Reading for this lecture:

CMR article RURO.

Interactive CMPy lab:

Block Entropy Curves

#### Motivation:

Previous: Measures of randomness of information source Block entropy H(L) Entropy rate  $h_{\mu}$ 

End point of next lectures:

Measures of memory & information storage

# Big Picture:

Complementary properties of a source.

Need both: Measures of randomness and structure.

How random?

Block entropy growth: H(L).

If L is large enough, we see linear rate of increase of H(L):

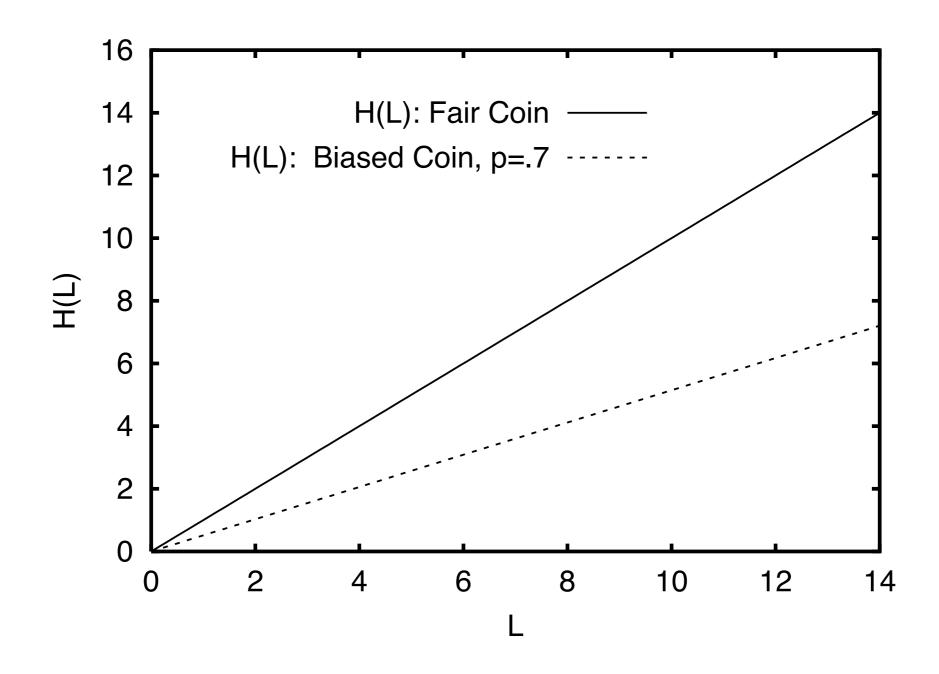
$$H(L) \propto h_{\mu}L$$

which is the entropy rate:

$$h_{\mu} = \lim_{L \to \infty} \left( H(L) - H(L-1) \right)$$

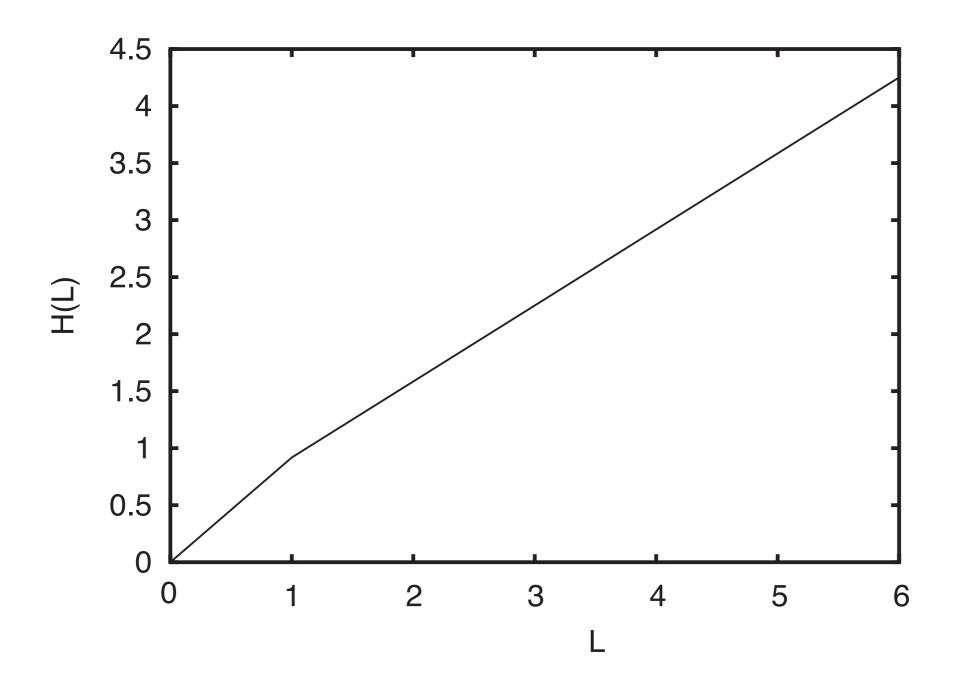
How random ...

# Fair and Biased Coins:



How random ...

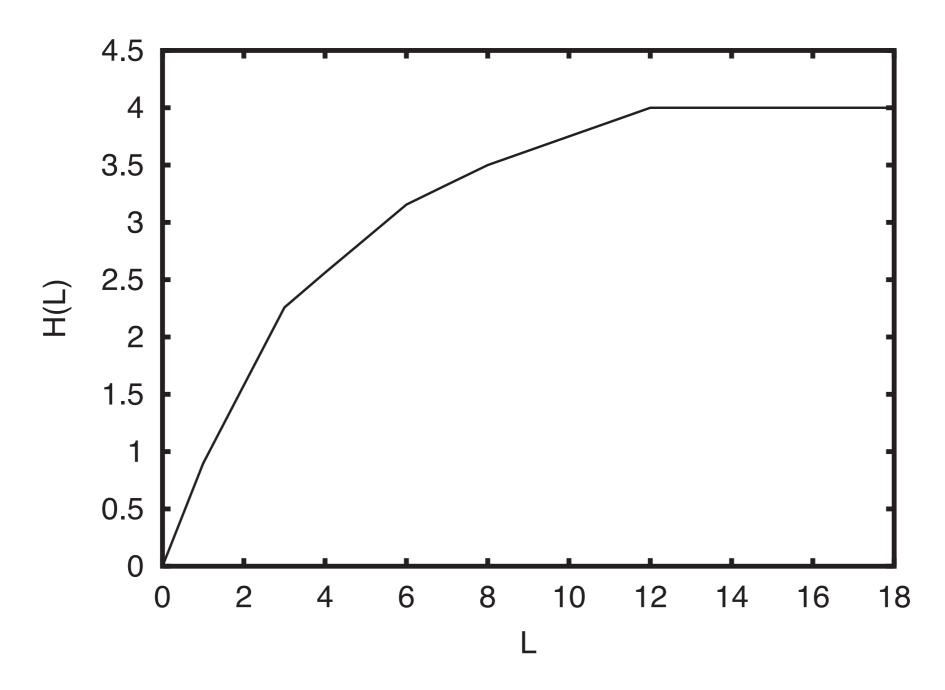
## Golden Mean Process:



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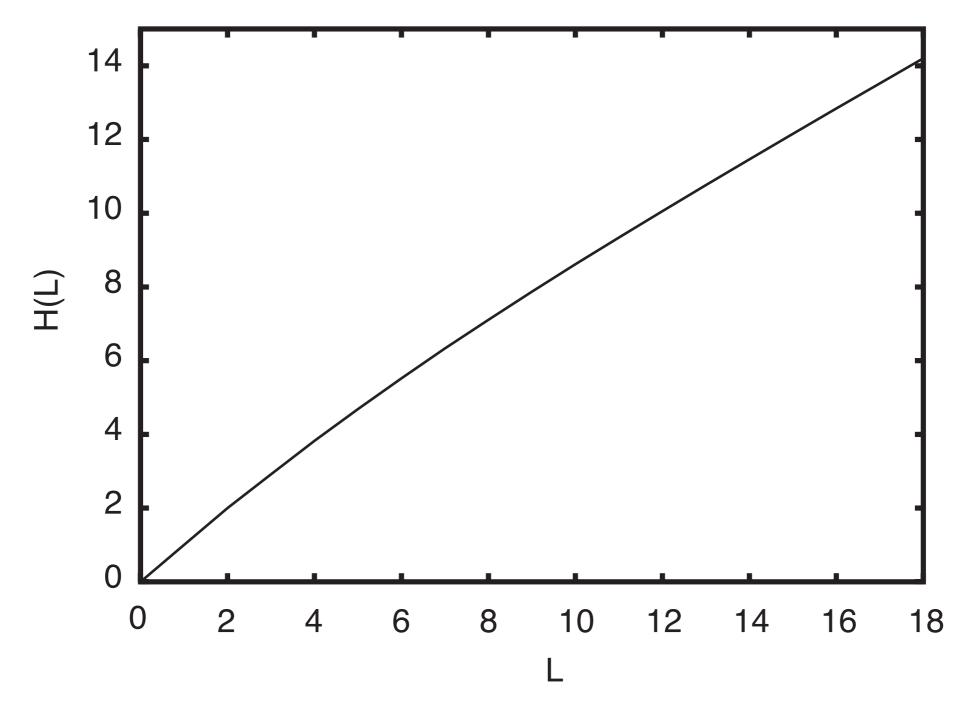
How random ...

## Period-16 Process:



How random ...

## **RRXOR Process:**



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How random ...

How large must L be to see the intrinsic randomness  $h_{\mu}$ ?

# **Entropy Convergence:**

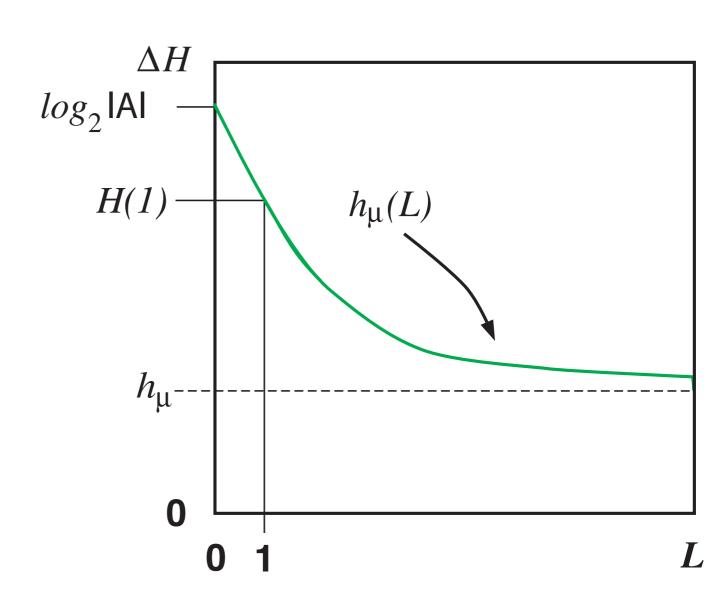
# Length-L entropy rate estimate:

$$h_{\mu}(L) = H(L) - H(L-1)$$

$$h_{\mu}(L) = \Delta H(L)$$

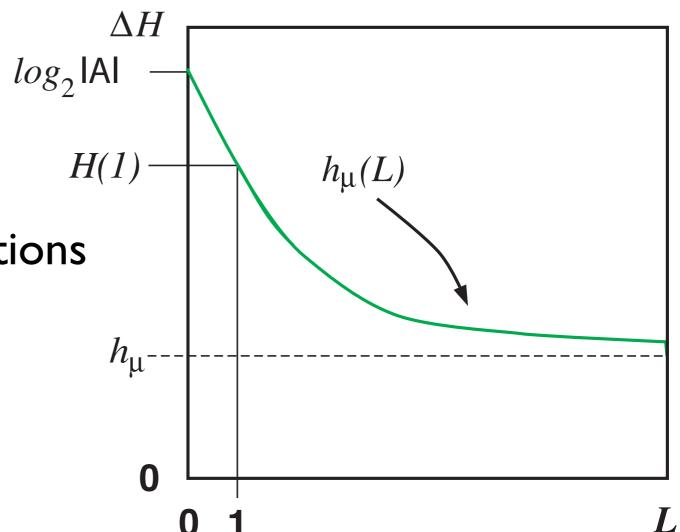
# Monotonic decreasing:

$$h_{\mu}(L) \le h_{\mu}(L-1)$$



Memory in Processes I ... Entropy Convergence ...

Process appears less random as account for longer correlations



# Entropy (rate) Loss is an Information Gain:

$$h_{\mu}(L) = \mathcal{D}(\Pr(s^L) || \Pr(s^{L-1}))$$

# Memory in Processes I ... Redundancy in Processes:

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

# Anatomy of Measurement:

$$\log_2 |\mathcal{A}|$$
 Information in single measurement  $\left\{\begin{array}{c} \\ \\ \\ \end{array}\right\} h_\mu$  Intrinsic Randomness

# Memory in Processes I ... Redundancy in Processes ...

$$\mathcal{R} = \lim_{L \to \infty} \mathcal{D}(\Pr(s^L) || U(s^L))$$

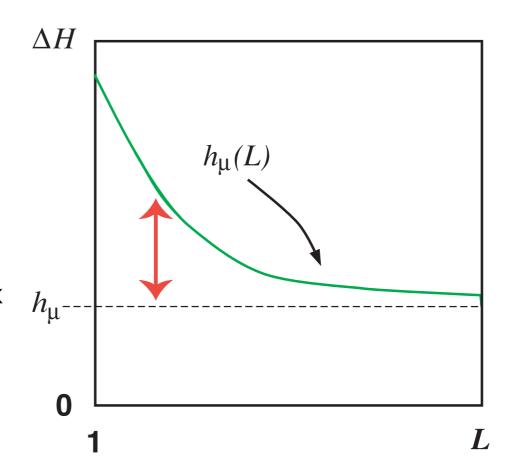
# Redundancy in words:

$$\mathcal{R}(L) = H(L) - h_{\mu}L$$

# H(L) H(L) $h_{\mu}L$ IID Approx

# Redundancy per symbol:

$$r(L) = \mathcal{R}(L) - \mathcal{R}(L-1) = h_{\mu}(L) - h_{\mu}$$



# Predictability Gain:

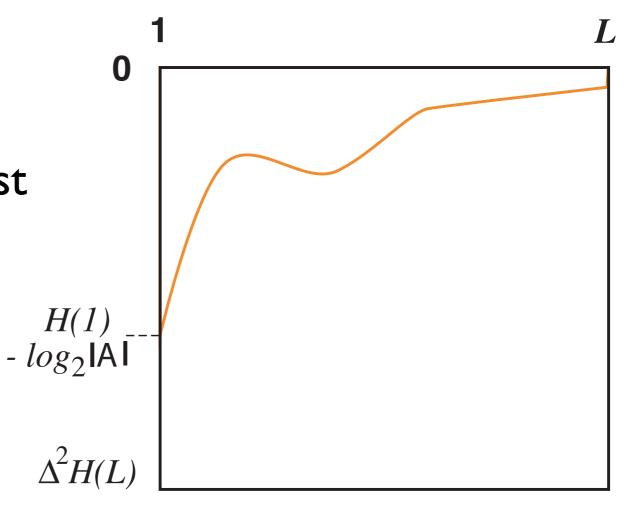
$$\Delta^2 H(L) = h_\mu(L) - h_\mu(L-1)$$

# Boundary condition:

$$\Delta^2 H(1) = H(1) - \log_2 |\mathcal{A}|$$

Predictability Gain ...

Rate at which unpredictability is lost



## Properties:

(I) H(L) Curvature:

$$\Delta^2 H(L) = H(L) - 2H(L-1) + H(L-2)$$

(2) H(L) Concavity:

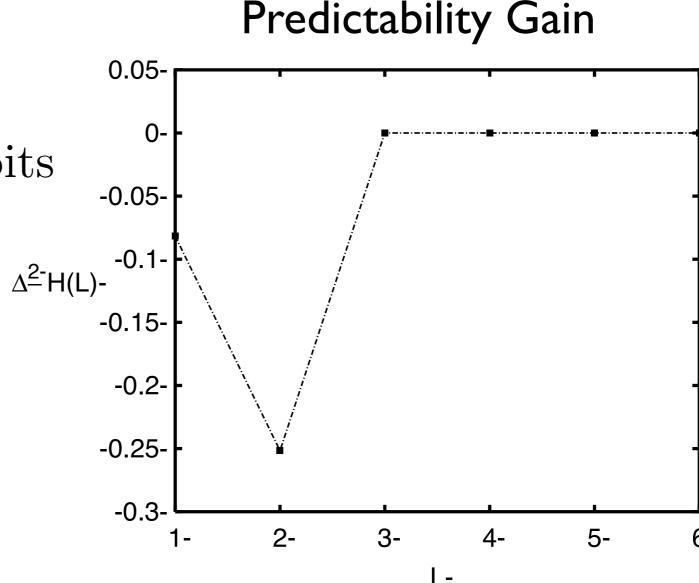
$$\Delta^2 H(L) \le 0$$

(3)  $|\Delta^2 H(L)| \gg 1 \Rightarrow$  L<sup>th</sup> measurement significant

Predictability Gain ...

Golden Mean Process:

$$\Delta^2 H(2) = -0.2516$$
 bits



Second measurement is informative: 00 restriction observed

# Memory in Processes I ... Entropy Hierarchy:

#### Take derivatives:

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

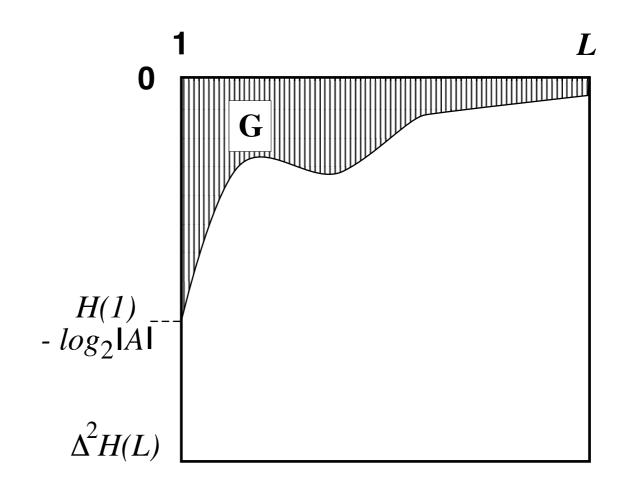
Now take integrals!

# **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

# **Excess Entropy:**

As entropy convergence:

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

# Excess Entropy ...

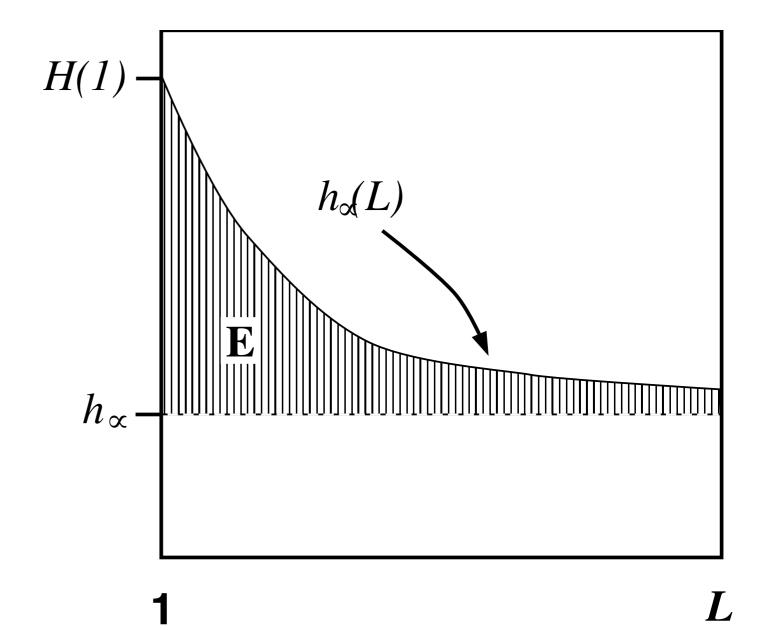
# As intrinsic redundancy:

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

# Properties:

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

(2) Positive:  $E \ge 0$ 



- (3) Controls convergence to actual randomness.
- (4) Slow convergence ⇔ Correlations at longer words.
- (5) Complementary to entropy rate.

Excess Entropy ...

Asymptote of entropy growth:

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

That is,

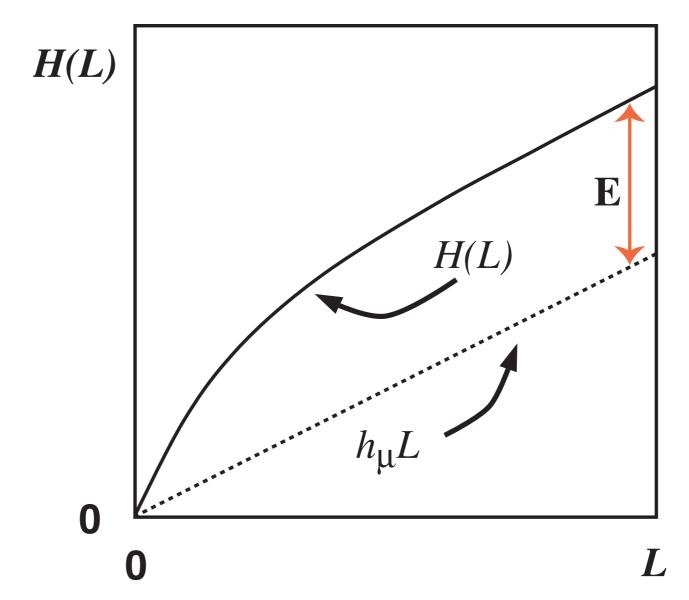
$$H(L) \propto \mathbf{E} + h_{\mu} L$$
  $H(L)$  Y-Intercept of entropy growth  $\mathbf{E}$ 

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

Excess Entropy ...

# Mutual information between past and future:

View process as a communication channel: Past to Future

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

# 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.

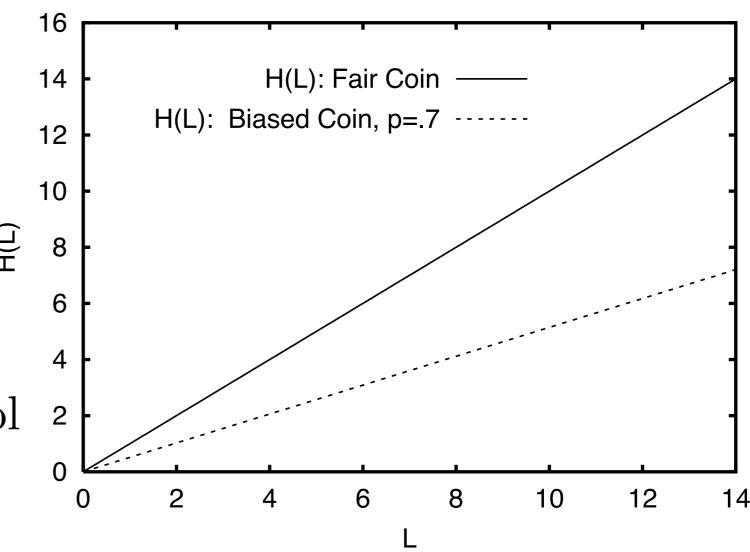
# Memory in Processes I ... Examples of Excess Entropy:

# 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

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

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

Meaning:

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

Examples of Excess Entropy ...

Period-16 Process:

 $(1010111011101110)^{\infty}$ 

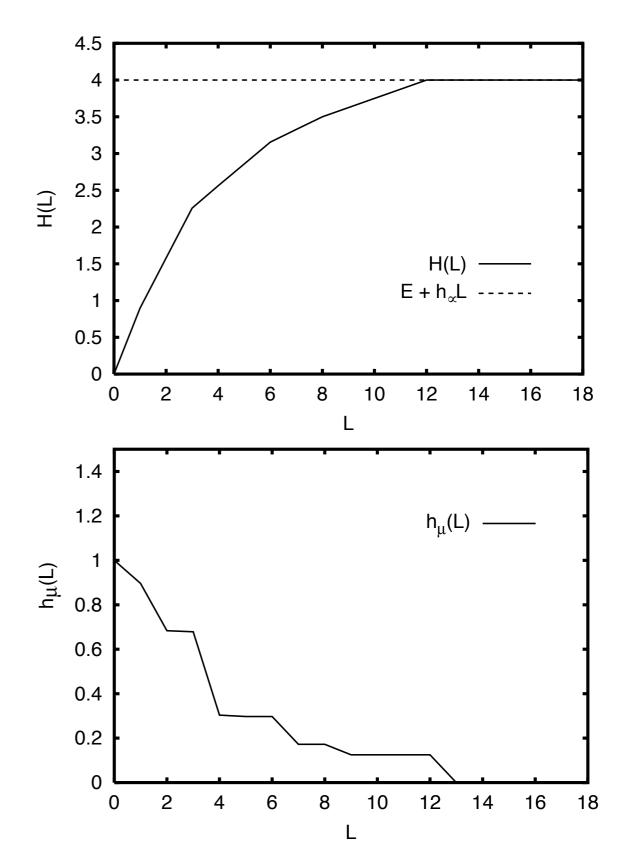
 $h_{\mu} = 0$  bits per symbol

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

#### 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 I ... Examples of Excess Entropy ...

#### Golden Mean Process:

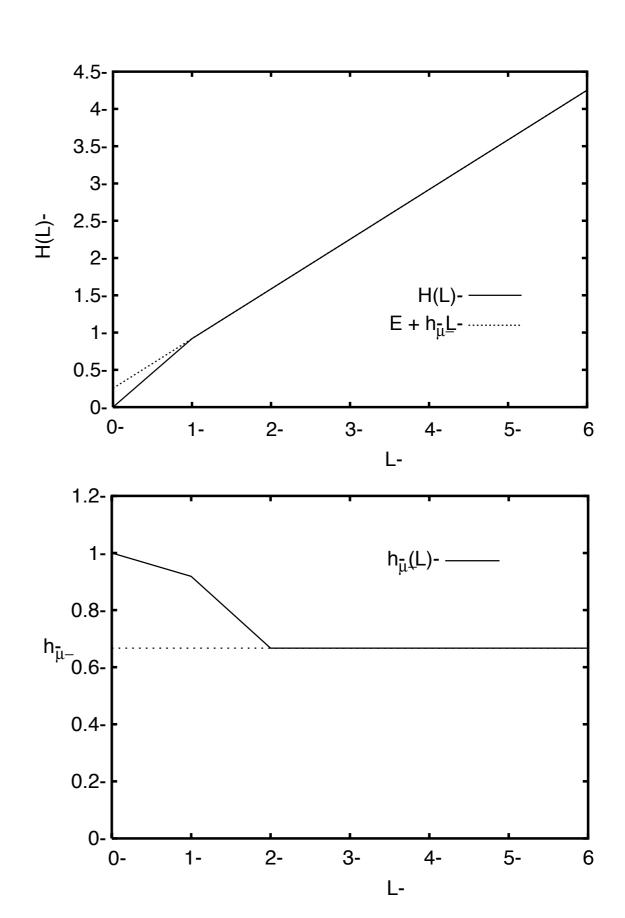
$$h_{\mu} = \frac{2}{3}$$
 bits per symbol

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

#### **R-Block Process:**

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

(Specifically, Spin-Block Process)



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

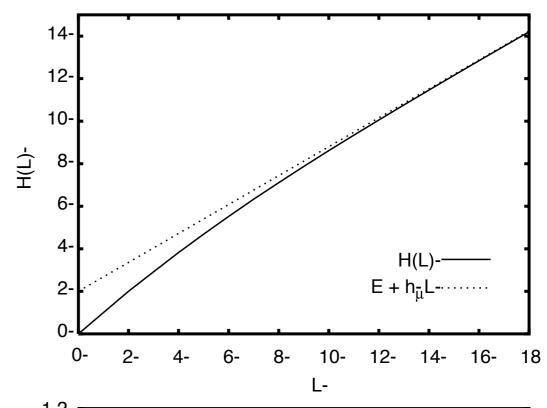
$$\mathbf{E} \approx 2.252 \text{ bits}$$

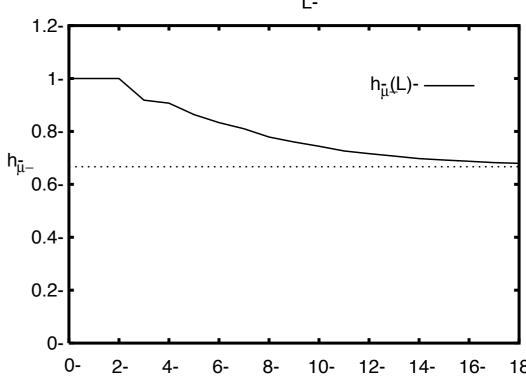
# Finitary processes: Exponential convergence:

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

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







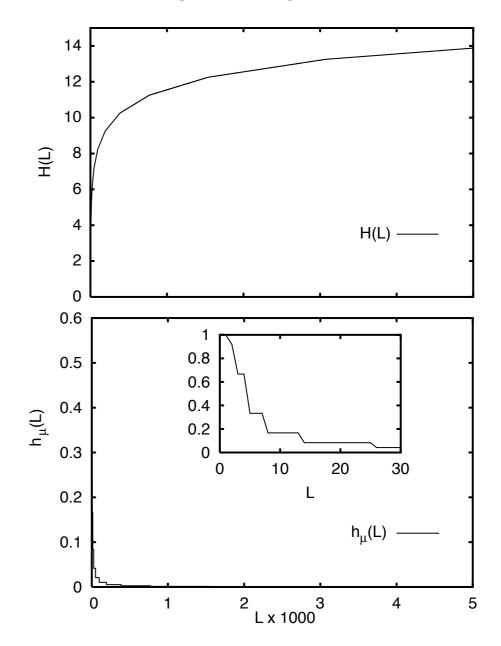
Memory in Processes I ...
Examples of Excess Entropy:
Infinitary Processes:

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

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

Reading for next lecture:

CMR article RURO.