## Physics in Medicine: Fundamentals of Analyzing Biomedical Signals

#### Klaus Lehnertz

## Topics:

- theory of nonlinear dynamical systems
- characterizing measures
- biosignals and recording of biosignals
- applications (medicine, physics, biology)

http://epileptologie-bonn.de/cms/front\_content.php?idcat=203

# Literature:

## Physics:

- E. Ott: *Chaos in dynamical systems*. 2nd ed., Cambridge University Press, Cambridge UK (2002)
- H. Kantz, T. Schreiber: *Nonlinear time series analysis*. 2nd ed., Cambridge University Press, Cambridge (2003)
- A. Pikovsky, M. Rosenblum, J. Kurths: Synchronization: A universal concept in nonlinear science, Cambridge University Press, Cambridge (2001)

#### Medicine:

- E. Basar, T.H. Bullock: *Brain Dynamics*. Series in Brain Dynamics Vol. 2, Springer, Berlin (1989)
- E. Basar: Chaos in Brain Function. Springer, Berlin (1990)
- E.R. Kandel, J.H. Schwartz, T.M. Jessell: *Principles of Neural Science*. 4th ed., Elsevier North Holland, New York (2000)

## Literature:

#### Signal processing, Statistics, Computing:

- A.V. Oppenheim; A.S. Willsky: Signals and systems. Prentice Hall, 1996
- J.S. Bendat, A.G. Piersol: *Random Data: Analysis and measurement procedures*. 4th ed., Wiley Interscience, New York, 2010
- B.R. Martin: Statistics for Physicists. Academic Press, London, New York, 1971
- P.R. Bevington: *Data reduction and error analysis for the physical sciences*. McGraw-Hill, New York, 2002.
- W.H. Press, B.P. Flannery, S.A. Teukolsky, W.T. Vetterling: *Numerical Recipes. The art of scientific computing*. 3rd ed., Cambridge University Press, Cambridge, 2007

TISEAN: Nonlinear Time Series Analysis

https://www.pks.mpg.de/~tisean/

## **TISEAN** Nonlinear Time Series Analysis

Rainer Hegger Holger Kantz Thomas Schreiber

Go to Version 3.0.1 (released March 2007)

Go to Version 2.1 (released December 2000)

#### Introduction

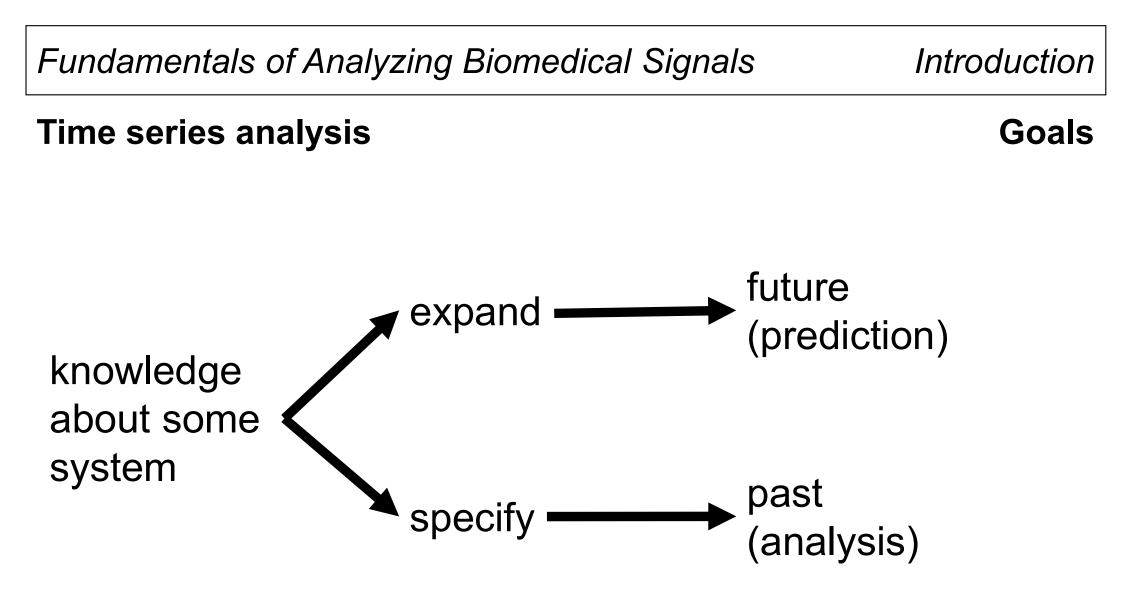
## Historical overview:

- 1778 P. Laplace (Laplace's demon, "everything is predictable")
- 1880 W. Sierpinski (non-Euclidian geometry, "mathematical monster")
- 1892 H. Poincare (three-body problem, dimension of manifolds)
- 1919 F. Hausdorff (extension of notion of dimension)
- 1963 E. Lorenz (weather forecasting)
- 1967 B. Mandelbrot (fractals, self similarity)
- 1975 J. Yorke (deterministic chaos)
- 1977/78 routes into chaos:
  - S. Grossmann/S. Thomae (period doubling)
  - M. Feigenbaum (Feigenbaum constant),
  - Newhouse-Ruelle-Takens route
- 1981 D. Ruelle (strange attactors),
  - P. Grassberger / I. Procaccia (correlation dimension)
  - F. Takens (state space reconstruction)
- since 1990 nonlinear time series analysis



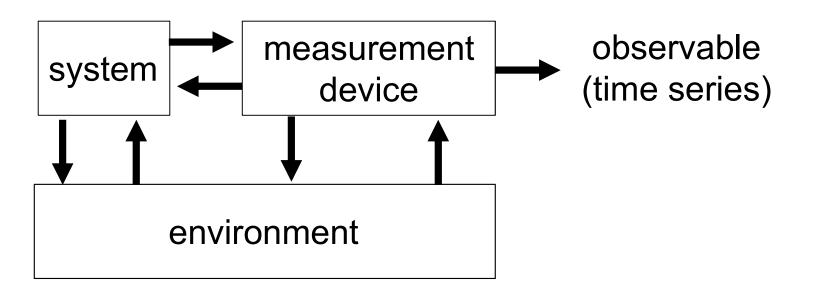






#### **Time series analysis**

#### **Measurement**



note: interactions !

- what is a suitable device ?
- what is a good observable ?
- what is a suitable environment?
- interfaces ?

#### Time series analysis

**Time series** 

*time series*: - sequence of data (length *N*)

- measurement or simulation (model)
- time-dependent

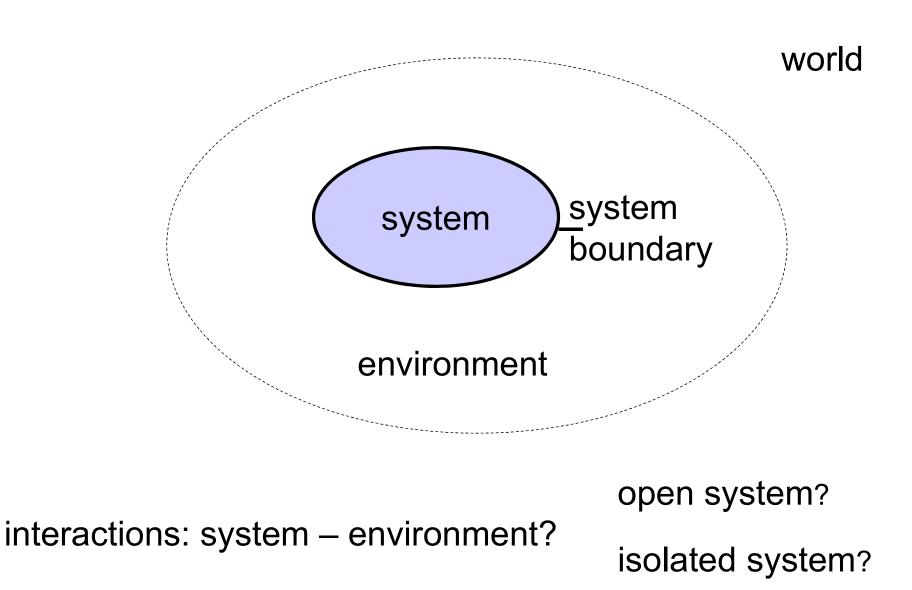
$$(v_i, v_{i+\Delta t}, \ldots, v_{N\Delta t})$$

∆t - temporal distance between successive data points
- sampling interval (measurement)

Fundamentals of Analyzing Biomedical Signals		als Introduction
Time series analysis		Time series
	experiment	model simulation
length of time series	(mostly) limited	user-defined
sampling interval	limited	user-defined
precision	A/D converter noise	user-defined

Introduction

#### **System**



#### Introduction

### **Dynamical system**

- system under influence of some force ( $\delta \nu \nu \alpha \mu \nu \sigma$  = force)
- time-dependent system states
- state changes depend on current state

#### deterministic

same initial states ↓ same evolution

#### stochastic

same initial states ↓ random evolution

### **Dynamical system**

- characterized by time-dependent state variables  $\mathbf{x}(t) \in \mathbb{R}^d$
- temporal evolution of state variables:

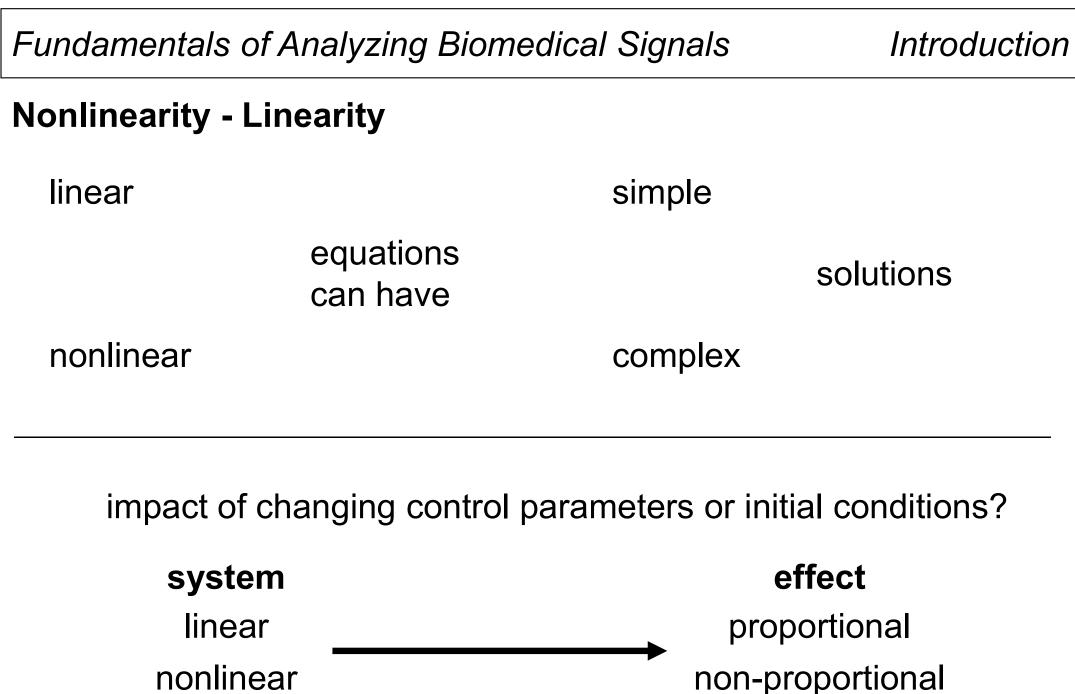
*continuous case:* set of (first-order) ordinary differential equations with initial conditions  $\mathbf{x}(0) = \mathbf{x}(t_0)$ 

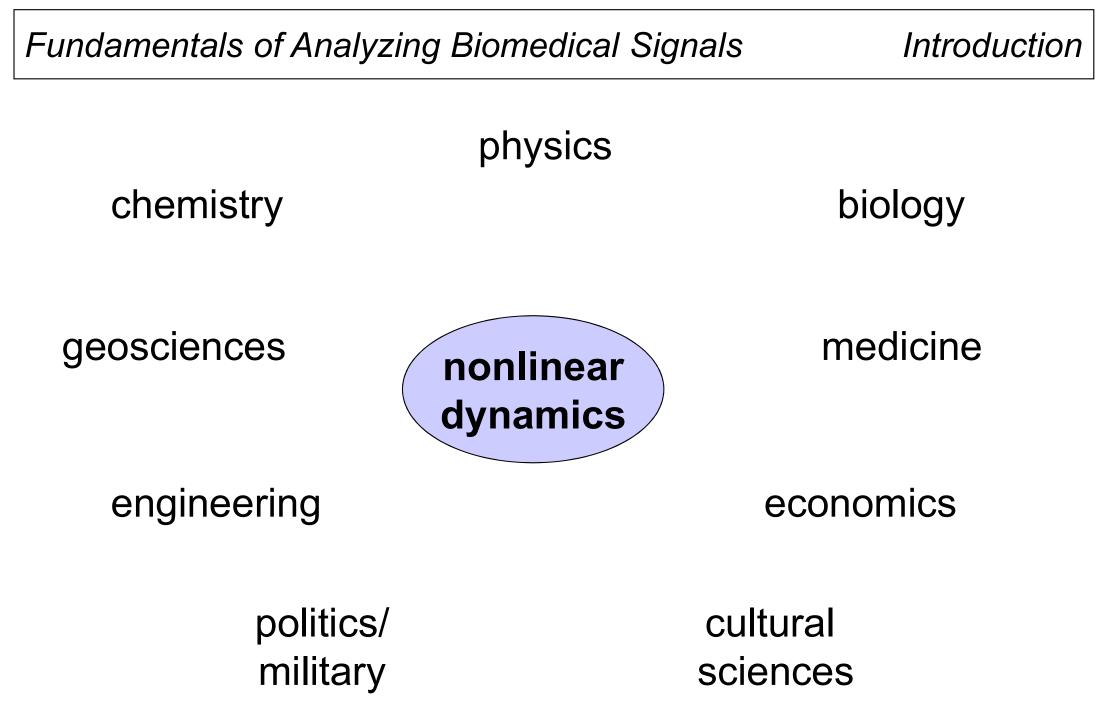
$$\frac{\mathrm{d}\mathbf{x}(t)}{\mathrm{d}t} = f(t, \mathbf{x}(t), \beta)$$

*discrete case:* set of difference equations (mapping) with initial conditions  $\mathbf{x}_0 = \mathbf{x}_{t_0}$ 

$$\mathbf{x}_{t+\Delta t} = F(T, \mathbf{x}_t, \beta)$$

with d = dimension of system;  $\beta$  = control parameter; f, F = nonlinear functions in case of nonlinear systems





# condensed matter physics

pattern formation phase transitions spin waves

#### fluid mechanics

transition to turbulent motion crystal growth surface of liquids

#### laser physics

laser instabilities semiconductor laser coupled laser

#### mechanics

nonlinear oscillators coupled/forced pendulums magneto-mechanic oscillators torsion bar

nonlinear dynamics in physics

optics

opto-galvanic systems nonlinear optics

#### acoustics

sound generation with:

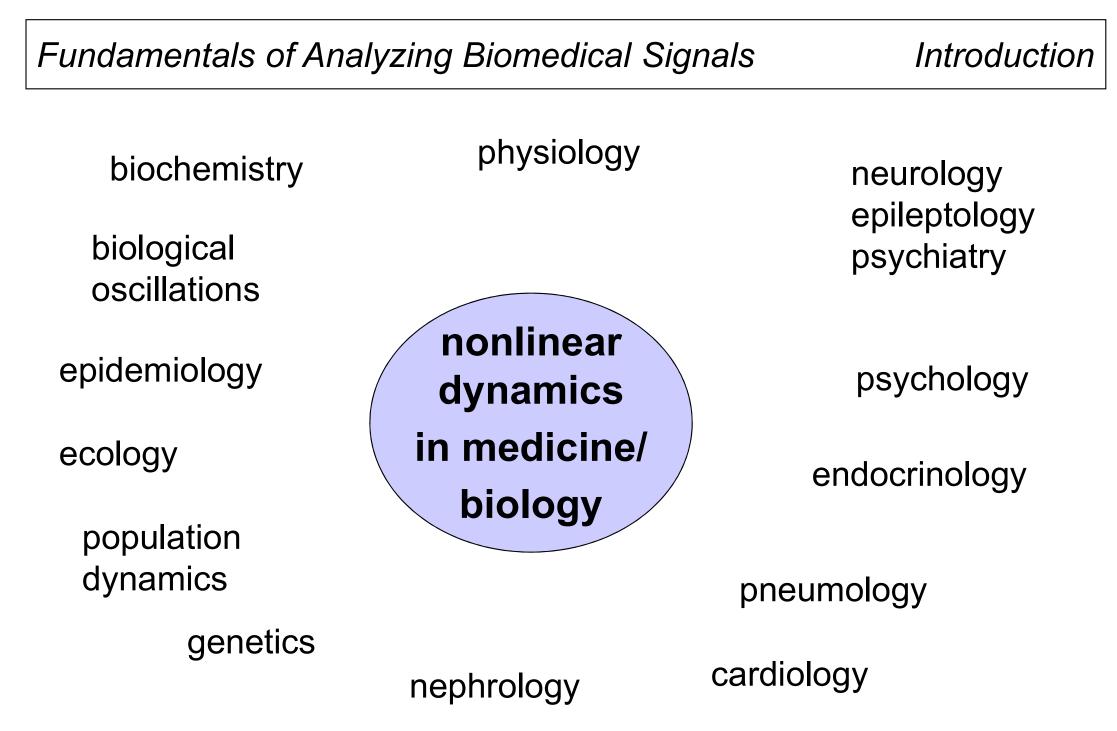
- laser
- musical instruments

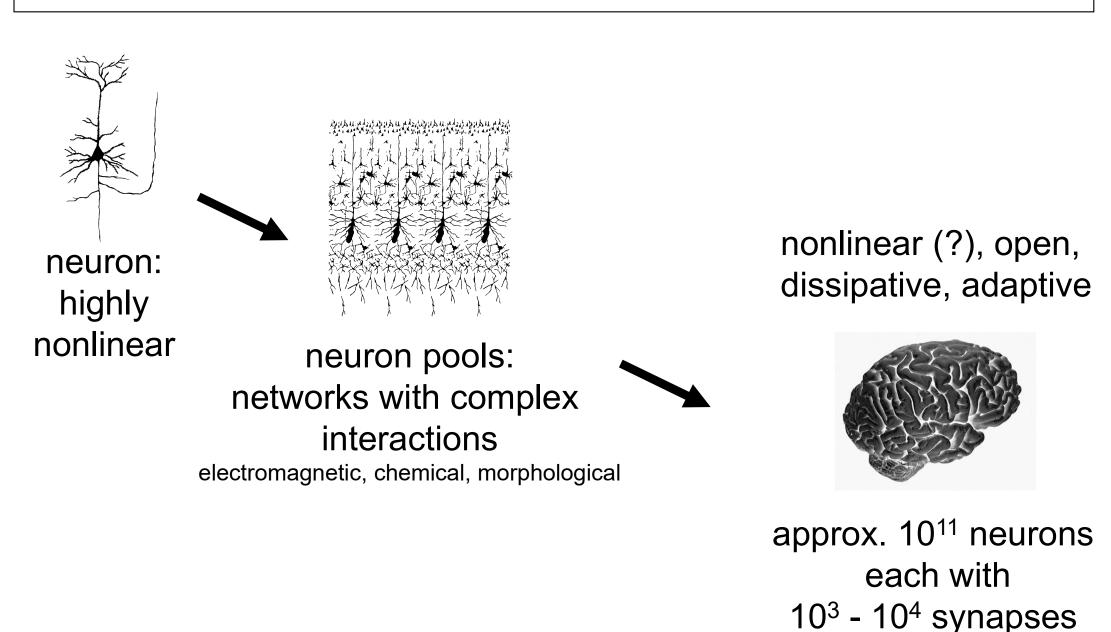
#### astrophysics

solar system motion of stars sun spots pulsar/quasar distribution of galaxies

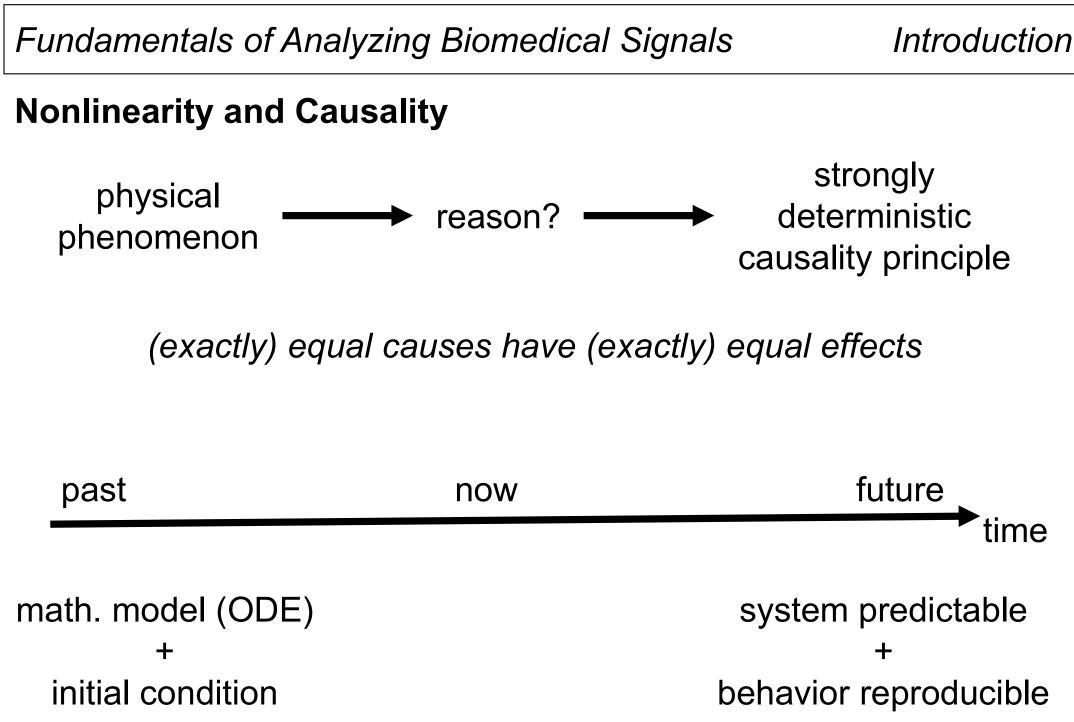
#### plasma physics

oscillations in gas discharges pattern formation plasma waves

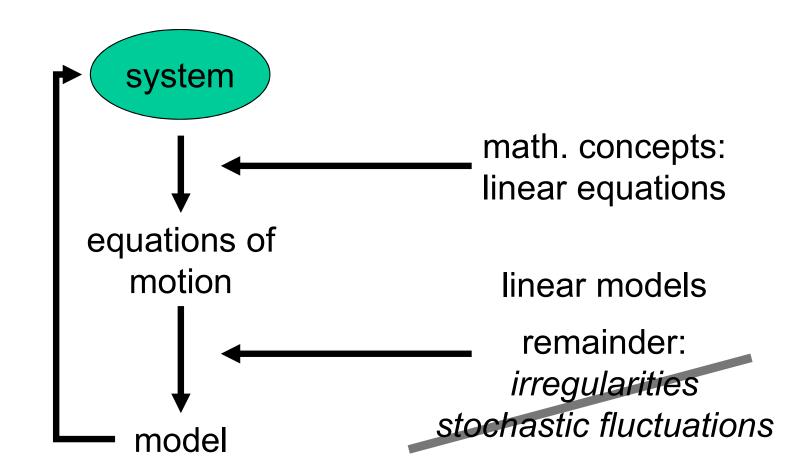




Introduction



#### The pragmatic perspective of a linear nature



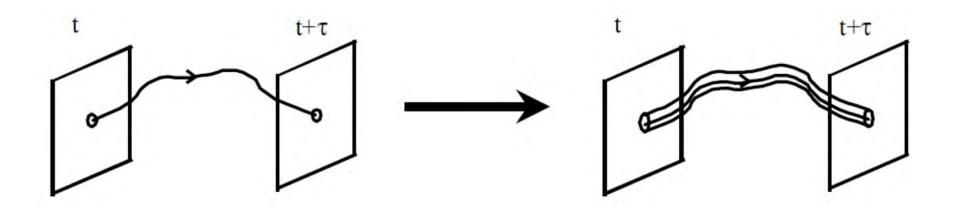
this perspective challenged by Poincaré and Sierpinski

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#### **Nonlinearity and Causality**

**Linear Systems** 

weak causality: equal causes  $\rightarrow$  equal effects strong causality: similar causes  $\rightarrow$  similar effects



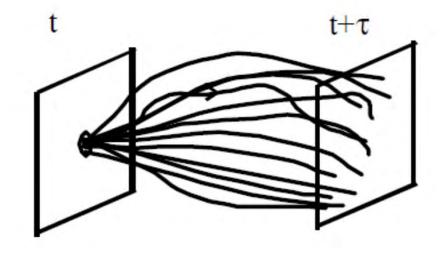
strong idealization; does not account for experimental conditions includes weak causality; accounts for experimental conditions: tiny deviations from initial conditions, weak perturbations, systematic errors, ...

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#### **Nonlinearity and Causality**

**Nonlinear Systems** 

#### violation of strong causality: similar causes $\rightarrow$ vastly different effects



- sensitive dependence on initial conditions
- deterministic chaos
- pattern formation
- "the whole is more than the sum of its parts" (Aristoteles)
- self-organization

#### **Processes and their Characteristics**

regular process	chaotic process	stochastic process
deterministic	deterministic	stochastic (noise/randomness)
long-term predictable	predictable	non-predictable
strong causality	violation of strong causality	no causal relationships
	nonlinearity	

#### **Deterministic Chaos**

Chaos (colloquially)

- disordered state and irregularity

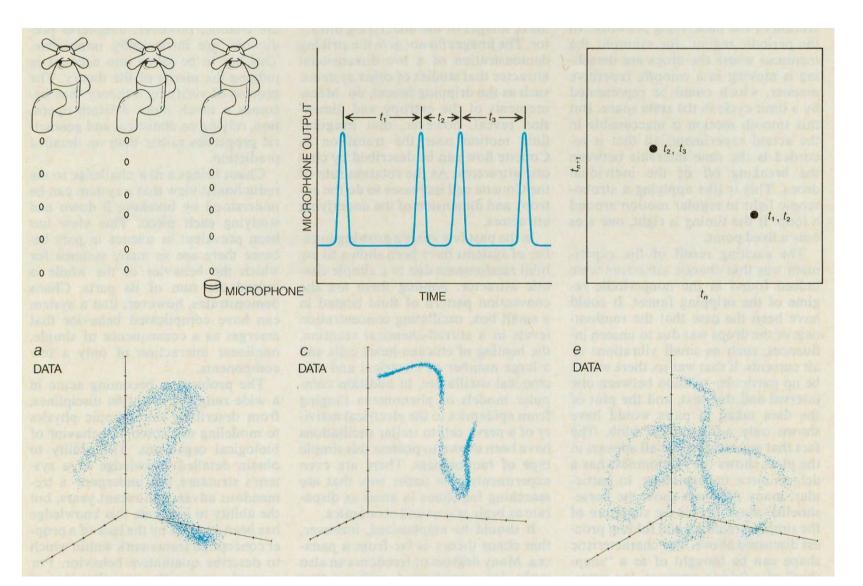
#### **Deterministic Chaos**

- irregular (non-periodic) behavior
- non-predictable or for some short time horizon only
- deterministic equations of motion (in contrast to stochasticity)
- instabilities and recurrences

#### Introduction

#### **Deterministic Chaos**

dripping faucet



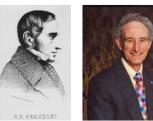


The Science Frontier Express Series

Introduction

#### **Deterministic Chaos**

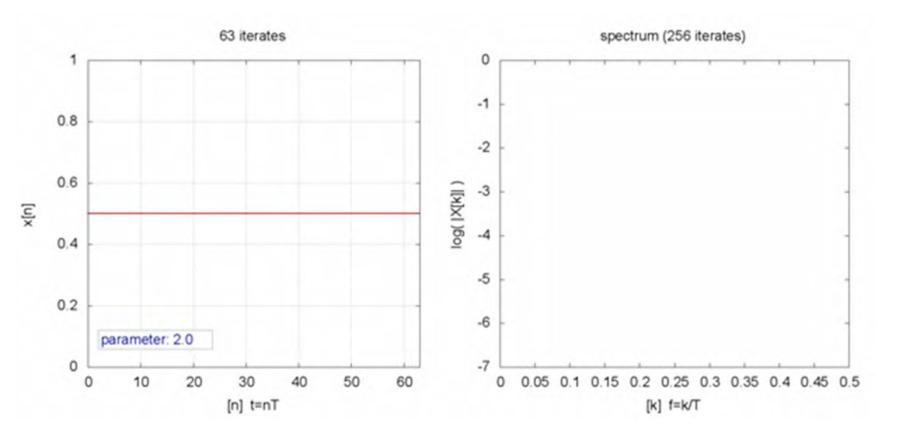
#### period doubling



"chaotic behavior can arise from very simple non-linear dynamical equations": *logistic map* (model for population growth, 1837)

 $x_{n+1} = rx_n(1-x_n); \ x_n \in [0,1]; \ r \in [0,4]$ 

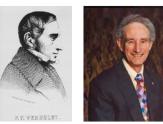
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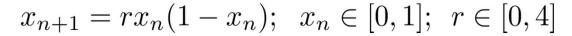
#### **Deterministic Chaos**

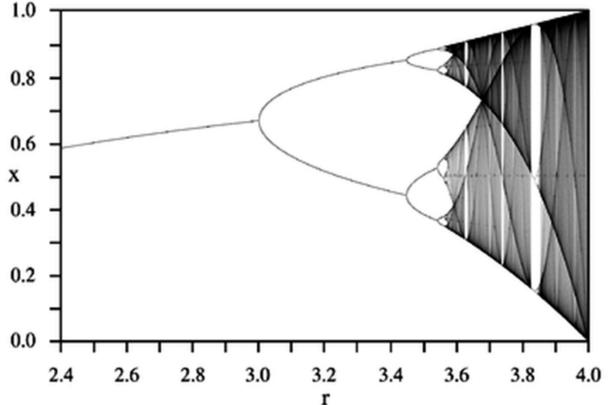
#### period doubling



"chaotic behavior can arise from very simple non-linear dynamical equations": *logistic map* (model for population growth, 1837)

P.F. Verhulst R. May





bifurcation diagram

- period-doubling route to chaos
- period-3 implies chaos
- islands of stability
- periodicities within chaos
- self-similarity

### **Deterministic Chaos**

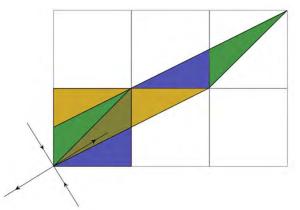


V. Arnold (1937-2010)

## Arnold's cat map:

chaotic map from the torus into itself:  $\Gamma:\mathbb{T}^2\to\mathbb{T}^2$ 

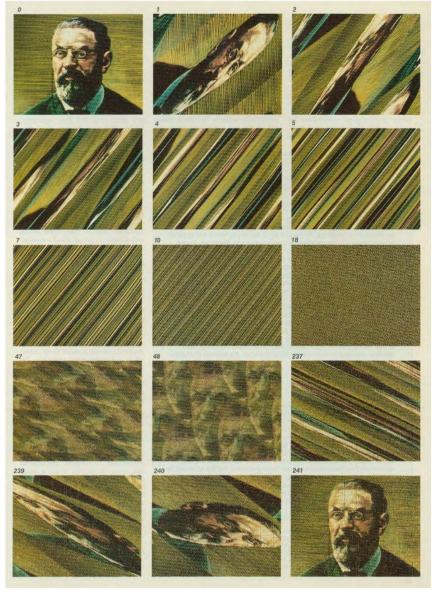
$$\Gamma: (x, y) \to (2x + y, x + y) \mod 1$$



deterministic operations:

- stretching
- bending
- folding (nonlinear)

#### recurrence and self-similarity



#### Nonlinear dynamical systems

- can be described by nonlinear ODEs. However, no analytic solutions exist!
- show qualitatively rich dynamics:
   drastic changes upon changes of control parameters (bifurcation)
   deterministic chaos
- long-term behavior can be assessed by investigating the phase-space (state-space)