

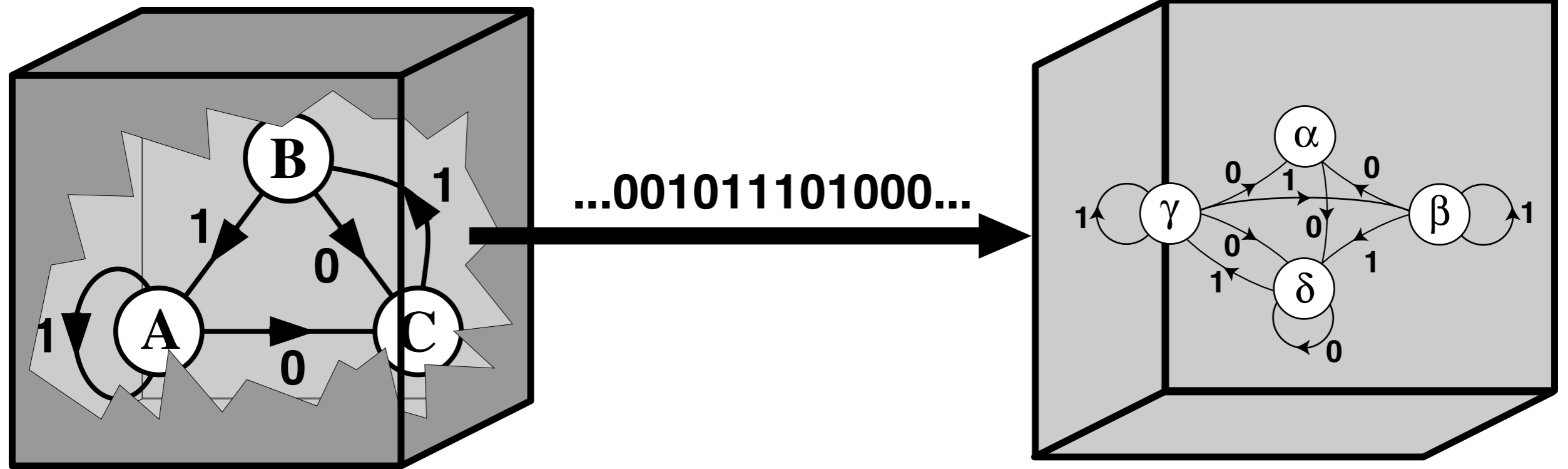
# What We Didn't Cover: The Present and Future

Reading for this lecture:

*Lecture Notes* and cited papers.

# Topics & Applications...

## Course narrative:



**System**

**Instrument**

**Process**

**Modeller**

Forms of Chaos:

Deterministic sources  
of novelty

Mechanisms that produce  
unpredictability

Sensitive dependence on  
initial condition

Sensitive dependence on  
parameter

Measurement Theory:

Partitions

Optimal Instrument:

$$\max_{\{P\}} h_{\mu}$$

$$\min_{\{P\}} C_{\mu}$$

How random?

$$\lambda_{\max}, H(L), h_{\mu},$$

$$\mathbf{G}, \mathcal{R}$$

How structured?

$$C_{\mu}, \mathbf{E}, \mathbf{T}, \chi, \Xi$$

Universal model:

$\epsilon$  – Machine

Pattern defined

Causal Architecture

Intrinsic Computation

# Topics & Applications...

What we didn't cover:

$\epsilon$ -Machine Enumeration

Infinite  $\epsilon$ -Machines & Generalized Hidden Markov Models

Hierarchical  $\epsilon$ -Machine Reconstruction

Quantum Processes & Quantum Machines

Rate Distortion Theory & Optimal Causal Inference

Cellular Automata Computational Mechanics

Spin Systems in 1D and 2D

Optimal Instrument Design

Complexity-Entropy Diagrams

# Topics & Applications...

What we didn't cover ...

## $\epsilon$ -Machine Enumeration

Every process has its  $\epsilon$ -machine presentation

Space of processes = Space of  $\epsilon$ -Machines

$\epsilon$ -Machines can be exactly enumerated

Alphabet

$n \backslash k$	2	3	4	5	6
1	3	7	15	31	63
2	7	141	1,873	20,925	213,997
3	78	15,598	1,658,606	136,146,590	
4	1,388	3,625,638			
5	35,186				
6	1,132,613				
7	43,997,426				
8	1,993,473,480				

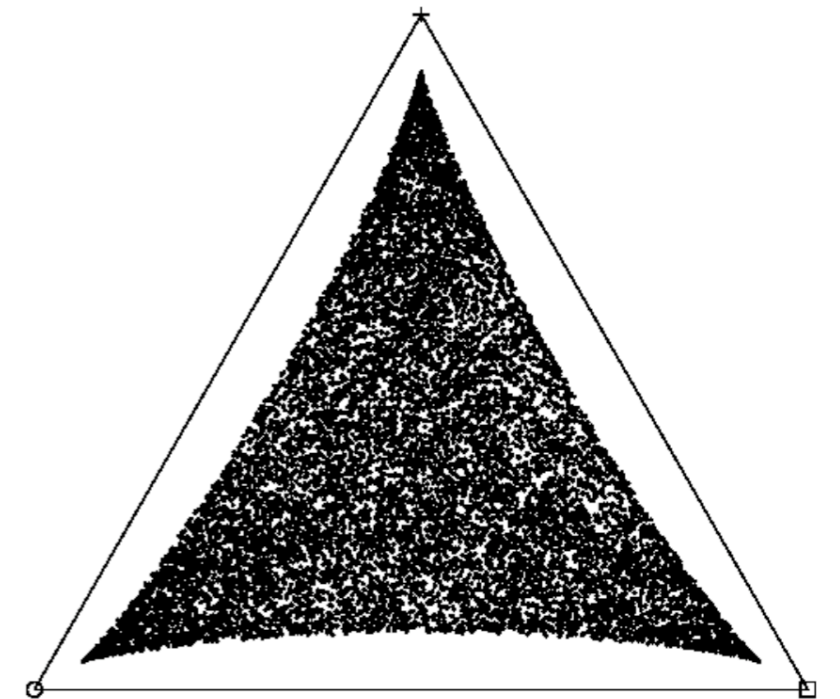
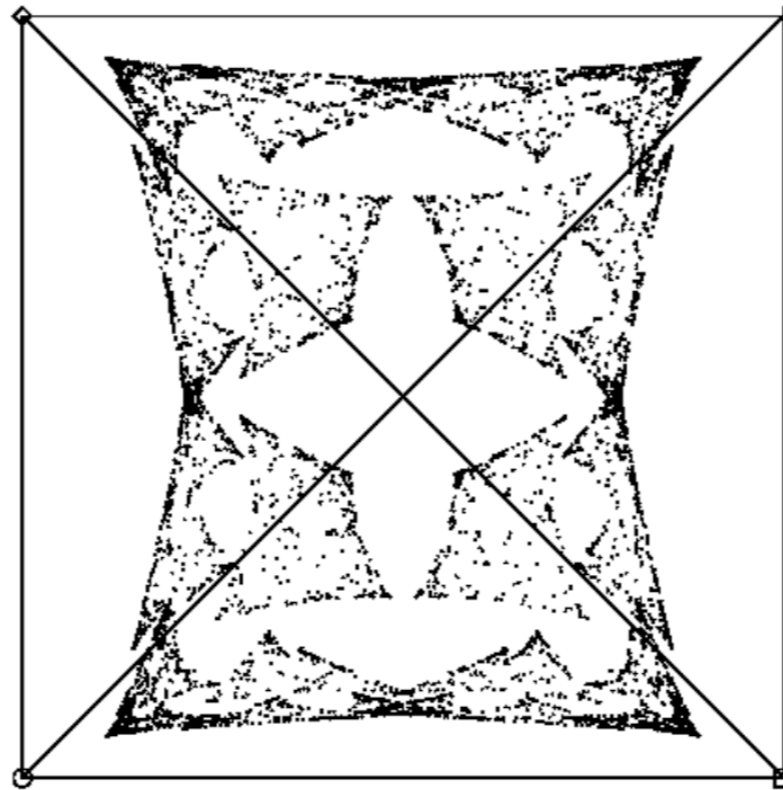
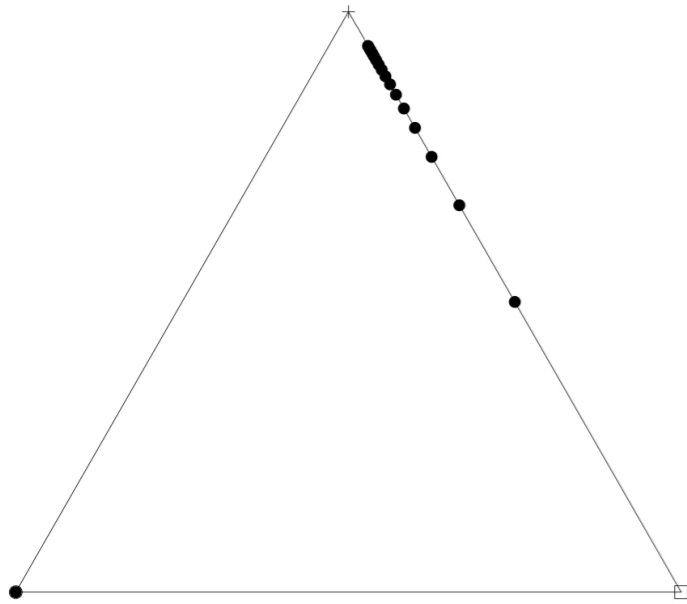
## Unknown unknowns!

B. D. Johnson, J. P. Crutchfield, C. J. Ellison, and C. S. McTague, "Enumerating Finitary Processes", (2010) submitted. Santa Fe Institute Working Paper 10-11-027. arXiv:1011.0036 [cs.FL].

# Topics & Applications...

What we didn't cover ...

## Infinite $\epsilon$ -Machines & Generalized Hidden Markov Models



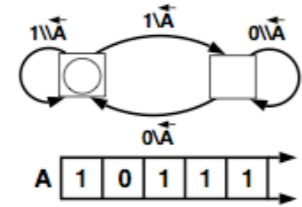
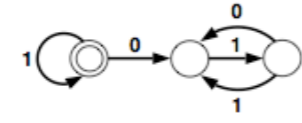
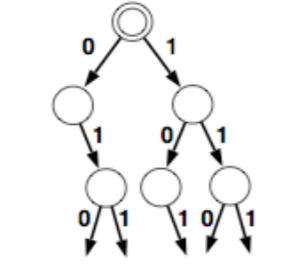
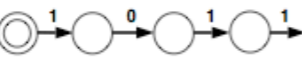
D. Upper, Ph.D. Dissertation (Mathematica, Berkeley, February 1997).

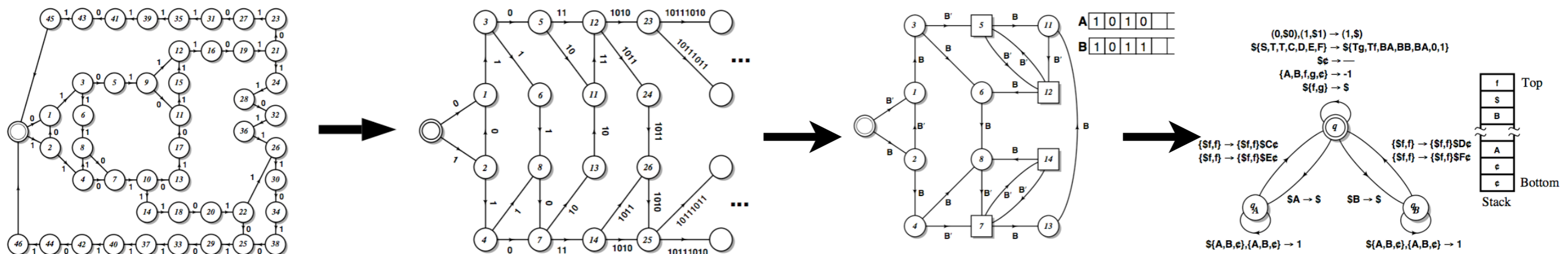
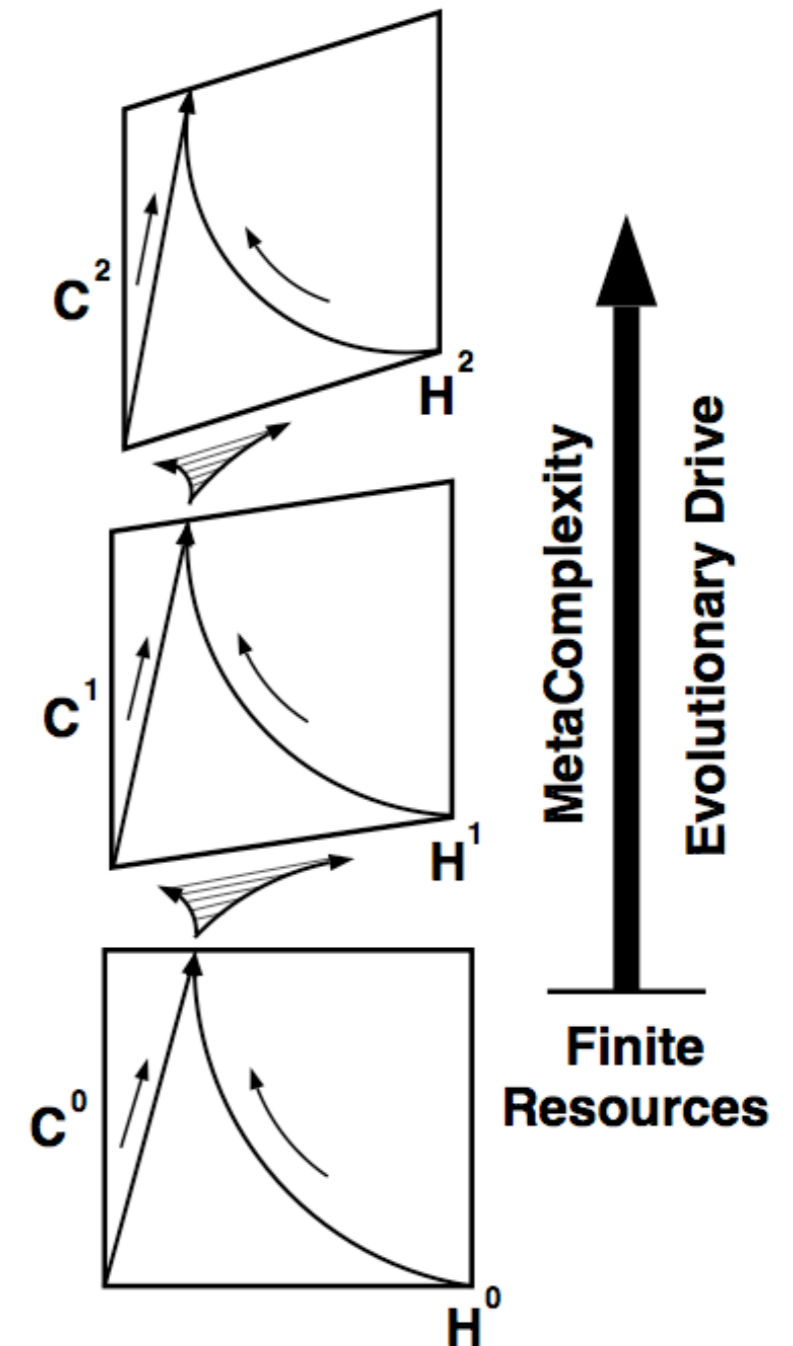
J. P. Crutchfield, "Calculi of Emergence:", *Computation, Dynamics, and Induction*, *Physica D* **75** (1994) 11-54.

# Topics & Applications...

What we didn't cover ...

## Hierarchical $\epsilon$ -Machine Reconstruction

Level	Model Class	Machine	Model Size, if class is appropriate	Equivalence Relation
...	...	...	...	...
3	String Production		$\mathcal{O}(\ V\  + \ E\  + \ P\ )$	Finitary-Recursive Conditional Independence
2	Finite Automata		$\mathcal{O}(\ V\  + \ E\ )$	Conditional Independence
1	Tree		$\mathcal{O}(\ A\ ^D)$	Block Independence
0	Data Stream		$m$	Measurement

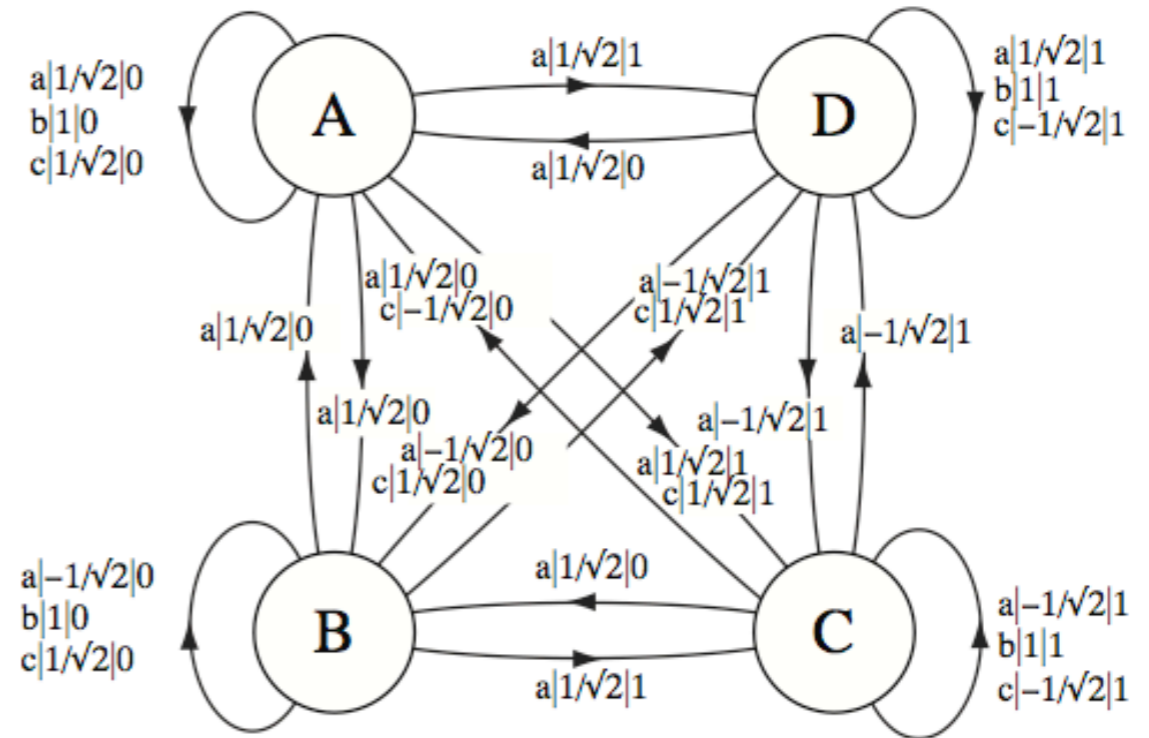
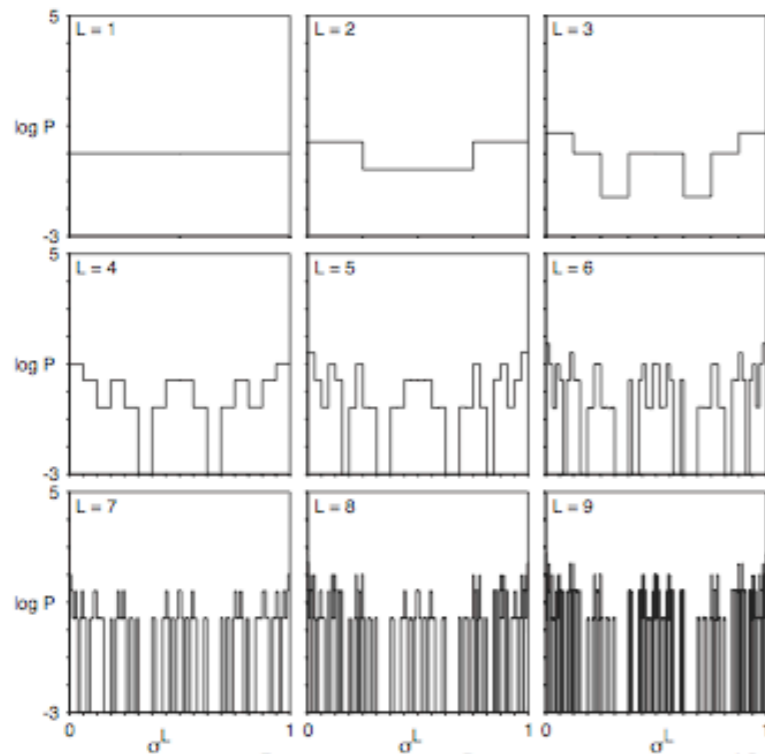
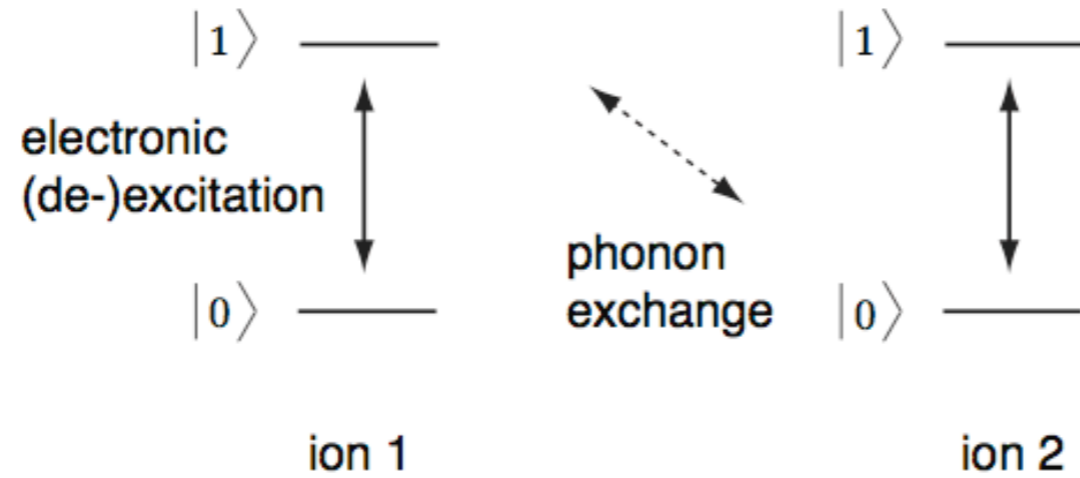


J. P. Crutchfield, "Calculi of Emergence:", Computation, Dynamics, and Induction", Physica D **75** (1994) 11-54.

# Topics & Applications...

What we didn't cover ...

## Quantum Processes & Quantum Machines

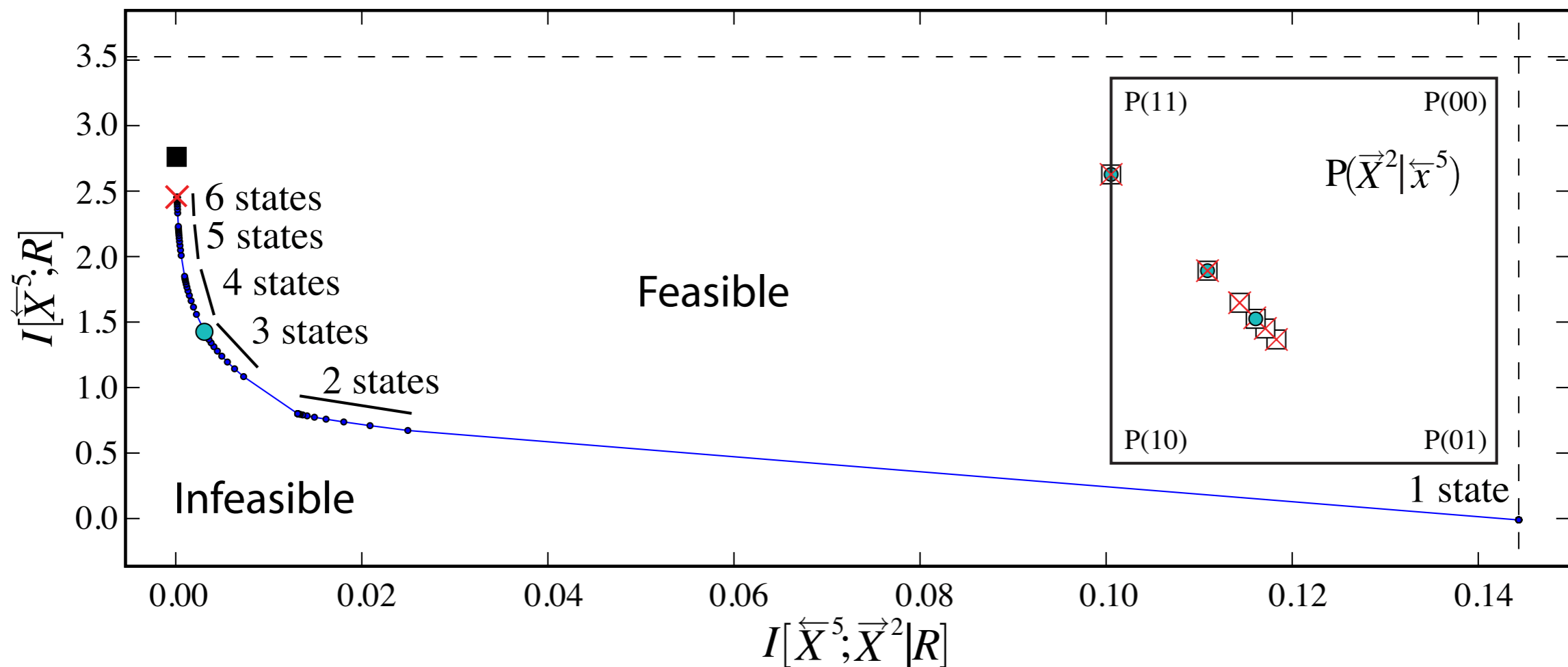


# Topics & Applications...

What we didn't cover ...

## Rate Distortion Theory & Optimal Causal Inference

$$\min_{\Pr(\mathcal{R}|\bar{X})} \left( I[\bar{X}; \mathcal{R}] + \beta I[\bar{X}; \bar{X}^2 | \mathcal{R}] \right)$$



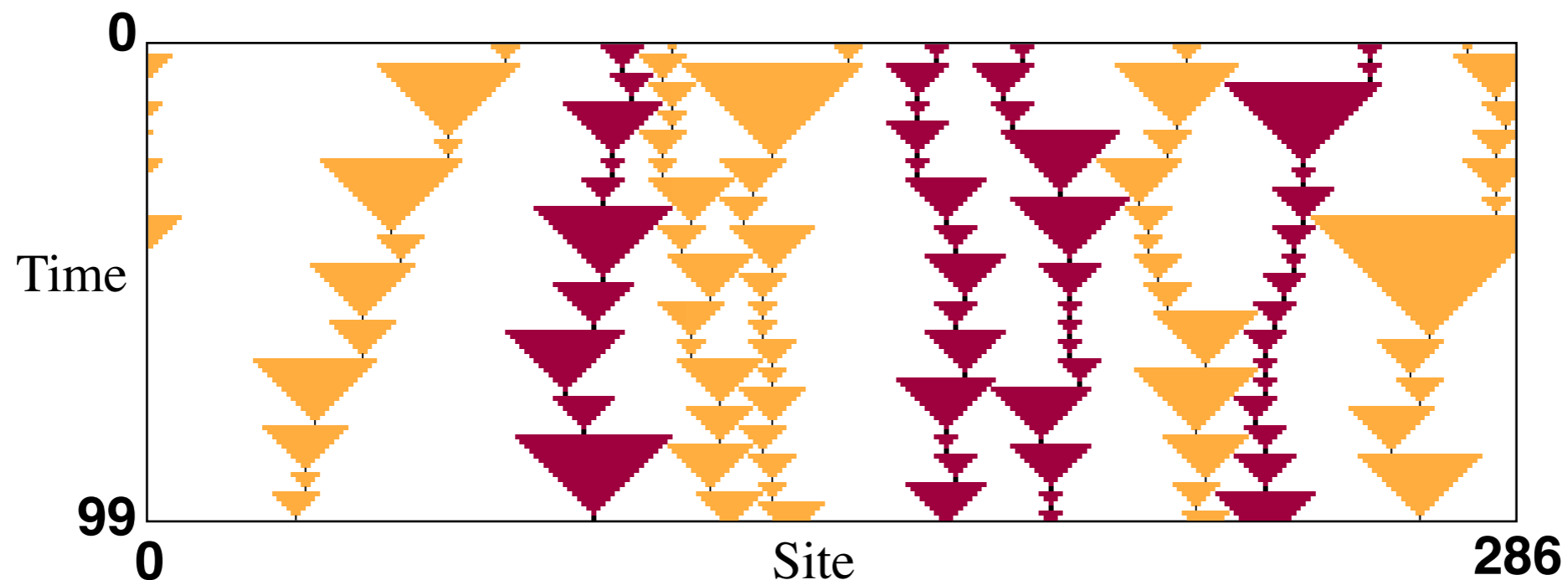
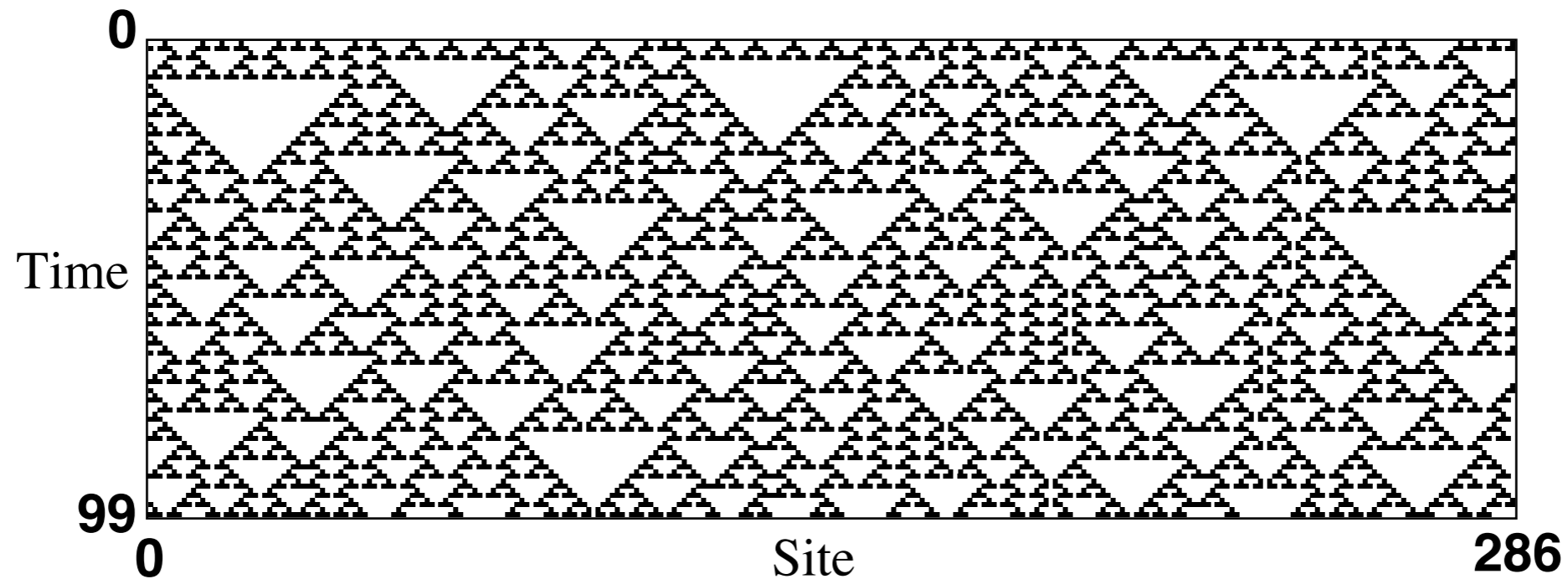
S. Still, J. P. Crutchfield, and C. J. Ellison "Optimal Causal Inference: Estimating Stored Information and Approximating Causal Architecture", CHAOS **20**:3 (2010) 037111.



# Topics & Applications...

What we didn't cover ...

## Cellular Automata Computational Mechanics:

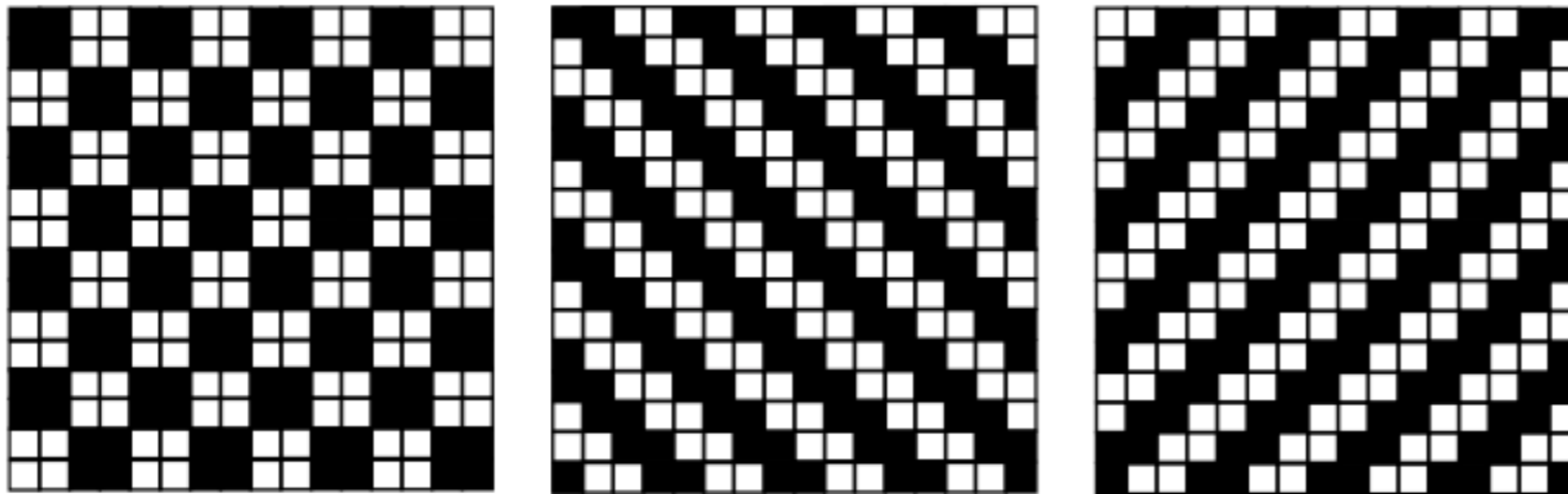


J. E. Hanson and J. P. Crutchfield, "Computational Mechanics of Cellular Automata: An Example", *Physica D* 103 (1997) 169-189.

# Topics & Applications...

What we didn't cover ...

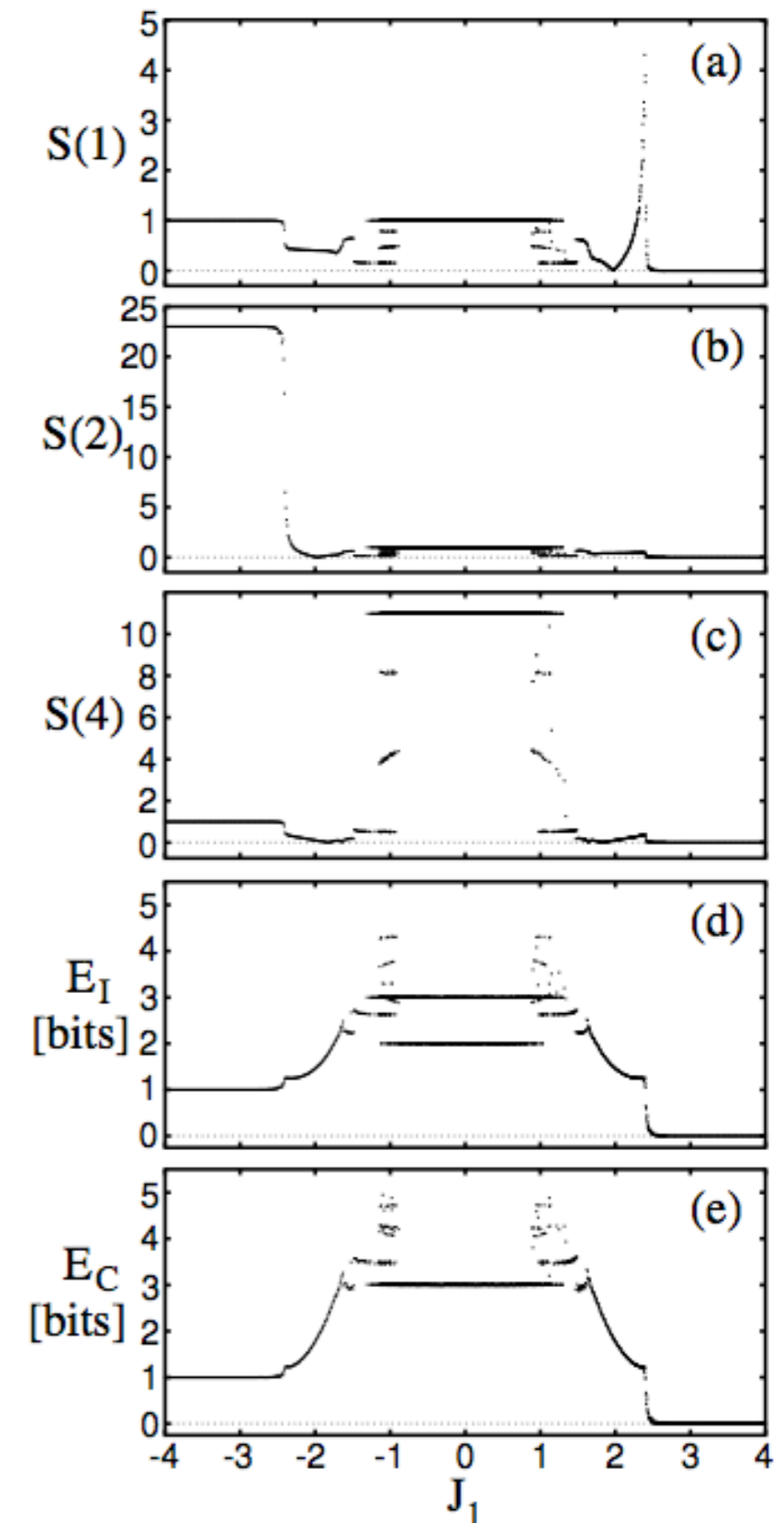
## Spin Systems in 1D and 2D



$$\mathbf{E}_I \equiv \lim_{M, N \rightarrow \infty} I \left[ \begin{array}{c} \uparrow \\ N \\ \downarrow \end{array} \begin{array}{|c|c|c|c|} \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline \end{array} ; \begin{array}{|c|c|c|c|} \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline \end{array} \begin{array}{c} \uparrow \\ N \\ \downarrow \end{array} \right]$$

$$H(M, N) = H \left[ \begin{array}{c} \leftarrow M \rightarrow \\ \begin{array}{|c|c|c|c|} \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline \end{array} \begin{array}{c} \uparrow \\ N \\ \downarrow \end{array} \right]$$

$$\sim \mathbf{E}_S + \mathbf{E}_S^x M + \mathbf{E}_S^y N + h_\mu MN$$



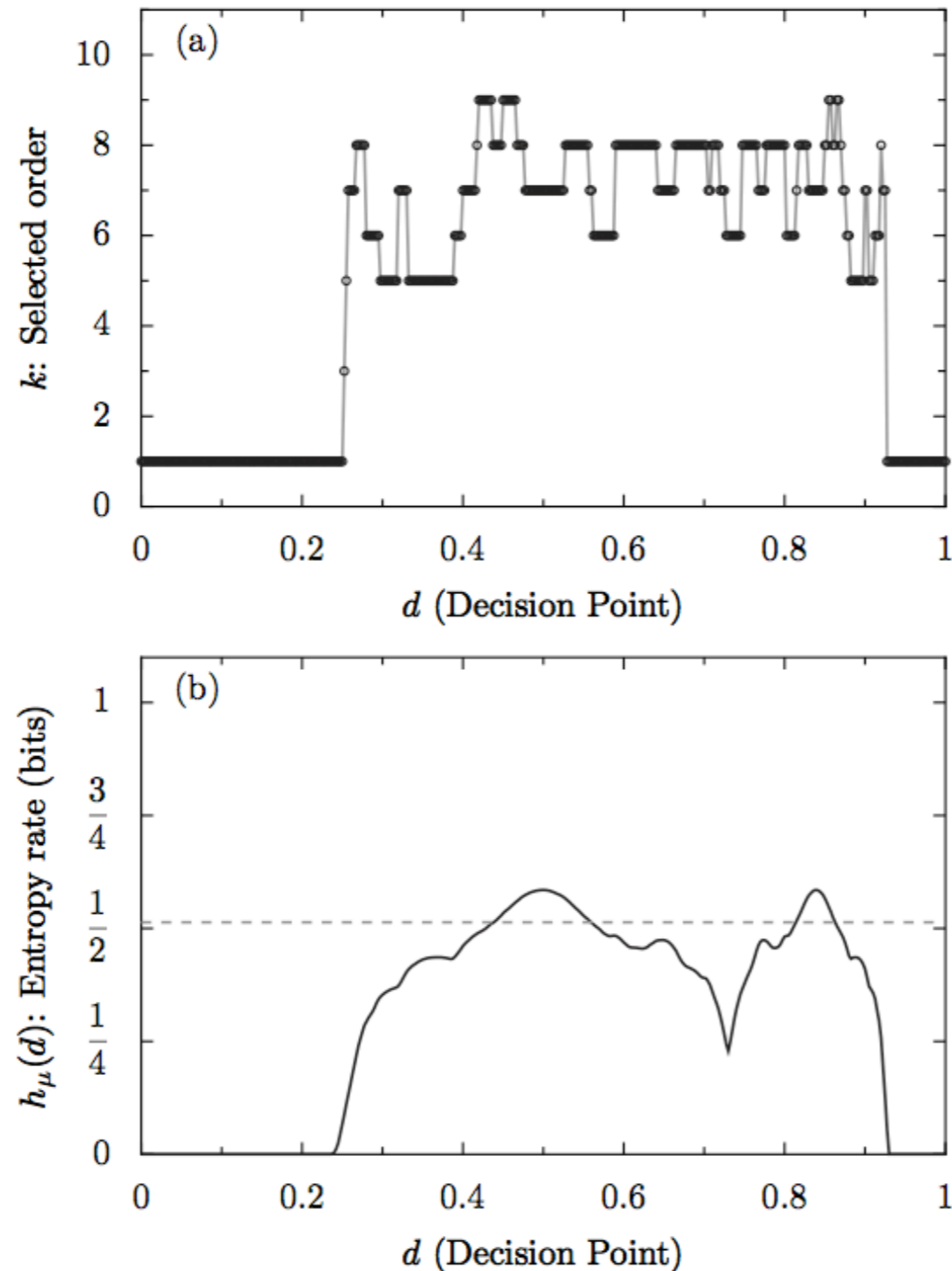
**E: Generalized  
Order Parameter!**

D. P. Feldman and J. P. Crutchfield, "Structural Information in Two-Dimensional Patterns: Entropy Convergence and Excess Entropy", Physical Review E **67** (2003) 051104.

# Topics & Applications...

What we didn't cover ...

## Optimal Instrument Design

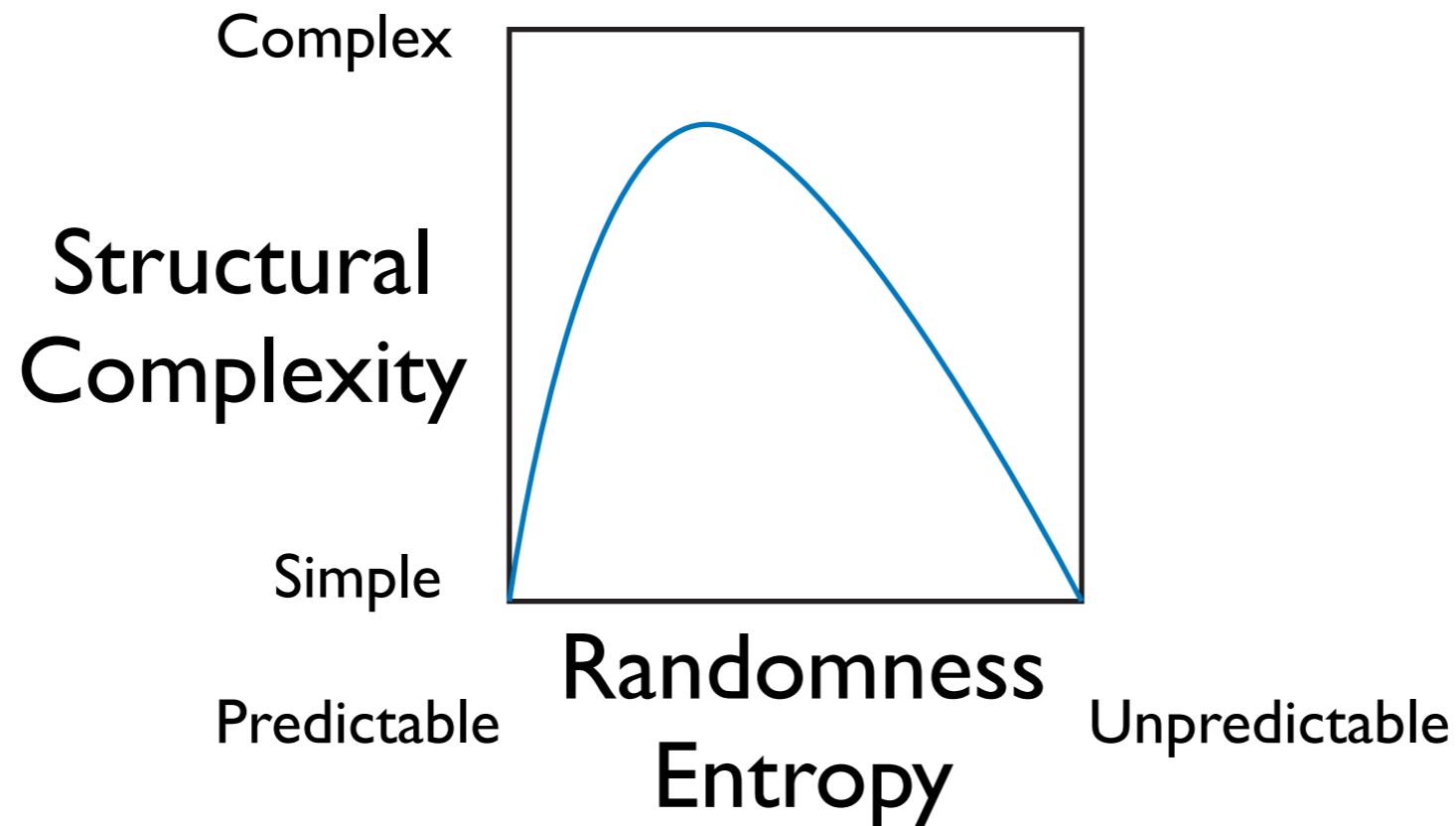


C. C. Strelhoff and J. P. Crutchfield, "Optimal Instruments and Models for Noisy Chaos", CHAOS 17 (2007) 043127.

# Topics & Applications...

What we didn't cover ...

## Complexity-Entropy Diagrams: Analyze a class of processes



Analogous to Thermodynamic Phase Diagram (gas, liquid, solid).

But uses only intrinsic computation properties.

A wide diversity of Complexity-Entropy Diagrams.

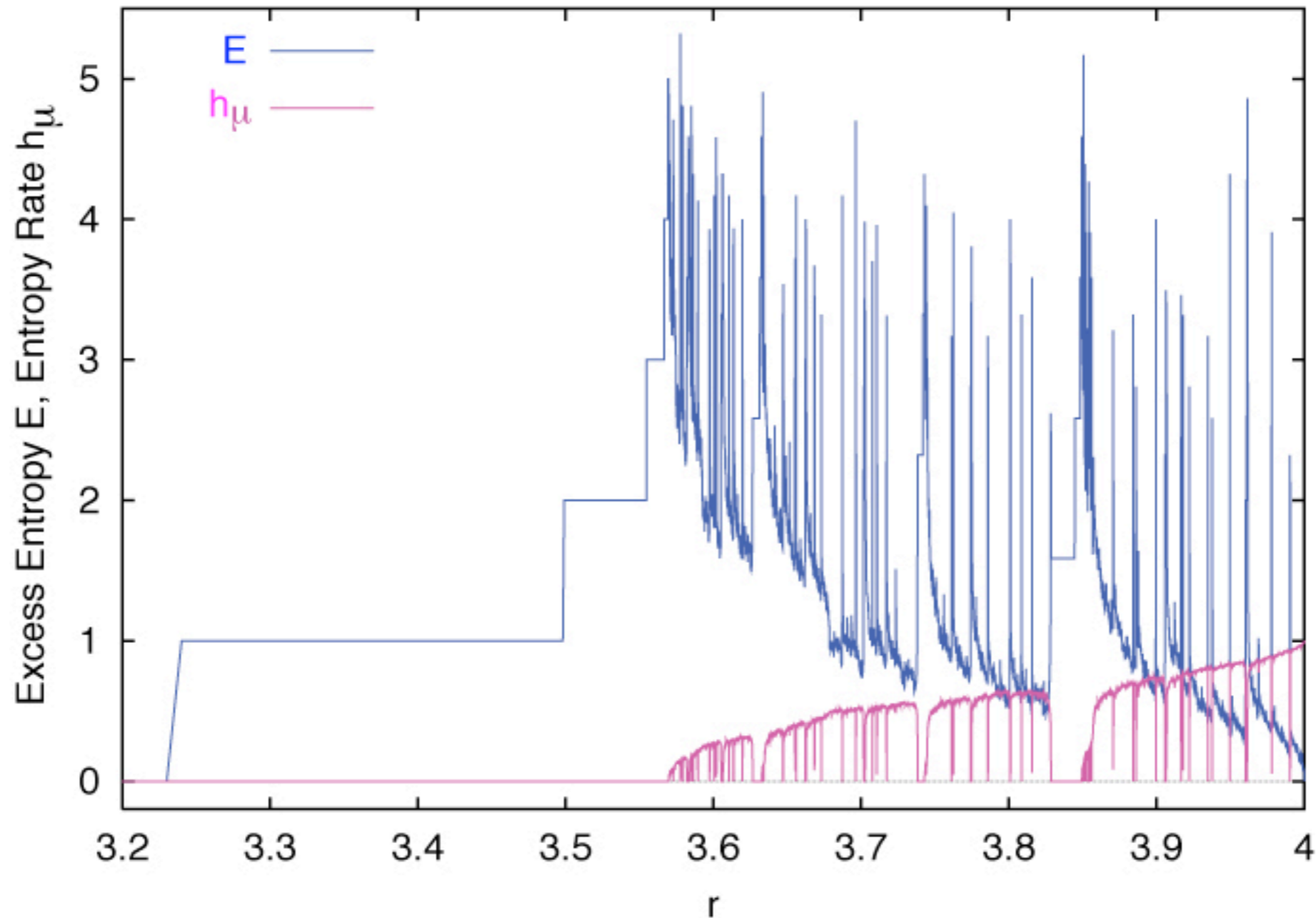
D. P. Feldman, Carl S. McTague, and J. P. Crutchfield, "The Organization of Intrinsic Computation: Complexity-Entropy Diagrams and the Diversity of Natural Information Processing", CHAOS 18:4 (2008) 53-73.

# Topics & Applications...

What we didn't cover ...

Complexity-Entropy Diagrams:  
Analyze a class of processes

## Logistic Map

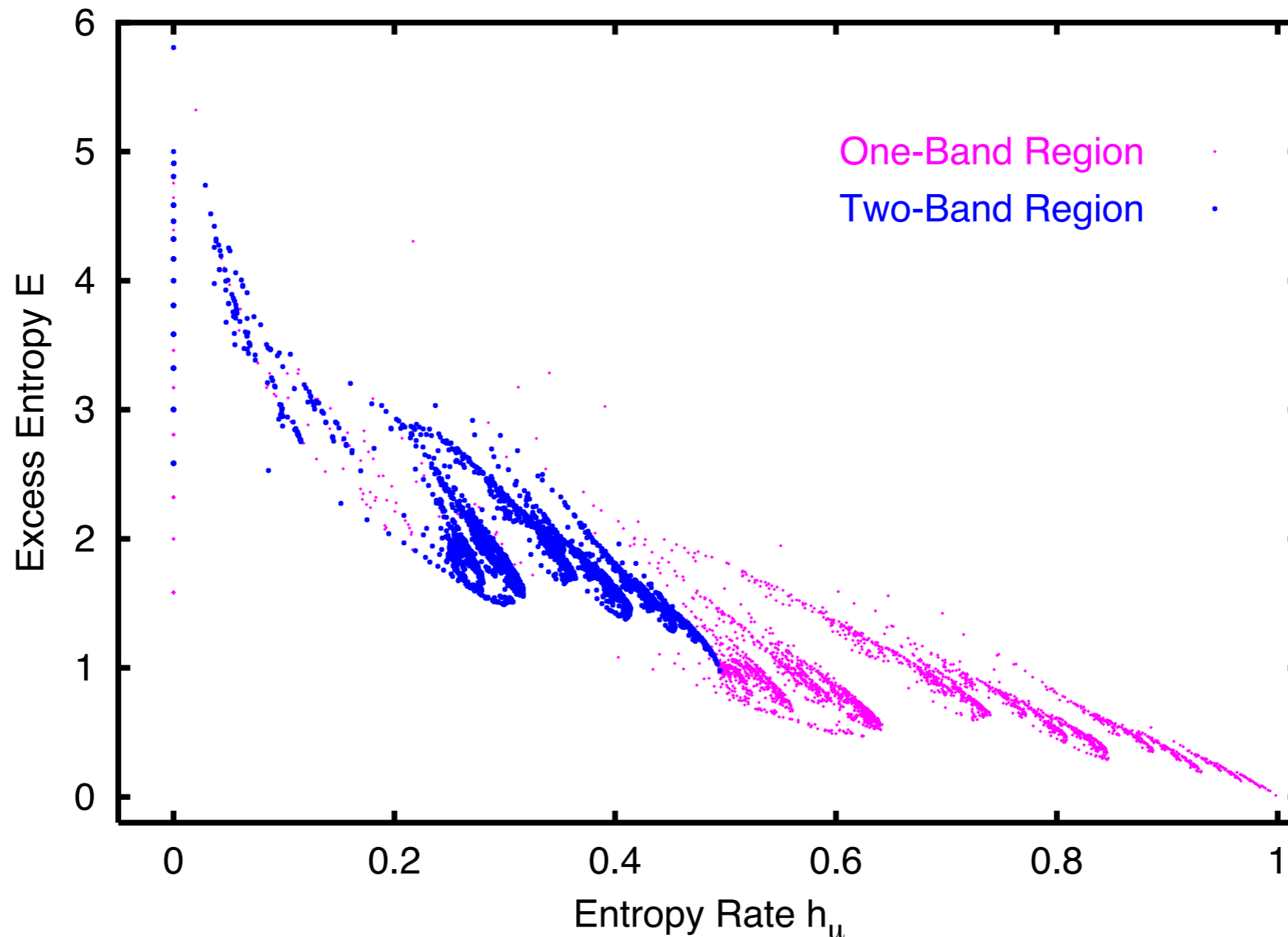


# Topics & Applications...

What we didn't cover ...

Complexity-Entropy Diagrams:  
Analyze a class of processes

## Logistic Map



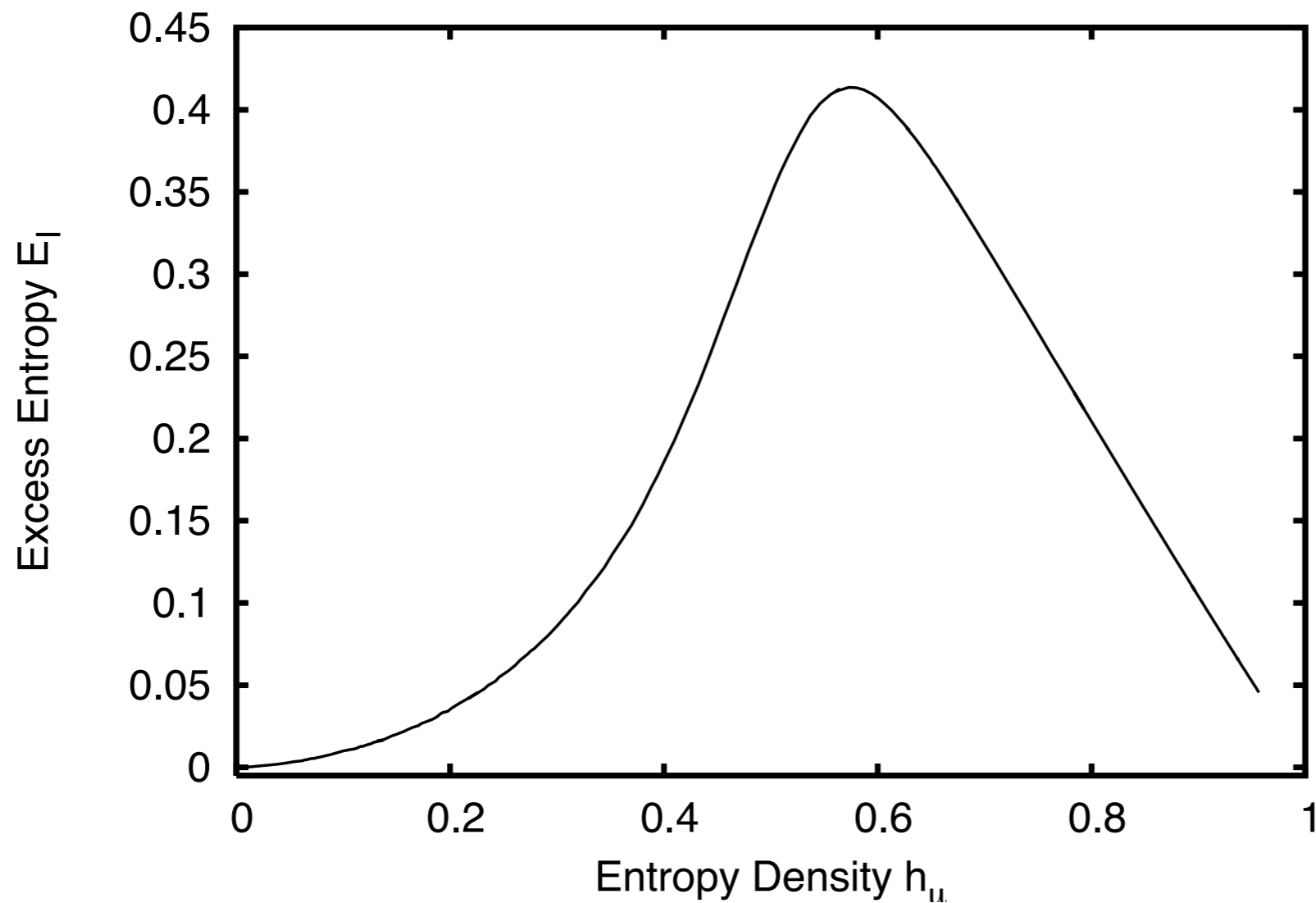
# Topics & Applications...

What we didn't cover ...

Complexity-Entropy Diagrams:

Analyze a class of processes

## 2D Ising Spin System



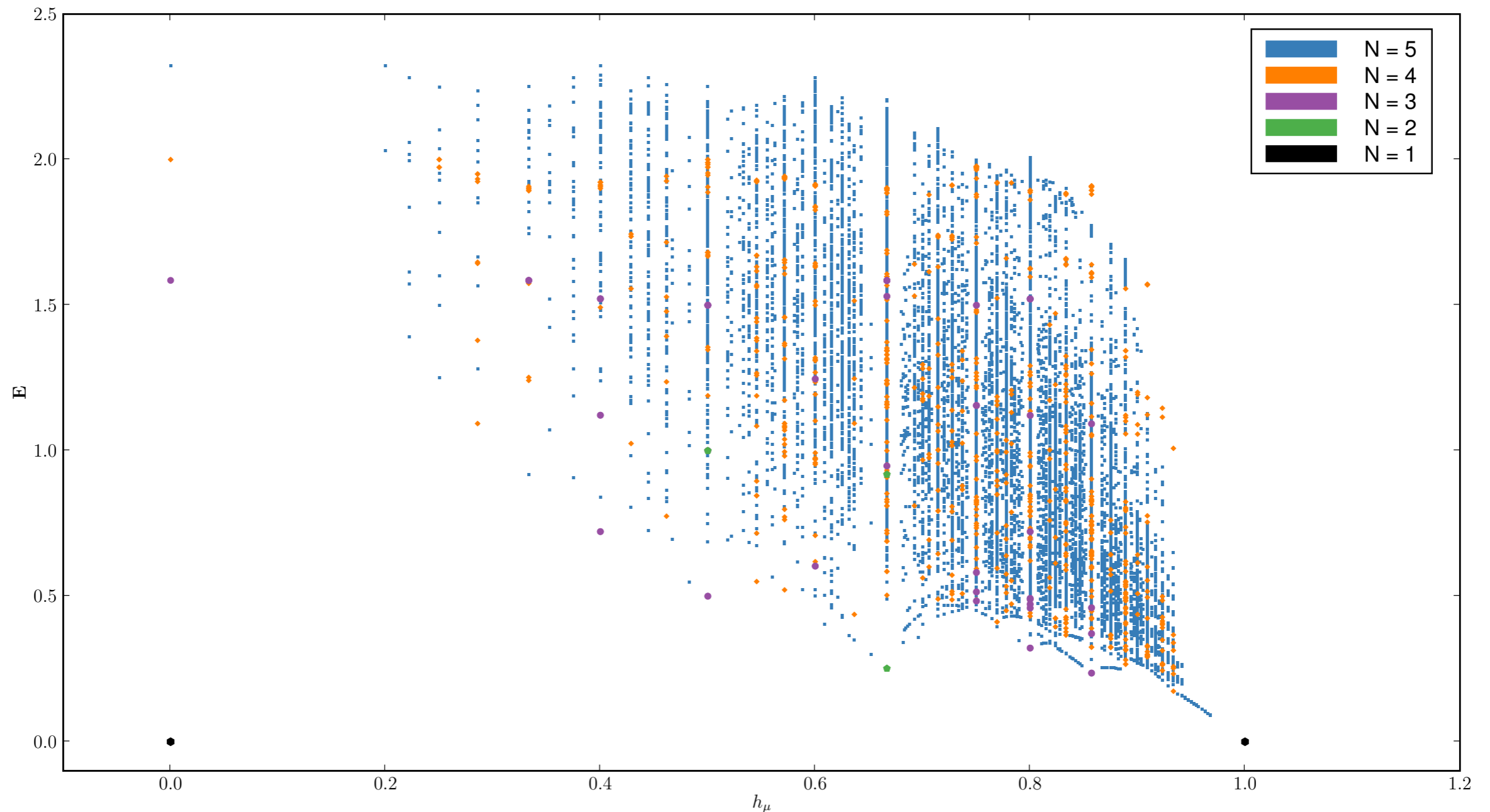
# Topics & Applications...

## Complexity-Entropy Diagram:

Analyze a class of processes ...

## $\epsilon$ -Machines

$E$  vs  $h_\mu$





# Topics & Applications...

## Computational Mechanics Applications by Others:

Chaotic Dynamical Systems

Symbolic Dynamics

Statistical Mechanical Models: Spin systems (Ising, glasses, ...)

Cellular Automata

Hidden Markov Models

Chaotic Crystallography

Hydrodynamics: Dripping faucet, turbulence

Quantum Dynamical Systems

Single Molecule Dynamics

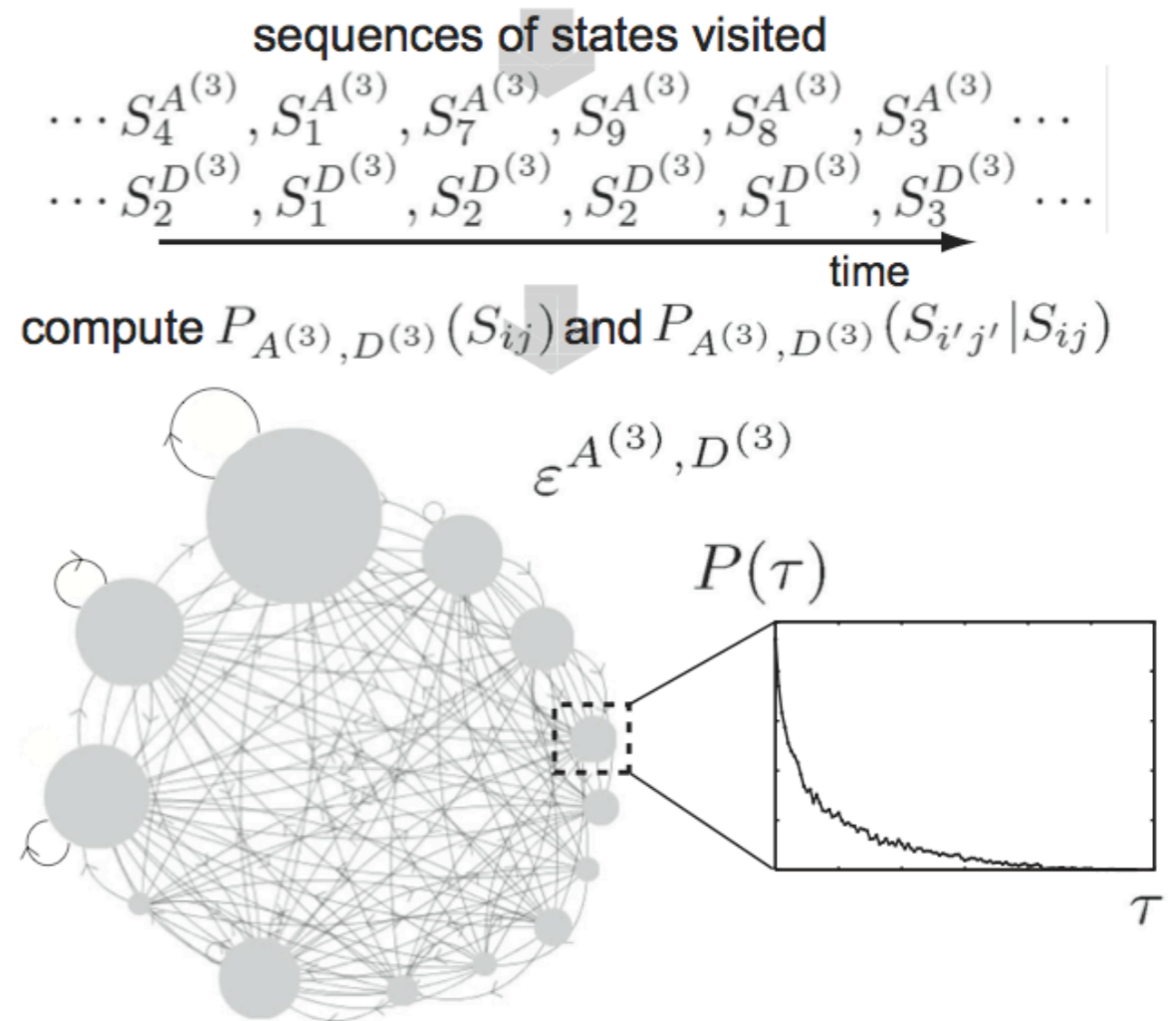
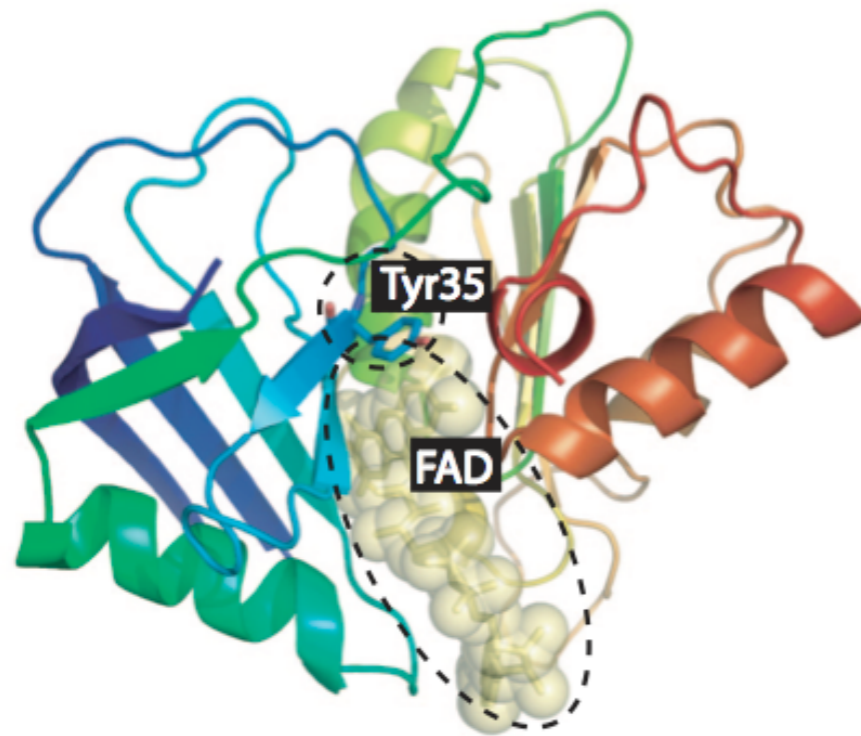
# Topics & Applications...

## Molecular Dynamics Spectroscopy:

### Multiscale complex network of protein conformational fluctuations in single-molecule time series

Chun-Biu Li<sup>\*†‡</sup>, Haw Yang<sup>§¶</sup>, and Tamiki Komatsuzaki<sup>\*†¶||</sup>

<sup>\*</sup>Nonlinear Sciences Laboratory, Department of Earth and Planetary Sciences, Faculty of Science, Kobe University, Nada, Kobe 657-8501, Japan; <sup>†</sup>Core Research for Evolutional Science and Technology (CREST), Japan Science and Technology Agency (JST), Kawaguchi, Saitama 332-0012, Japan; <sup>‡</sup>Department of Chemistry, University of California, Berkeley, CA 94720; and <sup>¶</sup>Physical Biosciences Division, Lawrence Berkeley National Laboratory, Berkeley, CA 94720



C.-B. Li, H. Yang, & T. Komatsuzaki, Proc. Natl. Acad. Sci USA **105**:2 (2008) 536–541.

# Topics & Applications...

## Computational Mechanics Research, Ongoing:

Novel materials

Quantum chaotic dynamics & measurement effects

Continuous processes

Spatiotemporal processes

Interactive learning

Network dynamics

Neurobiological processes

Multiagent systems

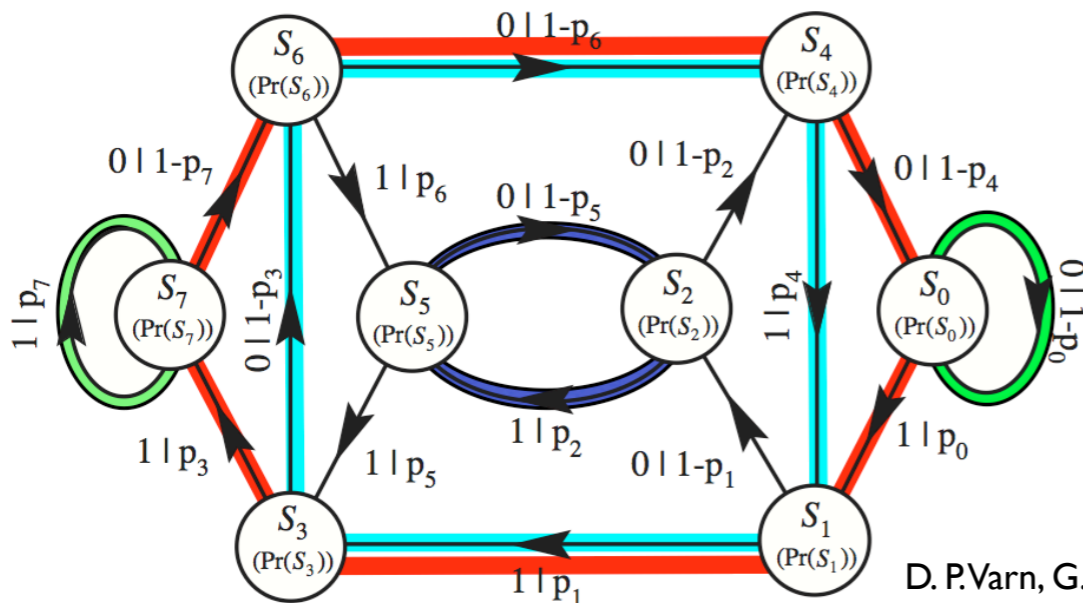
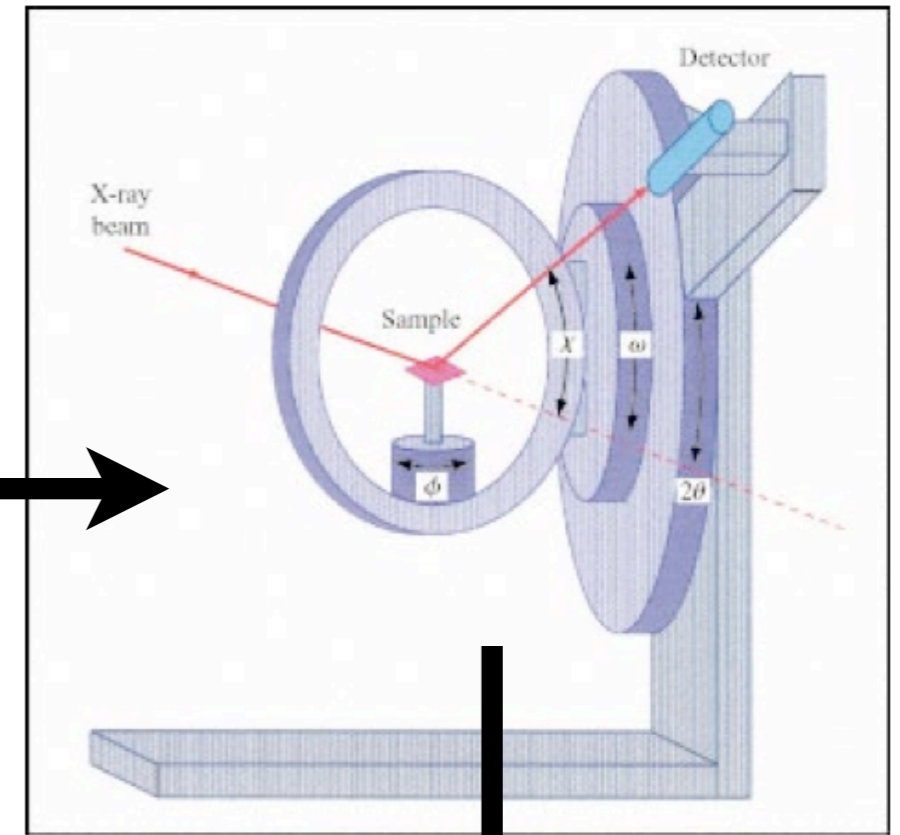
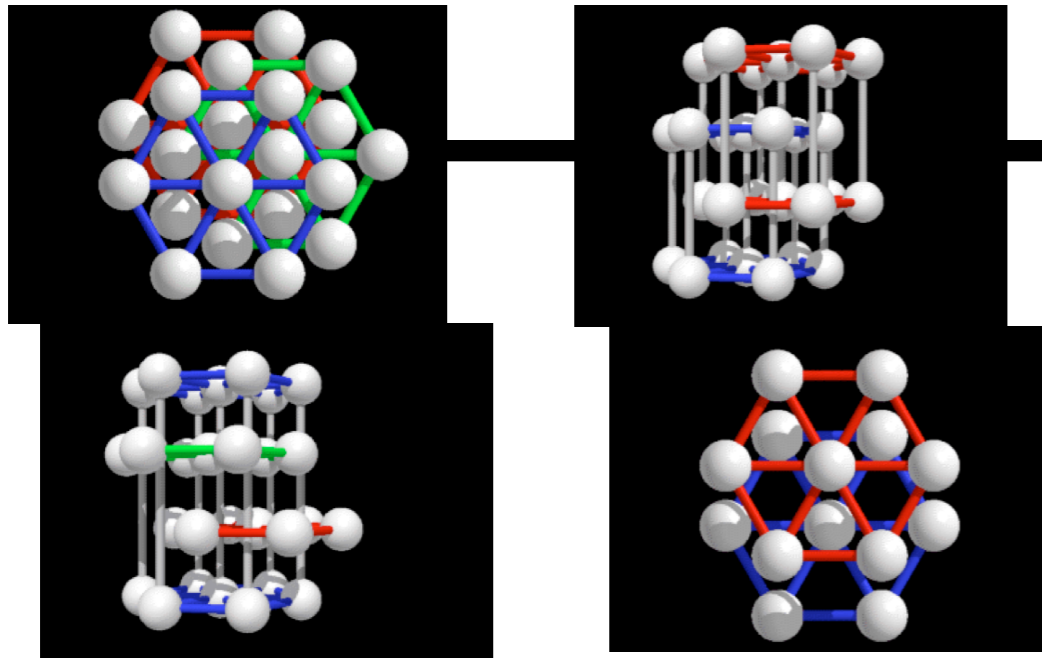
Distributed robotics

Evolutionary dynamics

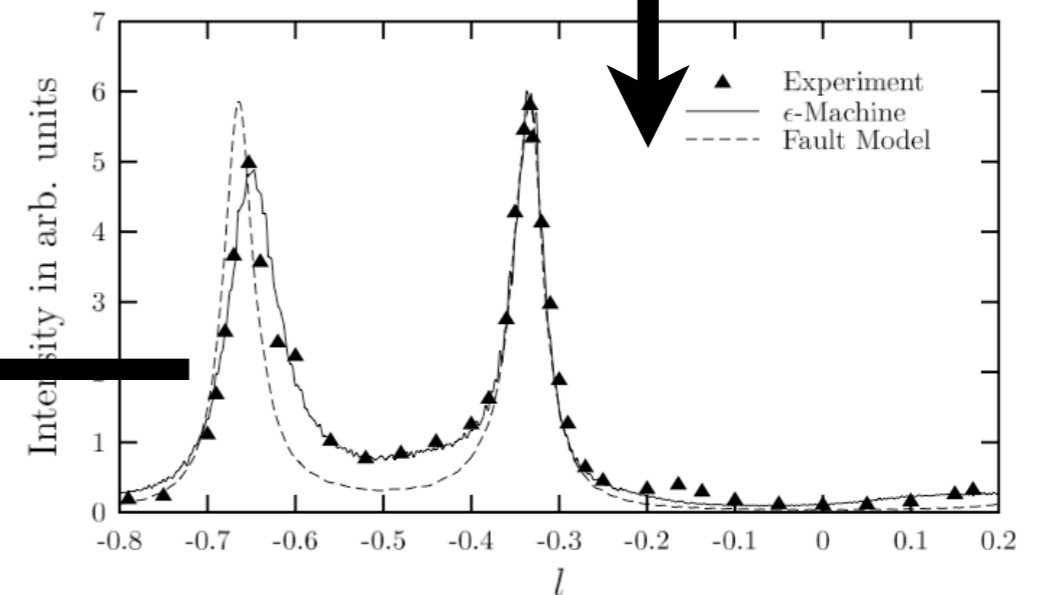
...

# Topics & Applications...

## Computational Mechanics Research ... Complex Materials & $\epsilon$ -Machine Spectral Reconstruction



$\epsilon$ -MSR



D. P. Varn, G. S. Canright, and J. P. Crutchfield, "epsilon-Machine spectral reconstruction theory: A direct method for inferring planar disorder and structure from X-ray diffraction studies", *Acta Cryst. Sec. A* **69**:2 (2013) 197-206.

# Topics & Applications...

## Computational Mechanics Research ...

Novel materials:

Use eMSR to design

Run eMSR backwards

Hypothesis: Structure key to properties.

But we can now analyze structure.

E.g., Design polytypes = Exotic semiconductors

# Topics & Applications...

Computational Mechanics Research ...

Spectral Decomposition of Intrinsic Computation:

Exact, closed-form expression of all information measures  
using  $\varepsilon$ -machine presentation.

How?

Mixed states + Eigenspectrum of  $\varepsilon$ -machine.

# Topics & Applications...

## Computational Mechanics Research ...

### Spectral decomposition of intrinsic computation ...

Recall mixed state expression:

$$\begin{aligned} h_\mu(L) &= H[X_{L-1} | X_0^{L-1}, S_0 \sim \mu_0(\lambda)] \\ &= H[X_{L-1} | (R_{L-1} | R_0 = \mu_0(\lambda))] \end{aligned}$$

Need only track evolution of mixed states via  $T^L$ .

Track analytically using spectral decomposition:

$$T^L = \sum_{\lambda \in \Lambda_T} \lambda^L T_\lambda$$

Eigenvalues:  $\Lambda_T \equiv \{\lambda \in \mathbb{C} : \det(\lambda I - T) = 0\}$

Projection operators:  $T_\lambda = \prod_{\substack{\zeta \in \Lambda_T \\ \zeta \neq \lambda}} \frac{T - \zeta I}{\lambda - \zeta}$

J. P. Crutchfield, C. J. Ellison, P. Riechers in preparation.

# Topics & Applications...

## Computational Mechanics Research ...

### Spectral decomposition of intrinsic computation ...

$$\begin{aligned} \text{Excess entropy: } \mathbf{E} &\equiv \sum_{L=1}^{\infty} [h_{\mu}(L) - h_{\mu}] \\ &= \sum_{\substack{\lambda \in \Lambda_T \\ |\lambda| < 1}} \frac{1}{1 - \lambda} \langle \delta_{\pi} | T_{\lambda} | H(\{T^{(s)}\}) \rangle \end{aligned}$$

### Synchronization information:

$$\mathbf{S} = \begin{cases} \sum_{\substack{\lambda \in \Lambda_T \\ |\lambda| < 1}} \frac{1}{1 - \lambda} \langle \delta_{\pi} | T_{\lambda} | H[\eta] \rangle & \text{if } \langle \pi_T | H[\eta] \rangle = 0 \\ \infty & \text{otherwise} \end{cases}$$

$$\begin{aligned} \text{Transient information: } \mathbf{T} &\equiv \sum_{L=1}^{\infty} L [h_{\mu}(L) - h_{\mu}] \\ &= \sum_{\substack{\lambda \in \Lambda_T \\ |\lambda| < 1}} \frac{1}{(1 - \lambda)^2} \langle \delta_{\pi} | T_{\lambda} | H(\{T^{(s)}\}) \rangle \end{aligned}$$

J. P. Crutchfield, C. J. Ellison, P. Riechers in preparation.



# Topics & Applications...

## Computational Mechanics Research ...

$\varepsilon$ -Transducers: Input-output functions!

$\varepsilon$ -Machines: Behaviors and distributions over them

Now, mappings on stochastic behaviors

Optimal, minimal, unique presentation of mappings

Noisy communication channel:



Input process

$\dots, X_{-1}, X_0, X_1, \dots$

Output process

$\dots, Y_{-1}, Y_0, Y_1, \dots$

Joint process

$\dots, (X, Y)_{-1}, (X, Y)_0, (X, Y)_1, \dots$

# Topics & Applications...

## Computational Mechanics Research ...

### $\epsilon$ -Transducers ...

Transitions:  $x|p|y$

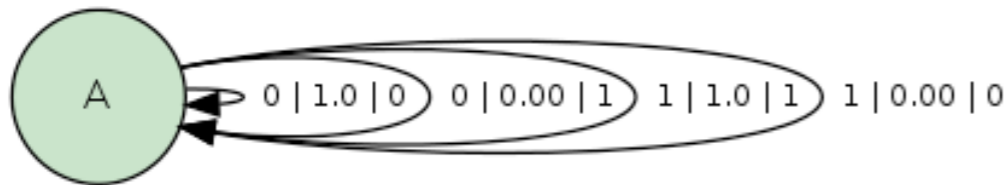
$x$  is input symbol,

$y$  is output symbol, and

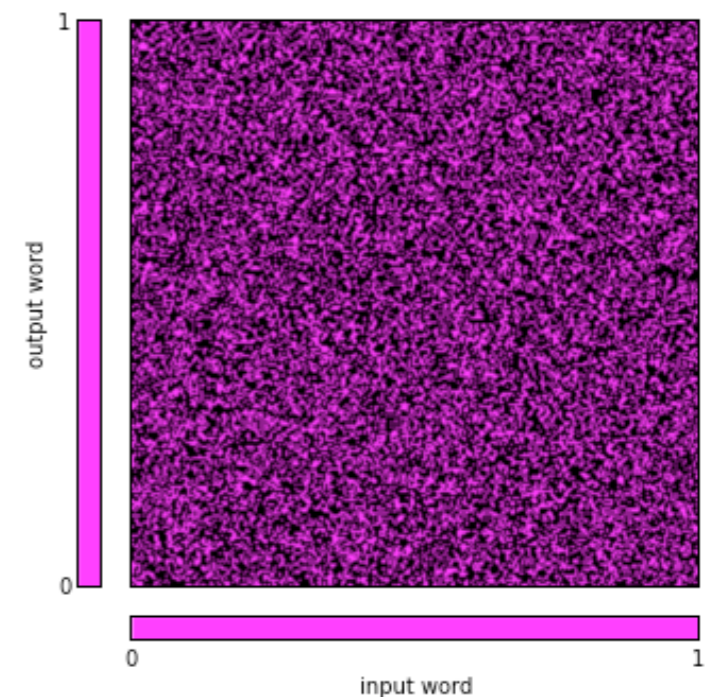
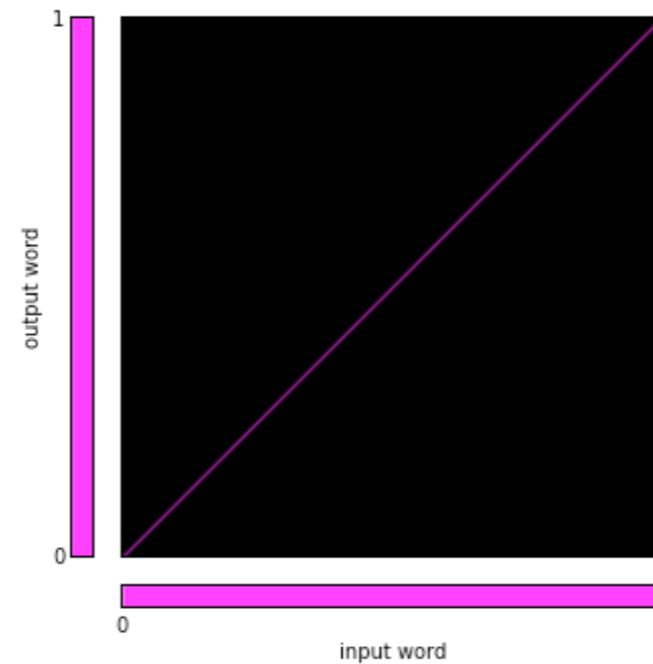
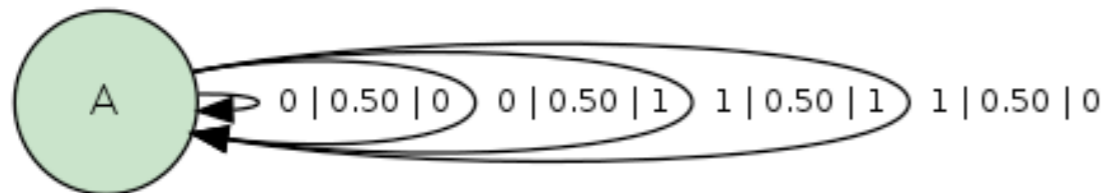
$p$  is probability

Memoryless channels:

Identity channel.



All-to-fair channel.



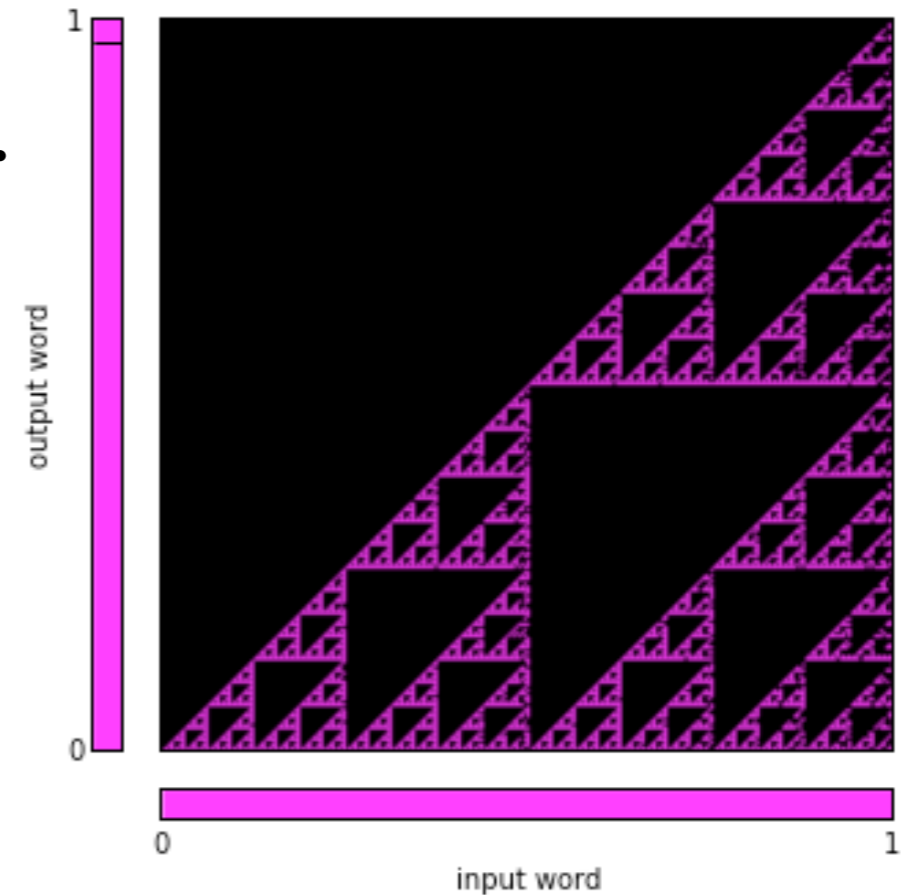
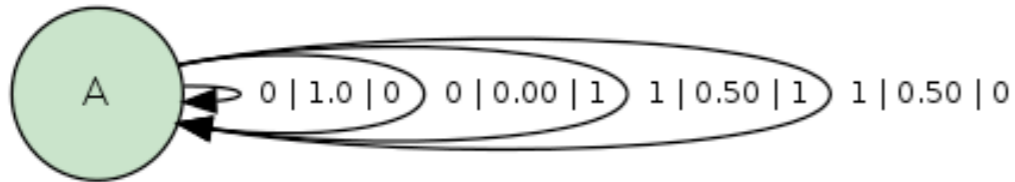
# Topics & Applications...

## Computational Mechanics Research ...

### $\epsilon$ -Transducers ...

Memoryless channels ...

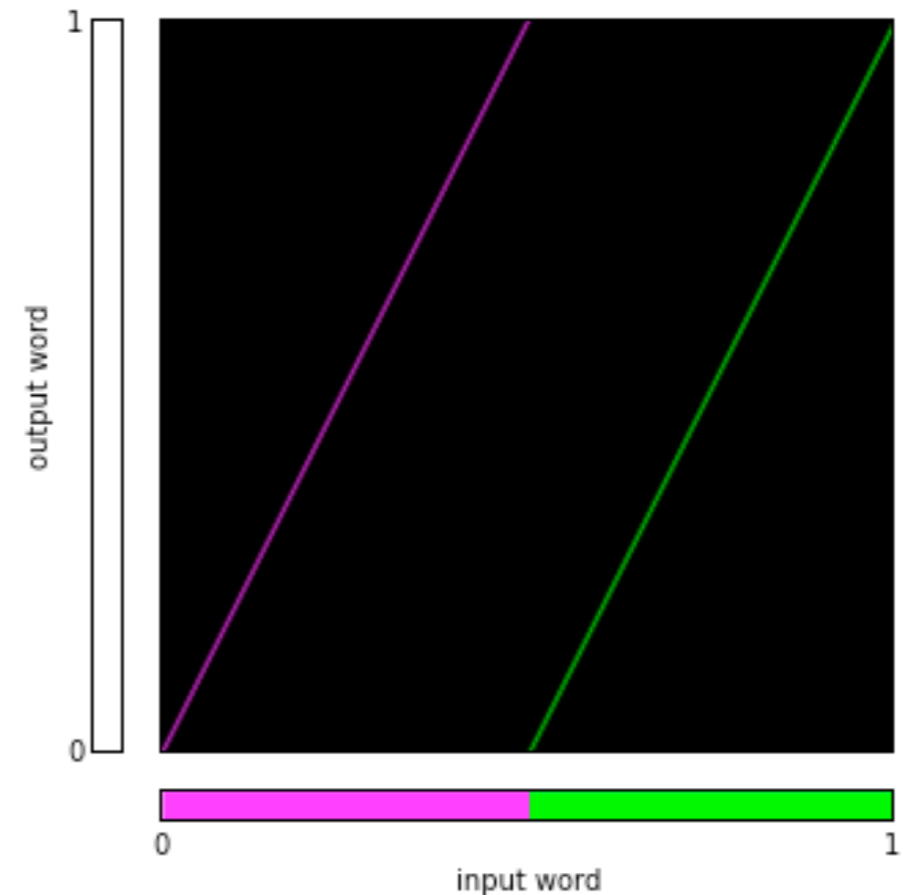
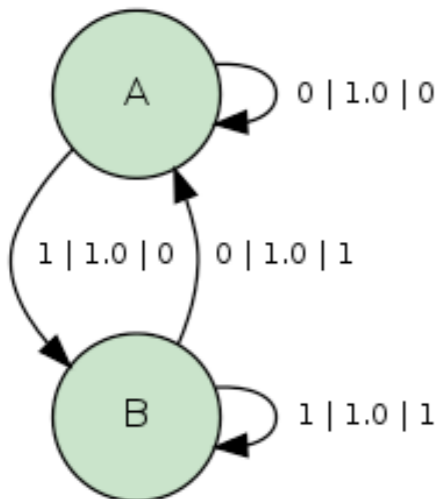
Z Channel: Fuzz out ones, transmit zeros.



Memoryful Channel:

Delay-by-1 Channel:

Input is order-1 Markov, output nonsynchronizing.



# Topics & Applications...

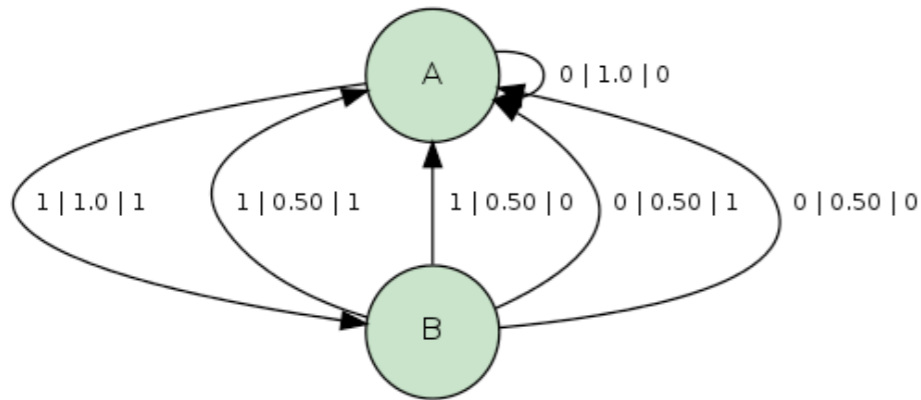
## Computational Mechanics Research ...

### $\epsilon$ -Transducers ...

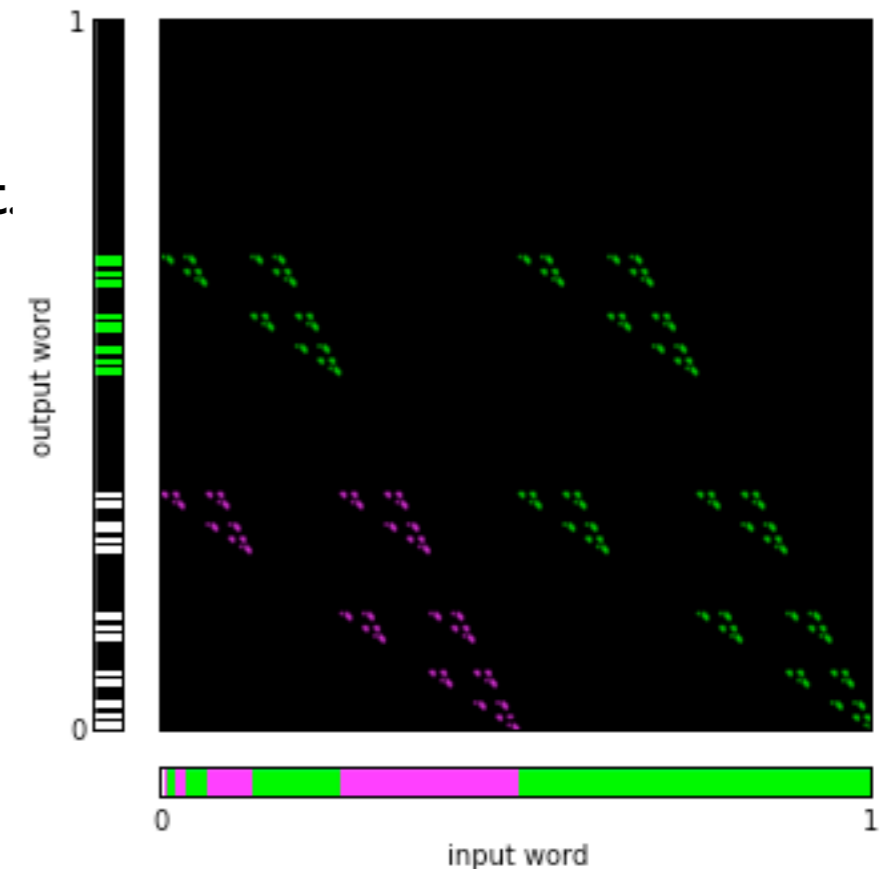
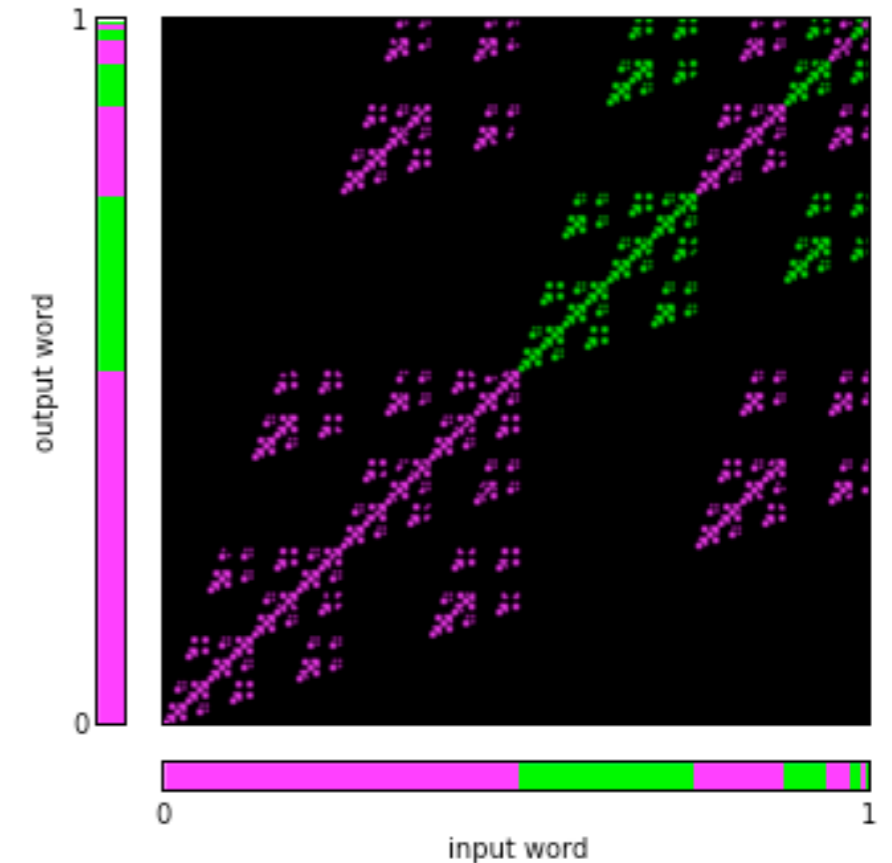
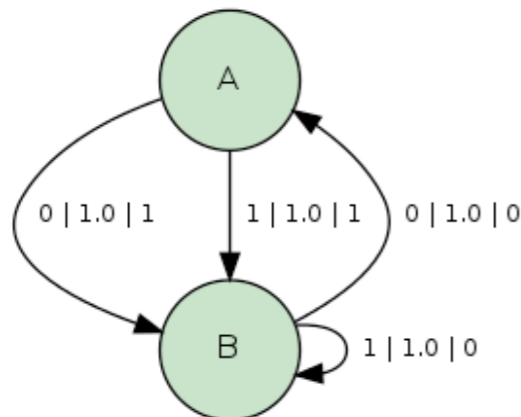
Memoryful channels ...

Even-to-Fair: Map Even Process to Fair Coin.

Random bits from Even Process, but fuzz out pairs of 1s.



Feedback-NOR: NOR of the previous output & current input. Strictly SOPIC on input, nonsynchronizing on output, but order-1 Markov on joint symbols.

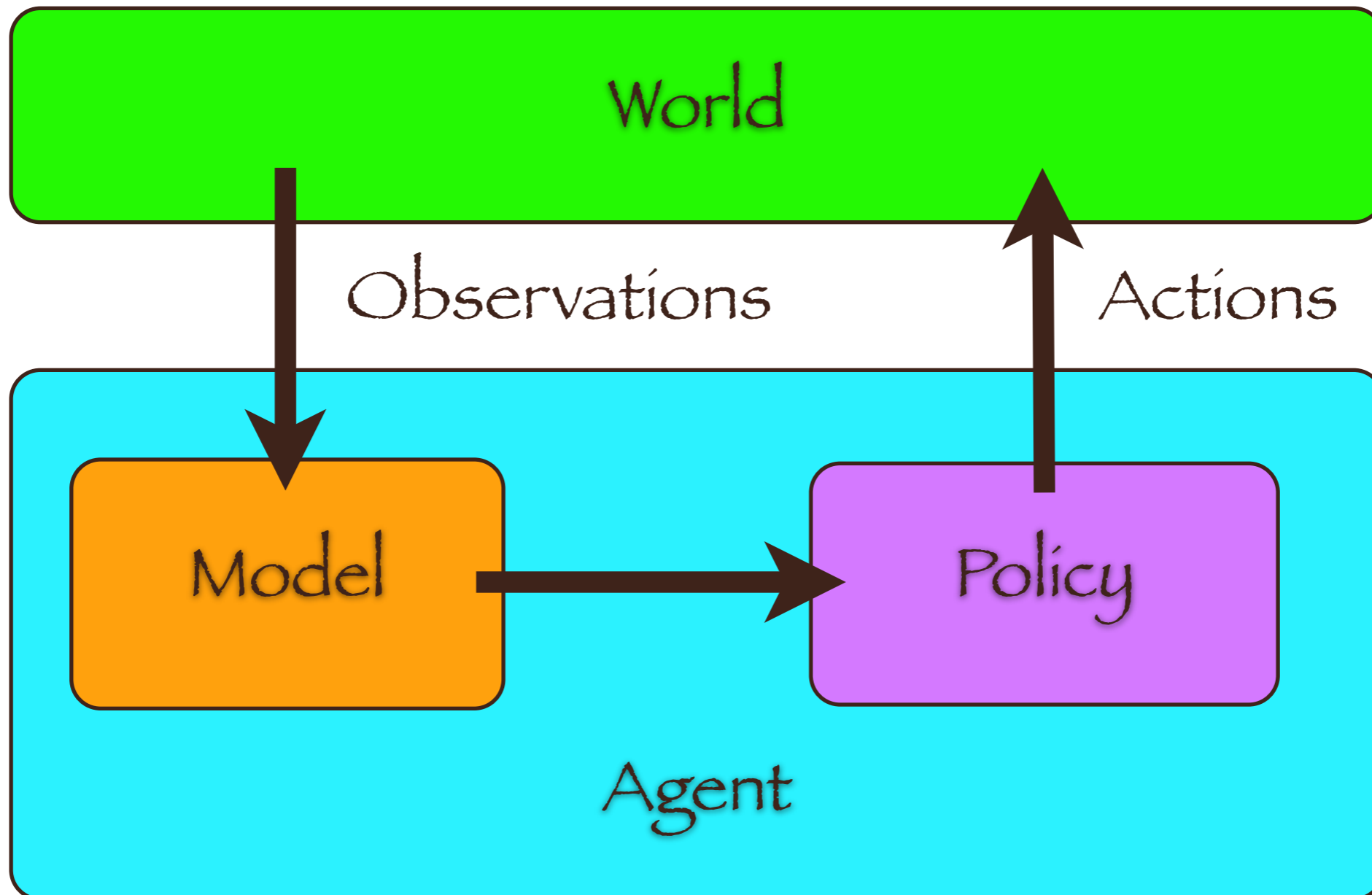


# Topics & Applications...

## Computational Mechanics Research ...

### Interactive Learning:

#### The Feedback Loop



# Topics & Applications...

## Computational Mechanics Research ...

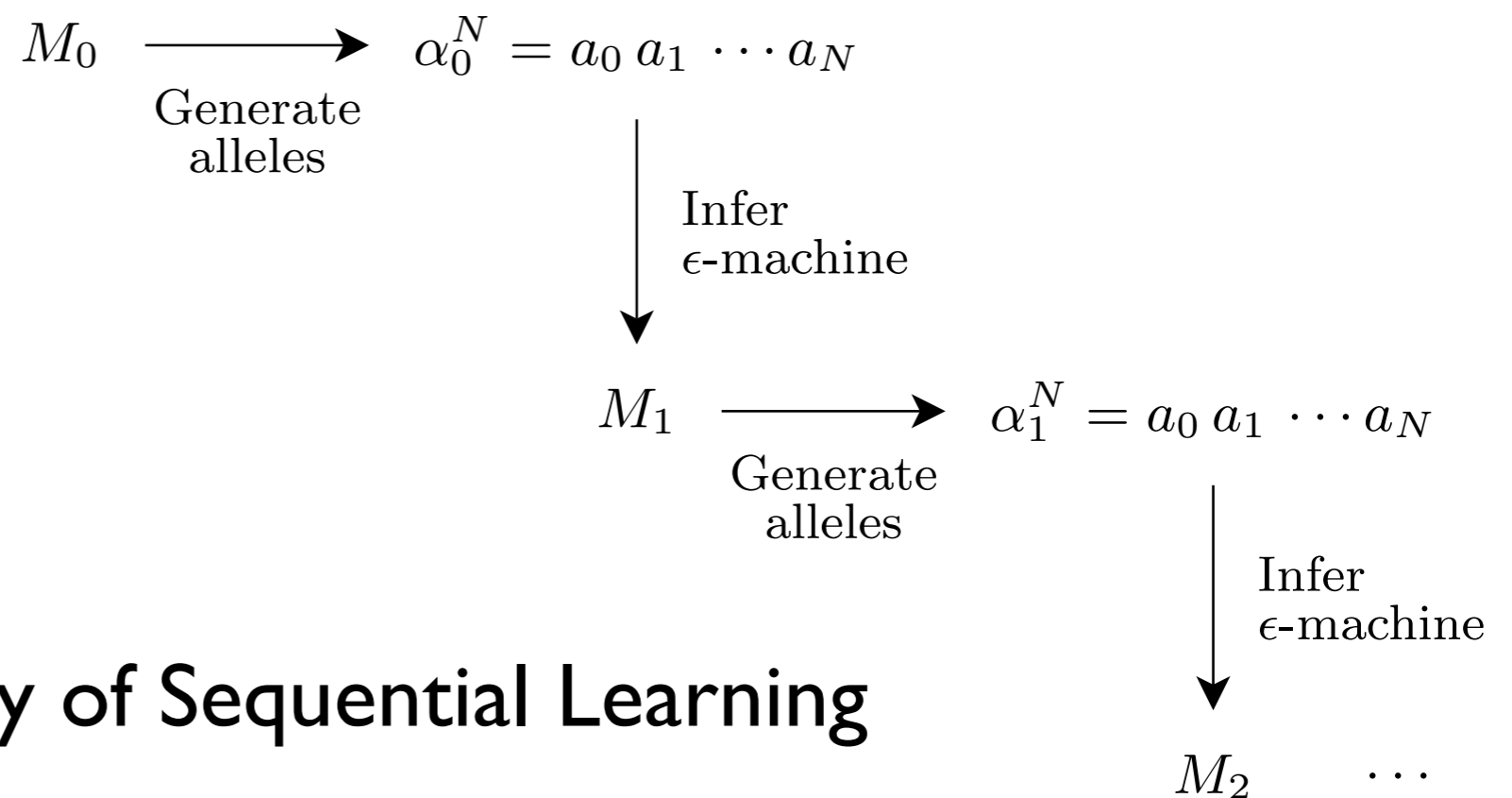
### The Evolution of Language:

Heard over radio during WWI:

*“Send Three- and Four-Pence, We’re Going to a Dance”*

Intended message:

*“Send reinforcements we’re going to advance”*



### Game of Telephone:

## Mathematical Theory of Sequential Learning

J. P. Crutchfield and S. Whalen, "Structural Drift: The Population Dynamics of Sequential Learning", PLoS Computational Biology **8**:6 (2012) e1002510.

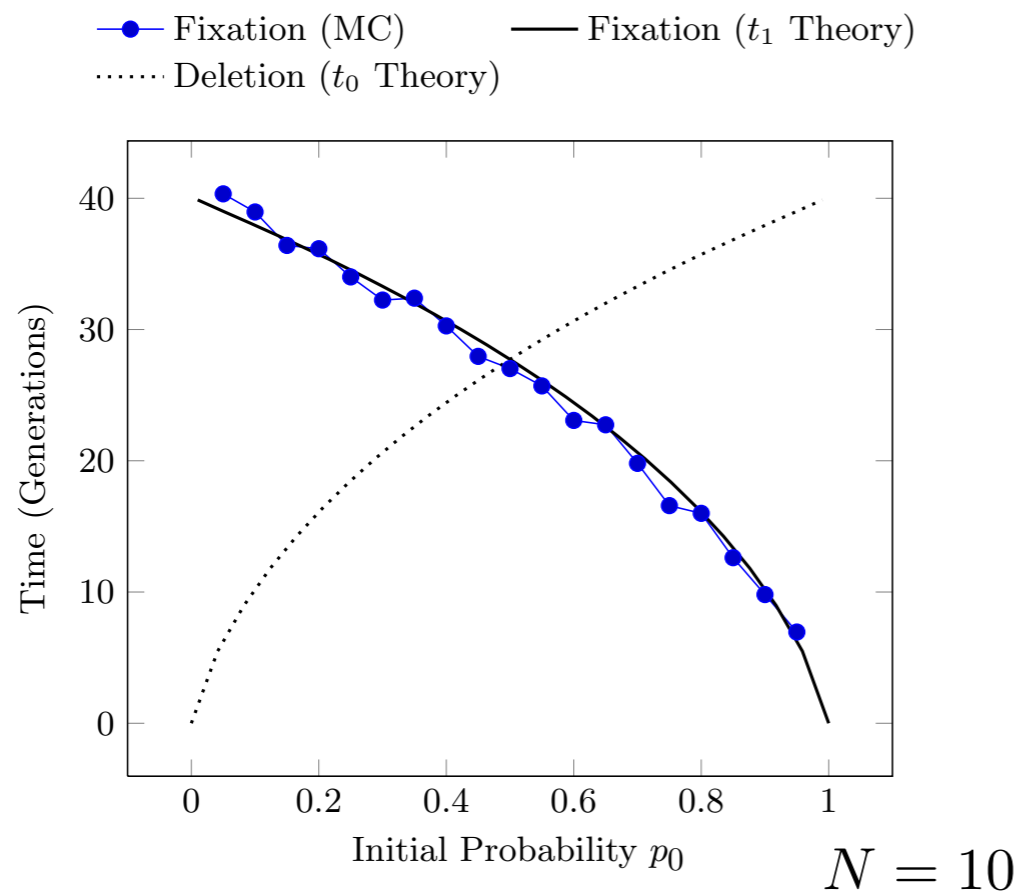
# Topics & Applications...

Computational Mechanics Research ...

The Evolution of Language ...

Mathematical Theory of Sequential Learning

## Genetic Drift



(Theory: Kimura)

$$t_1(p_0) = -\frac{1}{p_0} [4N_e(1 - p_0) \log(1 - p_0)]$$

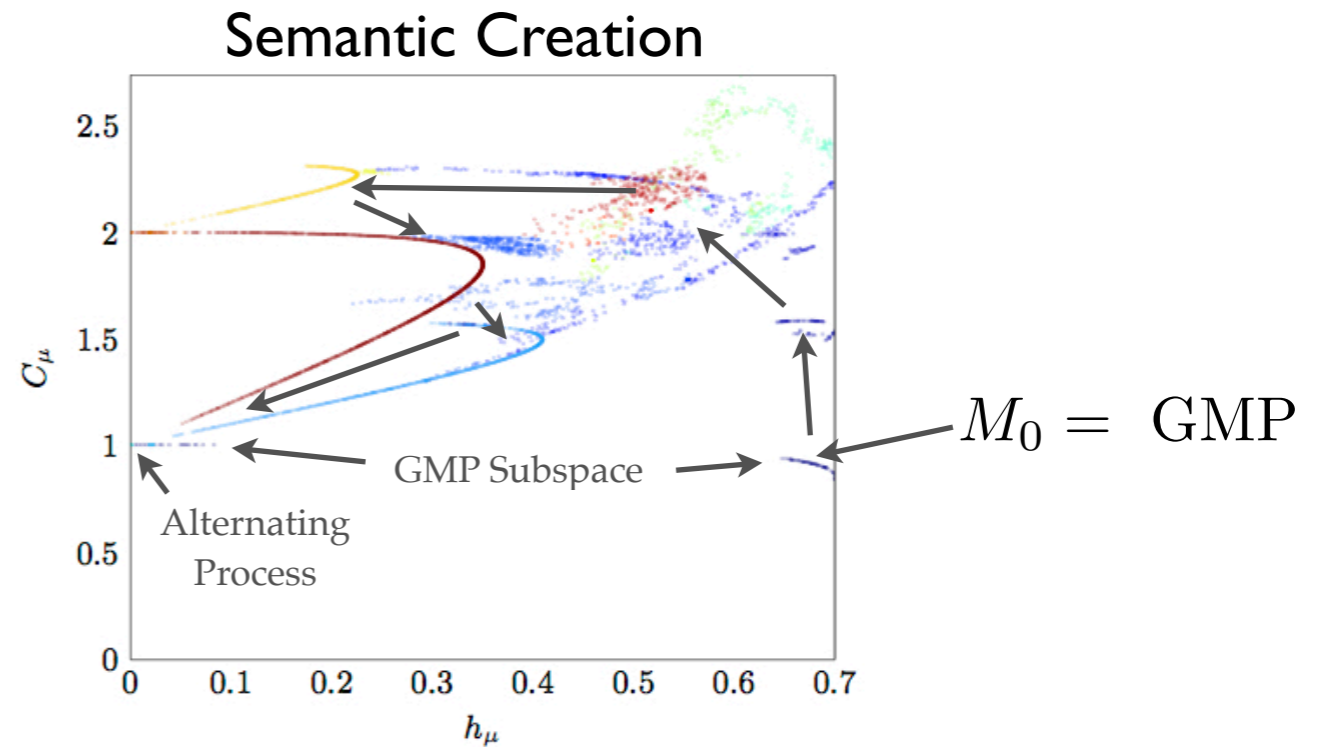
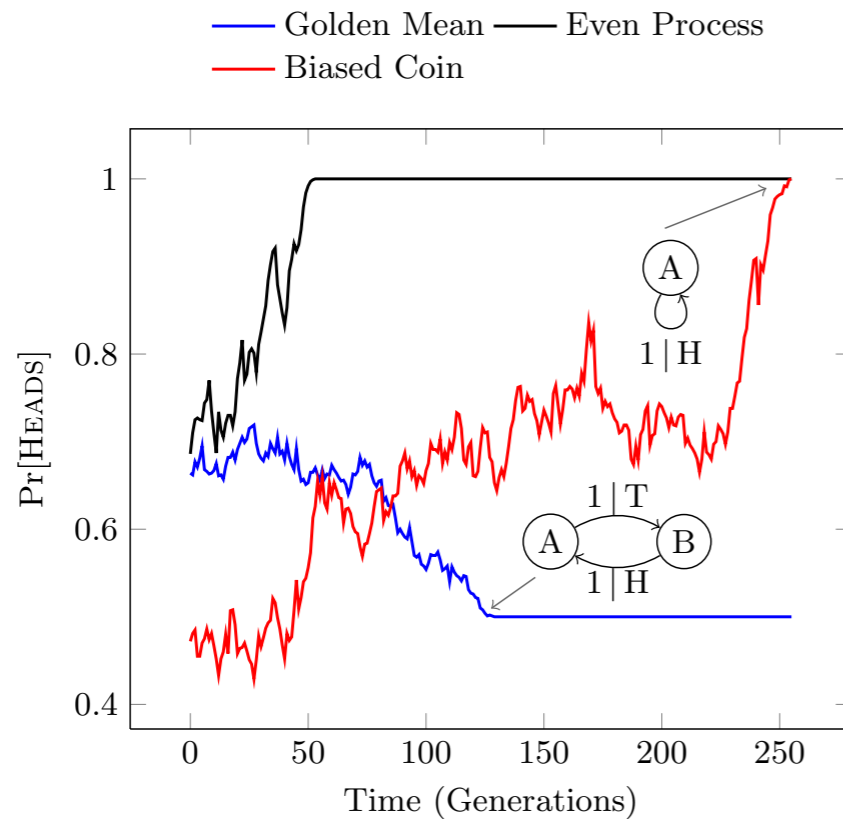
# Topics & Applications...

## Computational Mechanics Research:

### The Evolution of Language ...

### Sequential learning as a diffusion in space of $\epsilon$ -Machines

### Structural Drift





# Topics & Applications...

## Computational Mechanics Research: The Evolution of Language ...

Start with an English phrase:

find equilibrium

translationparty.com

my talk is almost done

let's go!

私の話はほとんど行われて

into Japanese

Most of my talk was done

back into English

私の話のほとんど行われていた

back into Japanese

I was almost done

back into English

ほとんど行われていた

back into Japanese

Little had been done

back into English

ほとんど行われていた

back into Japanese

Little had been done

back into English

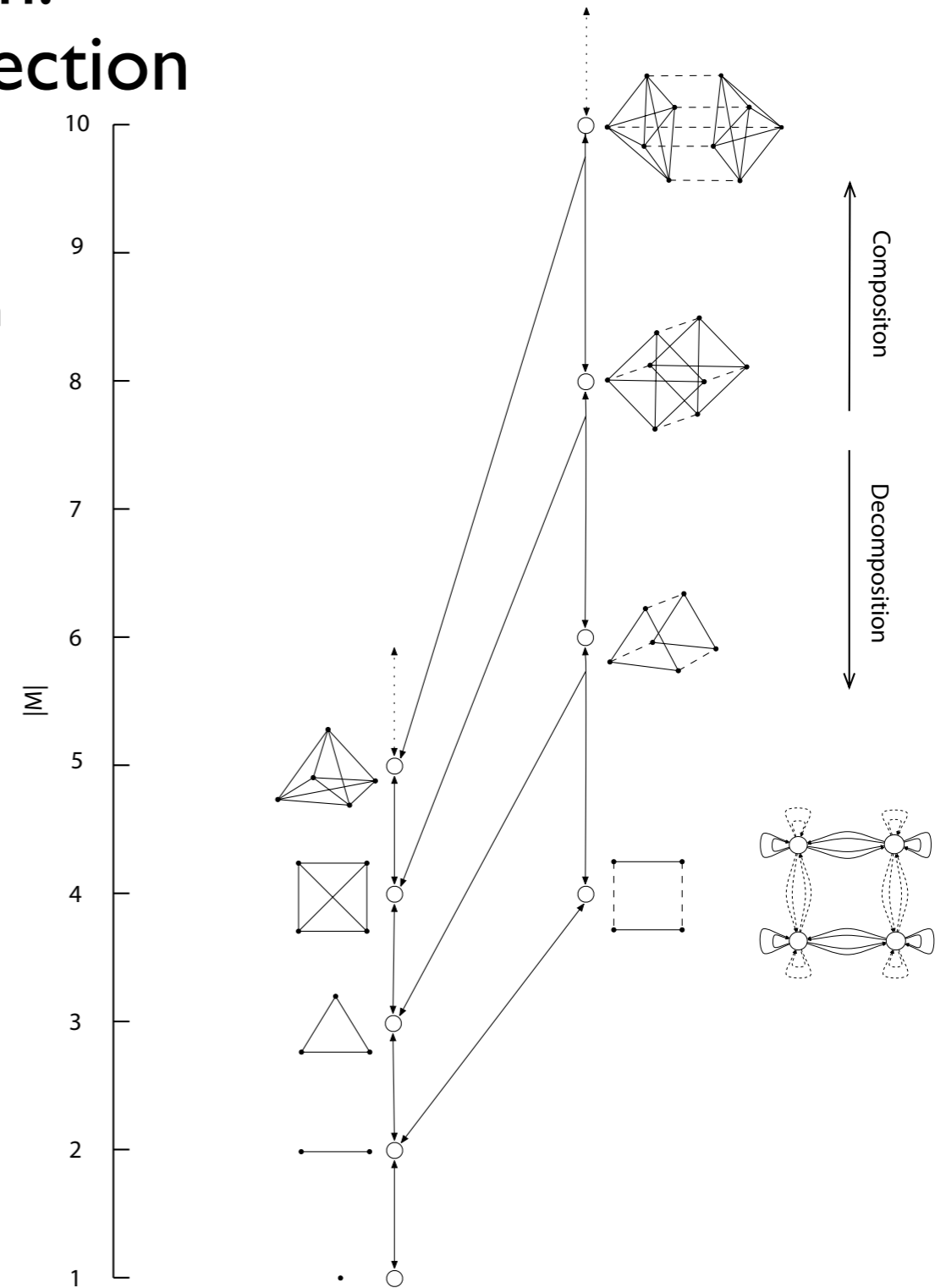
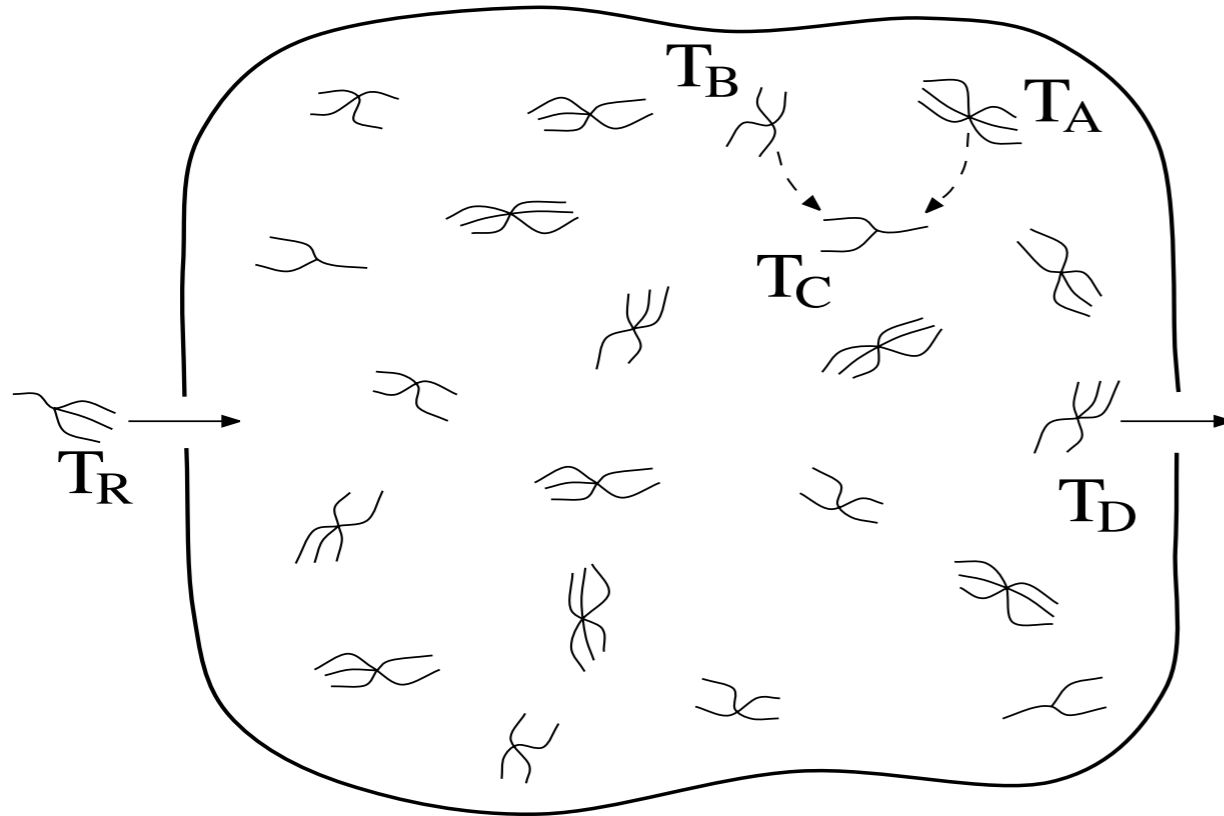
**Equilibrium found!**

You've done this before, haven't you.

# Topics & Applications...

## Computational Mechanics Research: Emergence of Evolutionary Selection

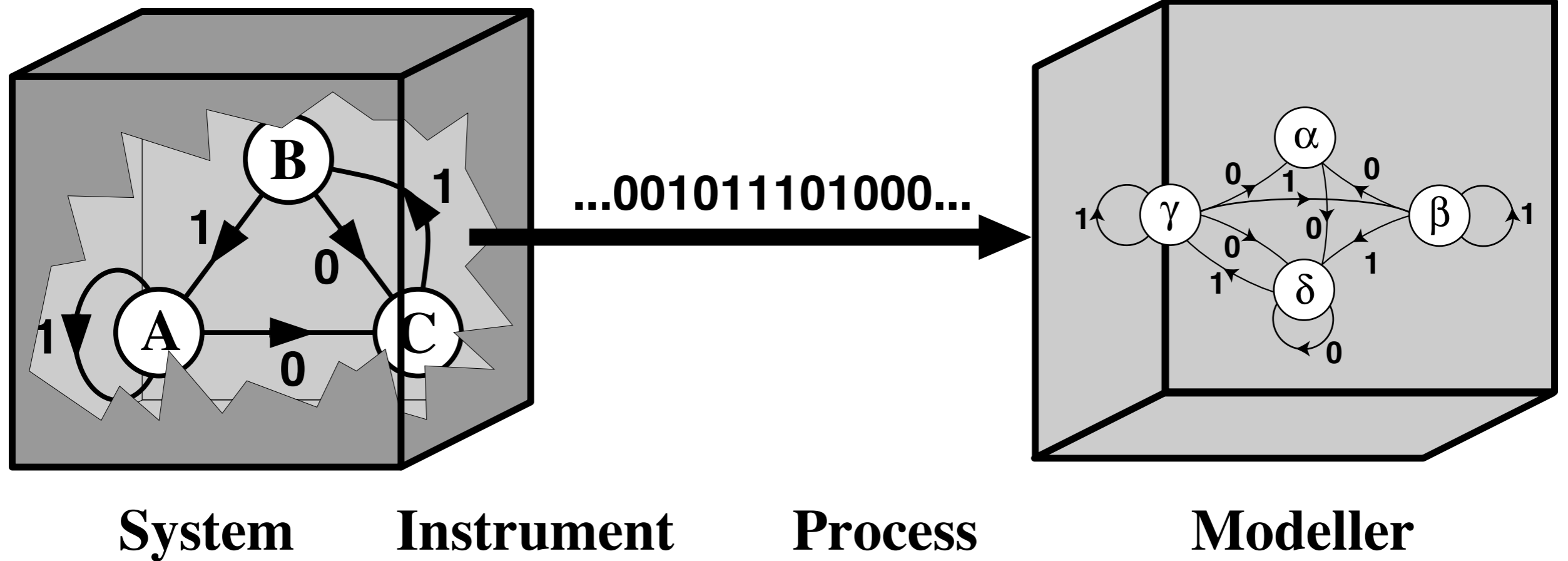
Self-Organizing Networks  
Evolutionary Theory:  
From the Primordial Soup to Natural Selection



J. P. Crutchfield and O. Goernerup, "Objects That Make Objects: The Population Dynamics of Structural Complexity", J. Royal Society Interface **3** (2006) 345-349.

# Topics & Applications...

## The Message:



How Nature is Organized  
is How Nature Computes

# Topics & Applications...

Further:

*Computational Mechanics Archive:*

<http://csc.ucdavis.edu/~cmg/>

*Dynamics of Learning Research Group Mail List:*

<http://lists.csc.ucdavis.edu/mailman/listinfo/dynlearn>