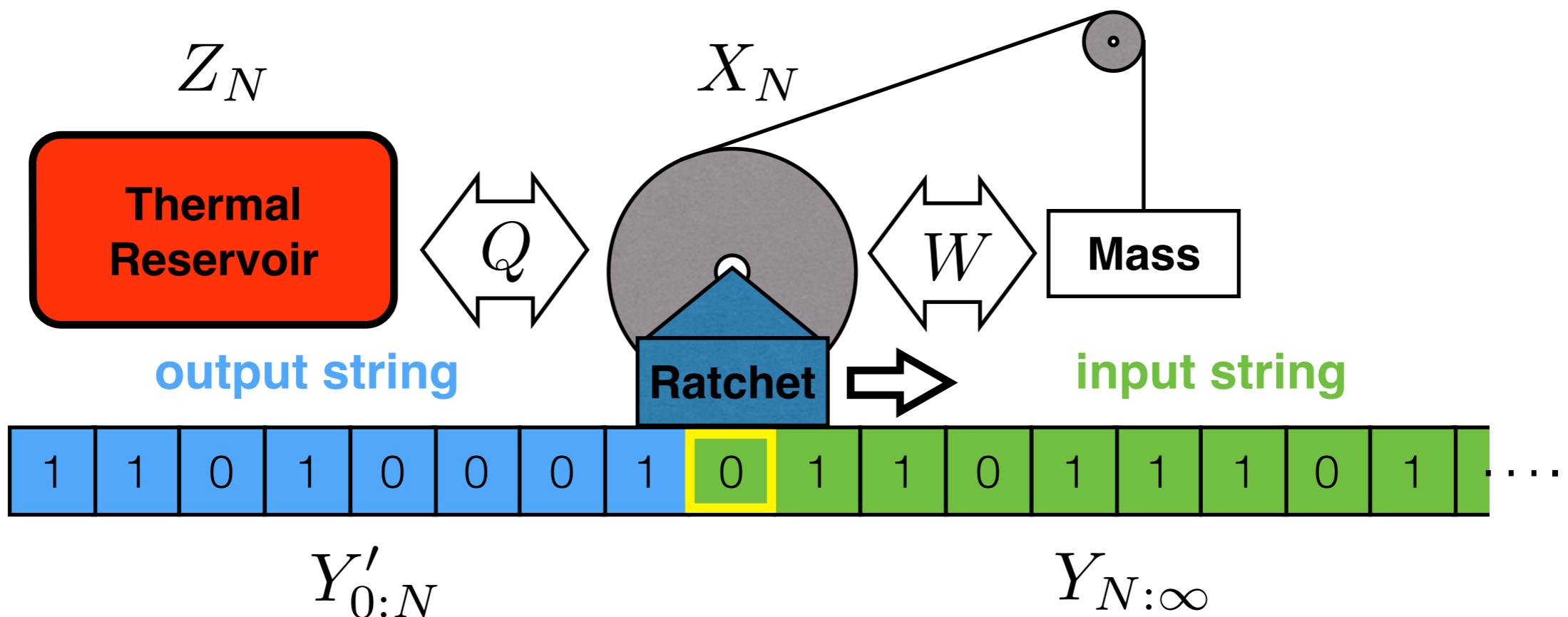
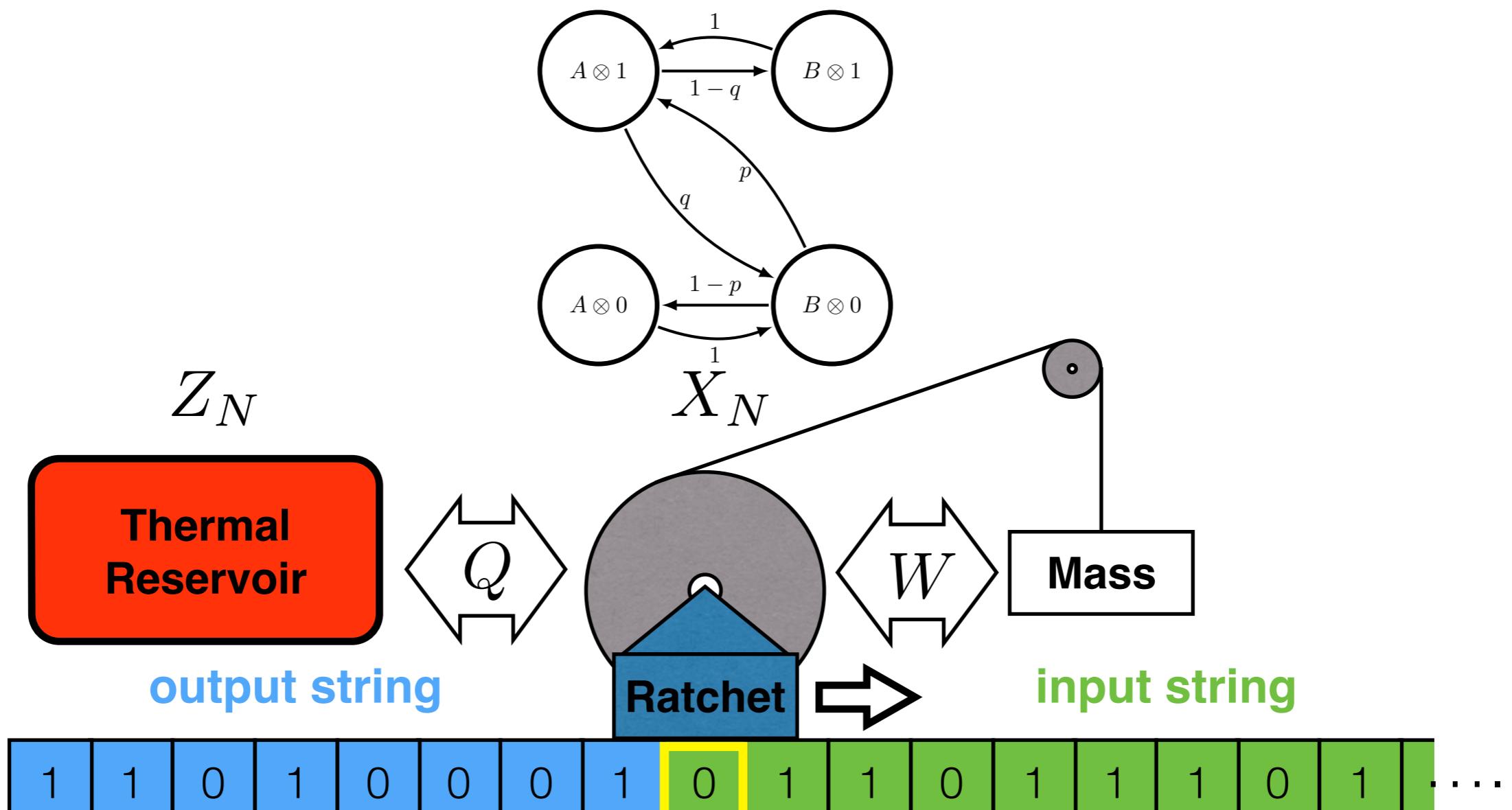


# Memory and correlation in autonomous Maxwellian Demons

# Alec Boyd, Dibyendu Mandal, James Crutchfield

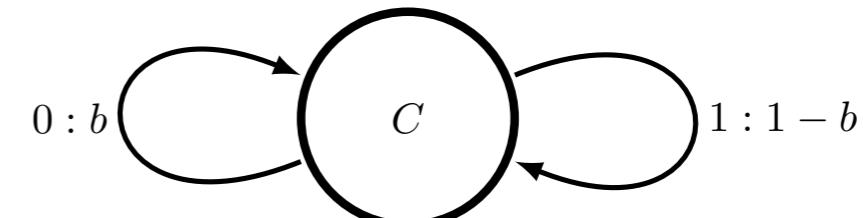
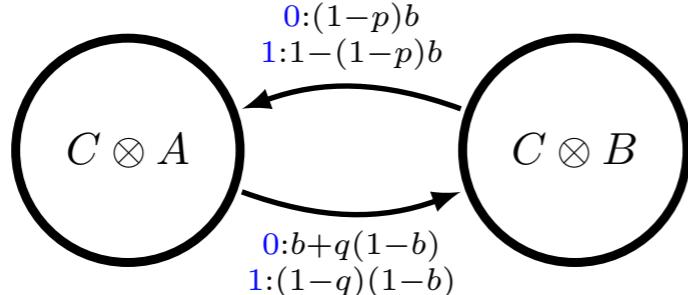


# Basic Setup



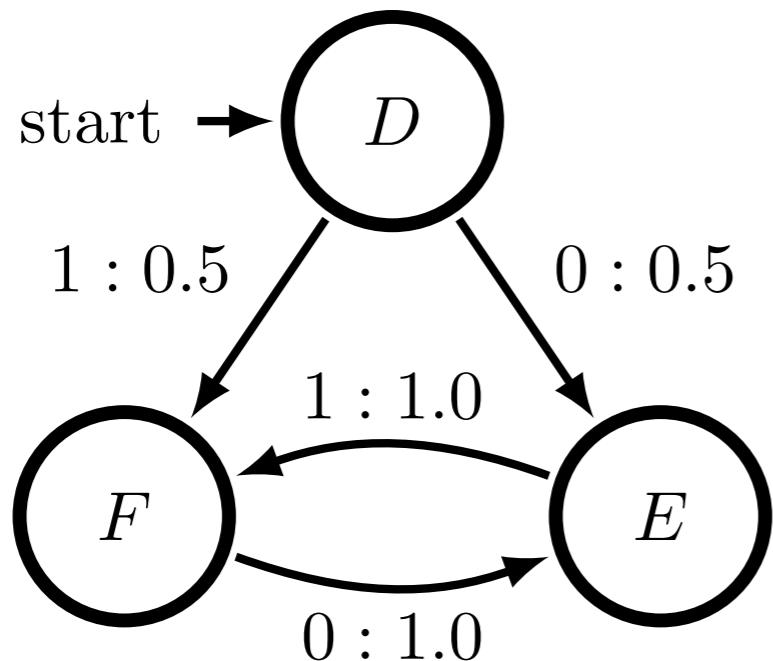
$Y'_{0:N}$

$Y_{N:\infty}$

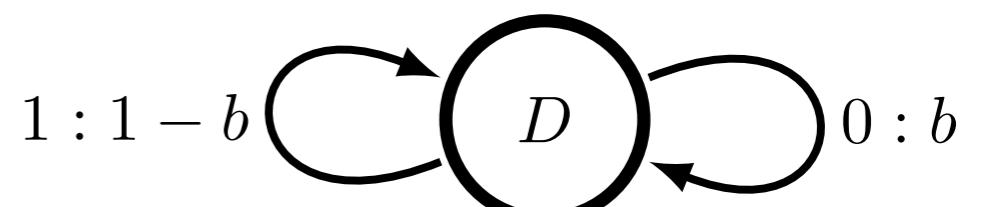


# Memory in Bit Strings

Memoryful:



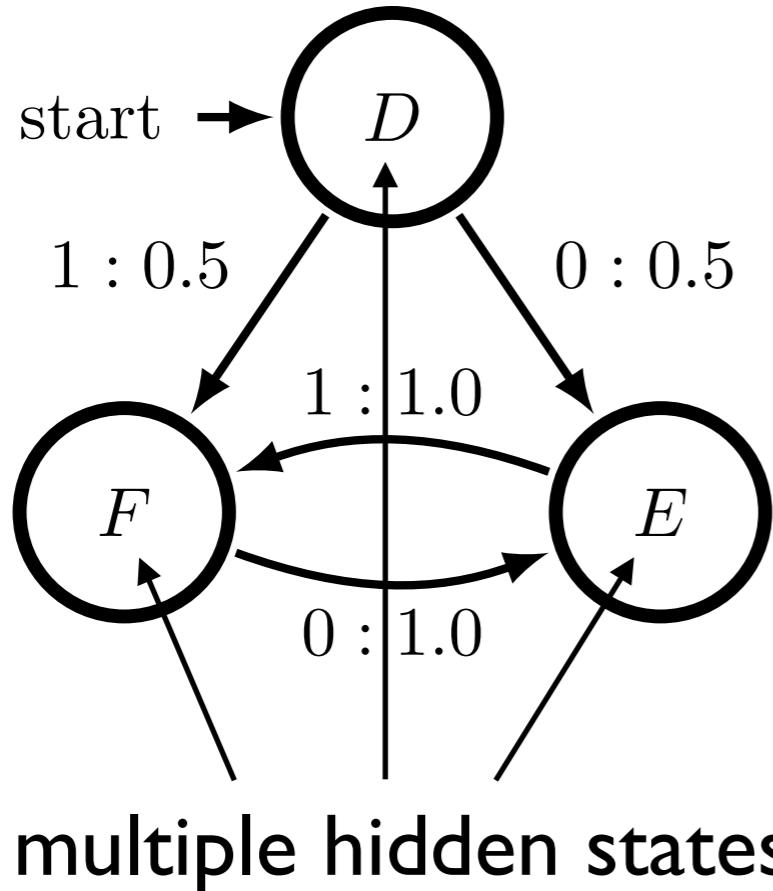
Memoryless:



Bit strings generated by HMMs

# Memory in Bit Strings

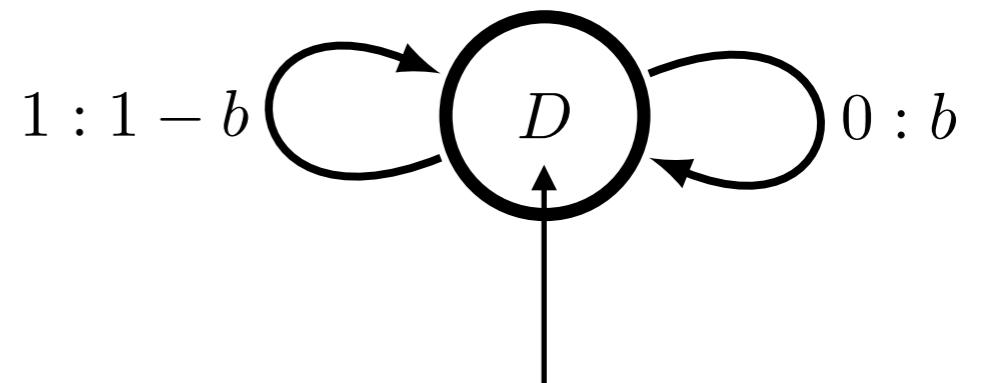
**Memoryful:**



$$T_{s_i \rightarrow s_{i+1}}^{(y_i)} = \Pr(Y_i = y_i, S_{i+1} = s_{i+1} | S_i = s_i)$$

$$\Pr(Y_{0:\infty} = y_{0:\infty}) = \sum_{s_{0:\infty} \in \mathcal{S}^\infty} \Pr(S_0 = s_0) \prod_{i=0}^{\infty} T_{s_i \rightarrow s_{i+1}}^{(y_i)}$$

**Memoryless:**



**single hidden state**

$$\Pr(Y_{0:\infty} = y_{0:\infty}) = \prod_{i=0}^{\infty} \Pr(Y_i = y_i)$$

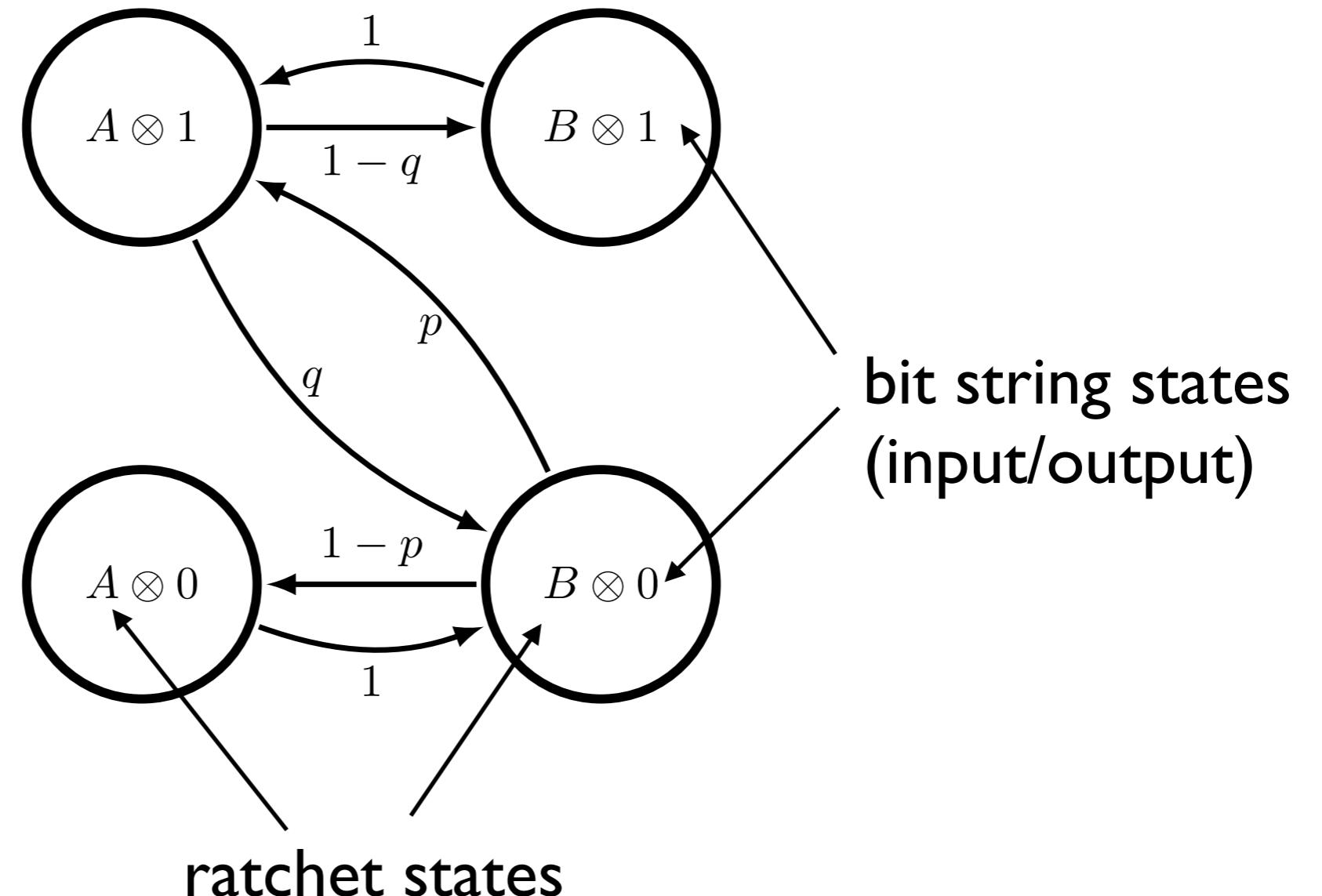
$$\Pr(Y_i = y) = \Pr(Y_j = y) \forall i, j$$

**Bit strings generated by HMMs**

# Ratchet Bit Interaction

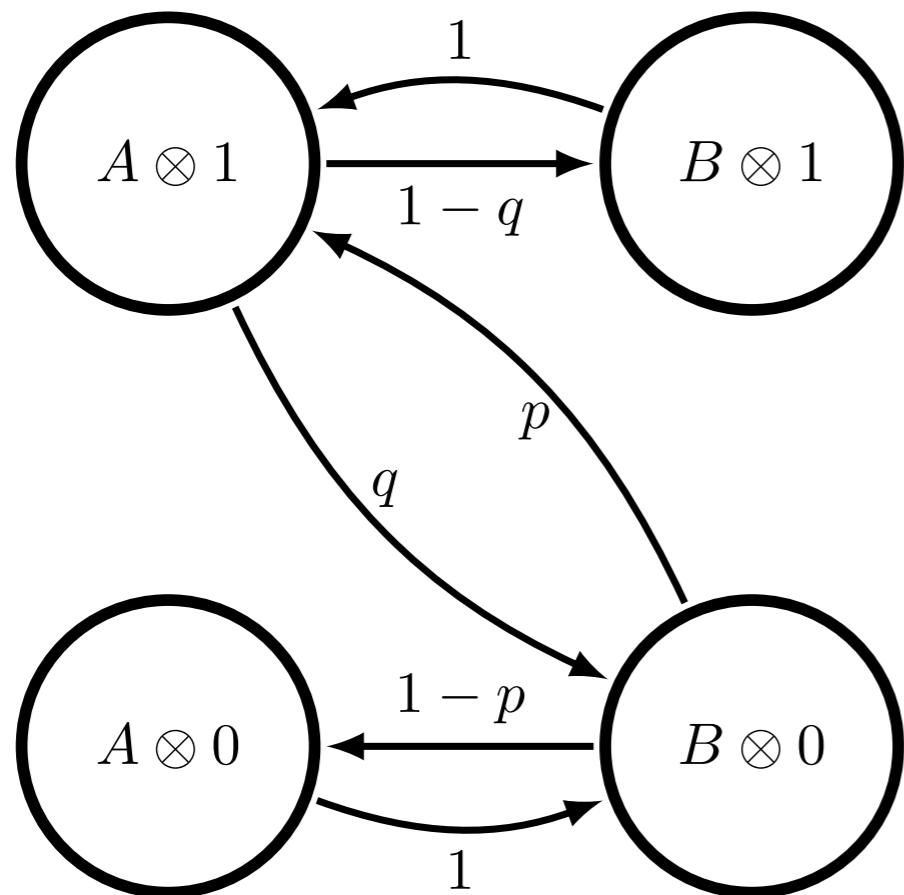
Ratchet and bit interact by evolving according to Markovian chain

$$M_{x_i \otimes y_i \rightarrow x_{i+1} \otimes y_i} = \Pr(X_{i+1} = x_{i+1}, Y'_i = y'_i | X_i = x_i, Y_i = y_i)$$

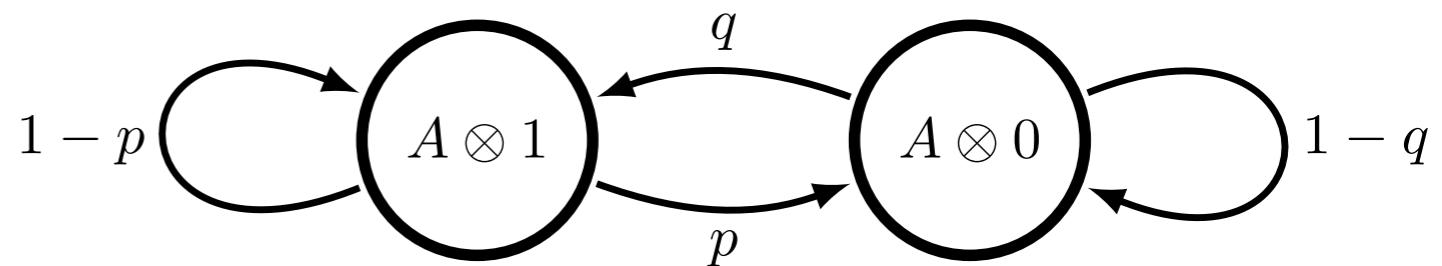


# Memory in Ratchets

Memoryful:

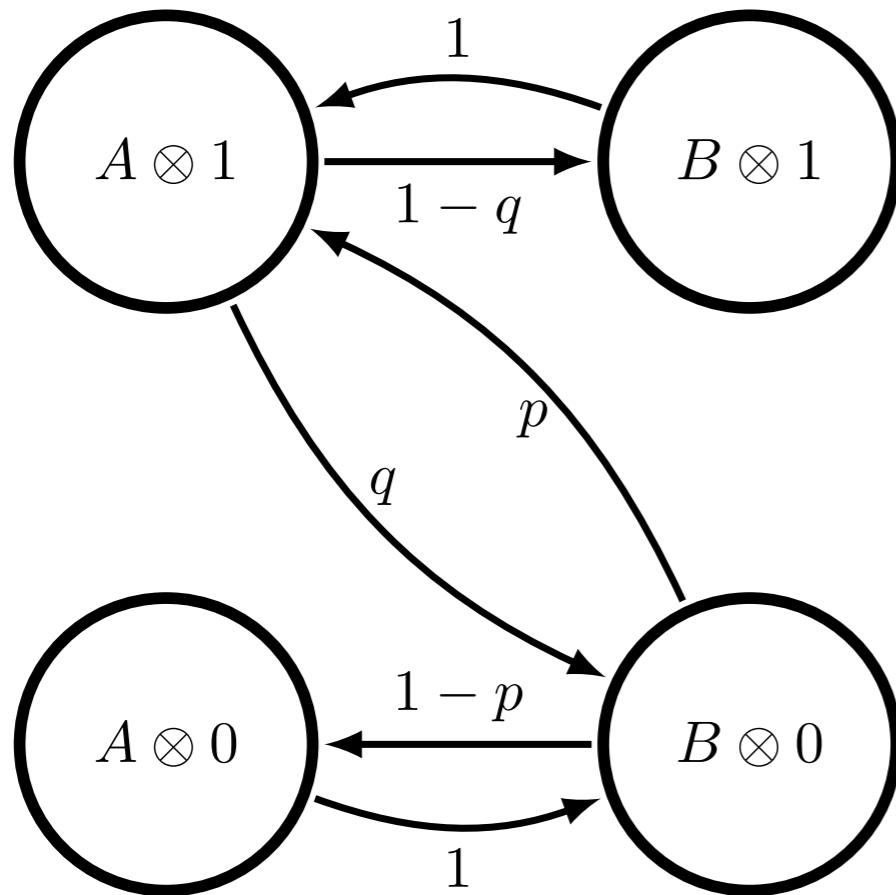


Memoryless:

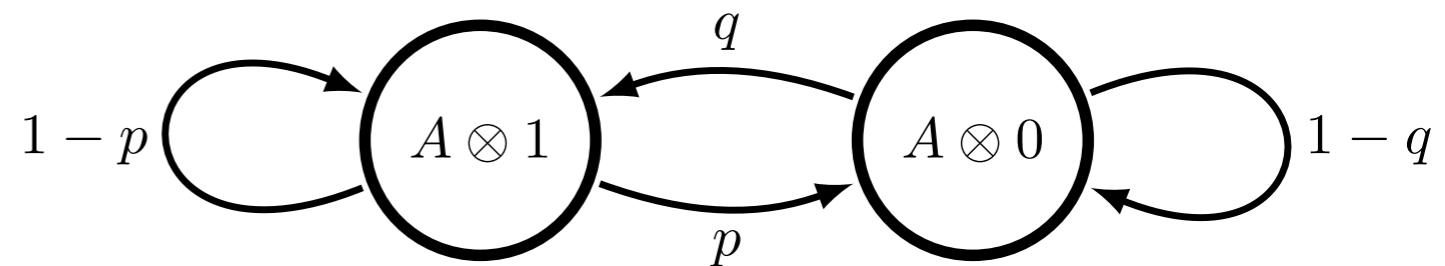


# Memory in Ratchets

Memoryful:



Memoryless:



Both have two bit states:

$$\mathcal{Y} = \{0, 1\}$$

multiple ratchet states:

$$\mathcal{X} = \{A, B\}$$

one ratchet state:

$$\mathcal{X} = \{A\}$$

# Memory and Information Bounds

Two candidate bounds on work production:

$$W \leq k_B T \ln 2 \Delta h_\mu$$

$$W \leq k_B T \ln 2 \Delta H(1)$$

## Ratchet

	memoryless	memoryful
memoryless	$W \leq \Delta H(1) = \Delta h_\mu$	$W \leq \Delta h_\mu \leq \Delta H(1)$
memoryful	$W \leq \Delta H(1) \leq \Delta h_\mu$	$W \leq \Delta h_\mu$ <del><math>W \leq \Delta H(1)</math></del>

Input

$$(k_B T \ln 2 = 1)$$

# Work Production

$$\langle W \rangle = \sum_{x,x' \in \mathcal{X} \atop y,y' \in \mathcal{Y}} \pi_{x,y} M_{x \otimes y \rightarrow x' \otimes y'} \Delta E_{x \otimes y \rightarrow x' \otimes y'}$$

The stationary distribution is highly dependent on the input

$$\pi_{x,y} = \lim_{N \rightarrow \infty} \Pr(X_N = x, Y_N = y)$$

Remaining factors are dependent on the ratchet transitions

$$\Delta E_{x \otimes y \rightarrow x' \otimes y'} = k_B T \ln \frac{M_{x' \otimes y' \rightarrow x \otimes y}}{M_{x \otimes y \rightarrow x' \otimes y'}}$$

# Memoryless Ratchets

Work done is only dependent on input symbol bias:

$$\langle W \rangle = \sum_{x,x' \in \mathcal{X} \atop y,y' \in \mathcal{Y}} \pi_{x,y} M_{x \otimes y \rightarrow x' \otimes y'} \Delta E_{x \otimes y \rightarrow x' \otimes y'}$$

$\uparrow$

$$\lim_{N \rightarrow \infty} \Pr(Y_N = y)$$

Work done is insensitive to order beyond single symbols. (Need memory to leverage correlations)

# Memoryless Ratchets

**ANY INPUT:** Change in single symbol entropy is a stricter and accurate bound (assumes no “feedback”, meaning no memory of the past outputs)

N. Merhav, J. Stat. Mech.: Theor. Exp. (2015) P06037

$$W \leq k_B T \ln 2 \Delta H(1) \leq k_B T \ln 2 \Delta h_\mu$$

**MEMORYLESS INPUT:** The outputs are also memoryless, so

$$\Delta h_\mu = h'_\mu - h_\mu = H[Y'_i] - H[Y_i] = \Delta H(1)$$

$$\langle W \rangle \leq k_B T \ln 2 \Delta H(1) = k_B T \ln 2 \Delta h_\mu$$

# Memoryless Ratchets

**ANY INPUT:** Change in single symbol entropy is a stricter and accurate bound (assumes no “feedback”, meaning no memory of the past outputs)

N. Merhav, J. Stat. Mech.: Theor. Exp. (2015) P06037

$$W \leq k_B T \ln 2 \Delta H(1) \leq k_B T \ln 2 \Delta h_\mu \quad \checkmark$$

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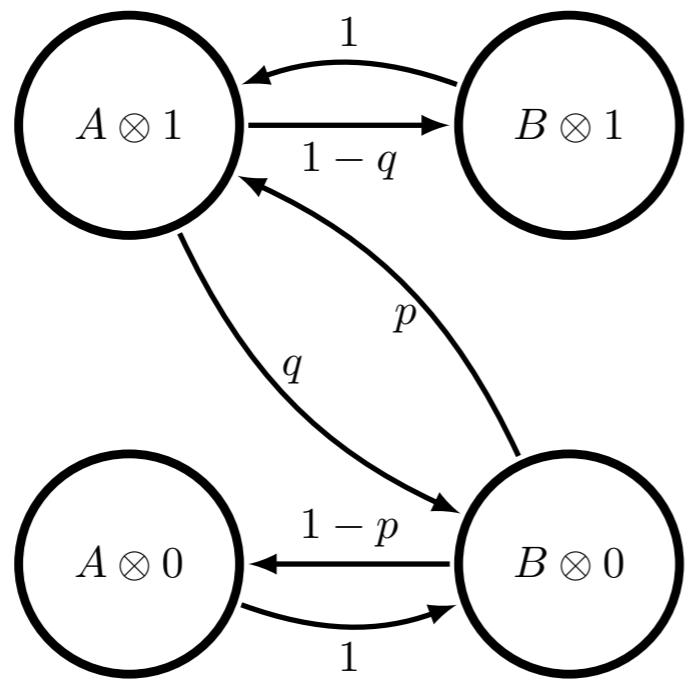
$$\langle W \rangle \leq k_B T \ln 2 \Delta H(1) = k_B T \ln 2 \Delta h_\mu \quad \checkmark$$

# Memoryful Ratchets Memoryless Inputs

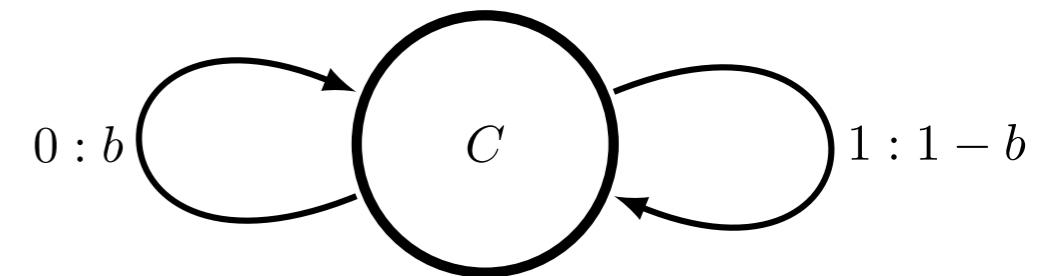
This is the case we've already demonstrated

A. B. Boyd, D. Mandal, and J. P. Crutchfield, arXiv: 1507.01537

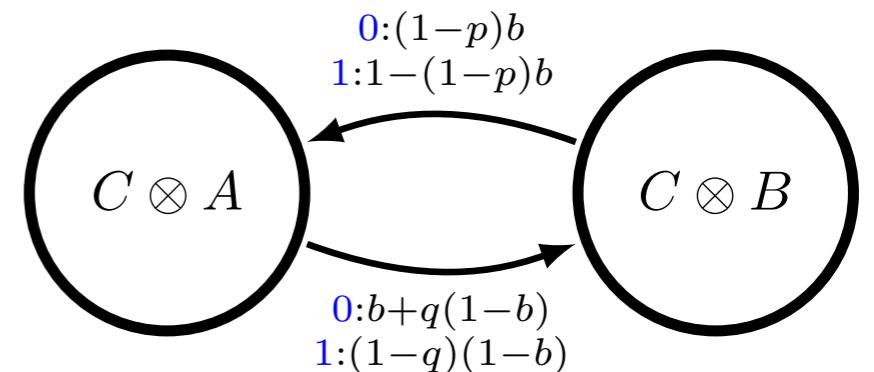
Ratchet:



In:



Out:



Input is IID, but output isn't necessarily.

$$\Delta h_\mu = h'_\mu - h_\mu = h'_\mu - H[Y_i] \leq H[Y'_i] - H[Y_i] = \Delta H(1)$$

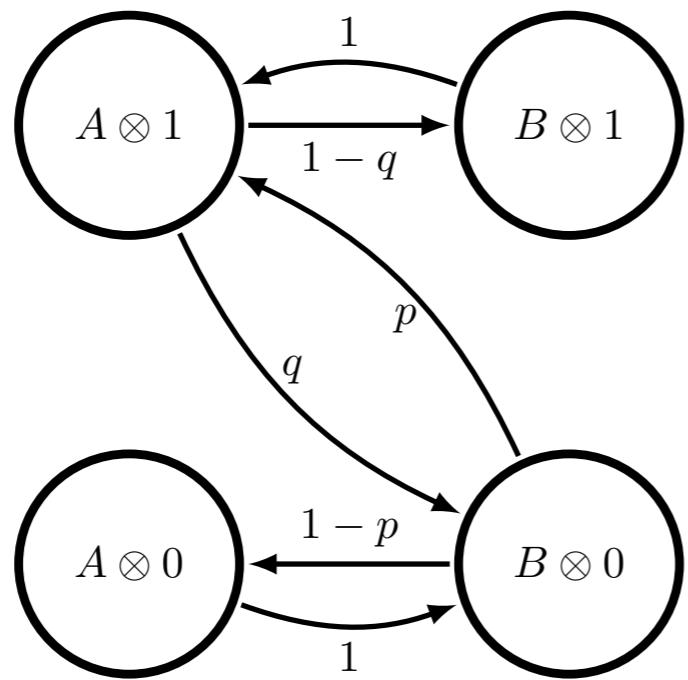
$$\langle W \rangle \leq k_B T \ln 2 \Delta h_\mu \leq k_B T \ln 2 \Delta H(1)$$

# Memoryful Ratchets Memoryless Inputs

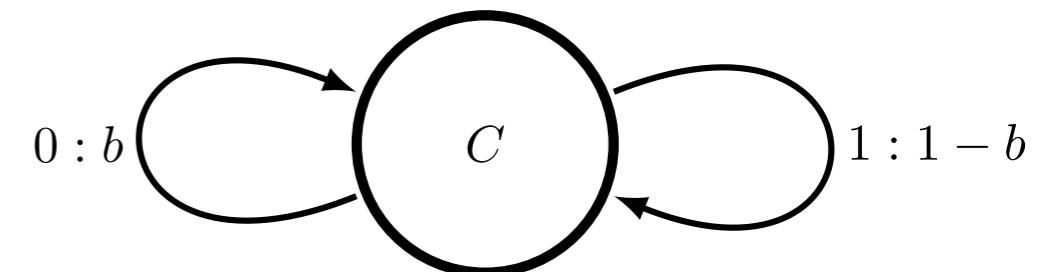
This is the case we've already demonstrated

A. B. Boyd, D. Mandal, and J. P. Crutchfield, arXiv: 1507.01537

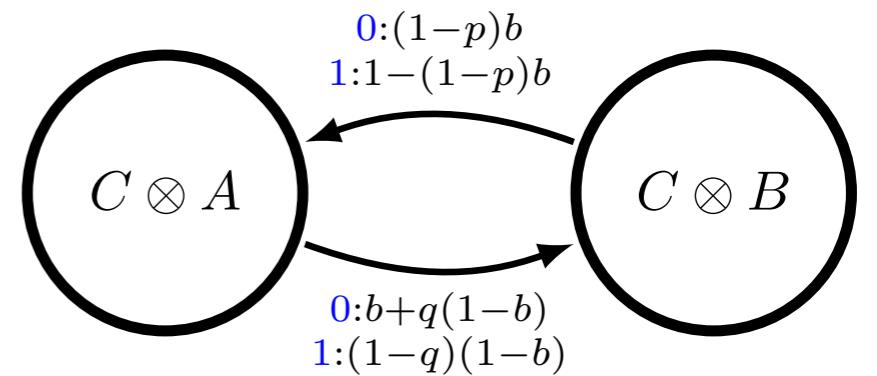
Ratchet:



In:



Out:



Input is IID, but output isn't necessarily.

$$\Delta h_\mu = h'_\mu - h_\mu = h'_\mu - H[Y_i] \leq H[Y'_i] - H[Y_i] = \Delta H(1)$$

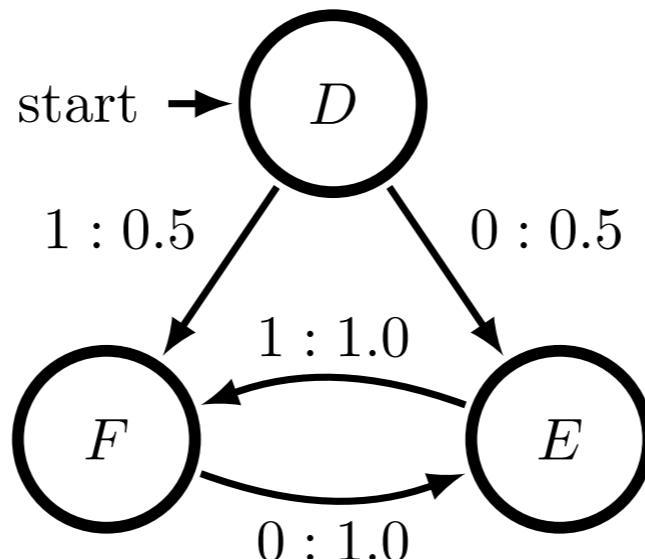
$$\langle W \rangle \leq k_B T \ln 2 \Delta h_\mu \leq k_B T \ln 2 \Delta H(1) \quad \checkmark$$

# Memoryful Ratchets Memoryful Inputs

Theorem:  $\langle W \rangle \leq k_B T \ln 2 \Delta h_\mu$  proven for all finite ratchets.

-Unknown if single symbol bounds hold.

-Consider correlated inputs: period-2



$$\Pr(Y_{0:\infty} = 010101\dots) = \Pr(Y_{0:\infty} = 101010\dots) = 1/2$$

$$h_\mu = 0$$

$$H[Y_i] = 1$$

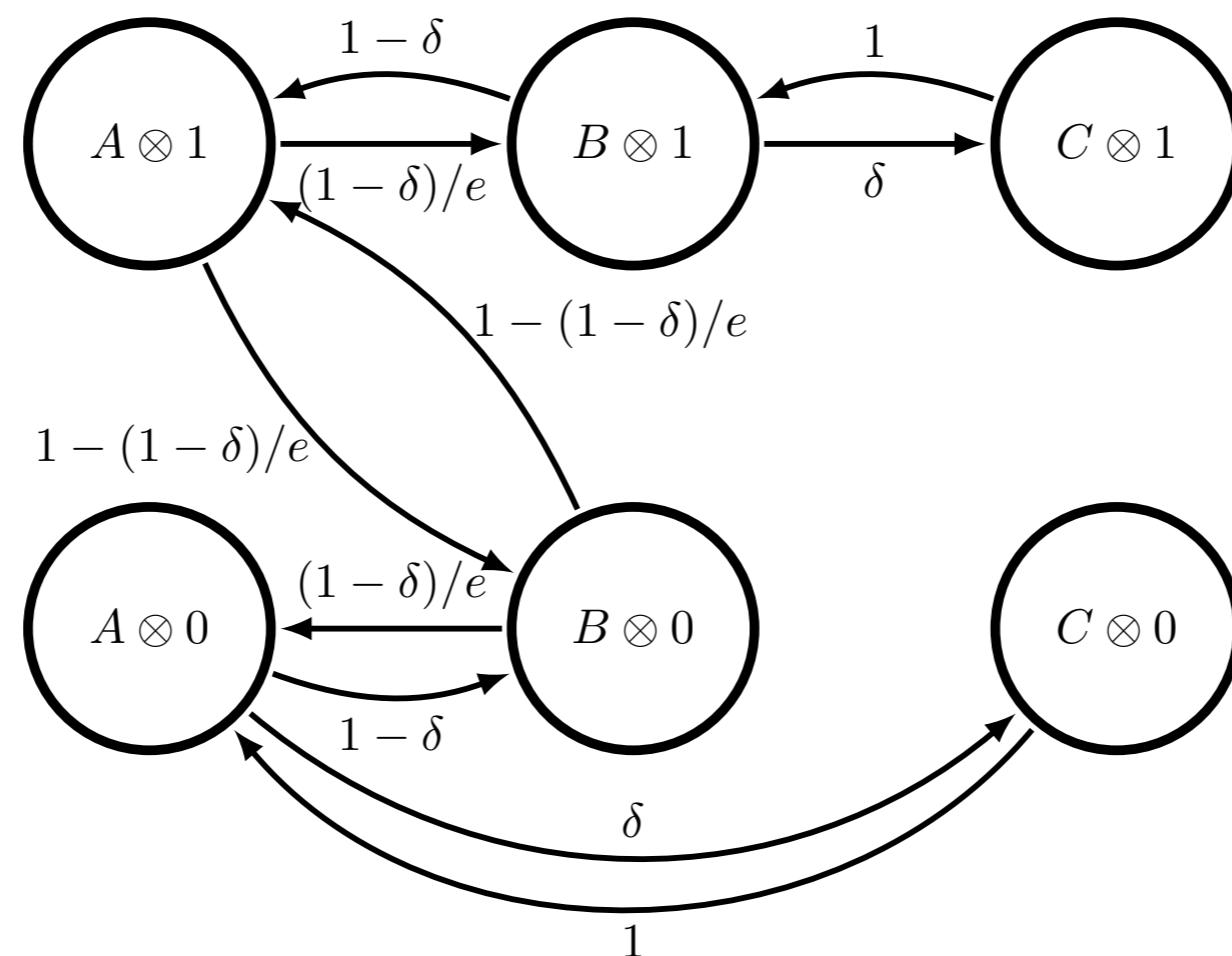
# Memoryful Ratchets Memoryful Inputs

Single symbol bound:  $\Delta H(1) \leq 0$   
suggests no work can be done.

Entropy rate bound:  $\Delta h_\mu \geq 0$   
suggests work can be done.

# Memoryful Ratchets Memoryful Inputs

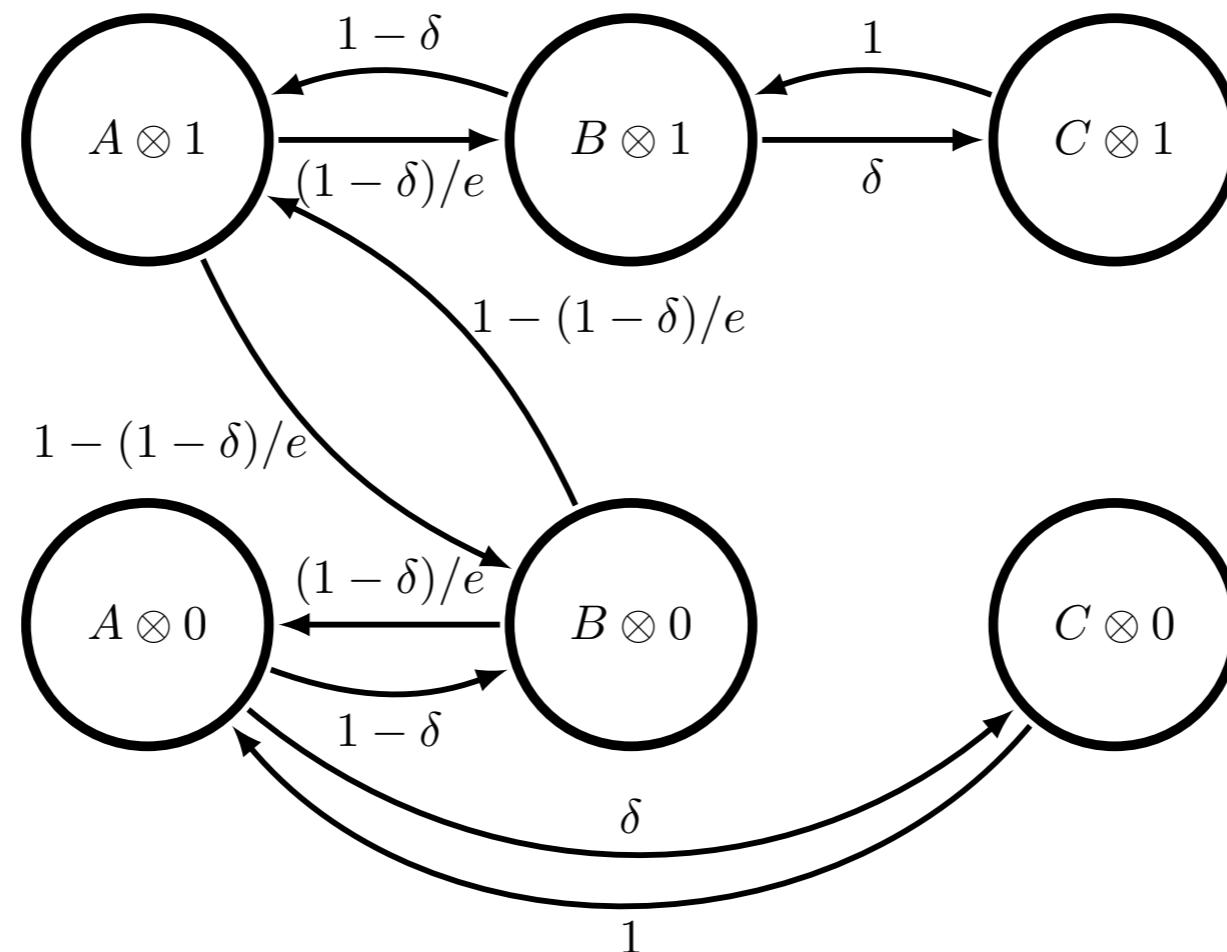
Period-2 Ratchet:



$$\langle W \rangle = \frac{1 - \delta}{e} > 0$$

# Memoryful Ratchets Memoryful Inputs

Period-2 Ratchet:



$$\langle W \rangle = \frac{1 - \delta}{e} > 0 \quad \checkmark$$

$$\langle W \rangle \leq \Delta h_\mu = H((1 - \delta)/e) \quad \checkmark$$

$$\cancel{\langle W \rangle \leq k_B T \ln 2\Delta H(1)} \quad \checkmark$$

# Memory and Information Bounds

Two candidate bounds on work production:

$$W \leq k_B T \ln 2 \Delta h_\mu$$

$$W \leq k_B T \ln 2 \Delta H(1)$$

## Ratchet

	memoryless	memoryful
memoryless	$W \leq \Delta H(1) = \Delta h_\mu$	$W \leq \Delta h_\mu \leq \Delta H(1)$
memoryful	$W \leq \Delta H(1) \leq \Delta h_\mu$	$W \leq \Delta h_\mu$ <del><math>W \leq \Delta H(1)</math></del>

Input

$$(k_B T \ln 2 = 1)$$

# Acknowledgements

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- James P. Crutchfield (UC Davis)
- Dibyendu Mandal (UC Berkeley)



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