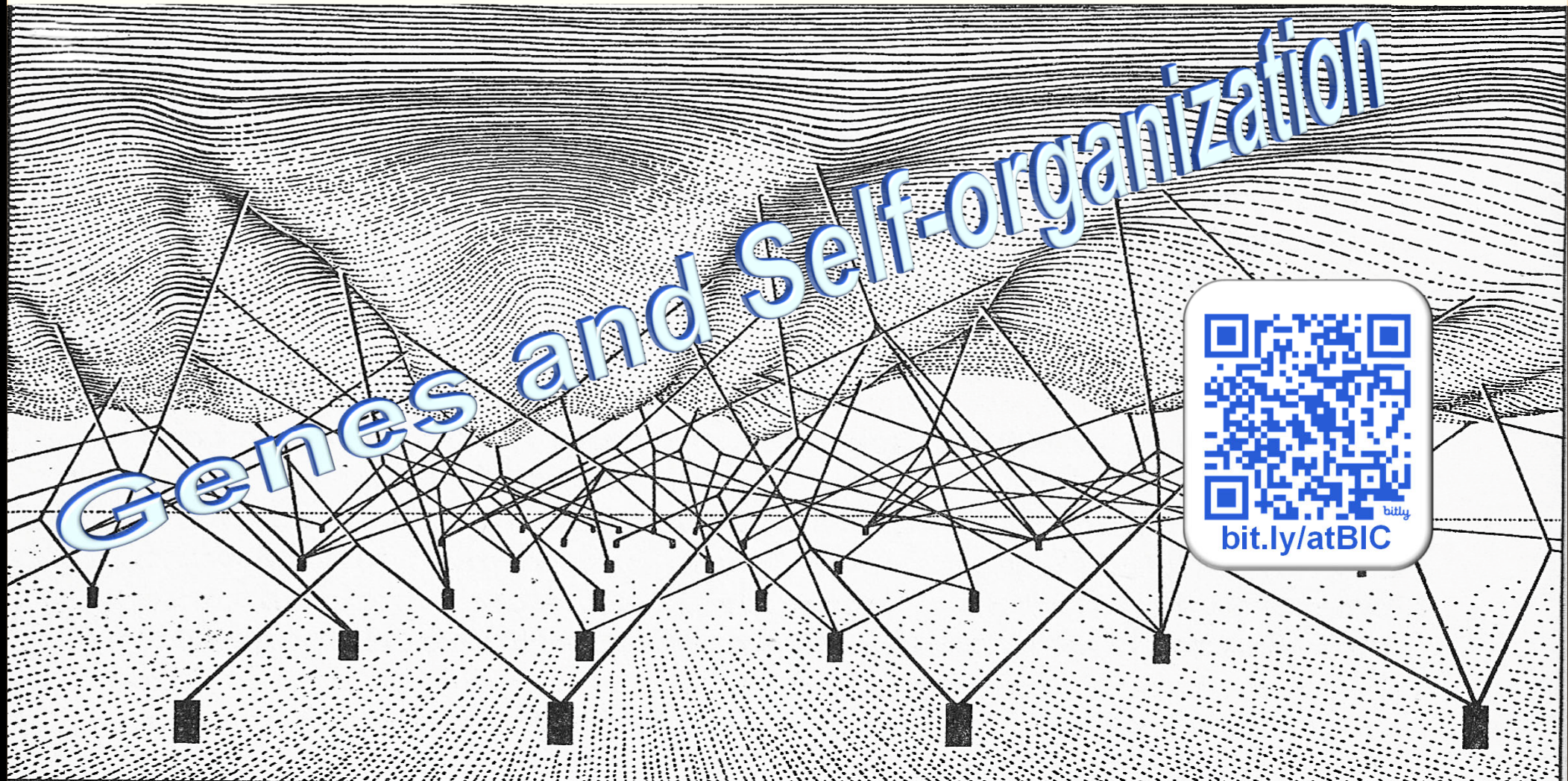


Genes and Self-organization



key events coming up

- **Labs: 35% (ISE-483)**
 - Complete 5 (best 4 graded) assignments based on algorithms presented in class
 - Lab 3: March 11th
 - Cellular Automata and Boolean Networks (Assignment 3)
 - Delivered by SSIE583 Group 3
 - Due: March 18th
- **SSIE – 583 -Presentation and Discussion: 25%**
 - Present and lead the discussion of an article related to the class materials
 - Enginet students post/send video or join by Zoom
 - **Dates TBA**
 - Conrad, M. [1990]. "The geometry of evolution." *Biosystems* **24**: 61-81.
 - Mario Franco
 - Stanley, Kenneth O., Jeff Clune, Joel Lehman, and Risto Miikkulainen. "Designing Neural Networks through Neuroevolution." *Nature Machine Intelligence* **1**, no. 1 (January 2019): 24–35.
 - Jessica Lasebikan
 - Lindgren, K. [1991]. "Evolutionary Phenomena in Simple Dynamics." In: *Artificial Life II*. Langton et al (Eds). Addison-wesley, pp. 295-312.
 - Akshay Gangadhar
 - Salahshour, Mohammad. "Interaction between Games Give Rise to the Evolution of Moral Norms of Cooperation." *PLOS Computational Biology* **18**, no. 9 (September 29, 2022): e1010429
 - Srikanth Iyer
 - Discussion by all



until now

- **Class Book**

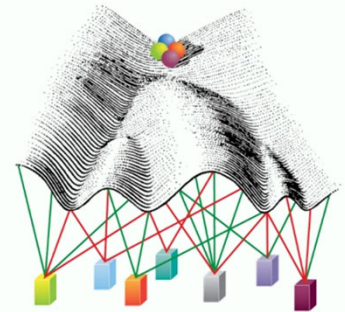
- Floreano, D. and C. Mattiussi [2008]. *Bio-Inspired Artificial Intelligence: Theories, Methods, and Technologies*. MIT Press. Preface, **Chapter 2**.
 - Nunes de Castro, Leandro [2006]. *Fundamentals of Natural Computing: Basic Concepts, Algorithms, and Applications*. Chapman & Hall. **Chapter 1**, pp. 1-23. Chapter 7, sections 7.1-7.4, **Appendix B.3.1**, **Chapter 2**, Chapter 8, sections 8.1, 8.2, 8.3.10

- **Lecture notes**

- Chapter 1: What is Life?
- Chapter 2: The logical Mechanisms of Life
- Chapter 3: Formalizing and Modeling the World
- Chapter 4: Self-Organization and Emergent Complex Behavior
 - posted online @ <http://informatics.indiana.edu/rocha/i-bic>

- **Papers and other materials**

- Dynamical Systems
 - Kauffman, S.A. [1969]. "Metabolic stability and epigenesis in randomly constructed genetic nets". *Journal of Theoretical Biology* 22(3):437-467.
- Optional
 - Prusinkiewicz and Lindenmeyer [1996] *The algorithmic beauty of plants*.
 - Chapter 1
 - Flake's [1998], *The Computational Beauty of Life*. MIT Press.
 - Chapters 10, 11, 14 – Dynamics, Attractors and chaos



bit.ly/atBIC

■ Projects

- Due by May 6th in Brightspace, “Final Project Paper” assignment
 - ALIFE 2023
 - Not to submit to actual conference due date (April 3rd , 2024)
 - <https://2024.alife.org/>
 - 8 pages, author guidelines:
 - https://2024.alife.org/call_paper.html
 - MS Word and Latex/Overleaf templates
 - Preliminary ideas **by March 15**
 - Submit to “Project Idea” assignment in Brightspace.
- Individual or group
 - With very definite tasks assigned per member of group

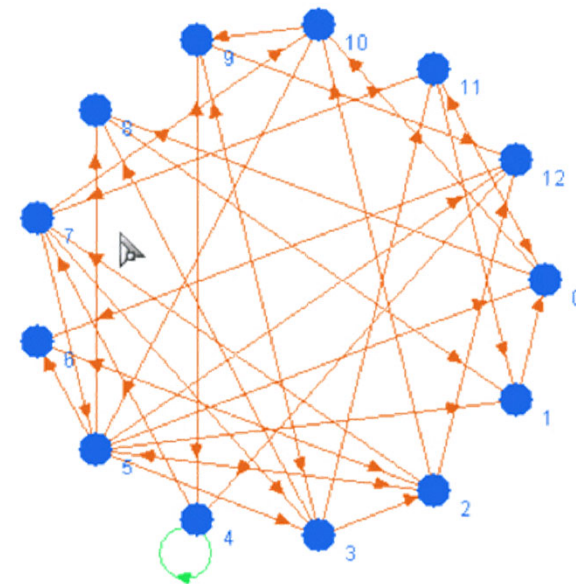
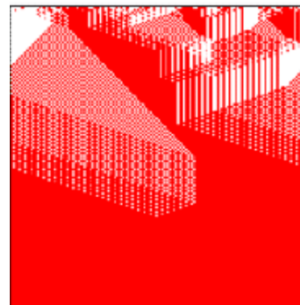
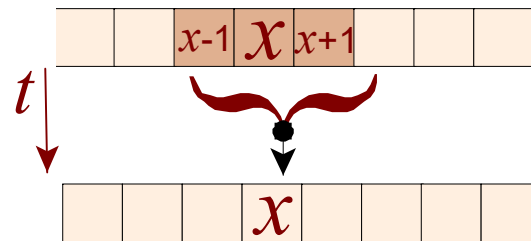
ALIFE 2024

Tackle a real problem using bio-inspired algorithms, such as those used in the labs.



examples

Cellular Automata



NK Boolean Network (N=13, K=3)

dynamical models of regulation from qualitative data

the drosophila segment polarity network



Based on the ODE model of von Dassow et al. (2000), consists of 4-cell parasegments, each cell with 15 interacting genes and proteins.

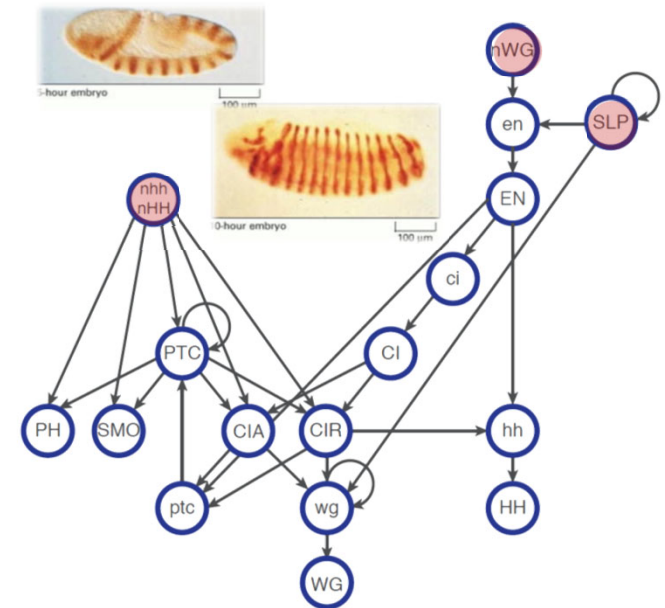
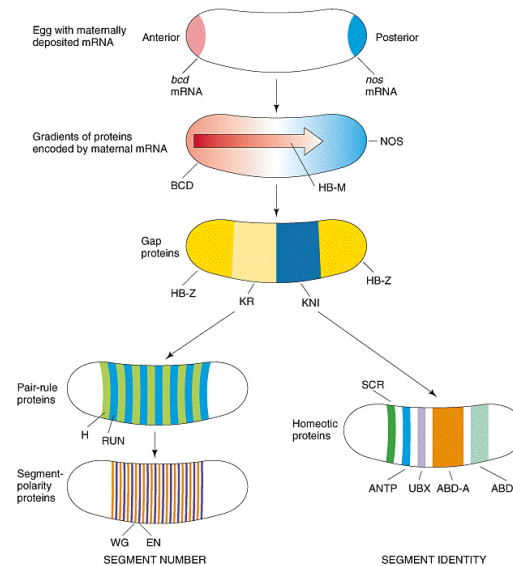
2⁶⁰ network configurations

Reproduces wild-type and mutant gene expression patterns in development of fruit fly

2 intercellular inputs: **nhh** (*hedgehog*), **nWG** (*wingless*)

1 intracellular input: **SLP** (*sloppy paired*)

Node	State – TransitionFunction
SLP_i^{t+1}	$\leftarrow 0 \text{ if } i=1 \vee i=2; 1 \text{ if } i=3 \vee i=4;$
wg_i^{t+1}	$\leftarrow (CIA_i^t \wedge SLP_i^t \wedge \neg CIR_i^t) \vee (wg_i^t \wedge (CIA_i^t \vee SLP_i^t) \wedge \neg CIR_i^t)$
WG_i^{t+1}	$\leftarrow wg_i^t$
en_i^{t+1}	$\leftarrow (WG_{i-1}^t \vee WG_{i+1}^t) \wedge \neg SLP_i^t$
EN_i^{t+1}	$\leftarrow en_i^t$
hh_i^{t+1}	$\leftarrow EN_i^t \wedge \neg CIR_i^t$
HH_i^{t+1}	$\leftarrow hh_i^t$
ptc_i^{t+1}	$\leftarrow CIA_i^t \wedge \neg EN_i^t \wedge \neg CIR_i^t$
PTC_i^{t+1}	$\leftarrow ptc_i^t \vee (PTC_i^t \wedge \neg HH_{i-1}^t \wedge \neg HH_{i+1}^t)$
PH_i^t	$\leftarrow PTC_i^t \wedge (HH_{i-1}^t \vee HH_{i+1}^t)$
SMO_i^t	$\leftarrow \neg PTC_i^t \vee (HH_{i-1}^t \vee HH_{i+1}^t)$
ci_i^{t+1}	$\leftarrow \neg EN_i^t$
CI_i^{t+1}	$\leftarrow ci_i^t$
CIA_i^{t+1}	$\leftarrow CI_i^t \wedge (\neg PTC_i^t \vee hh_{i-1}^t \vee hh_{i+1}^t \vee HH_{i-1}^t \vee HH_{i+1}^t)$
CIR_i^{t+1}	$\leftarrow CI_i^t \wedge PTC_i^t \wedge \neg hh_{i-1}^t \wedge \neg hh_{i+1}^t \wedge \neg HH_{i-1}^t \wedge \neg HH_{i+1}^t$



Albert & Othmer [2003]. *J. Theor. Bio.* 223: 1-18.

dynamical models of regulation from qualitative data

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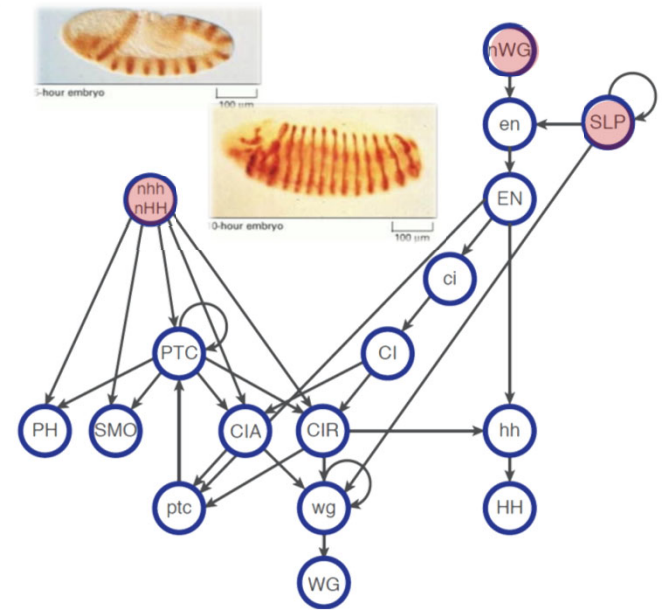
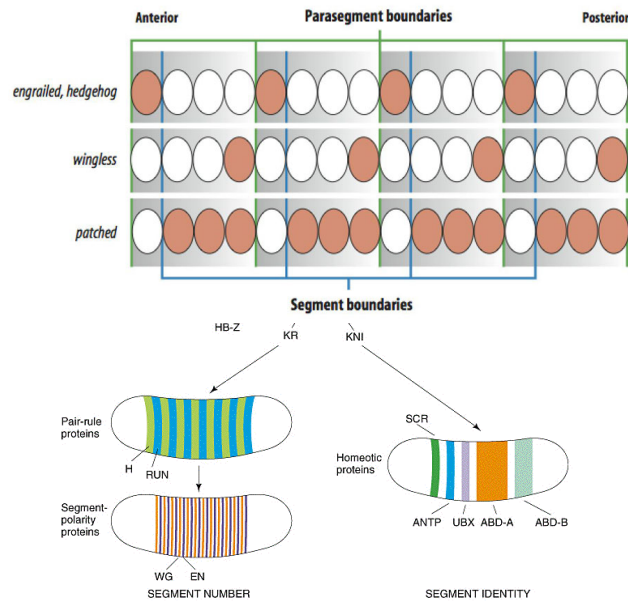
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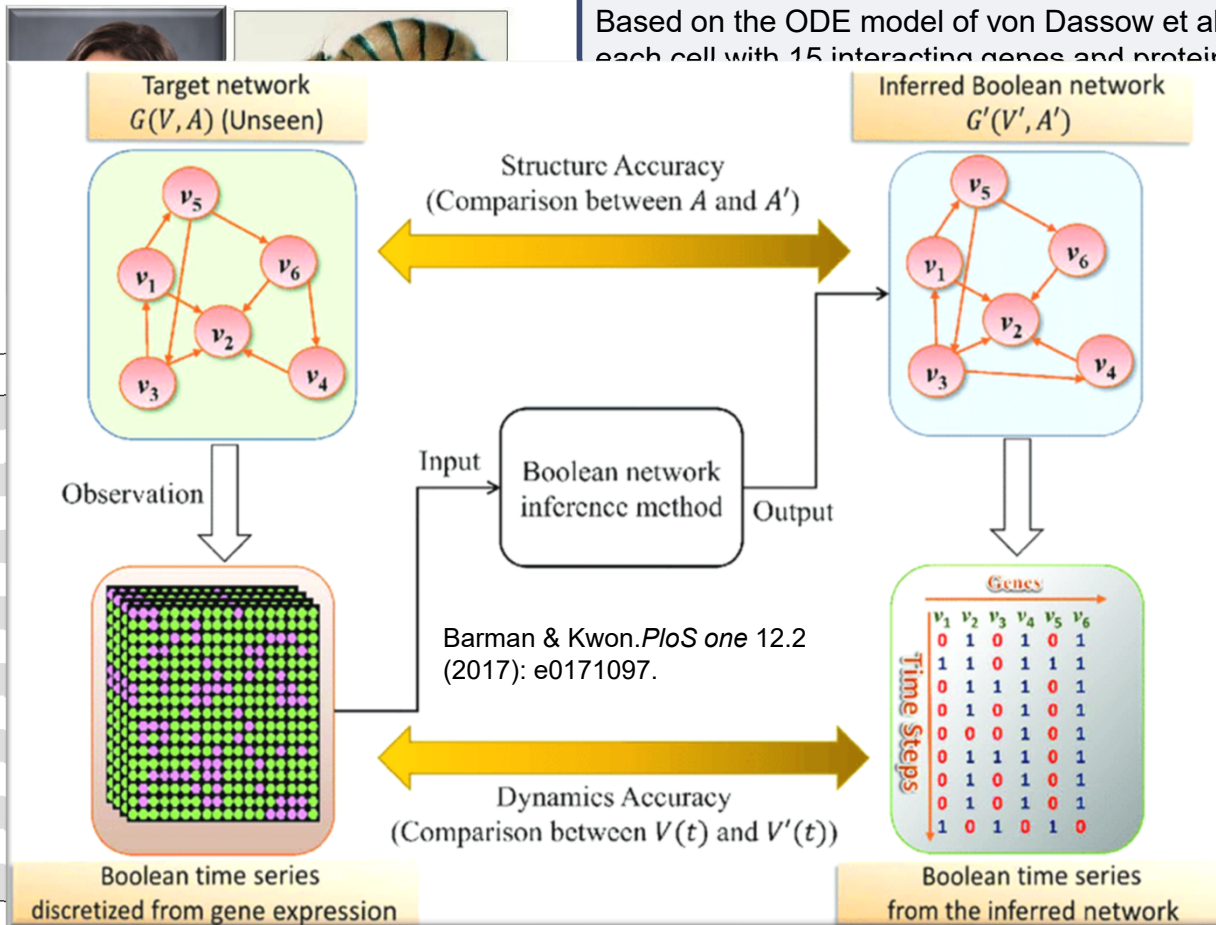
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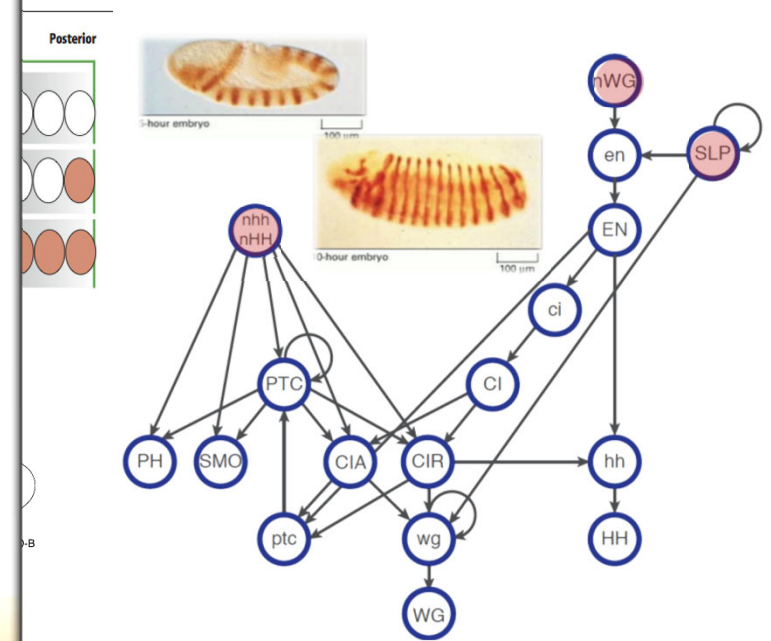
Albert & Othmer [2003]. *J. Theor. Bio.* 223: 1-18.

the drosophila segment polarity network

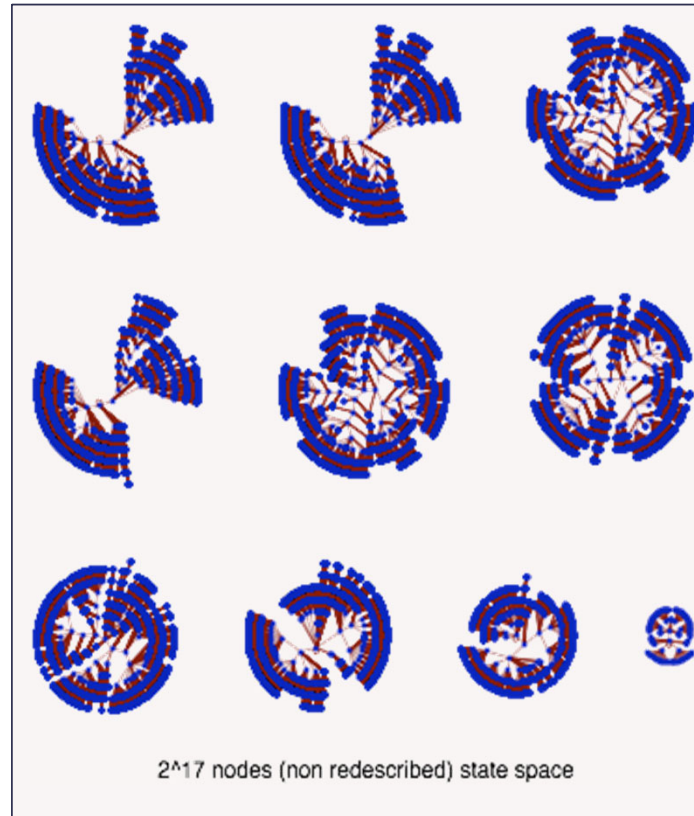
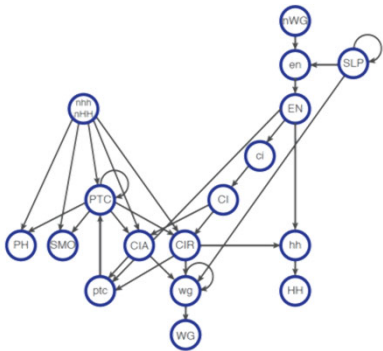
Based on the ODE model of von Dassow et al. (2000), consists of 4-cell parasegments, each cell with 15 interacting genes and proteins.



expression patterns in development of fruit fly (*wingless*)



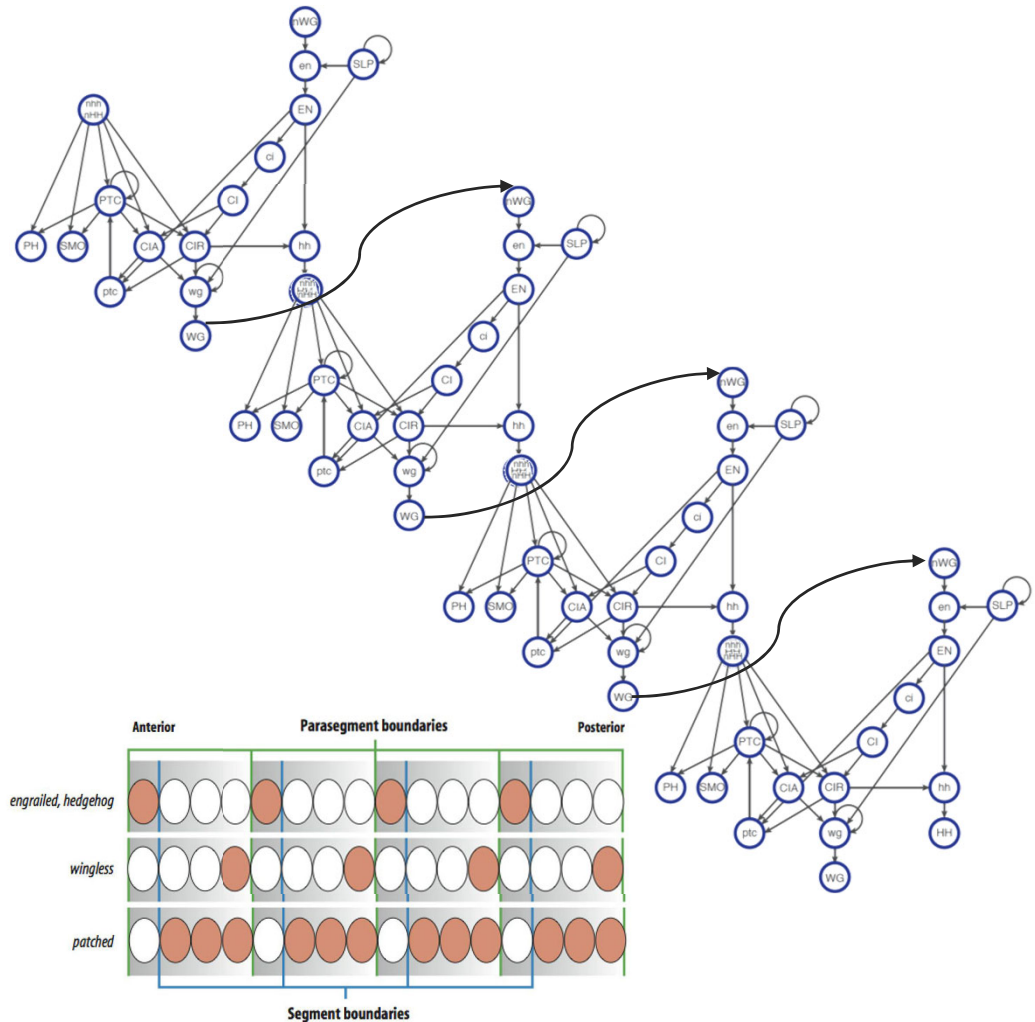
cell-types in a spatial arrangement



- Observing the state-transition graph
 - converges to one of 10 possible stable configurations
 - Steady-state attractors
- observed experimentally
 - wildtype
 - plus 3 variants
 - broad stripe
 - no-segmentation
 - Ectopic
 - plus 3 variants

Albert & Othmer [2003]. *J. Theor. Bio.* **223**: 1-18.

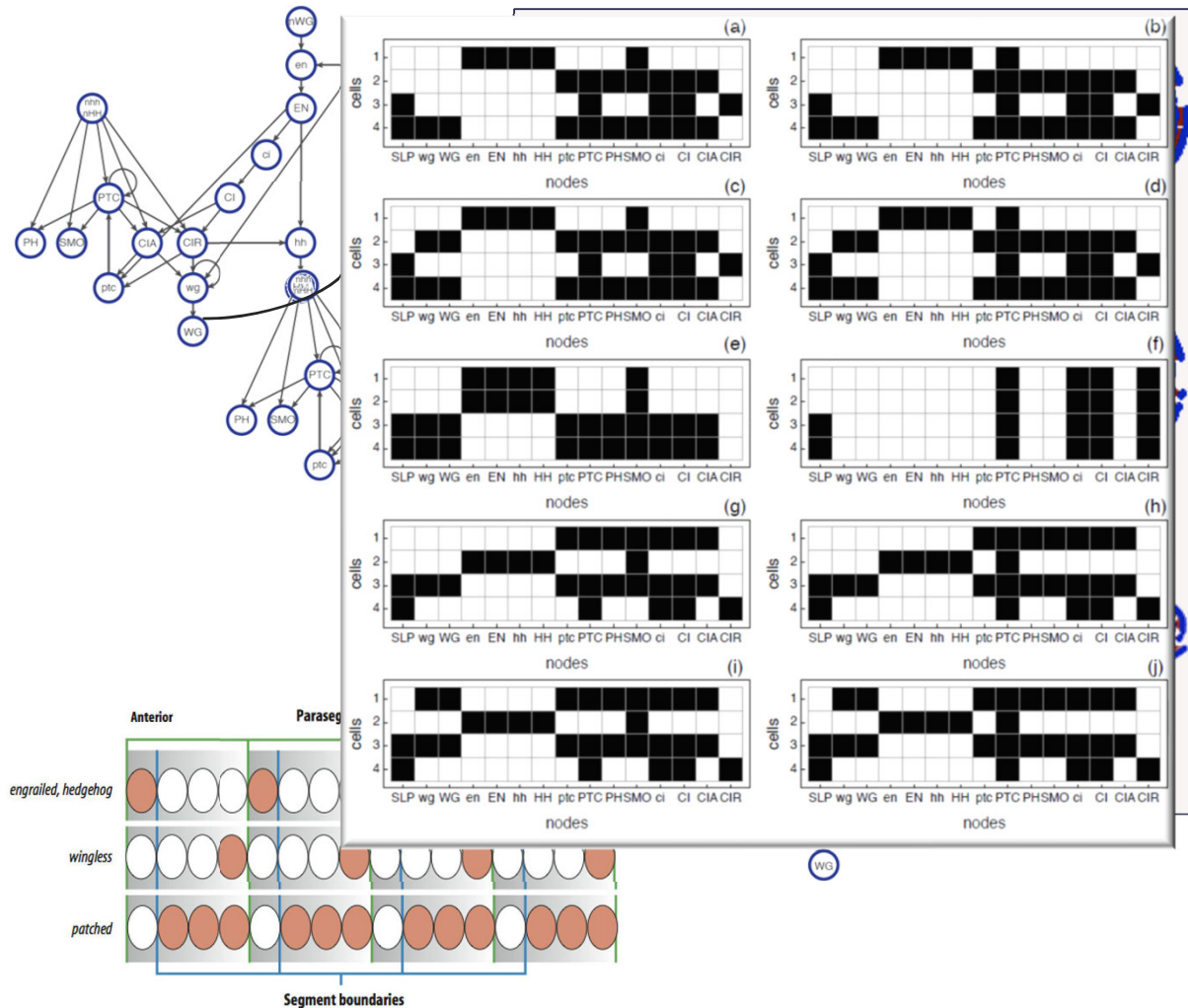
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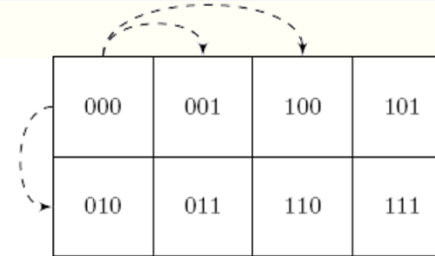
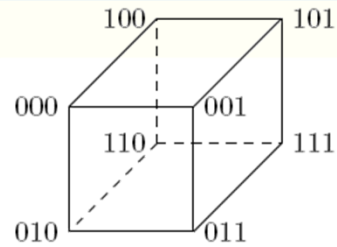


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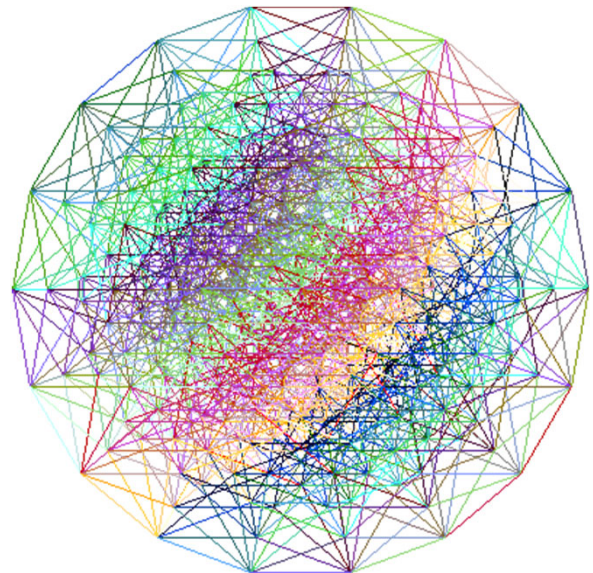
Albert & Othmer [2003]. *J. Theor. Bio.* **223**: 1-18.

higher-dimensional hyper-cube

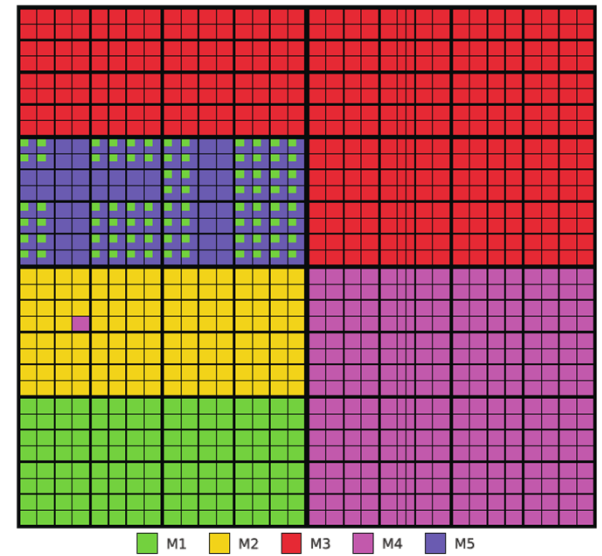
representations



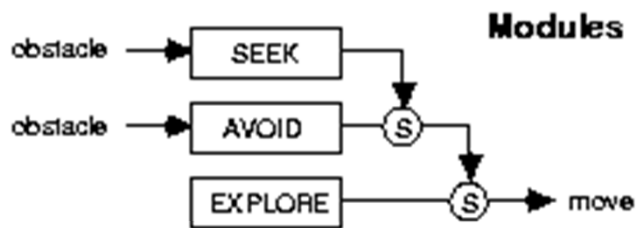
9-dimension representation



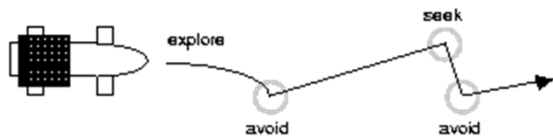
Planar representation
16 state drosophila segment polarity network
(from Willadsen and Wiles , 2007)



Robot example

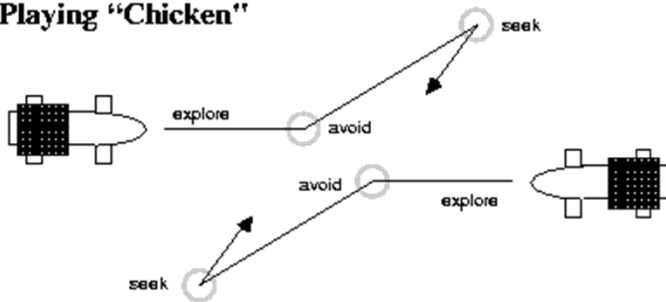


Typical Path



wall

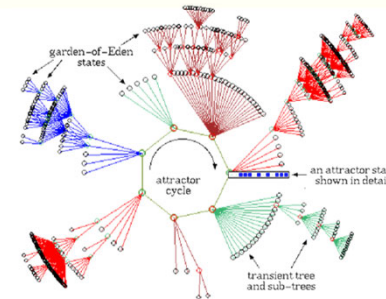
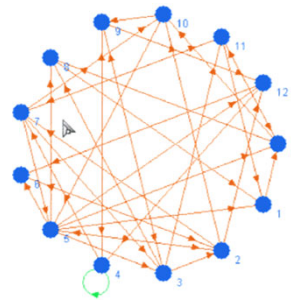
Playing "Chicken"



Jonathan Connell 's Muramator

- Emergent Behavior from system/environment coupling
 - Classifies Walls and Other Robots
 - Self-organization
 - Embodied cognition

self-organization



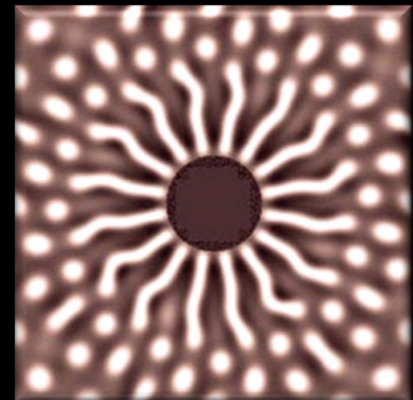
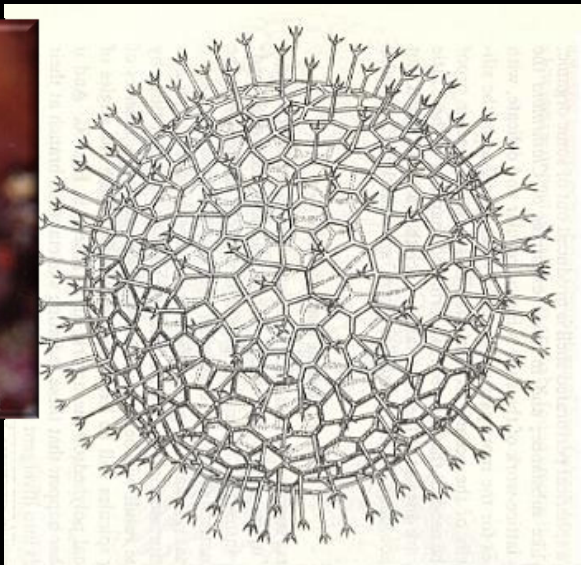
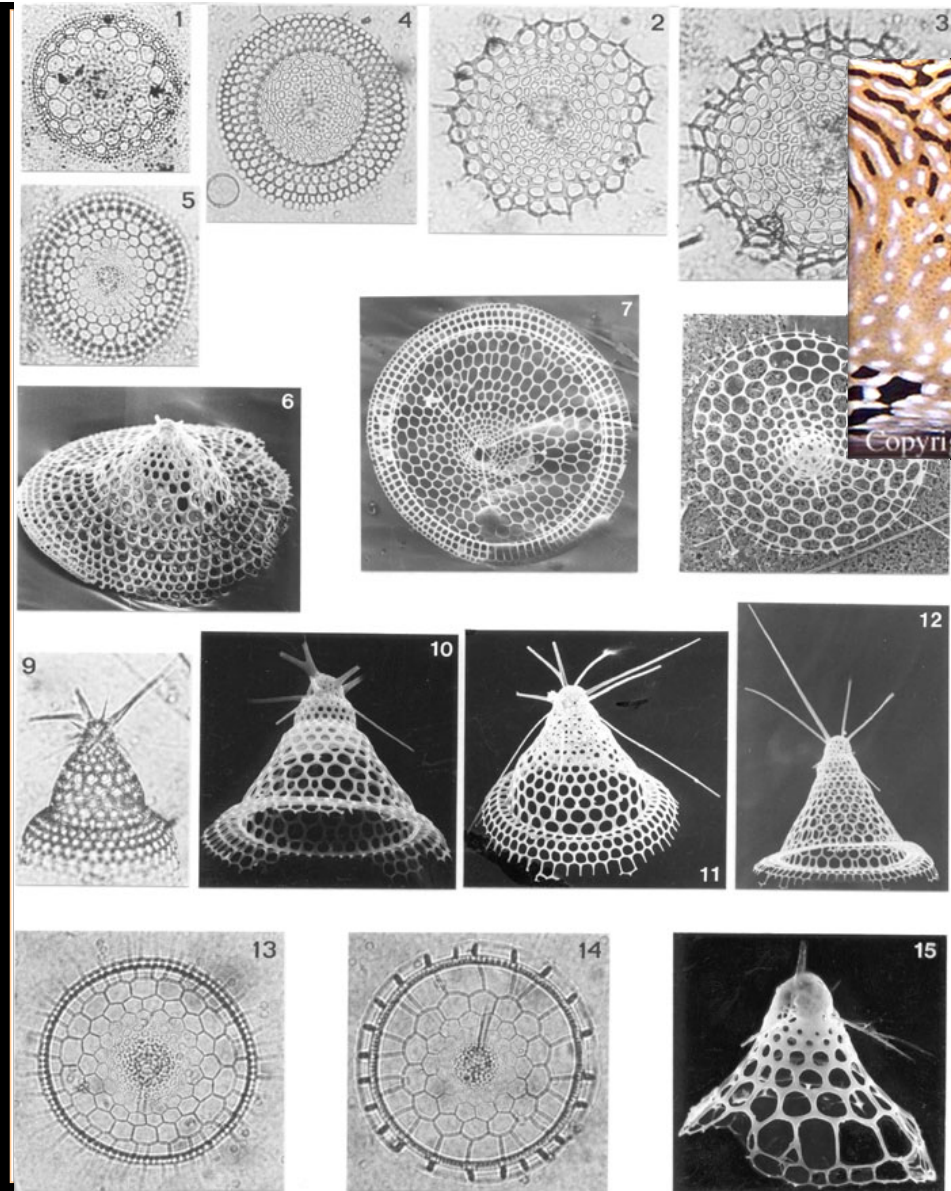
- **Discrete dynamical systems**
 - Extremely large number of coupled elements
 - Systems of binary variables (0,1), coupled to one another in a network
 - The activity of each element depends on previous state of other elements
 - Simplifies continuous systems while maintaining essential behavior
- **Statistical properties of sets of networks**
 - Understanding of macroscopic, emergent properties
 - Similar to temperature
- **Typically irreversible**

self-organization

- Genetic regulatory networks
 - Genes are on or off
 - Development, morphogenesis
 - Attractors interpreted as different cell types
- Classification in Immune networks
- Representation in artificial neural networks
- Stable patterns of species abundances in ecosystems

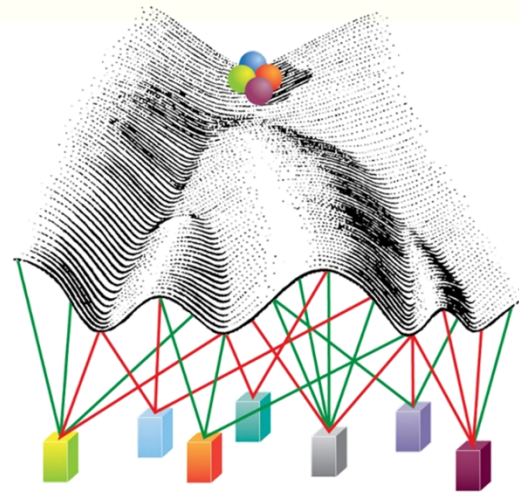
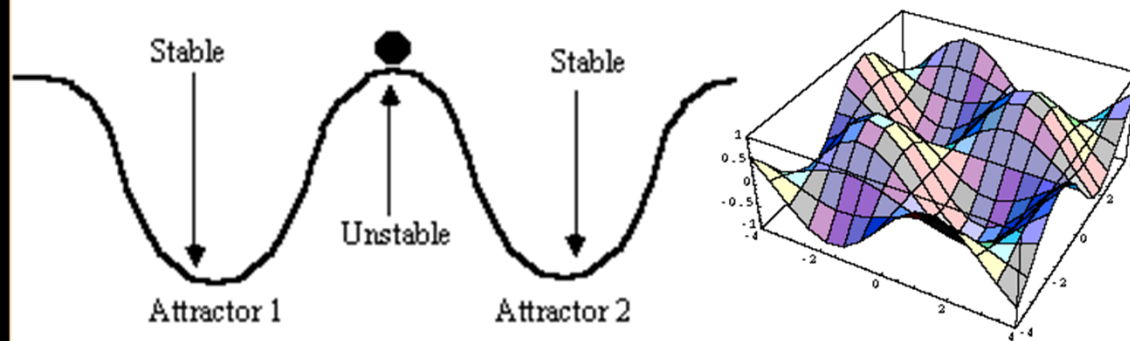


Order is the raw material for evolution: how much of life is Natural Selection and how much is self-organization?
(credit assignment problem)



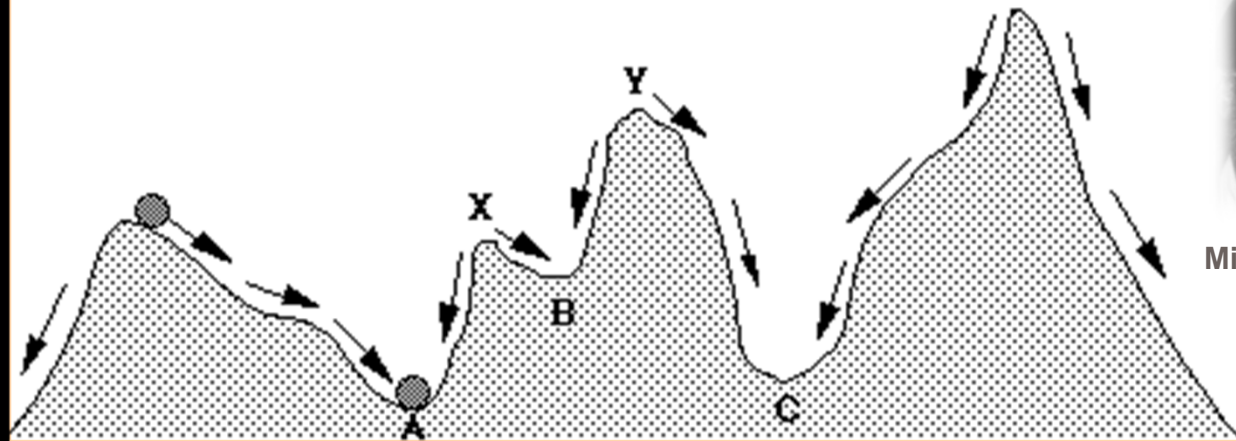
(energy) landscape metaphor

- Phase-space as landscape
 - State of the system as a drop of water released in hills and valleys

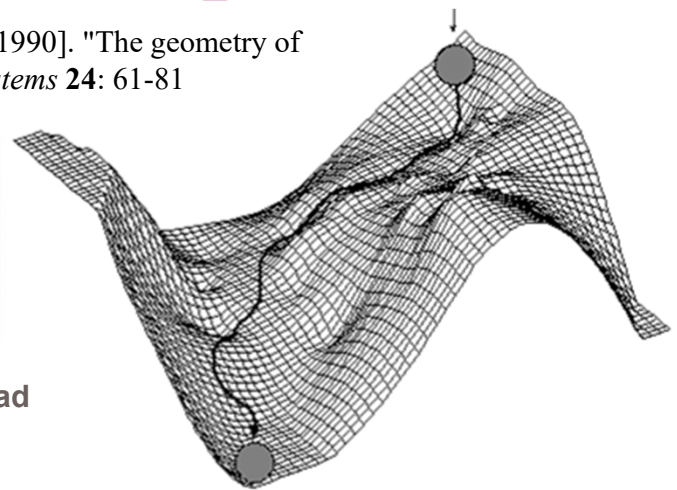


Waddington CH (1942). *Nature*. **150** (3811):563–565

See: Conrad, M. [1990]. "The geometry of evolution." *Biosystems* **24**: 61-81

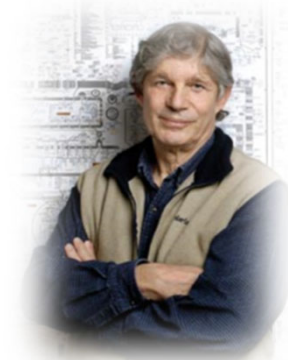
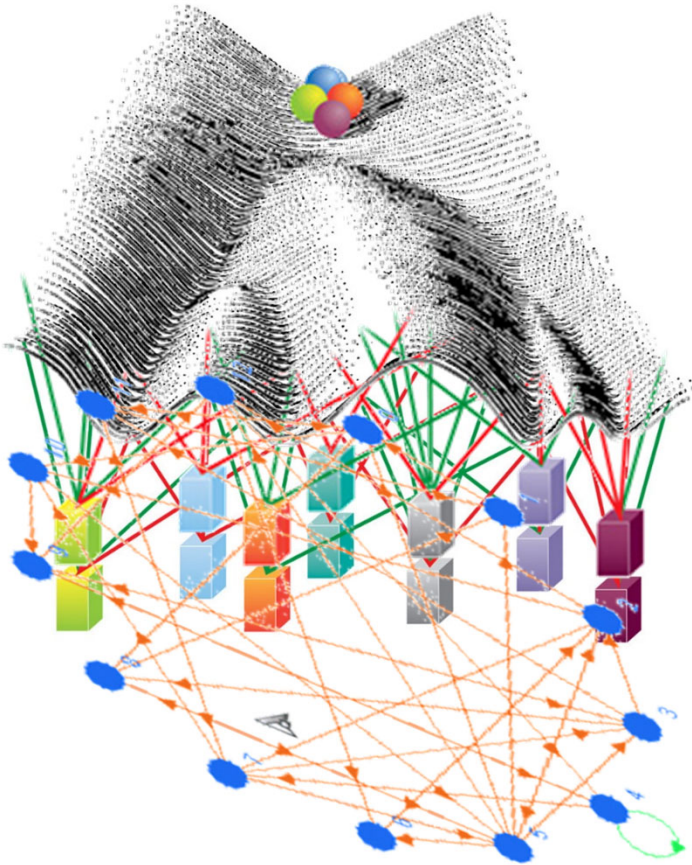


Michael Conrad



self-organization as a key mechanism for order and robustness

evolution does not need to encode all details and is constrained



Kauffman, S. A. (1984). *Phys. D Nonlinear Phen.* **10**, 145–156.

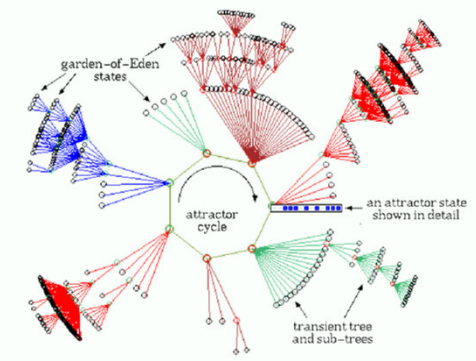
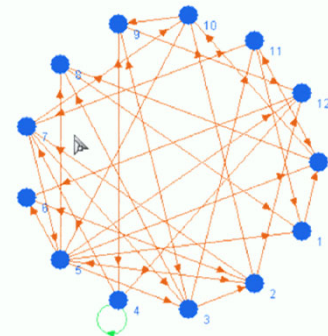
Waddington CH (1942). *Nature.* **150** (3811):563–565

robustness of phenotypes is the result of a **buffering** of the developmental process.

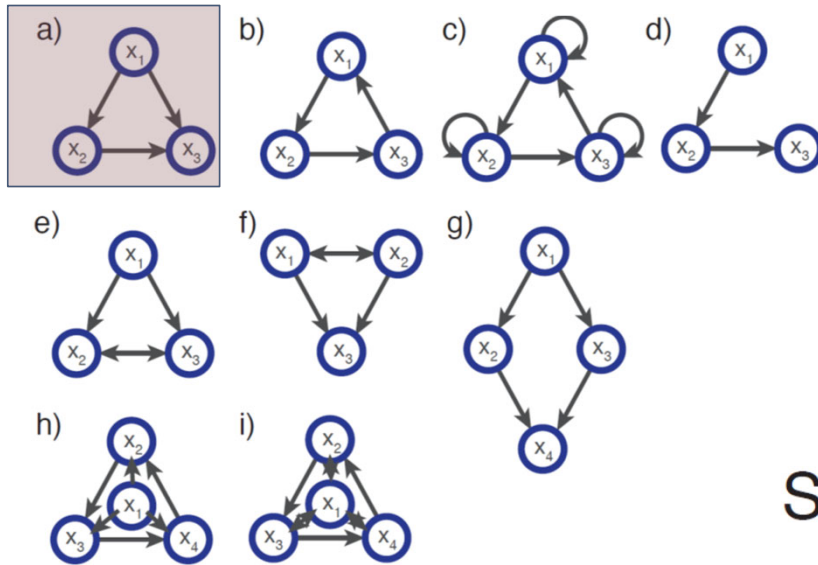
dynamics of gene networks provides buffering (**self-organization**).

definitions

- **basin of attraction**
 - All states in trajectories leading to an attractor (state cycle)
- **length of cycle**
 - Number of states in cycle
 - 1 to 2^N
- **perturbation (minimal)**
 - Flipping of one node to the opposite state
- **Damage**
 - Change in behavior from a perturbation
- **Structural perturbation**
 - Permanent in connections or Boolean rules in the network



ensemble dynamics for same structure



different Boolean networks for same structure

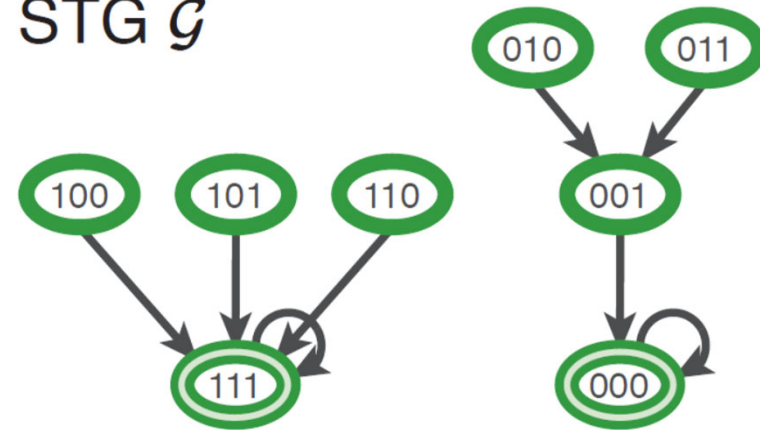
$$\prod_{i=1}^N 2^{2^{k_i}}$$

a) 64

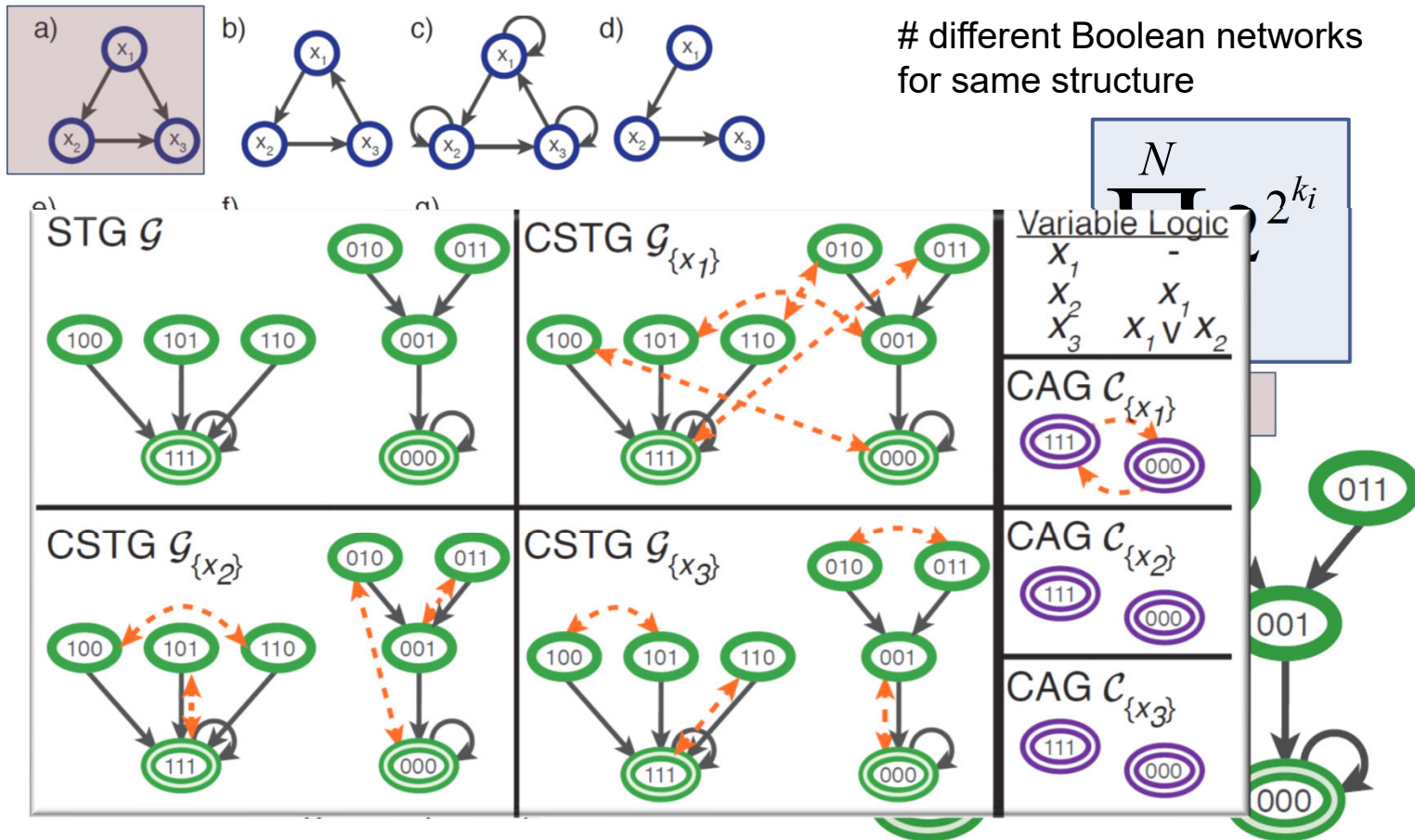
Variable Logic

x_1 -
 x_2 x_1
 x_3 $x_1 \vee x_2$

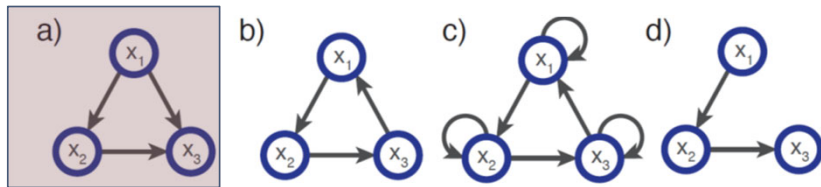
STG \mathcal{G}



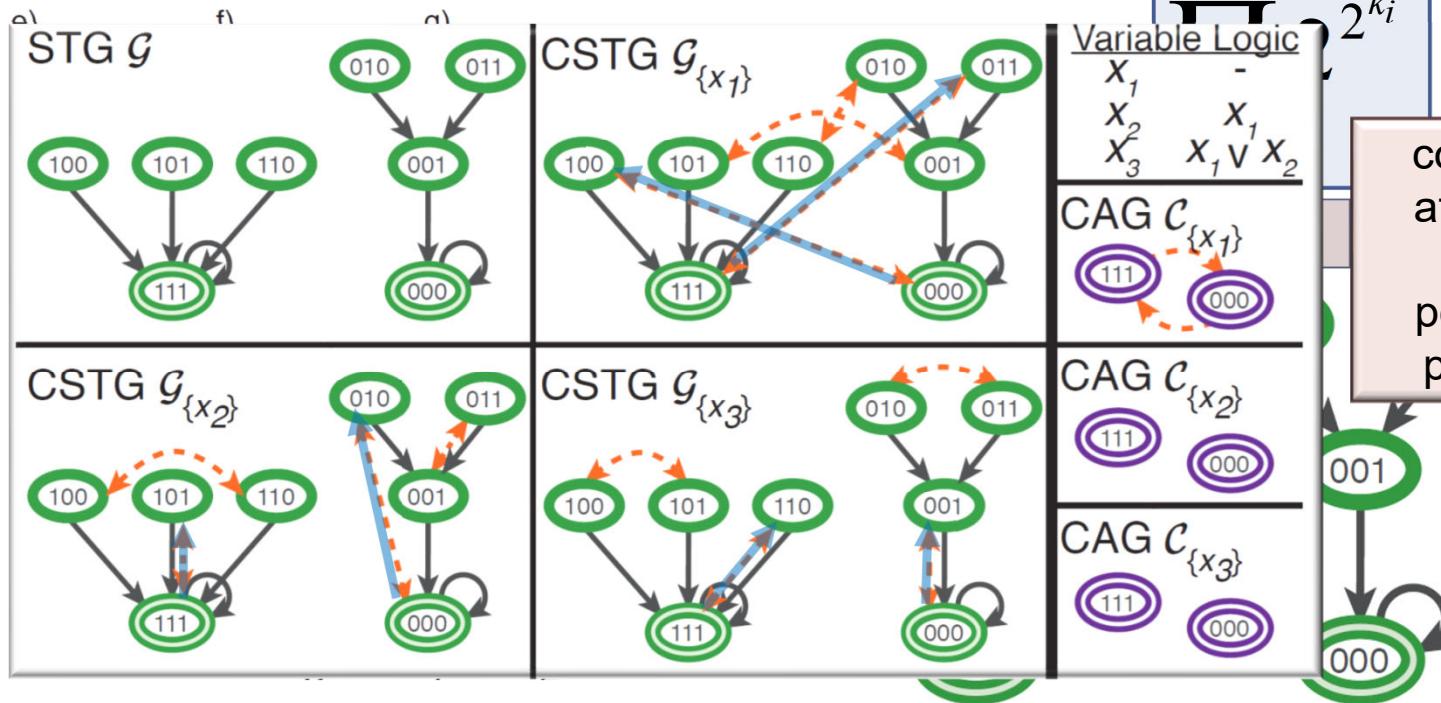
ensemble dynamics for same structure



ensemble dynamics for same structure



different Boolean networks for same structure

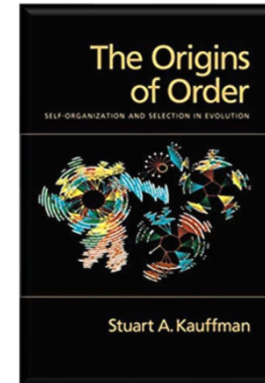


Of NK-Boolean Networks

- Random networks
 - Started with random initial conditions
 - Self-organization is not a result of special initial conditions
- Statistical analysis
 - $K \leq 2$
 - Steady state, ordered, crystallization
 - ($5 \leq K$ to) $K=N$
 - Disordered, chaotic
 - Mean length of cycles: $0.5 \times 2^{N/2}$
 - Mean number of cycles: N/e
 - High reachability, sensitive to perturbation
 - Number of other state cycles system can reach after perturbation
 - $K=2$
 - Mean length: $n^{1/2}$
 - Mean number of cycles: $n^{1/2}$
 - Low reachability
 - Percolation of frozen clusters (isolated subsets)
 - Not very sensitive to perturbation



Kauffman, S. A. (1984). *Phys. D Nonlinear Phen.* **10**,145–156.

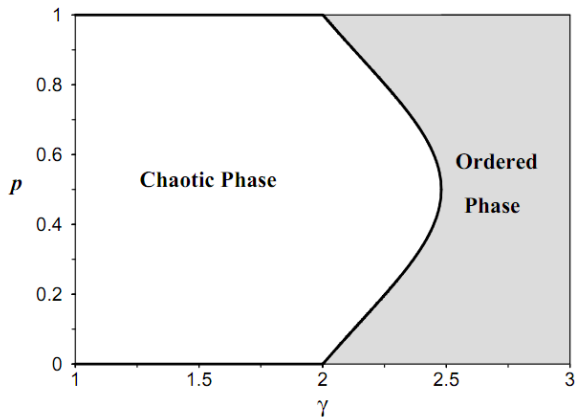


Kauffman, SA. *J. theoretical biology* **22.3** (1969): 437-467.

criticality

- $2 \leq K \leq 5$
 - Good for evolvability?
 - Some changes with large repercussions
 - Best capability to perform information exchange
 - Information can be propagated more easily
- Problems with analysis
 - Network topology is random
 - Not scale-free, as later explored by Aldana
 - Real genetic networks tend to have lower values of K (in ordered regime)
 - Genes as simply Boolean may be oversimplification
 - Though a few states can approximate very well continuous data

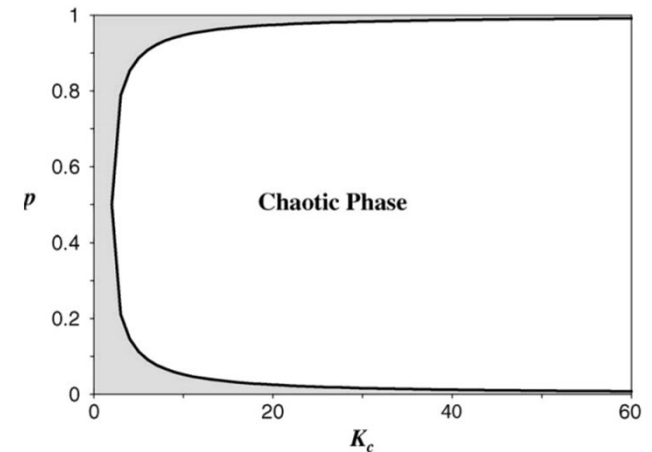
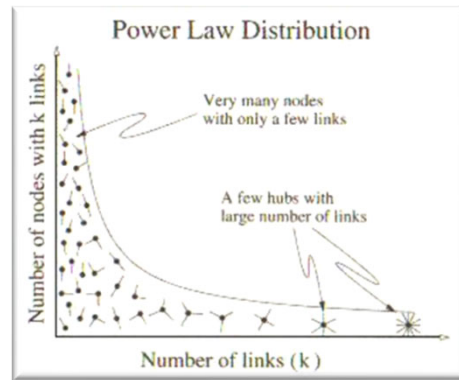
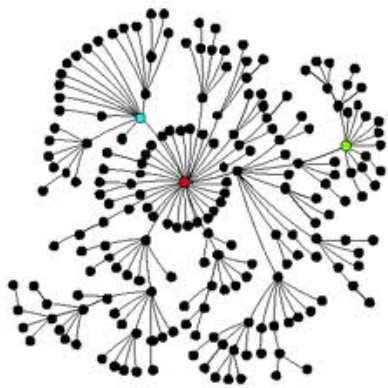
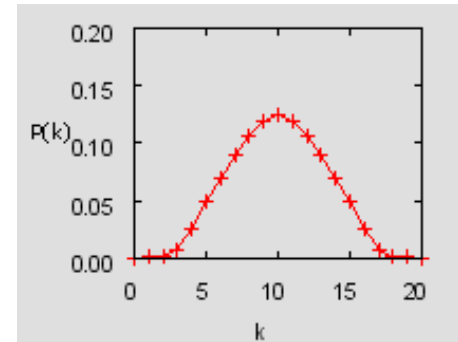
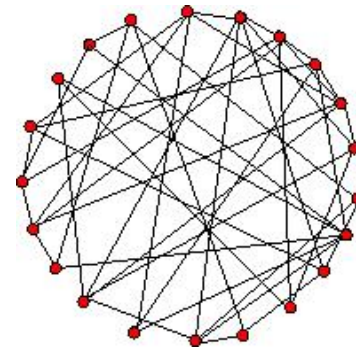
dynamical behavior of ensembles of networks



scale-free topology

$$P(k) = c \cdot k^{-\gamma}$$

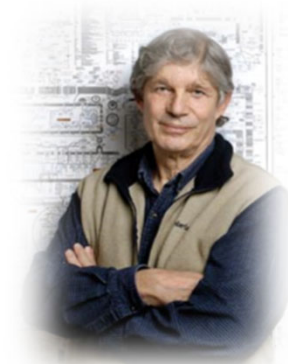
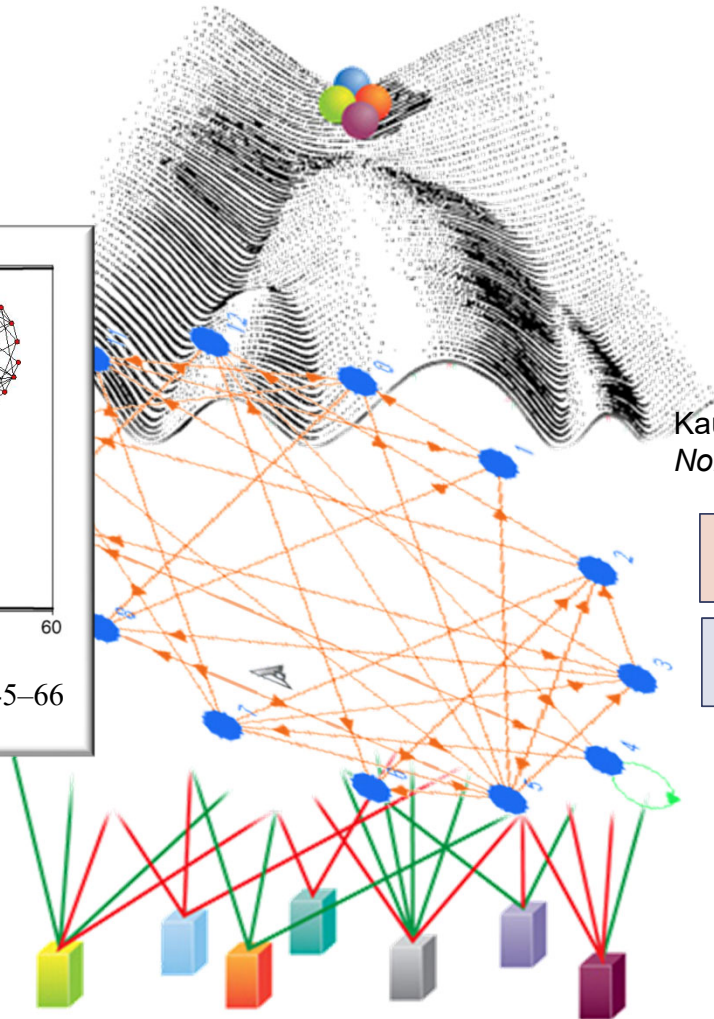
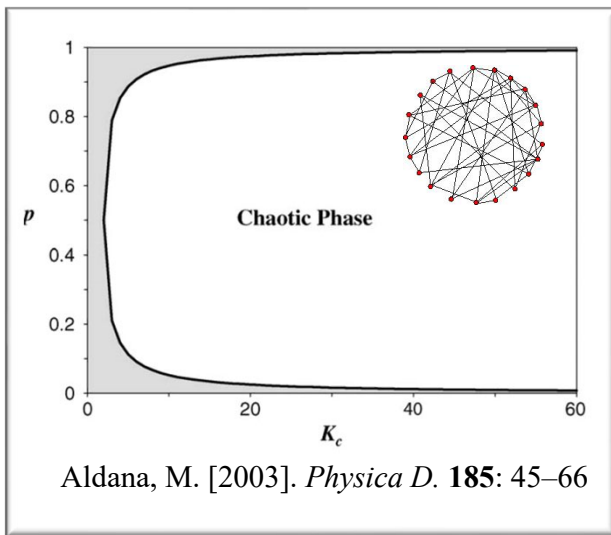
Random topology



Aldana, M. [2003]. *Physica D*. 185: 45–66

self-organization easily chaotic

evolution requires life in critical regime which is small, how come life is not chaotic?



Kauffman, S. A. (1984). *Phys. D Nonlinear Phen.* **10**, 145–156.



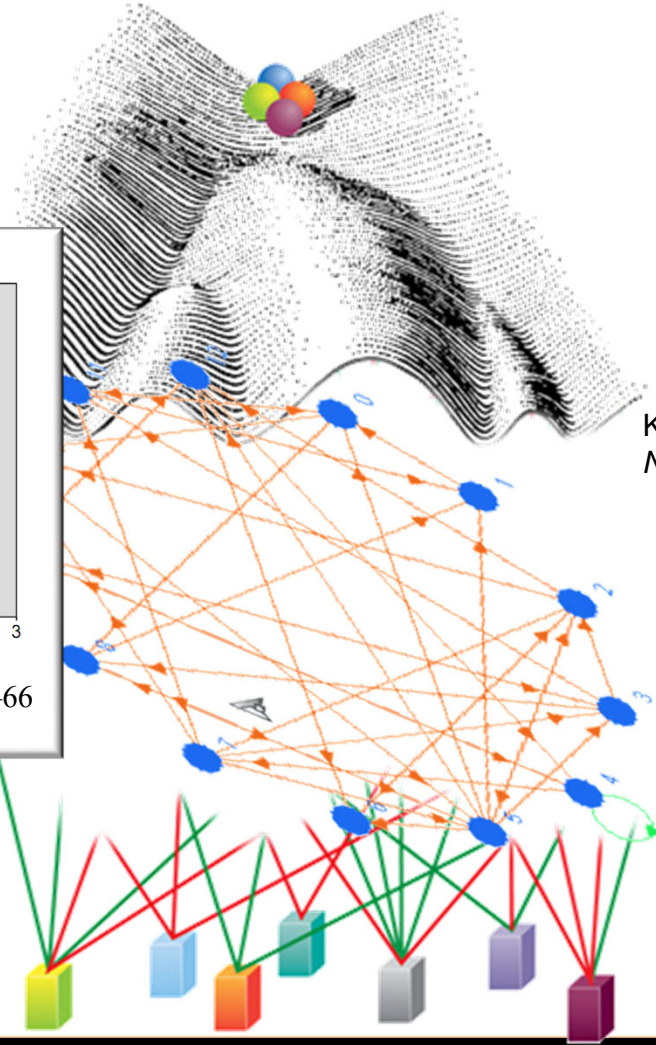
Waddington CH (1942). *Nature.* **150** (3811):563–565

robustness of phenotypes is the result of a **buffering** of the developmental process.

dynamics of gene networks provides buffering (**self-organization**). But still easily chaotic.

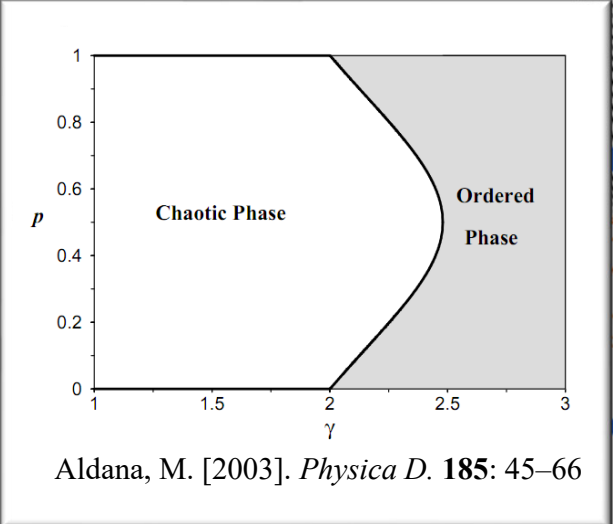
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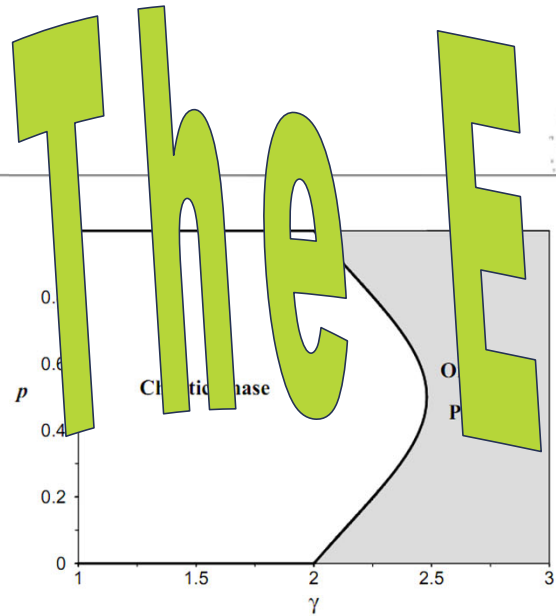
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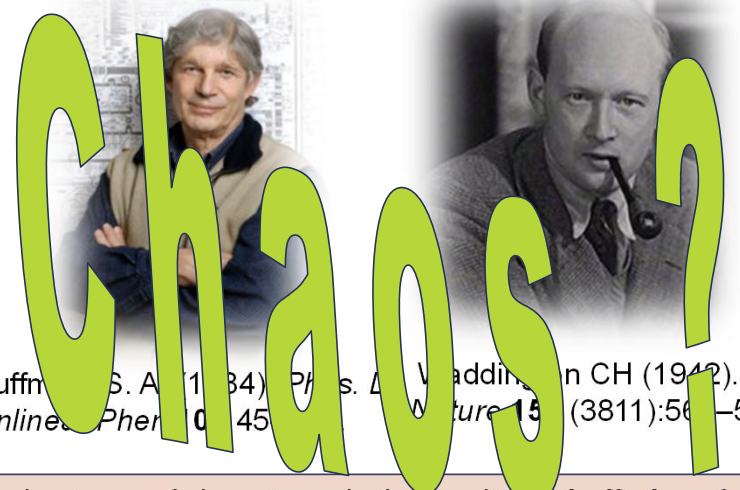
Structure (**topological organization**), can provide larger stable or critical universe, but still easily chaotic.

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Kauffman, S. A. (1984). *Phys. L.* Waddington, CH (1942). *Nonlinear Phen.* **0**: 45. *Nature* **15** (3811):561–565

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The Criticality Hypothesis

readings

■ Class Book

- Floreano, D. and C. Mattiussi [2008]. *Bio-Inspired Artificial Intelligence: Theories, Methods, and Technologies*. MIT Press.
 - Chapter 2.

■ Lecture notes

- Chapter 1: What is Life?
- Chapter 2: The logical Mechanisms of Life
- Chapter 3: Formalizing and Modeling the World
- Chapter 4: Self-Organization and Emergent Complex Behavior
 - posted online @ <http://informatics.indiana.edu/rocha/i-bic>

■ Papers and other materials

● Optional

- Nunes de Castro, Leandro [2006]. *Fundamentals of Natural Computing: Basic Concepts, Algorithms, and Applications*. Chapman & Hall.
 - Chapter 2, all sections
 - Chapter 7, sections 7.3 – Cellular Automata
 - Chapter 8, sections 8.1, 8.2, 8.3.10
- Flake's [1998], *The Computational Beauty of Life*. MIT Press.
 - Chapters 10, 11, 14 – Dynamics, Attractors and chaos

