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Memory in Processes ... Entropy Hierarchy:

Take derivatives:

- (I) Block entropy: H(L)
- (2) Entropy rate: $h_{\mu}(L) = \Delta H(L)$
- (3) Predictability gain: $\Delta h_{\mu}(L) = \Delta^2 H(L)$

Now take integrals!

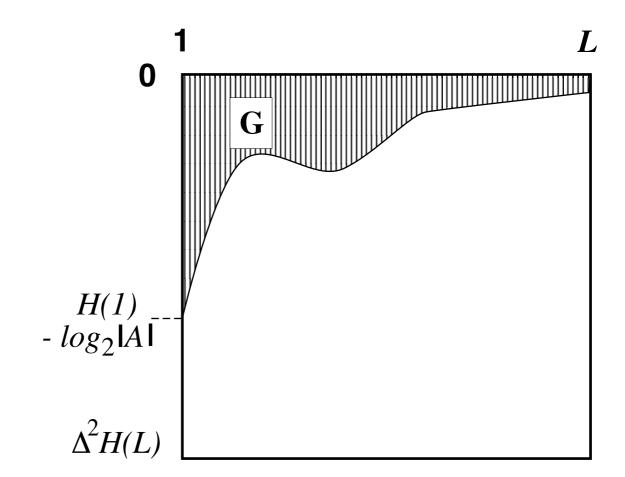
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Total Predictability:

$$\mathbf{G} = \sum_{L=1}^{\infty} \Delta^2 H(L)$$

Redundancy:

$$-\mathbf{G} = \mathcal{R} = \log_2 |\mathcal{A}| - h_{\mu}$$



Interpretation:

- (I) Account for all correlations to see intrinsic randomness
- (2) Until that point, correlations appear as excess randomness

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Excess Entropy:

As entropy convergence:

$$\mathbf{E} = \sum_{L=1}^{\infty} [h_{\mu}(L) - h_{\mu}]$$

$$(\Delta L = 1 \text{ symbol})$$

As intrinsic redundancy:

$$\mathbf{E} = \sum_{L=1}^{\infty} r(L)$$

Properties:

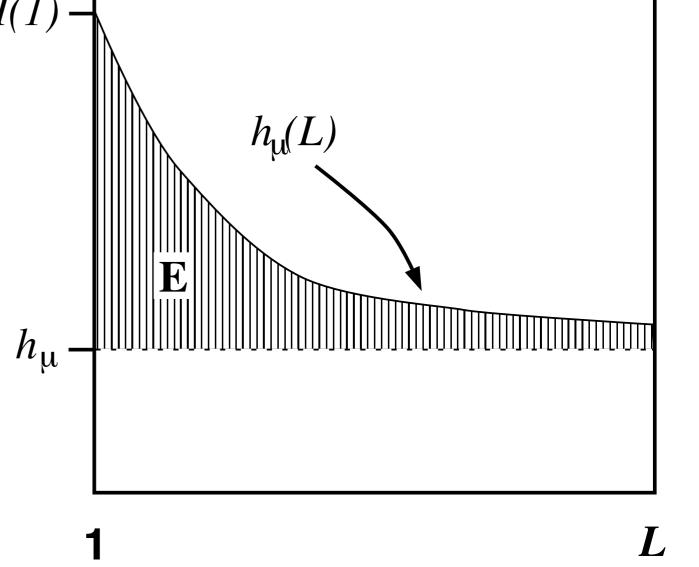
(2) Positive: E > 0

(I) Units: $\mathbf{E} = [\mathrm{bits}]$

(3) Controls convergence to actual randomness.

(4) Slow convergence ⇔ Correlations at longer words.

(5) Complementary to entropy rate.



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Excess Entropy ...

Asymptote of entropy growth:

$$\mathbf{E} = \lim_{L \to \infty} [H(L) - h_{\mu}L]$$

That is,

$$H(L) \propto \mathbf{E} + h_{\mu}L$$
 $H(L)$ Y-Intercept of entropy growth \mathbf{E}

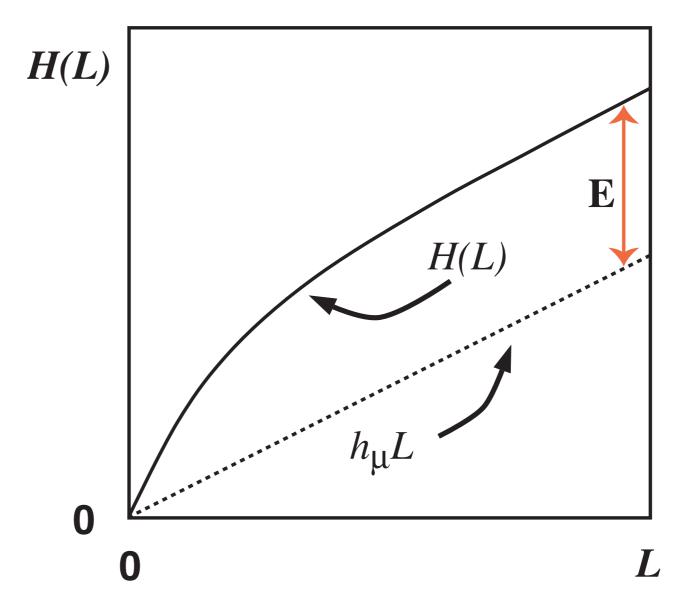
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Excess Entropy ...

Cost of Amnesia:

Forget what you know:

Information needed to recover predicting with error $hicksim h_{\mu}$



Cf. Memoryless Source: IID at same entropy rate

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Excess Entropy ...

Mutual information between past and future:

View process as a communication channel: Past to Future

$$\mathbf{E} = I(\overset{\leftarrow}{S}; \vec{S})$$

Property:

Symmetric in time

Interpretation:

Information that process communicates from past to future. Reduction in uncertainty about the future, given the past. Reduction in uncertainty about the past, given the future.

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Memory in Processes ... Examples of Excess Entropy:

Fair Coin:

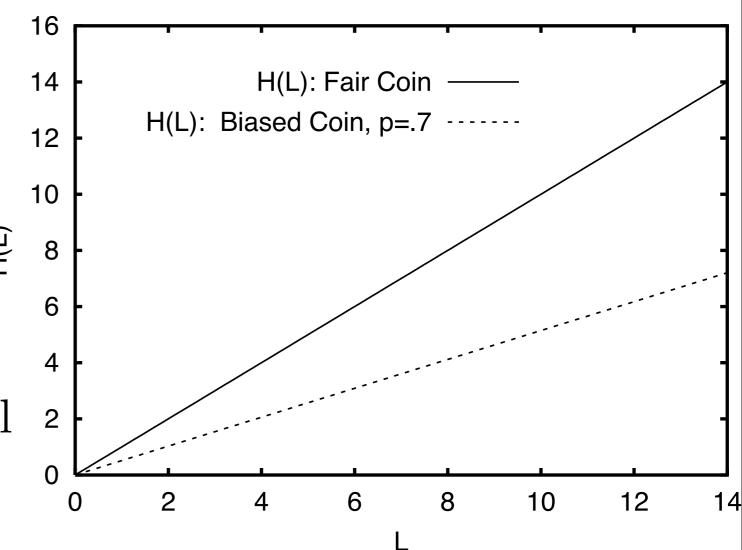
$$h_{\mu} = 1$$
 bit per symbol

$$\mathbf{E} = 0$$
 bits

Biased Coin:

$$h_{\mu} = H(p)$$
 bits per symbol

$$\mathbf{E} = 0$$
 bits



Any IID Process:

$$h_{\mu} = H(X)$$
 bits per symbol

$$\mathbf{E} = 0$$
 bits

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Examples of Excess Entropy ...

Period-2 Process: 010101010101

$$H(1) = 1$$
 $H(2) = 1$
 $H(3) = 1$

$$h_{\mu}(1) = 1$$
 $h_{\mu}(2) = 0$
 $h_{\mu}(3) = 0$

$$h_{\mu} = 0$$
 bits per symbol

$$\mathbf{E} = 1$$
 bit

Meaning:

I bit of phase information 0-phase or I-phase?

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Examples of Excess Entropy ...

Period-16 Process:

 $(1010111011101110)^{\infty}$

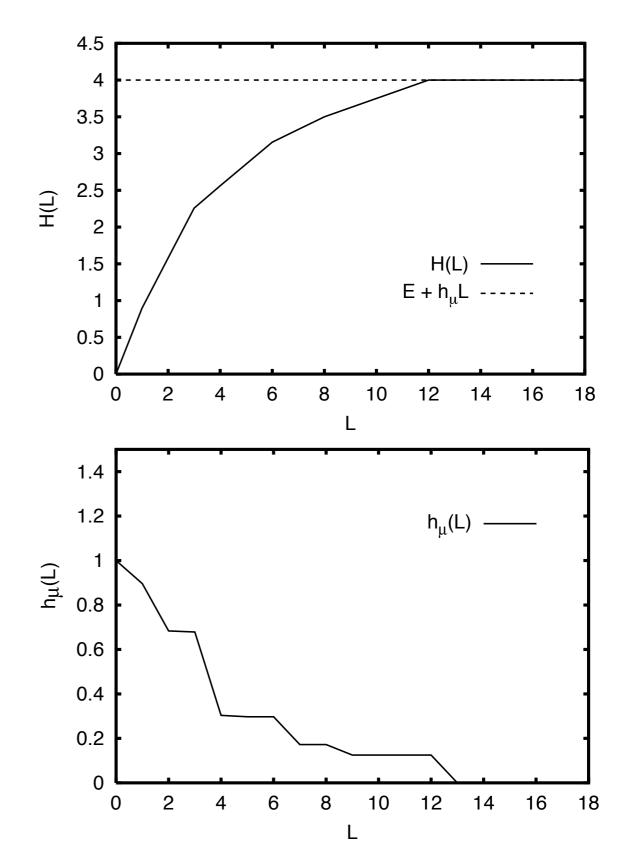
 $h_{\mu} = 0$ bits per symbol

$$\mathbf{E} = 4 \text{ bits}$$

Period-P Processes:

 $h_{\mu} = 0$ bits per symbol

$$\mathbf{E} = \log_2 P$$
 bits



Cf., entropy rate does not distinguish periodic processes!

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Memory in Processes ... Examples of Excess Entropy ...

Golden Mean Process:

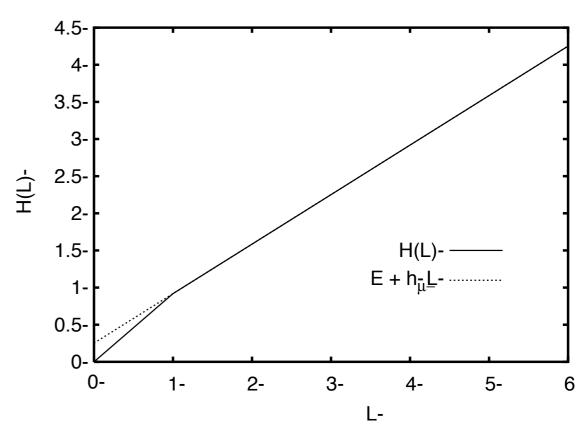
$$h_{\mu} = \frac{2}{3}$$
 bits per symbol

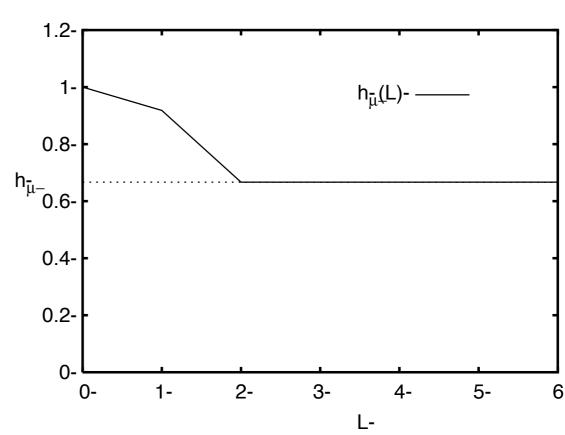
$$\mathbf{E} \approx 0.2516 \text{ bits}$$

R-Block Markov Chain:

$$\mathbf{E} = H(R) - R \cdot h_{\mu}$$

(E.g., ID Ising Spin System)





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Examples of Excess Entropy:

Finitary Processes: Exponential entropy convergence

Random-Random XOR (RRXOR) Process:

$$S_t = S_{t-1} \text{ XOR } S_{t-2}$$

$$h_{\mu} = \frac{2}{3}$$
 bits per symbol

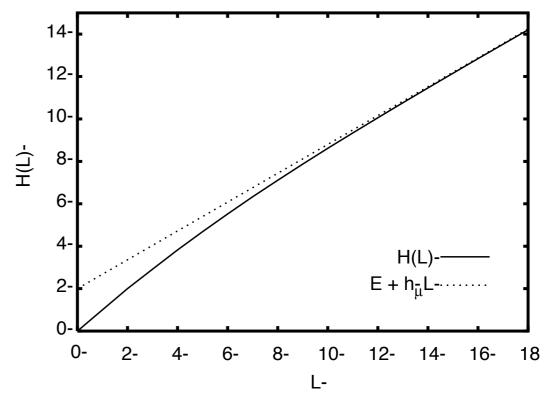
$$\mathbf{E} \approx 2.252 \text{ bits}$$

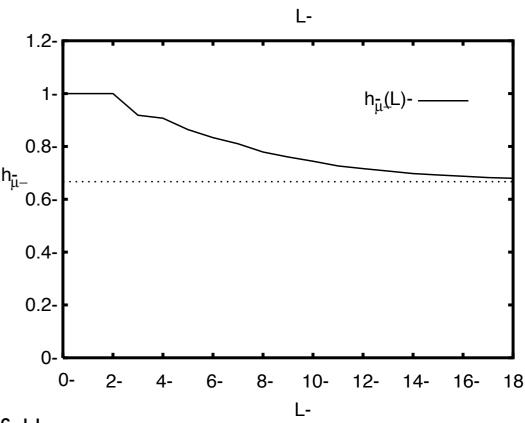
Finitary processes: Exponential convergence:

$$h_{\mu}(L) - h_{\mu} \approx 2^{-\gamma L}$$

$$\mathbf{E} = \frac{H(1) - h_{\mu}}{1 - 2^{-\gamma}}$$

$$\gamma \approx 0.30$$





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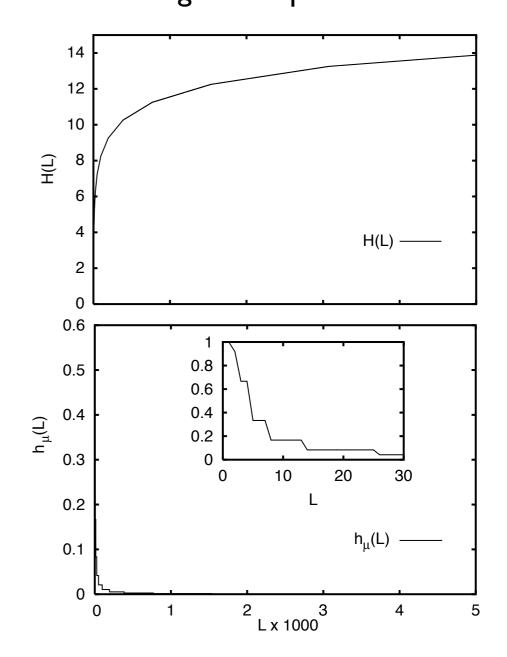
Memory in Processes ...
Examples of Excess Entropy:
Infinitary Processes:

$$\mathbf{E} o \infty$$

Excess entropy can diverge:
Slow entropy convergence
Long-range correlations
(e.g., at phase transitions)

Morse-Thue Process:

A context-free language
From Logistic map at onset of chaos



 $h_{\mu} = 0$ bits per symbol

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Memory in Processes ... Synchronization:

Problem Statement:

You have a correct model of a process, but you don't know it's current state.

Question: How much information must you extract from measurements to know which hidden state the process is in?

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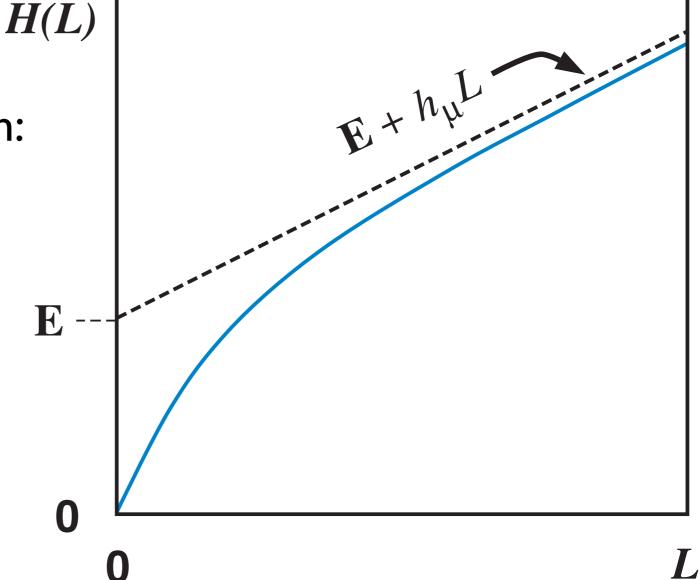
Transient Information:

Synchronized to source when:

$$L \ge L'$$

you have

$$H(L) \approx \mathbf{E} + h_{\mu}L$$



Synchronized:

At length L^\prime at which you see true entropy rate.

Extracted sufficient information to do optimal prediction.

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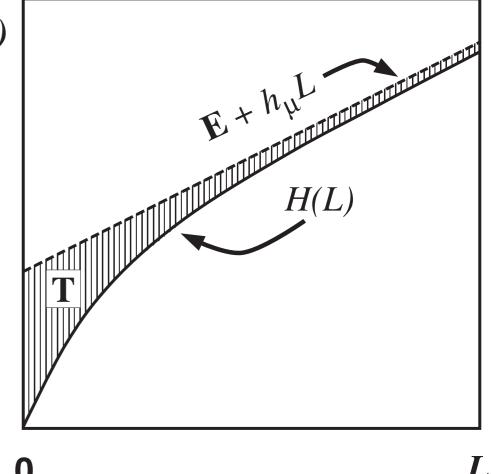
Transient Information ...

H(L)

How much information to extract?

Transient Information:

$$\mathbf{T} = \sum_{L=0}^{\infty} \left[\mathbf{E} + h_{\mu} L - H(L) \right]$$



Controls convergence to synchronization. Units: bits x symbols

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Memory in Processes ... Example of Transient Information: Tahitian Vacation (3 days)! Weather has a 5 day cycle: Two days of rain, followed by three of sun Weather is exactly predictable: $h_{\mu} = 0$ bits per day Weather has memory: $\mathbf{E} = \log_2 5 \text{ bits}$ But, How to pack?

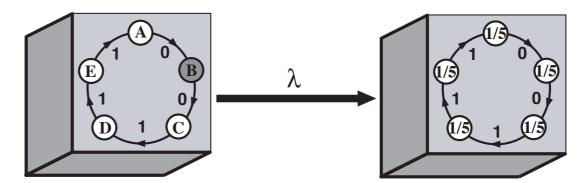
```
How to pack?
What to pack?
What to wear on trip?
Dressed appropriately for arrival?
```

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0 = Rain

I = Sun

No weather reports yet.



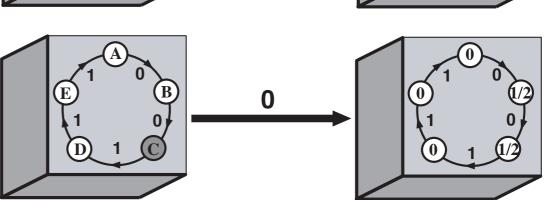
Weather **Tahiti Update Reports** Traveler's Model

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0 = Rain I = Sun

No weather reports yet.

Rain!



Tahiti

Weather Reports

Update Traveler's Model

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0 = Rain I = Sun

No weather reports yet. Rain! 01 Sun! Weather **Update Tahiti Reports** Traveler's Model

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0 = Rain I = Sun

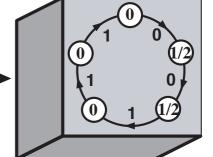
No weather reports yet.

^

E 1 0 B λ
D 1 C

Rain!

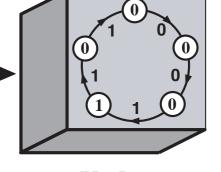
E 1 0 B 1 C



Sun!

(E) 1 (B) 1 (C)

Tahiti



Pack umbrella, wear shorts on plane

Weather Reports

01

Update Traveler's Model

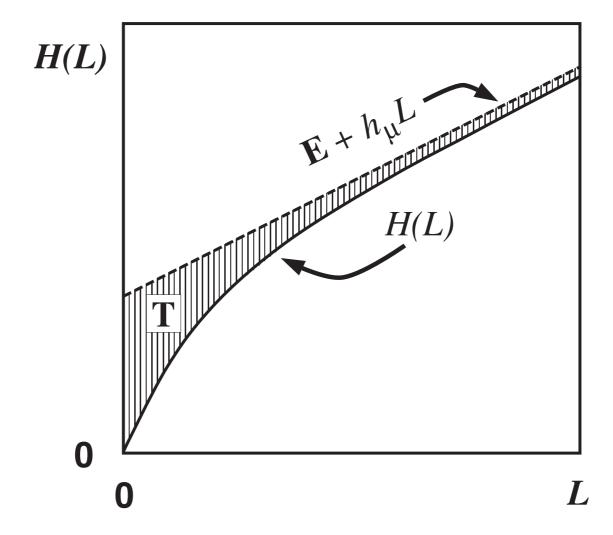
 $T \approx 4.073 \text{ bit} \times \text{symbols}$

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Memory in Processes ...

Transient Information ...

How to interpret?



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Transient Information ...

Synchronization information:

Observer has correct model of a Markov chain: $\mathcal{M} = \{V, T\}$

Observer Synchronized to Process:

$$\mathbf{T}(L) \equiv \mathbf{E} + h_{\mu}L - H(L) = 0$$

Observer knows with certainty in which state the process is:

$$Pr(v_0, v_1, \dots, v_k) = (0, \dots, 1, \dots, 0)$$

Average per-symbol uncertainty is exactly h_{μ} .

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Transient Information ...

Synchronization information ...

Average state-uncertainty:

$$\mathcal{H}(L) \equiv -\sum_{s^L \in \mathcal{A}^L} \Pr(s^L) \sum_{v \in \mathcal{V}} \Pr(v|s^L) \log_2 \Pr(v|s^L)$$

Synchronization information:

$$\mathbf{S} \equiv \sum_{L=0}^{\infty} \mathcal{H}(L)$$

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Memory in Processes ...
Transient Information ...
Synchronization information ...

Theorem: For a R-block Markov process, the synchronization information is given by:

$$\mathbf{S} = \mathbf{T} + \frac{1}{2}R(R+1)h_{\mu}$$

Corollary: For periodic process:

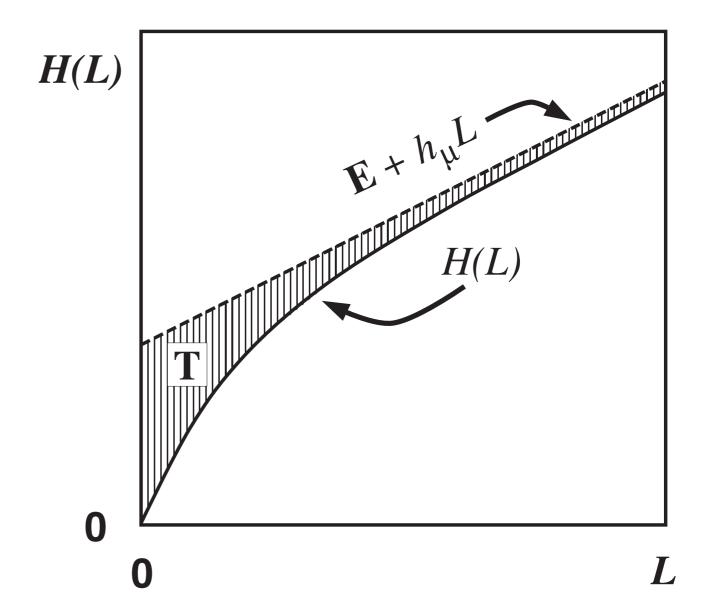
$$S = T$$

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Memory in Processes ...

Transient Information ...

How to interpret?



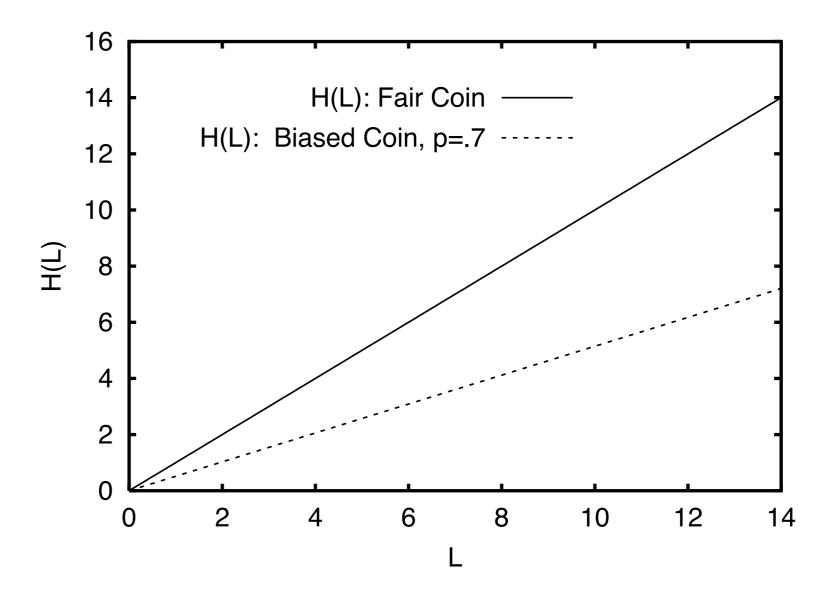
- I. Total uncertainty observed while synchronizing.
- 2. Information to extract to be synchronized.

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Memory in Processes ...

Examples of Transient Information:

Fair & Biased Coins & IID Processes: $\mathbf{T} = 0$



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Memory in Processes ... Examples of Transient Information ...

Period-5 Processes:

There are three distinct:

$$(11000)^{\infty}$$

 $(10101)^{\infty}$
 $(10000)^{\infty}$

All:

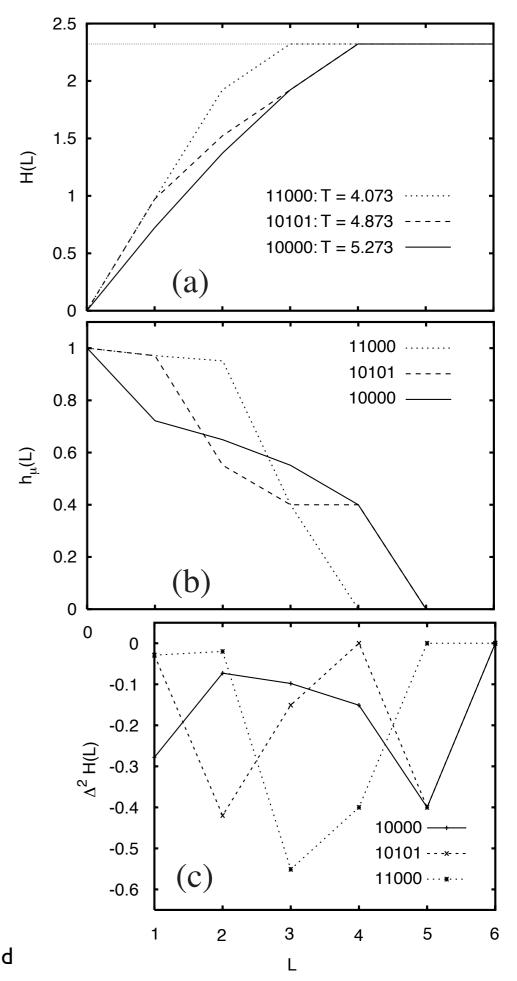
Predictable: $h_{\mu} = 0 \, \, {\rm bits}$ Memory: ${\bf E} = \log_2 5 \, \, {\rm bits}$

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Memory in Processes ... Examples of Transient Information ...

Period-5 Processes ...

But different ways to sync:



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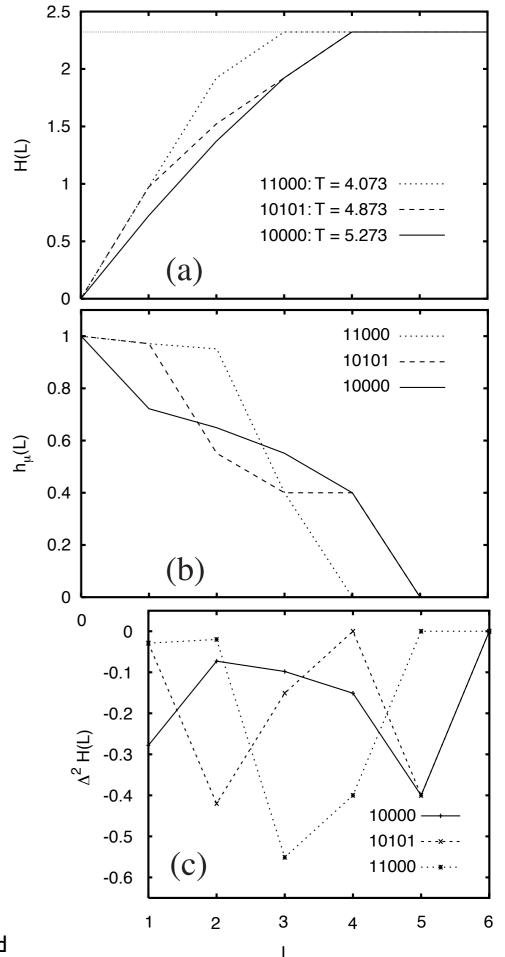
Wednesday, June 22, 2011 26

Memory in Processes ... Examples of Transient Information ...

Period-5 Processes ...

But different ways to sync:

$$(11000)^{\infty}$$
 $\mathbf{T} \approx 4.073$ bit × symbols $(10101)^{\infty}$ $\mathbf{T} \approx 4.873$ bit × symbols $(10000)^{\infty}$ $\mathbf{T} \approx 5.273$ bit × symbols



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Memory in Processes ...

Examples of Transient Information ...

Period-P Processes:

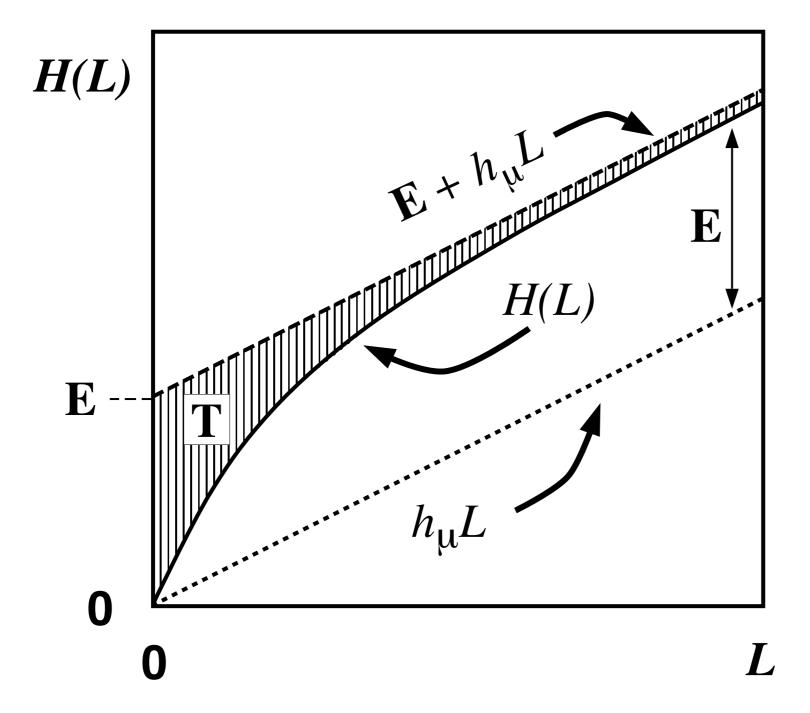
Entropy rate vanishes.

Excess entropy same for all.

But T distinguishes periodic processes.

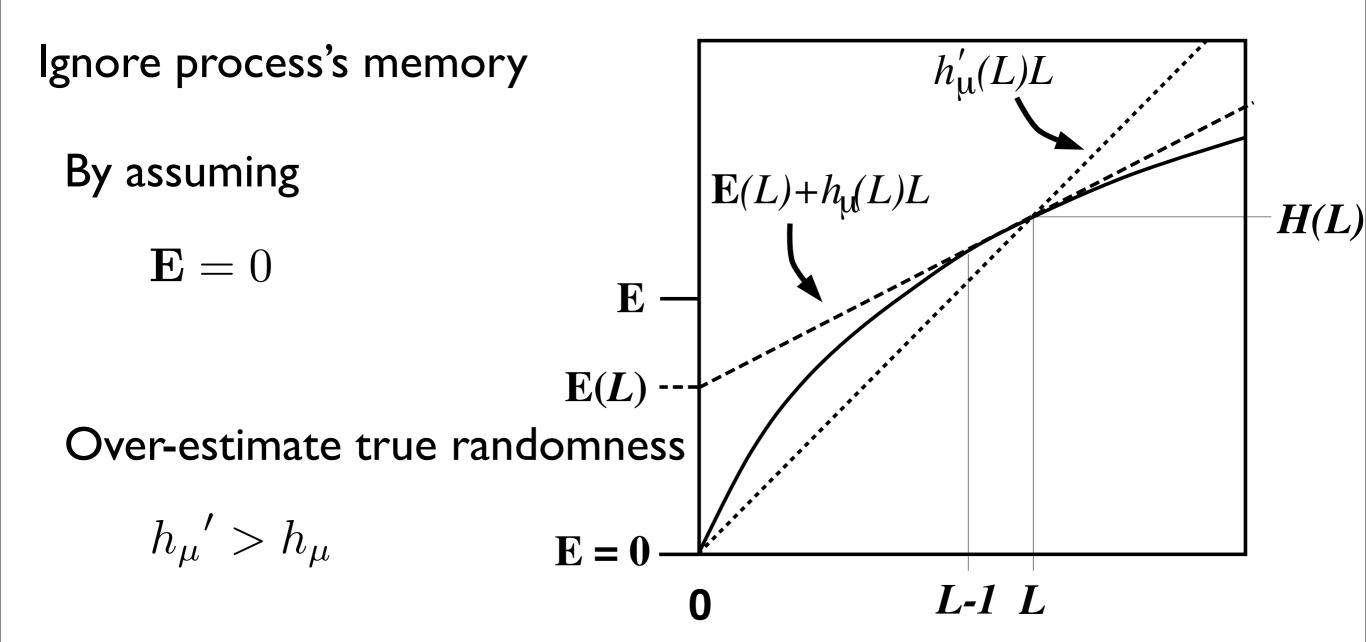
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Information-Entropy Roadmap for a Stochastic Process:



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Regularities Unseen, Randomness Observed:



Lesson:

Structure (**E** & **T**) converted to apparent randomness (h_{μ}) .

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Calculus of the Entropy Hierarchy:

Via Discrete-Time Derivatives and Integrals

Level	Gain (Derivative)	Information (Integral)
0	Block Entropy $H(L)$	Transient Information $\mathbf{T} = \sum_{L=1}^{\infty} [\mathbf{E} + h_{\mu}L - H(L)]$
	Entropy Rate Loss $h_{\mu}(L) = \Delta H(L)$	Excess Entropy $\mathbf{E} = \sum_{L=1}^{\infty} [h_{\mu}(L) - h_{\mu}]$
2	Predictability Gain $\Delta^2 H(L)$	Total Predictability (Redundancy) $G = -R$
•••	•••	•••

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What is information?

Depends on the question!

Uncertainty, surprise, randomness,

Compressibility.

Transmission rate.

Memory, apparent stored information, Synchronization.

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RGJ Lecture:

Process info diagram

Anatomy of a Bit:
Process i-Diagram
Meaning of the atoms

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