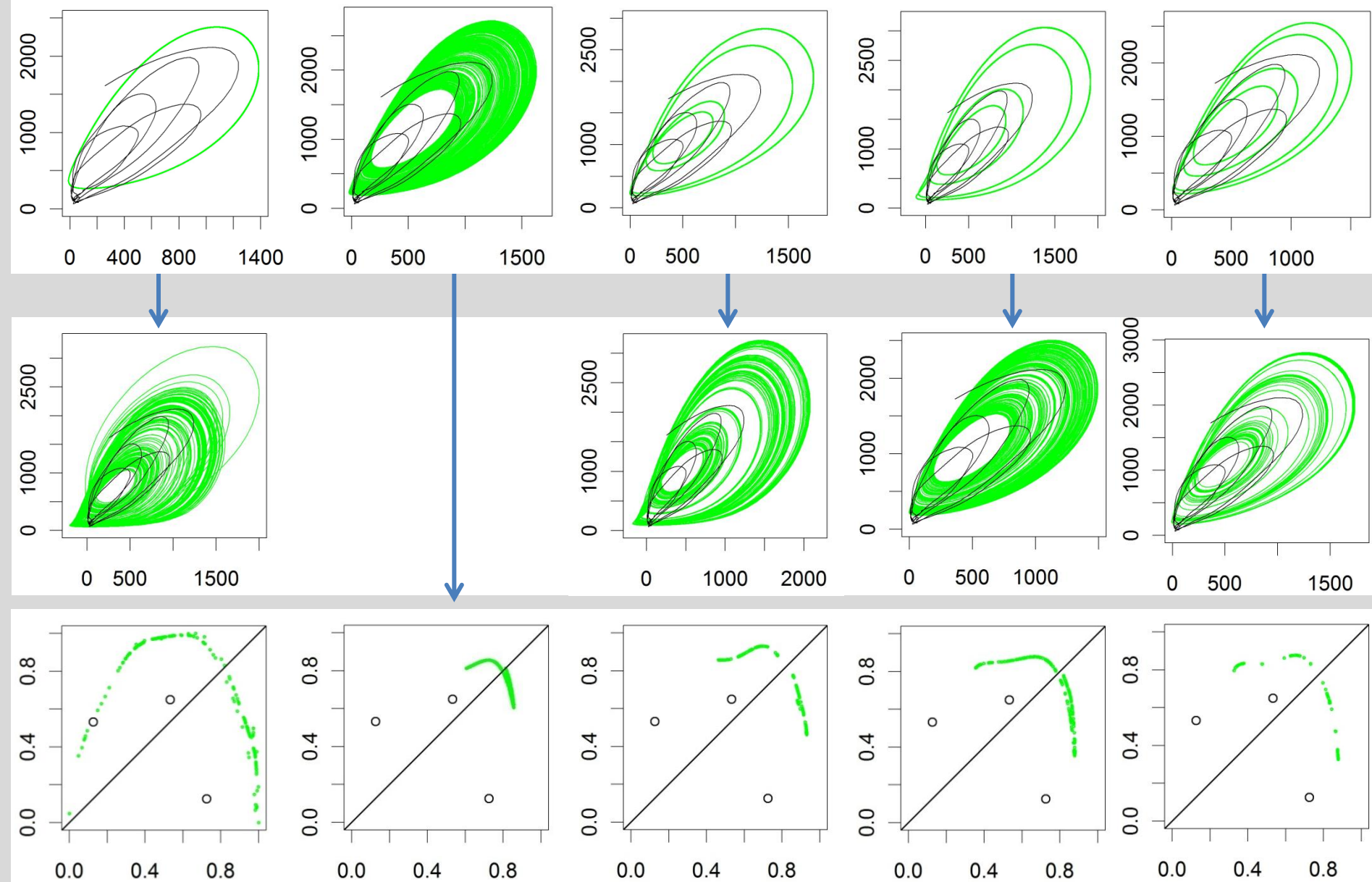


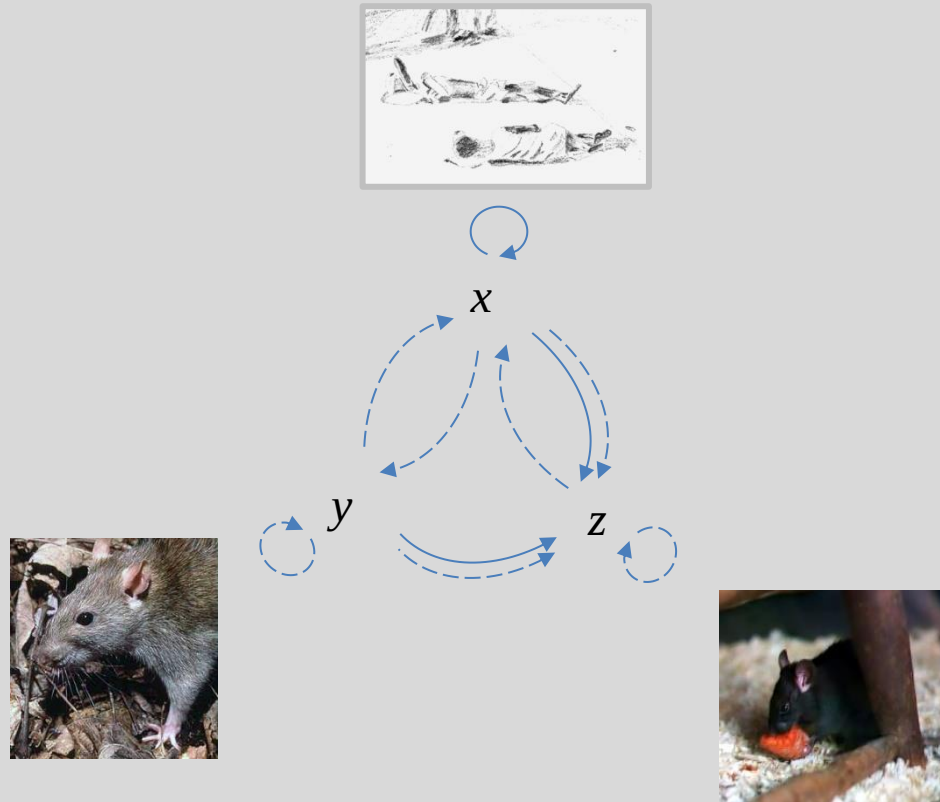
Number of captured infected *M. Rattus*



Original data from The Advisory Committee
“Reports on plague investigation in India”, J. Hyg. 12 (1912).

Multivariate analysis

 M_1 M_2 M_3 M_4 M_5 



$$\begin{cases} \dot{x} = a_1 z(y - \alpha z) - a_3 x \\ \dot{y} = b_1 y^2 - b_2 xy \\ \dot{z} = c_1 z^2 + c_2 y - c_3 xz + c_5 x(y - \beta) \end{cases}$$

- Multivariate ODE directly obtained from observed time series
- All the terms could be interpreted
- A good tool for coupling detection

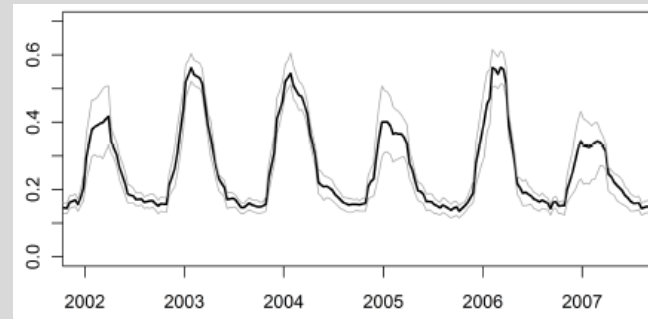
Aggregated signal



*Phase
synchronisation*



AVHRR data (8km, aggr. 100km)



Safi province (Morocco)

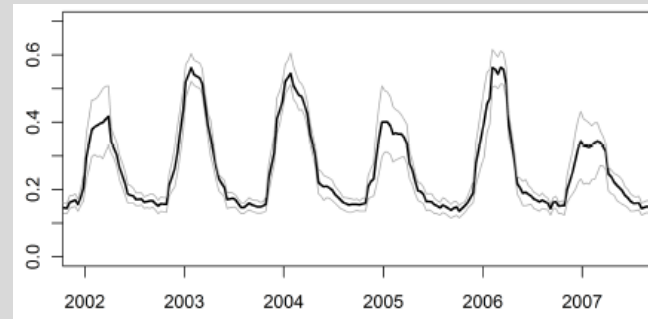
Aggregated signal



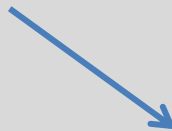
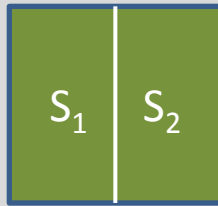
*Phase
synchronisation*



AVHRR data (8km, aggr. 100km)



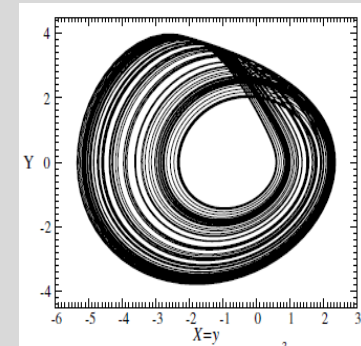
Safi province (Morocco)



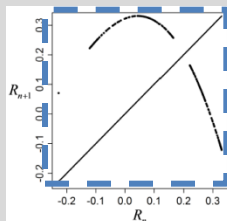
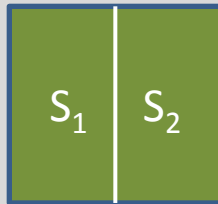
What can be the effect of aggregation?

Analytical point of view

$$\begin{cases} \frac{dx}{dt} = -y - z \\ \frac{dy}{dt} = x + ay \\ \frac{dz}{dt} = b + z(x - c) \end{cases}$$

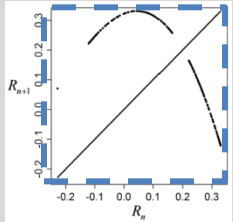


Rössler-y

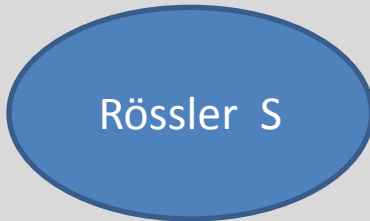


O. Rössler

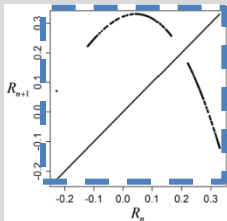
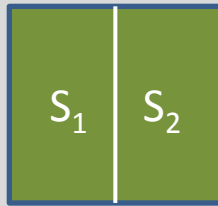
Analytical point of vue



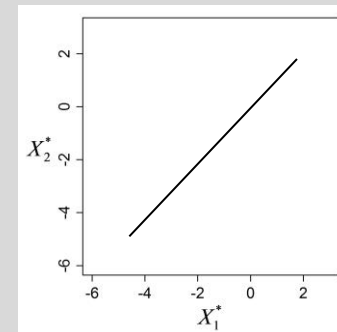
synchro.-y
full



Rössler-y

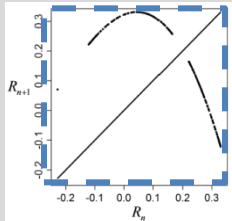


Cross portrait



$$y_k = y$$

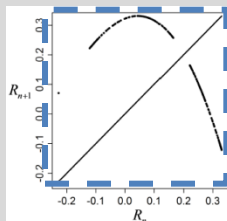
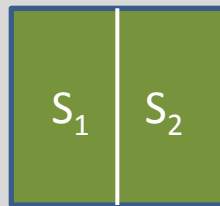
Analytical point of vue



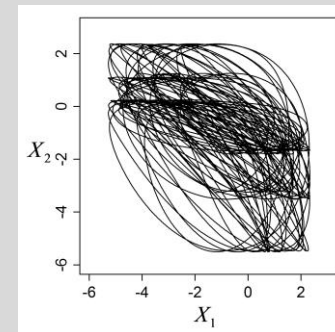
synchro.-y
no



Rössler-y

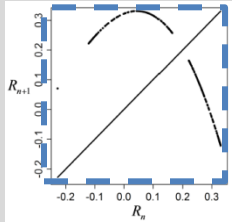


Cross portrait



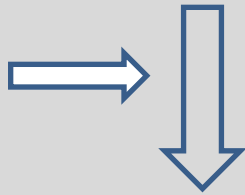
$$y_k = \beta_k(t) y$$

Analytical point of vue

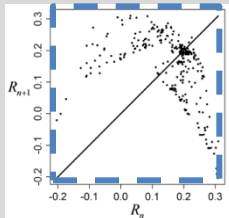
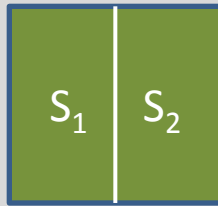


Rössler S

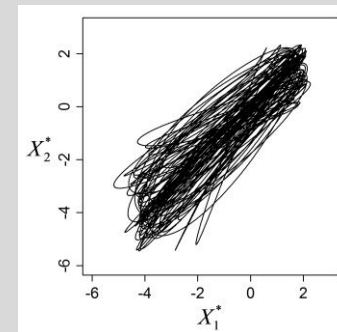
synchro.-y
in phase



Rössler-y

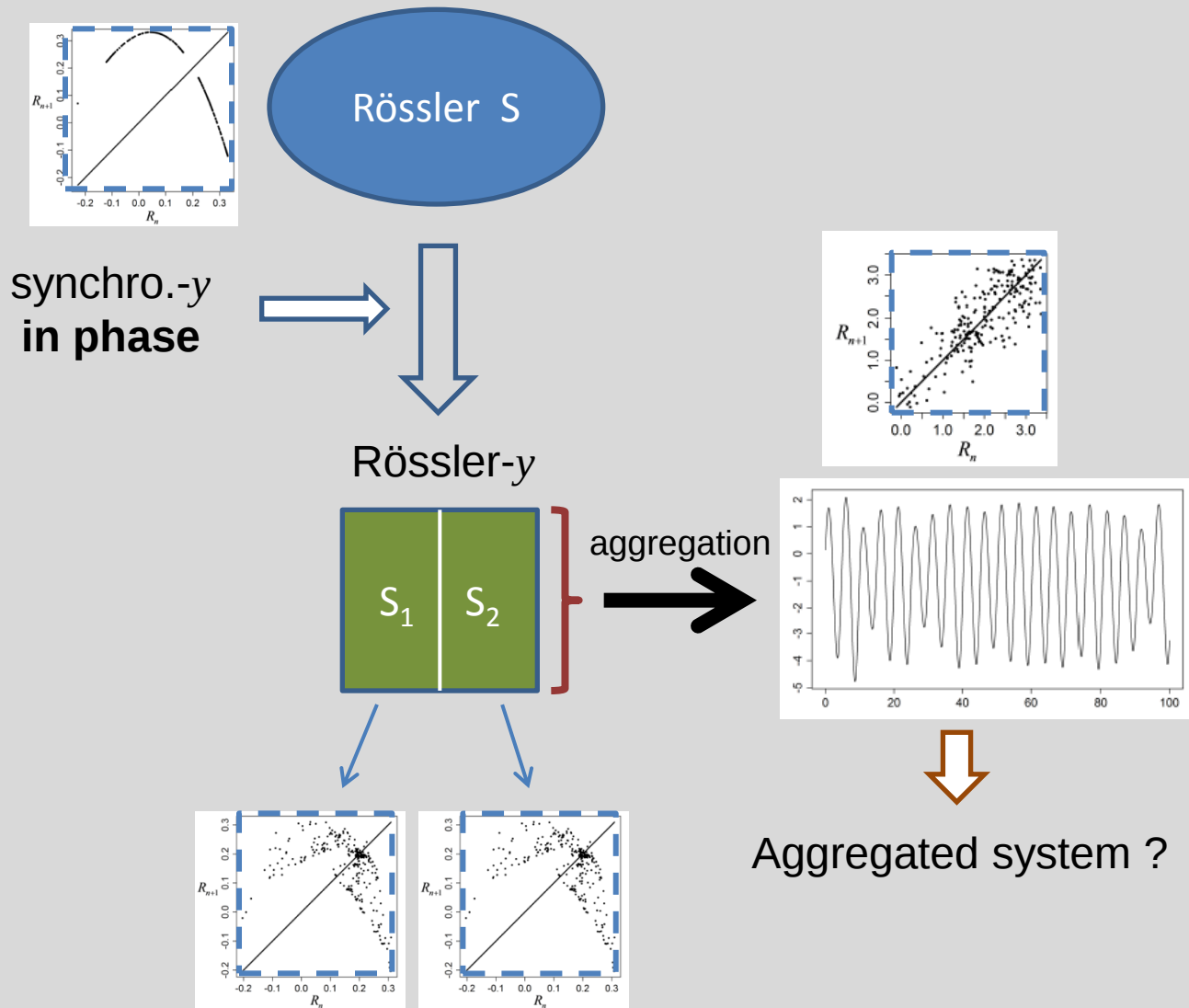


Cross portrait

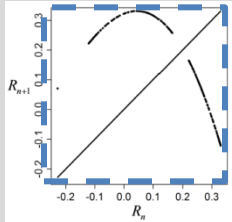


$$y_k \approx \beta_k y \quad \text{locally}$$

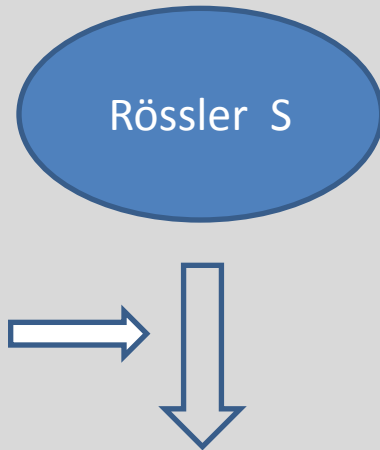
Analytical point of vue



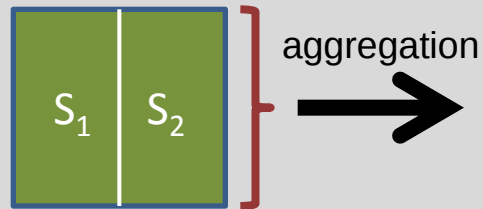
Analytical point of vue



synchro.-y
in phase

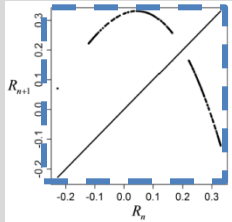


Rössler-y



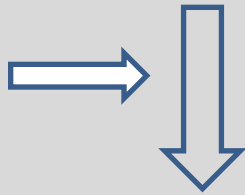
$$S_k \begin{cases} \dot{x}_k = -y_k - z_k \\ \dot{y}_k = x_k + a_k y_k + \alpha_k (y_k - y) \\ \dot{z}_k = b + z_k (x_k - c) \end{cases}$$

Analytical point of vue

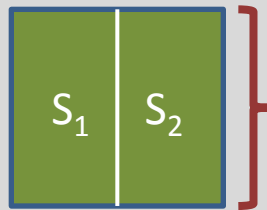


Rössler S

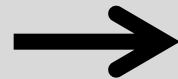
synchro.-y
in phase



Rössler-y



aggregation



?

$$\approx \begin{cases} \dot{x} = -y - z \\ \dot{y} = x + ay \\ \dot{z} = 2b + z[(1 - \gamma)x - c] \end{cases}$$

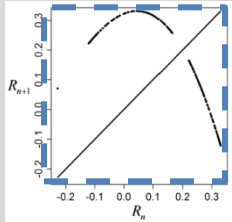


$$S_k \begin{cases} \dot{x}_k = -y_k - z_k \\ \dot{y}_k = x_k + a_k y_k + \alpha_k (y_k - y) \\ \dot{z}_k = b + z_k (x_k - c) \end{cases}$$

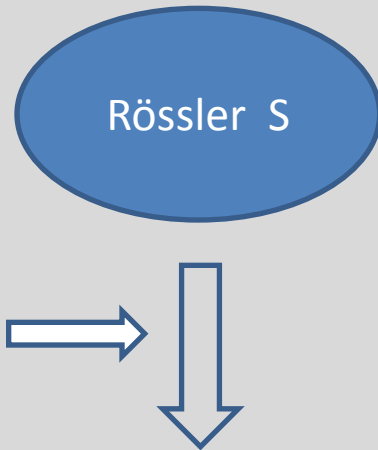
$\Rightarrow$ décalage dans le diagramme de bifurcation

$\Rightarrow$ seuls les invariants topologiques peuvent être conservés

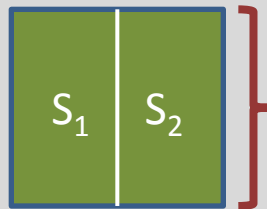
Analytical point of vue



synchro.-y
in phase



Rössler-y



aggregation



?

$$\approx \begin{cases} \dot{x} = -y - z \\ \dot{y} = x + ay \\ \dot{z} = 2b + z[(1 - \gamma)x - c] \end{cases}$$



$$S_k \begin{cases} \dot{x}_k = -y_k - z_k \\ \dot{y}_k = x_k + a_k y_k + \alpha_k (y_k - y) \\ \dot{z}_k = b + z_k (x_k - c) \end{cases}$$

$\Rightarrow$ décalage dans le diagramme de bifurcation

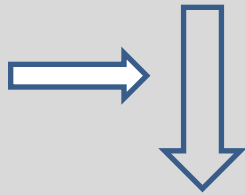
$\Rightarrow$ seuls les invariants topologiques peuvent être conservés

Can be generalized?

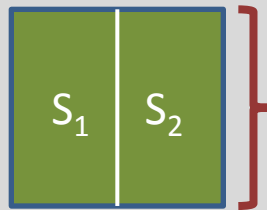
Analytical point of vue

Canonical
form

synchro.
in phase



Can be generalized?



aggregation

S_k

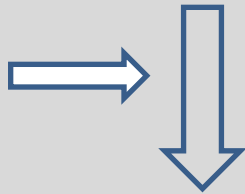
$$\begin{cases} \dot{X}_1 = X_2 \\ \dot{X}_2 = X_3 \\ \dot{X}_n = F(X_1, X_2, \dots, X_n) \end{cases}$$

$$F(X, Y, Z) = \alpha_{000} + \sum_{i+j+k=1} \alpha_{ijk} X^i Y^j Z^k + \sum_{2 \leq i+j+k \leq q} \alpha_{ijk} X^i Y^j Z^k$$

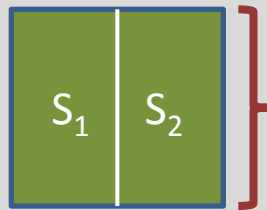
Analytical point of vue

Canonical
form

synchro.
in phase



Can be generalized?



aggregation

S_k

$$\begin{cases} \dot{X}_1 = X_2 \\ \dot{X}_2 = X_3 \\ \dot{X}_n = F(X_1, X_2, \dots, X_n) \end{cases}$$

$$\begin{cases} \dot{X} = Y \\ \dot{Y} = Z \\ \dot{Z} = 2\alpha_{000} + \sum_{i+j+k=1} \alpha_{ijk} X^i Y^j Z^k \\ \quad + (1 - \nu_2) \sum_{i+j+k=2} \alpha_{ijk} X^i Y^j Z^k + \dots \\ \quad + (1 - \nu_n) \sum_{i+j+k=n} \alpha_{ijk} X^i Y^j Z^k + \dots \\ \quad + (1 - \nu_q) \sum_{i+j+k=q} \alpha_{ijk} X^i Y^j Z^k. \end{cases}$$

with

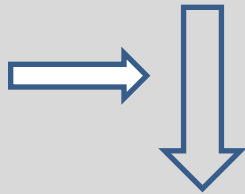
$$\nu_n = \sum_{i=2}^{n-1} r_i \beta^i / (1 + \beta)^n$$

$$F(X, Y, Z) = \alpha_{000} + \sum_{i+j+k=1} \alpha_{ijk} X^i Y^j Z^k + \sum_{2 \leq i+j+k \leq q} \alpha_{ijk} X^i Y^j Z^k$$

Analytical point of vue

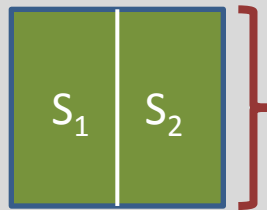
Canonical
form

synchro.
in phase



Can be generalized to

- N attractors
- n dimensions
- any polynomial form (of any degree q)



aggregation

S_k

$$\begin{cases} \dot{X}_1 = X_2 \\ \dot{X}_2 = X_3 \\ \dot{X}_n = F(X_1, X_2, \dots, X_n) \end{cases}$$

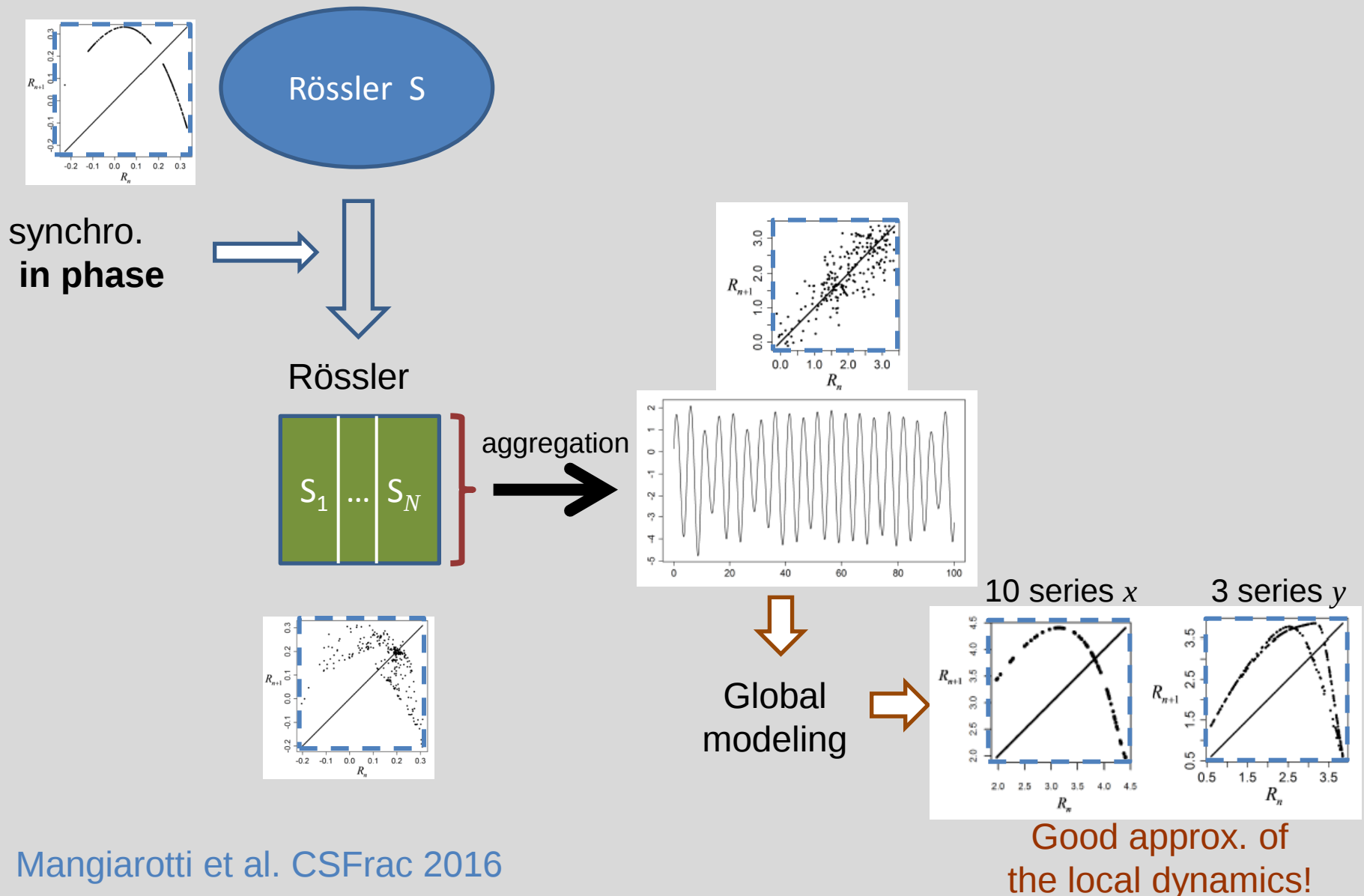
$$\begin{cases} \dot{X} = Y \\ \dot{Y} = Z \\ \dot{Z} = 2\alpha_{000} + \sum_{i+j+k=1} \alpha_{ijk} X^i Y^j Z^k \\ \quad + (1 - \nu_2) \sum_{i+j+k=2} \alpha_{ijk} X^i Y^j Z^k + \dots \\ \quad + (1 - \nu_n) \sum_{i+j+k=n} \alpha_{ijk} X^i Y^j Z^k + \dots \\ \quad + (1 - \nu_q) \sum_{i+j+k=q} \alpha_{ijk} X^i Y^j Z^k. \end{cases}$$

with

$$\nu_n = \sum_{i=2}^{n-1} r_i \beta^i / (1 + \beta)^n$$

$$F(X, Y, Z) = \alpha_{000} + \sum_{i+j+k=1} \alpha_{ijk} X^i Y^j Z^k + \sum_{2 \leq i+j+k \leq q} \alpha_{ijk} X^i Y^j Z^k$$

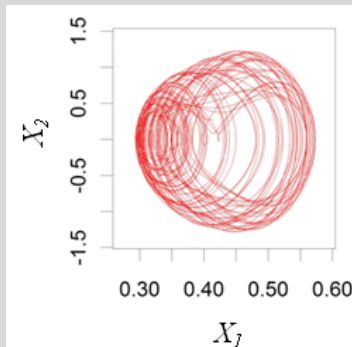
Practical point of vue



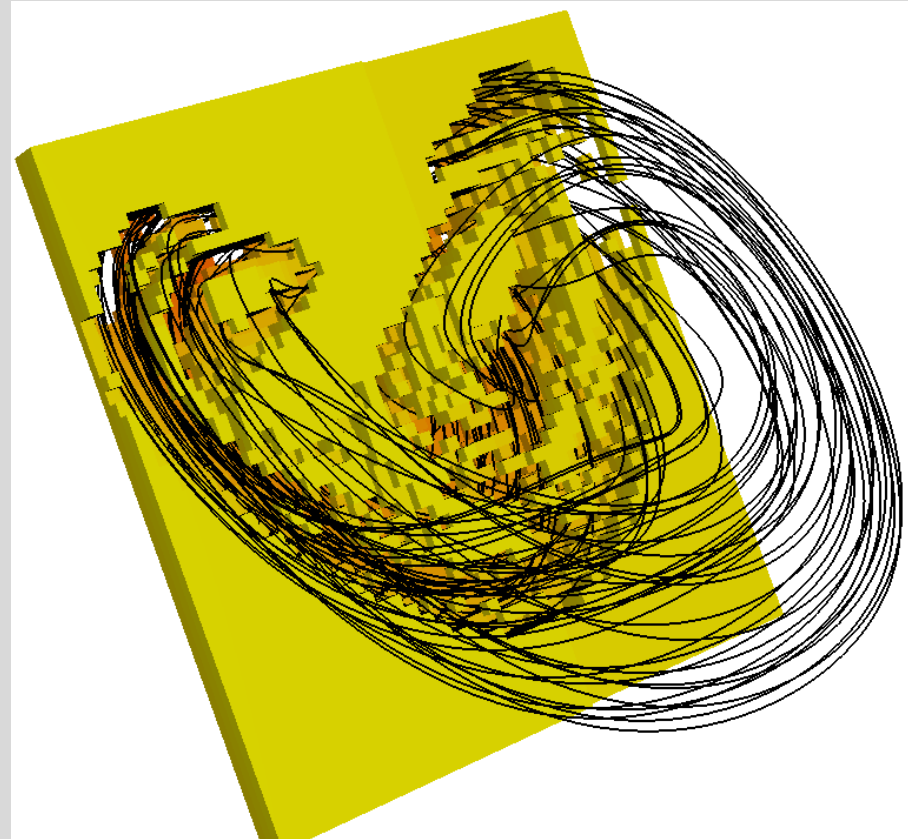
Topology of chaos

model

Phase portrait



Cereal crops attractor



To understand the dynamics of a chaotic system, it requires to understand the organisation of the flow in the phase space



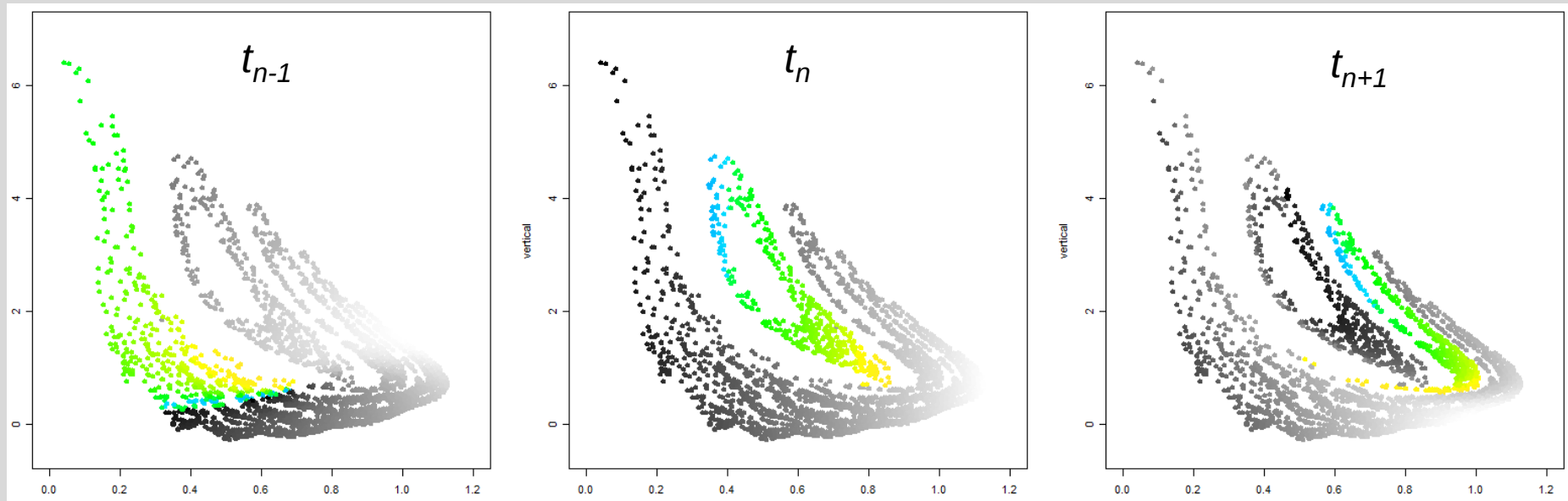
TOPOLOGY OF CHAOS

Poincaré section with color tracer

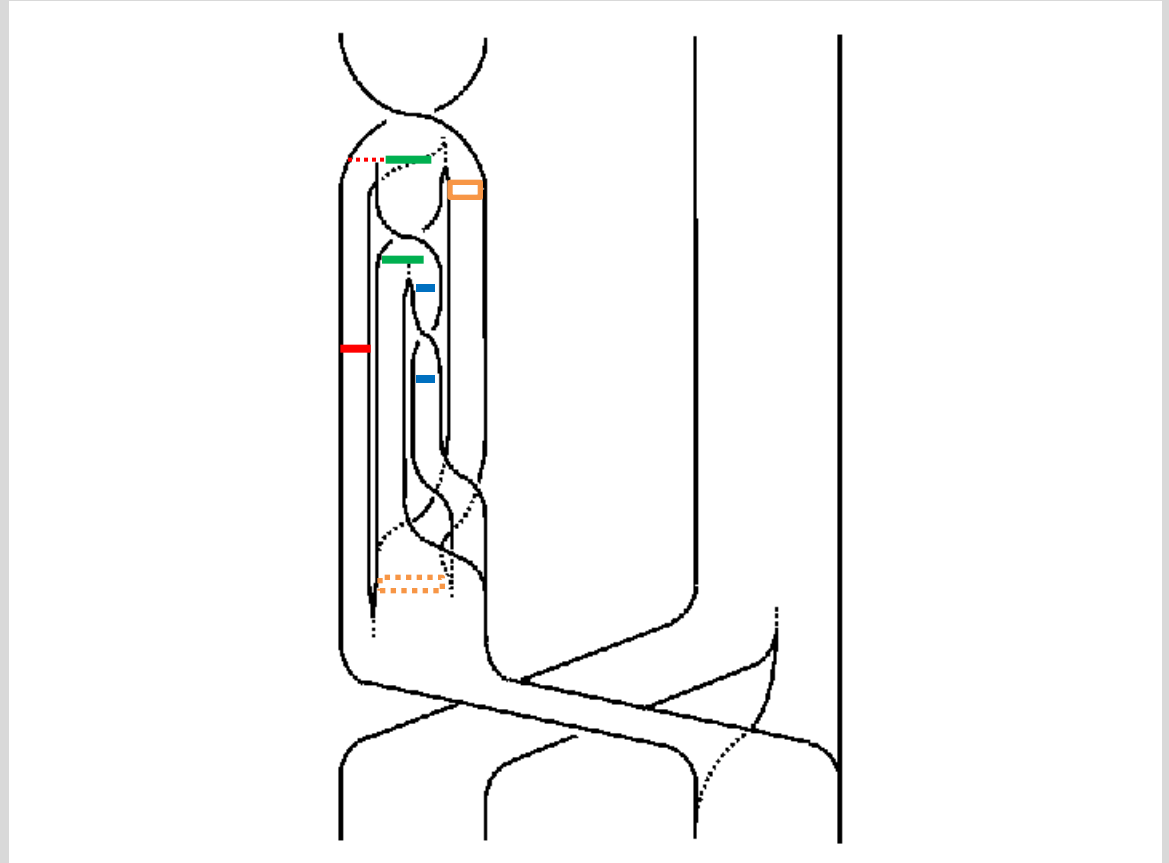
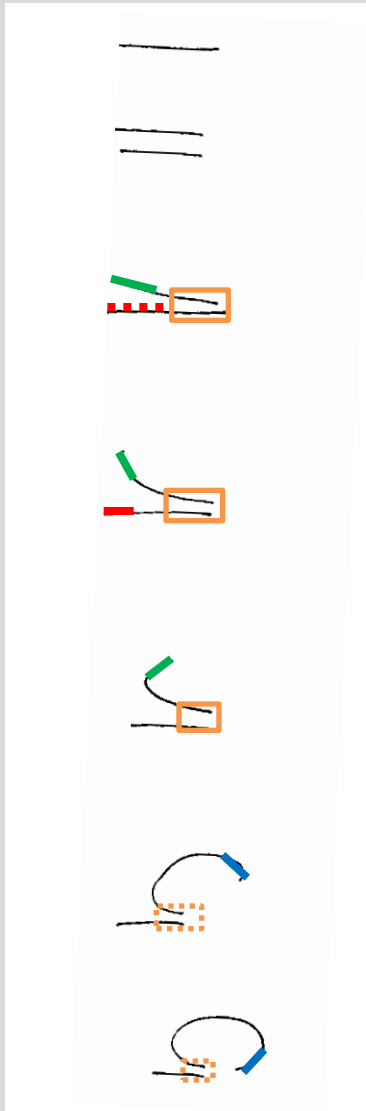
Back direction



Forth direction



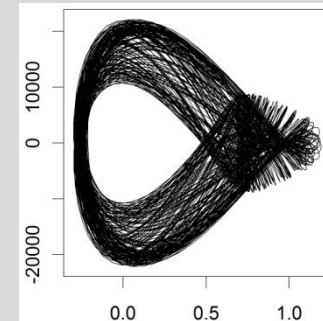
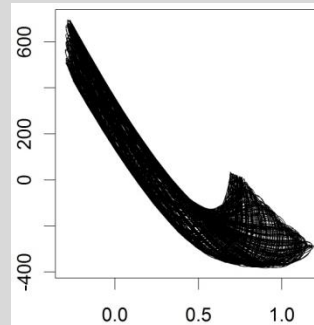
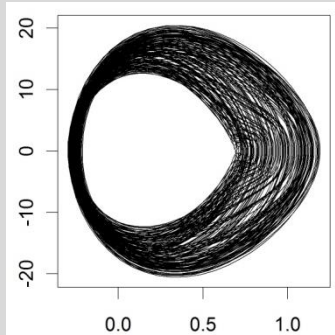
Template



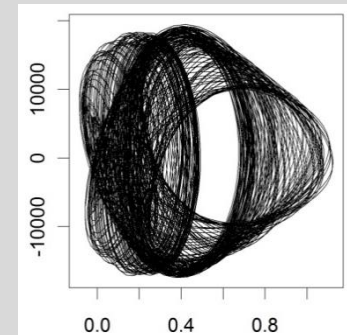
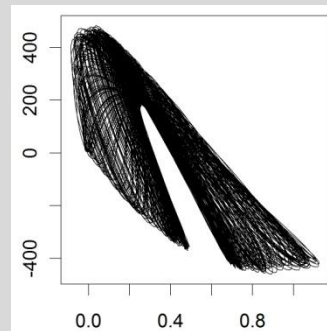
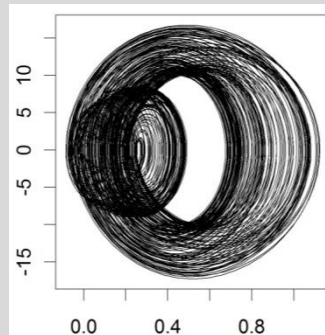
4D-models

Cereal crops (Al Ismailia)

model 1

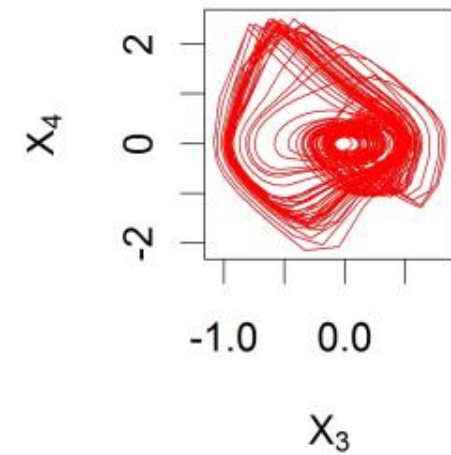
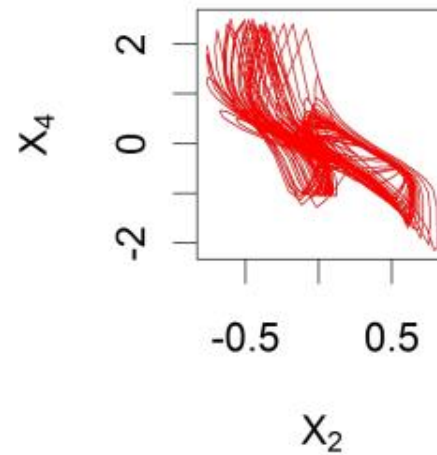
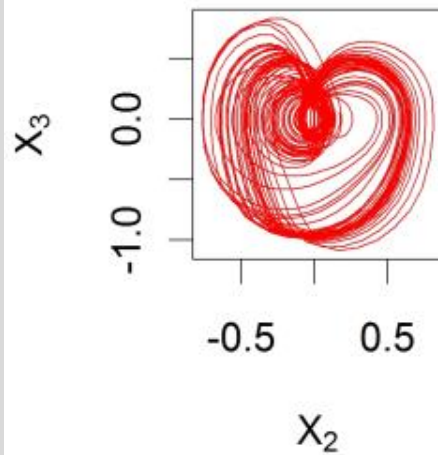
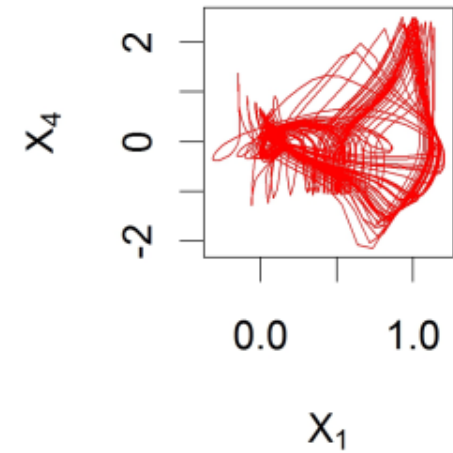
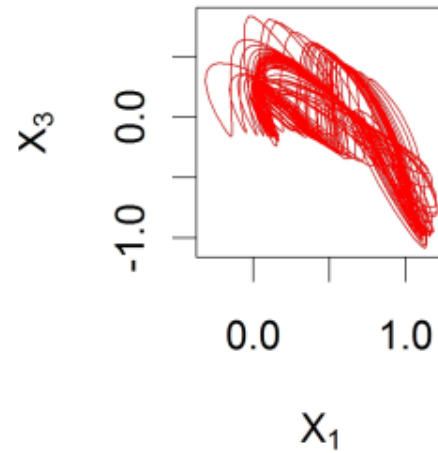
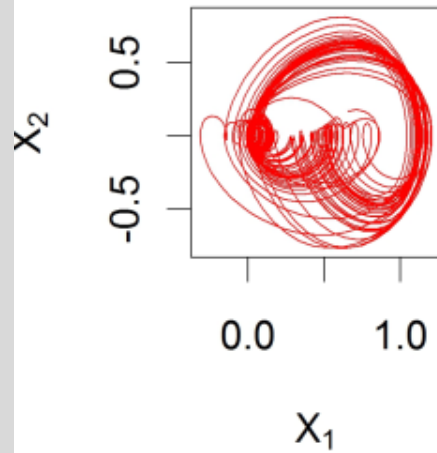


model 2



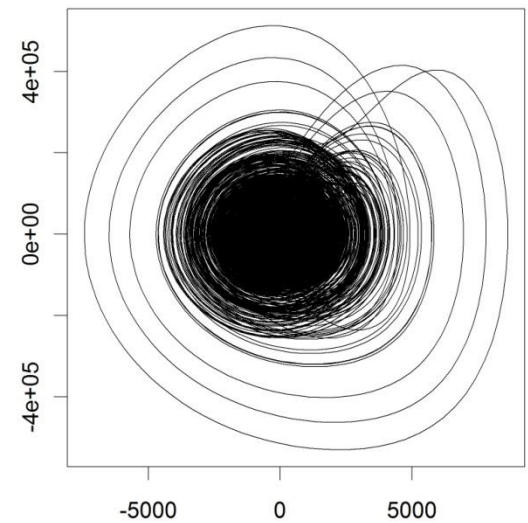
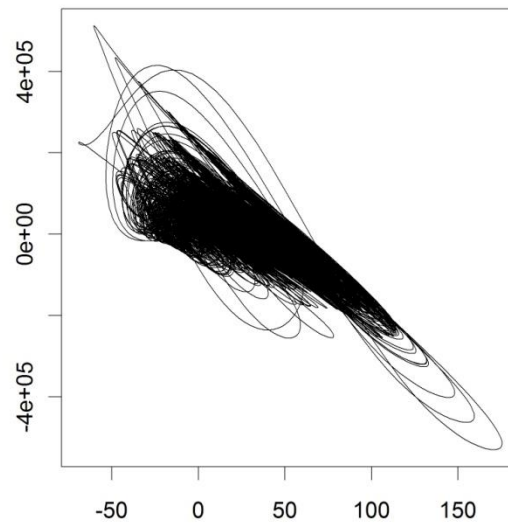
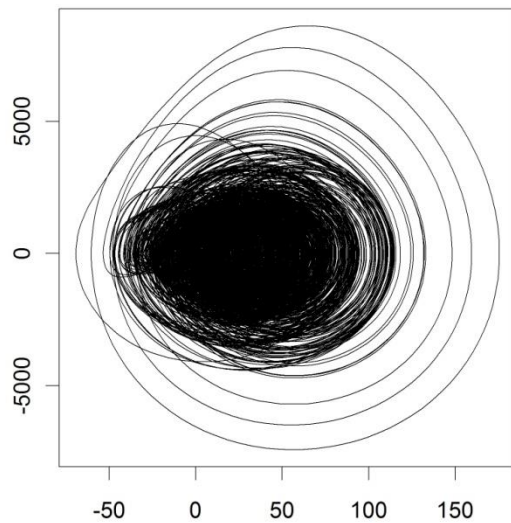
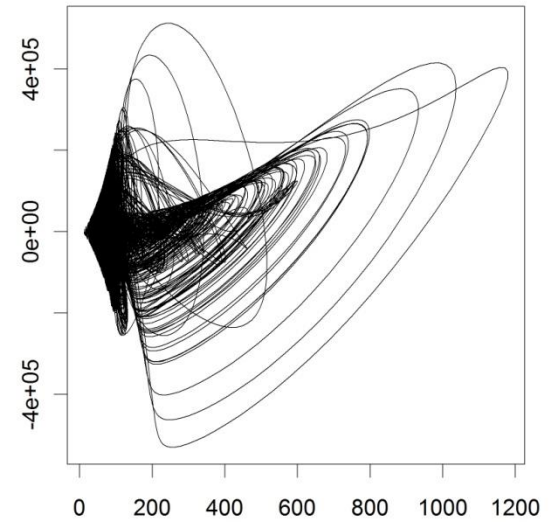
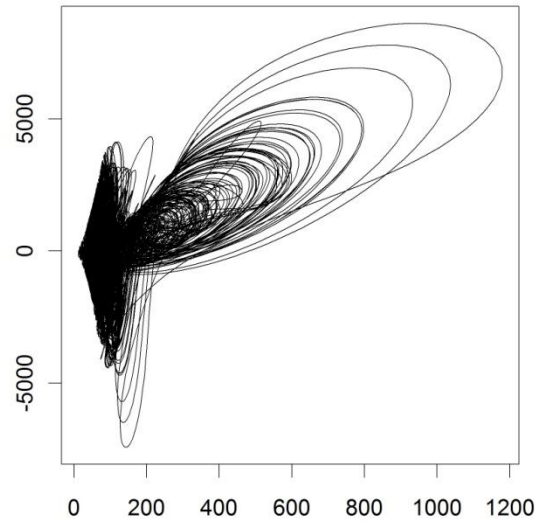
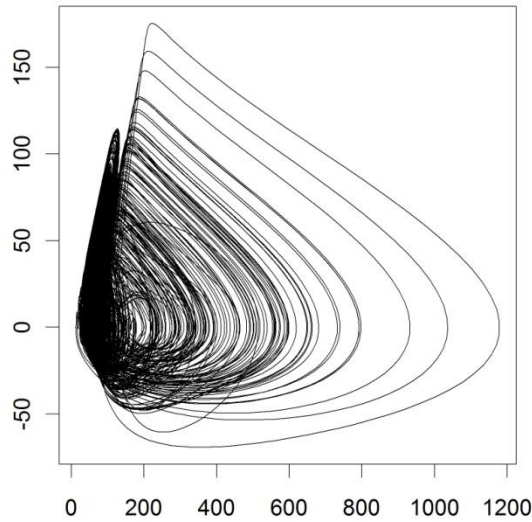
4D-models

E/E_0

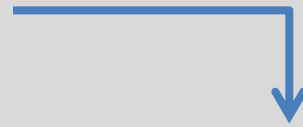


4D-models

Ebola attractor

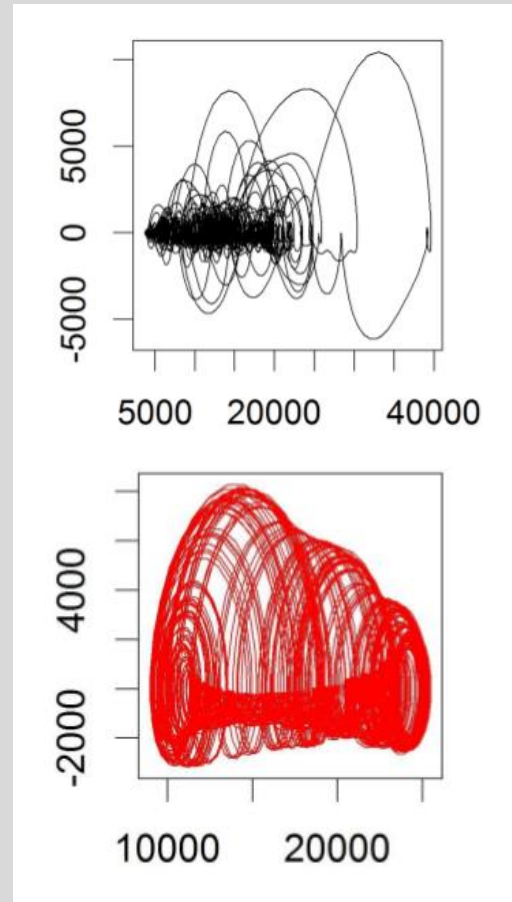


4D-models



Touvre

Karstic springs



Application to the Hénon attractor (1976)

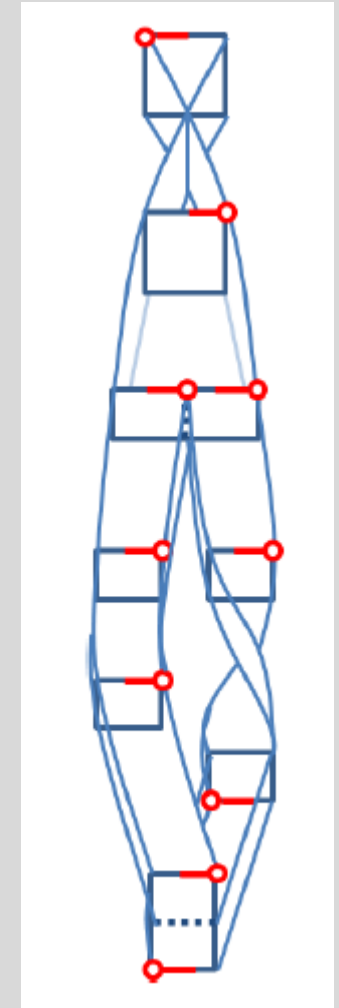
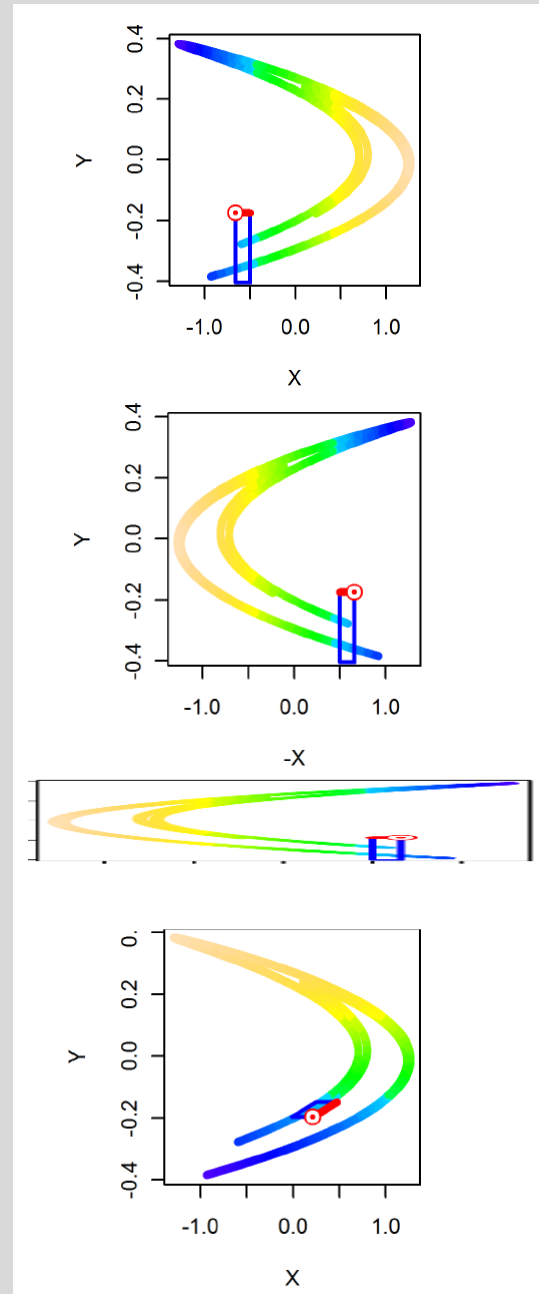


M. Hénon

$$\begin{cases} X_{i+1} = Y_i + 1 - aX_i^2 \\ Y_{i+1} = bX_i. \end{cases}$$

chronique of the torsions:

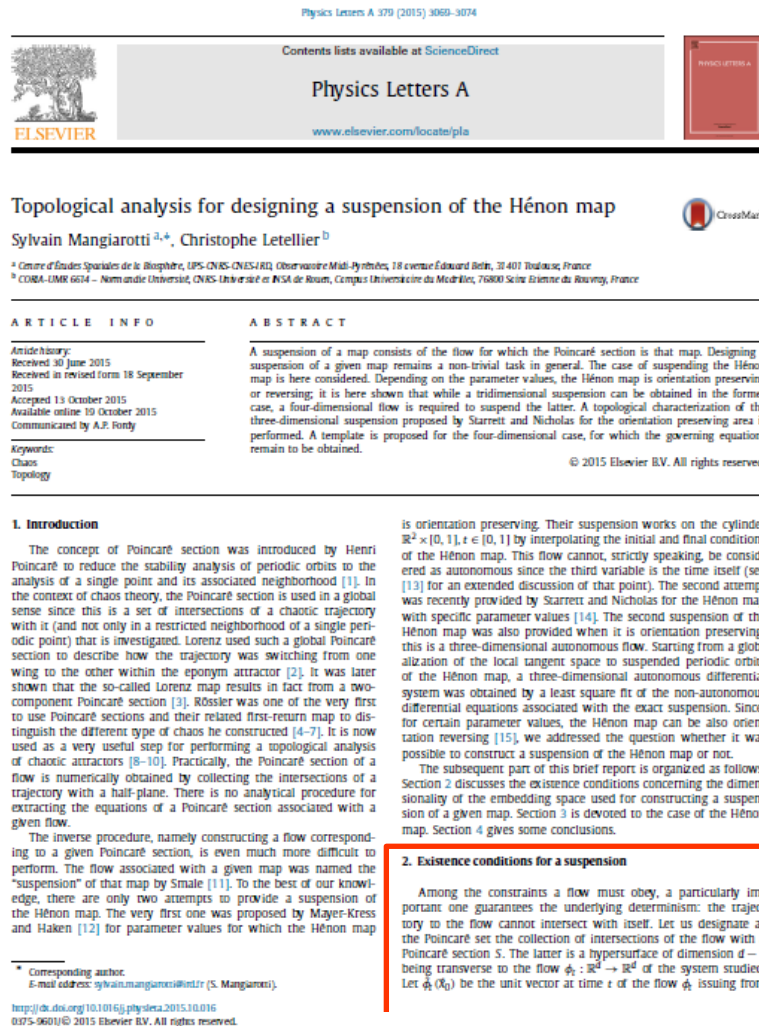
$$\left\{ (tz)^{\pm 4}, \begin{bmatrix} (tx)^{-4} & (tx)^{-2} \\ (tx)^{-2} & 1 \end{bmatrix} \right\}$$



Mangiarotti
HDR 2014

Topological analysis for designing a suspension of the Hénon map

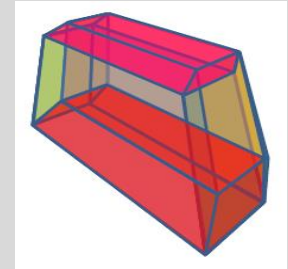
Mangiarotti & Letellier, Phys. Letters A 2015



Existence condition for a suspension

4D objects

- Visualization tools
- Extraction of multi-dimensional structures



Collaboration with J.-P. Jessel at [IRIT](#) (UPS)



Modelling platform GPoM



Acknowledgments

M. Huc, L. Drapeau

R. Coudret, F. Le Jean, M. Chassan, N. Martin



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TATA INSTITUTE OF FUNDAMENTAL RESEARCH

*Summer Research Program on
Dynamics of Complex Systems*
Discussion Meeting 21-23 July 2016



Thank you for your interest

