

Lecture 4 Langton's Investigation

Under what conditions can we expect a complex dynamics of information to emerge spontaneously and come to dominate the behavior of a CA?

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Approach

- Investigate 1D CAs with:
 - random transition rules
 - starting in random initial states
- Systematically vary a simple parameter characterizing the rule
- Evaluate qualitative behavior (Wolfram class)

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Assumptions

- Periodic boundary conditions
 - no special place
- Strong quiescence:
 - if all the states in the neighborhood are the same, then the new state will be the same
 - persistence of uniformity
- Spatial isotropy:
 - all rotations of neighborhood state result in same new state
 - no special direction
- Totalistic [not used by Langton]:
 - depend only on sum of states in neighborhood
 - implies spatial isotropy

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Langton's Lambda

- Designate one state to be quiescent state
- Let K = number of states
- Let $N = 2r + 1$ = area of neighborhood
- Let $T = K^N$ = number of entries in table
- Let n_q = number mapping to quiescent state
- Then

$$\lambda = \frac{T - n_q}{T}$$

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Range of Lambda Parameter

- If *all* configurations map to quiescent state:
 $\lambda = 0$
- If *no* configurations map to quiescent state:
 $\lambda = 1$
- If every state is represented *equally*:
 $\lambda = 1 - 1/K$
- A sort of measure of “excitability”


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Example

- States: $K = 5$
- Radius: $r = 1$
- Initial state: random
- Transition function: random (given λ)

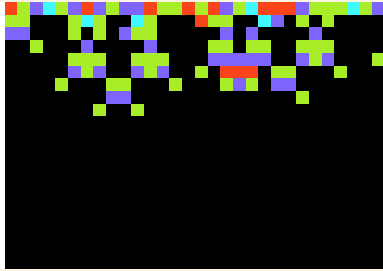
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Class I ($\lambda = 0.2$)

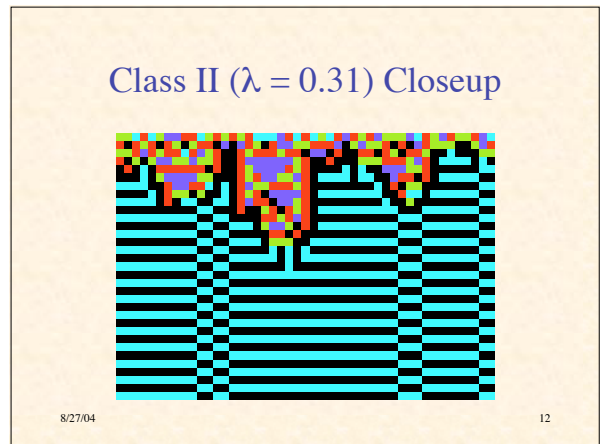
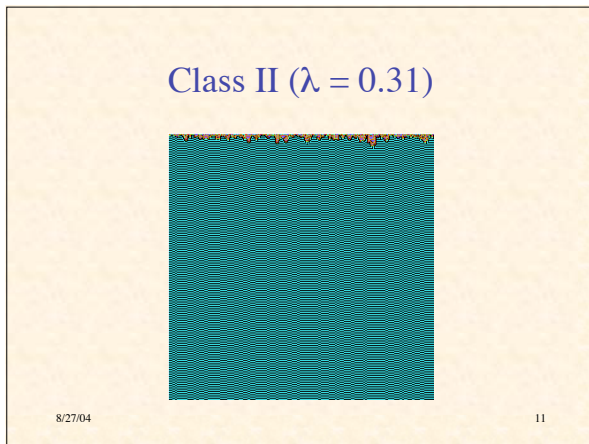
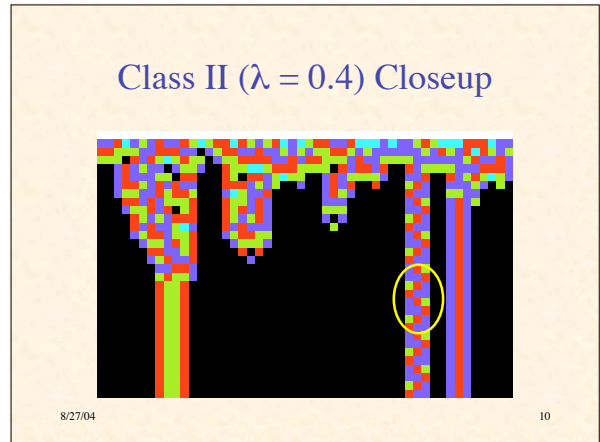
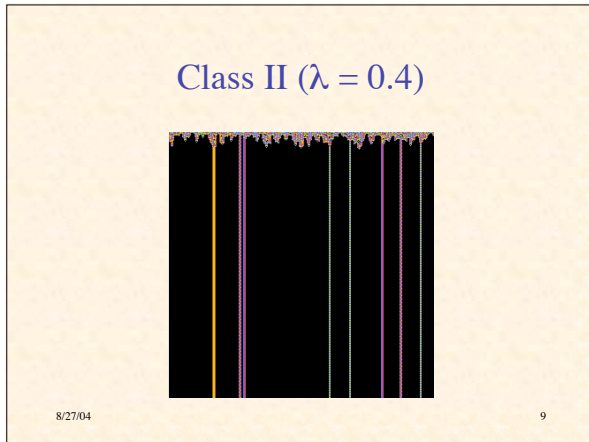


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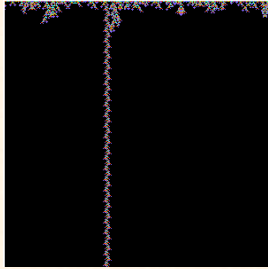
Class I ($\lambda = 0.2$) Closeup



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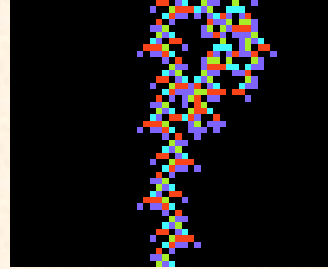
Class II ($\lambda = 0.37$)



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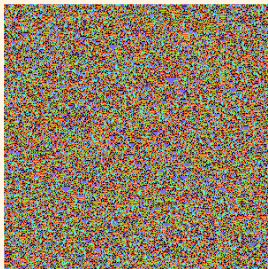
Class II ($\lambda = 0.37$) Closeup



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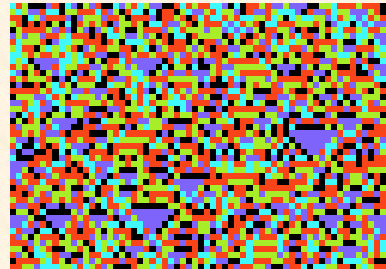
Class III ($\lambda = 0.5$)



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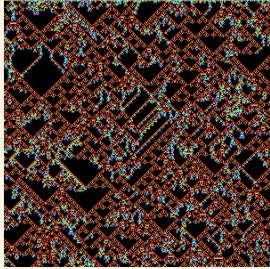
Class III ($\lambda = 0.5$) Closeup



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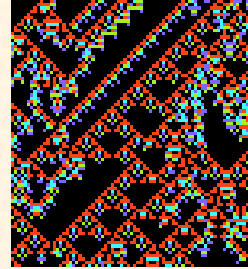
Class IV ($\lambda = 0.35$)



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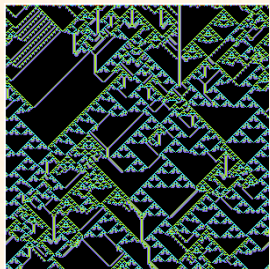
Class IV ($\lambda = 0.35$) Closeup



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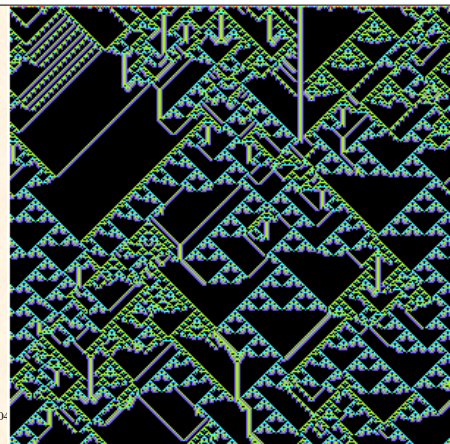
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Class IV ($\lambda = 0.34$)



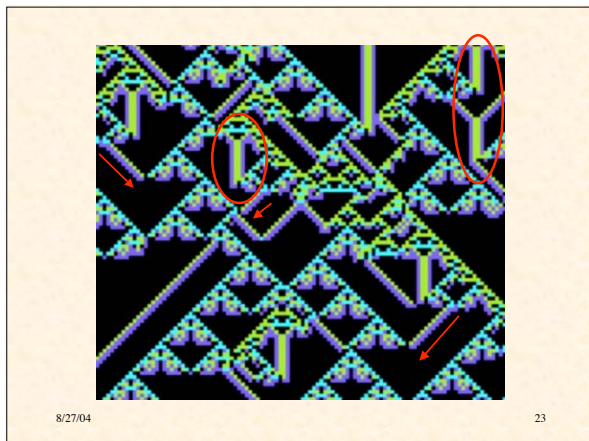
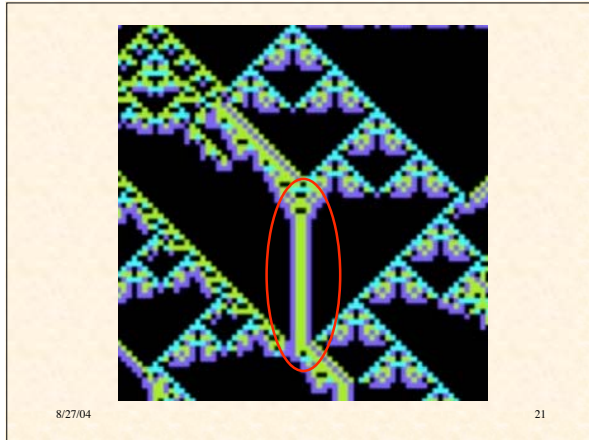
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Class IV Shows Some of the Characteristics of Computation

- Persistent, but not perpetual storage
- Terminating cyclic activity
- Global transfer of control/information

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λ of Life

- For Life, $\lambda \approx 0.273$
- which is near the critical region for CAs with:
 - $K = 2$
 - $N = 9$

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Transient Length (I, II)

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Transient Length (III)

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Shannon Information (very briefly!)

- Information varies directly with surprise
- Information varies inversely with probability
- Information is additive
- ∴ The information content of a message is proportional to the negative log of its probability

$$I\{s\} = -\lg \Pr\{s\}$$

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Entropy

- Suppose have source S of symbols from ensemble $\{s_1, s_2, \dots, s_N\}$
- Average information per symbol:

$$\sum_{k=1}^N \Pr\{s_k\} I\{s_k\} = \sum_{k=1}^N \Pr\{s_k\} (-\lg \Pr\{s_k\})$$
- This is the *entropy* of the source:

$$H\{S\} = -\sum_{k=1}^N \Pr\{s_k\} \lg \Pr\{s_k\}$$

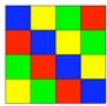
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Maximum and Minimum Entropy

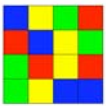
- Maximum entropy is achieved when all signals are equally likely
 No ability to guess; maximum surprise
 $H_{\max} = \lg N$
- Minimum entropy occurs when one symbol is certain and the others are impossible
 No uncertainty; no surprise
 $H_{\min} = 0$

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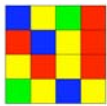
Entropy Examples



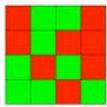
$H = 2.0$ bits



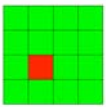
$H = 2.0$ bits



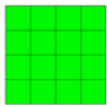
$H = 1.9$ bits



$H = 1.0$ bits



$H = 0.3$ bits



$H = 0.0$ bits

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Entropy of Transition Rules

- Among other things, a way to measure the uniformity of a distribution

$$H = -\sum_i p_i \lg p_i$$
- Distinction of quiescent state is arbitrary
- Let n_k = number mapping into state k
- Then $p_k = n_k / T$

$$H = \lg T - \frac{1}{T} \sum_{k=1}^K n_k \lg n_k$$

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Entropy Range

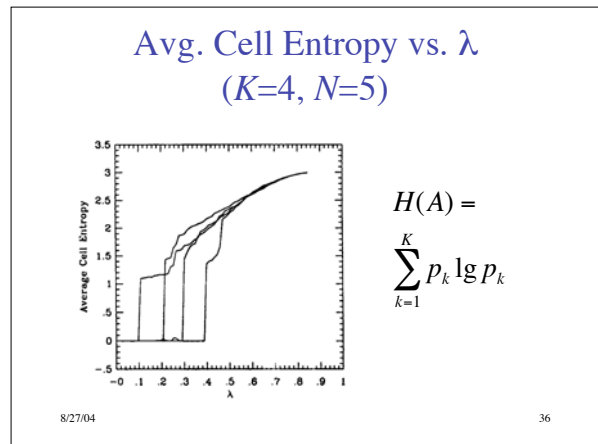
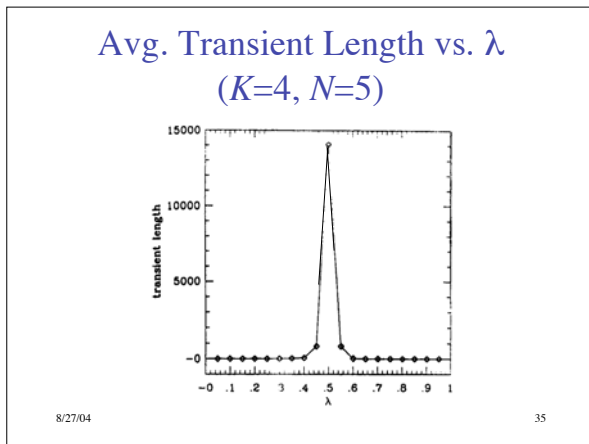
- Maximum entropy ($\lambda = 1 - 1/K$):
 uniform as possible
 all $n_k = T/K$
 $H_{\max} = \lg K$
- Minimum entropy ($\lambda = 0$ or $\lambda = 1$):
 non-uniform as possible
 one $n_s = T$
 all other $n_r = 0$ ($r \neq s$)
 $H_{\min} = 0$

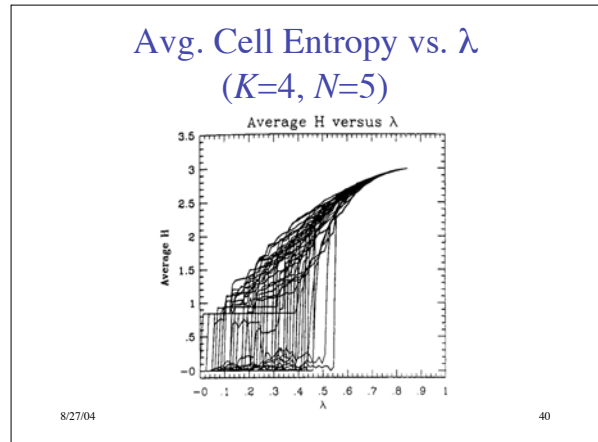
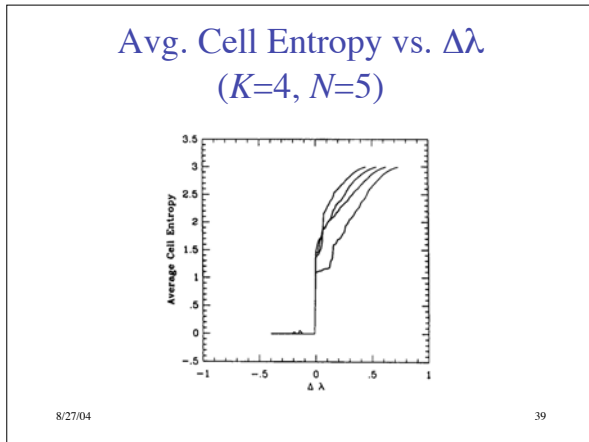
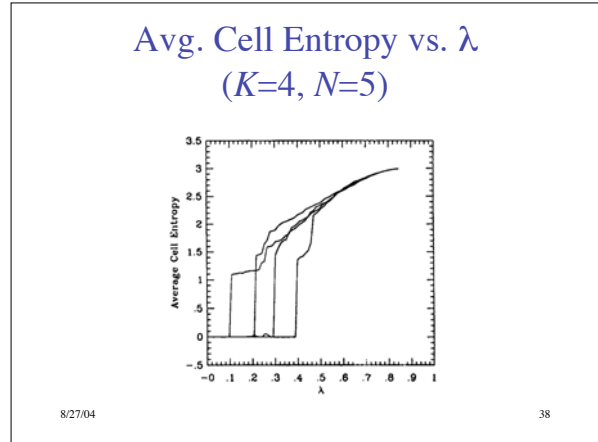
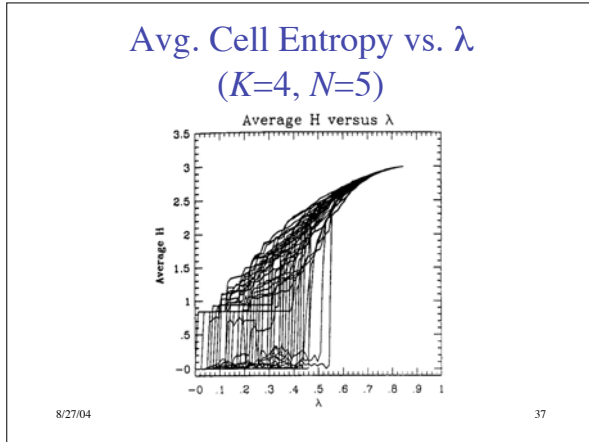
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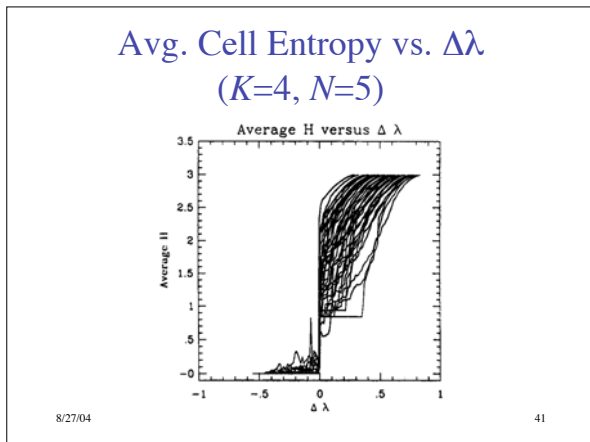
Further Investigations by Langton

- 2-D CAs
- $K = 8$
- $N = 5$
- 64×64 lattice
- periodic boundary conditions

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Entropy of Independent Systems

- Suppose sources A and B are independent
- Let $p_j = \Pr\{a_j\}$ and $q_k = \Pr\{b_k\}$
- Then $\Pr\{a_j, b_k\} = \Pr\{a_j\} \Pr\{b_k\} = p_j q_k$

$$H(A,B) = \sum_{j,k} \Pr(a_j, b_k) \lg \Pr(a_j, b_k)$$

$$= \sum_{j,k} p_j q_k \lg(p_j q_k) = \sum_{j,k} p_j q_k (\lg p_j + \lg q_k)$$

$$= \sum_j p_j \lg p_j + \sum_k q_k \lg q_k = H(A) + H(B)$$

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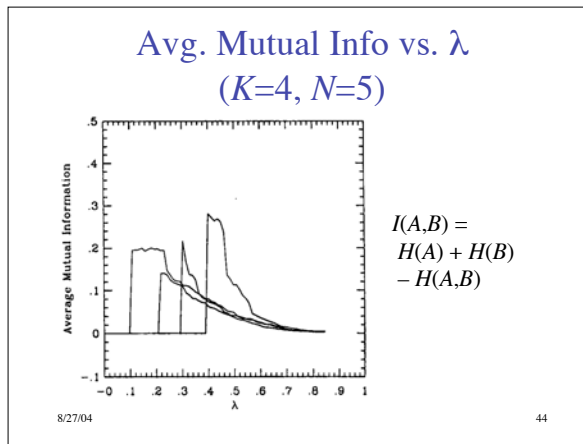
Mutual Information

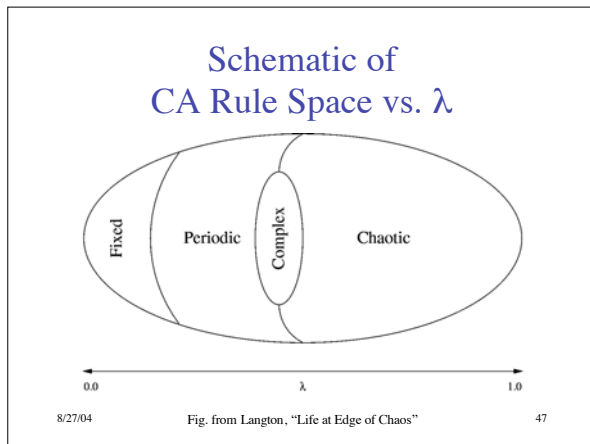
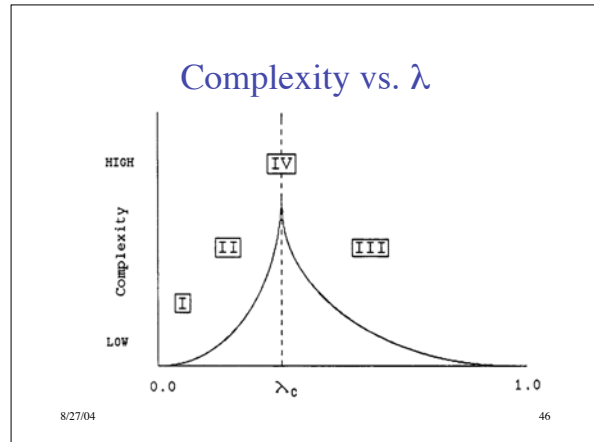
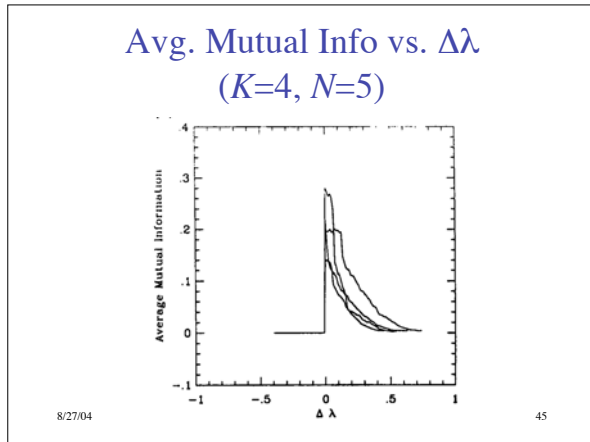
- *Mutual information* measures the degree to which two sources are not independent
- A measure of their correlation

$$I(A,B) = H(A) + H(B) - H(A,B)$$

- $I(A,B) = 0$ for completely independent sources
- $I(A,B) = H(A) = H(B)$ for completely correlated sources

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Additional Bibliography

1. Langton, Christopher G. "Computation at the Edge of Chaos: Phase Transitions and Emergent Computation," in *Emergent Computation*, ed. Stephanie Forrest. North-Holland, 1990.
2. Langton, Christopher G. "Life at the Edge of Chaos," in *Artificial Life II*, ed. Langton et al. Addison-Wesley, 1992.
3. Emmeche, Claus. *The Garden in the Machine: The Emerging Science of Artificial Life*. Princeton, 1994.

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