

# Quantifying Self-Organization and Coherent Structures with Statistical Complexity

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# References

Shalizi, Shalizi & Haslinger, PRL 93 (2004): 118701 = [nlin.AO/0409024](#)

Shalizi, DMTCS “AB” (2003): 11 = [math.PR/0305160](#)

Shalizi, Rouquier, Haslinger, Moore & Shalizi, in prep.

- “I know it when I see it”
- Disputes: turbulence, ecology
- Does self-organizing  $\Rightarrow$  irreversible?  
Yes: Prigogine, Nicolis; Haken; etc.  
No: D’Souza, Margolus; Smith
- *Not* self-organized criticality (necessarily)
- Why not just use entropy?

# Entropy

- Entropy = log of volume in phase space

$$S = k \log W$$

- Much to do with heat, work, etc., etc.
- Nothing to do with organization
  - Low entropy disorganized systems (low-temperature stat. mech.)
  - High entropy organized things (organisms)
  - Organization  $\uparrow$  *because* entropy  $\uparrow$  (self-assembly)

# If Not Entropy, What?

- Self-organization = complexity increases
- Complexity = description length (Kolmogorov, von Neumann, etc.)
- Probabilistic descriptions, statistical complexity (Rissanen, many others)
- Complexity = information required for prediction (Grassberger)
- Role of minimal sufficient statistics (Crutchfield & Young)

# Local Statistics

- Local statistics are summaries of the past light-cone
- Maximize information in statistic (sufficiency)  $\Rightarrow$  maximally accurate prediction
- Minimize information needed by statistic (minimality)



- past A and past B equivalent iff

$$\Pr(\text{Future}|A) = \Pr(\text{Future}|B)$$

- $[A]$  = all pasts equivalent to A
- Statistic (“causal state”):

$$\varepsilon(\text{past}) = [\text{past}]$$

- Sufficient, minimal, unique, recursive, etc.
- Complexity = information needed to fix state

$$C = I[\text{past}; \varepsilon(\text{past})]$$

System has self-organized between  $t_1$  and  $t_2$  if

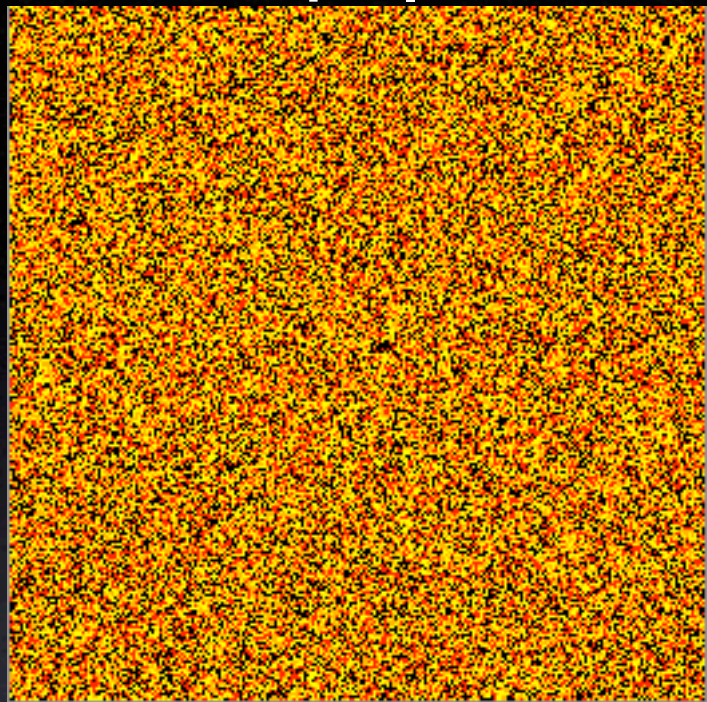
$$(I) C(t_1) < C(t_2)$$

(II) the increase is not caused by outside input

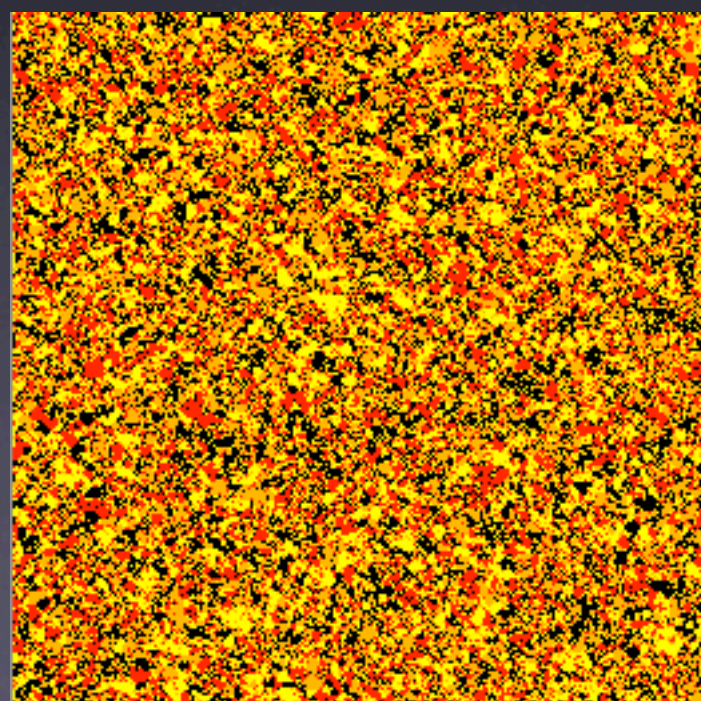
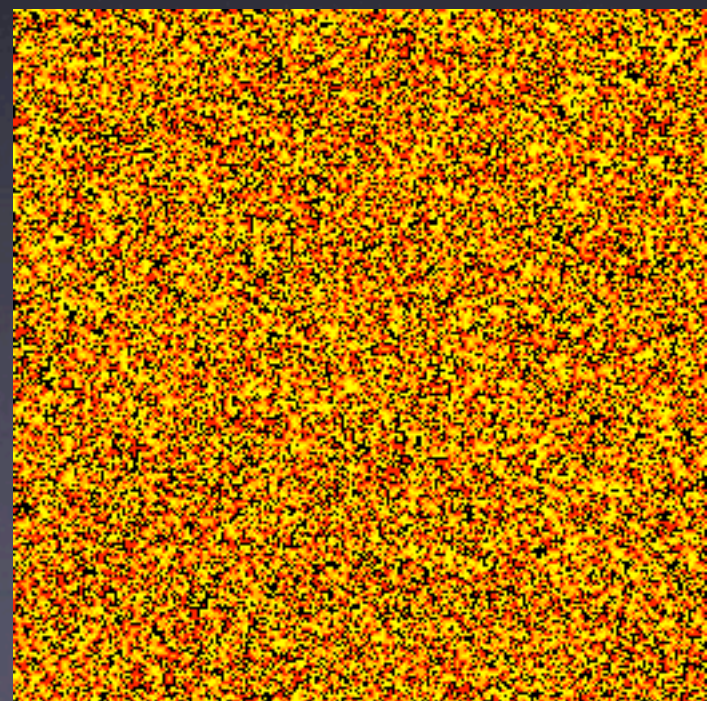
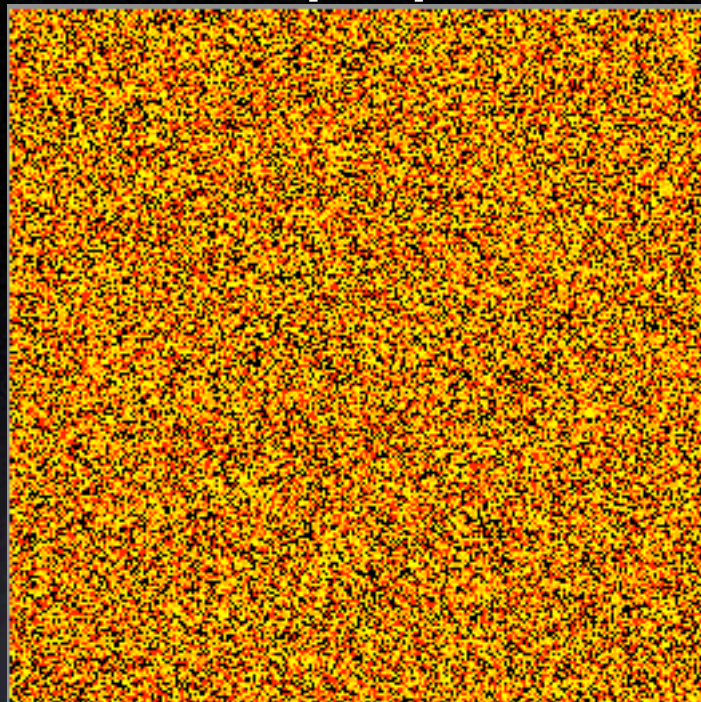
# Cyclic Cellular Automata

- Qualitative model of excitable media
- $K$  colors; a cell of color  $k$  switches to  $k+1 \pmod{K}$  if at least  $T$  neighbors are already of that color
- Analytic theory for structures formed (Griffeath et al.)
- Spirals, “turbulence”, local oscillation, fixation

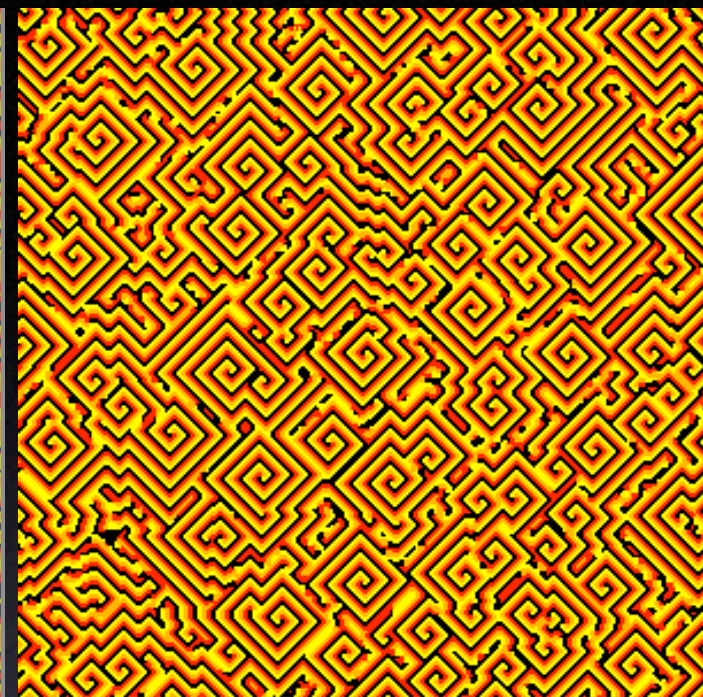
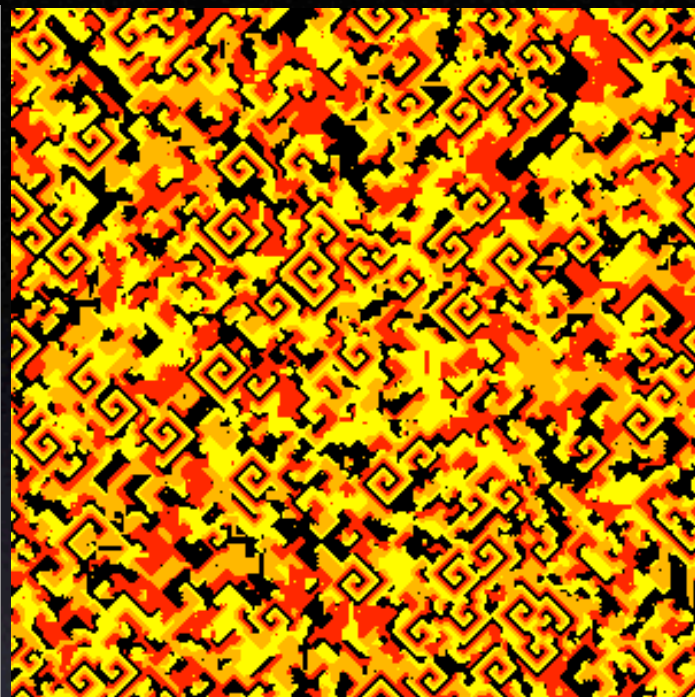
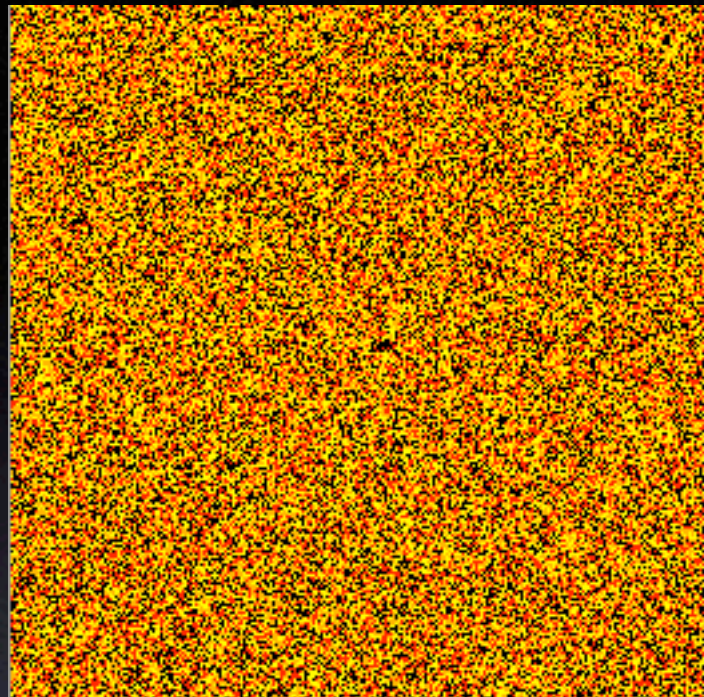
$T=1$



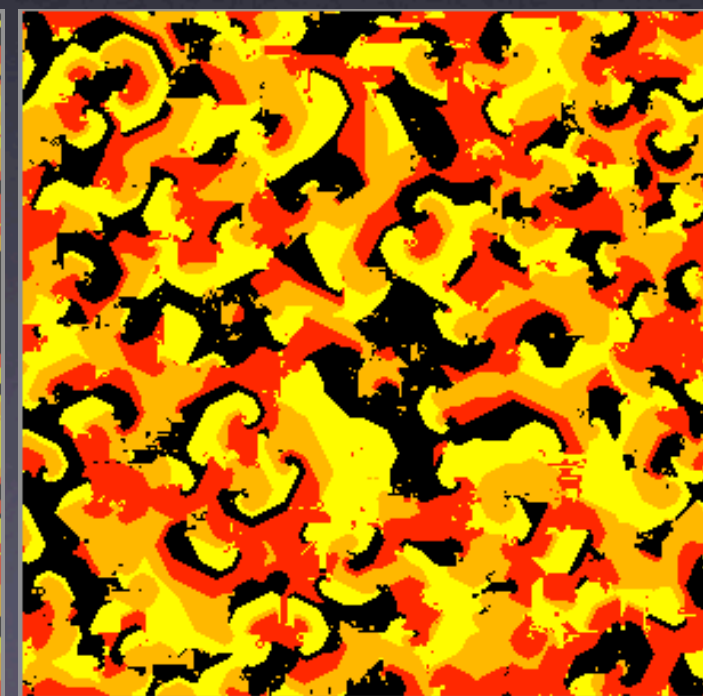
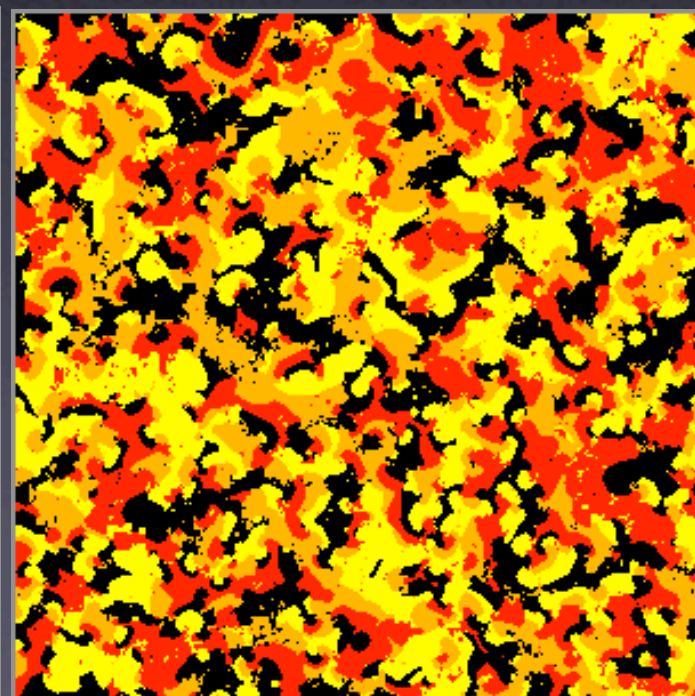
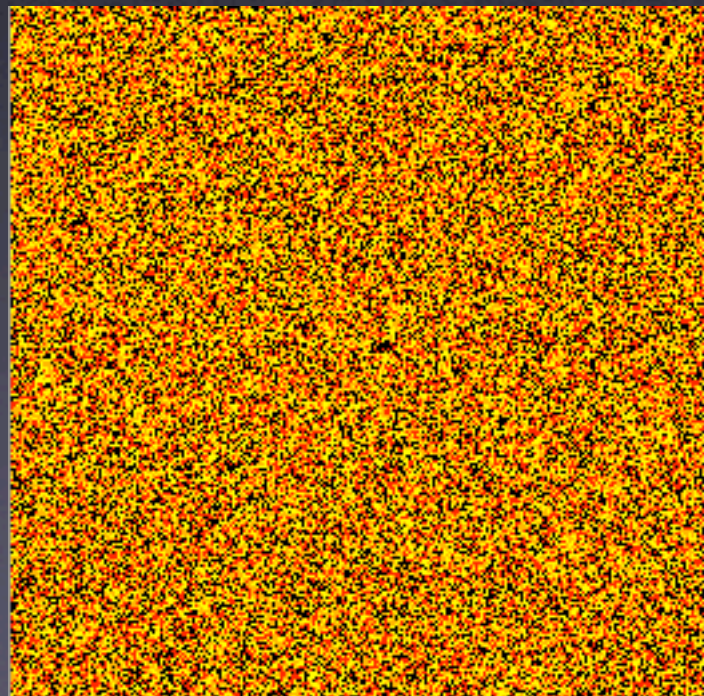
$T=4$



T=2



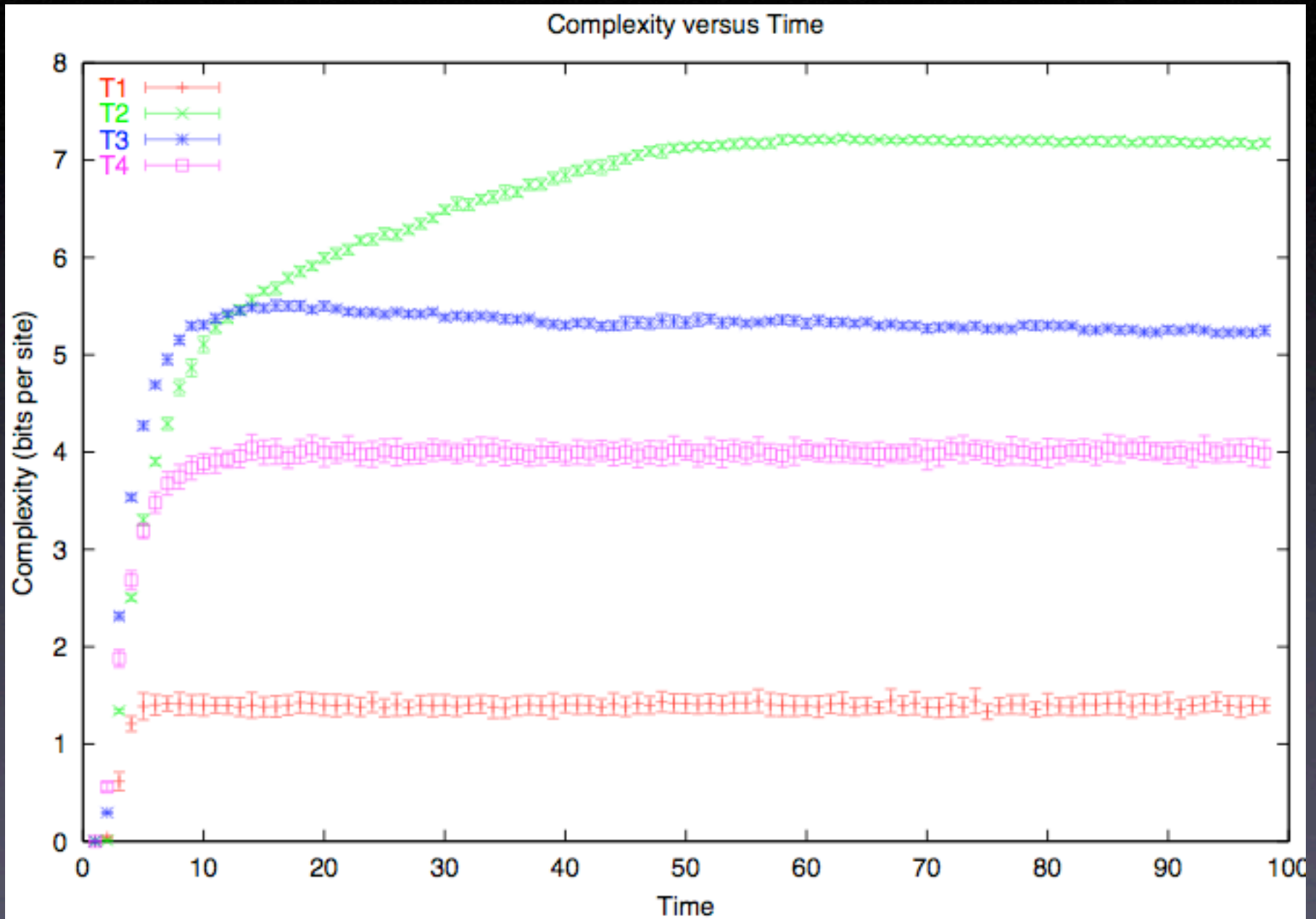
T=3



## Identification of states from data

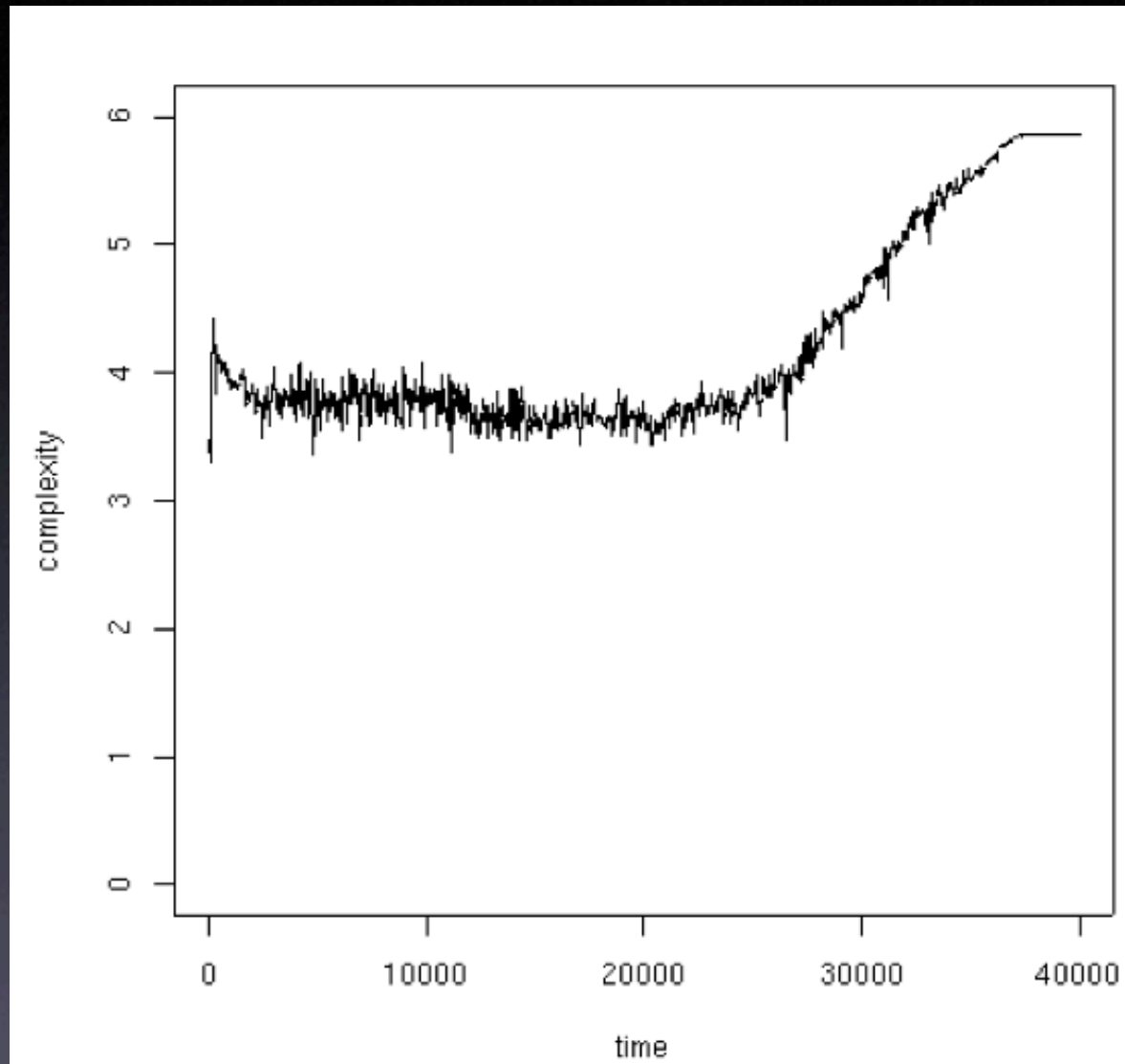
- Clustering of past light-cone configurations by similarity of empirical distributions for future
- Consistent estimates of states,  $C(t)$

# CCA



(300x300, n = 30)

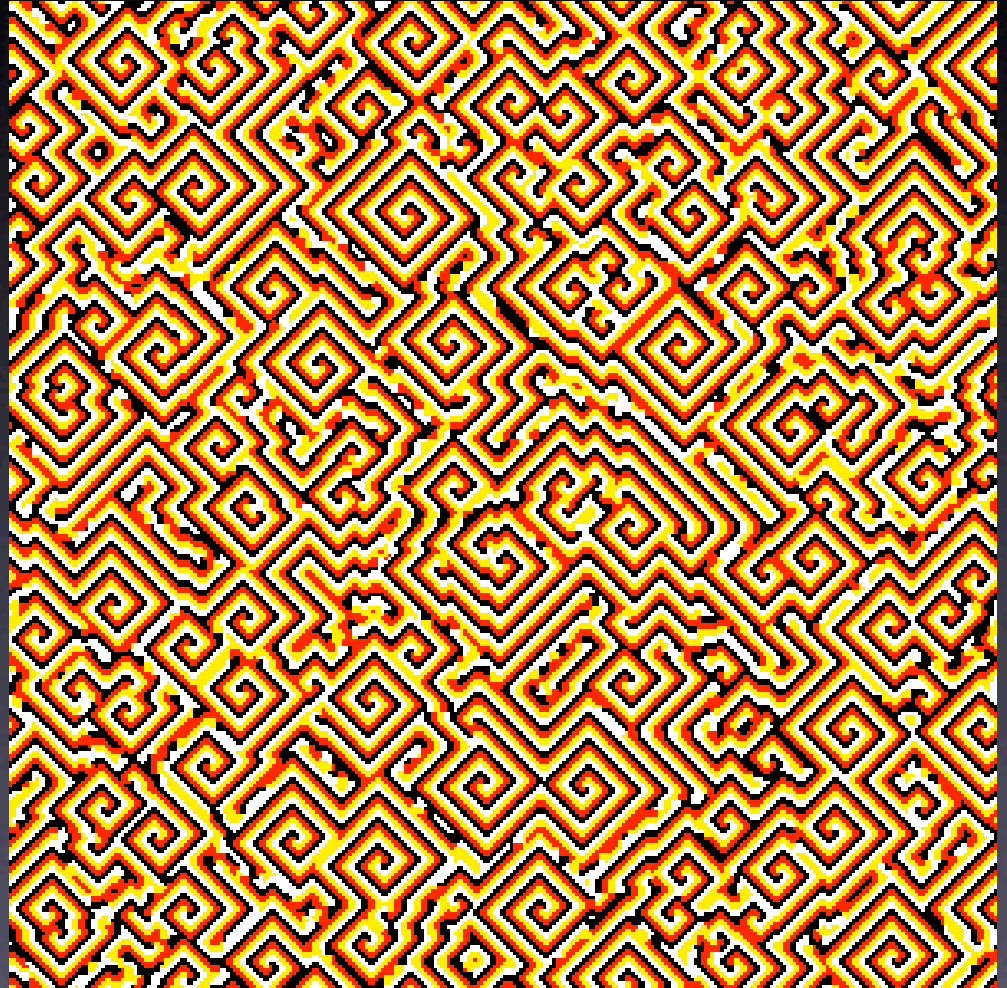
# BTW Sandpile



(supra-threshold relaxation,  $300 \times 300$ ,  $n=1$ )

# Finding Coherent Structures

- Spatially extended, temporally persistent
- Generated by the microdynamics (“emergent”)
- More efficient and more comprehensible descriptions



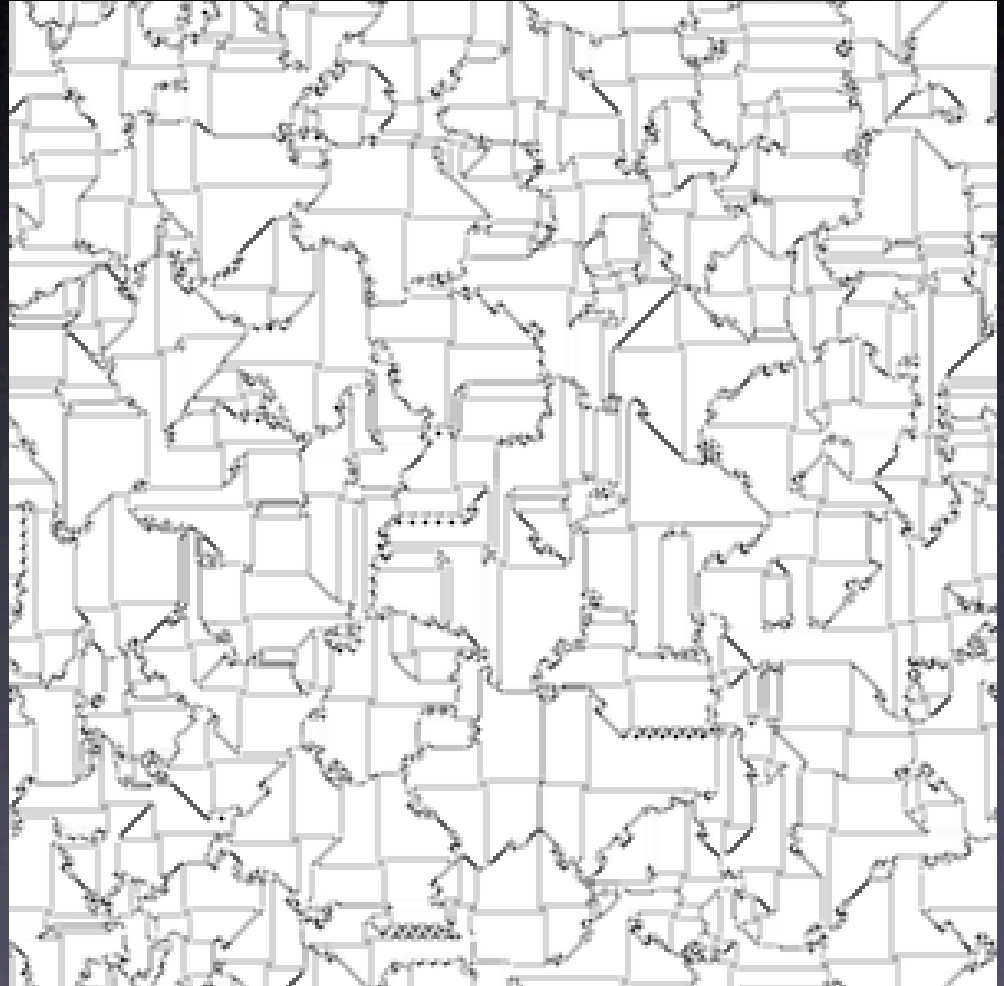
# Order parameters

- OP measures symmetry breaking

$$\Phi = f(\text{OP})$$

$$-\log(\text{Pr}(\text{config})) \sim \int f(\text{OP})$$

- Structures = defects in OP field
- OP found by trial and error

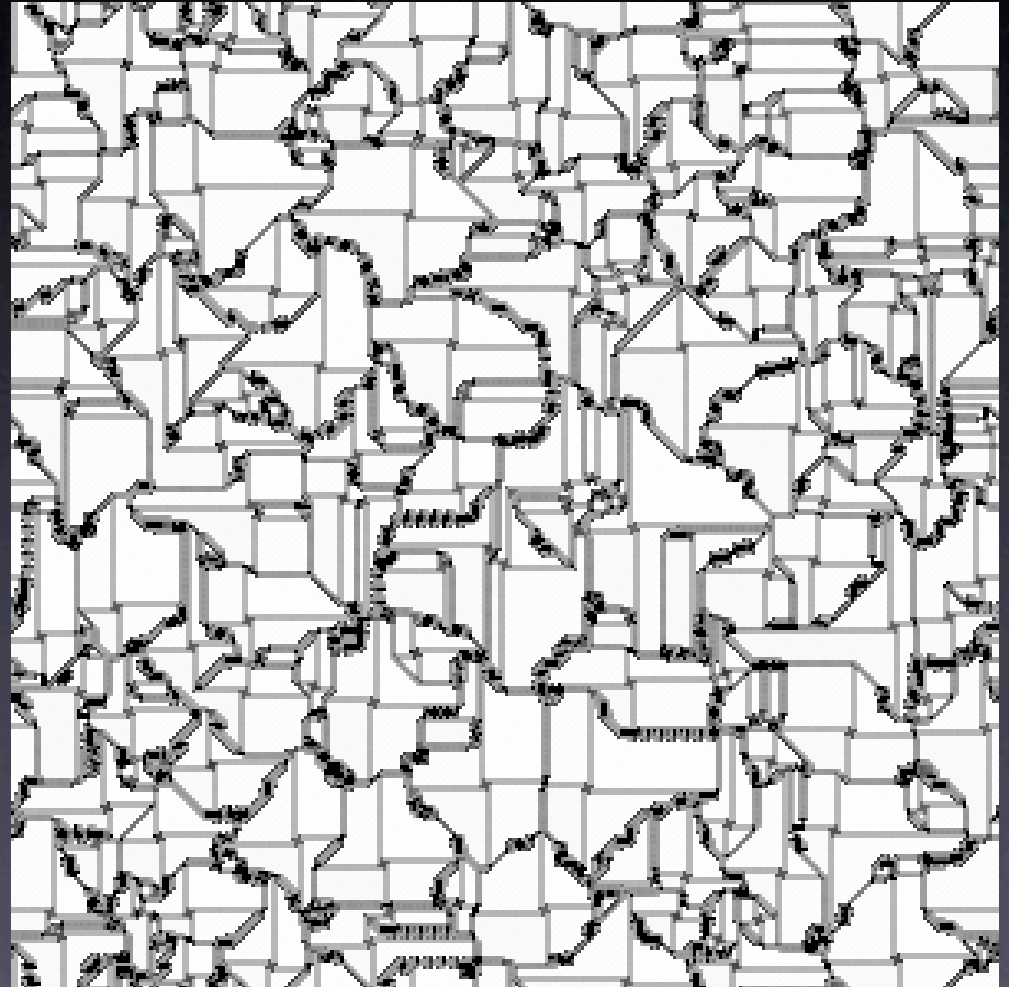


# Complexity field

- Local description length

$$C = -\log(\text{Pr}(\text{state}))$$

- Automatic; no tradition needed



# Summary

- Statistical complexity = information needed for optimal prediction
- Self-organization = internally-caused rise in complexity
- This definition works in practice
- The local complexity field lets us identify coherent structures automatically