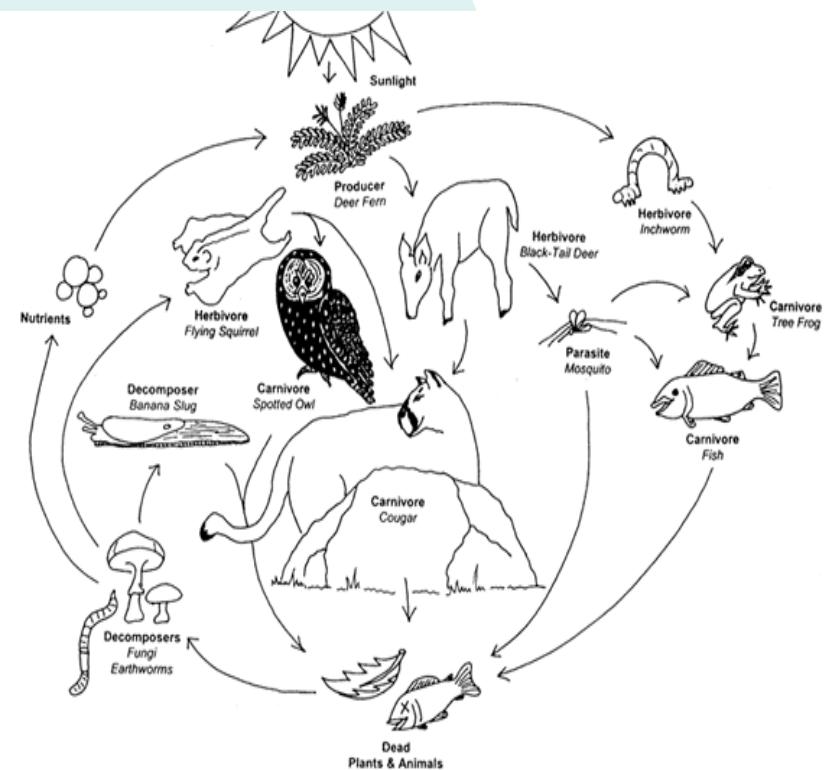


Simple Models of Complex Chaotic Systems

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Presented at the

AAPT Topical Conference on
Computational Physics in Upper
Level Courses

At Davidson College (NC)

On July 28, 2007

Collaborators

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California - Davis



q **Konstantinos
Chlouverakis**, Univ
Athens (Greece)



Background

- q Grew out of an multi-disciplinary **chaos course** that I taught 3 times
- q Demands computation
- q Strongly motivates students
- q Used now for physics undergraduate research projects (~20 over the past 10 years)

Minimal Chaotic Systems

❑ 1-D map (quadratic map)

$$x_{n+1} = 1 - x_n^2$$

❑ Dissipative map (Hénon)

$$x_{n+1} = 1 - ax_n^2 + bx_{n-1}$$

❑ Autonomous ODE (jerk equation)

$$\ddot{x} + a\ddot{x} - \dot{x}^2 + x = 0$$

❑ Driven ODE (Ueda oscillator)

$$\ddot{x} + x^3 = \sin \omega t$$

❑ Delay differential equation (DDE)

$$\dot{x} = x_{t-\tau} - x_{t-\tau}^3$$

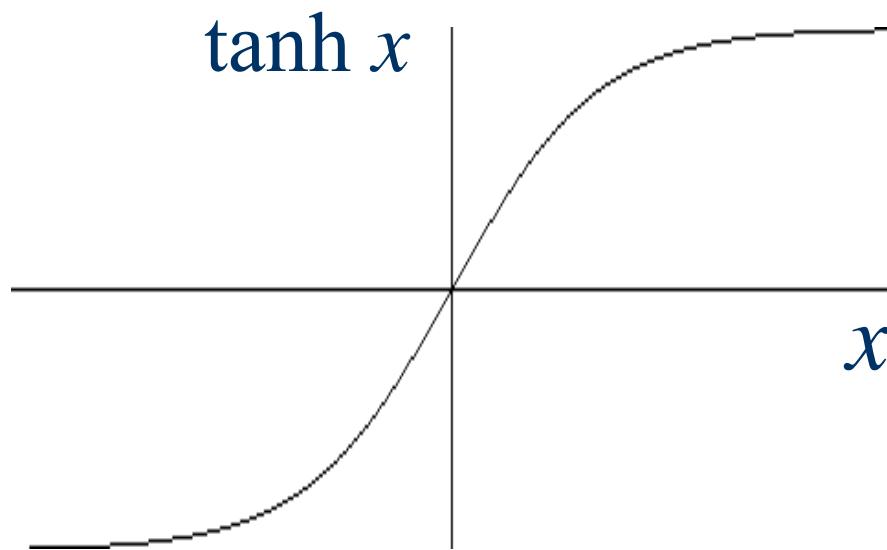
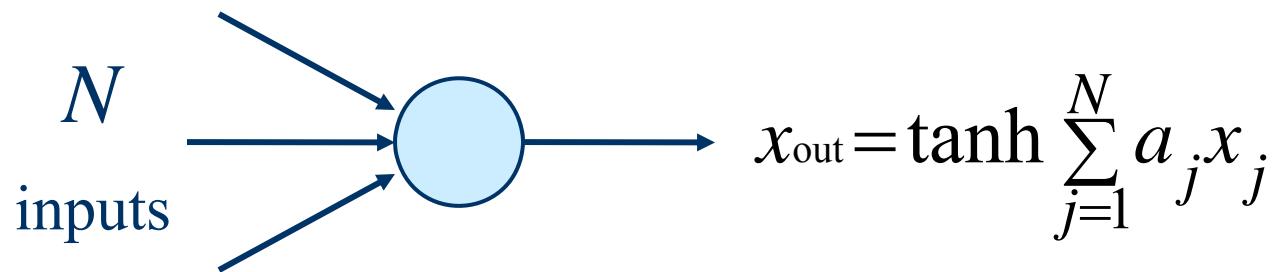
❑ Partial diff eqn (Kuramoto-Sivashinsky)

$$\partial_t u + u \partial_x u + a \partial_x^2 u + \partial_x^4 u = 0$$

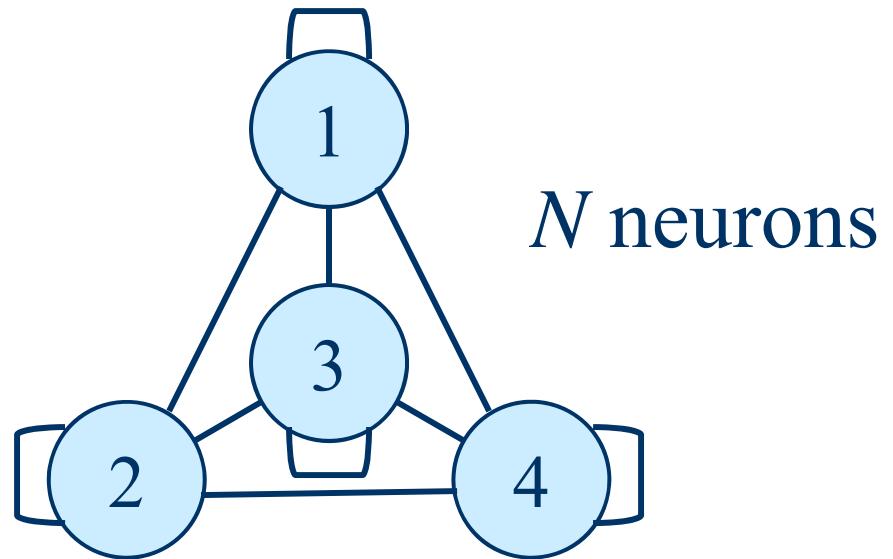
What is a complex system?

- q Complex ≠ complicated
- q Not real and imaginary parts
- q Not very well defined
- q Contains many interacting parts
- q Interactions are nonlinear
- q Contains feedback loops (+ and -)
- q Cause and effect intermingled
- q Driven out of equilibrium
- q Evolves in time (not static)
- q Usually chaotic (perhaps weakly)
- q Can self-organize and adapt

A Physicist's Neuron



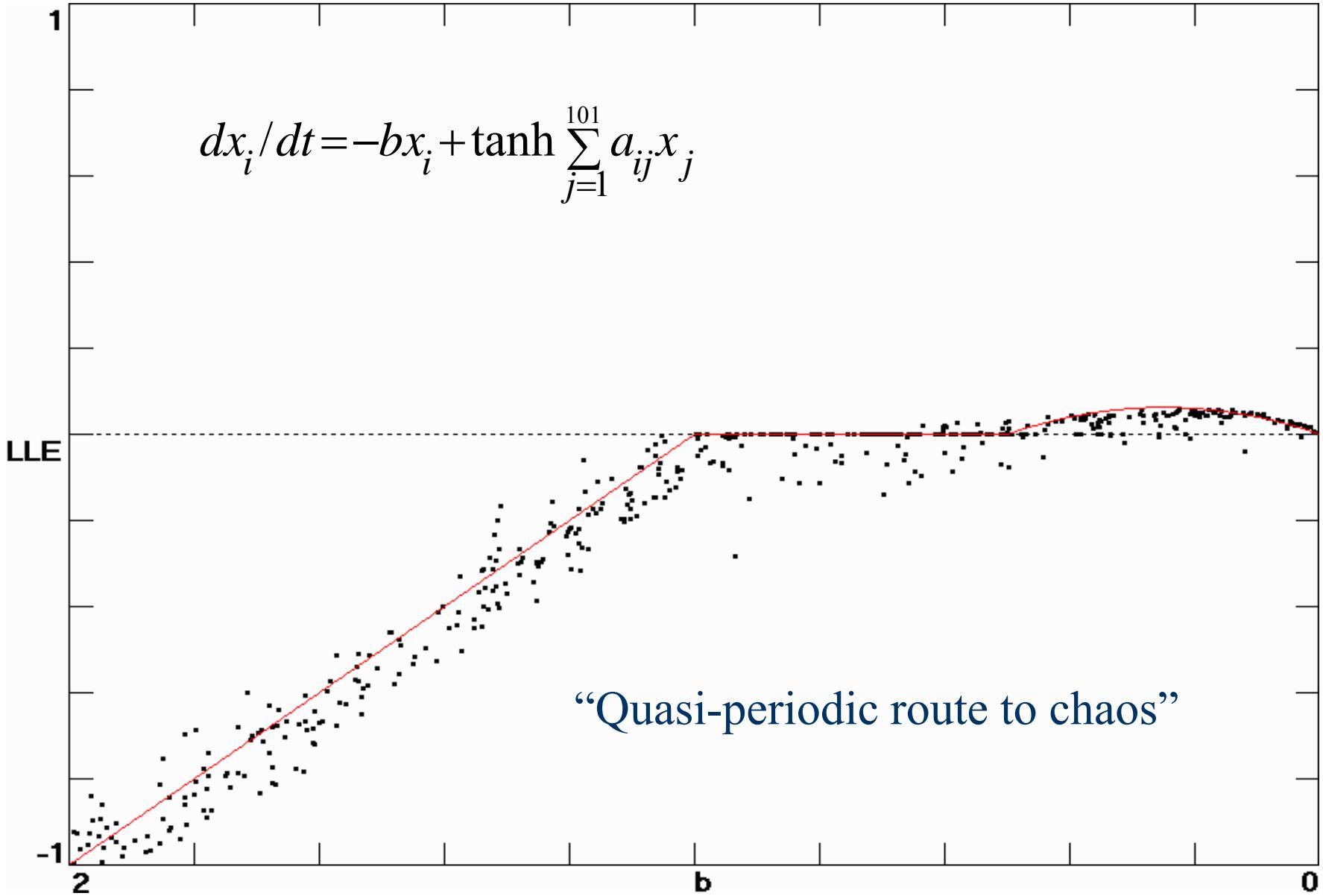
A General Model (artificial neural network)

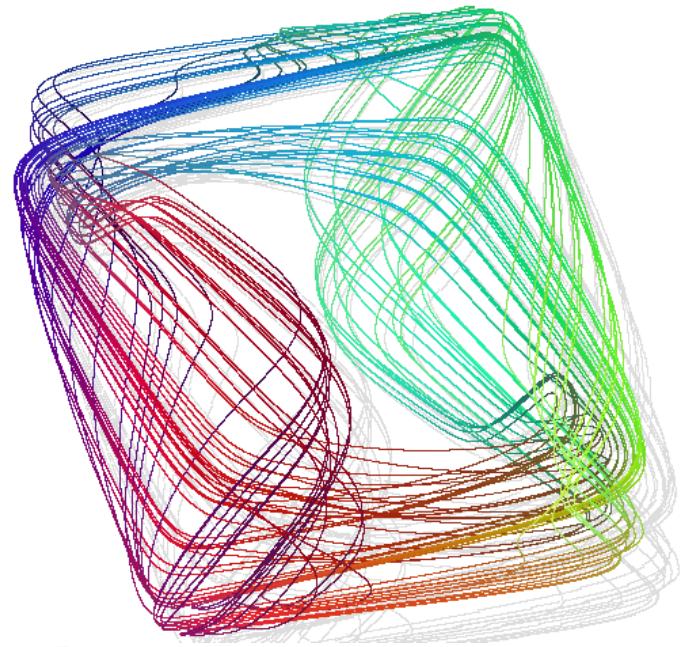
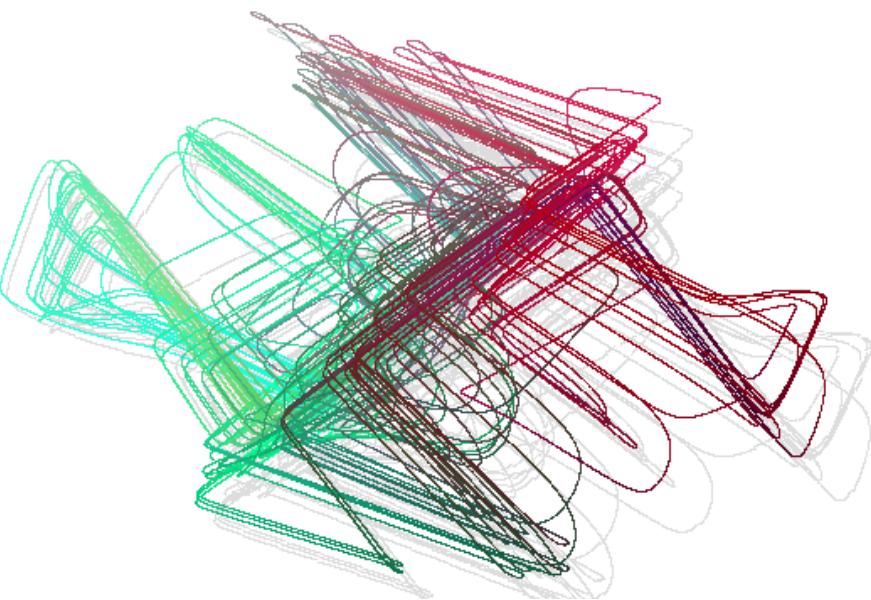


$$\dot{x}_i = -b_i x_i + \tanh \sum_{\substack{j=1 \\ j \neq i}}^N a_{ij} x_j$$

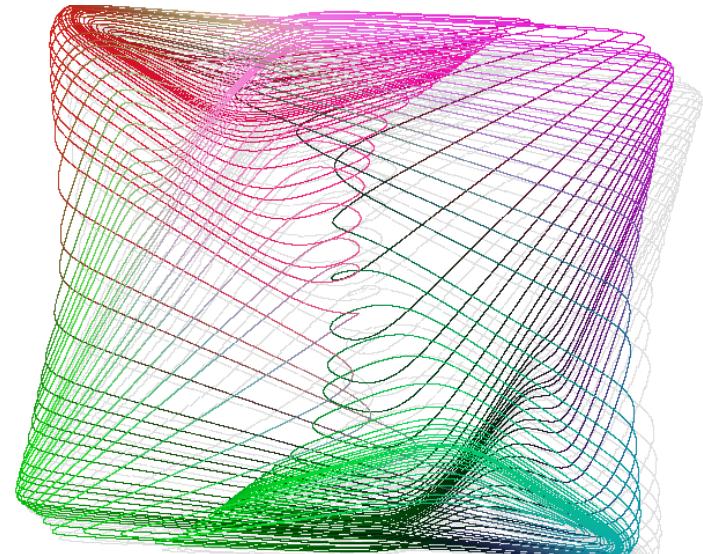
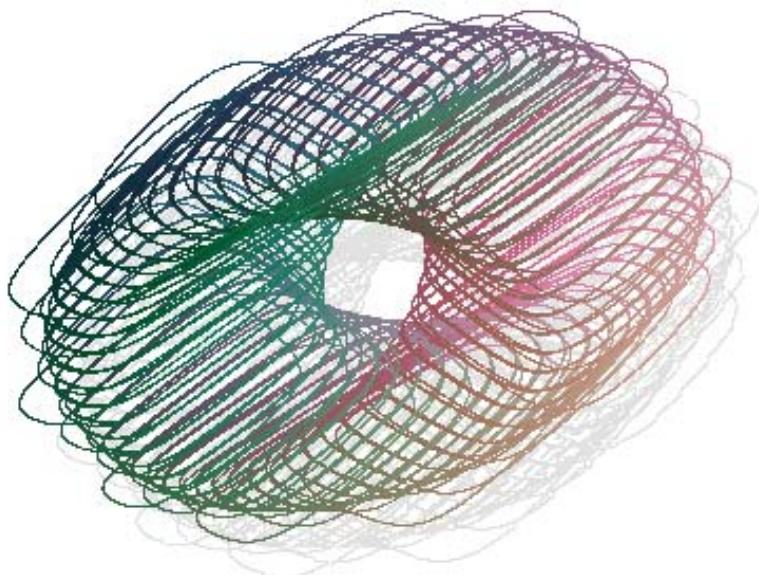
“Universal approximator,” $N \rightarrow \infty$

Route to Chaos at Large N (=101)

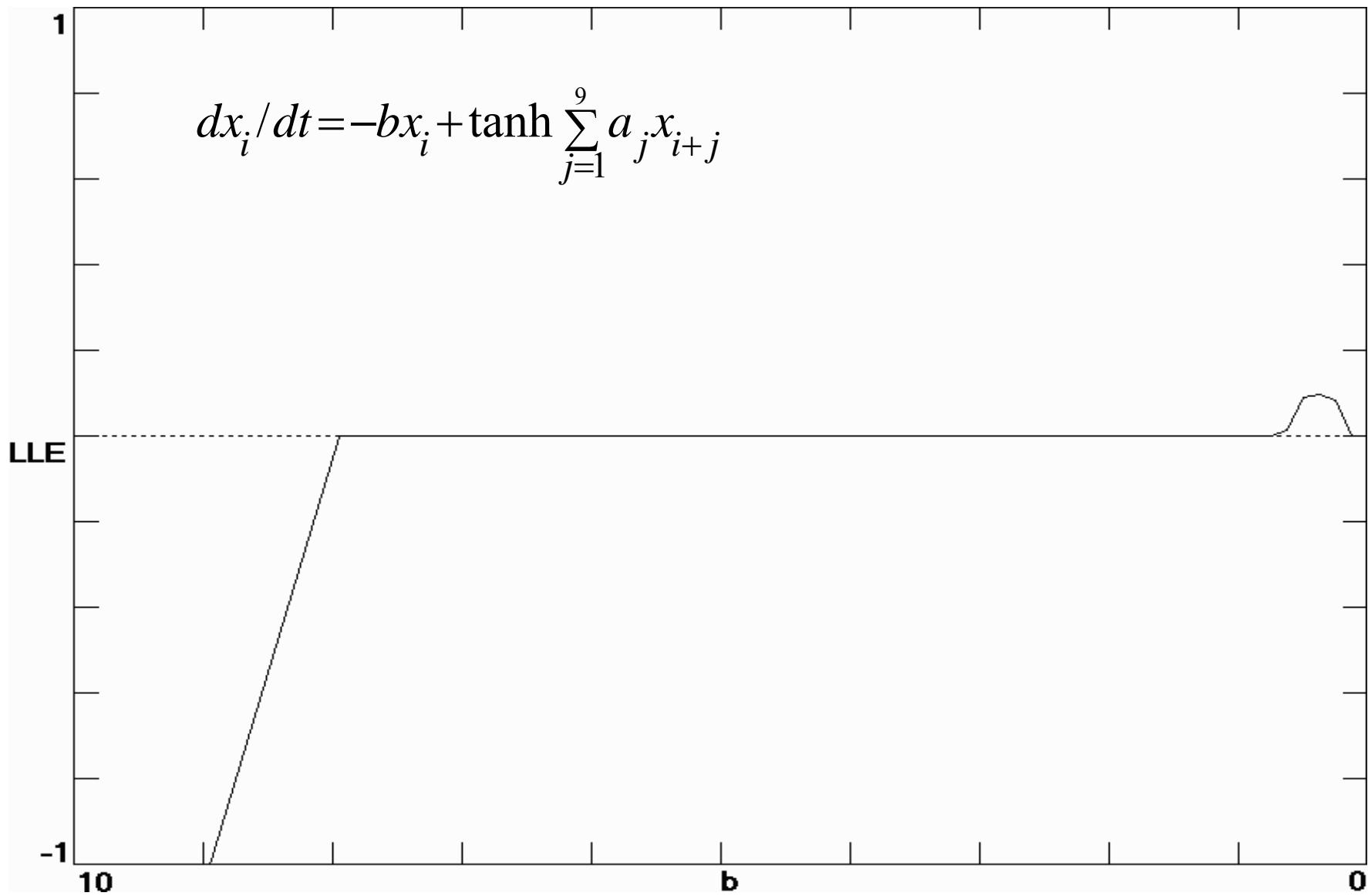




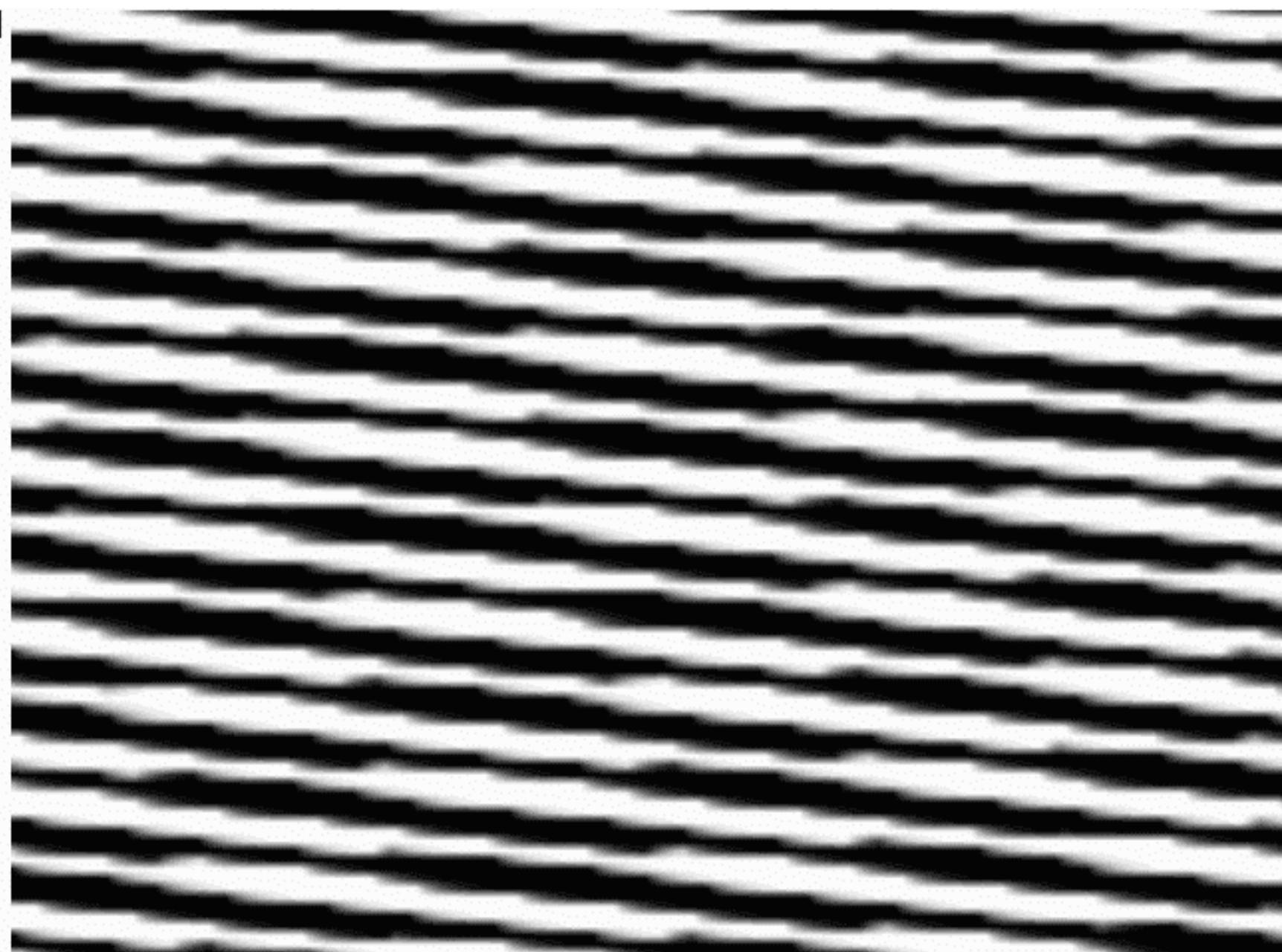
Strange Attractors



Sparse Circulant Network ($N=101$)



101



i

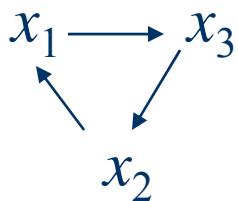
1

0

t

200

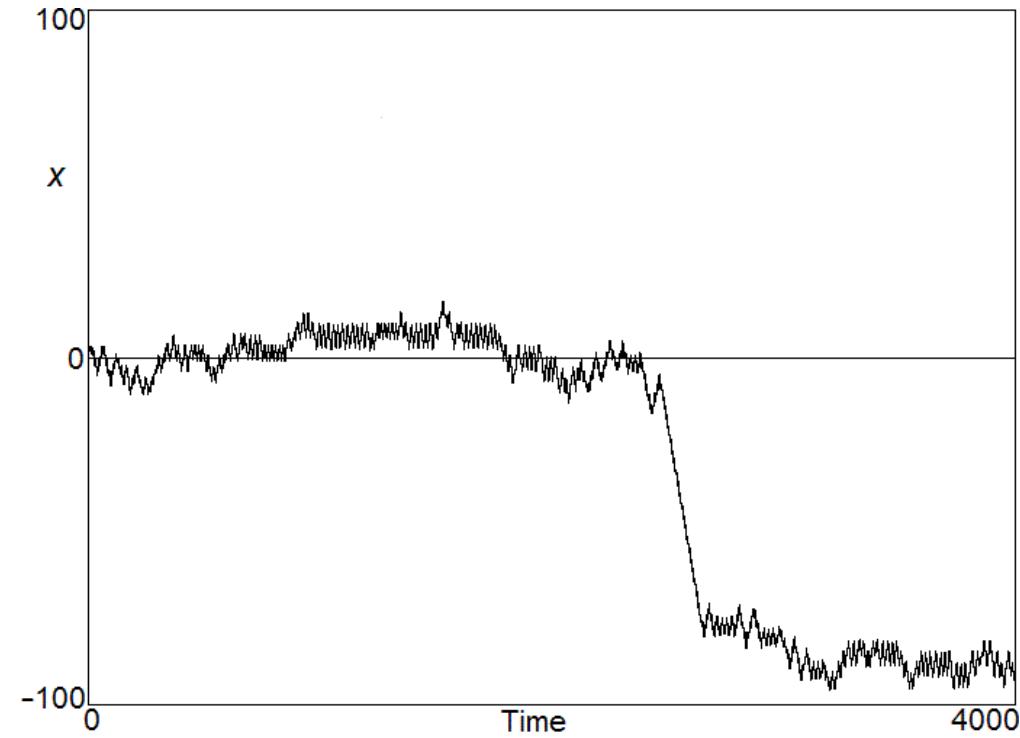
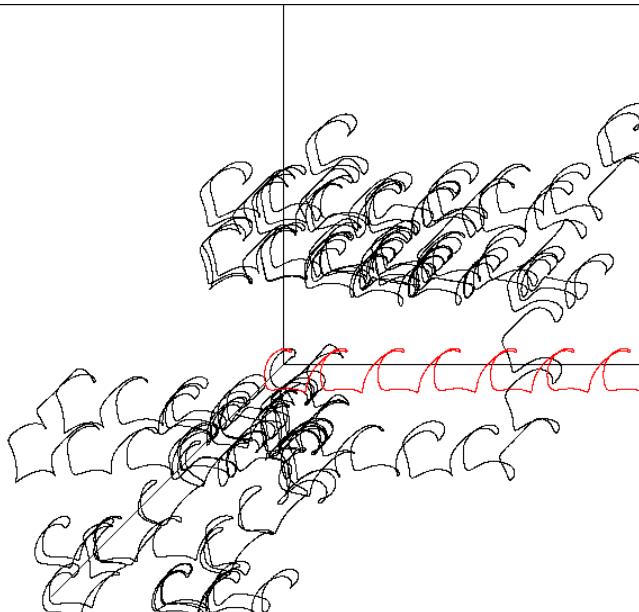
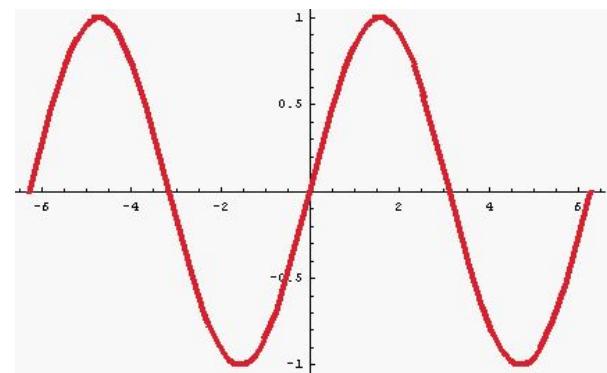
Labyrinth Chaos



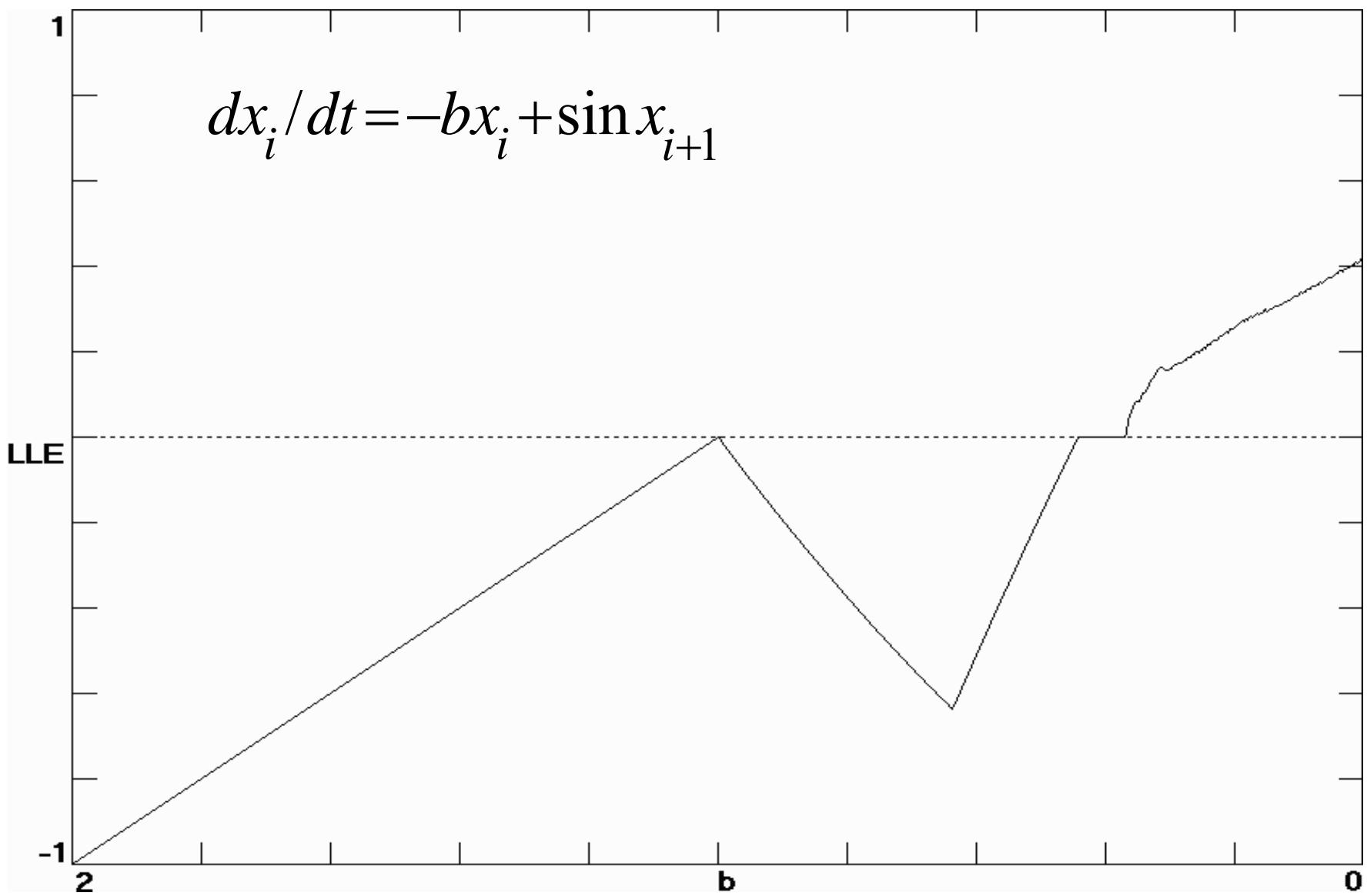
$$dx_1/dt = \sin x_2$$

$$dx_2/dt = \sin x_3$$

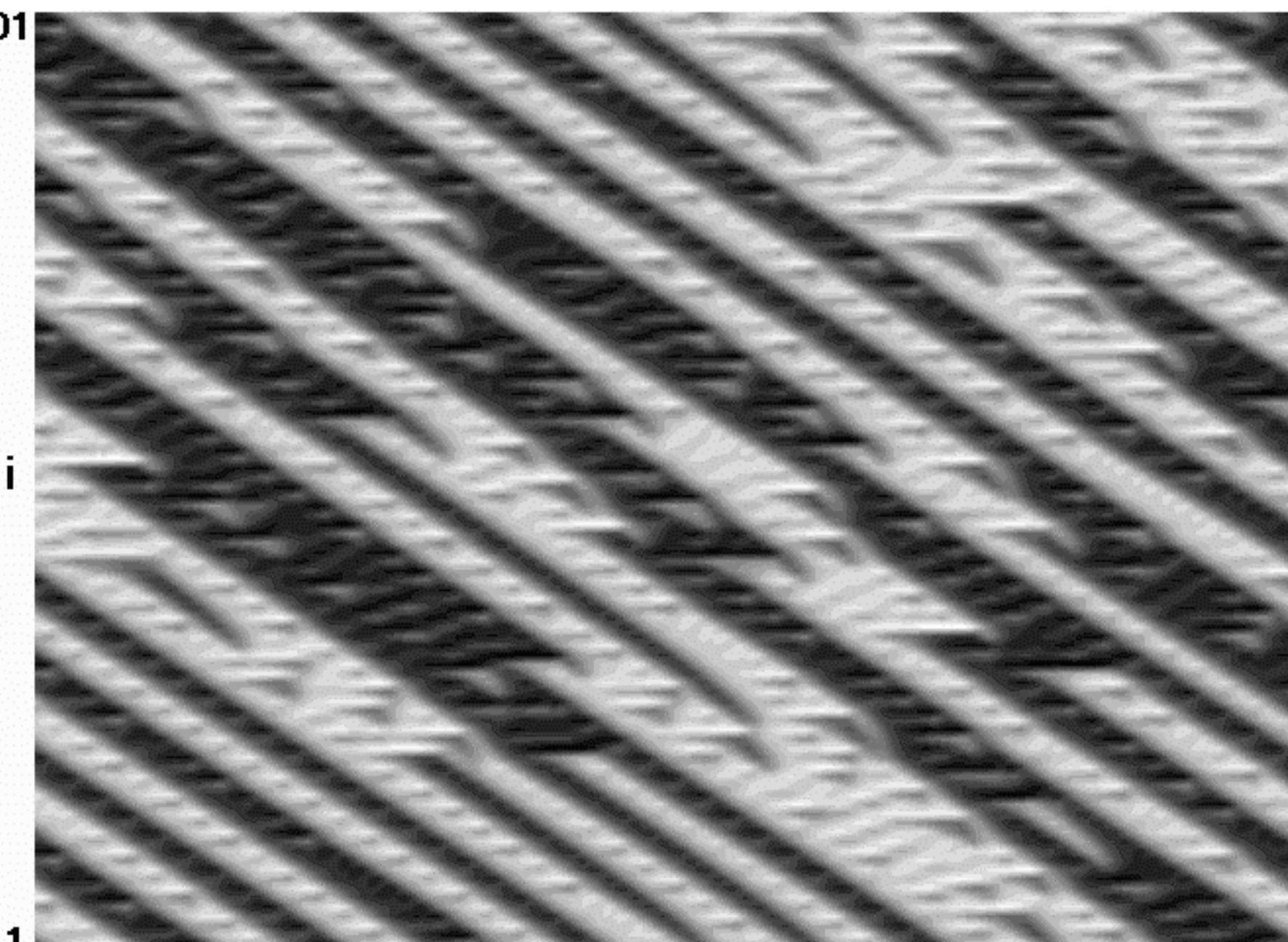
$$dx_3/dt = \sin x_1$$



Hyperlabyrinth Chaos ($N=101$)



101



i

1

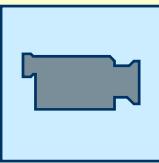
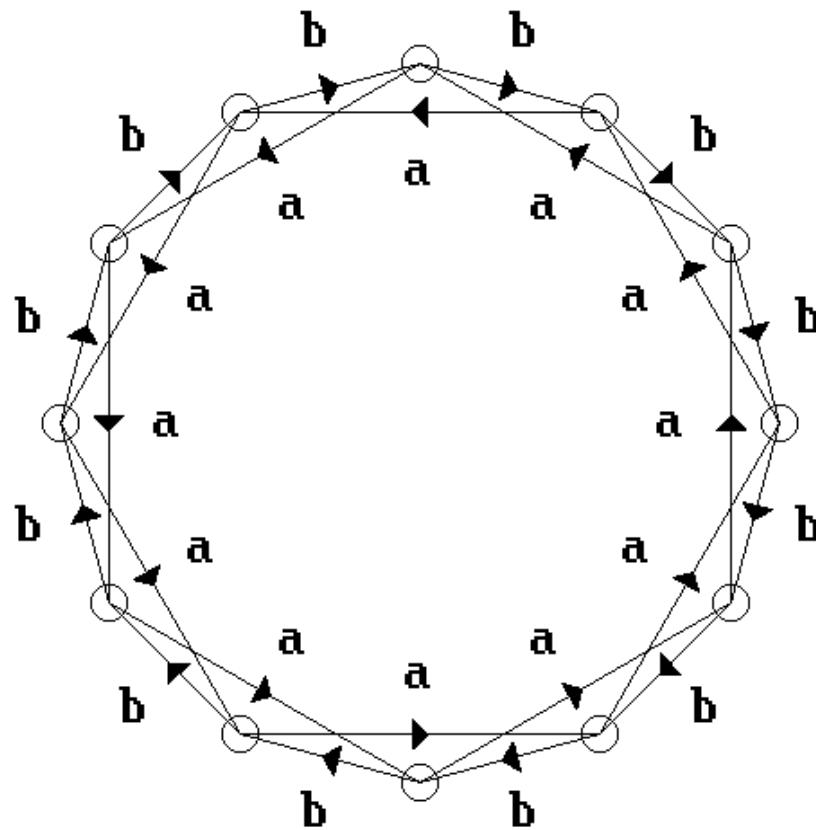
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t

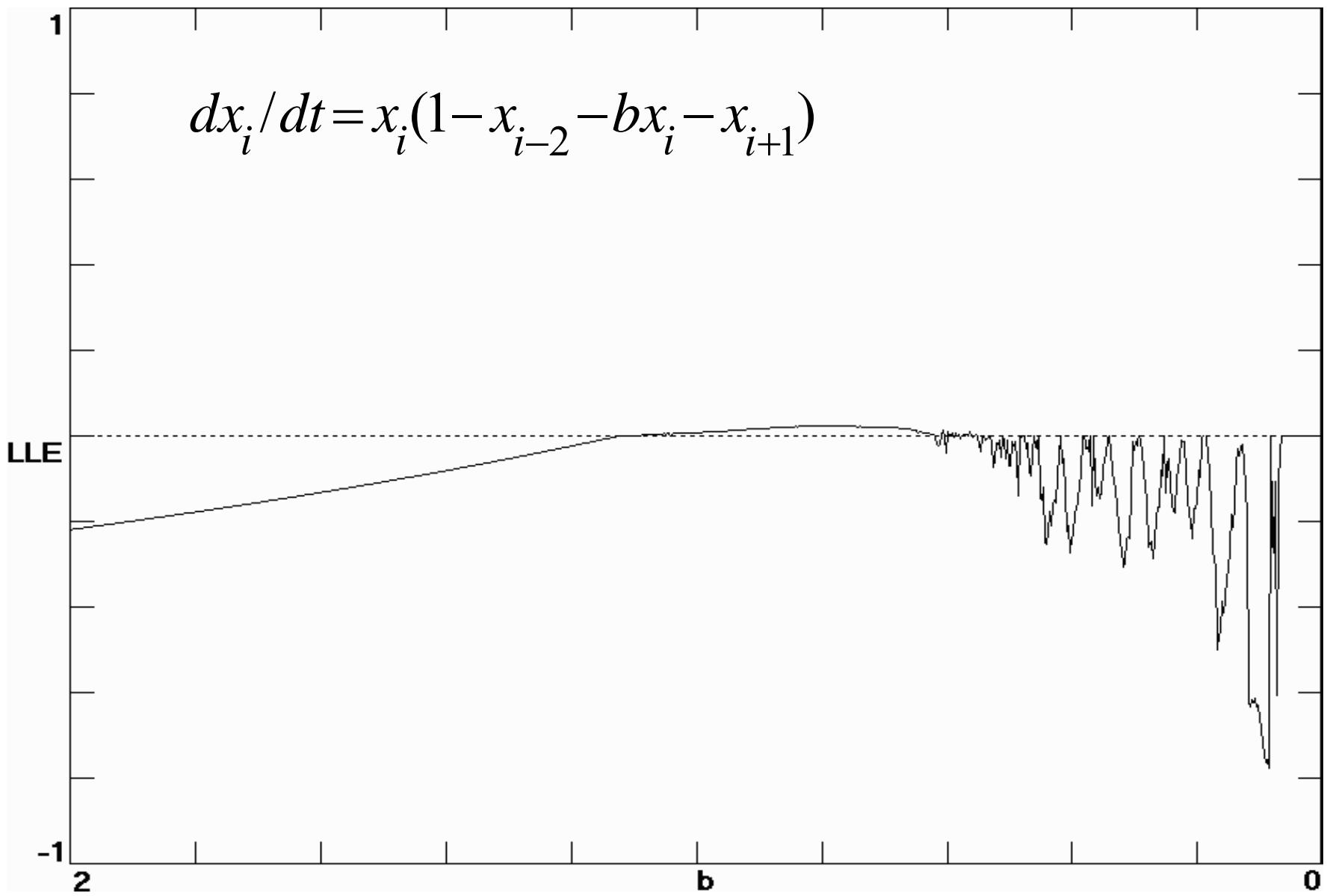
200

Minimal High-D Chaotic L-V Model

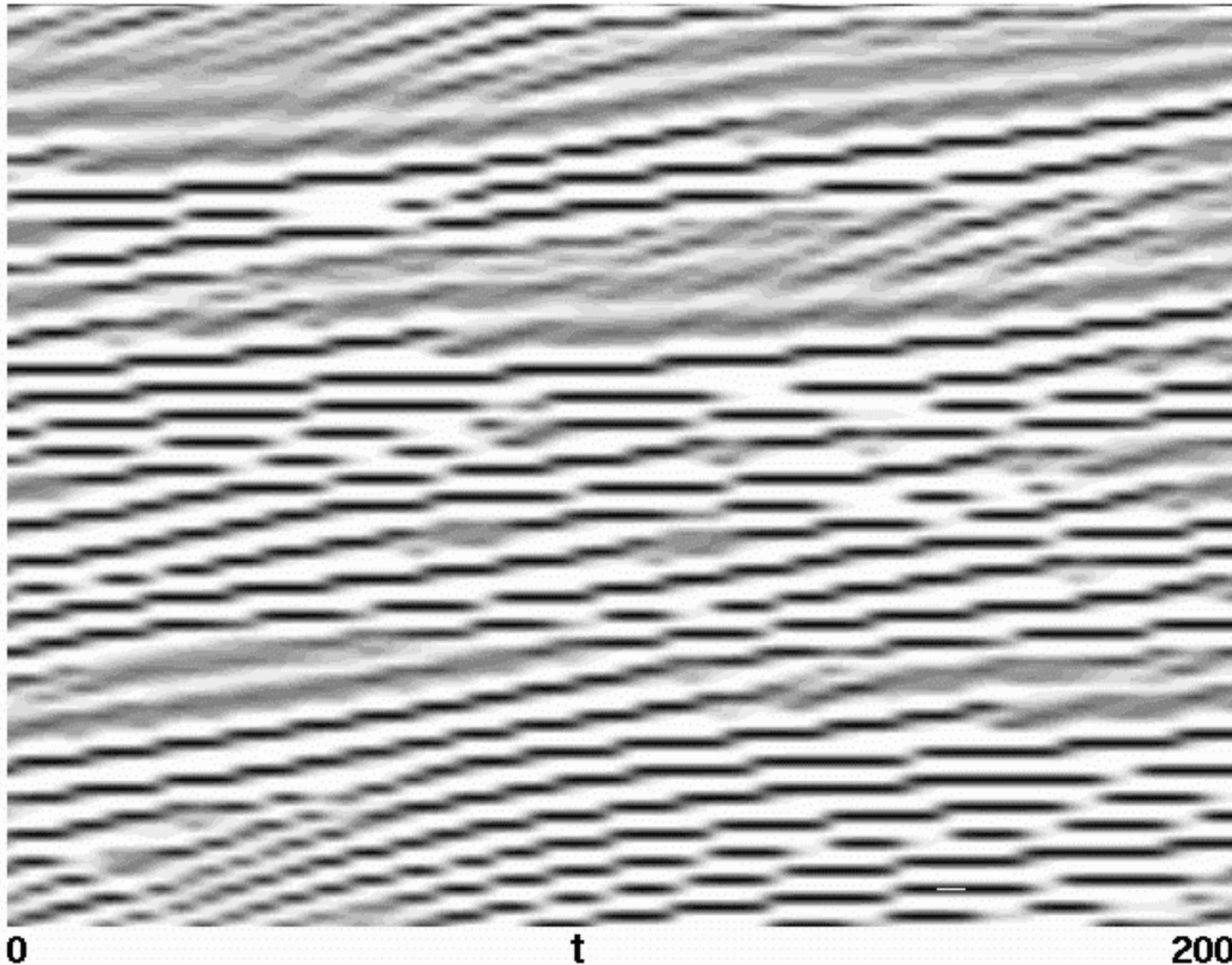
$$dx_i/dt = x_i(1 - x_{i-2} - x_i - x_{i+1})$$



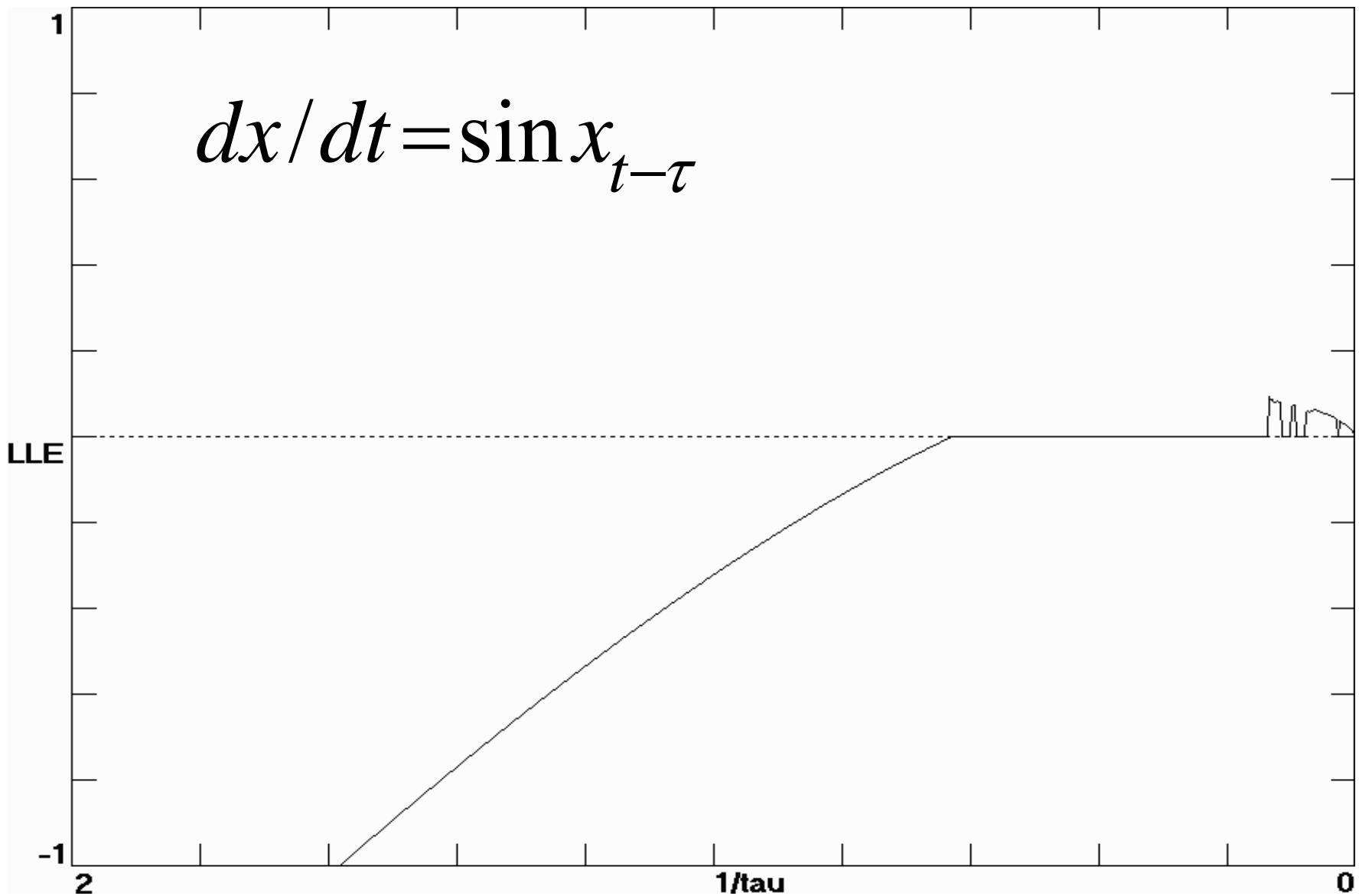
Lotka-Volterra Model ($N=101$)



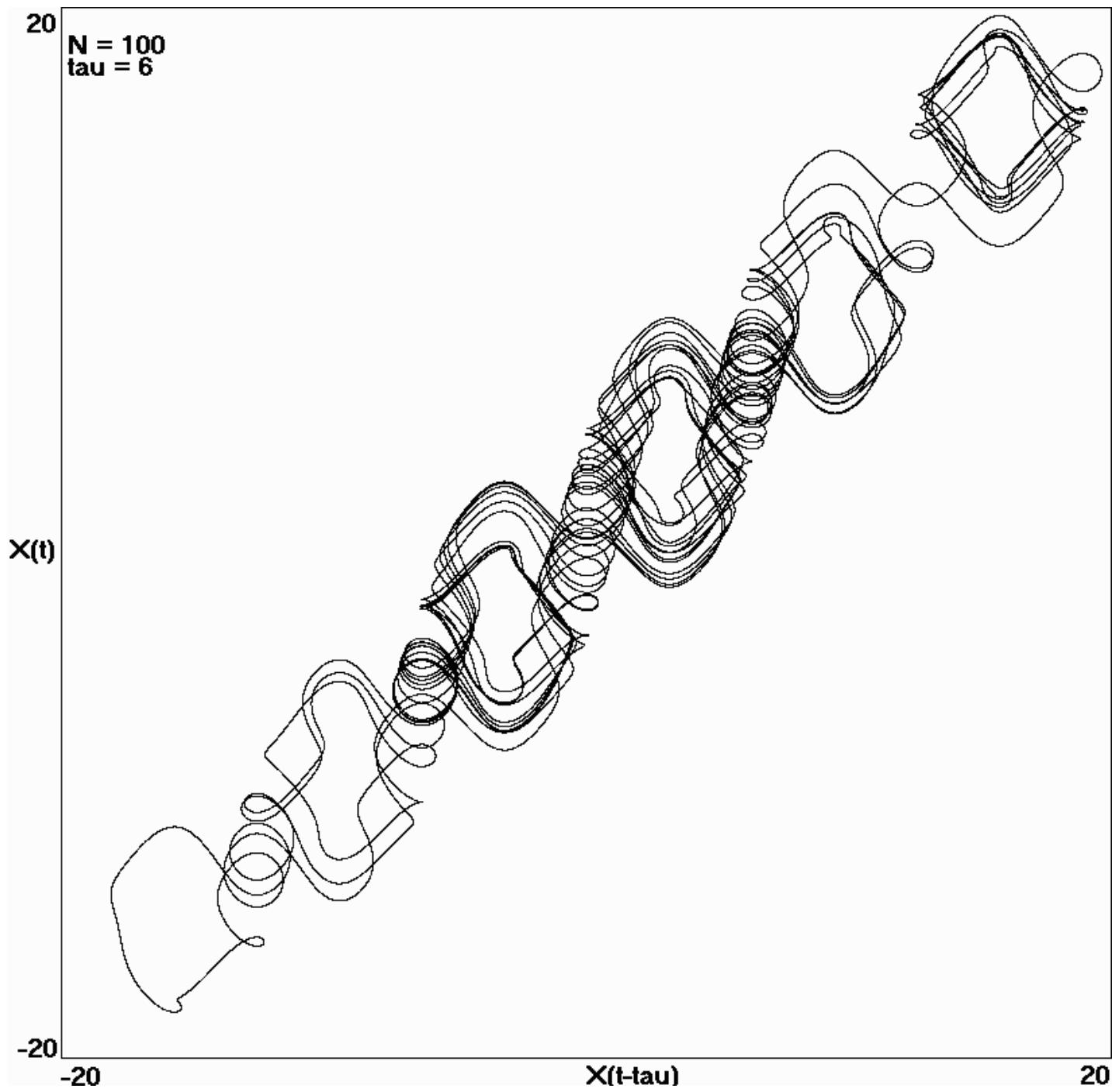
101



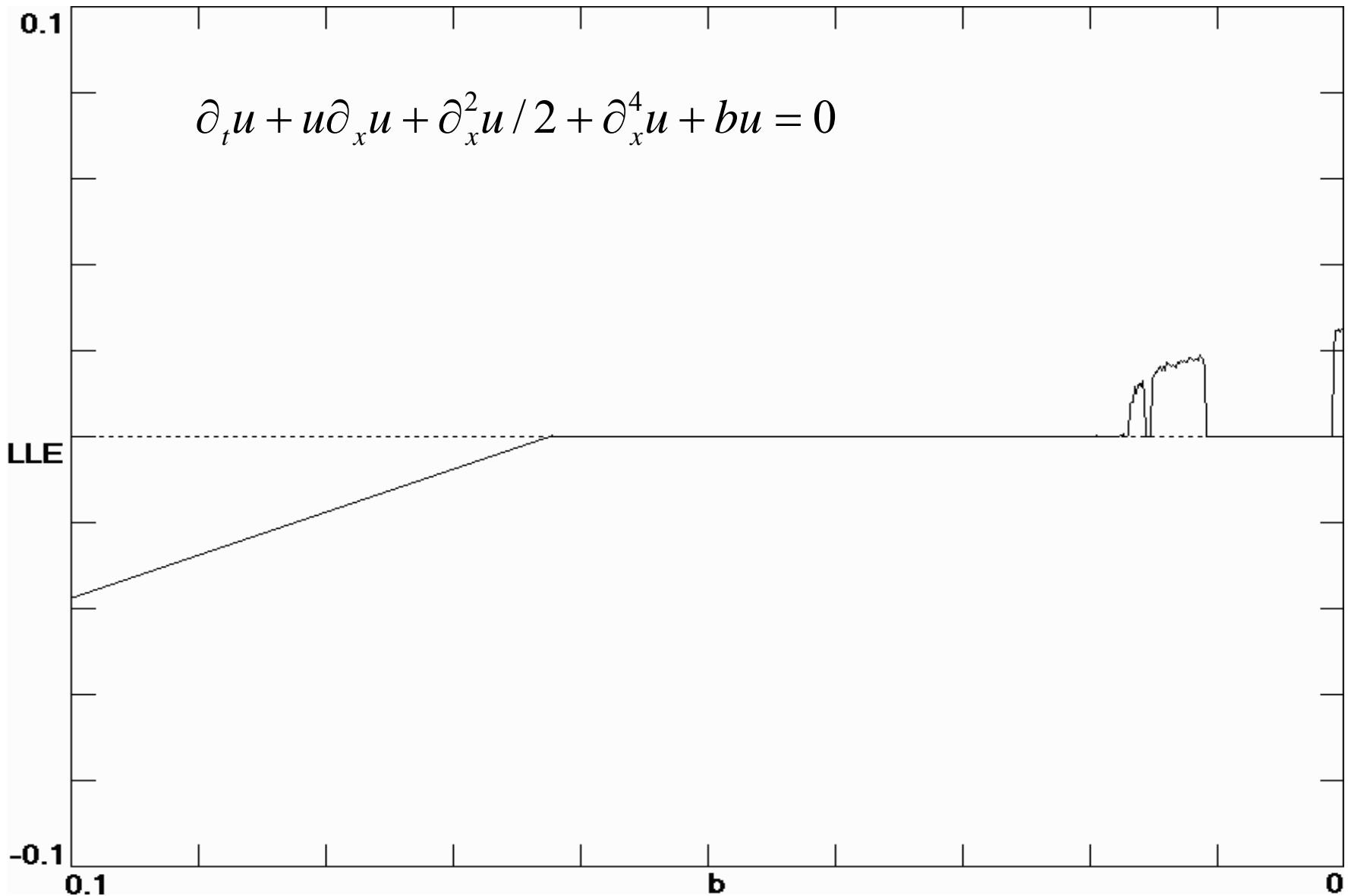
Delay Differential Equation



N = 100
tau = 6



Partial Differential Equation



101

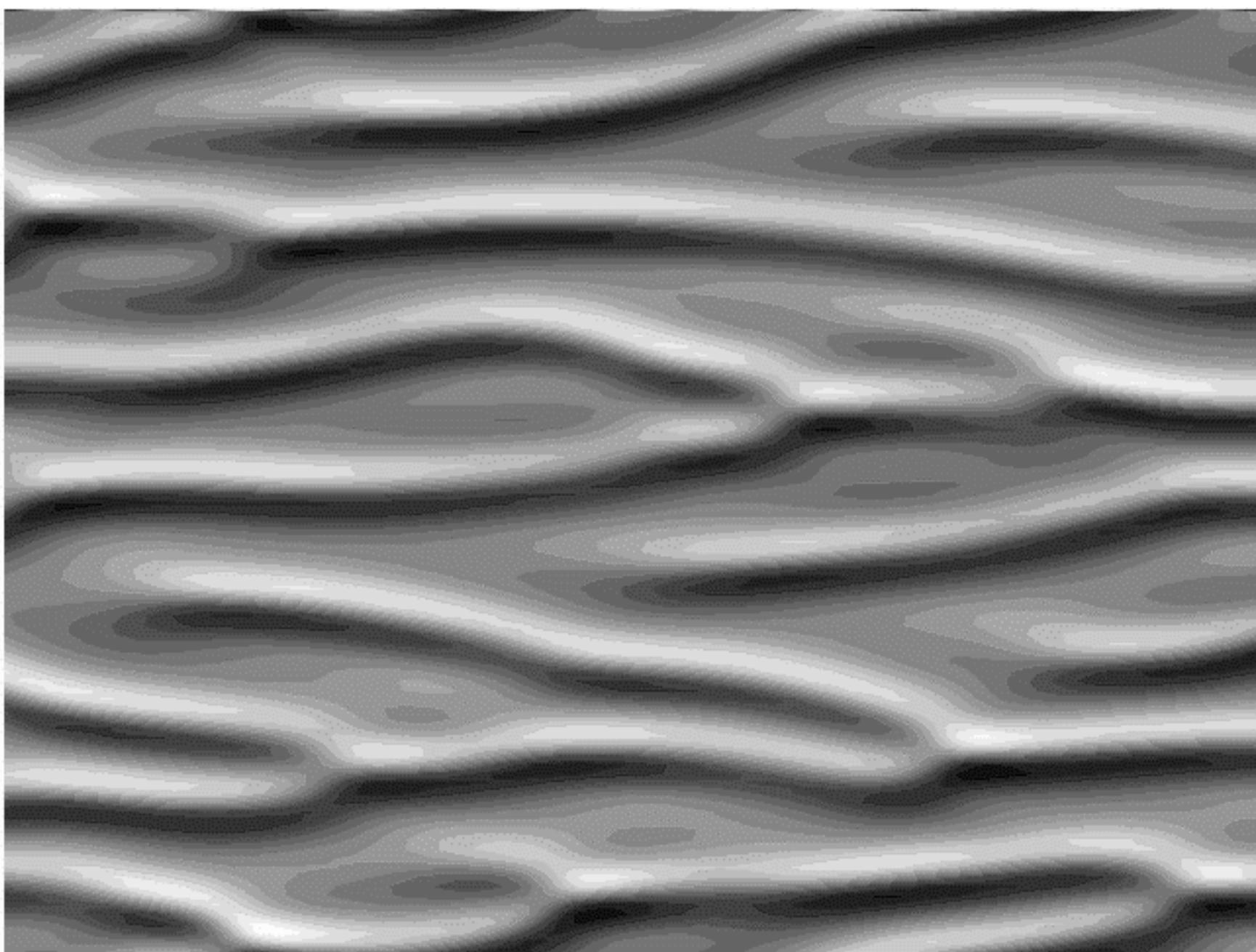
i

1

0

t

200



Summary of High- N Dynamics

- q Chaos is common for highly-connected networks
- q Sparse, circulant networks can also be chaotic (but the parameters must be carefully tuned)
- q Quasiperiodic route to chaos is usual
- q Symmetry-breaking, self-organization, pattern formation, and spatio-temporal chaos occur
- q Maximum attractor dimension is of order $N/2$
- q Attractor is sensitive to parameter perturbations, but dynamics are not

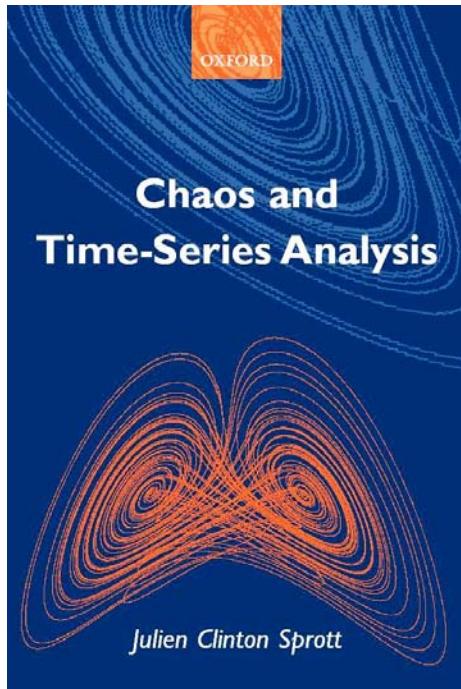
Shameless Plug

Chaos and Time-Series Analysis

J. C. Sprott

Oxford University Press (2003)

ISBN 0-19-850839-5



An introductory text for advanced undergraduate and beginning graduate students in all fields of science and engineering

References

- q [http://sprott.physics.wisc.edu/
lectures/davidson.ppt \(this talk\)](http://sprott.physics.wisc.edu/lectures/davidson.ppt)
- q [http://sprott.physics.wisc.edu/chao
stsa/ \(my chaos textbook\)](http://sprott.physics.wisc.edu/chaosfaqs.html)
- q [sprott@physics.wisc.edu \(contact
me\)](mailto:sprott@physics.wisc.edu)