

**Lorenz Lecture – AGU'05**

**MPI für PKS, Dresden, 27 Feb. 2006**

# **The Earth as a Complex System, and a Simple Way of Looking at It**

**Michael Ghil**

Ecole Normale Supérieure, Paris, &  
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Work with *D. Dee* (NASA Goddard), *V. Keilis-Borok* (IGPP, UCLA, & MITP, Moscow),  
*A. Mullhaupt* (Wall Street), *P. Pestiaux* (TotalFina, France),  
*A. Saunders* (UCLA), & *I. Zaliapin* (IGPP, UCLA, & MITP, Moscow).

Edward Norton Lorenz  
born May 23, 1917



Jule Gregory Charney  
January 1, 1917 – June 16, 1981

# Motivation

## 1. *Components*

- solid earth (crust, mantle)
- fluid envelopes (atmosphere, ocean, snow & ice)
- living beings on and in them (fauna, flora, people)

## 2. *Complex feedbacks*

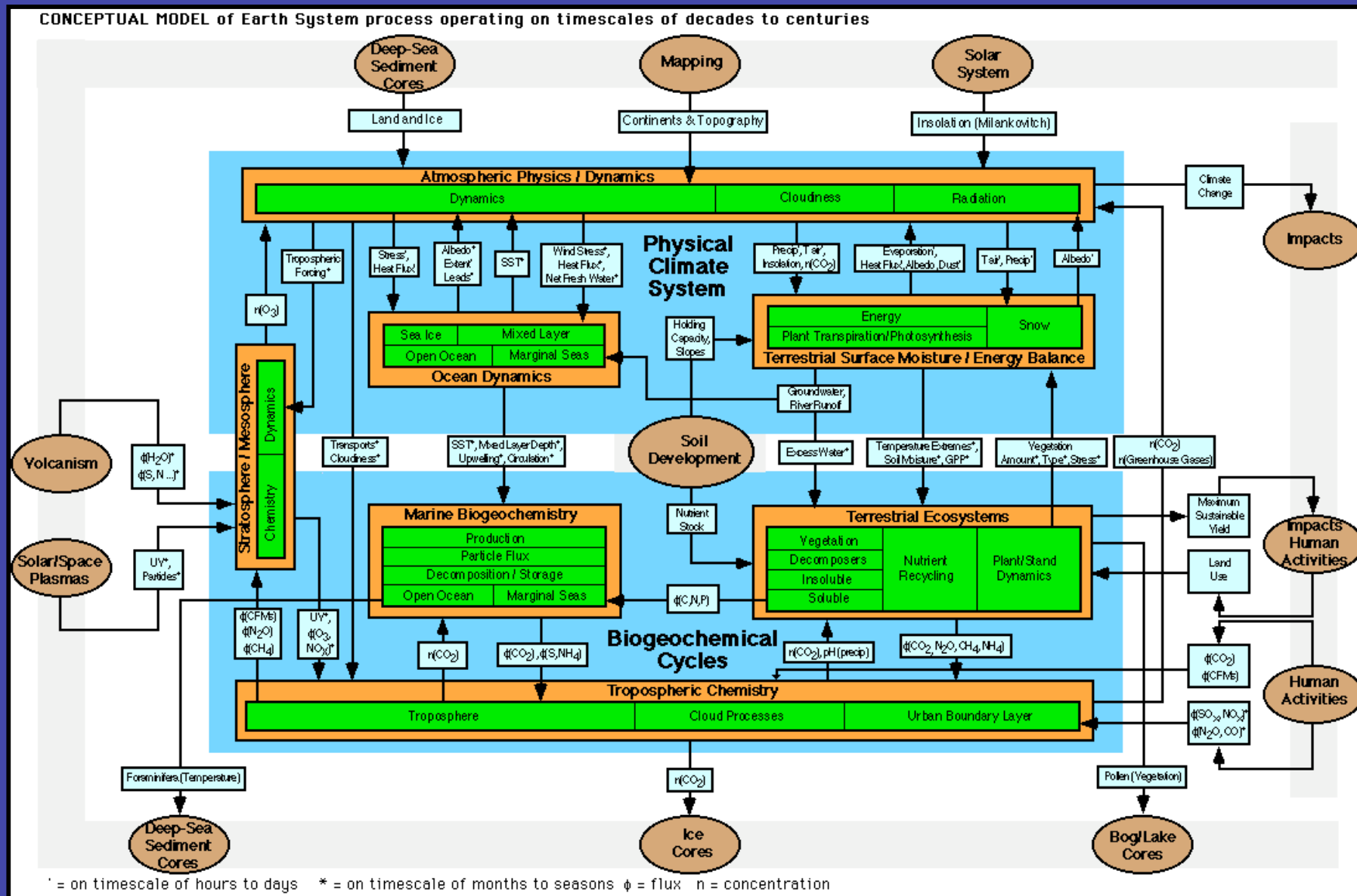
- **positive** and **negative**
- **nonlinear** - small pushes, big effects?

## 3. *Approaches*

- **reductionist**
- **holistic**

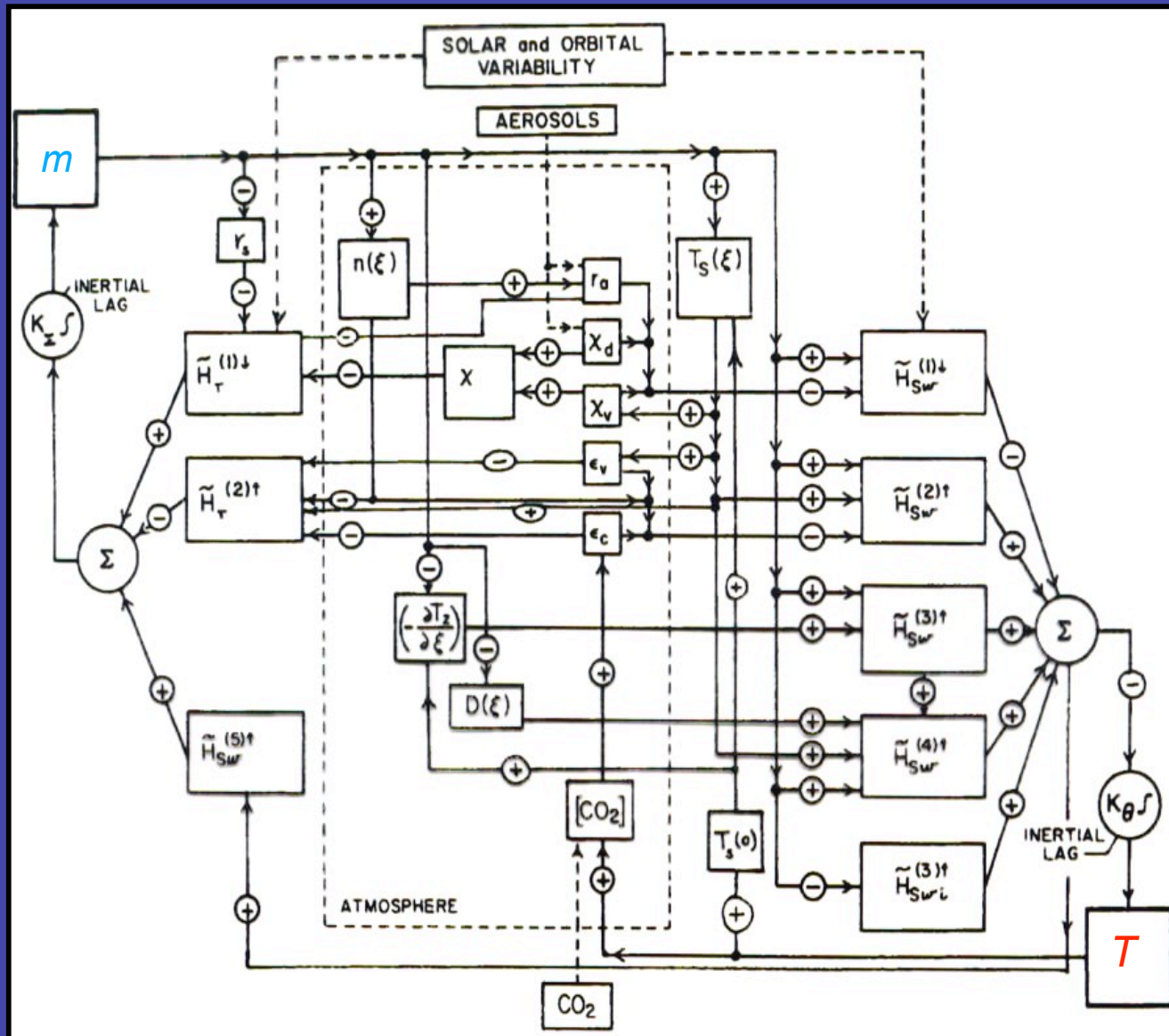
## 4. *What to do?* - Let's see!

# F. Bretherton's "horrendogram" of Earth System Science



# The climate system on long time scales

“Ambitious” diagram



Flow diagram showing feedback loops contained in the dynamical system for ice-mass  $m$  and ocean temperature variations  $T$ .

Constants for ODE & PDE models are poorly known. Mechanisms and effective delays are easier to ascertain.

# Introduction

## *Binary systems*

Examples: Yes/No, True/False (ancient Greeks)

## *Classical logic* (*Tertium not datur*)

Boolean algebra (19<sup>th</sup> cent.)

Propositional calculus (20<sup>th</sup> cent.)

(syllogisms as trivial examples)

## *Genes: on/off*

Descriptive – Jacob and Monod (1961)

Mathematical *genetics* – L. Glass, S. Kauffman, M. Sugita (1960s)

*Symbolic dynamics* of differentiable dynamical systems (DDS): S. Smale (1967)

## *Switches: on/off, 1/0*

Modern *computation* (EE & CS)

- cellular automata (CAs) J. von Neumann (1940s, 1966), S. Ulam,

Conway (the game of life), S. Wolfram (1970s, '80s)

- spatial increase of complexity –

infinite number of channels

- conservative logic Fredkin & Toffoli (1982)

- kinetic logic: importance of distinct delays

to achieve temporal increase in complexity (synchronization,

operating systems & parallel computation), R. Thomas (1973, 1979,...)

# Introduction (cont.)

*M.G.*'s immediate motivation:

*Climate dynamics* – complex interactions  
(reduce to binary), C. Nicolis (1982)

*Joint work* on developing and applying BDEs to climate dynamics  
with D. Dee, A. Mullhaupt & P. Pestiaux (1980s)  
& with A. Saunders (late 1990s)

*Work* of L. Mysak and associates (early 1990s)

*Recent applications* to solid-earth geophysics  
(*earthquake modeling and prediction*)  
with V. Keilis-Borok and I. Zaliapin

*Recent applications* to the biosciences  
(*genetics and micro-arrays*)  
Oktem, Pearson & Egiazarian (2003) *Chaos*  
Gagneur & Casari (2005) *FEBS Letters*

# Outline

## **What for BDEs?**

- *life is sometimes too complex for ODEs and PDEs*

## **What are BDEs?**

- *formal models of complex feedback webs*
- *classification of major results*

## **Applications to climate modeling**

- *paleoclimate – Quaternary glaciations*
- *interdecadal climate variability in the Arctic*
- *ENSO – interannual variability in the Tropics*

## **Applications to earthquake modeling**

- *colliding-cascades model of seismic activity*
- *intermediate-term prediction*

## **Concluding remarks**

- *bibliography*
- *future work*



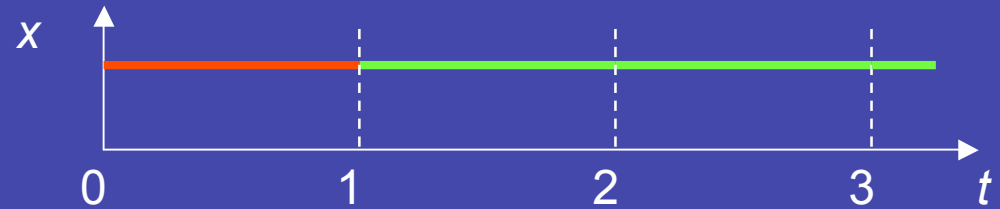
# What are BDEs?

## Short answer:

Maximum simplification of nonlinear dynamics  
(non-differentiable time-continuous dynamical system)

## Longer answer:

1)  $x \in B = \{0, 1\}$   
 $x(t) = x(t - 1)$   
 (simplest EBM:  $x = T$ )



2)  $x(t) = \bar{x}(t - 1)$

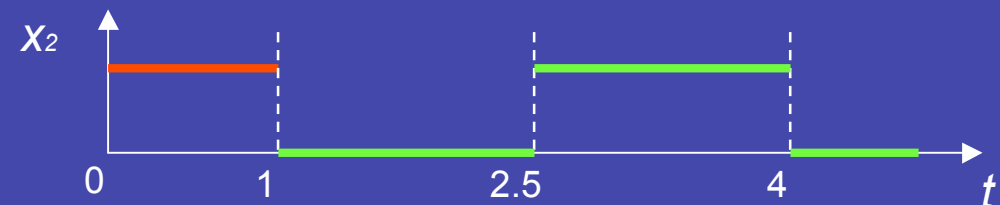


3)  $x_1, x_2 \in B = \{0, 1\}; 0 < \theta \leq 1$

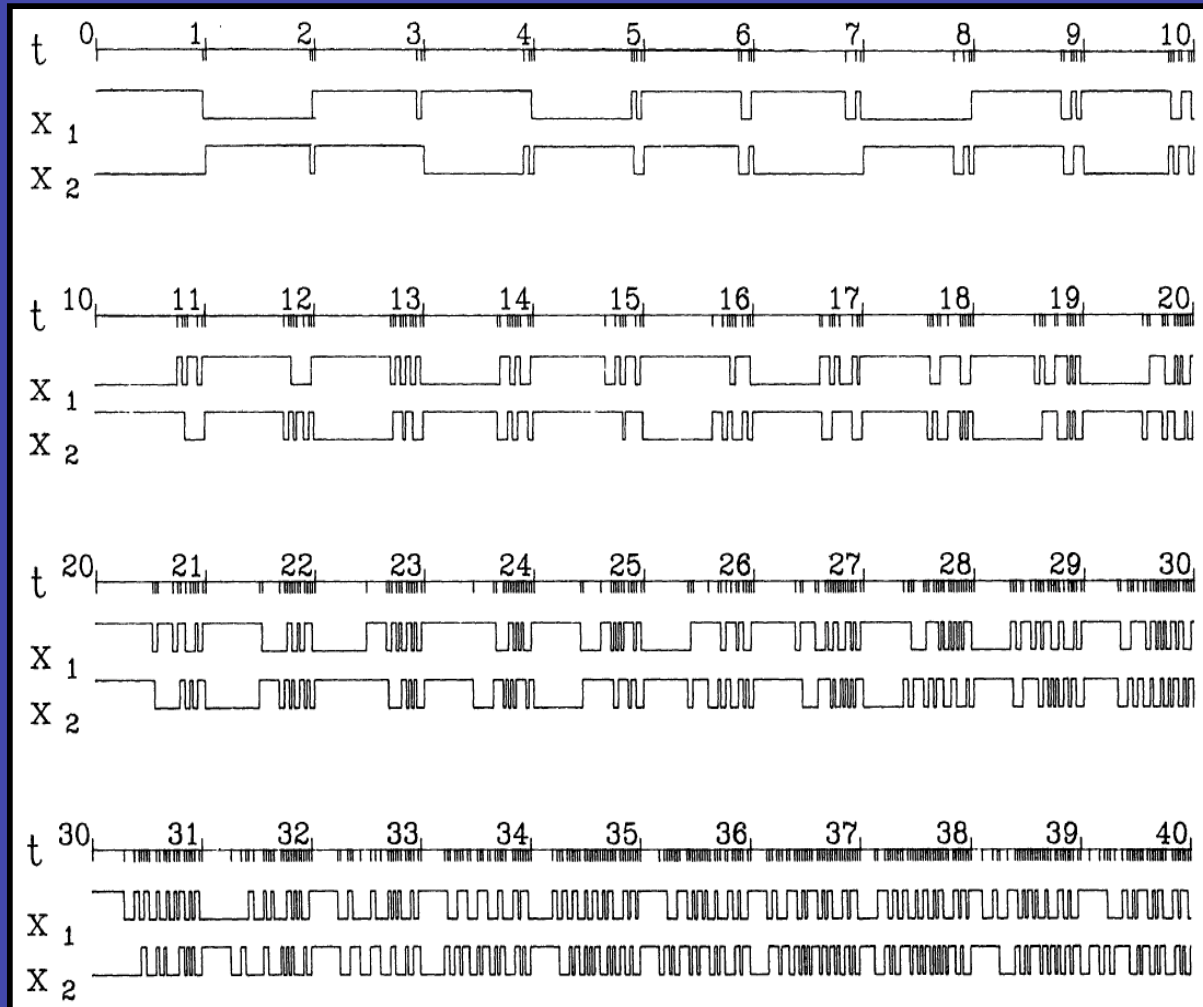
$$\begin{cases} x_1(t) = x_2(t - \theta), \theta = 1/2 \\ x_2(t) = \bar{x}_1(t - 1) \end{cases}$$



Eventually periodic with  
a period =  $2(1+\theta)$   
(simplest OCM:  $x_1=m, x_2=T$ )



$$\begin{cases} x_1(t) = x_2(t - \theta) \\ x_2(t) = x_2(t - 1) \nabla x_1(t - \theta) \end{cases} \quad \theta \text{ is irrational}$$



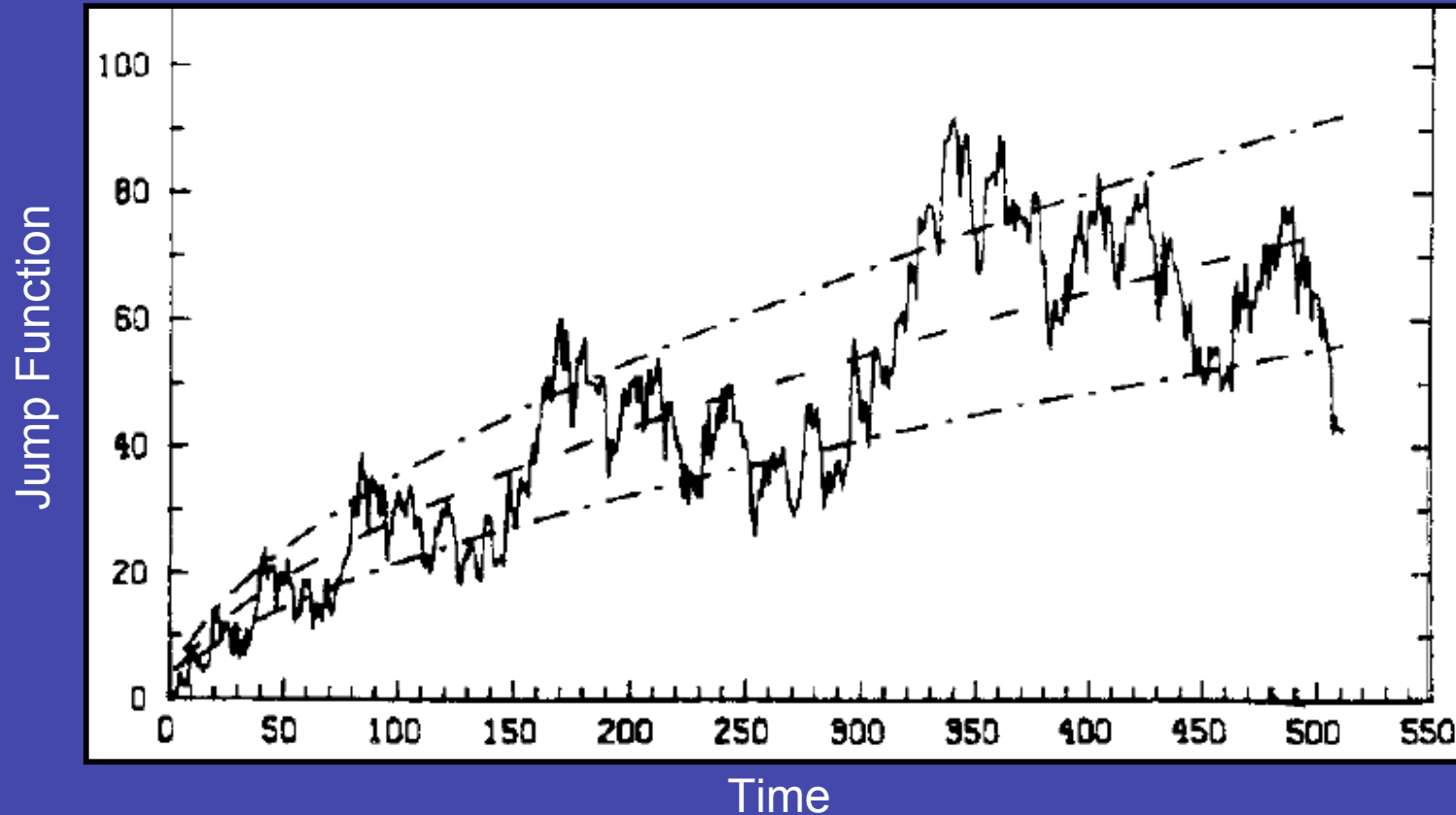
Increase in complexity!

**Evolution:** biological, cosmogonic, historical

But how much?

Aperiodic solutions with *increasing complexity*

$$x(t) = x(t-1)\nabla x(t-\theta), \quad \theta = \frac{\sqrt{5}-1}{2} = \text{"golden ratio"}$$

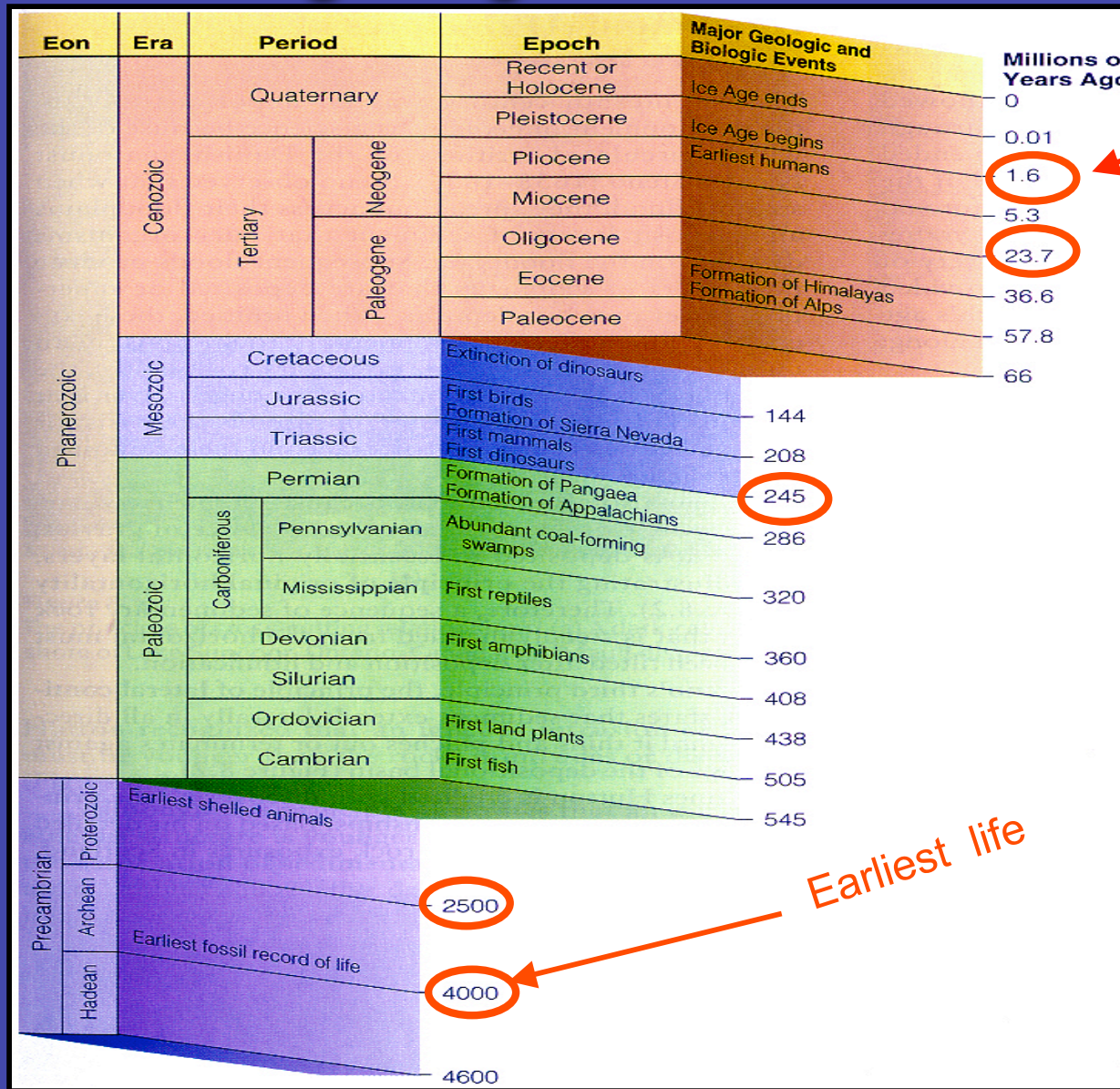


### Theorem:

Conservative BDEs with irrational delays have aperiodic solutions with a *power-law increase in complexity*.

N.B. Log-periodic behavior!

# The geological time scale



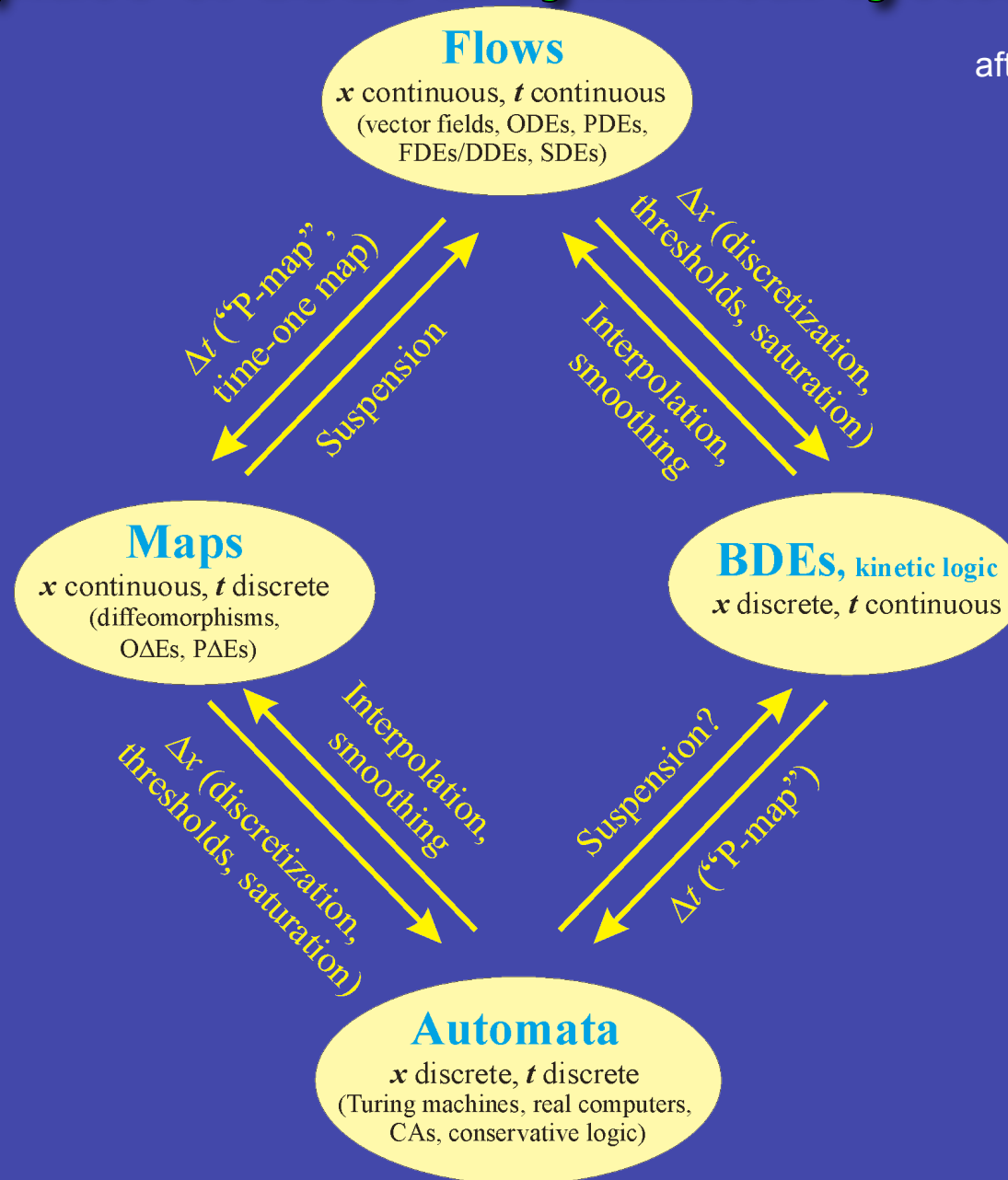
Ice age begins

Earliest life

Density of events  $\cong \log(t)$

# The place of BDEs in dynamical system theory

after A. Mullhaupt (1984)



# Classification of BDEs

**Definition:** A BDE is *conservative* if its solutions are immediately periodic, i.e. no transients; otherwise it is *dissipative*.

**Remark:** Rational vs. irrational delays.

**Example:**

1) Conservative

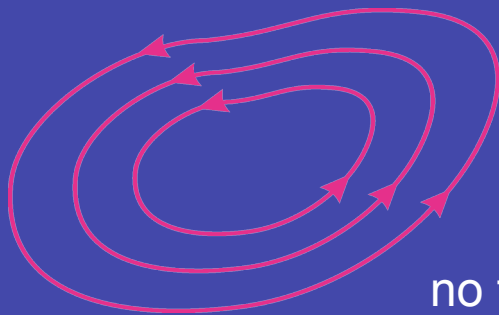
$$x(t) = \bar{x}(t - 1)$$

2) Dissipative

$$x(t) = x(t - 1) \wedge x(t - \theta)$$

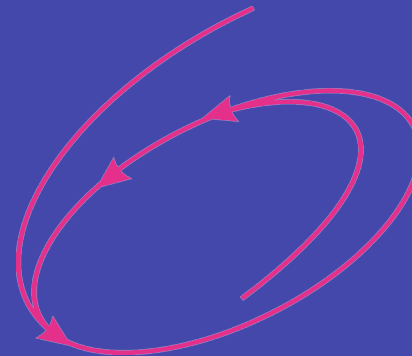
**Analogy with ODEs**

Conservative – Hamiltonian



no transients

Dissipative – limit cycle



attractor

**Examples.** Convenient shorthand for scalar 2<sup>nd</sup> order BDEs

$$x = y \circ z \Leftrightarrow x(t) = x(t-1) \circ x(t-\theta)$$

### 1. Conservative

$$x = y \nabla z = y \oplus z = y + z \pmod{2}$$

$$x = y \Delta z = 1 \oplus y \oplus z$$

Remarks: i) Conservative  $\equiv$  linear (mod 2)

ii)  $\exists$  few conservative connections ( $\sim$  ODEs)

### 2. Dissipative

$$x = y \wedge z \xrightarrow{\sim} x \rightarrow 0$$

$$x = y \vee z \xrightarrow{\sim} x \rightarrow 1$$

## Theorem

Conservative  $\leftarrow$  reversible  
 $\leftarrow$  invertible

# Classification of BDEs

## Structural stability & bifurcations

### Theorem

BDEs with periodic solutions only are structurally stable, and conversely

*Remark.* They are dissipative.

Meta-theorems, by example.

The asymptotic behavior of

$$x(t) = x(t - \theta) \wedge \bar{x}(t - \tau)$$

is given by

$$x(t) = x(t - \theta)$$

Hence, if  $\tau < \theta = 1$  then solutions are asymptotically periodic;

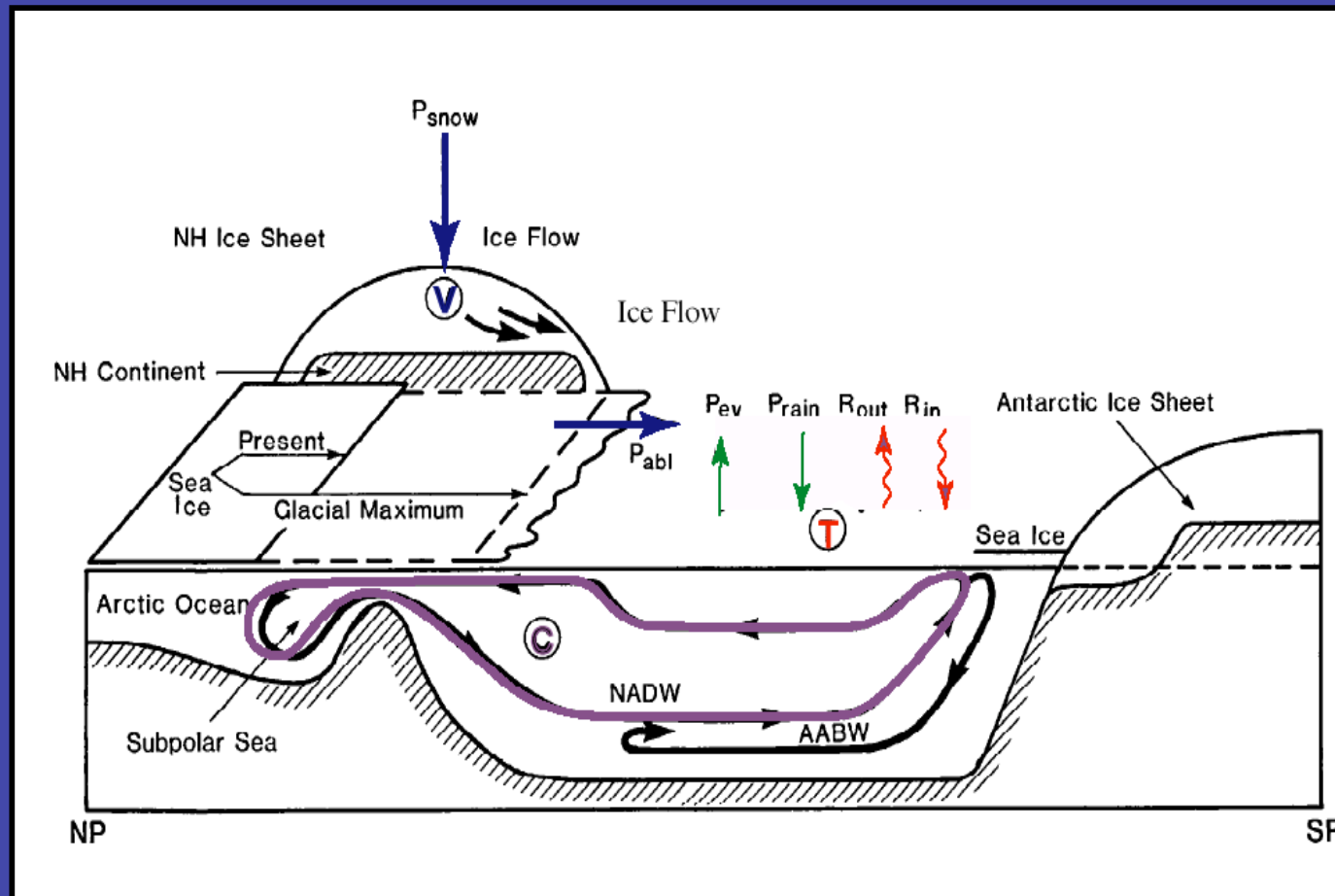
if however  $\theta < \tau = 1$  then solutions tend asymptotically to 0.

Therefore, as  $\theta$  passes through  $\tau$ , one has Hopf bifurcation.



# Paleoclimate application

Thermohaline circulation and glaciations

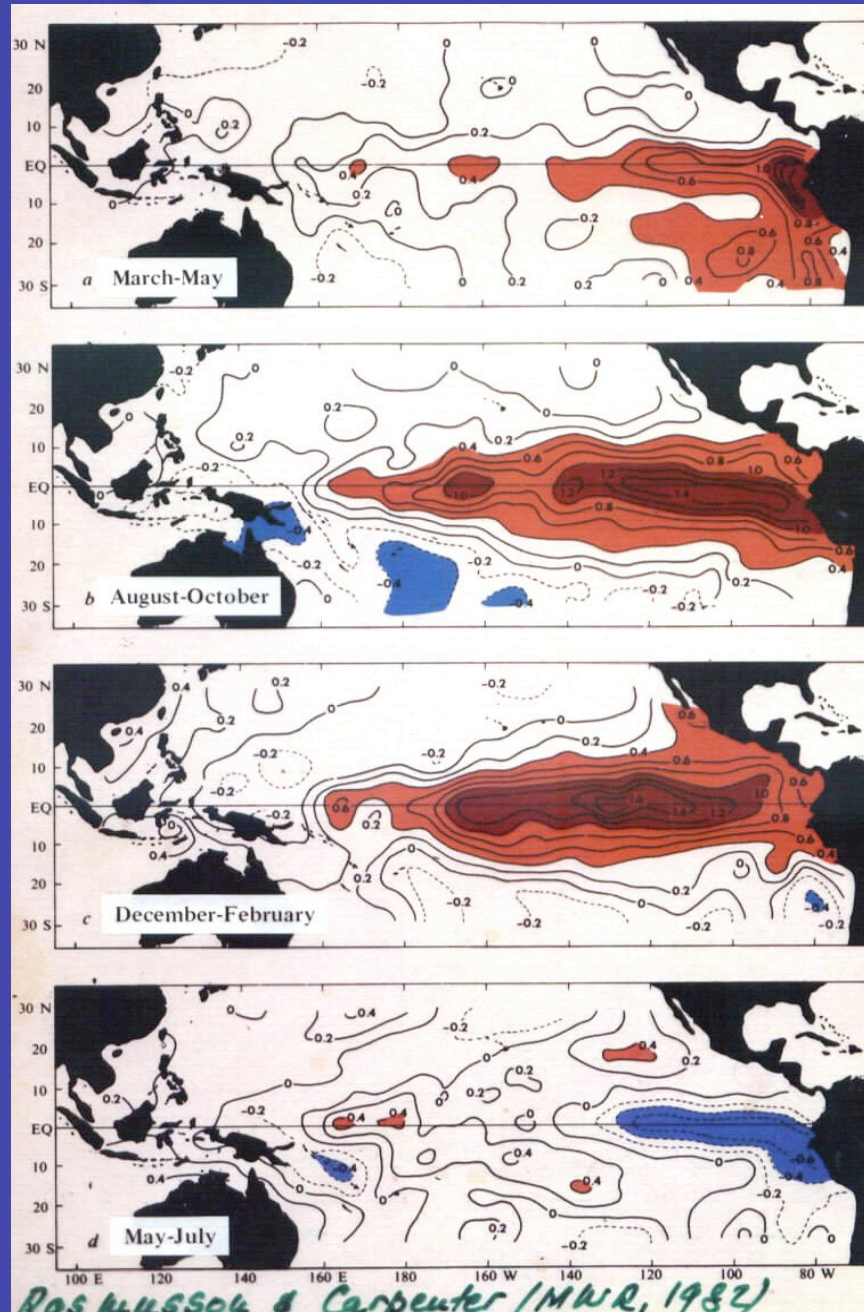


## Logical variables

- $T$  - global surface temperature;
- $V_N$  - NH ice volume,  $V_N = V$ ;
- $V_S$  - SH ice volume,  $V_S = 1$ ;
- $C$  - deep-water circulation index

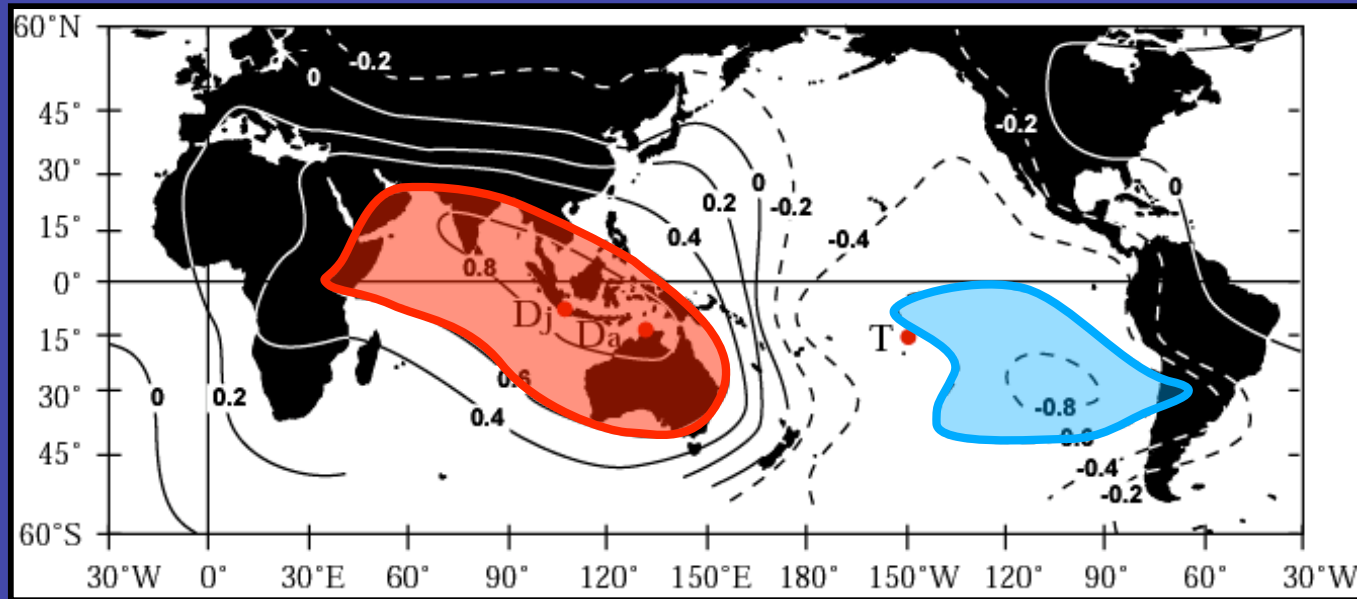
M. Ghil, A. Mullhaupt, & P. Pestiaux,  
*Climate Dyn.*, 2, 1-10, 1987.

# Spatio-temporal evolution of ENSO episode



# Scalar time series that capture ENSO variability

The large-scale Southern Oscillation (SO) pattern associated with El Niño (EN), as originally seen in surface pressures



Neelin (2006) *Climate Modeling and Climate Change*, after Berlage (1957)

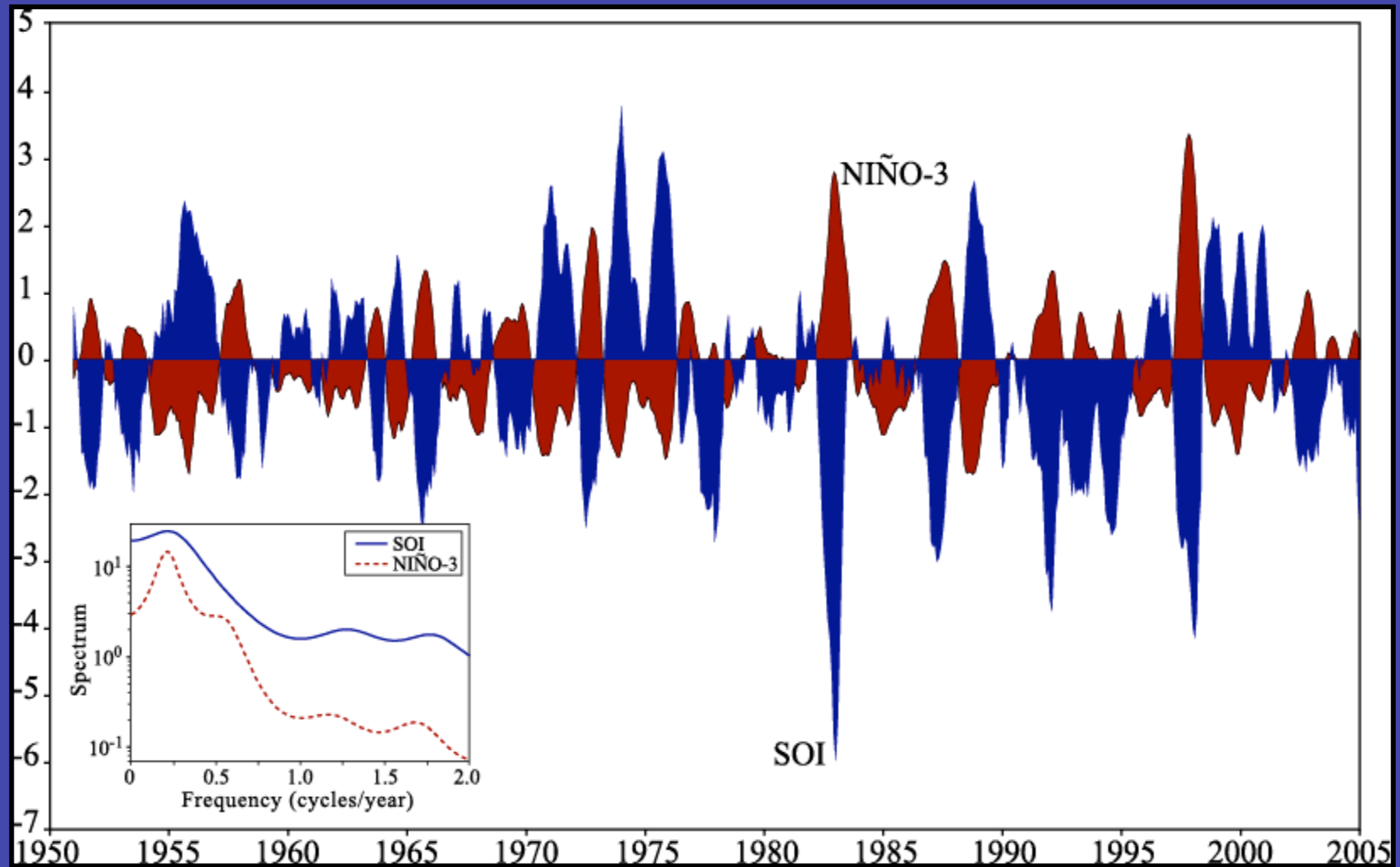
## **Southern Oscillation:**

The seesaw of sea-level pressures  $p_s$  between the two branches of the Walker circulation

Southern Oscillation Index (SOI) = normalized difference between  $p_s$  at Tahiti (T) and  $p_s$  at Darwin (Da)

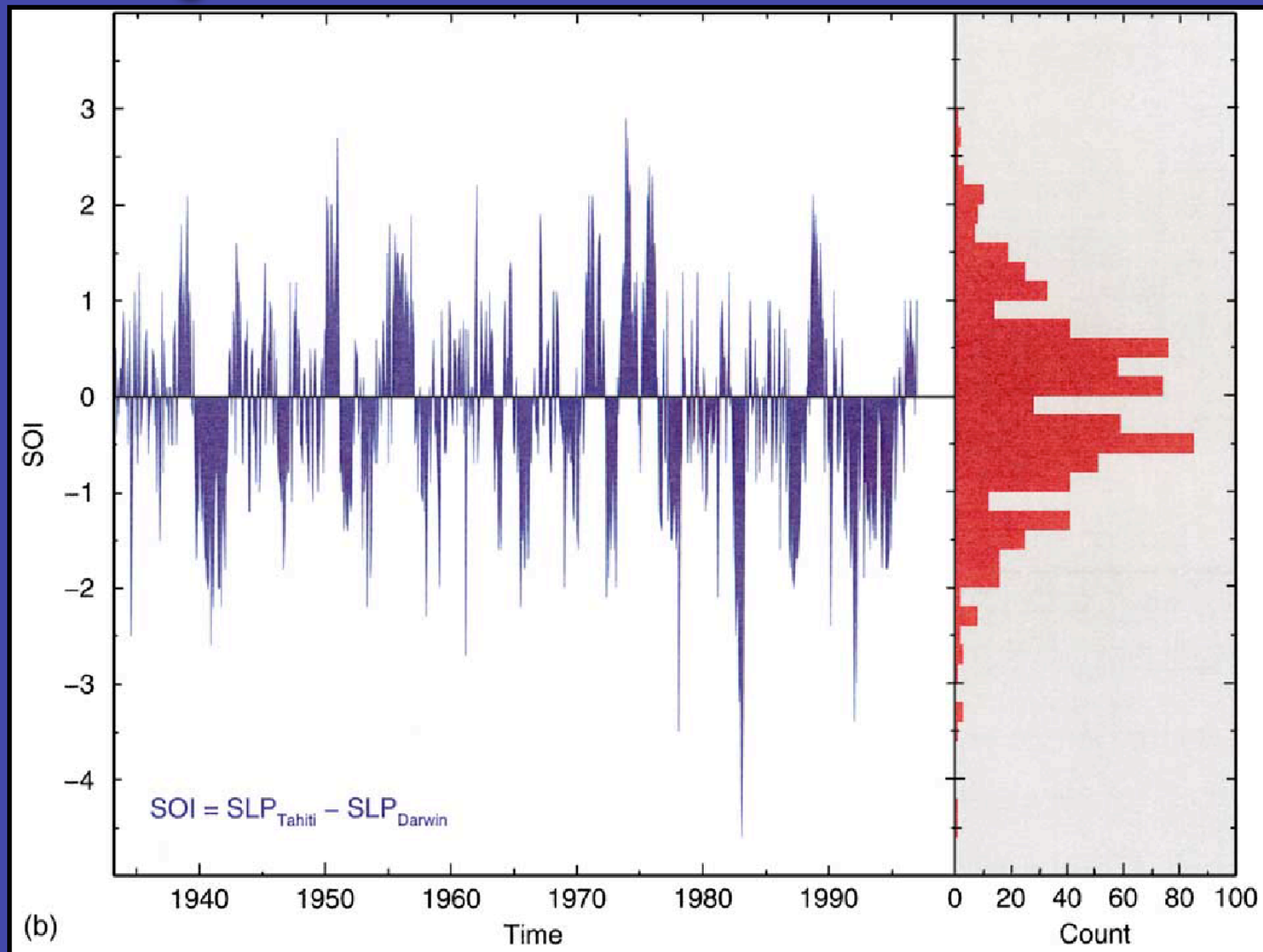
# Scalar time series that capture ENSO variability

Time series of *atmospheric pressure*  
and *sea surface temperature* (SST) indices



Data courtesy of NCEP's Climate Prediction Center  
Neelin (2006) *Climate Modeling and Climate Change*

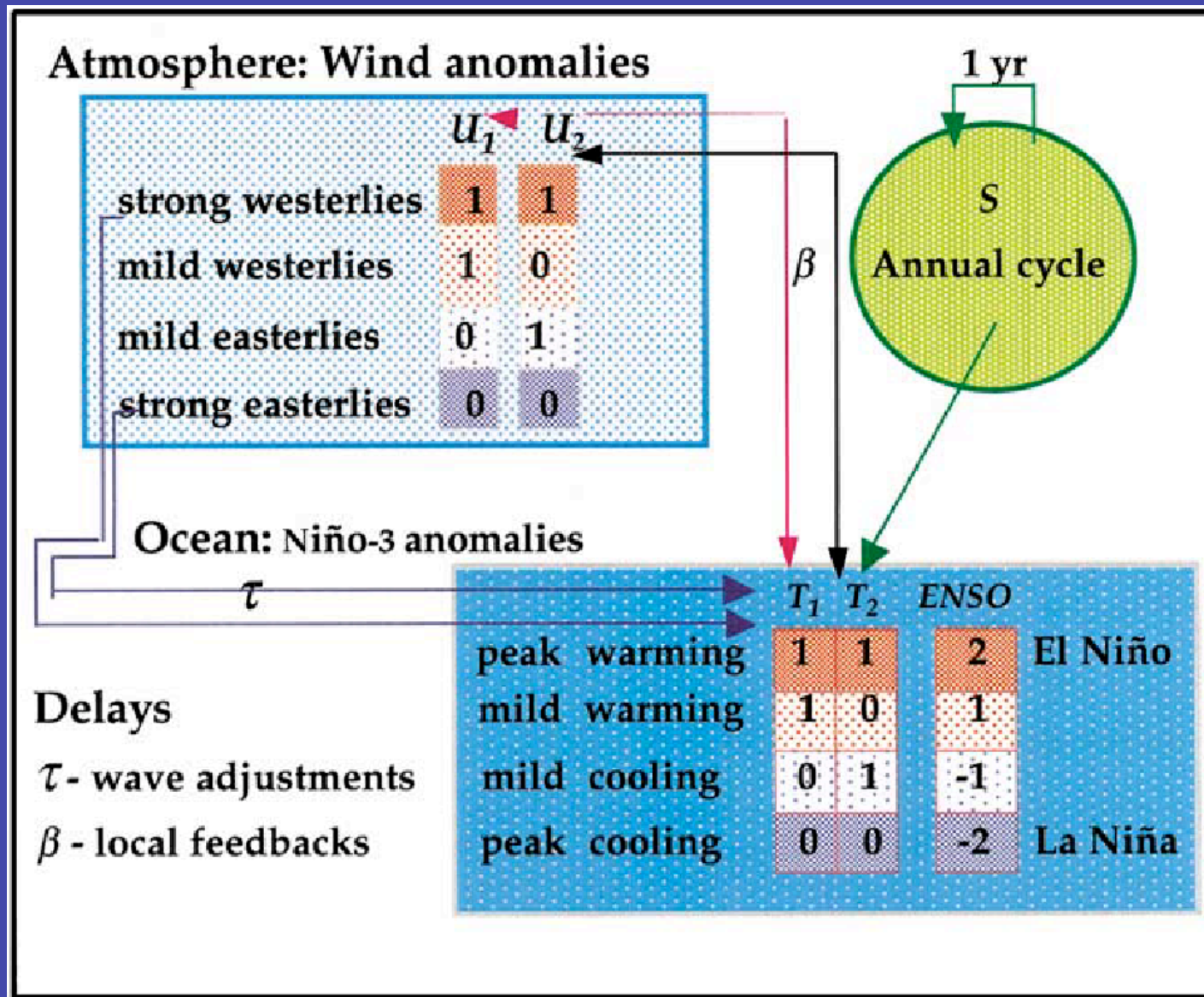
# Histogram of size distribution for ENSO events



(b)

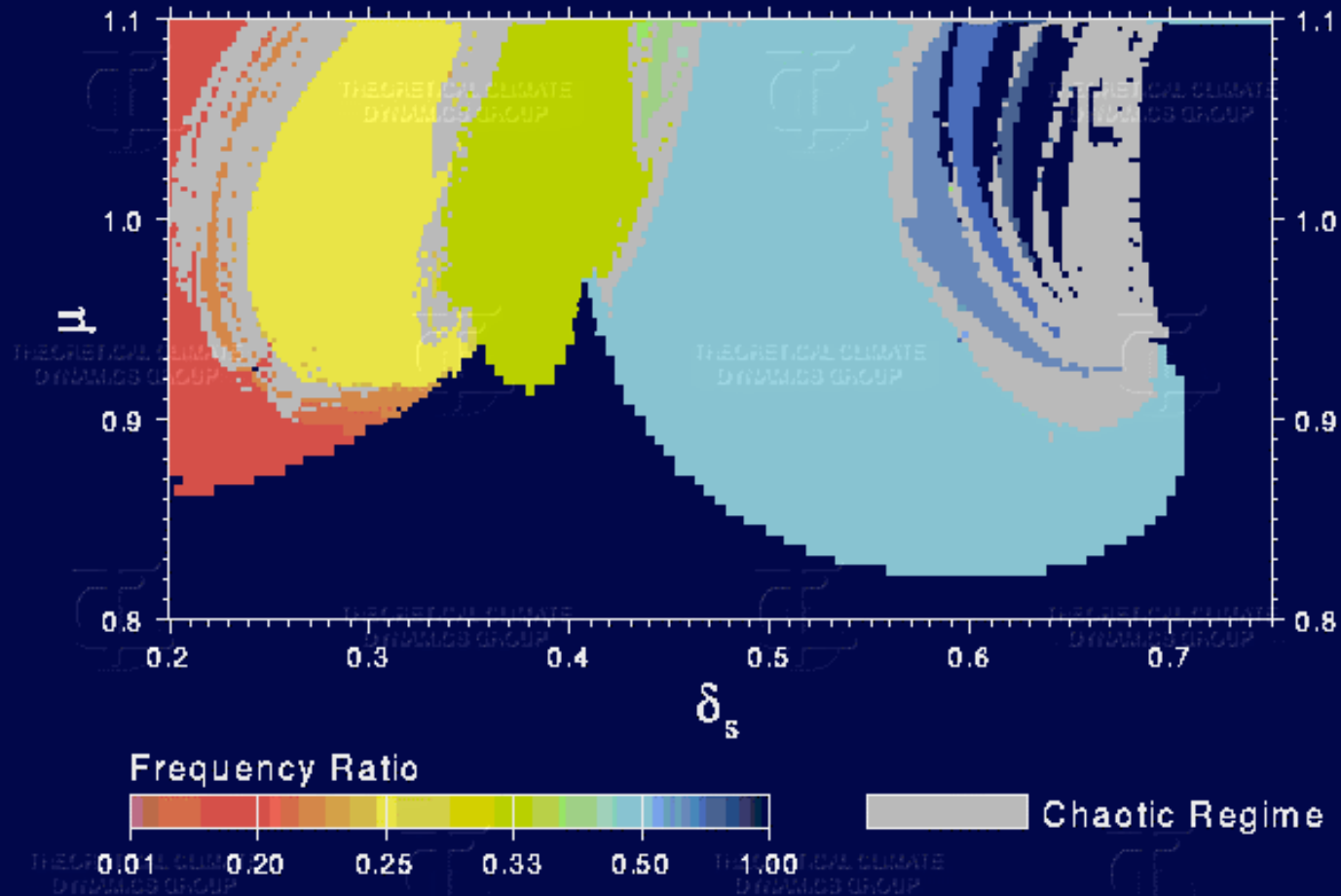
A. Saunders & M. Ghil, *Physica D*, 160, 54–78, 2001  
(courtesy of Pascal Yiou)

# BDE Model for ENSO: Formulation

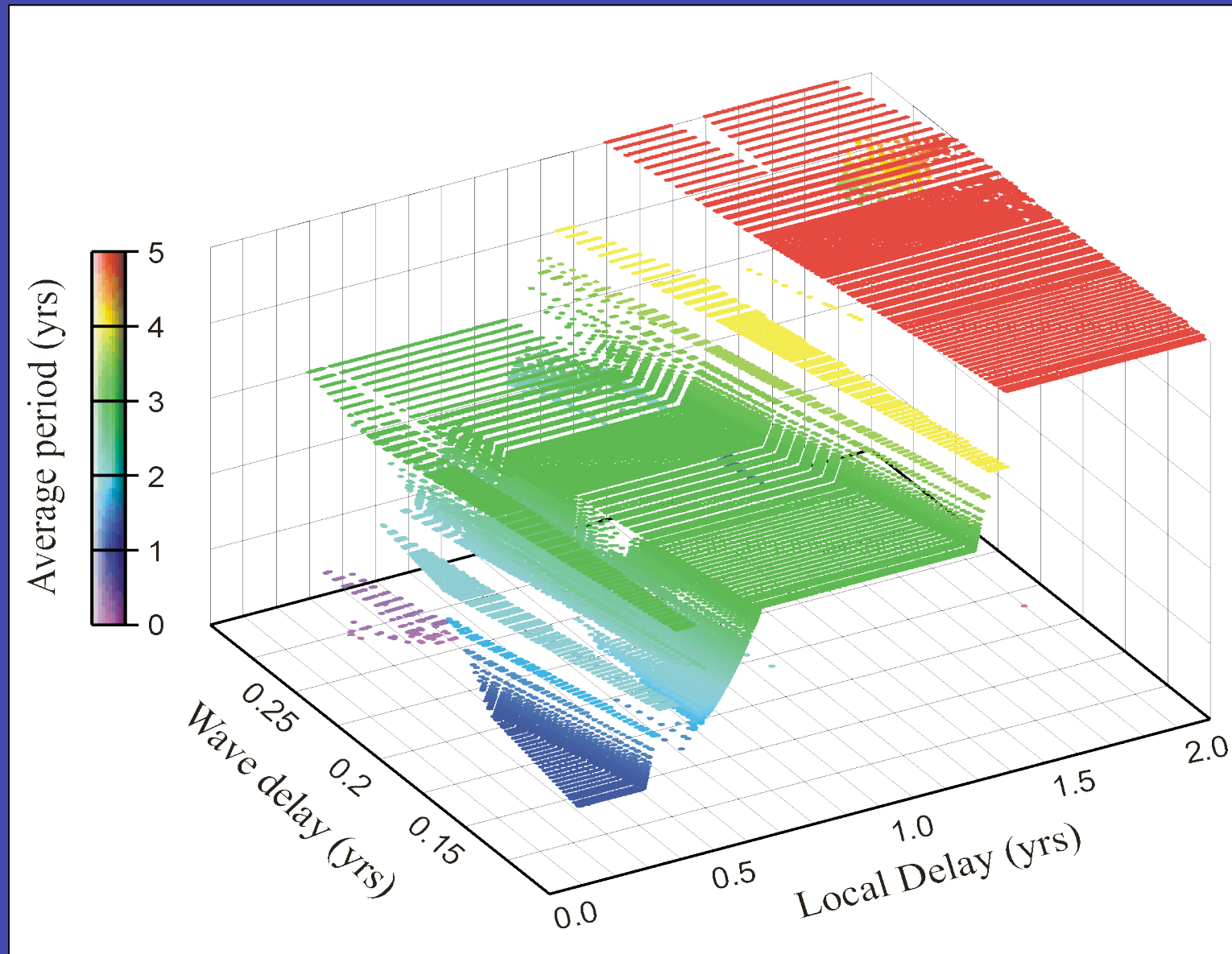


# Devil's Bleachers in a 1-D ENSO Model

Ratio of ENSO frequency to annual cycle

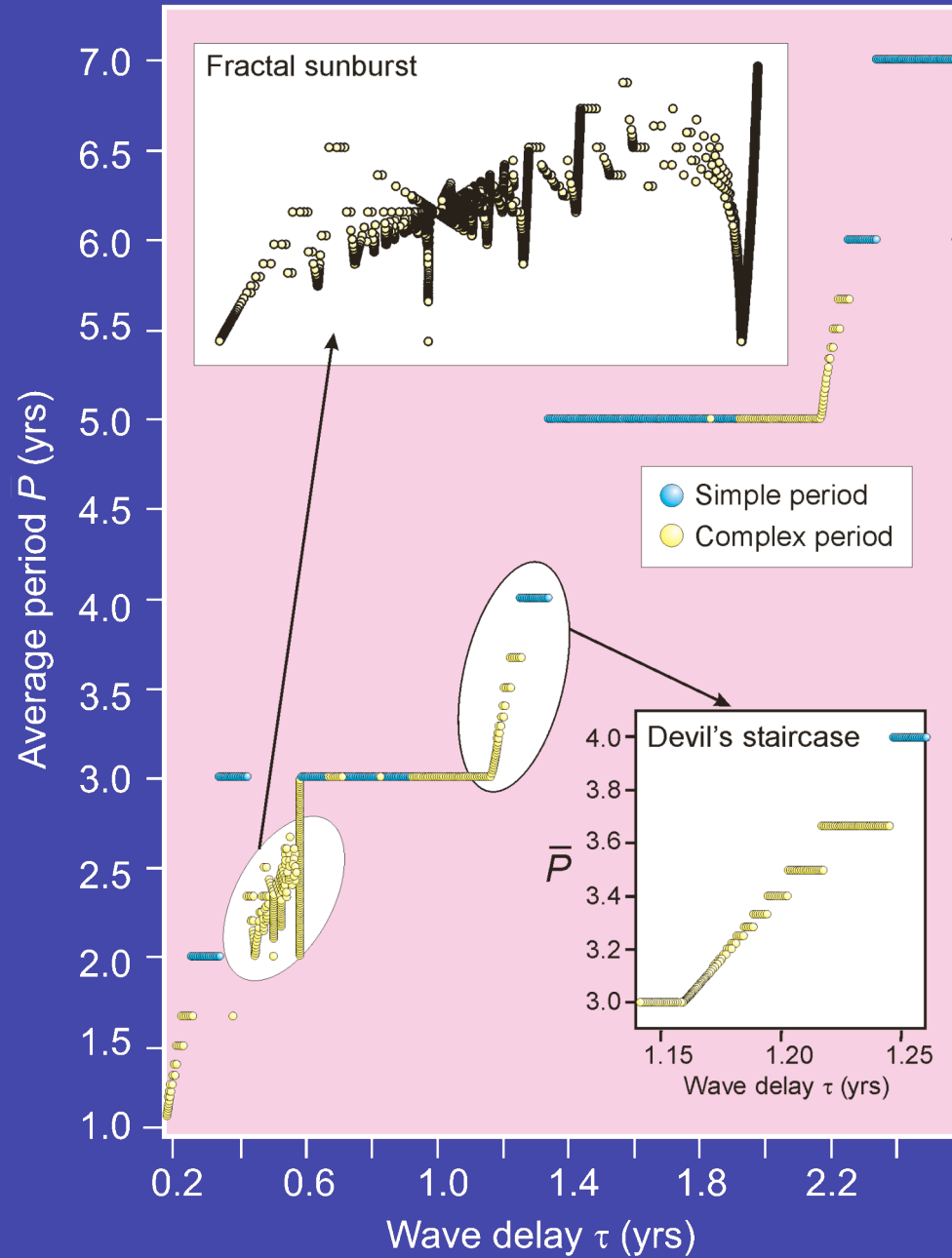


# Devil's Bleachers in the BDE Model of ENSO

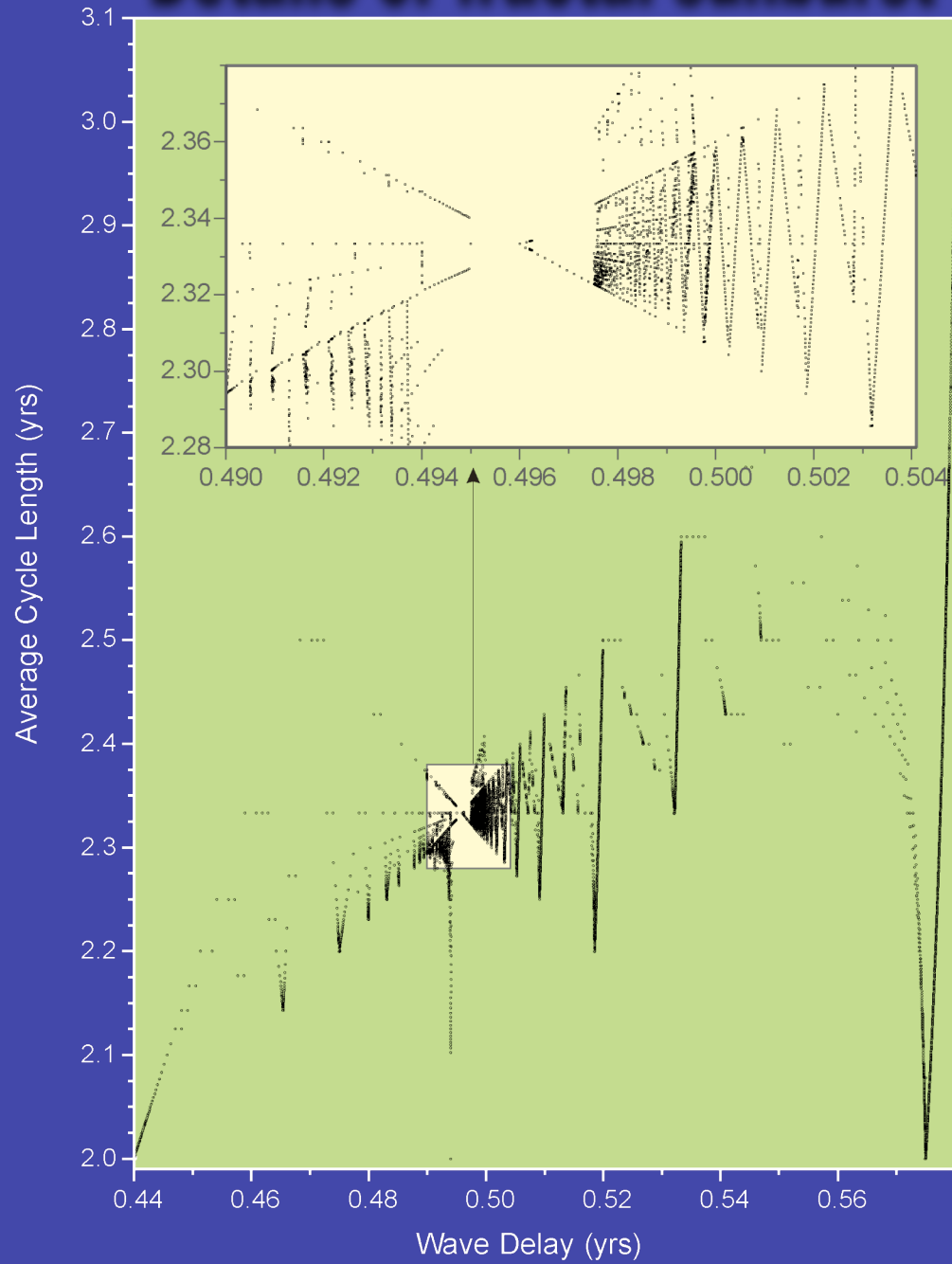




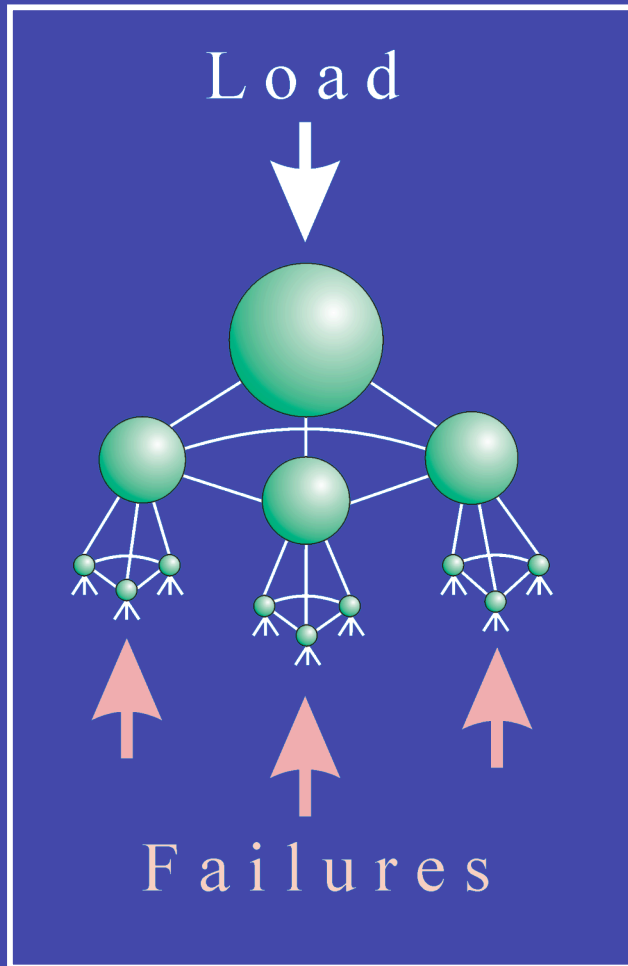
# Devil's staircase and fractal sunburst



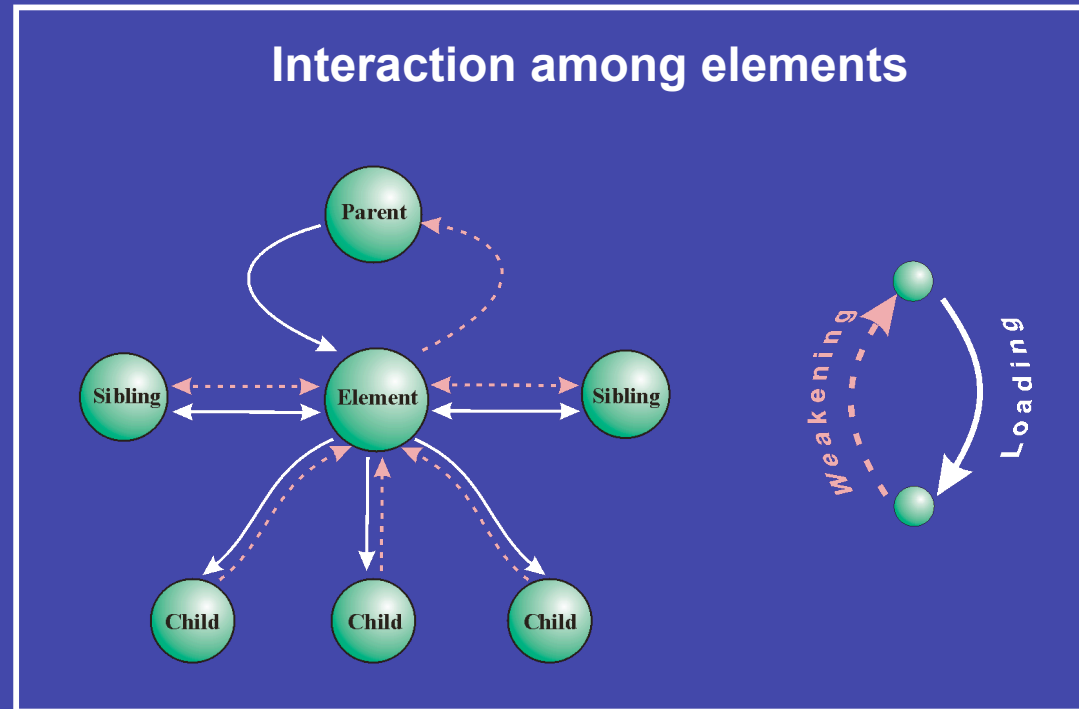
# Details of fractal sunburst



# Colliding-Cascade Model

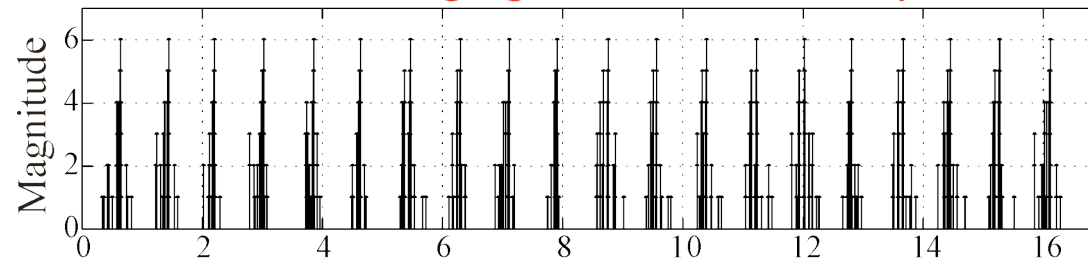


1. Hierarchical structure
2. Loading by external forces
3. Elements' ability to fail & heal

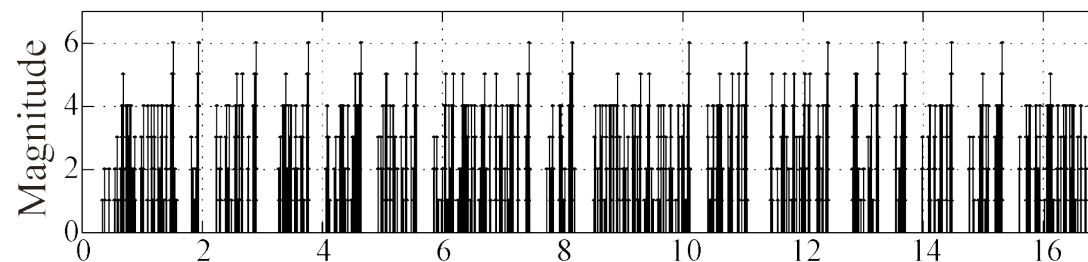


# BDE model of colliding cascades: Three seismic regimes

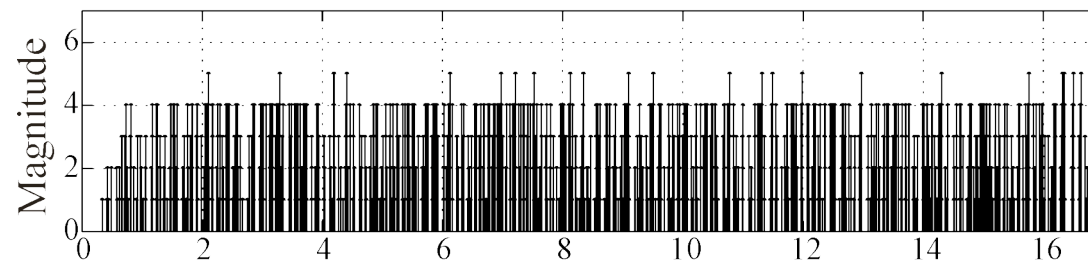
**H: High periodic seismicity**



**I: Intermittent seismicity**



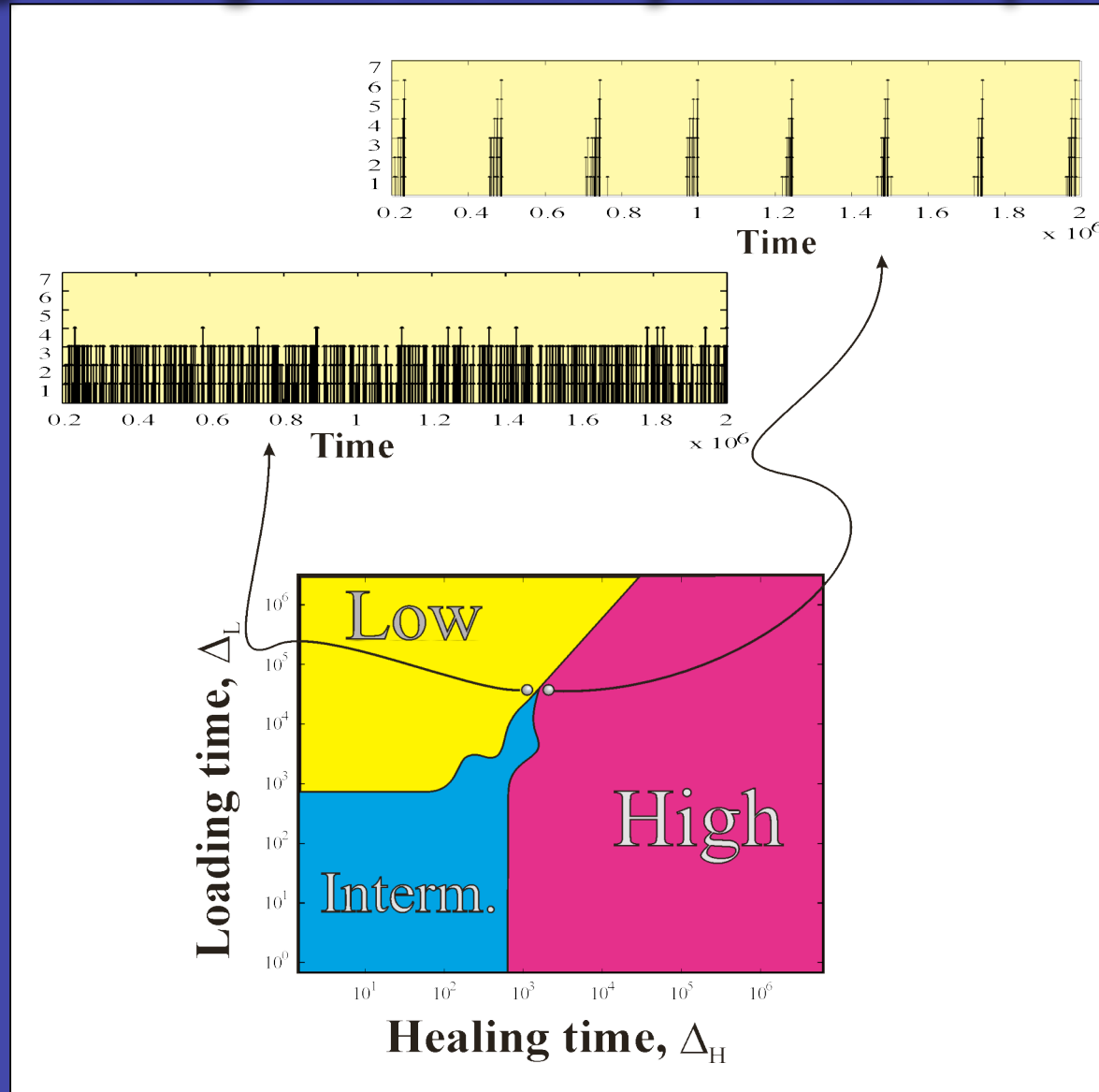
**L: Low seismicity**



Time

# BDE model of colliding cascades

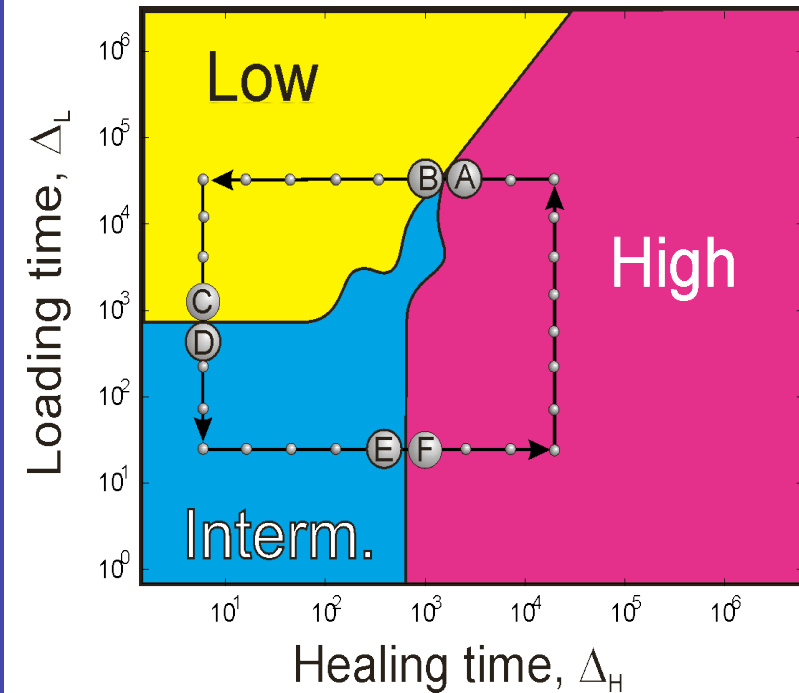
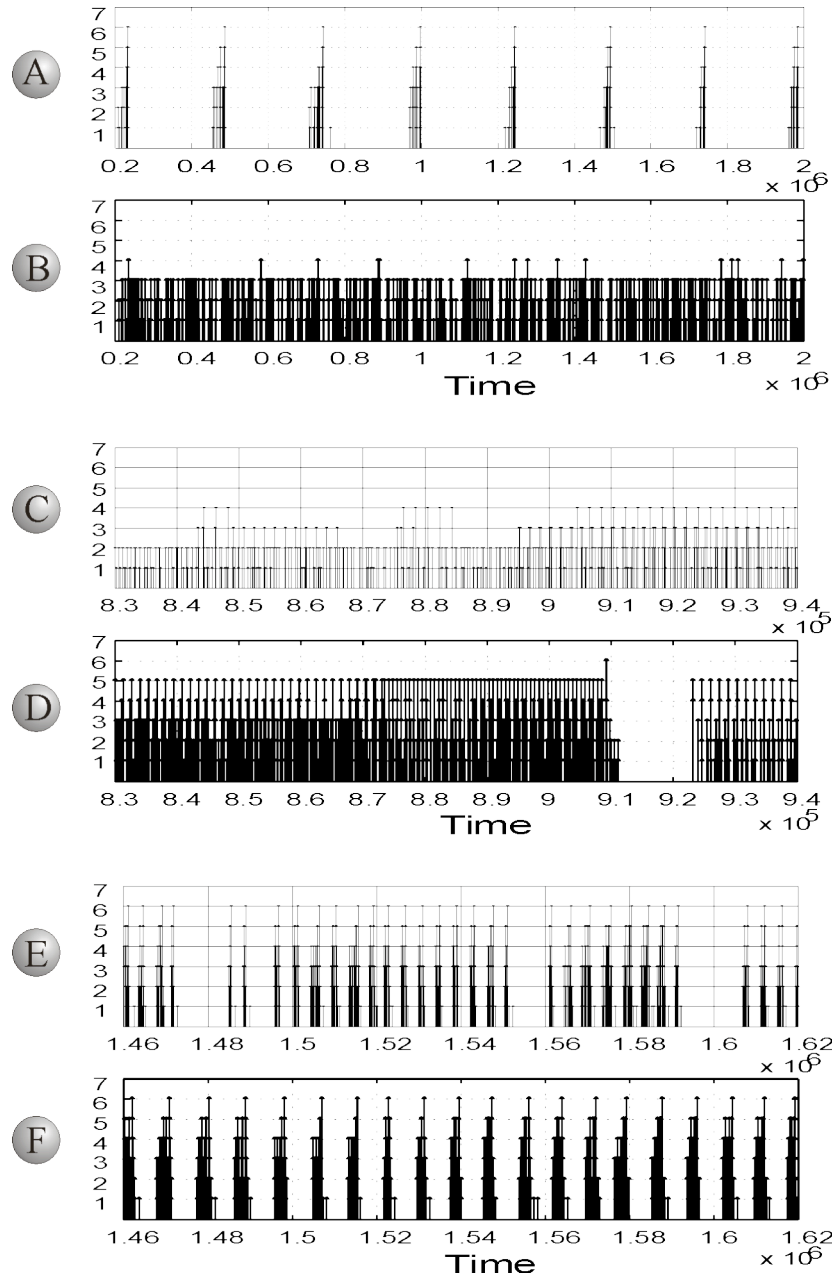
## Regime diagram: Instability near the triple point



# BDE model of colliding cascades

## Regime diagram:

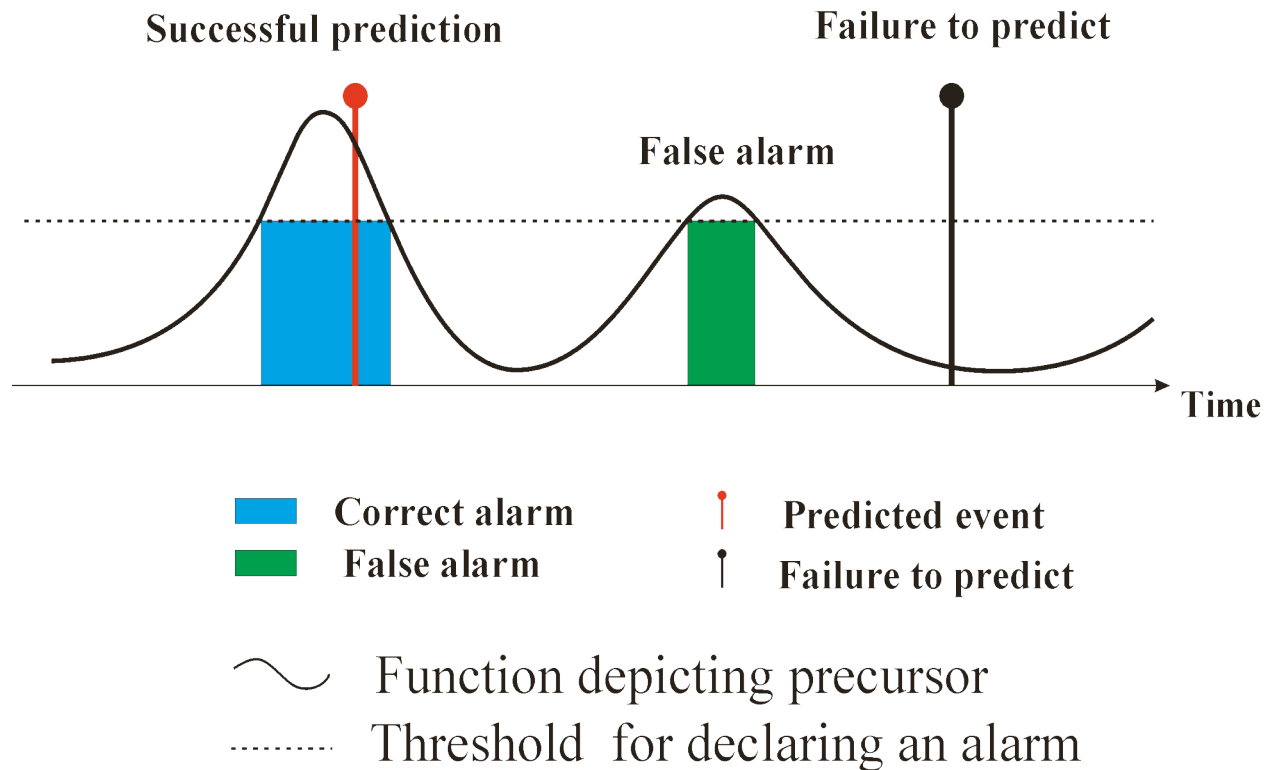
### Transition between regimes



I. Zaliapin, V. Keilis-Borok & M. Ghil  
(2003a, *J. Stat. Phys.*)

# Forecasting algorithms for natural and social systems: Can we beat statistics-based approach?

## Possible outcomes of prediction



Ghil and Robertson (2002, *PNAS*)  
Keilis-Borok (2002, *Annu. Rev. Earth Planet. Sci.*)

# Minimax prediction strategy

$P$  – set of parameters for precursor  $\Pi$   
(e.g. magnitude threshold, time window, *etc.*)

$\Pi_t(P)$  – Boolean alarm process

$\tau(P)$  – fractional time covered by alarms

$n(P)$  – fractional number of unpredicted target events

$f(P)$  – fractional number of false alarms

## Minimax prediction strategy 1:

$$P = \arg \min [f(P)]$$

$$A_{\text{collective}} = \Pi_1 \vee \Pi_2 \vee \dots \vee \Pi_n$$

## Minimax prediction strategy 2:

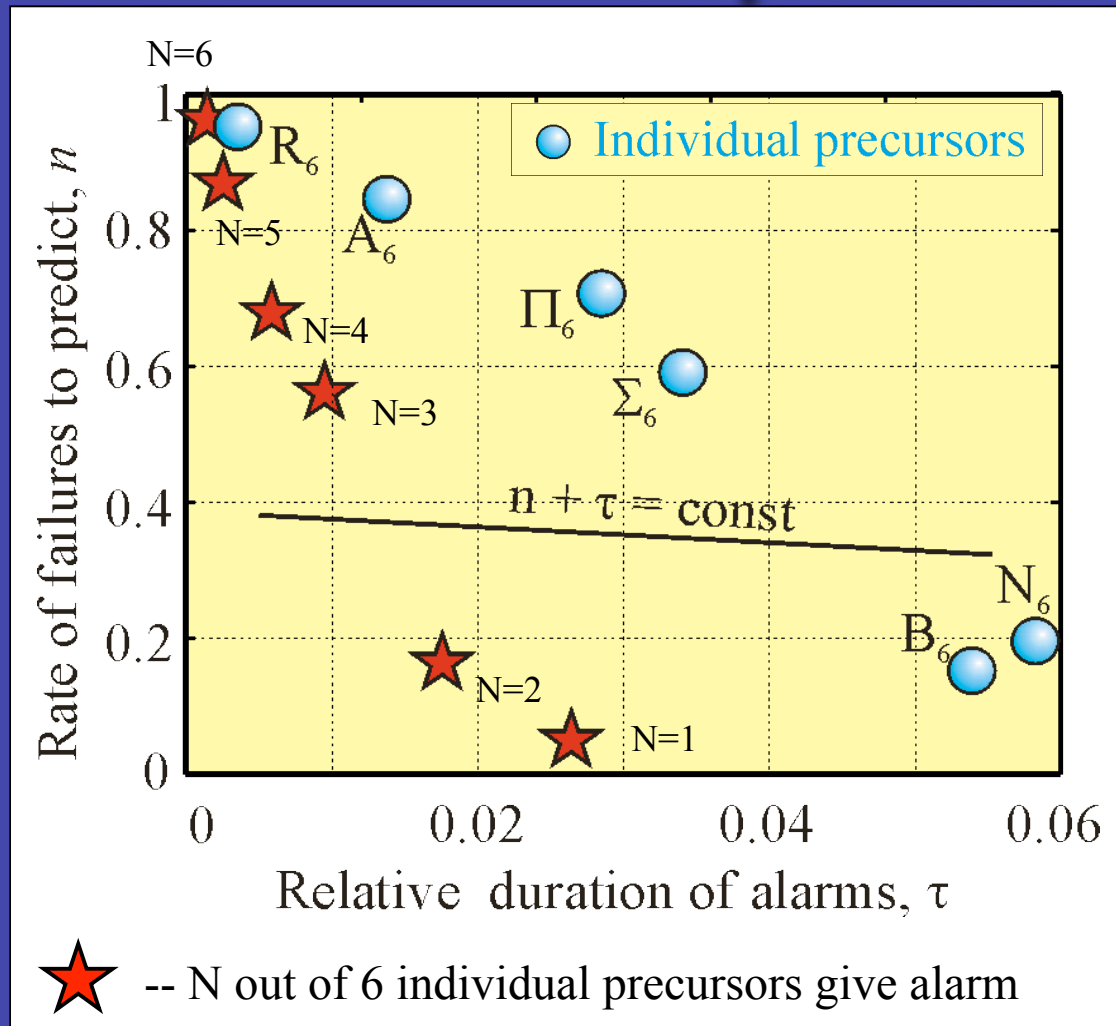
$$P = \arg \min [n(P)]$$

$$A_{\text{collective}} = \Pi_1 \wedge \Pi_2 \wedge \dots \wedge \Pi_n$$



# BDE model

# Minimax prediction strategy 1



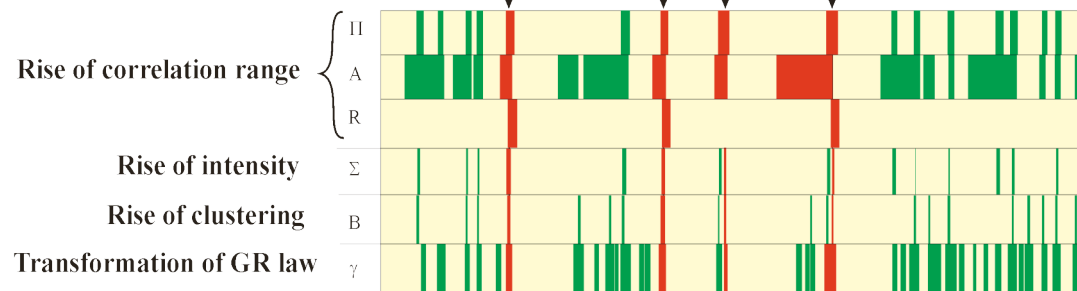
Individual patterns are tuned to eliminate false alarms at the cost of having more failures to predict. Collectively, errors of both kinds are drastically reduced.

# BDE model

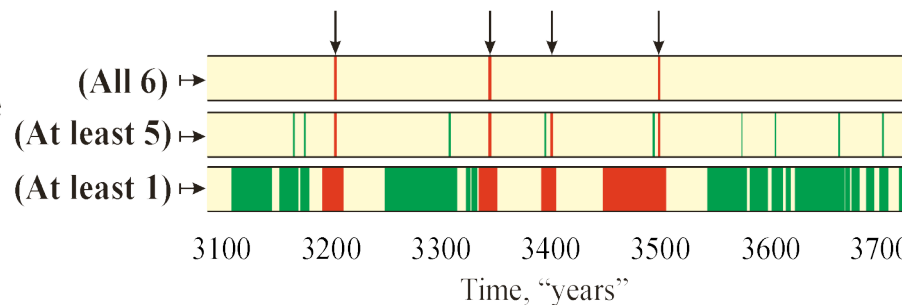
# Minimax prediction strategy 2

Minimax prediction strategy in BDE model:  
voting of individual premonitory patterns.

## Individual alarms



## Collective alarms



False alarm Correct alarm Large earthquake

Individual patterns are tuned to eliminate failures to predict  
at the cost of having more false alarms.  
Collectively, errors of both kinds are drastically reduced.

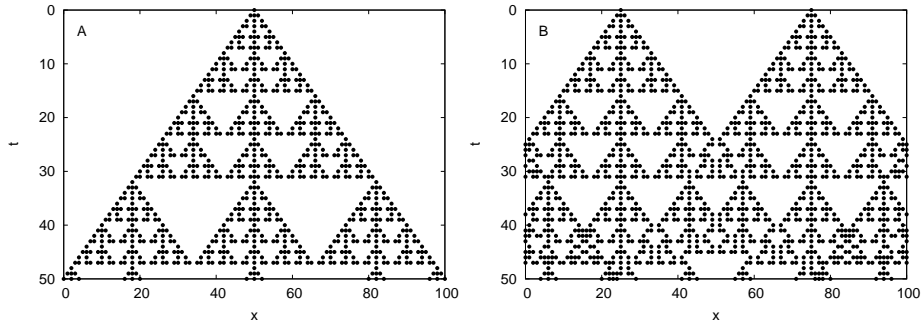


Fig. 13. Solutions of the “partial BDE” (34): (a) for a single nonzero site at  $t = 0$ ; and (b) the collision of two “waves,” each originating from such a site. For the space and time steps  $\theta_t = \theta_z = 1$ , this BDE is equivalent to the elementary cellular automaton (ECA) with rule 150; empty sites ( $u_i(j) = 0$ ) in white and occupied sites ( $u_i(j) = 1$ ) in black.

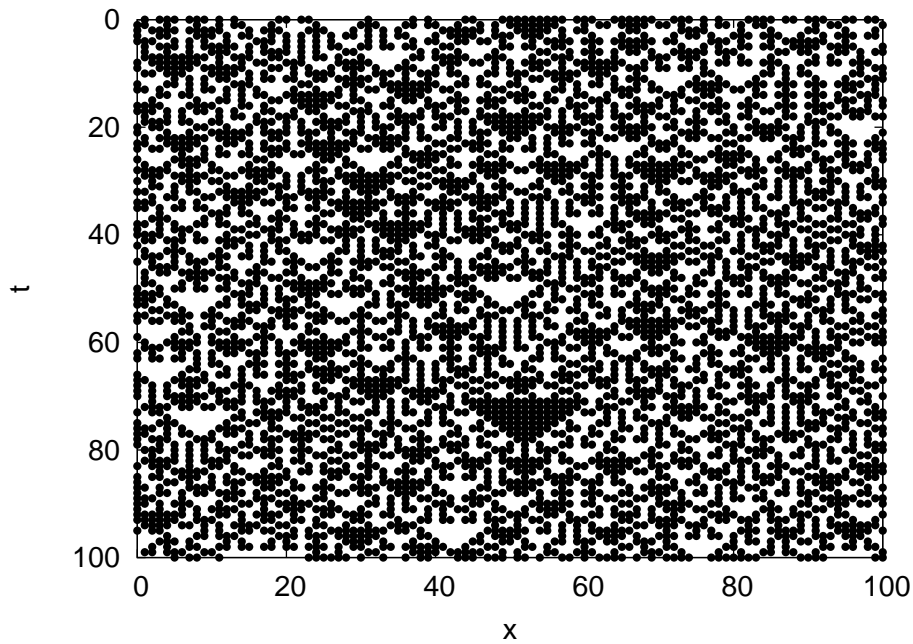


Fig. 14. The solution of the BDE (34) starting from a random initial state of length  $N = 100$ . The qualitative behavior is characterized by “triangles” of empty or occupied sites but without any recurrent pattern; this behavior does not depend on the particular random initial state.

**Table 2.** Results on partial BDEs

PDE $\partial_t v =$	Approx.	PBDE $u_i(t + \theta_t) =$	ECA rule	ECA class	Behavior
$\partial_z v$	$\nabla$ (I order)	$u_{i-1}(t)$	170	—	conservative
$\partial_{zz} v$	$\vee$	$[u_{i-1}(t) \vee u_{i+1}(t)] \nabla u_i(t)$	54	I	dissipative
$\partial_{zz} v$	$\wedge$	$[u_{i-1}(t) \wedge u_{i+1}(t)] \nabla u_i(t)$	108	II	dissipative
$\partial_{zz} v$ $\partial_z v$	$\nabla$ (I order) $\nabla$ (II order)	$[u_{i-1}(t) \nabla u_{i+1}(t)] \nabla u_i(t)$	150	III	dissipative

Summary of results on the partial BDEs obtained from the different considered approximations for the spatial derivative in the parabolic and hyperbolic PDEs. Temporal derivative is always approximated by the  $\nabla$  operator. From top to bottom, these are the equations (29), (35), (36) and (33) discussed in the text, respectively. Notice that, though all but the first considered PBDEs are dissipative, only the last one, *i.e.* Eq. (33), displays chaotic behavior in the limit of infinite lattice size.

# Short BDE bibliography

## Theory

Dee & Ghil (1984, *SIAM J. Appl. Math.*)

Ghil & Mullhaupt (1985, *J. Stat. Phys.*)

## Applications to climate

Ghil et al. (1987, *Climate Dyn.*)

Mysak et al. (1990, *Climate Dyn.*),

Darby & Mysak (1993, *Climate Dyn.*),

Saunders & Ghil (2001, *Physica D*)

## Applications to solid-earth problems

Zaliapin, Keilis-Borok & Ghil (2003a, b, *J. Stat. Phys.*)

## Applications to genetics

Oktem, Pearson & Egiazarian (2003, *Chaos*)

Gagneur & Casari (2005, *FEBS Letters*)

## Applications to the socio-economic and computer sciences?

## Review paper

Ghil & Zaliapin (2005) A novel fractal way: Boolean delay equations and their applications to the Geosciences,  
*Invited for book honoring B.Mandelbrot 80<sup>th</sup> birthday*

# Concluding remarks

1. BDEs have **rich behavior**:  
periodic, quasi-periodic, aperiodic, increasing complexity
2. BDEs are relatively **easy to study**
3. BDEs are natural in a **digital world**
4. Two types of **applications**
  - strictly discrete (genes, computers)
  - saturated, threshold behavior (nonlinear oscillations, climate dynamics, population biology, earthquakes)
5. Can **provide insight** on a very qualitative level  
(~ symbolic dynamics)
6. **Generalizations** possible  
(spatial dependence – “partial” BDEs;  
stochastic delays &/or connectives)

# Conclusions

Hmmm, this is interesting!



But what does it all mean?

**Needs more work!!!**

