### THE LANDAUER LIMIT AND THERMODYNAMICS OF BIOLOGICAL SYSTEMS

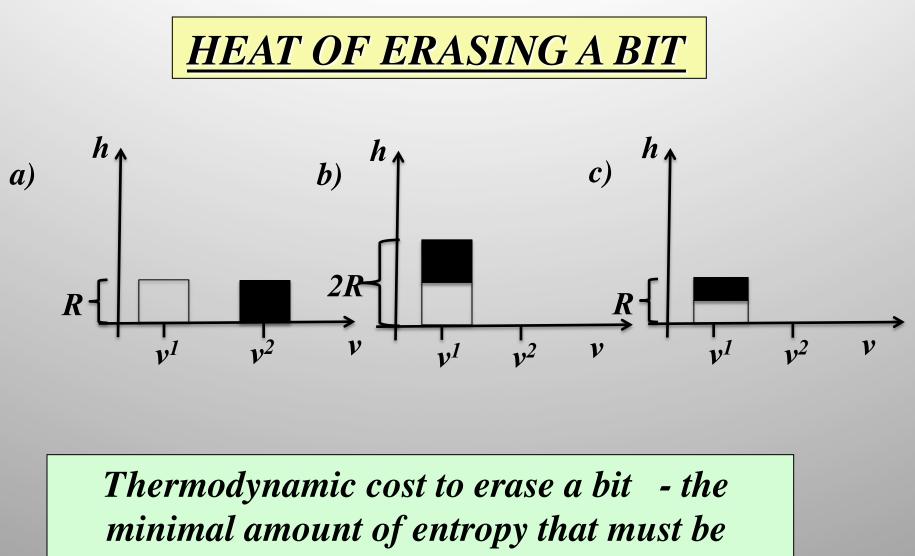
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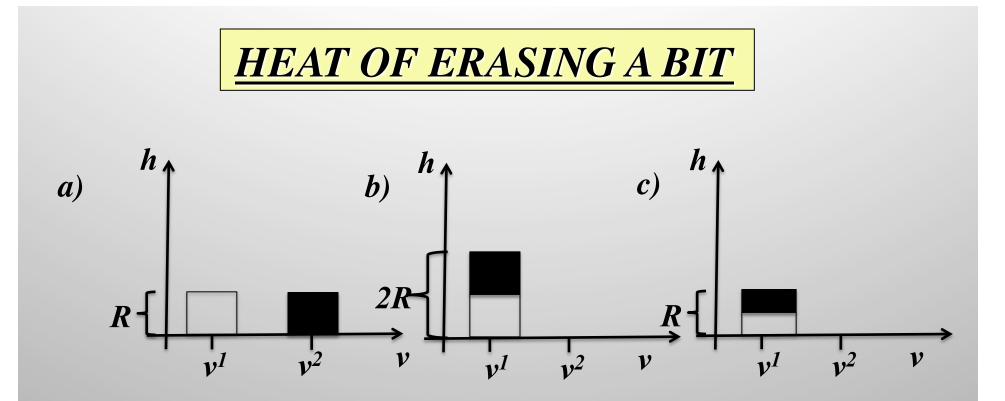


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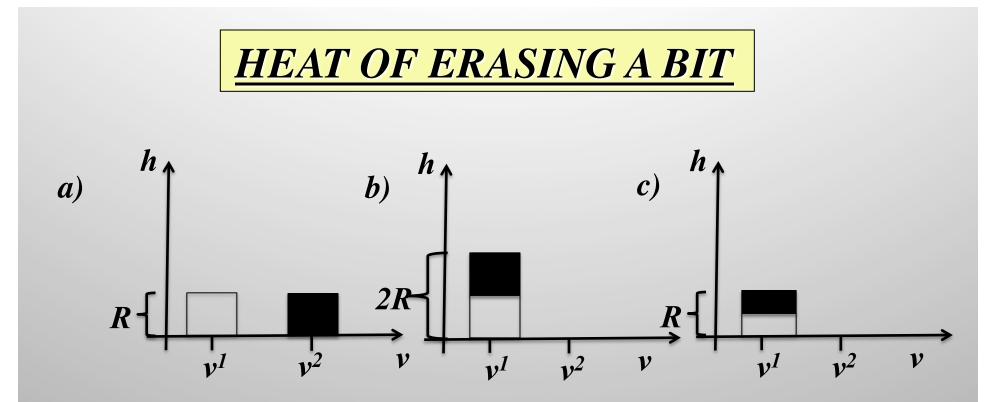


expelled to the environment - is ln[2]

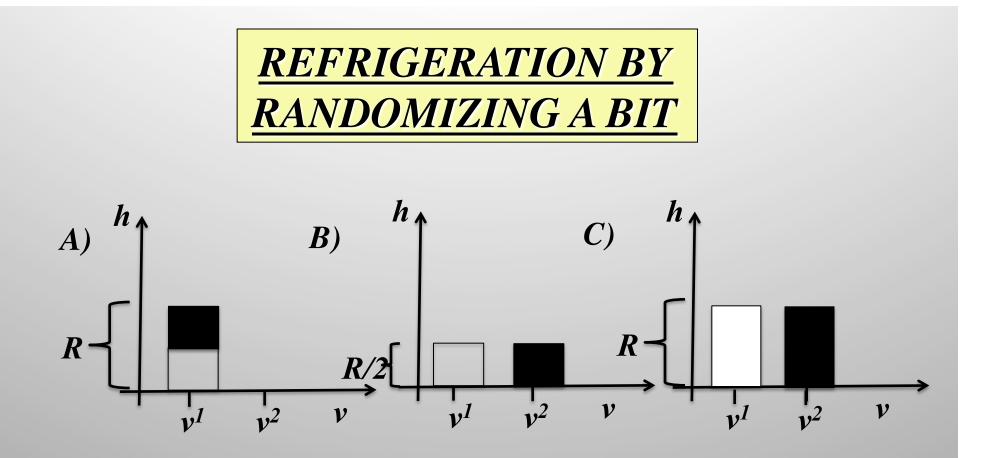


- Crucially, <u>DO</u> know precise pre-erasure value of bit
  - After all, a computer is useless if don't know its initial state

Bennett, 2003: "If erasure is applied to random data, the operation may be thermodynamically reversible ... but if it is applied to known data, it is thermodynamically irreversible."



- Crucially, <u>DO</u> know precise pre-erasure value of bit
  - After all, a computer is useless if don't know its initial state
- In fact, the prior distribution over the pre-erasure value and in particular the entropy of that prior is irrelevant
- Requires careful engineering to make this property hold
- "Local detailed balance" does not hold



- Example: Adiabatic demagnetization
- Exploited in modern engineering:
  - Noisy error correction computing
  - Real (not "pseudo") random number generators

# SOME ERASING AND SOME RANDOMIZATION

- What is the thermodynamic cost for an arbitrary conditional distribution from X = {0, 1, 2, 3} into itself?
- E.g., what if
  - 0 and 1 go to 0 (as in bit erasure);

*i.e.*,  $P(0 \mid 0) = P(0 \mid 1) = 1$ 

- 2 goes to 0 with probability .8, stays the same otherwise;
  i.e., P(0 | 2) = .8, P(2 | 2) = .2
- 3 goes to 2 with probability .4, and to 0 with probability .6;
  i.e., P(2 | 3) = .5, P(0 | 3) = .6

$$\mathbb{E}(cost) = \sum_{v_t, v_{t+1}} \pi(v_{t+1} \mid v_t) P(v_t) \ln \left[ \sum_{v'_t} \pi(v_{t+1} \mid v_t) \right]$$
  
=  $I(W_{t+1}; V_{t+1}) - I(W_t; V_t)$ 

**THERMODYNAMIC COST** 

- where v<sub>t</sub> is the observable v's value at time t;
- $\pi(. | .)$  is the conditional distribution of dynamics;
- I(.;.) is mutual information;
- W is unobserved degrees of freedom;

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- where v<sub>t</sub> is the observable v's value at time t;
- $\pi(. | .)$  is the conditional distribution of dynamics;

Example: In a 2-to-1 map,  $\pi(0 \mid 0) = \pi(0 \mid 1) = 1$ , so expected cost equals  $\ln[2]$ 

Given the evolution kernel  $\pi(. | .)$ , as one varies  $P(v_t)$ :

 $0 \leq \mathbb{E}(cost) + H(V_t) + H(V_{t+1}) \leq \log[|V|] - \max_{v_t} \left| KL(\pi(V_{t+1} \mid a) \mid| \Pi^+(V_{t+1})) \right|$ 

- where v<sub>t</sub> is the observable v's value at time t;
- **H(.)** is Shannon entropy;
- KL(. || .) is KL divergence;

• 
$$\Pi^{+}(\mathbf{v}_{t+1}) = \sum_{v_{t}} \pi(v_{t+1} | v_{t}) / |\mathbf{V}|$$

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- KL(. || .) is KL divergence;
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Example: In a 2-to-1 map both bounds are tight:

Thermodynamic cost is the drop in Shannon entropies over V

#### **K'TH ORDER MARKOV CHAINS**

For a k'th order Markov chain, thermodynamic cost during a single step is bounded below by

 $L \equiv H(V_t \mid V_{t+1}, \dots, V_{t+k-1}) - H(V_{t+k} \mid V_{t+1}, \dots, V_{t+k-1})$ 

and above by

Very messy expression

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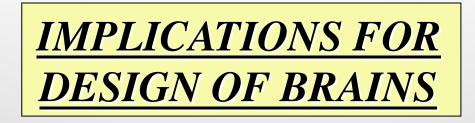
If P(v) has reached stationarity, lower bound is

 $\mathbb{E}(cost) = I(V_t; V_{t+1}, \dots, V_{t+k-1}) - I(V_{t+k}; V_{t+1}, \dots, V_{t+k-1})$ 

(Cf. Still et al. 2012)

## **THERMODYNAMIC COST**

- Tons of other fun results, including:
- 1) Second law: Thermodynamic cost is non-negative for any process that goes from distribution A to distribution B and then back to distribution B
- 2) Examples of coarse-graining (in the statistical physics sense) that increase thermodynamic cost
- 3) Examples of coarse-graining that decrease thermo. cost
- 4) Implications for optimal compiler design
- 5) Analysis of thermodynamic cost of Hidden Markov Models



- $P(x_t)$  a dynamic process outside of a brain;
- Natural selection favors brains that:
  - (generate v<sub>t</sub>'s that) predict future of x accurately;
    but ...
  - not generate heat that needs to be dissipated;
  - *not require free energy from environment (need to create all that heat)*

Natural selection favors brains that:

Accurately predict future (*quantified with a fitness function*);
 Using a prediction program with minimal thermo. cost



- Natural selection favors (phenotypes of) a prokaryote that:
  - (generate v<sub>t</sub>'s that) maximize fitness;

*but* ...

- *not* generate heat that needs to be dissipated;
- *not* require free energy from environment (need to create all that heat)

#### Natural selection favors prokaryotes that:

Behave as well as possible (*quantified with a fitness function*);
 While implementing behavior with minimal thermo. cost

$$\mathbb{E}(cost) = \sum_{v_t, v_{t+1}} \pi(v_{t+1} \mid v_t) P(v_t) \ln \left[ \sum_{v'_t} \pi(v_{t+1} \mid v_t) \right]$$

- where v<sub>t</sub> is the observable v's value at time t;
- $\pi(. | .)$  is the conditional distribution of dynamics;

N.b., thermodynamic cost varies with t:

For what kernels  $\pi(.|.)$  does thermo. cost increase with time?

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- where v<sub>t</sub> is the observable v's value at time t;
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*Plug in*  $\pi$ (. | .) *of terrestrial biosphere:* 

Does thermo. cost of biosphere behavior increase with time?

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- where v<sub>t</sub> is the observable v's value at time t;
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*Plug in*  $\pi$ (. | .) *of terrestrial biosphere:* 

How far is thermo. cost of biosphere from upper bound of solar free energy flux?