



Cellular Automata and beyond ...

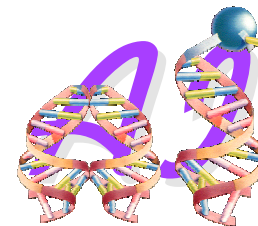
The World of Simple Programs

Christian Jacob

Department of Computer Science

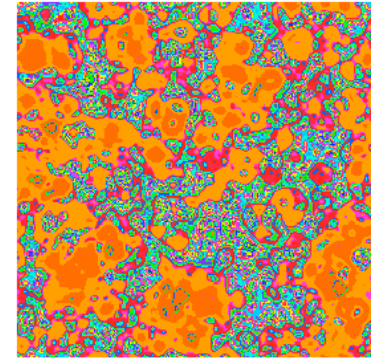
University of Calgary

CPSC 601.73 — Winter 2003

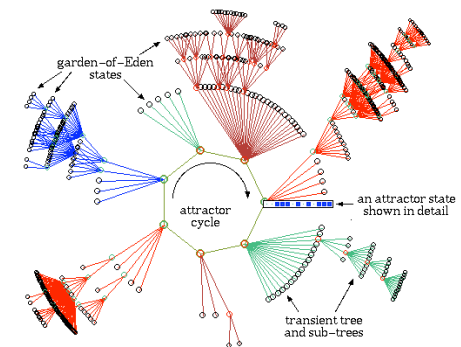


Cellular Automata

Lindenmayer
Systems



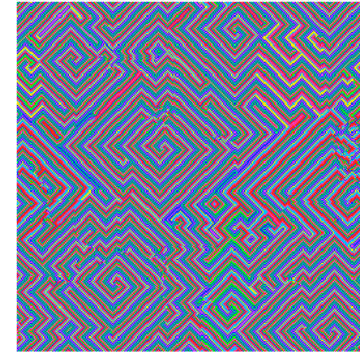
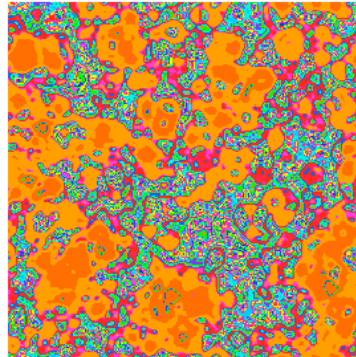
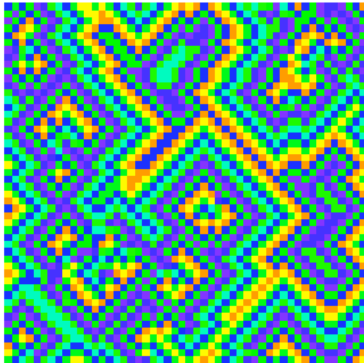
Random Boolean
Networks



Classifier Systems

Cellular Automata

Global Effects from Local Rules



Cellular Automata

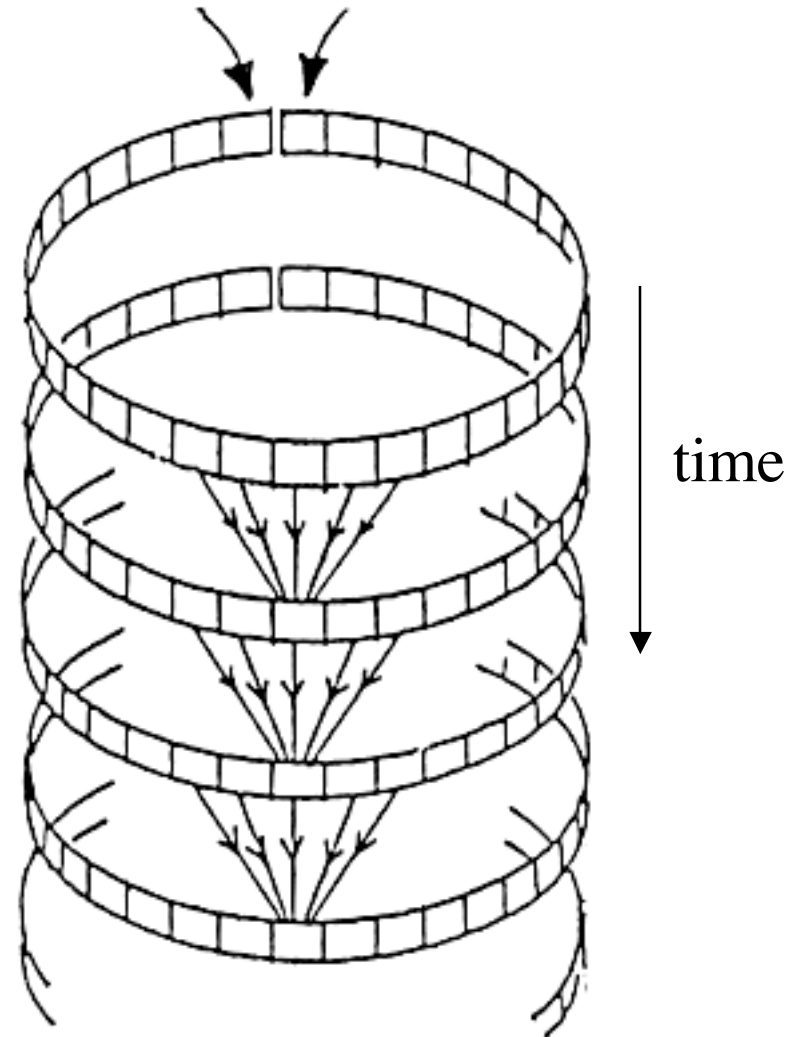
- The CA space is a lattice of cells (usually 1D, 2D, 3D) with a particular geometry.
- Each cell contains a variable from a limited range of values (e.g., 0 and 1).
- All cells update synchronously.
- All cells use the same updating rule (in uniform CA), depending only on local relations.
- Time advances in discrete steps.

One-dimensional Finite CA Architecture

- Neighbourhood size:
 $K = 5$

local connections
per cell

- Synchronous
update in discrete
time steps



A. Wuensche: The Ghost in the Machine, Artificial Life III, 1994.

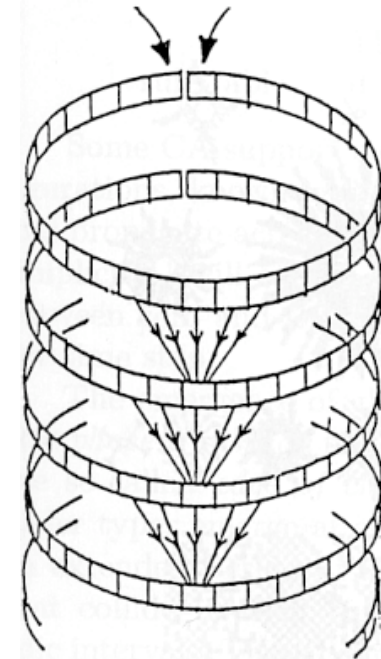
Time Evolution of Cell i with K -Neighbourhood

$$C_i^{(t+1)} = f(C_{i-[K/2]}^{(t)}, \dots, C_{i-1}^{(t)}, C_i^{(t)}, C_{i+1}^{(t)}, \dots, C_{i+[K/2]}^{(t)})$$

With periodic boundary conditions:

$$x < 1 : C_x = C_{N+x}$$

$$x > N : C_x = C_{x-N}$$



Value Range and Update Rules

- For V different states (= values) per cell there are V^K permutations of values in a neighbourhood of size K .
- The update function f can be implemented as a lookup table with V^K entries, giving V^{V^K} possible rules.

V^K {

00000: 1 ... V

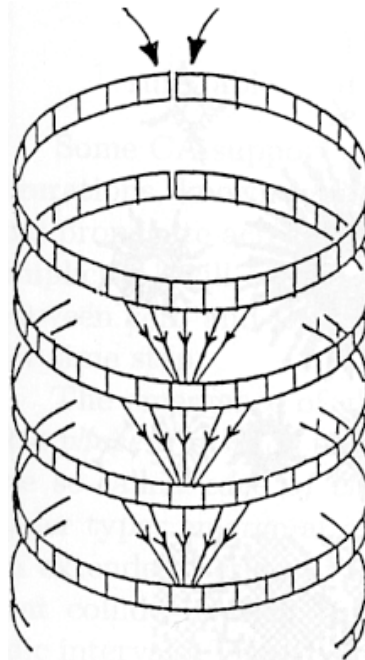
00001: _

00010: _

...

11110: _

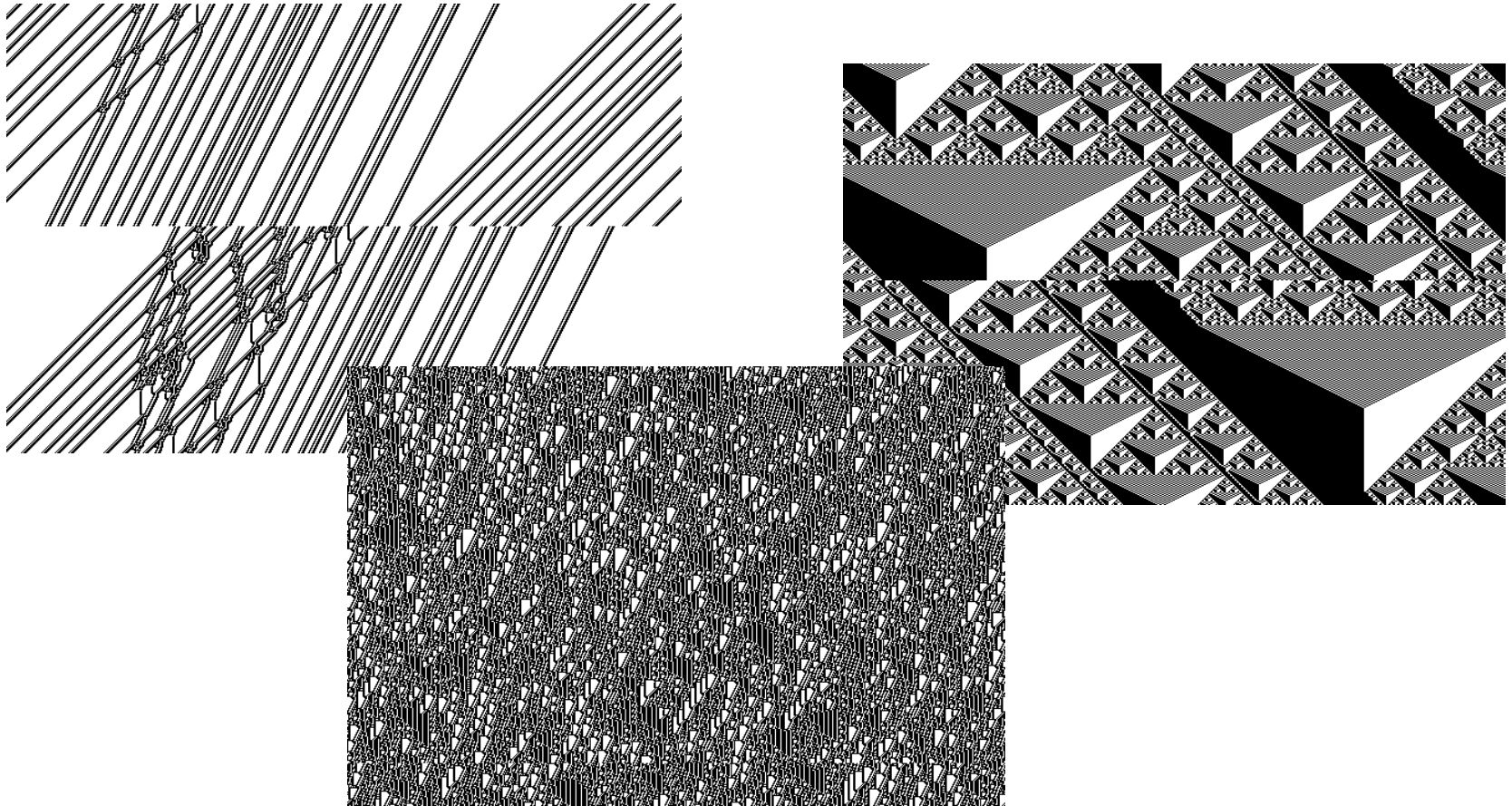
11111: _



v	K	v^K	V^{v^K}
2	3	8	256
2	5	32	4.3×10^9
2	7	128	3.4×10^{38}
2	9	512	1.3×10^{154}

Cellular Automata: Local Rules — Global Effects

Demos



History of Cellular Automata

- Alternative names:
 - Tessellation automata
 - Cellular spaces
 - Iterative automata
 - Homogeneous structures
 - Universal spaces
- John von Neumann (1947)
 - Tries to develop abstract model of self-reproduction in biology (from investigations in cybernetics; Norbert Wiener)
- J. von Neumann & Stanislaw Ulam (1951)
 - 2D self-reproducing cellular automaton
 - 29 states per cell
 - Complicated rules
 - 200,000 cell configuration
 - (Details filled in by Arthur Burks in 1960s.)

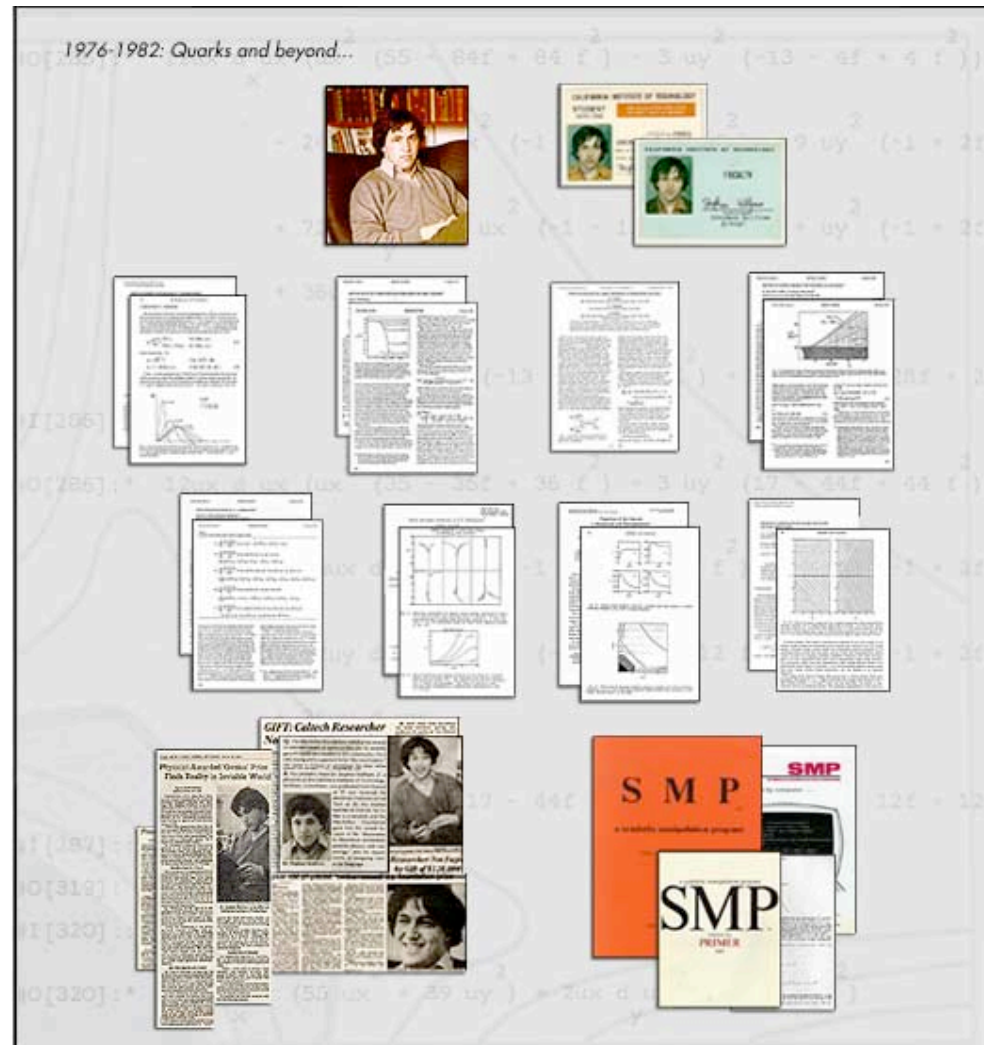
History of Cellular Automata (2)

- Threads emerging from J. von Neumann's work:
 - Self-reproducing automata (spacecraft!)
 - Mathematical studies of the essence of
 - Self-reproduction and
 - Universal computation.
- CAs as Parallel Computers (end of 1950s / 1960s)
 - Theorems about CAs (analogies to Turing machines) and their formal computational capabilities
 - Connecting CAs to mathematical discussions of dynamical systems (e.g., fluid dynamics, gases, multi-particle systems)
- 1D and 2D CAs used in electronic devices (1950s)
 - Digital image processing (with so-called cellular logic systems)
 - Optical character recognition
 - Microscopic particle counting
 - Noise removal

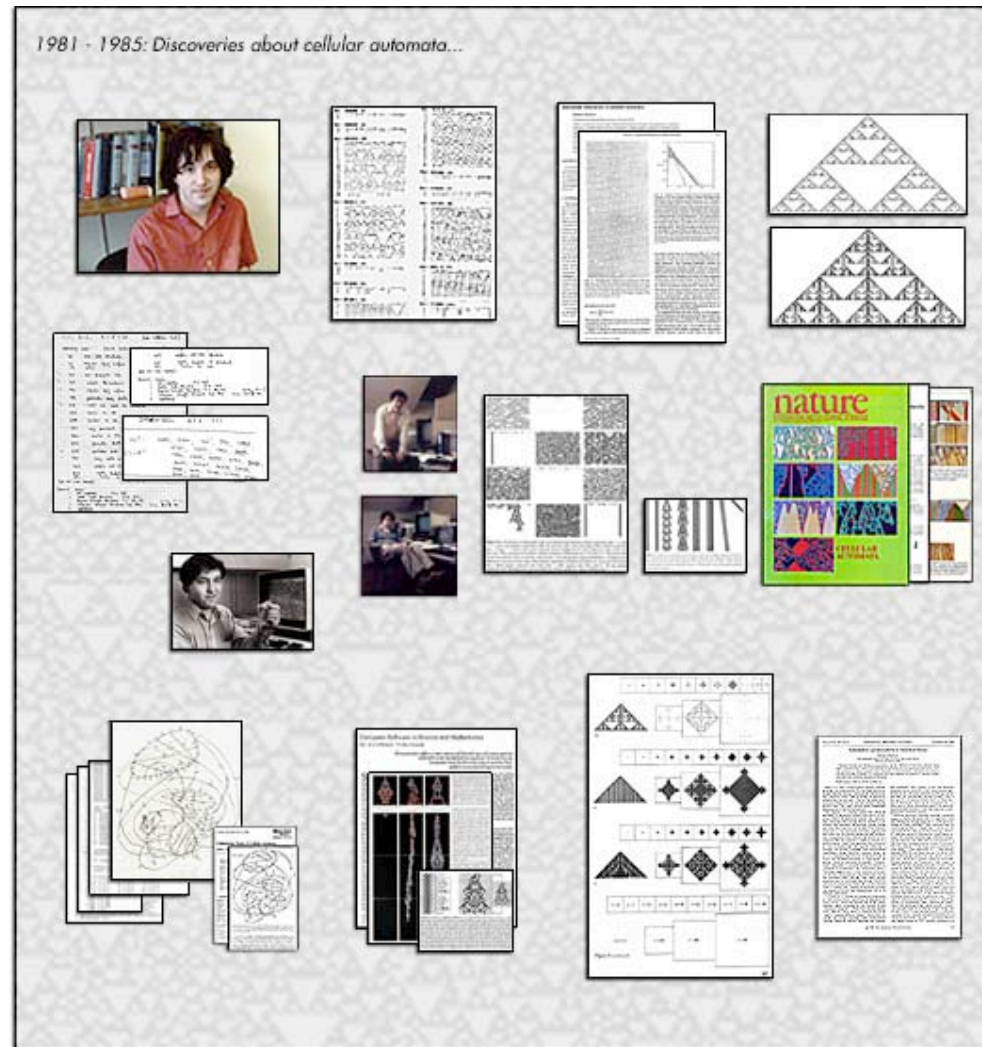
History of Cellular Automata (3)

- Stanislaw Ulam at Los Alamos Laboratories
 - 2D cellular automata to produce recursively defined geometrical objects (evolution from a single black cell)
 - Explorations of simple growth rules
- Specific types of Cas (1950s/60s)
 - 1D: optimization of circuits for arithmetic and other operations
 - 2D:
 - Neural networks with neuron cells arranged on a grid
 - Active media: reaction-diffusion processes
- John Horton Conway (1970s)
 - Game of Life (on a 2D grid)
 - Popularized by Martin Gardner: *Scientific American*

Stephen Wolfram's World of CAs



Stephen Wolfram's World of CAs



Stephen Wolfram's World of CAs



Stephen Wolfram's World of CAs



Example Update Rule

- $V = 2, K = 3$
- The rule table for rule 30:

111 110 101 100 011 010 001 000



0 0 0 1 1 1 1 0

128 64 32 16 8 4 2 1

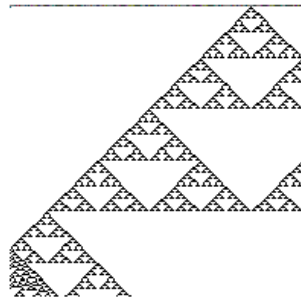
$$16 + 8 + 4 + 2 = 30$$

See [examples](#) ...

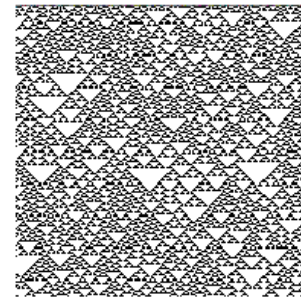
CA Demos

- *Evolvica* CA [Notebooks](#)

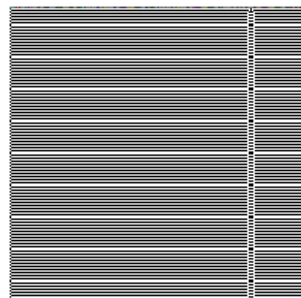
22: {0, 0, 0, 1, 0, 1, 1, 0}



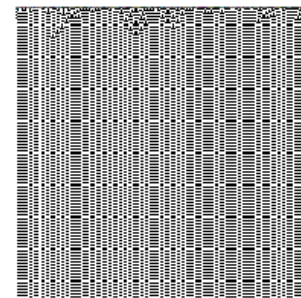
22: {0, 0, 0, 1, 0, 1, 1, 0}



37: {0, 0, 1, 0, 0, 1, 0, 1}

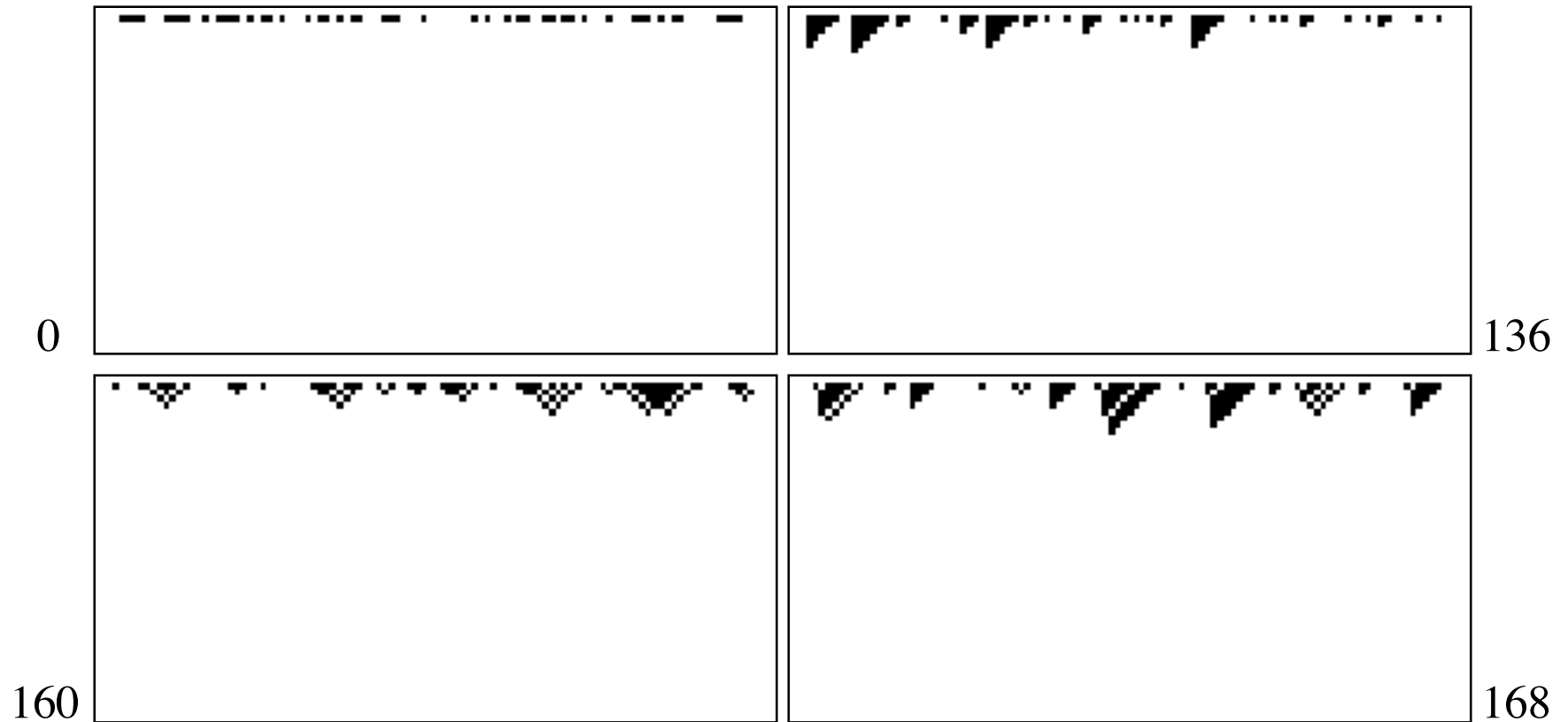


37: {0, 0, 1, 0, 0, 1, 0, 1}



Four Wolfram Classes of CA

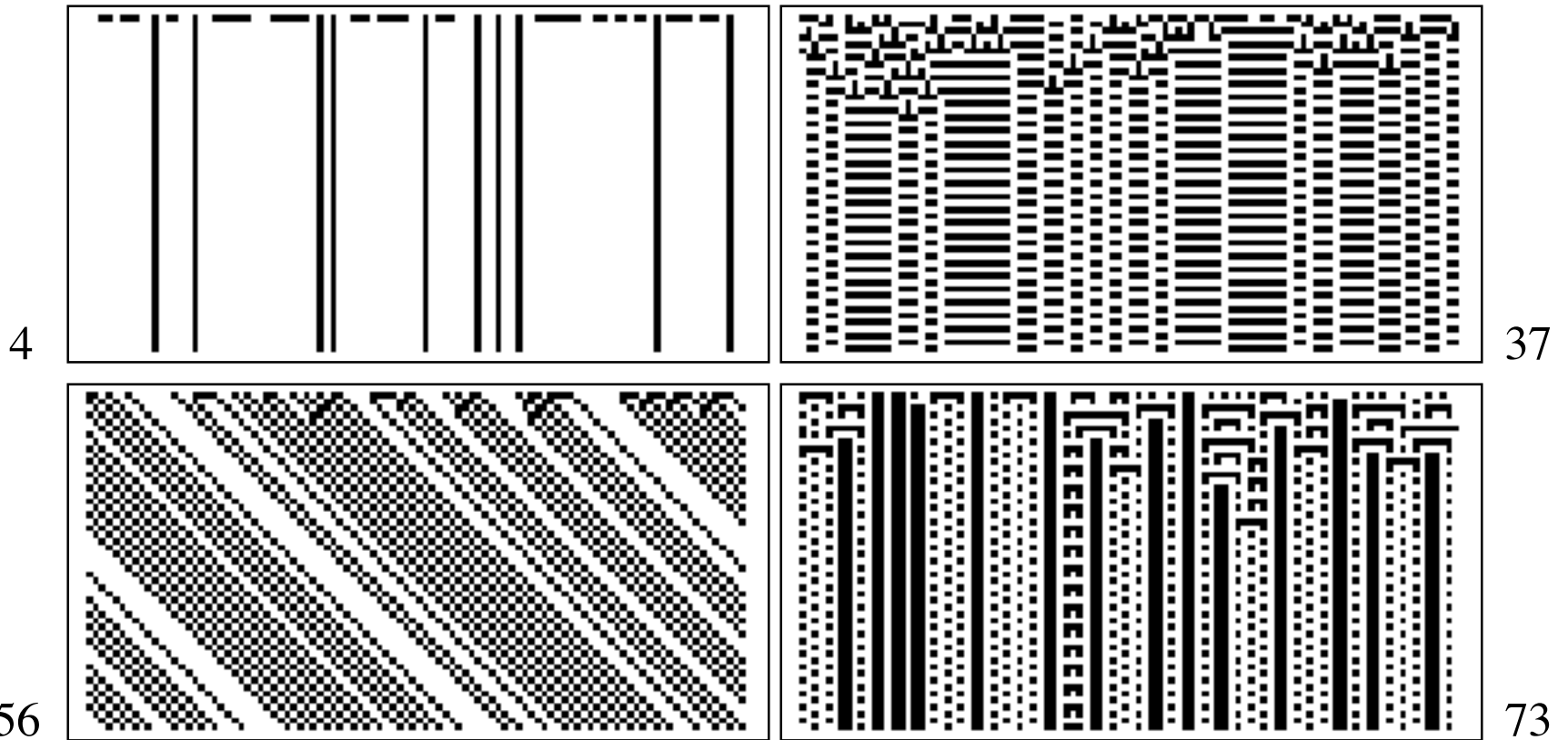
- **Class 1:**
A fixed, homogeneous, state is eventually reached
(e.g., rules 0, 8, 128, 136, 160, 168).



Four Wolfram Classes of CA

- **Class 2:**

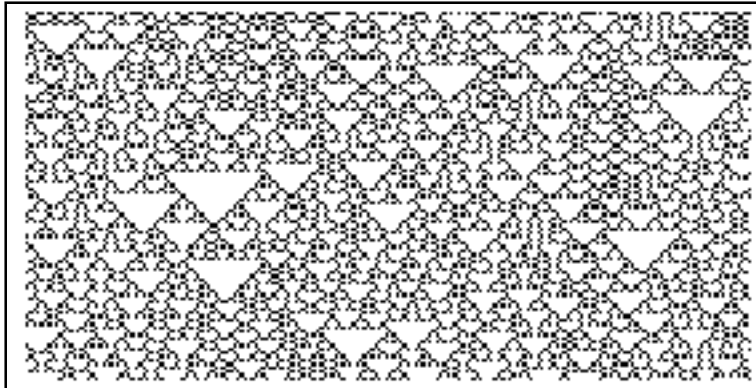
A pattern consisting of separated periodic regions is produced (e.g., rules 4, 37, 56, 73).



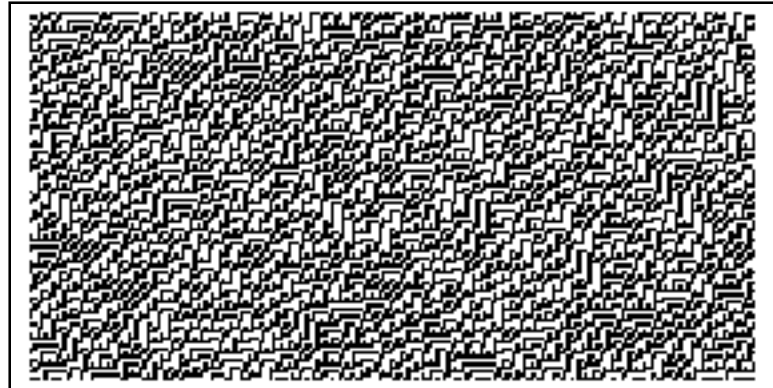
Four Wolfram Classes of CA

- **Class 3:**
A chaotic, aperiodic, pattern is produced
(e.g., rules 18, 45, 105, 126).

18



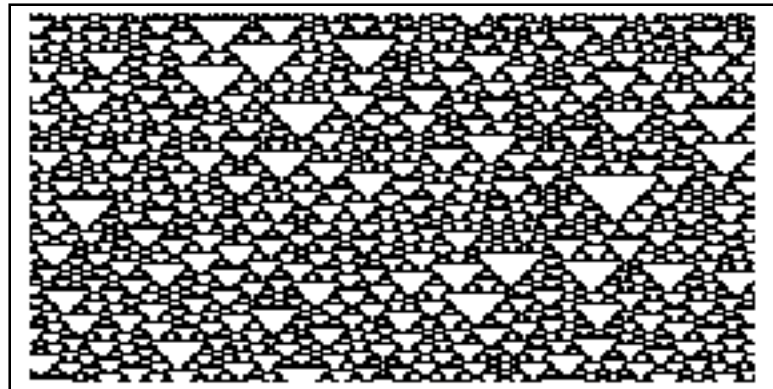
45



105



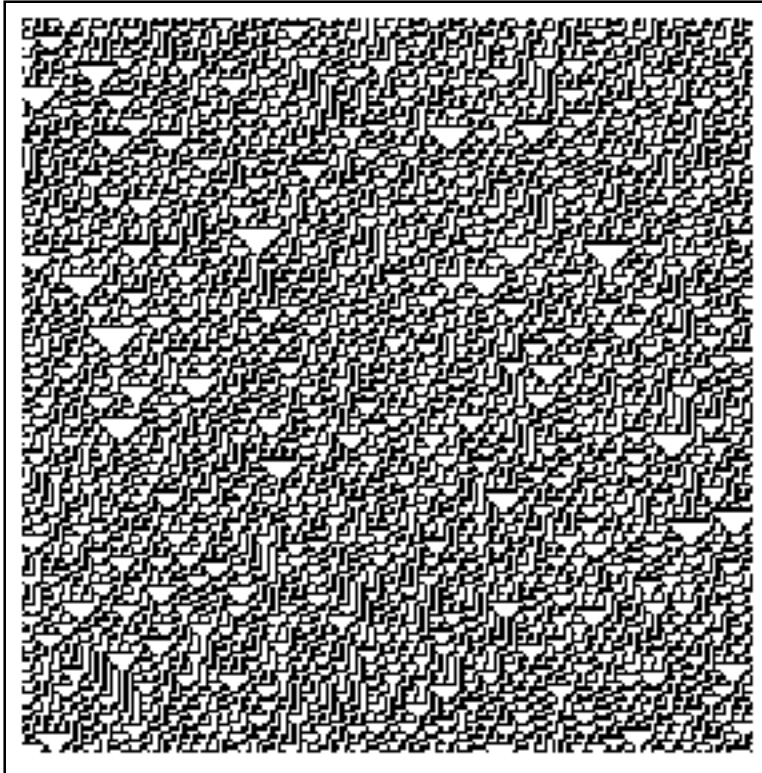
126



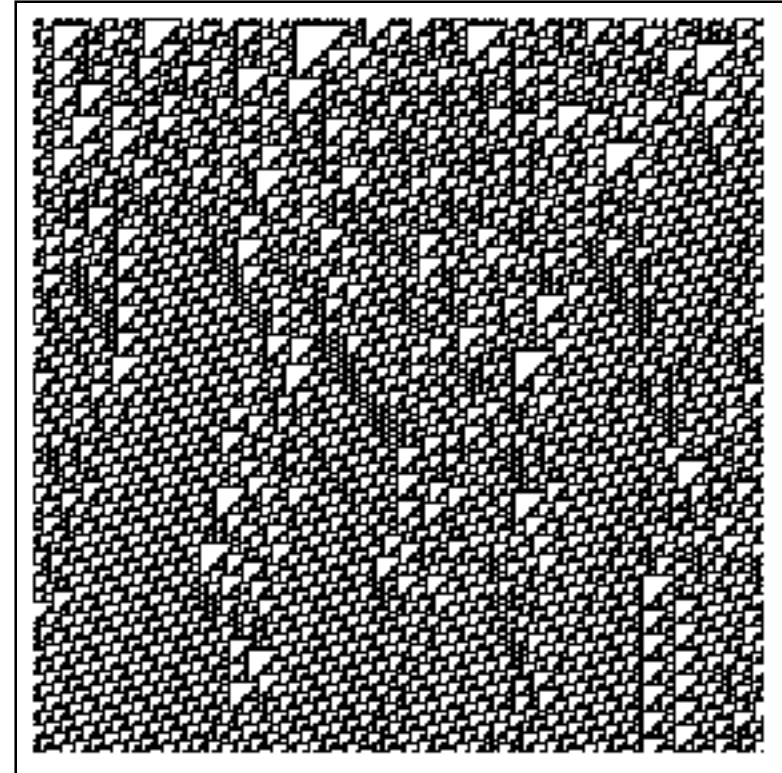
Four Wolfram Classes of CA

- **Class 4:**
Complex, localized structures are generated
(e.g., rules 30, 110).

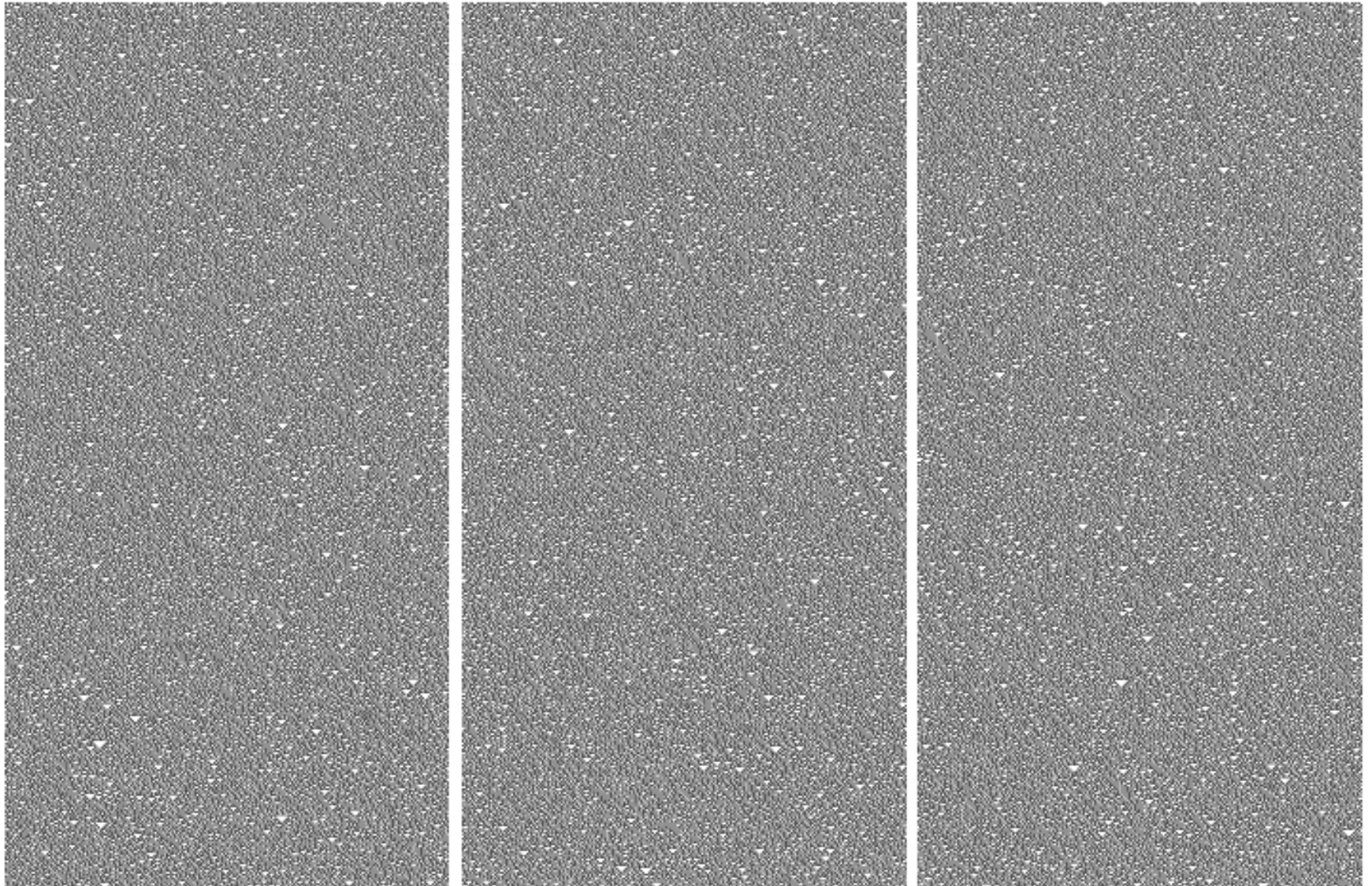
30



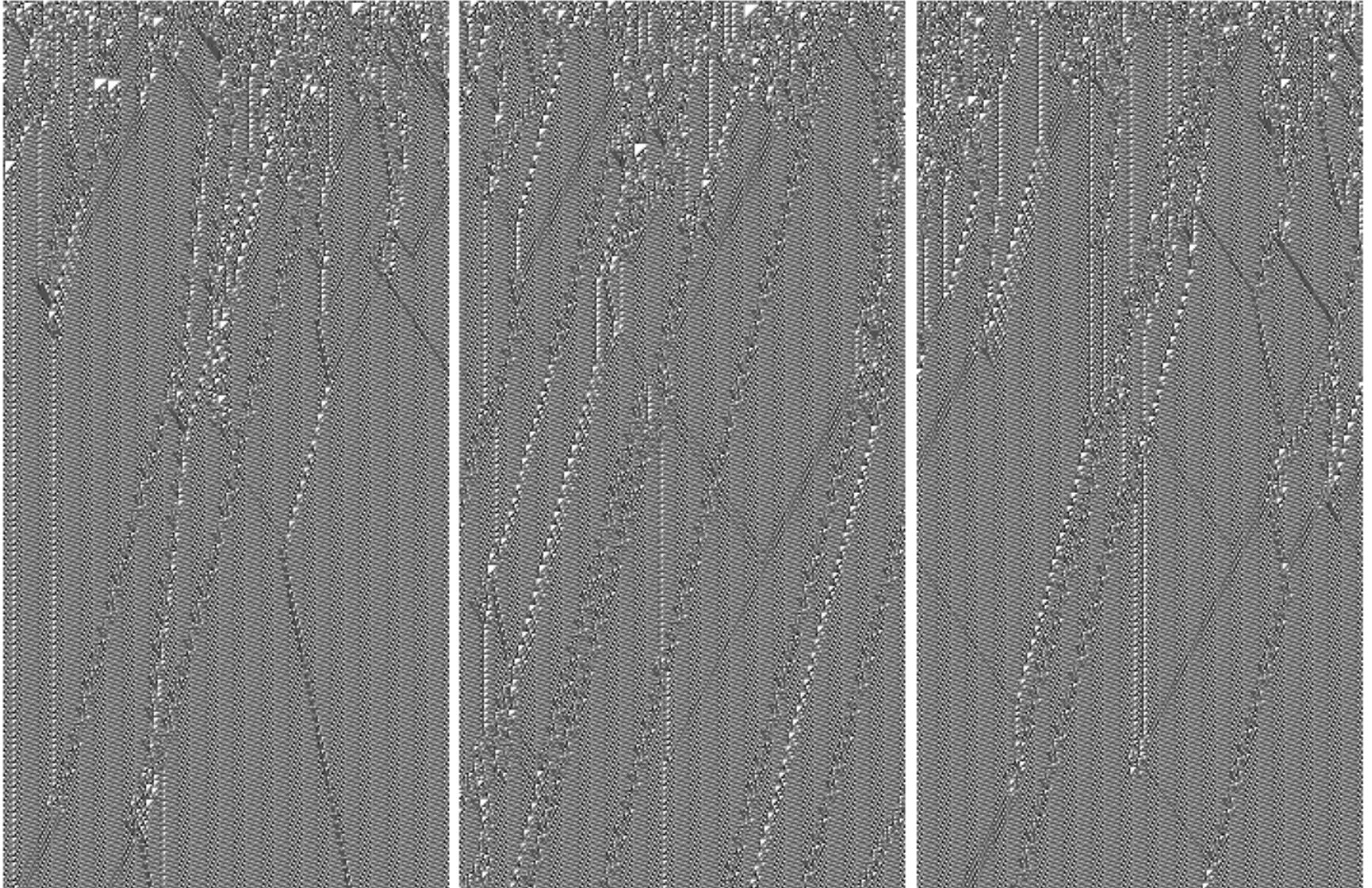
110



Class 4: Rule 30



Class 4: Rule 110



Further Classifications of CA Evolution

- Wolfram classifies CAs according to the patterns they evolve:

- 1. Pattern disappears with time.
- 2. Pattern evolves to a fixed finite size.
- 3. Pattern grows indefinitely at a fixed speed.
- 4. Pattern grows and contracts irregularly.

–3/text.html: Fig. 1

- Qualitative Classes

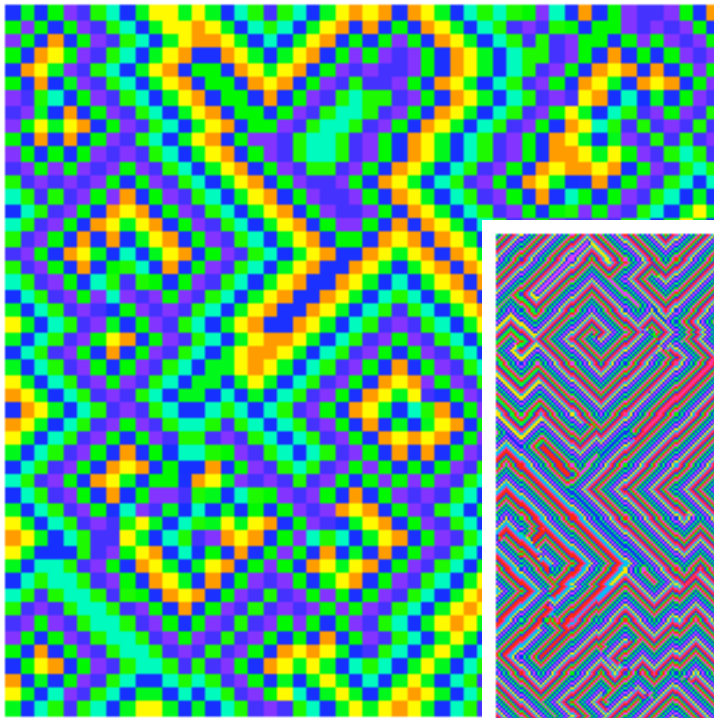
- 1. Spatially homogeneous state
- 2. Sequence of simple stable or periodic structures
- 3. Chaotic aperiodic behaviour
- 4. Complicated localized structures, some propagating

–85-cellular/7/text.html: Fig. 3 (first row)

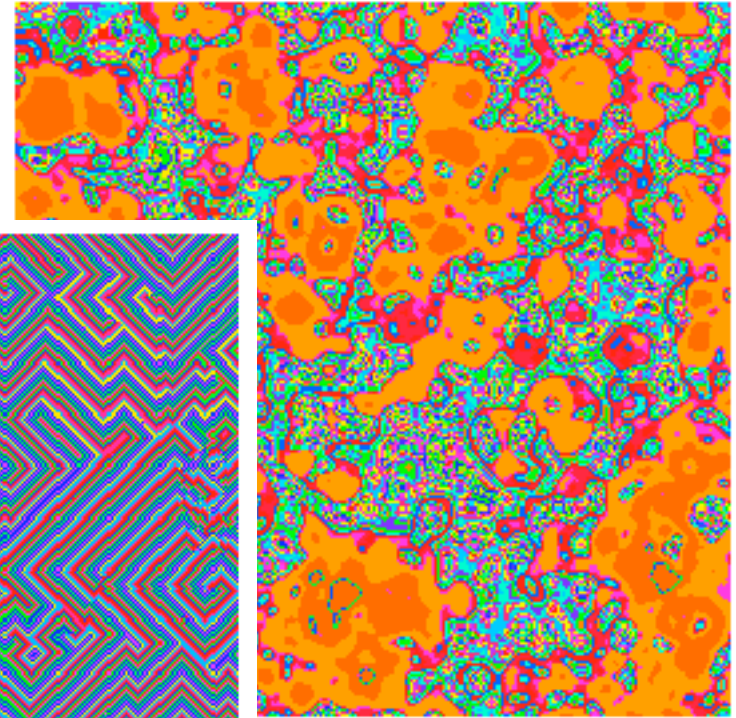
Further Classifications of CA Evolution (2)

- Classes from an Information Propagation Perspective
 - 1. No change in final state
 - 2. Changes only in a finite region
 - 3. Changes over an ever-increasing region
 - 4. Irregular changes
- Degrees of Predictability for the Outcome of the CA Evolution
 - 1. Entirely predictable, independent of initial state
 - 2. Local behavior predictable from local initial state
 - 3. Behavior depends on an ever-increasing initial region
 - 4. Behavior effectively unpredictable

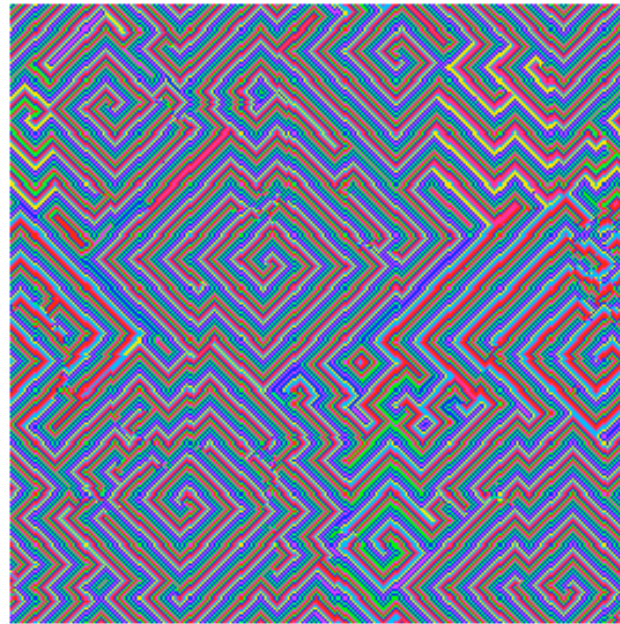
2-D CA: Emergent Pattern Formation in Excitable Media



Neuron excitation



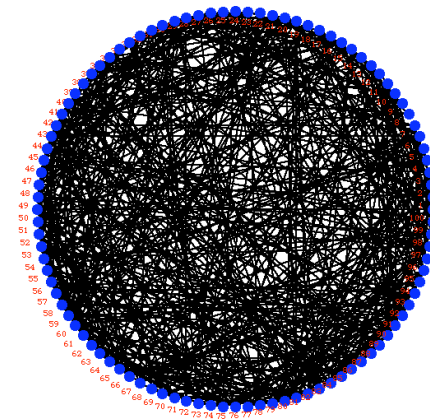
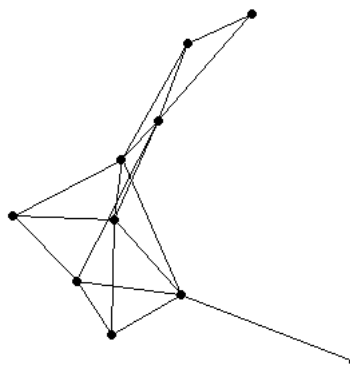
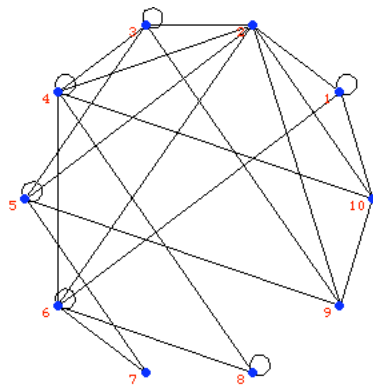
Hodgepodge



Neuron excitation (relaxed)

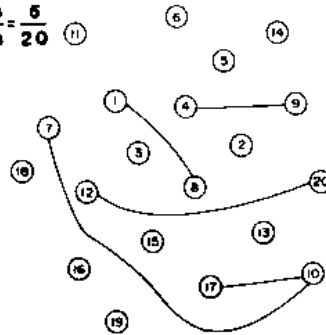
Random Boolean Networks

Generalized Cellular Automata

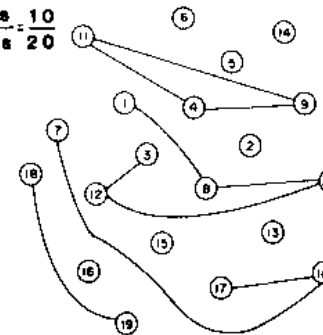


Crystallization of Connected Webs

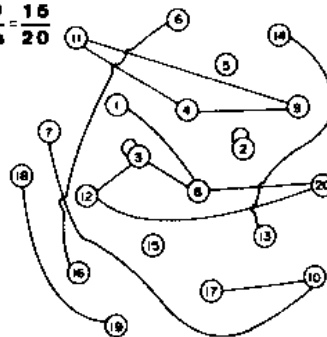
Edges = 5
Nodes = 20



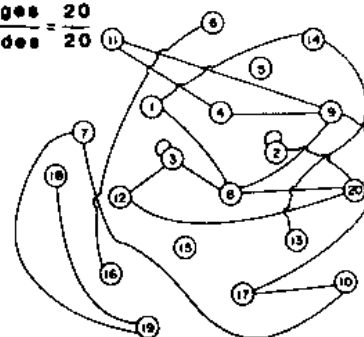
Edges = 10
Nodes = 20



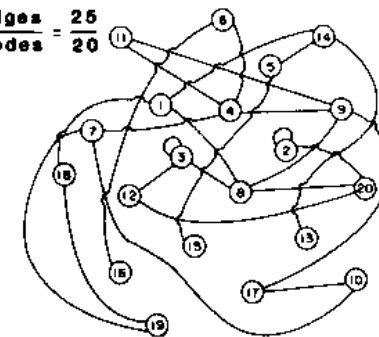
Edges = 15
Nodes = 20



Edges = 20
Nodes = 20



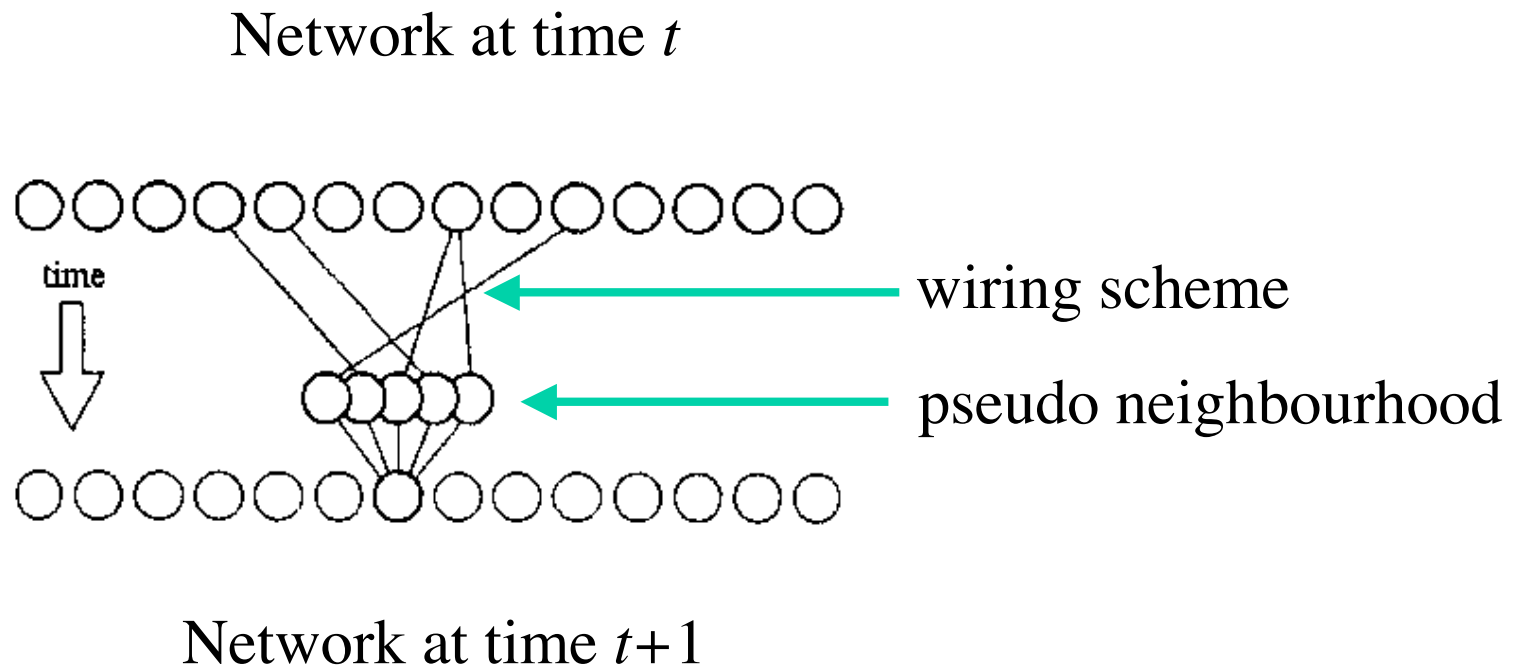
Edges = 25
Nodes = 20



[S. Kauffman: At Home in the Universe]

Random Nets Demo

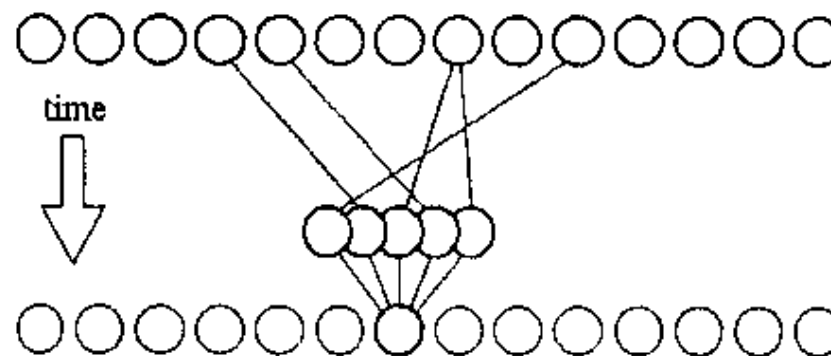
Random Network Architecture



Time Evolution of the i -th Cell

- Cell i is connected to K cells $w_{i1}, w_{i2}, \dots, w_{iK}$; with w_{ij} from $\{1, \dots, N\}$.
- N^K possible alternative wiring options.
- Update rule for cell i :

$$C_i^{(t+1)} = f_i(C_{w_{i1}}^{(t)}, C_{w_{i2}}^{(t)}, \dots, C_{w_{iK}}^{(t)})$$



Wiring/Rule Schemes

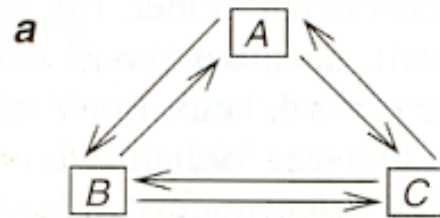
- A random network of size N with neighbourhood size K can be assigned

$$S = (N^K)^N \square (V^{V^K})^N$$

alternative wiring and rule schemes.

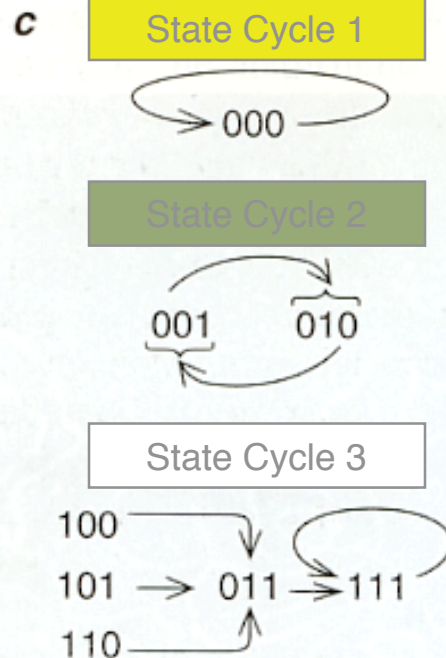
- Example:
 $V = 2, N = 16, K = 15; S = 2^{832}$.

States and Cycles



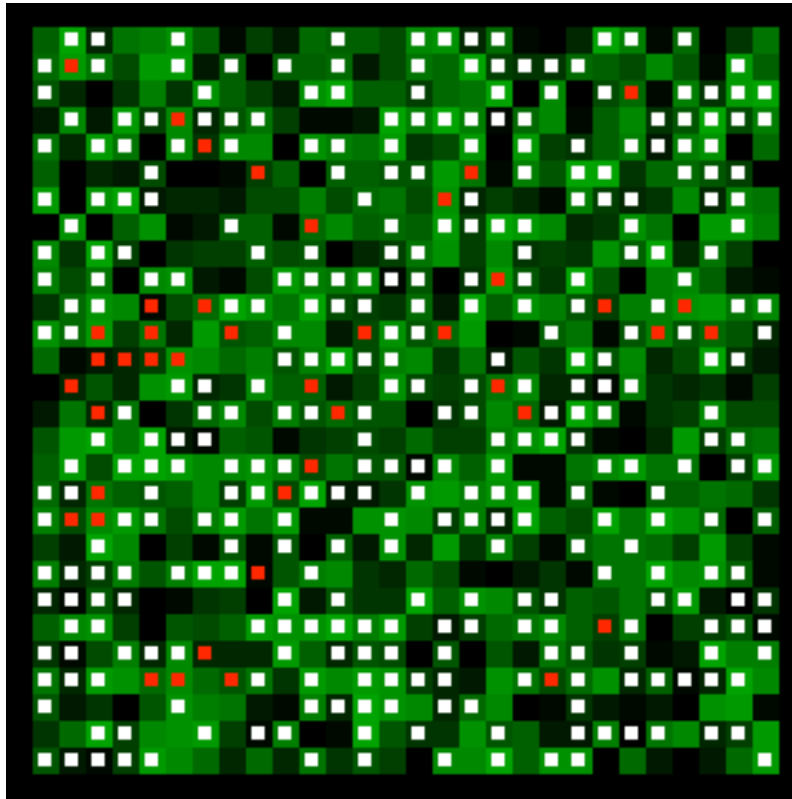
b

System State			Following State		
A	B	C	A	B	C
0	0	0	0	0	0
0	0	1	0	1	0
0	1	0	0	0	1
1	0	0	0	1	1
1	0	1	0	1	1
1	1	0	0	1	1
0	1	1	1	1	1
1	1	1	1	1	1

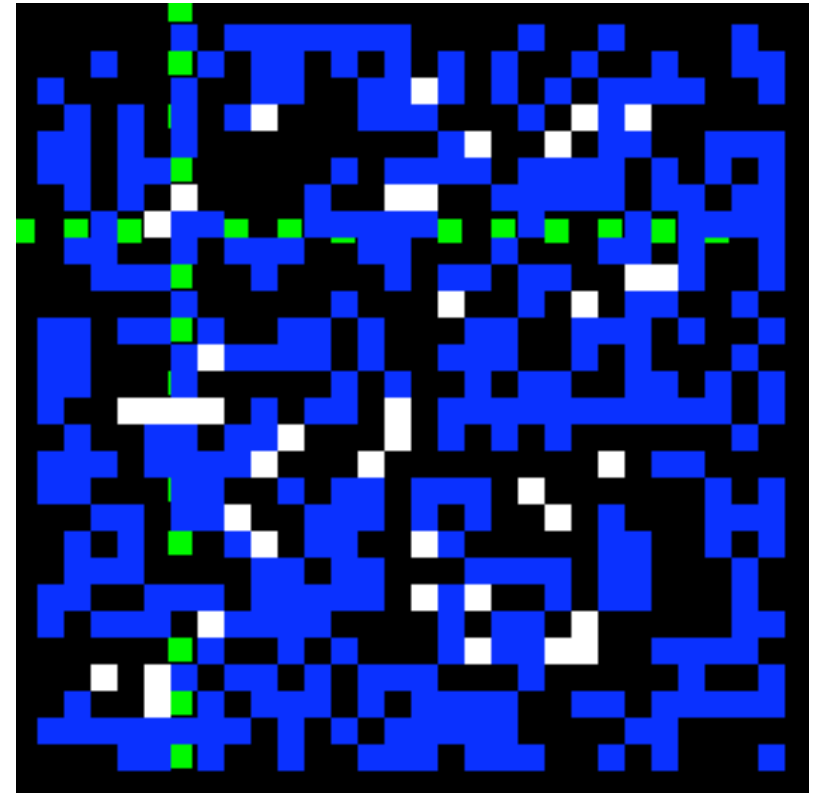


[S. Kauffman: Leben am Rande des Chaos]

Kauffman's Random Boolean Networks



Boolean functions represented by shades of green.
Stuart Kauffman used this network to investigate the
interaction of proteins within living systems.

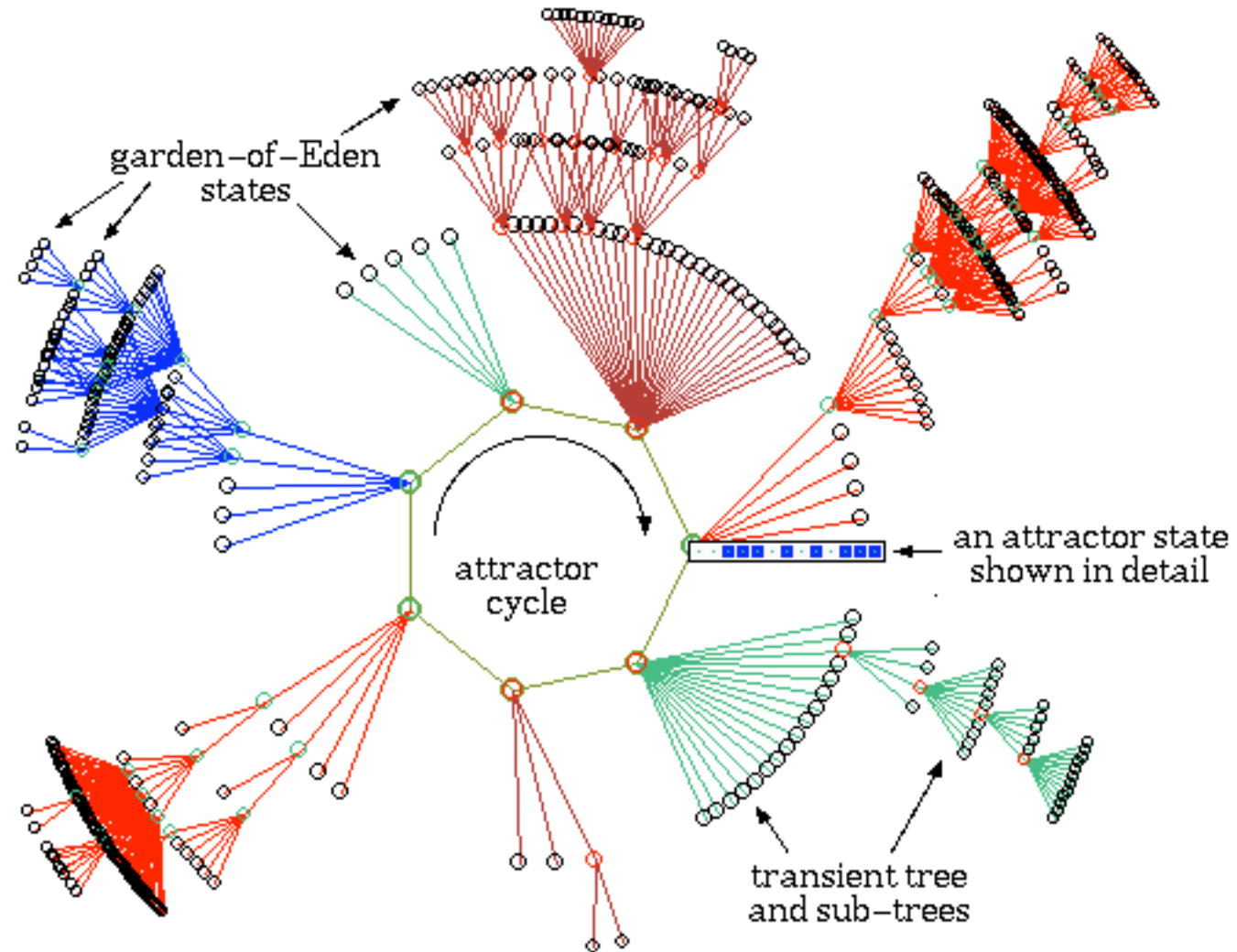


Binary values that have changed are white.
Unchanged values are blue.

These networks settle very quickly into an oscillatory
state.

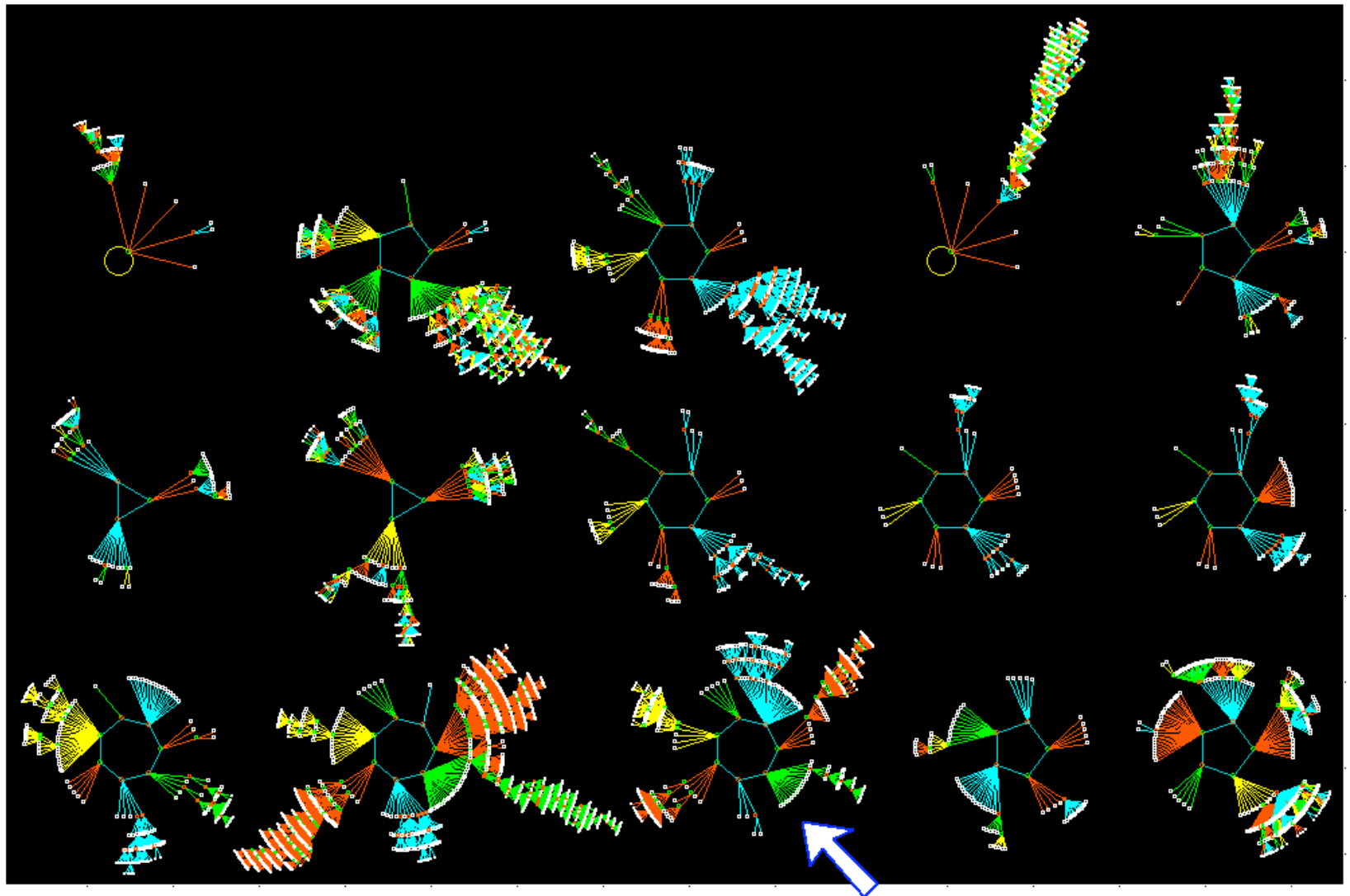
http://members.rogers.com/fmobrien/experiments/boolean_net/BooleanNetworkApplet_both.html

Attractor Cycles



[A. Wuensche, Discrete Dynamics Lab]

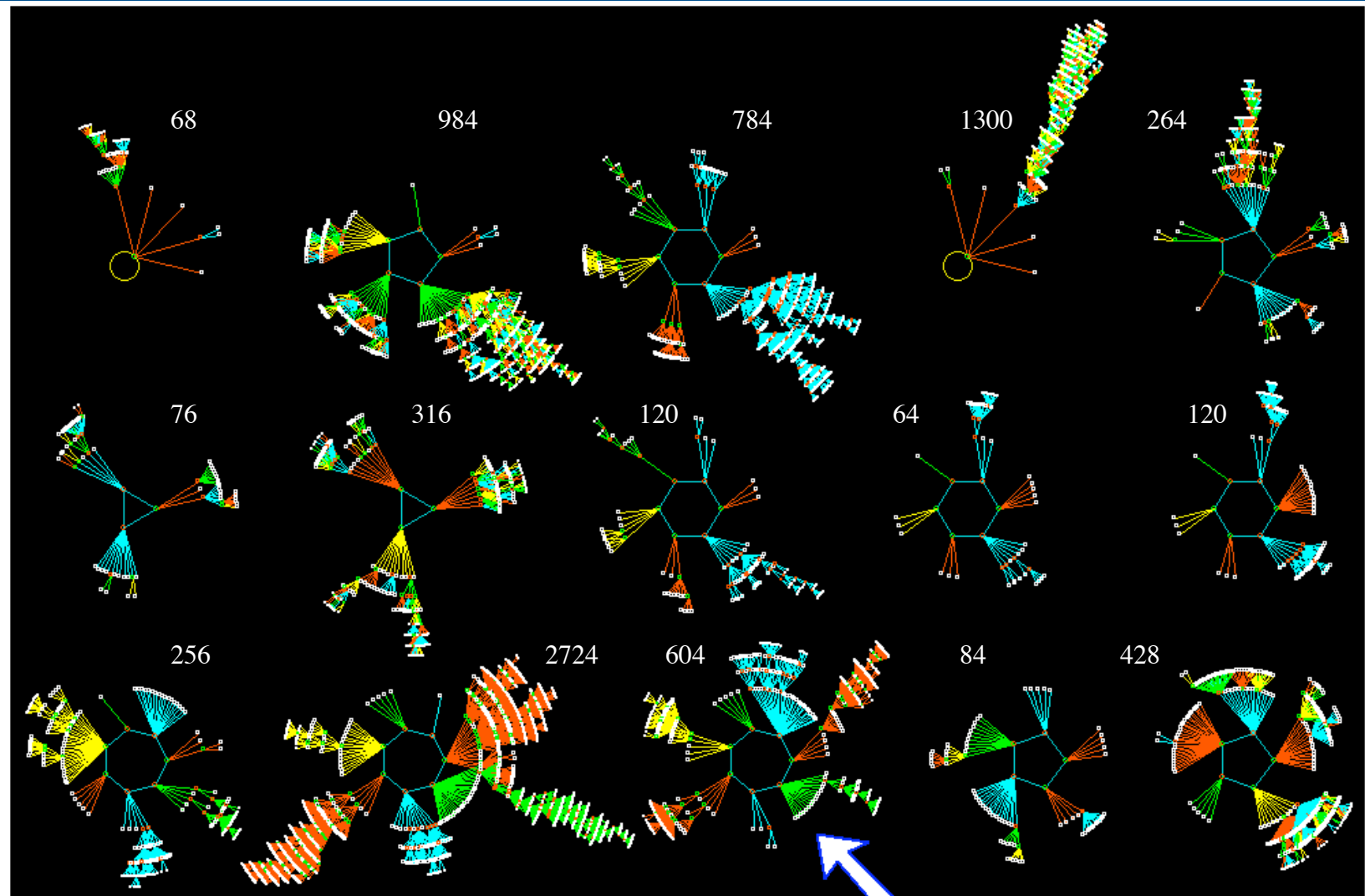
Basin of Attraction Field



Nodes: $n = 13$; Connectivity: $k = 3$; States: $2^{13} = 8192$

[A. Wuensche, Discrete Dynamics Lab]

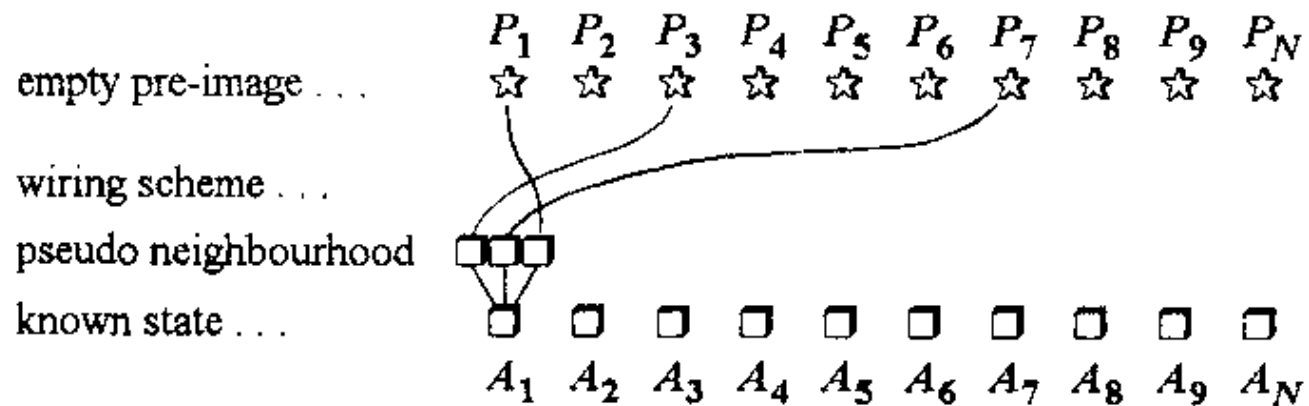
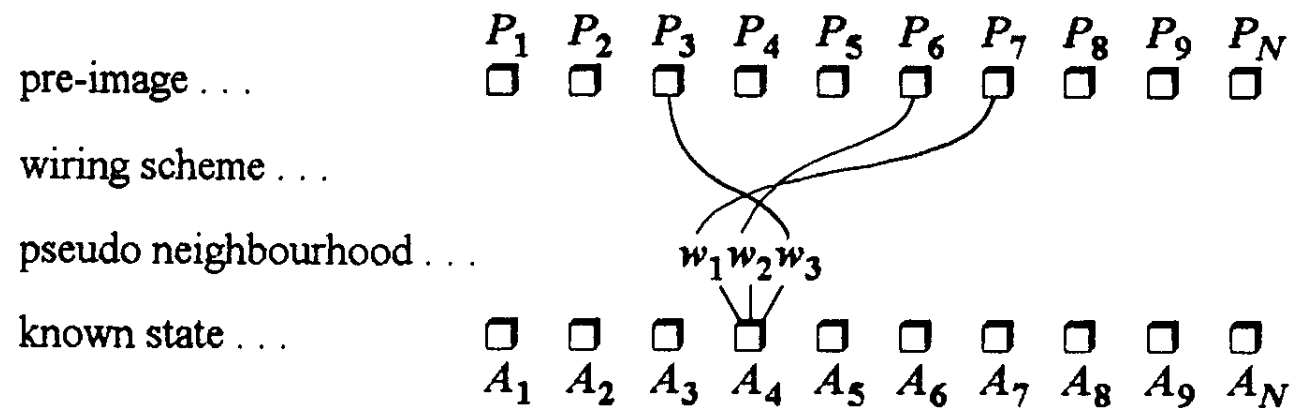
Basin of Attraction Field



Nodes: $n = 13$; Connectivity: $k = 3$; States: $2^{13} = 8192$

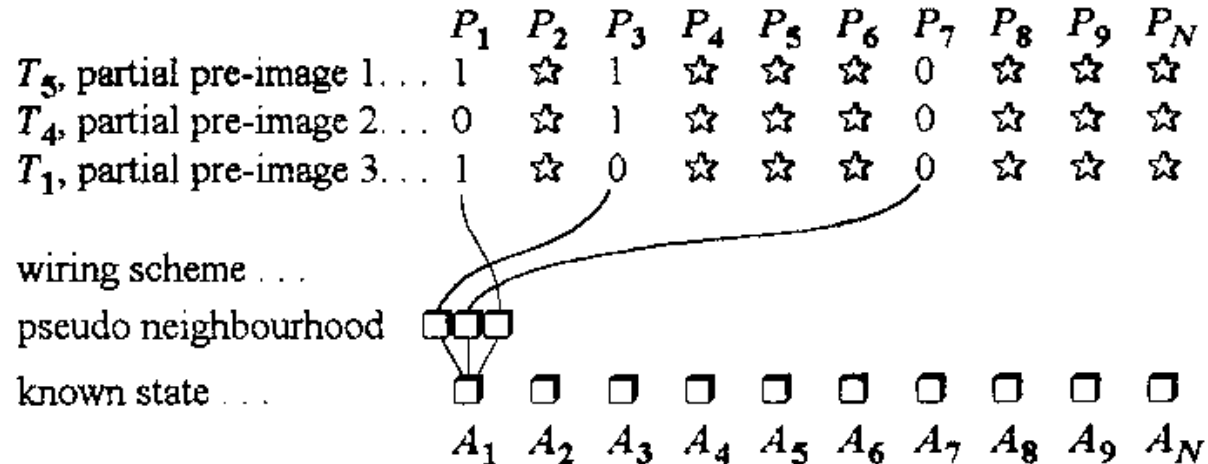
[A. Wuensche, Discrete Dynamics Lab]

Calculating Pre-Images



Calculating Pre-Images (2)

111 110 101 100 011 010 001 000 ...neighborhoods
 rule table... 0 0 1 1 0 0 1 0 ...outputs (0 or 1)
 T_7 T_6 T_5 T_4 T_3 T_2 T_1 T_0



Mutations on Random Boolean Networks

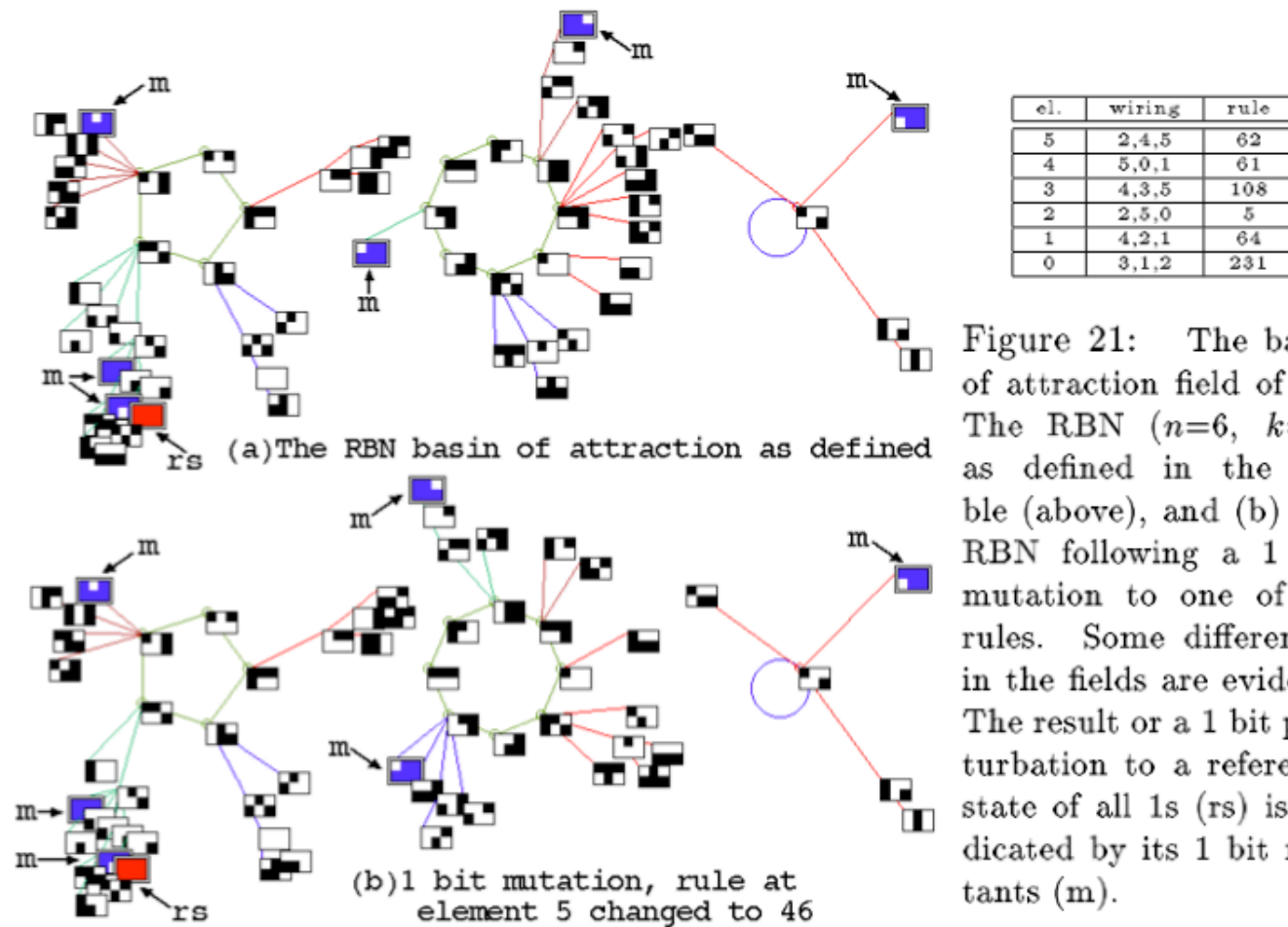
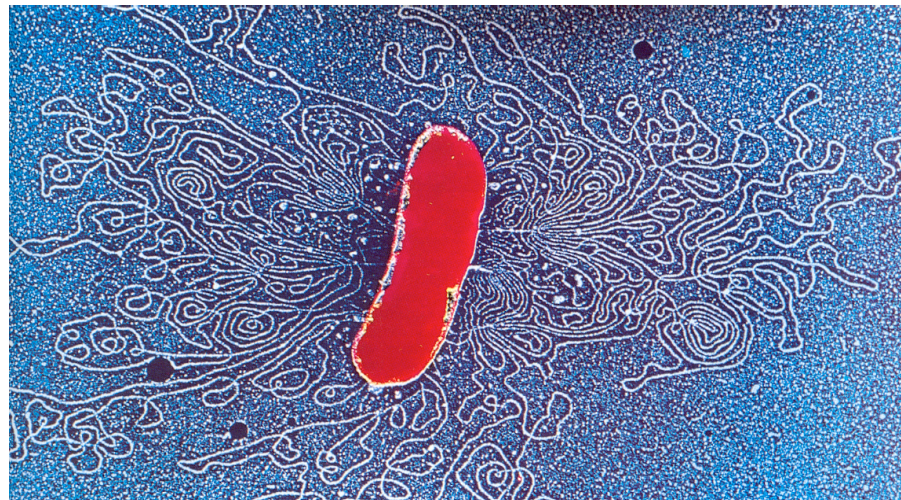


Figure 21: The basin of attraction field of (a) The RBN ($n=6$, $k=3$) as defined in the table (above), and (b) the RBN following a 1 bit mutation to one of its rules. Some differences in the fields are evident. The result of a 1 bit perturbation to a reference state of all 1s (rs) is indicated by its 1 bit mutants (m).

[A. Wuensche 98]

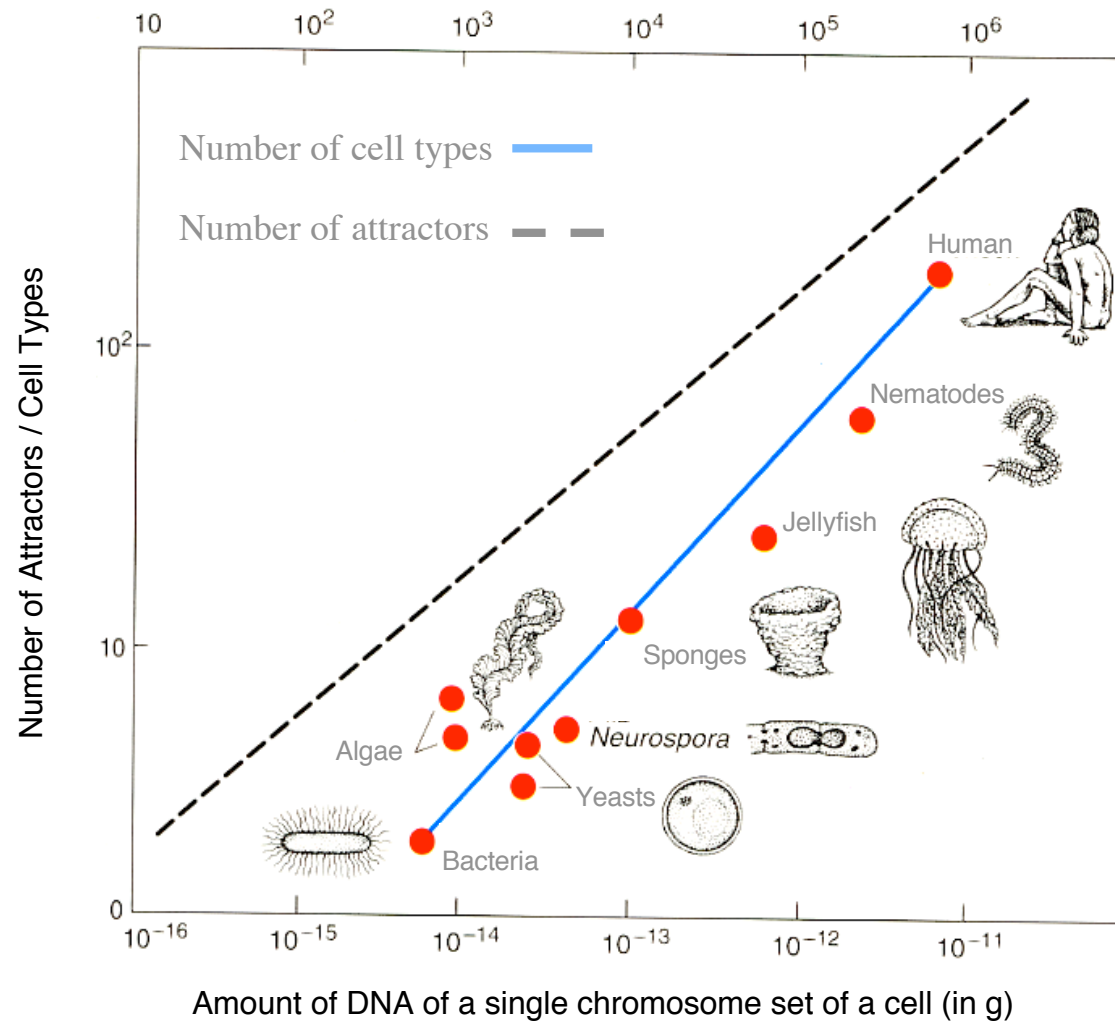
Attractor = Cell Type ?

- From the set of all possible gene activation patterns, the regulatory network selects a specific sequence of activations over time.
- A **differentiated cell** doesn't change its type any more.
 - Hence, only a constrained set of genes is active
 - = state cycle
 - = **attractor**?



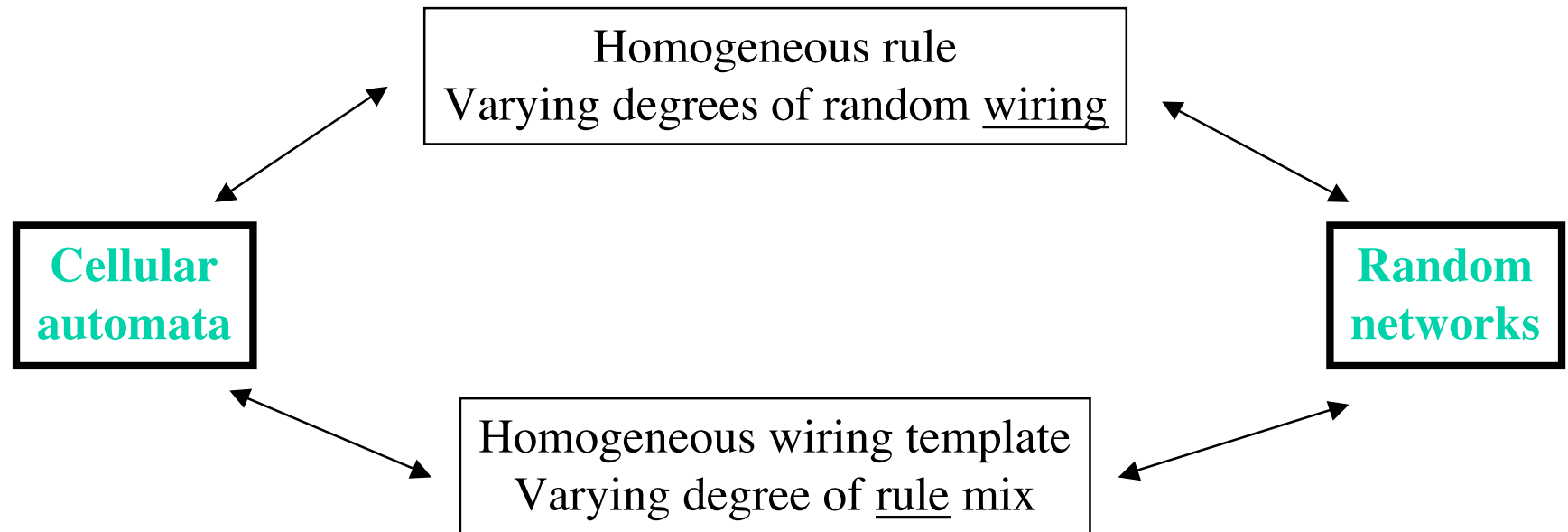
[S. Kauffman: Leben am Rande des Chaos]

Cell Types vs. Attractors



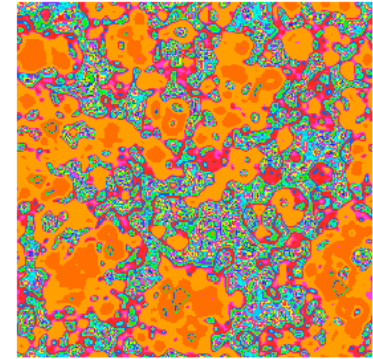
[S. Kauffman: Leben am Rande des Chaos]

Intermediate Architectures

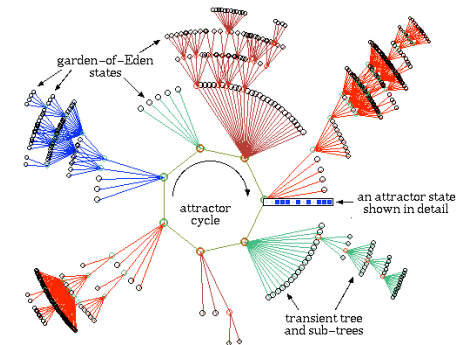


Cellular Automata

Lindenmayer
Systems



Random Boolean
Networks



Classifier Systems

References

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<http://www.santafe.edu/~wuensch/ddlab.html>