

Directional Computational Mechanics I

Reading for this lecture: CMR articles

TBA

PRATISP

MSP

IACP, IACPLCOCS

Directional Computational Mechanics

Agenda:

Forward and reverse processes

Forward and reverse ε -machines

Causal irreversibility

Joint process lattice

Calculating reverse ε -machine from forward

Forward and Reverse Processes

Forward and Reverse Processes

Chain: $\overleftrightarrow{X} = \overleftarrow{X}_t \overrightarrow{X}_t$

Past: $\overleftarrow{X}_t = \dots X_{t-3} X_{t-2} X_{t-1}$

Future: $\overrightarrow{X}_t = X_t X_{t+1} X_{t+2} \dots$

Future L-Block: $\overrightarrow{X}_t^L = X_t X_{t+1} \dots X_{t+L-1}$

Past L-Block: $\overleftarrow{X}_t^L = X_{t-L} \dots X_{t-2} X_{t-1}$


Process: $\mathcal{P} \sim \text{Pr}(\overleftrightarrow{X})$

$$\text{Pr}(\overleftrightarrow{X}) = \text{Pr}(\dots X_{-2} X_{-1} X_0 X_1 X_2 \dots)$$

Forward and Reverse Processes

Previously,

$$\overleftrightarrow{X} = \dots X_{-2} X_{-1} X_0 X_1 X_2 \dots$$


Scan direction

Forward process: Same as before.

$$\mathcal{P}^+ \sim \text{Pr}(\overleftrightarrow{X})$$

Forward entropy rate ~ Prediction:

$$h_{\mu}^+ = h_{\mu}(\mathcal{P}^+) = H[X_0 | \overleftarrow{X}_0]$$

Forward and Reverse Processes

Now, **reverse process**:

$$\overleftrightarrow{X} = \dots X_{-2} X_{-1} X_0 X_1 X_2 \dots$$

←
Scan direction

Reverse process: Scan measurements in reverse order

$$\mathcal{P}^- \sim \text{Pr}(\overleftarrow{Y})$$

where $Y_t \equiv X_{-t}$

Typically, we will still write in terms of X_t .

Reverse entropy rate \sim Retrodiction:

$$\begin{aligned} h_\mu^- &= h_\mu(\mathcal{P}^-) = H[Y_0 | \overleftarrow{Y}_0] \\ &= H[X_{-1} | \overrightarrow{X}_0] \end{aligned}$$

(Note index shift)

Forward and Reverse Processes

In which time direction is process most predictable?

Neither!

Theorem: Entropy rate is time symmetric

$$h_{\mu}^{-} = h_{\mu}^{+}$$

Forward and Reverse Processes

In which time direction is process most predictable?

$$\begin{aligned}\text{Proof: } h_{\mu}^{+} &= H[X_0 | \overleftarrow{X}_0] \\ &= \lim_{L \rightarrow \infty} H[X_0 | X_{-L}, \dots, X_{-1}] \\ &= \lim_{L \rightarrow \infty} [H[X_{-L}, \dots, X_0] - H[X_{-L}, \dots, X_{-1}]]\end{aligned}$$

$$\text{Stationarity: } H[X_{-L}, \dots, X_{-1}] = H[X_{-L+1}, \dots, X_0]$$

$$\begin{aligned}h_{\mu}^{+} &= \lim_{L \rightarrow \infty} [H[X_{-L}, \dots, X_0] - H[X_{-L+1}, \dots, X_0]] \\ &= \lim_{L \rightarrow \infty} H[X_{-L} | X_{-L+1}, \dots, X_0] && \text{(Stationarity, again)} \\ &= \lim_{L \rightarrow \infty} H[X_{-1} | X_0, \dots, X_{L-1}] \\ &= H[X_{-1} | \overrightarrow{X}_0] \\ &= h_{\mu}^{-} \quad (= H[Y_0 | \overleftarrow{Y}_0])\end{aligned}$$

Forward and Reverse Processes

Does the communicated information differ?

Theorem: Excess entropy is time symmetric

$$\mathbf{E}(\mathcal{P}^+) = \mathbf{E}(\mathcal{P}^-)$$

Forward and Reverse Processes

Does the communicated information differ?

Proof sketch:

$$\begin{aligned}\mathbf{E}(\mathcal{P}^+) &= I[\overleftarrow{X}_0; \overrightarrow{X}_0] \\ &= \lim_{L \rightarrow \infty} I[\overleftarrow{X}_0^L; \overrightarrow{X}_0^L] \\ &= \lim_{L \rightarrow \infty} I[\overrightarrow{X}_0^L; \overleftarrow{X}_0^L] \\ &= \lim_{L \rightarrow \infty} I[\overleftarrow{Y}_0^L; \overrightarrow{Y}_0^L] \\ &= I[\overleftarrow{Y}_0; \overrightarrow{Y}_0] \\ &= \mathbf{E}(\mathcal{P}^-)\end{aligned}$$

$$I[W; Z] = I[Z; W]$$

Definition of Y_t

Forward and Reverse Processes

Conclusion:

Neither entropy rate nor excess entropy
detect temporal asymmetry

Does the stored information differ?

First, we need the ε -machine of the reverse process: $M(\overleftarrow{\mathcal{P}})$

Forward and Reverse ϵ -Machines

Forward and Reverse ϵ -Machines

Previously, $\overleftrightarrow{X} = \dots X_{-2} X_{-1} X_0 X_1 X_2 \dots$

$\xrightarrow{\hspace{10em}}$
Scan direction

Forward ϵ -Machine: M^+

Equivalence Relation $\overleftarrow{x} \sim^+ \overleftarrow{x}' : \epsilon^+(\overleftarrow{x})$

Forward Causal States: $\mathcal{S}^+ = \mathcal{P}^+ / \sim^+$

Complexity measures:

Forward entropy rate: $h_\mu^+ = H[X_0 | \mathcal{S}_0^+] = h_\mu$

Statistical complexity: $C_\mu^+ = H[\mathcal{S}^+]$

Forward crypticity: $\chi^+ = H[\mathcal{S}_0^+ | \overrightarrow{X}_0]$

Mystery wedge!
 $C_\mu - \mathbf{E}$

Forward and Reverse ϵ -Machines

Now, “reverse” ϵ -Machines: $\overleftrightarrow{X} = \dots X_{-2} X_{-1} X_0 X_1 X_2 \dots$
← Scan direction

Retrodictive equivalence relation: $\overrightarrow{x} \sim^- \overrightarrow{x}'$

$$\epsilon^-(\overrightarrow{x}) = \{ \overrightarrow{x}' : \Pr(\overleftarrow{X} | \overrightarrow{x}) = \Pr(\overleftarrow{X} | \overrightarrow{x}') \}$$

Retrodictive causal states: $\mathcal{S}^- = \mathcal{P}^- / \sim^-$

Reverse ϵ -Machine: $M^- = \{ \mathcal{S}^-, T^{(s)} \in \mathcal{A} \}$

Forward and Reverse ε -Machines

Reverse ε -Machines ...

$$\overleftarrow{X} = \dots X_{-2} X_{-1} X_0 X_1 X_2 \dots$$

\longleftarrow
Scan direction

Retrodictive entropy rate: $h_{\mu}^{-} = H[X_{-1} | \mathcal{S}_0^{-}] = h_{\mu}$

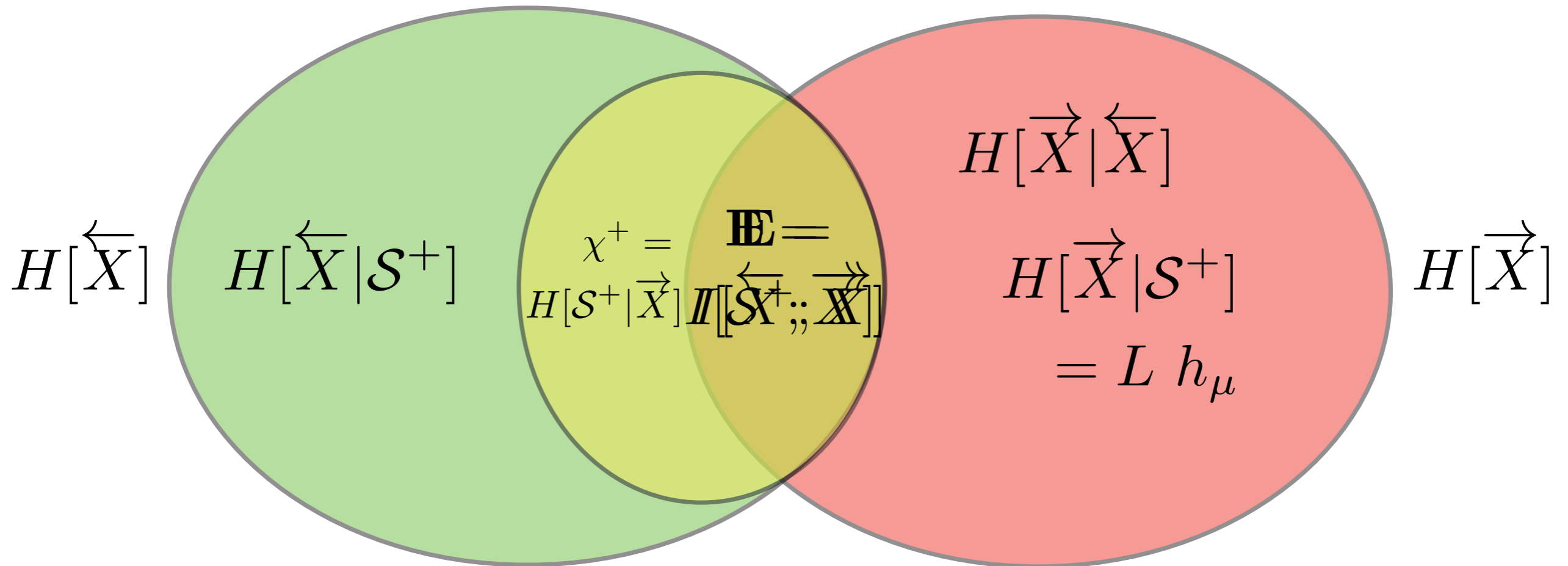
Reverse statistical complexity: $C_{\mu}^{-} = H[\mathcal{S}^{-}]$

Reverse crypticity: $\chi^{-} = H[\mathcal{S}_0^{-} | \overleftarrow{X}_0]$

Forward and Reverse ε -Machines

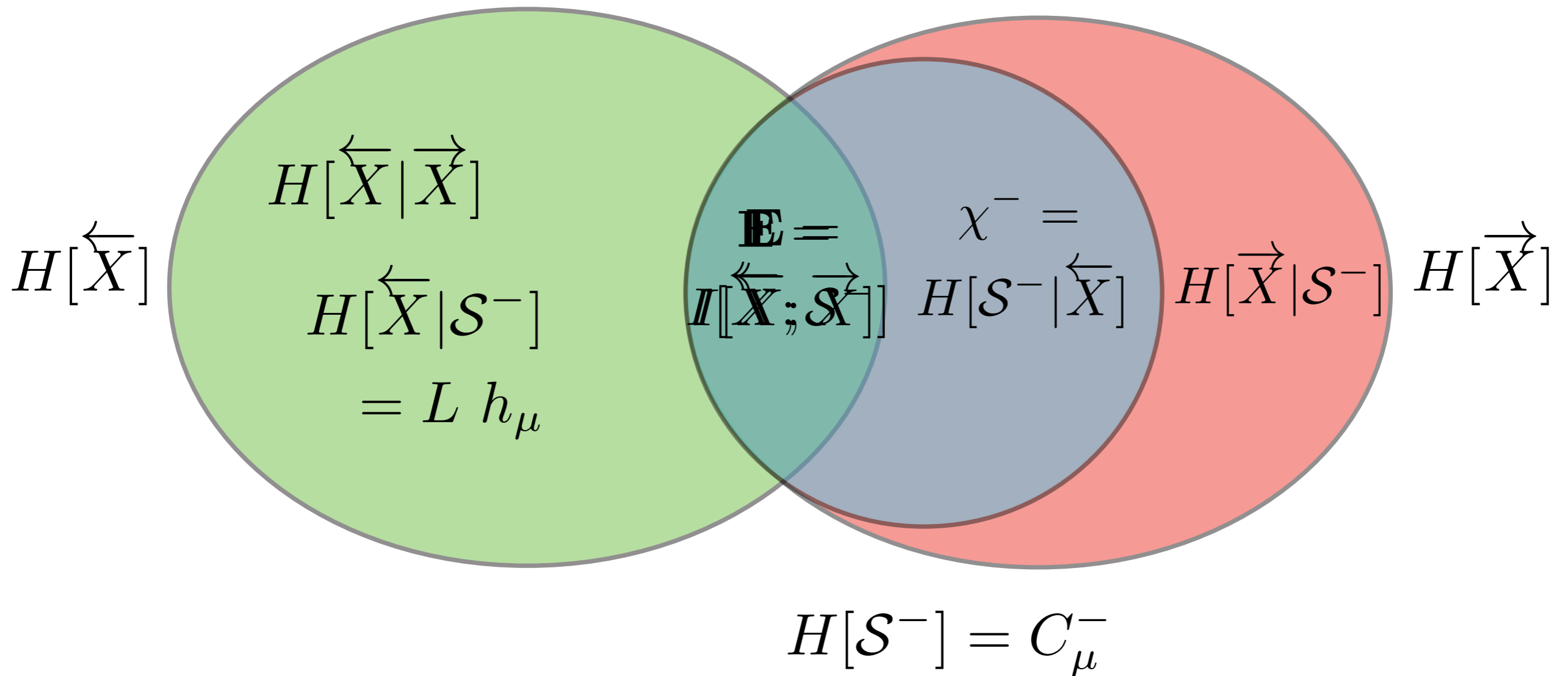
Forward process ε -machine I-diagram:

$$H[\mathcal{S}^+] = C_\mu^+$$



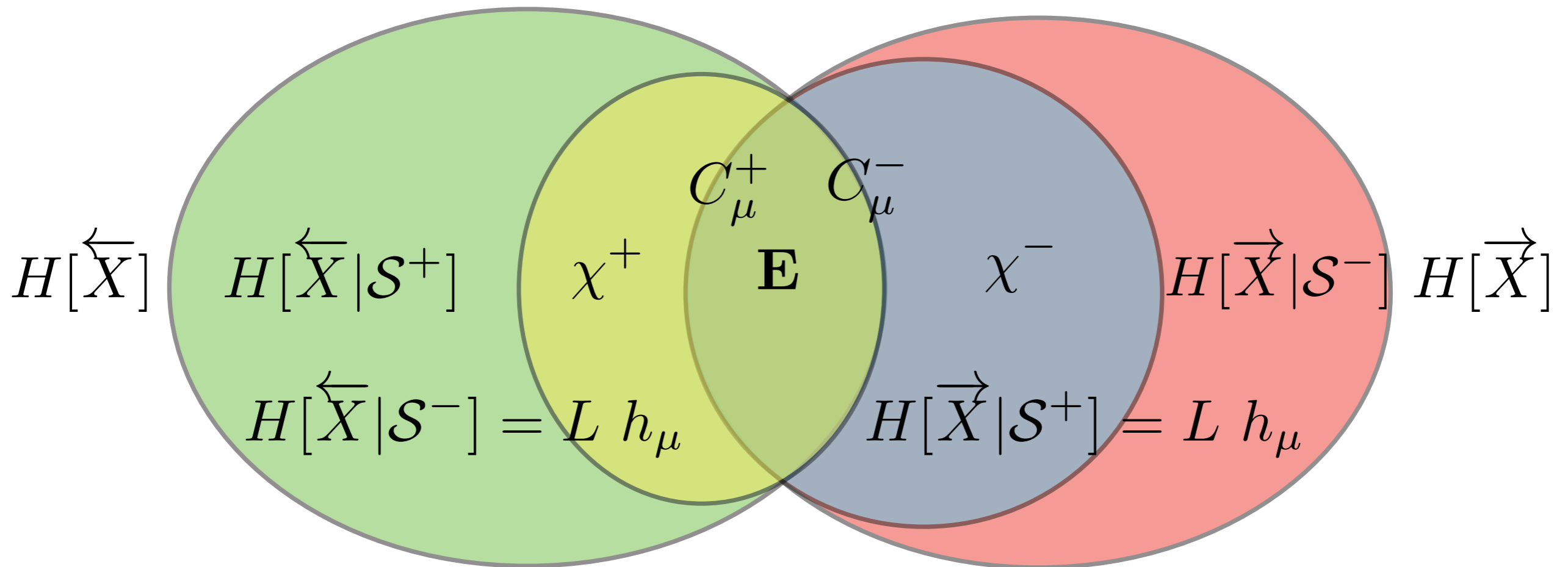
Forward and Reverse ε -Machines

Reverse process ε -machine I-diagram:



Forward and Reverse ε -Machines

Forward and reverse process ε -machine I-diagram:



Forward and Reverse ϵ -Machines

Two remaining mystery wedges!

“Error” associated with retrodiction using forward ϵ -machines:

$$H[\overleftarrow{X} | \mathcal{S}^+]$$

“Error” associated with prediction using reverse ϵ -machines:

$$H[\overrightarrow{X} | \mathcal{S}^-]$$

Forward and Reverse ε -Machines

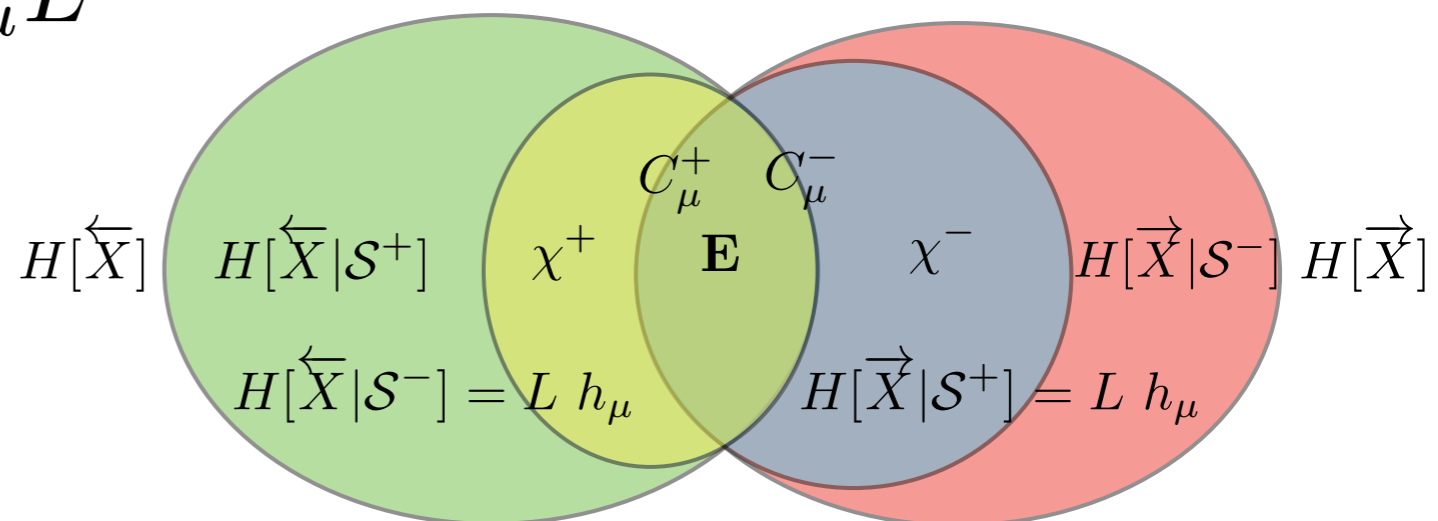
Mysterious, perhaps, but well characterized.

Corollary:

$$H[\overleftarrow{X}^L | \mathcal{S}^+] = h_\mu L - \chi^+$$

Proof sketch:

$$\begin{aligned} \chi^+ + H[\overleftarrow{X}^L | \mathcal{S}^+] &= H[\overleftarrow{X}^L | \overrightarrow{X}] \\ &= H[\overleftarrow{X}^L | \mathcal{S}^-] \\ &= h_\mu L \end{aligned}$$



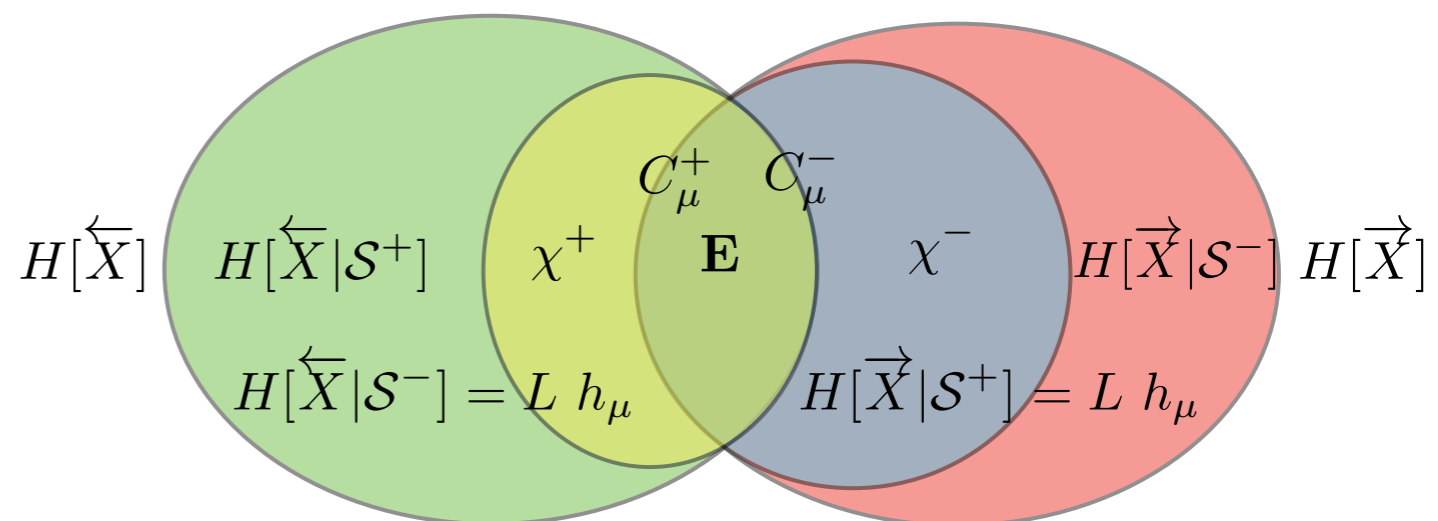
Forward and Reverse ε -Machines

Corollary:

$$H[\vec{X}^L | \mathcal{S}^-] = h_\mu L - \chi^-$$

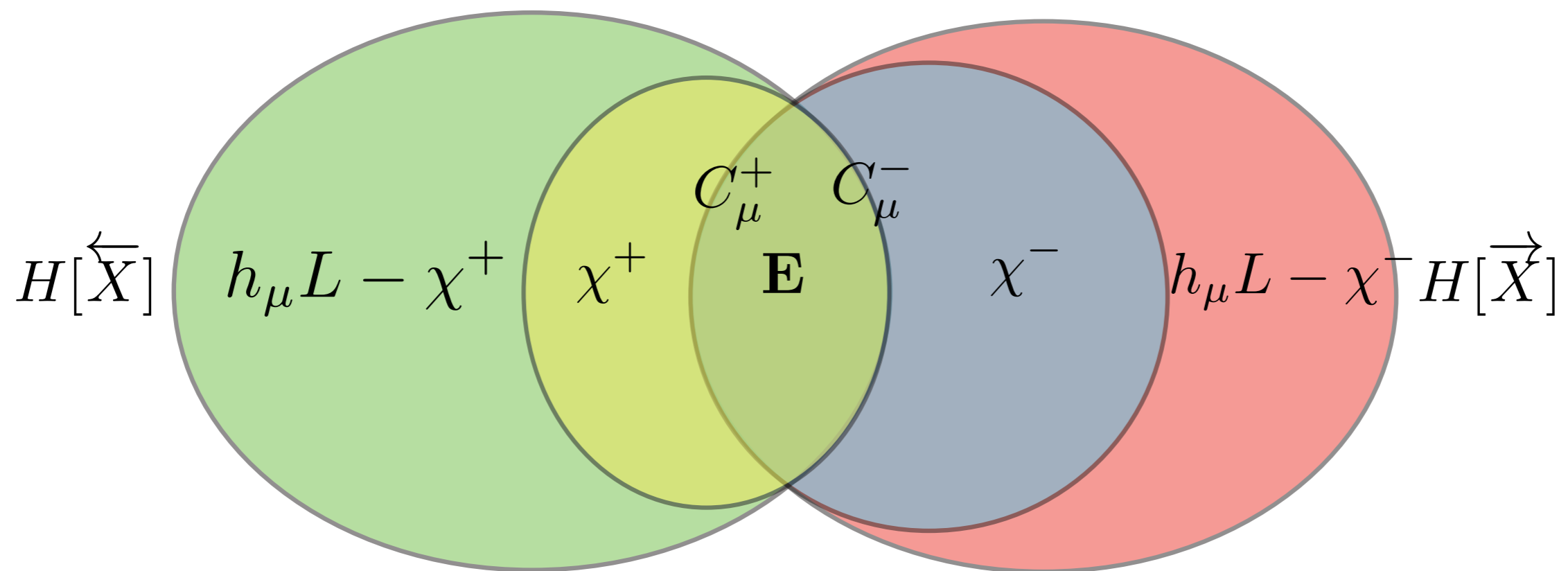
Proof sketch:

$$\begin{aligned} \chi^- + H[\vec{X}^L | \mathcal{S}^-] &= H[\vec{X}^L | \overleftarrow{X}] \\ &= H[\vec{X}^L | \mathcal{S}^+] \\ &= h_\mu L \end{aligned}$$



Forward and Reverse ε -Machines

Result: Characterized the entire I-diagram



Forward and Reverse Joint Processes

Joint Process Lattice:

A graphi-notational interlude

Forward and Reverse Joint Processes

Joint Processes:

Various:

- **State-symbol:**

$$\Pr(\overleftrightarrow{\mathcal{S}}, \overleftrightarrow{X}) = (\dots, (\sigma_t, x_t), \dots)$$

- **Forward state-symbol:**

$$\Pr(\overleftrightarrow{\mathcal{S}}^+, \overleftrightarrow{X}) = (\dots, (\sigma_t^+, x_t), \dots)$$

- **Reverse state-symbol:**

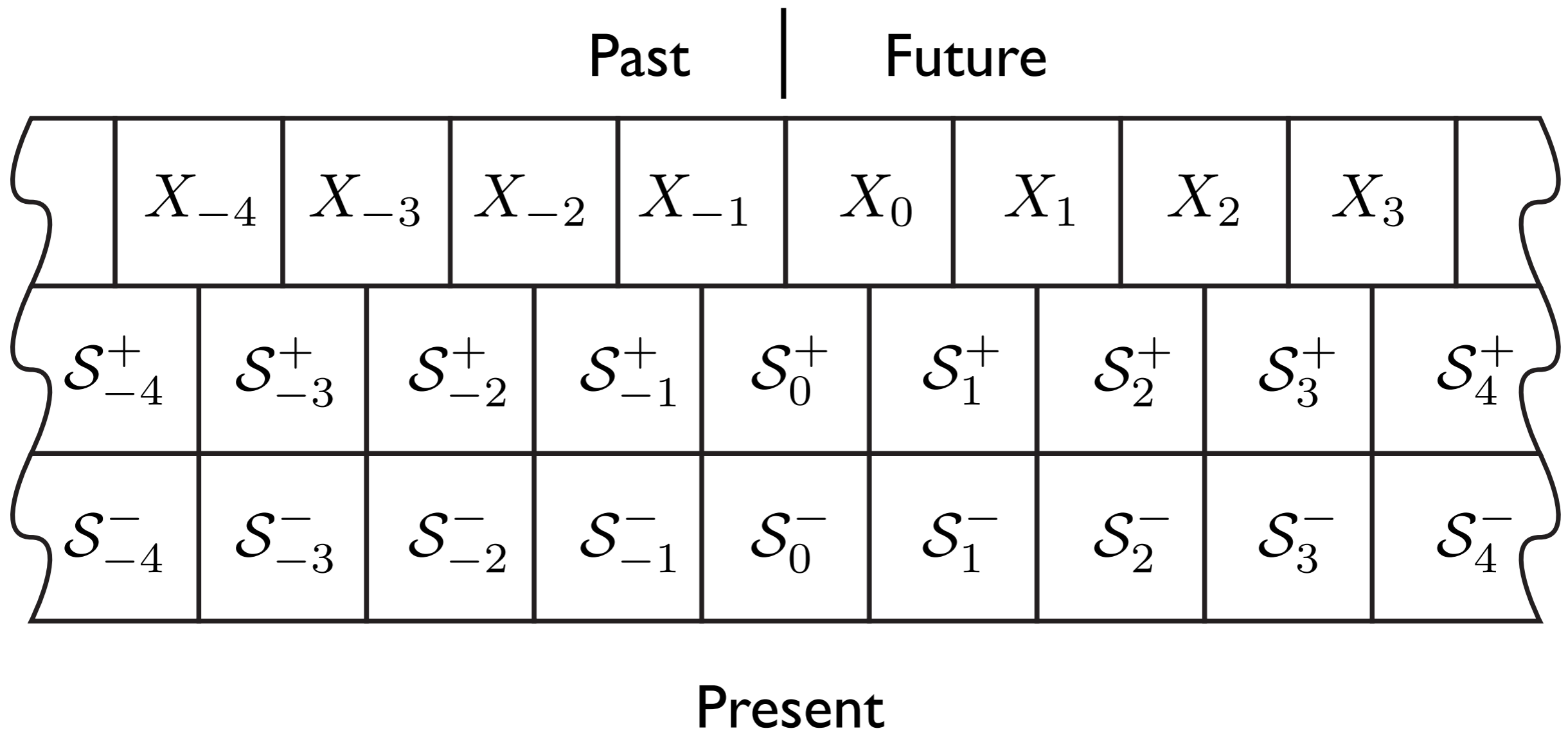
$$\Pr(\overleftrightarrow{\mathcal{S}}^-, \overleftrightarrow{X}) = (\dots, (\sigma_t^-, x_t), \dots)$$

- **Forward state, reverse state, symbol:**

$$\Pr(\overleftrightarrow{\mathcal{S}}^-, \overleftrightarrow{\mathcal{S}}^+, \overleftrightarrow{X}) = (\dots, (\sigma_t^-, \sigma_t^+, x_t), \dots)$$

Forward and Reverse Joint Processes

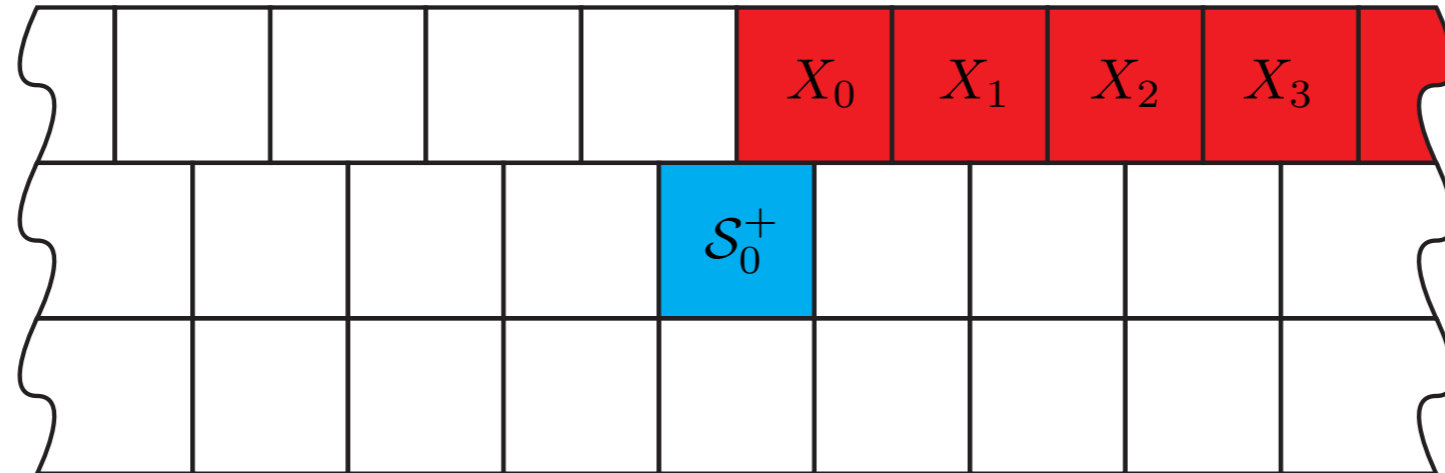
Joint Process Lattice: Guide book



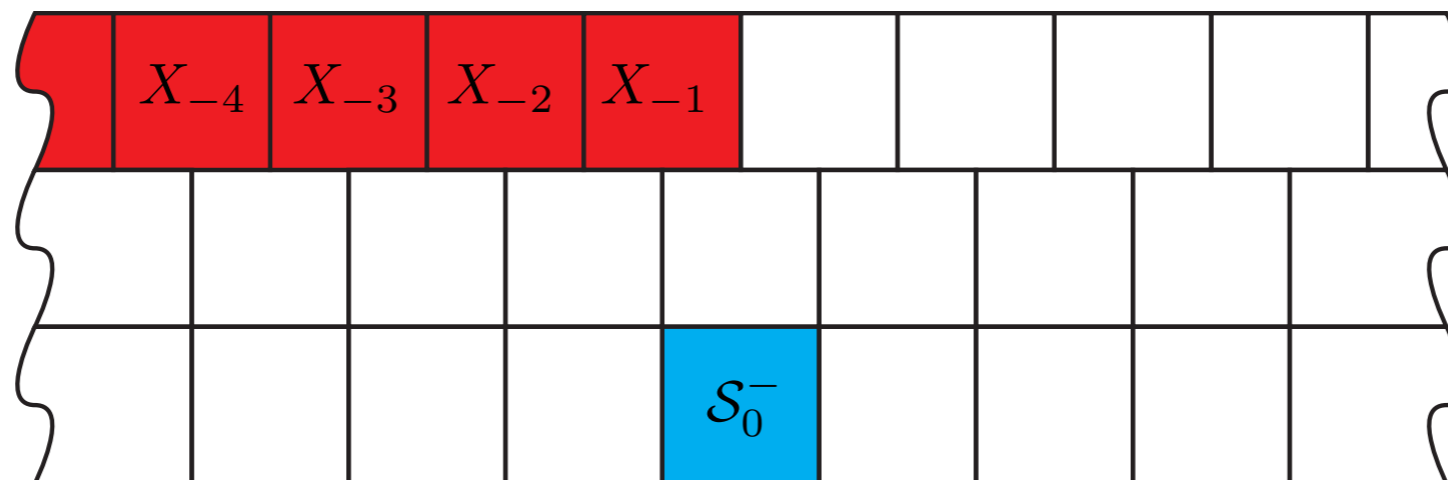
Forward and Reverse Joint Processes

Joint Process Lattice ...

Prediction using forward ε -machines: $H[\vec{X} | \mathcal{S}^+]$



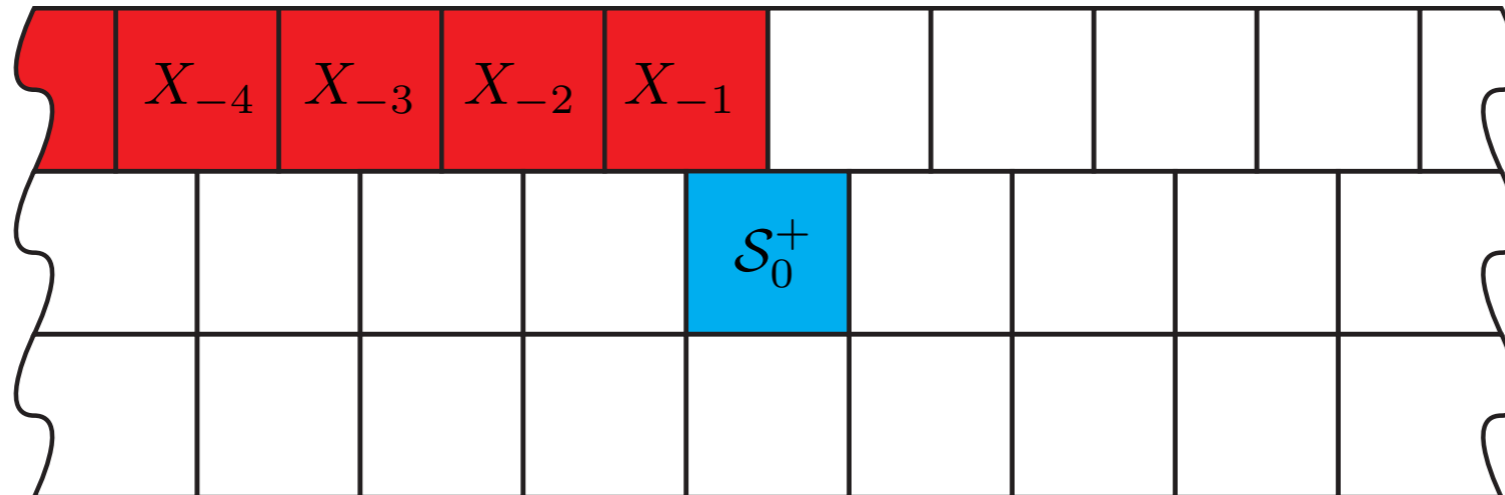
Retrodiction using reverse ε -machines: $H[\overleftarrow{X} | \mathcal{S}^-]$



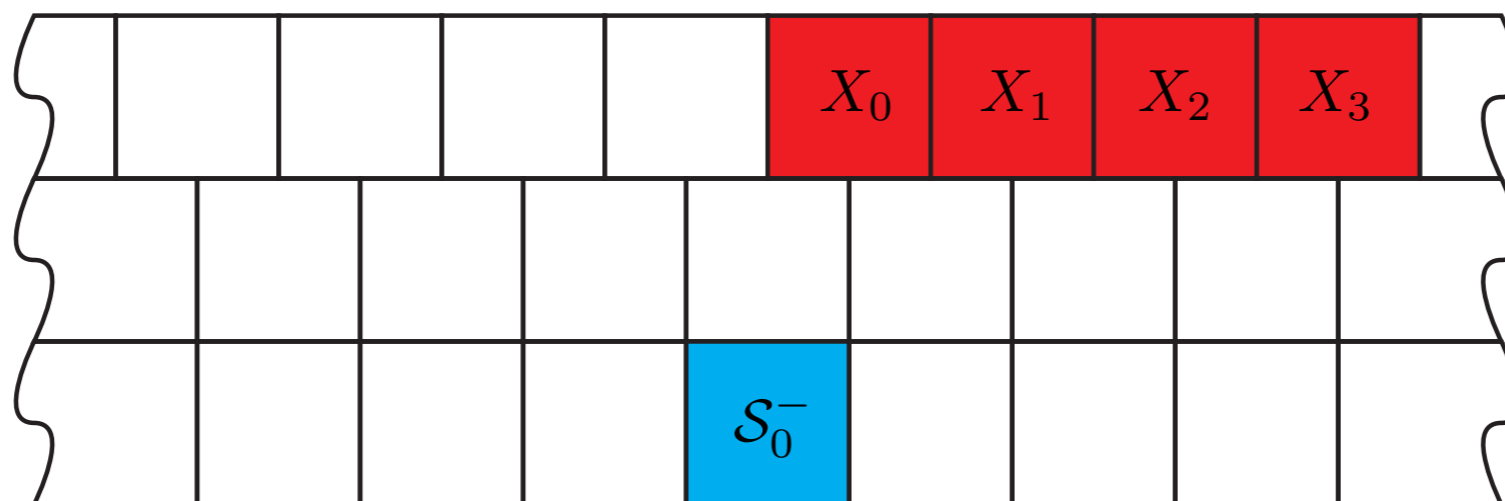
Forward and Reverse Joint Processes

Joint Process Lattice ...

Retrodiction using forward ε -machines: $H[\overleftarrow{X}^L | \mathcal{S}^+]$



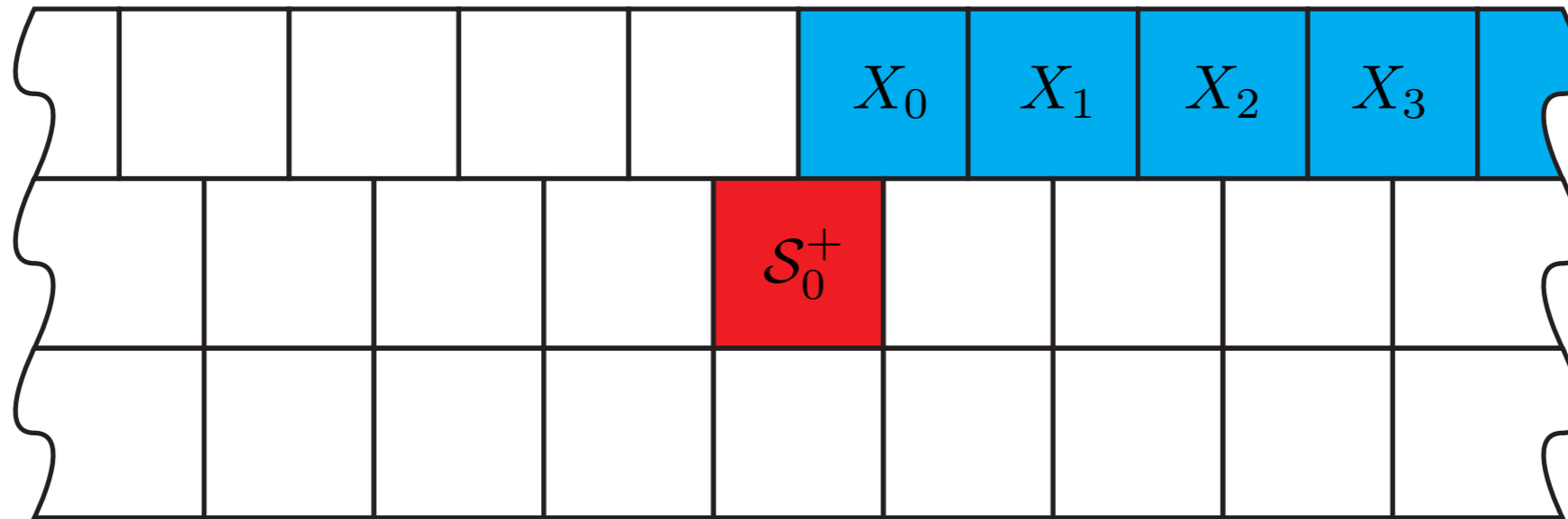
Prediction using reverse ε -machines: $H[\overrightarrow{X}^L | \mathcal{S}^-]$



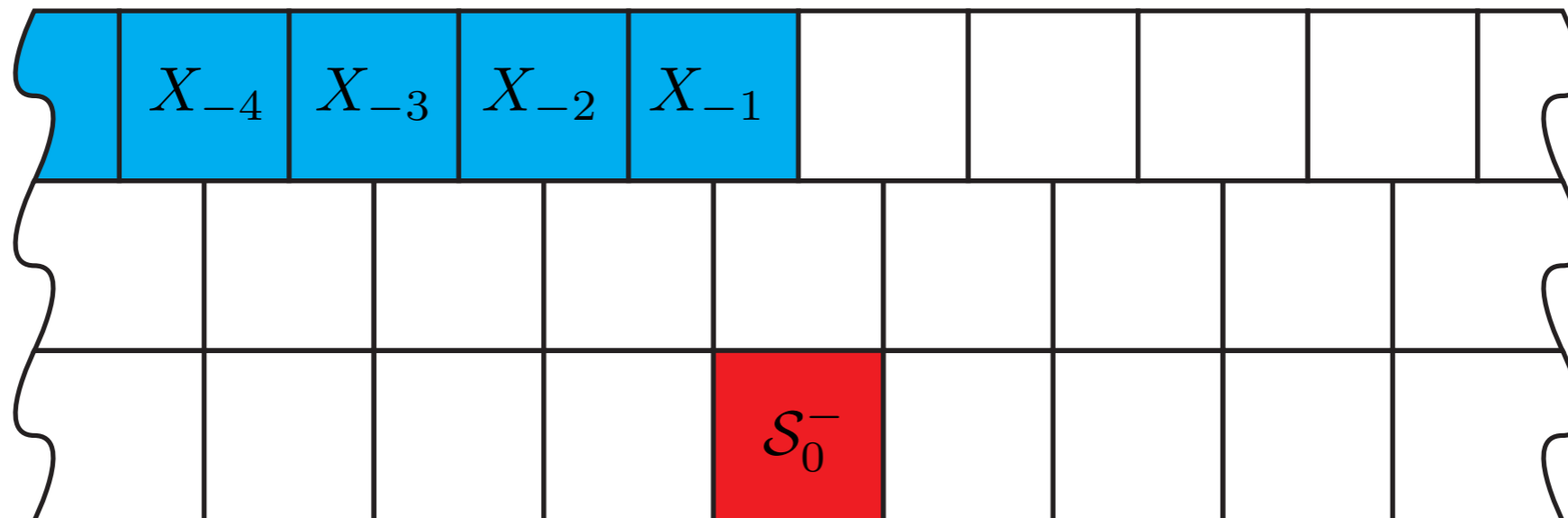
Forward and Reverse Joint Processes

Joint Process Lattice ...

Forward crypticity: $\chi^+ = H[S^+ | \vec{X}]$



Reverse crypticity: $\chi^- = H[S^- | \overleftarrow{X}]$



Causal Irreversibility

Causal Irreversibility

Time symmetric properties:

Forward and reverse entropy rates:

$$h_{\mu} = h_{\mu}^{+} = h_{\mu}^{-}$$

Excess entropy of forward and reverse processes.

$$\mathbf{E} = \mathbf{E}(\mathcal{P}^{+}) = \mathbf{E}(\mathcal{P}^{-})$$

Causal Irreversibility

Can the stored information differ?

Yes!

Theorem: ϵ -Machines need not be time symmetric

$$\overleftarrow{M} \neq \overrightarrow{M}$$

Corollary: Statistical complexity need not be time symmetric

$$C_{\mu}^{-} \neq C_{\mu}^{+}$$

Causal Irreversibility

Can the stored information differ?

Proof: By example ...

Misiurewicz parameter in the Logistic map:

First root, $r < 4$, where critical point is periodic and

$$f^4\left(\frac{1}{2}\right) = f^5\left(\frac{1}{2}\right)$$

Find

$$r \approx 3.9277370017867516$$

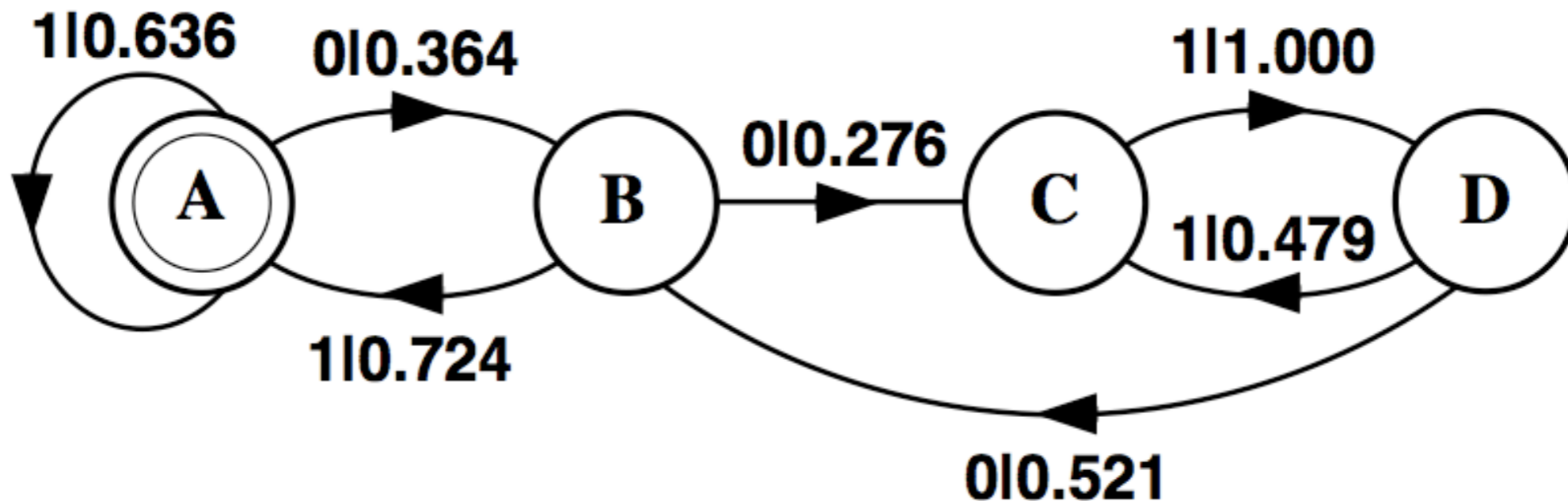
Use binary generating partition.

Causal Irreversibility

Can the stored information differ?

Proof: By example ... Misiurewicz parameter for Logistic map ...

Forward ϵ -machine:



$$h_{\mu}(\vec{\mathcal{P}}) \approx 0.81 \text{ bits/symbol}$$

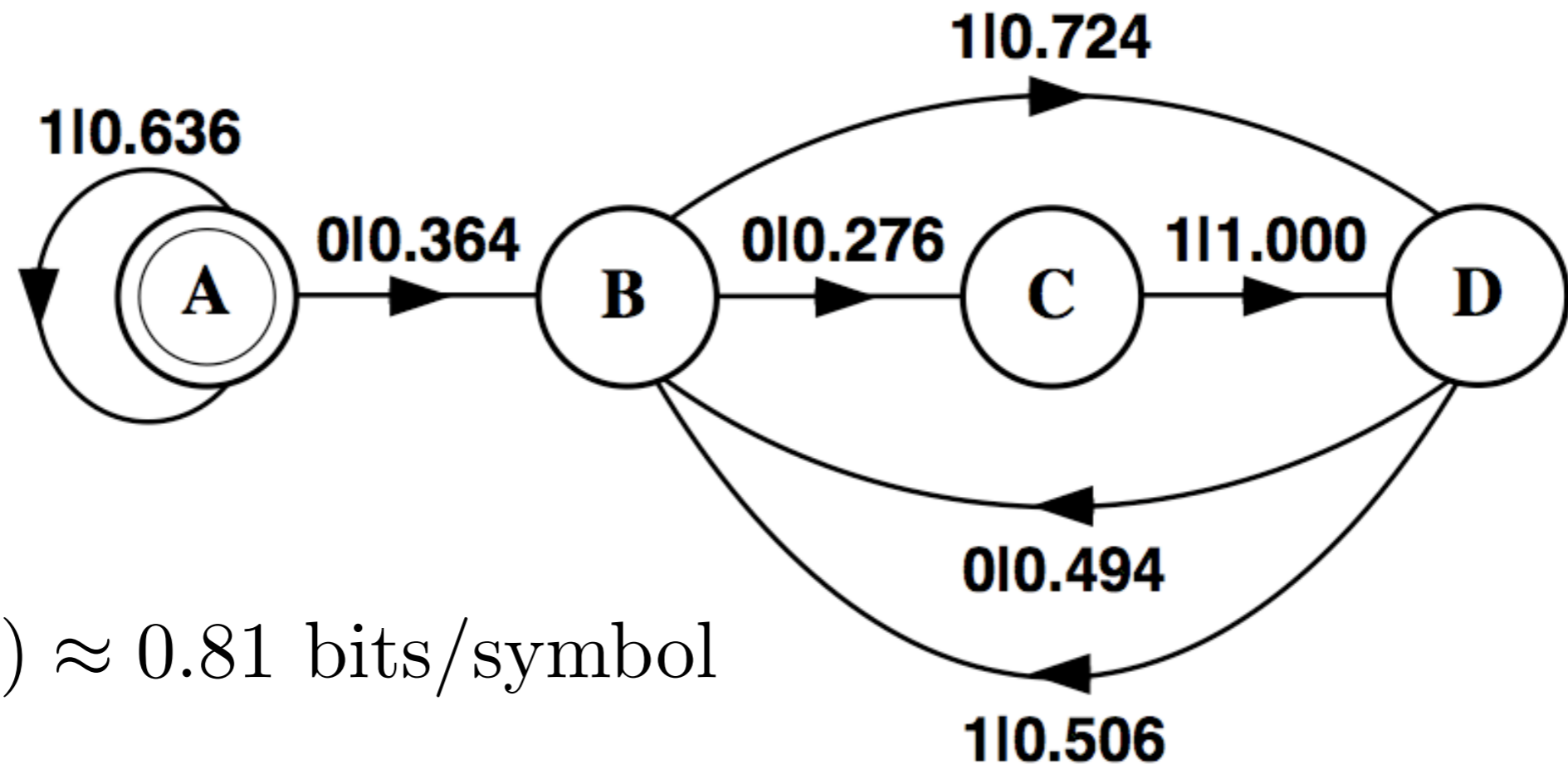
$$C_{\mu}(\vec{\mathcal{P}}) \approx 1.77 \text{ bits}$$

Causal Irreversibility

Can the stored information differ?

Proof: By example ... Misiurewicz parameter for Logistic map ...

Reverse ε -machine:



$$h_{\mu}(\overleftarrow{\mathcal{P}}) \approx 0.81 \text{ bits/symbol}$$

$$C_{\mu}(\overleftarrow{\mathcal{P}}) \approx 1.41 \text{ bits}$$

Causal Irreversibility

Temporal asymmetry in process structure.

Even if in (statistical) equilibrium!

Interpretation:

Stored information required to do optimal prediction can differ,
depending on scan (time) direction.

Causal Irreversibility

Alternative notions of reversibility:

Most familiar: Microscopically reversible processes

Physics invariant under $t \Leftrightarrow -t$.

Here, what would this mean?

Given measurement sequence:

$$w = x_0 x_1 \dots x_n$$

Reverse time order of measurements:

$$\begin{aligned} \tilde{w} &= y_0 y_1 \dots y_n, \quad y_i = x_{n-i} \\ &= x_n x_{n-1} \dots x_0 \end{aligned}$$

Causal Irreversibility

Alternative notions of reversibility ...

Microscopically reversible process:

$$\mathcal{P}^+ = \mathcal{P}^-$$

Given $w = \text{supp } \text{Pr}(w)$

$$\text{Pr}(w) = \text{Pr}(\tilde{w})$$

\mathcal{P}^+ 's word distribution

Causal Irreversibility

Alternative notions of reversibility ...

Causal reversibility:

$$C_{\mu}^{+} = C_{\mu}^{-}$$

Microscopic reversibility \Rightarrow Causal reversibility

$$\begin{aligned} \mathcal{P}^{+} = \mathcal{P}^{-} &\Leftrightarrow M^{+} = M^{-} \\ &\Rightarrow C_{\mu}^{+} = C_{\mu}^{-} \end{aligned}$$

Causal Irreversibility

Causal Irreversibility:

$$\begin{aligned}\Xi &= C_{\mu}(\mathcal{P}^+) - C_{\mu}(\mathcal{P}^-) \\ &= C_{\mu}^+ - C_{\mu}^-\end{aligned}$$

A measure of a process's time asymmetry

Misiurewicz process causal irreversibility:

$$\Xi(\mathcal{P}) \approx 0.36 \text{ bits}$$

Causal irreversibility \Rightarrow Microscopic irreversibility

$$\begin{aligned}\Xi \neq 0 &\Rightarrow C_{\mu}^+ \neq C_{\mu}^- \\ &\Rightarrow M^+ \neq M^- \\ &\Rightarrow \mathcal{P}^+ \neq \mathcal{P}^-\end{aligned}$$

Causal irreversibility less restrictive: Structural similarity

Causal Irreversibility

Crypticity and reversibility:

$$\text{Forward: } \chi^+ = H[\mathcal{S}^+ | \vec{X}]$$

$$\text{Reverse: } \chi^- = H[\mathcal{S}^- | \overleftarrow{X}]$$

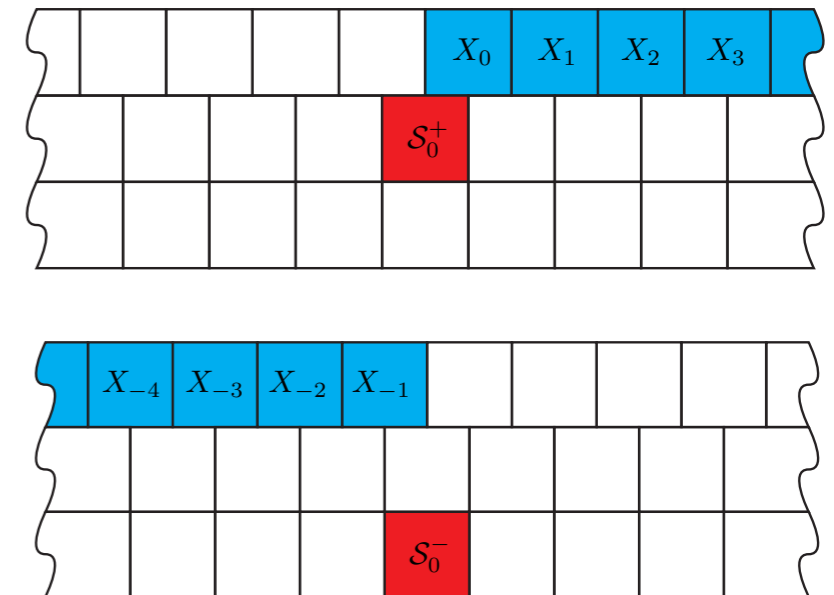
$$\text{Proposition: } \Xi \neq 0 \Rightarrow \chi^+ \neq \chi^-$$

$$\text{Proof sketch: } \chi^+ = C_{\mu}^+ - \mathbf{E}$$

$$\chi^- = C_{\mu}^- - \mathbf{E}$$

$$C_{\mu}^- \neq C_{\mu}^+$$

$$\Rightarrow \chi^+ \neq \chi^-$$

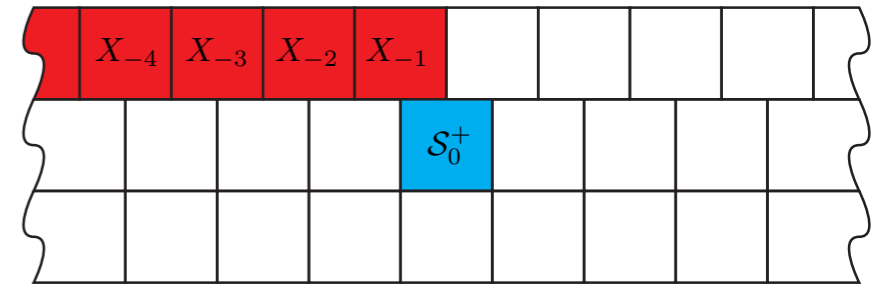


Causal Irreversibility

More corollaries:

Retrodiction with forward states

$$\begin{aligned} H[\overleftarrow{X}^L | \mathcal{S}^+] &= h_\mu L - \chi^+ \\ &= h_\mu L - C_\mu^+ + \mathbf{E} \\ &= \mathbf{E} + h_\mu L - C_\mu^+ \end{aligned}$$



$$H[\overleftarrow{X}^L | \mathcal{S}^+] > 0 \quad \text{RHS positive, only when: } L \geq \left\lceil \frac{\chi^+}{h_\mu} \right\rceil$$

~ cryptic order

In addition,

$$H[\overleftarrow{X}^L | \mathcal{S}^+] + C_\mu^+ = \mathbf{E} + h_\mu L$$

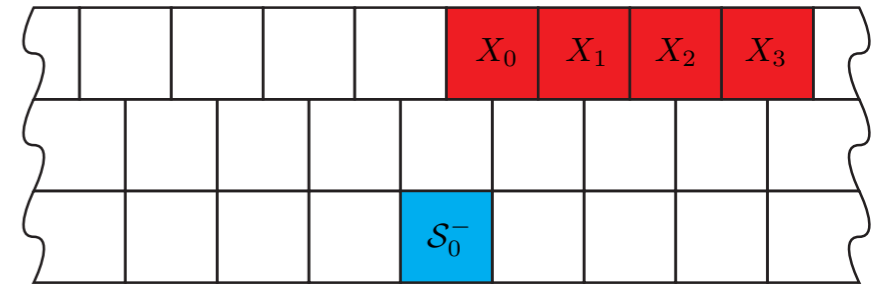
$$H[\overleftarrow{X}^L, \mathcal{S}^+] = \mathbf{E} + h_\mu L$$

Block-State Entropy

Causal Irreversibility

Similarly, prediction with reverse states

$$\begin{aligned}
 H[\vec{X}^L | \mathcal{S}^-] &= h_\mu L - \chi^- \\
 &= h_\mu L - C_\mu^- + \mathbf{E} \\
 &= \mathbf{E} + h_\mu L - C_\mu^-
 \end{aligned}$$



$$H[\vec{X}^L | \mathcal{S}^-] > 0 \quad \text{RHS positive, only when: } L \geq \left\lceil \frac{\chi^-}{h_\mu} \right\rceil$$

~ cryptic order

And also,

$$H[\vec{X}^L | \mathcal{S}^-] + C_\mu^- = \mathbf{E} + h_\mu L$$

$$H[\vec{X}^L, \mathcal{S}^-] = \mathbf{E} + h_\mu L$$

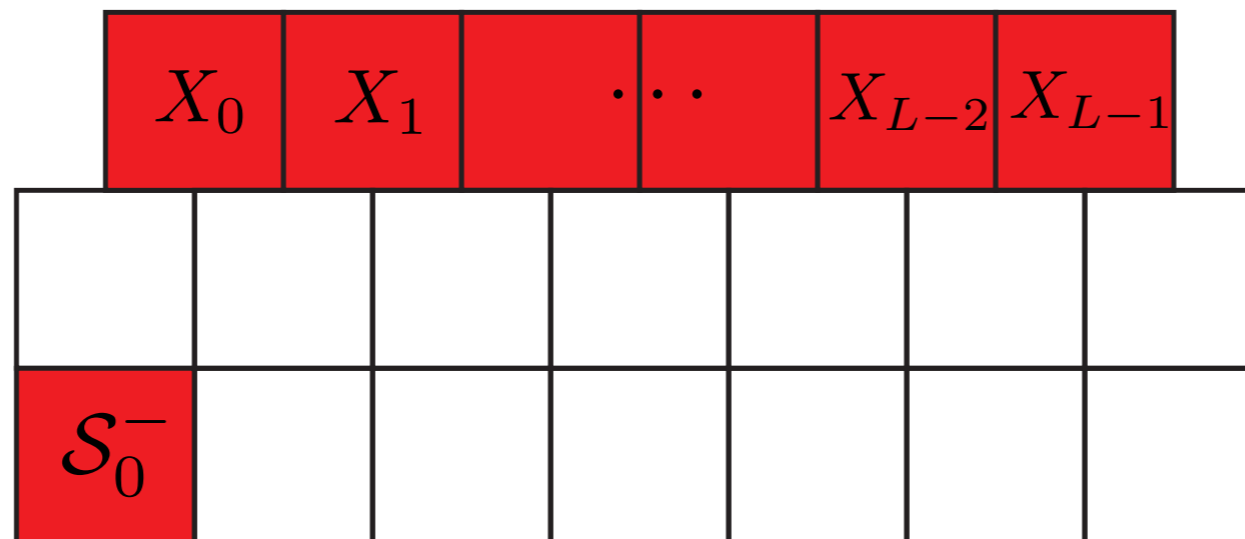
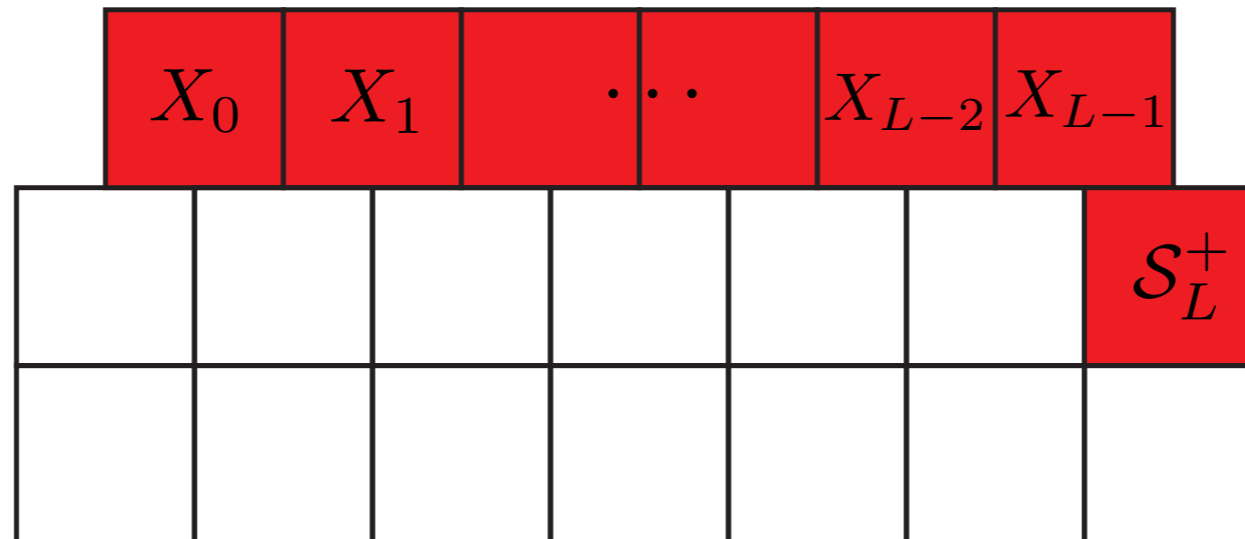
Block-State Entropy

Causal Irreversibility

So, in addition:

$$H[\vec{X}^L, \mathcal{S}^-] = H[\overleftarrow{X}^L, \mathcal{S}^+]$$

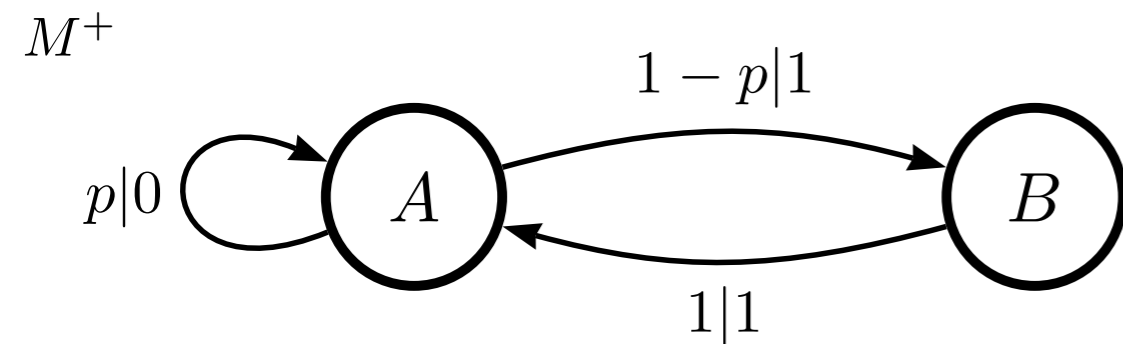
$$\vec{X}_0^L = \overleftarrow{X}_{L-1}^L$$



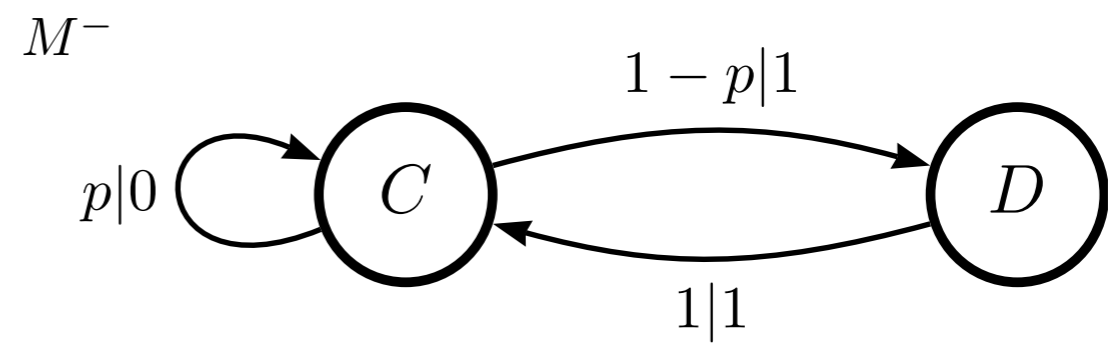
Causal Irreversibility

Example: Even Process

Forward ε -machine



Reverse ε -machine



Noncryptic: $\chi^+ = 0$

Noncryptic: $\chi^- = 0$

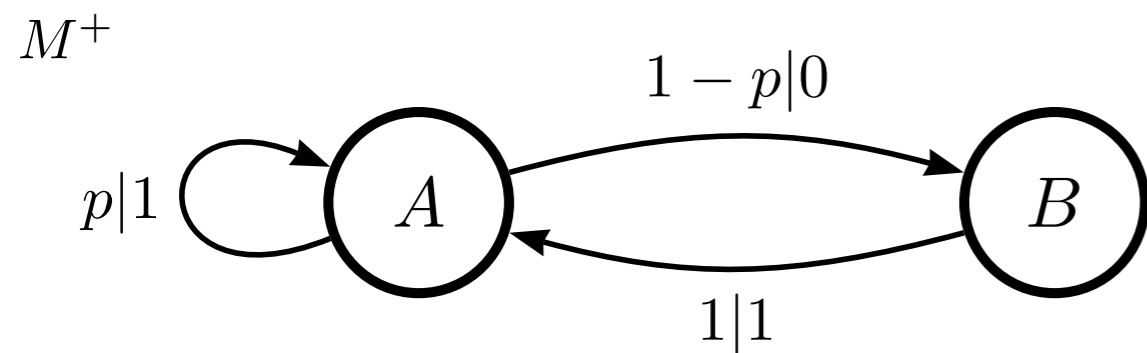
Microscopic reversible: $M^+ = M^- \Rightarrow \mathcal{P}^+ = \mathcal{P}^-$

Causally reversible: $\Xi = 0$

Causal Irreversibility

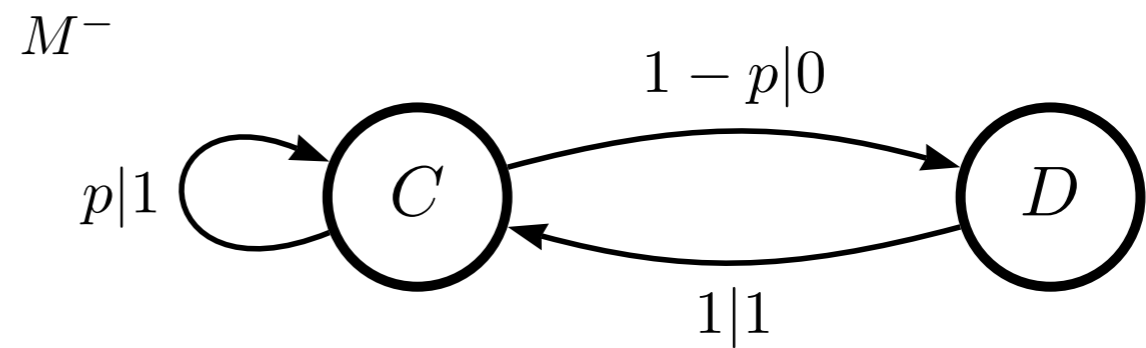
Example: Golden Mean Process

Forward ε -machine



Cryptic: $\chi^+ > 0$

Reverse ε -machine



Cryptic: $\chi^- > 0$

Microscopic reversible: $M^+ = M^- \Rightarrow \mathcal{P}^+ = \mathcal{P}^-$

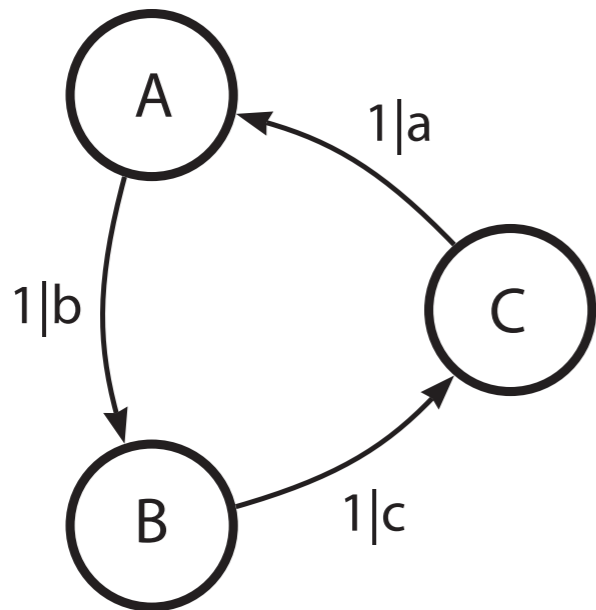
Causally reversible: $\Xi = 0$

So, crypticity not same as causal irreversibility.

Causal Irreversibility

Example: a-b-c Process

Forward ε -machine



Noncryptic: $\chi^+ = 0$

Microscopically irreversible:

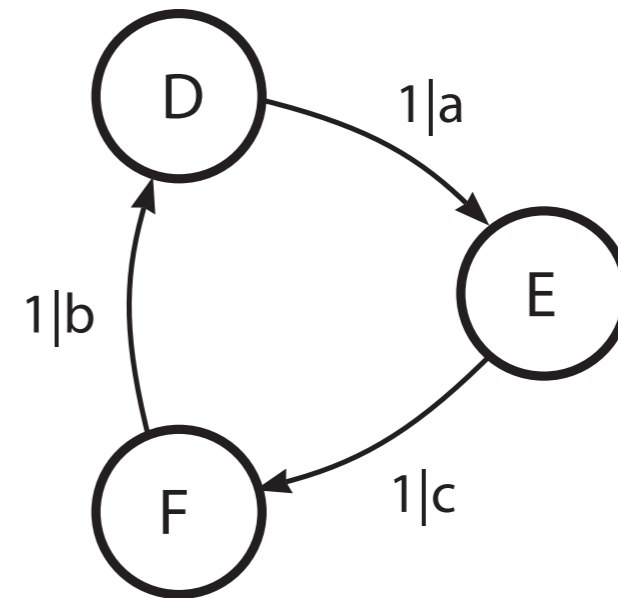
$$\Pr(ab|\mathcal{P}^+) = 1/3$$

$$\Pr(ab|\mathcal{P}^-) = 0$$

Causally reversible:

$$\Xi = 0$$

Reverse ε -machine

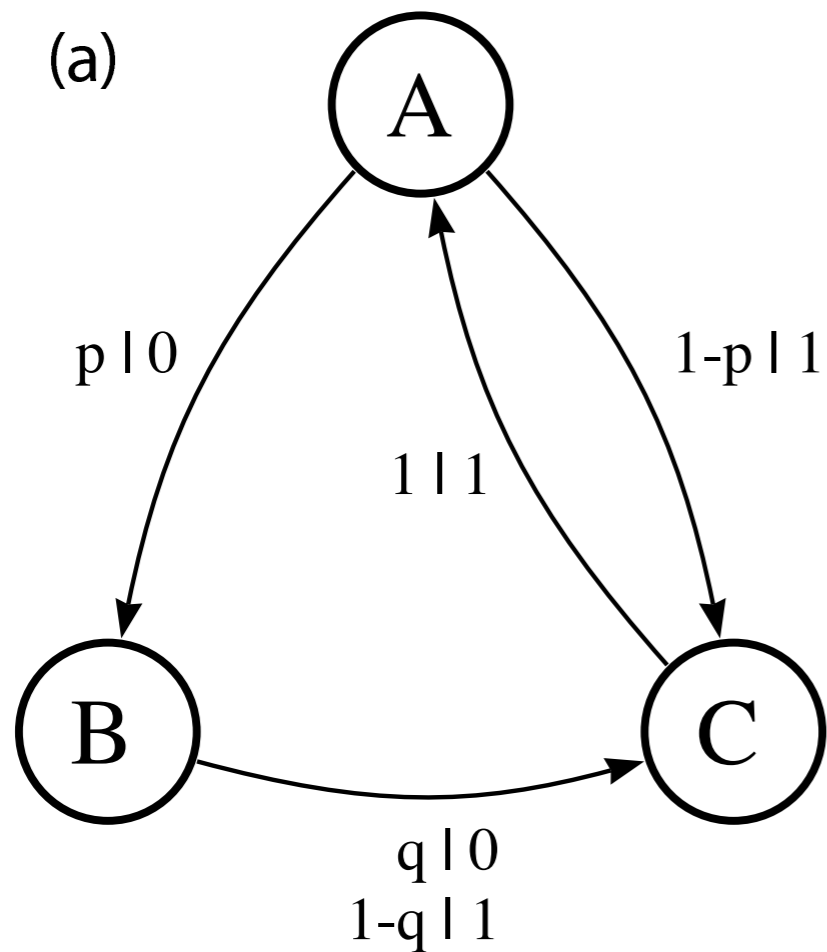


Noncryptic: $\chi^- = 0$

Causal Irreversibility

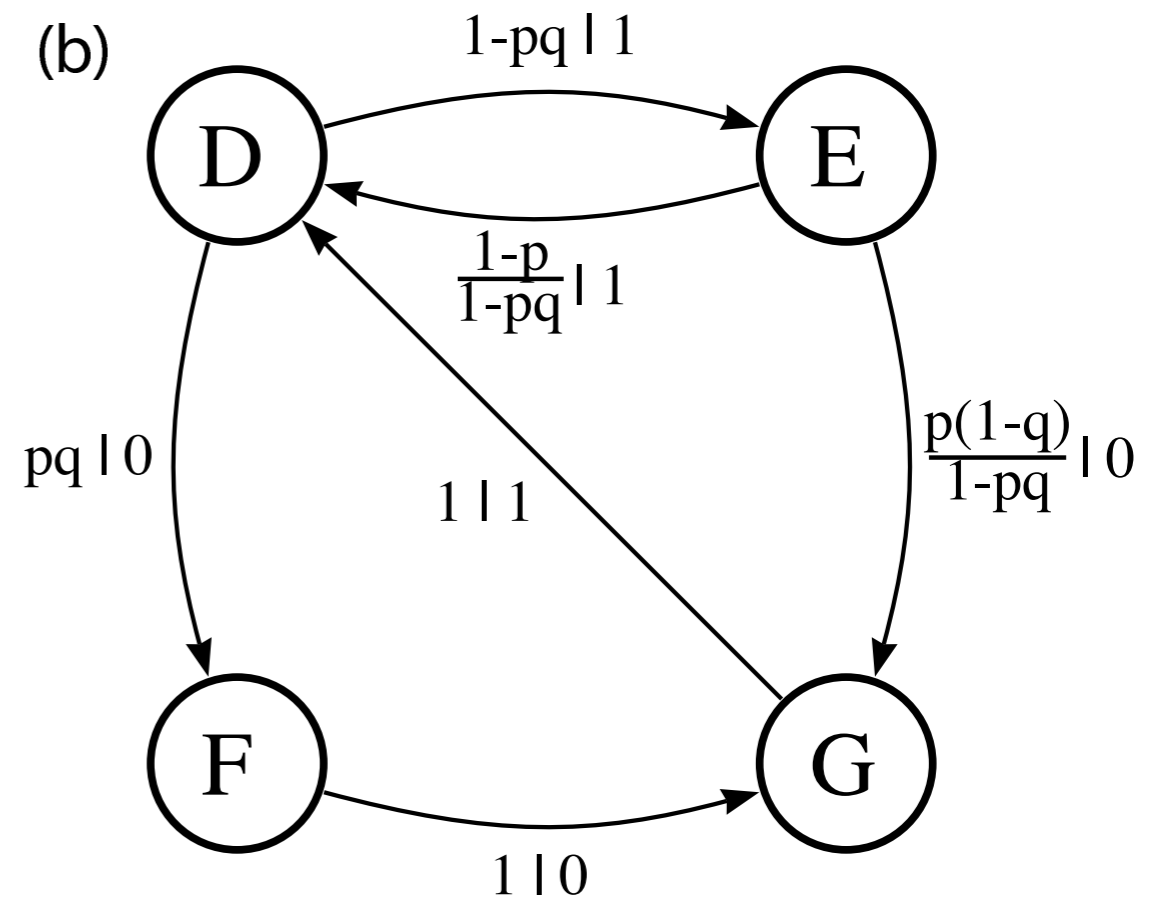
Example: Random Insertion Process

Forward ϵ -machine



Cryptic: $\chi^+ > 0$

Reverse ϵ -machine



Cryptic: $\chi^- > 0$

Causal Irreversibility

Example: Random Insertion Process ...

Forward ε -machine

$$\Pr(\mathcal{S}^+) = \begin{pmatrix} A & B & C \\ \frac{1}{p+2} & \frac{p}{p+2} & \frac{1}{p+2} \end{pmatrix}$$

Reverse ε -machine

$$\Pr(\mathcal{S}^-) = \begin{pmatrix} D & E & F & G \\ \frac{1}{p+2} & \frac{1-pq}{p+2} & \frac{pq}{p+2} & \frac{p}{p+2} \end{pmatrix}$$

Causal Irreversibility

Example: Random Insertion Process ...

Set $p = q = 1/2$

$$h_{\mu} = \frac{3}{5} \text{ bits/symbol}$$

$$C_{\mu}^{+} \approx 1.5219 \text{ bits}$$

$$C_{\mu}^{-} \approx 1.8464 \text{ bits}$$

Causally irreversible:

$$\Xi \approx -0.3245 \text{ bits}$$

Microscopically irreversible:

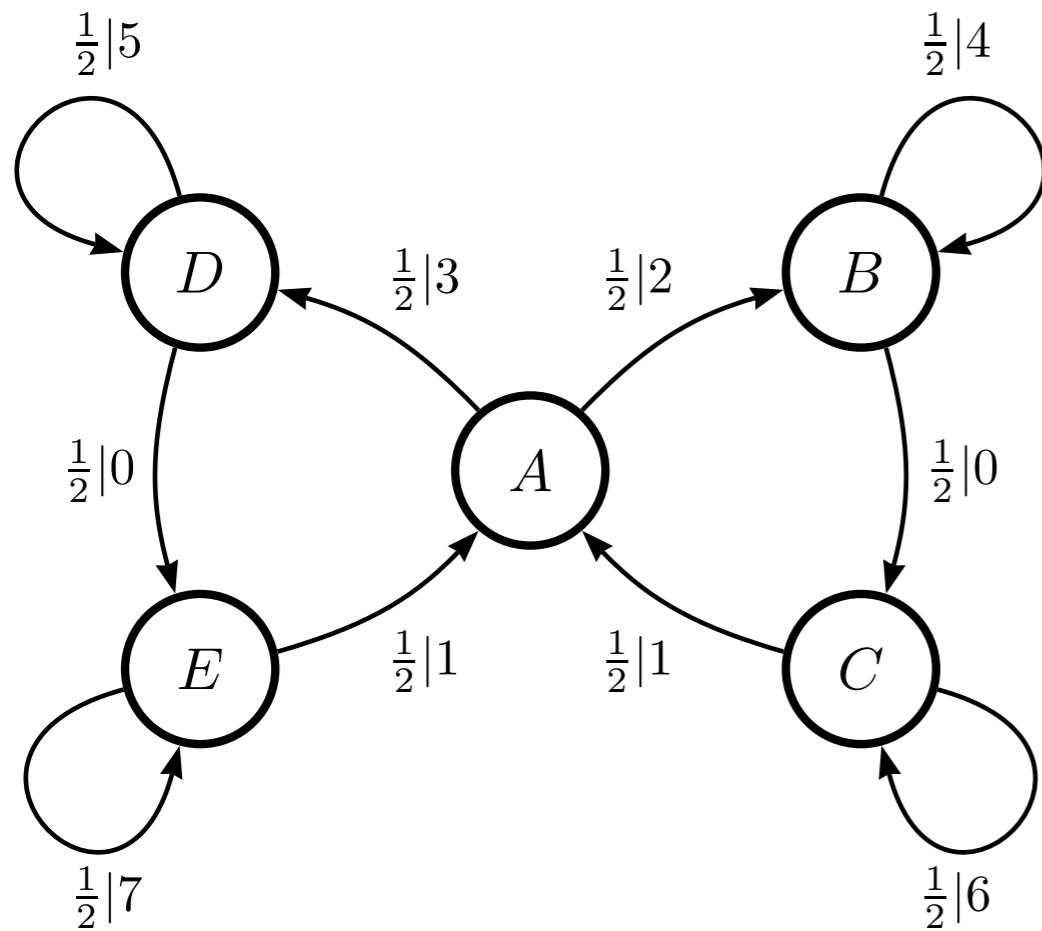
$$M^{+} \neq M^{-} \Rightarrow \mathcal{P}^{+} \neq \mathcal{P}^{-}$$

Causal Irreversibility

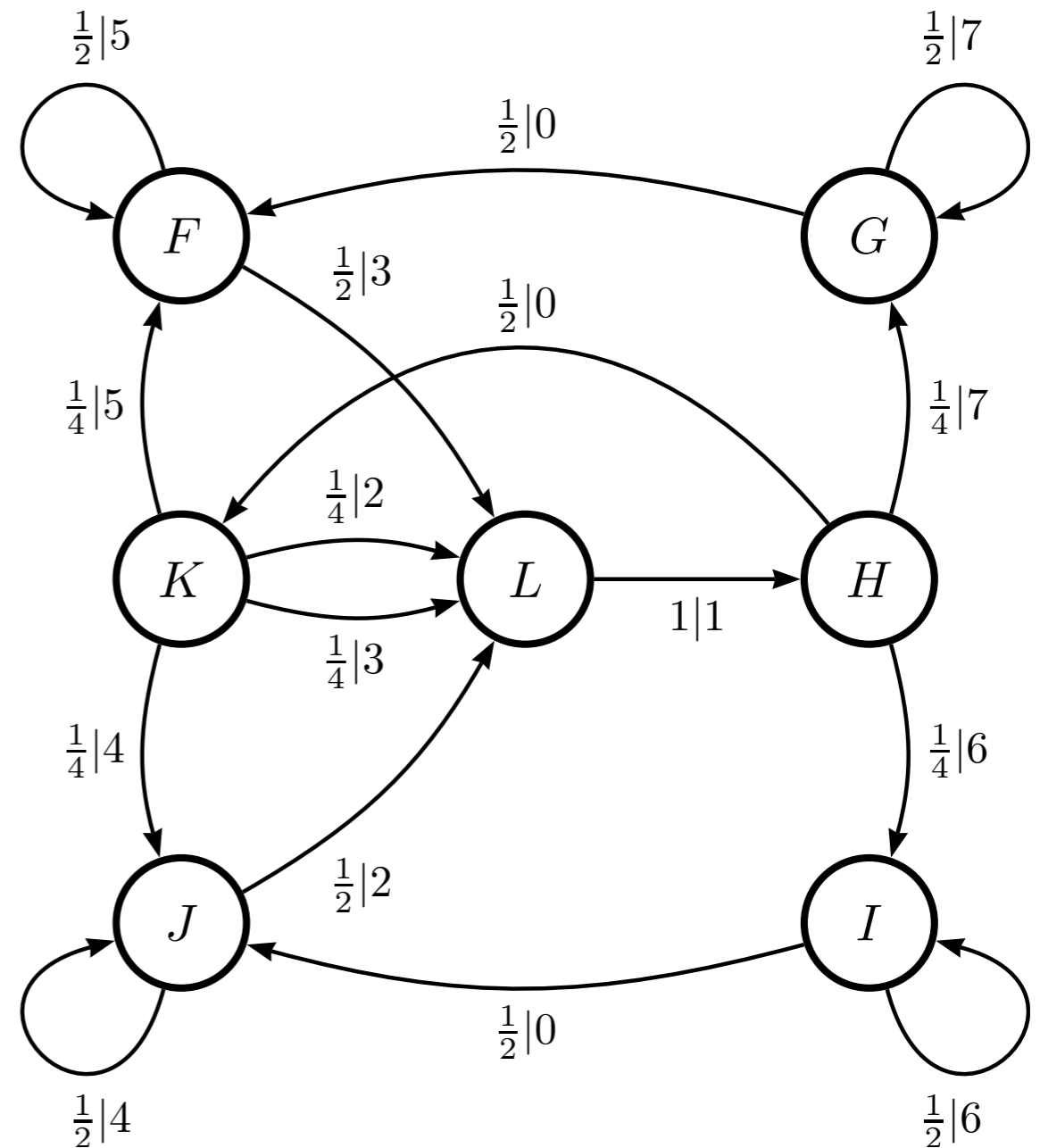
Example: Butterfly Process

$$\mathcal{A} = \{0, 1, \dots, 6\}$$

Forward ε -machine



Reverse ε -machine



Causal Irreversibility

Example: Butterfly Process

Microscopically irreversible: $M^+ \neq M^-$

$$\Pr(\mathcal{S}^+) = \left(\frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5}\right) \quad C_\mu^+ = \log_2 5 \text{ bits}$$

$$\Pr(\mathcal{S}^-) = (0.1, 0.2, 0.2, 0.15, 0.15, 0.1, 0.1) \\ C_\mu^- = 2.7464 \text{ bits}$$

Causally irreversible: $\Xi = -0.4245$ bits

2-Cryptic: $\chi^+ = \frac{3}{10}$ bits

$$\mathbf{E} = C_\mu^+ - \chi^+ = 2.0219 \text{ bits}$$

$$\chi^- = C_\mu^- - \mathbf{E} = 0.7245 \text{ bits}$$

Causal Irreversibility

Examples Summary

Process	Microscopically	Causally	Cryptic
Even	Rev	Rev	No
Golden Mean	Rev	Rev	Yes
a-b-c	Irrev	Rev	No
RIP	Irrev	Irrev	Yes
Butterfly	Irrev	Irrev	Yes

Directional ε -Machines

Directional ϵ -Machines

Two questions:

How to calculate reverse ϵ -machine given forward?

How to calculate forward and reverse state dependencies?

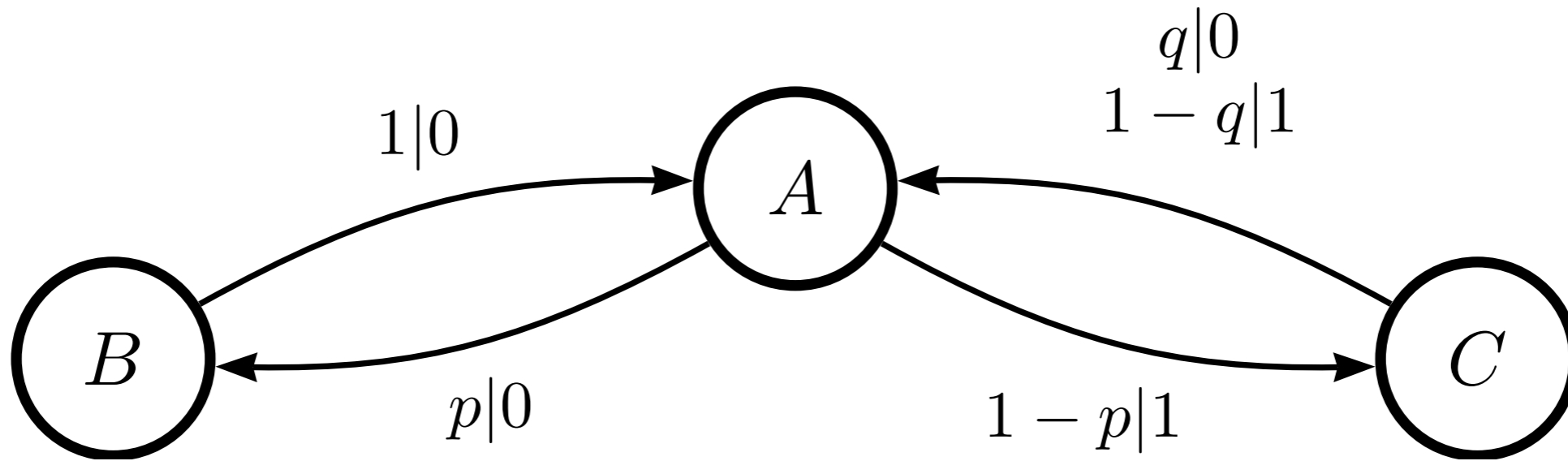
Answer:

1. Reverse time: transition arrow directions reversed.
2. Calculate mixed-state presentation of reversed machine.
3. Minimize to get ϵ -machine of reverse process.

Directional ϵ -Machines

Procedure by example:

Random Noisy Copy Process (RnC)



Directional ε -Machines

Procedure by example:

Random Noisy Copy Process (RnC)

$$\mathcal{S}^+ = \{A, B, C\}$$

$$T^{(0)} = \begin{array}{c} A \\ B \\ C \end{array} \begin{array}{ccc} A & B & C \\ \left(\begin{array}{ccc} 0 & p & 0 \\ 1 & 0 & 0 \\ q & 0 & 0 \end{array} \right) \end{array} \quad T^{(1)} = \begin{array}{c} A \\ B \\ C \end{array} \begin{array}{ccc} A & B & C \\ \left(\begin{array}{ccc} 0 & 0 & 1-p \\ 0 & 0 & 0 \\ 1-q & 0 & 0 \end{array} \right) \end{array}$$

$$\pi = \text{Pr}(\mathcal{S}^+) = \frac{1}{2} \begin{array}{ccc} A & B & C \\ \left(\begin{array}{ccc} 1 & p & 1-p \end{array} \right) \end{array}$$

Directional ε -Machines

Time-reversed presentation: $\widetilde{M}^+ = \mathcal{T}(M^+)$

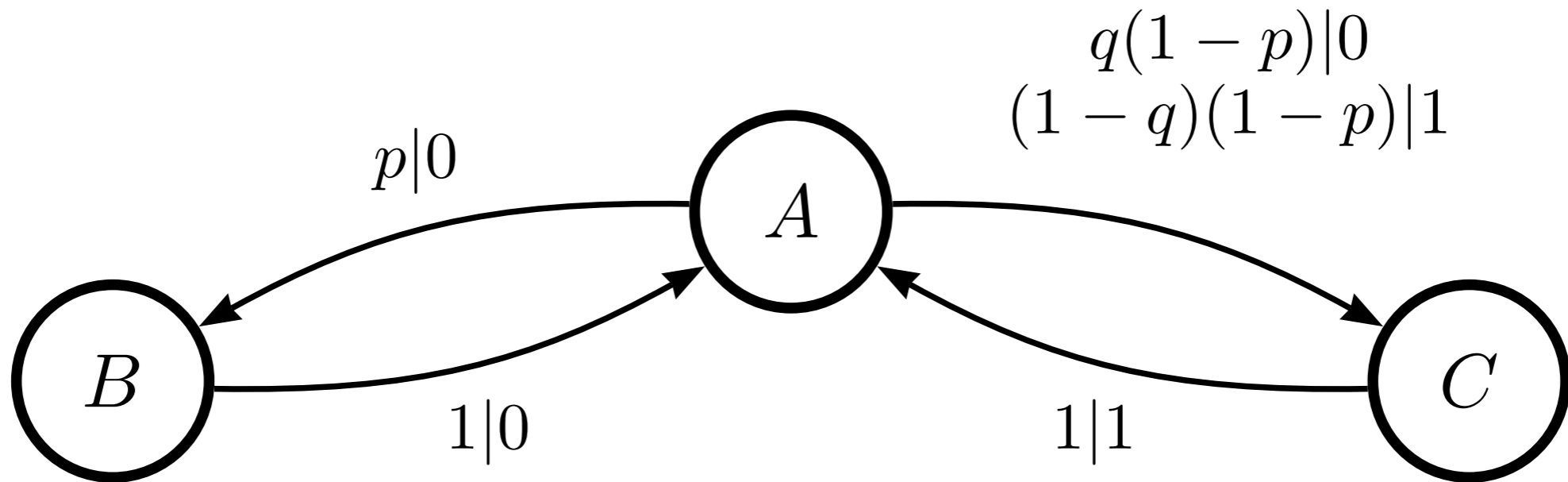
1. Reverse transitions.
2. Normalize to produce stochastic matrices:

$$\begin{aligned}\widetilde{T}_{\mathcal{R}'\mathcal{R}}^{(x)} &\equiv \Pr(X = x, \mathcal{R} | \mathcal{R}') \\ &= T_{\mathcal{R}\mathcal{R}'}^{(x)} \frac{\Pr(\mathcal{R})}{\Pr(\mathcal{R}')}\end{aligned}$$

$$\widetilde{T} = \sum_{\{x\}} \widetilde{T}^{(x)}$$

Directional ε -Machines

Time-reversed presentation: $\widetilde{M}^+ = \mathcal{T}(M^+)$



$$\widetilde{T}^{(0)} = \begin{matrix} & A & B & C \\ \begin{matrix} A \\ B \\ C \end{matrix} & \begin{pmatrix} 0 & p & q(1-p) \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \end{matrix} \quad \widetilde{T}^{(1)} = \begin{matrix} & A & B & C \\ \begin{matrix} A \\ B \\ C \end{matrix} & \begin{pmatrix} 0 & 0 & (1-q)(1-p) \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix} \end{matrix}$$

Stationary distribution of \widetilde{M}^+ : $\widetilde{\pi} = \pi$

Nonunifilar!

Directional ε -Machines

Mixed-state presentation: $M^- = \mathcal{U}(\widetilde{M}^+)$

$$\{\mathcal{S}^-\} \sim \{\nu(w) = \Pr(\mathcal{S}^+ | \vec{w}) : \vec{w} \in \mathcal{A}^*\}$$

Will generate transient and recurrent states.
Ignore transient states.

Start with $w = \lambda$

$$\nu(\lambda) = \Pr(\mathcal{S}^+) = \frac{1}{2} (1 \quad p \quad 1 - p)$$

Append symbols 0 and 1, calculate induced states:

$$\nu(0) \quad \nu(1)$$

Directional ε -Machines

Mixed-state presentation: $M^- = \mathcal{U}(\widetilde{M}^+)$

$$\begin{aligned}\nu(0) &= \Pr(\mathcal{S}_0^+ | X_0 = 0) \\ &= \frac{\widetilde{\pi} \widetilde{T}^{(0)}}{\widetilde{\pi} \widetilde{T}^{(0)} \mathbf{1}} \\ &= \frac{(p, p, q(1-p))}{2p + q(1-p)}\end{aligned}$$

$$\begin{aligned}\nu(1) &= \Pr(\mathcal{S}_0^+ | X_0 = 1) \\ &= \frac{\widetilde{\pi} \widetilde{T}^{(1)}}{\widetilde{\pi} \widetilde{T}^{(1)} \mathbf{1}} \\ &= \frac{(1, 0, 1-q)}{2-q}\end{aligned}$$

Directional ε -Machines

Mixed-state presentation: $M^- = \mathcal{U}(\tilde{M}^+)$

Extend words with 0 and 1;

calculate induced states that follow $\nu(0)$ and $\nu(1)$:

$$\nu(00) = \Pr(\mathcal{S}_0^+ | X_0^2 = 00) = \frac{\tilde{\pi} \tilde{T}^{(0)} \tilde{T}^{(0)}}{\tilde{\pi} \tilde{T}^{(0)} \tilde{T}^{(0)} \mathbf{1}}$$

$$\nu(01) = \Pr(\mathcal{S}_0^+ | X_0^2 = 01) = \frac{\tilde{\pi} \tilde{T}^{(1)} \tilde{T}^{(0)}}{\tilde{\pi} \tilde{T}^{(1)} \tilde{T}^{(0)} \mathbf{1}}$$

$$\nu(10) = \Pr(\mathcal{S}_0^+ | X_0^2 = 10) = \frac{\tilde{\pi} \tilde{T}^{(0)} \tilde{T}^{(1)}}{\tilde{\pi} \tilde{T}^{(0)} \tilde{T}^{(1)} \mathbf{1}}$$

$$\nu(11) = \Pr(\mathcal{S}_0^+ | X_0^2 = 11) = \frac{\tilde{\pi} \tilde{T}^{(1)} \tilde{T}^{(1)}}{\tilde{\pi} \tilde{T}^{(1)} \tilde{T}^{(1)} \mathbf{1}}$$

Directional ε -Machines

Mixed-state presentation: $M^- = \mathcal{U}(\widetilde{M}^+)$

Continue, until no new mixed states are produced.

For example,

$$\nu(1001) = \nu(111001)$$

The result is the set of mixed states for the reverse process.

$$\mathcal{S}^- = \{D, E, F\}$$

$$D \equiv \nu(1001) = \begin{matrix} & A & B & C \\ \begin{matrix} 0 & 0 & 1 \end{matrix} \end{matrix}$$

$$E \equiv \nu(100) = \begin{matrix} \begin{matrix} 1 & 0 & 0 \end{matrix} \end{matrix}$$

$$F \equiv \nu(10) = \begin{pmatrix} 0 & \frac{p}{p+q(1-p)} & \frac{q(1-p)}{p+q(1-p)} \end{pmatrix}$$

Directional ε -Machines

Mixed-state presentation: $M^- = \mathcal{U}(\widetilde{M}^+)$

Transition probabilities:

Consider $\nu(00) \rightarrow \nu(100)$

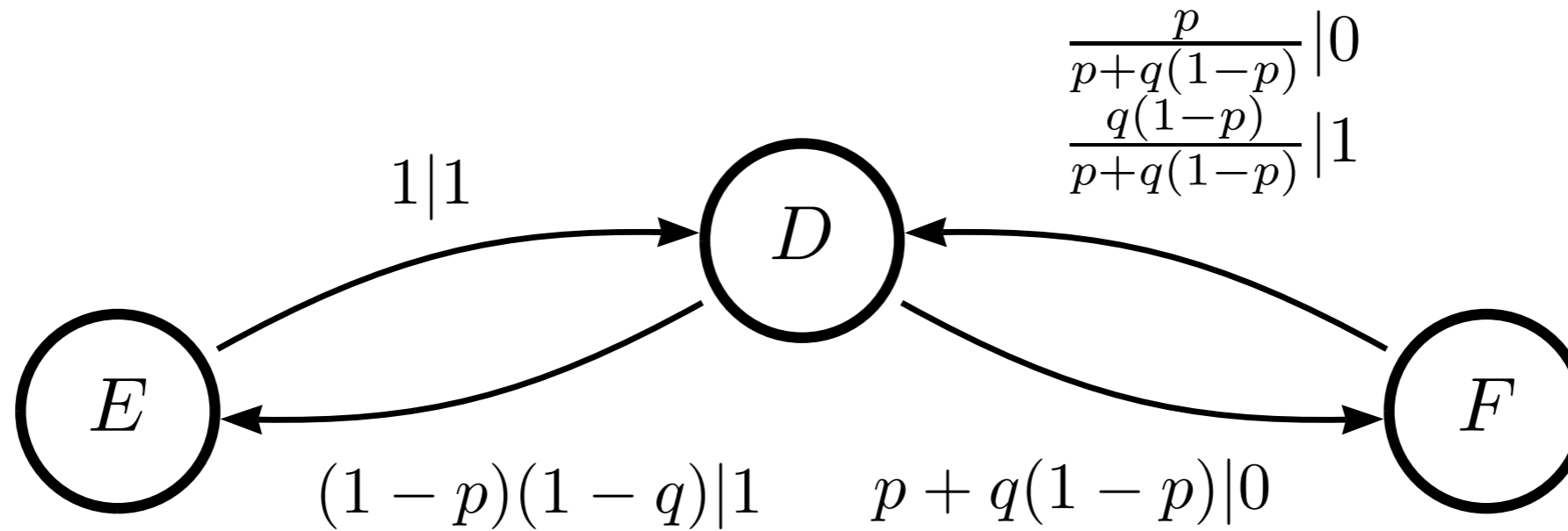
$$\begin{aligned}\Pr(1, \nu(100) | \nu(00)) &= \Pr(1 | 00) \\ &= \nu(00) \widetilde{T}^{(1)} \mathbf{1} \\ &= \frac{1 - p}{1 + p + q - pq}\end{aligned}$$

Do this for all possible state-to-state transitions.

Directional ε -Machines

Mixed-state presentation: $M^- = \mathcal{U}(\widetilde{M}^+)$

The result:

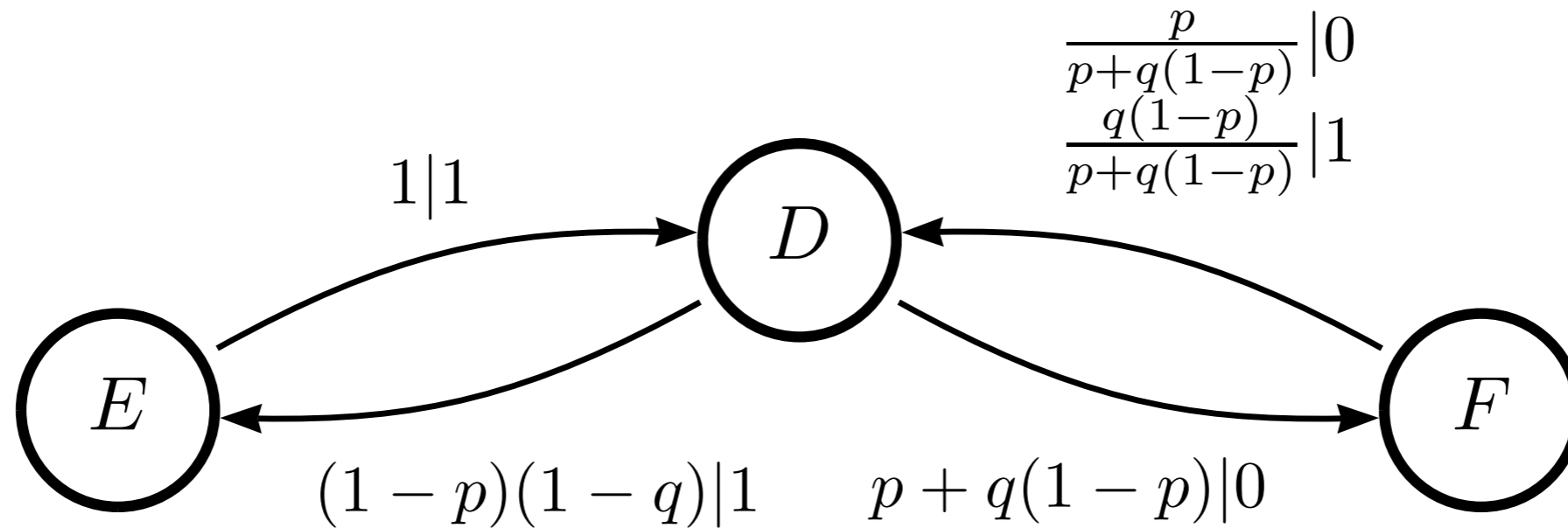


Note:

This is unifilar.

Directional ϵ -Machines

Minimize mixed-state presentation to get ϵ -machine:



For RnC, the MSP turned out to be minimal.

So, this is the ϵ M of the reverse process: $M^- = M(\mathcal{P}^-)$

Directional ε -Machines

Minimizing MSP to get ε -machine:

Recall MSP gives a unifilar presentation

Do recurrent portion of MSP

Recurrent states are a refinement of causal states.

So minimizing simply groups these.

Gives recurrent causal states of ε -machine.

Trick for general case: Strip transients and MSP again!

Minimization conjecture: Reverse \Rightarrow MSP \Rightarrow Reverse \Rightarrow MSP.

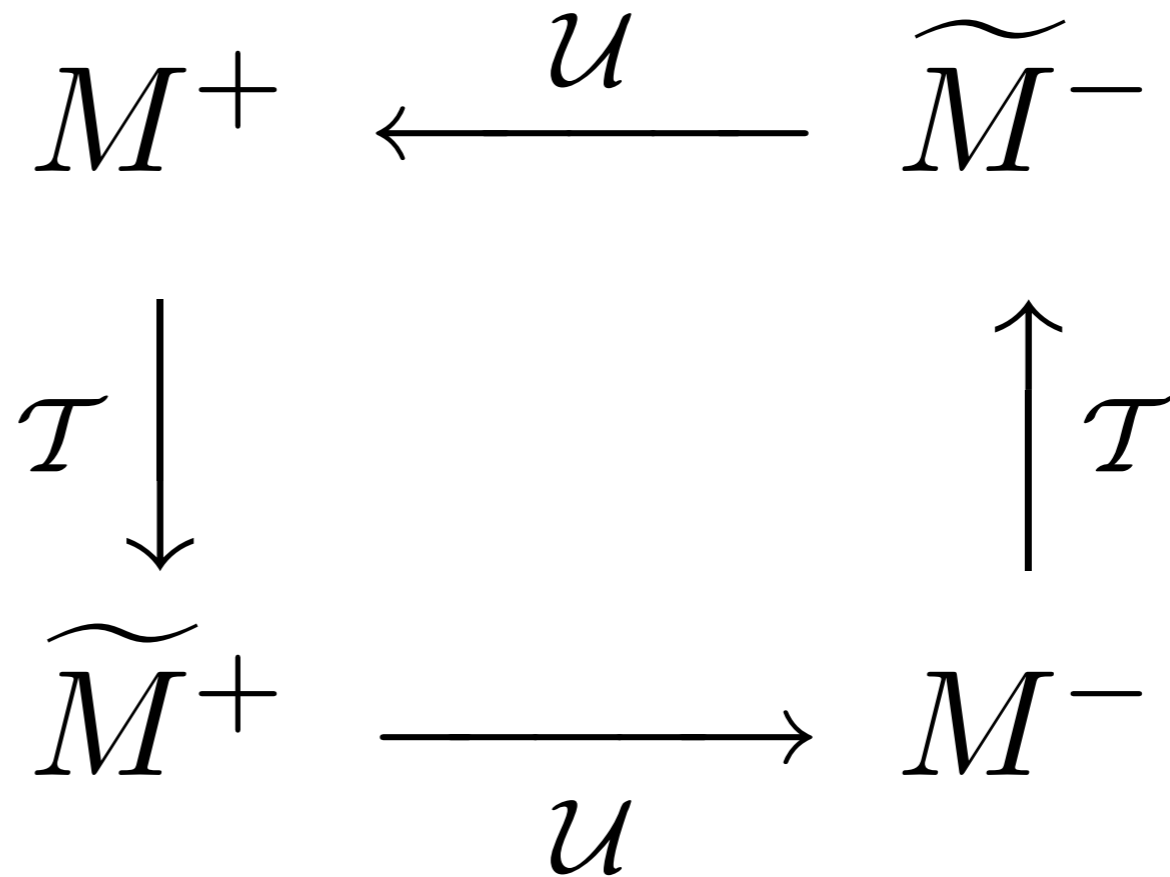
Directional ε -Machines

How to calculate $\Pr(\mathcal{S}^-, \mathcal{S}^+)$?

Track states through

Time-reversed
presentation

$$\widetilde{M}^+ = \mathcal{T}(M^+)$$



Mixed-state
presentation $M^- = \mathcal{U}(\widetilde{M}^+)$

Directional ε -Machines

How to calculate $\Pr(\mathcal{S}^-, \mathcal{S}^+)$?

Consider only the recurrent causal states: $\Pr(\mathcal{S}^-) > 0$

$$\mathcal{S}^- = \{D, E, F\}$$

$$D = \nu(1001) = \begin{array}{c} A \quad B \quad C \\ \left(\begin{array}{ccc} 0 & 0 & 1 \end{array} \right) = \Pr(\mathcal{S}^+ | \mathcal{S}^- = D)$$

$$E = \nu(100) = \begin{array}{c} A \quad B \quad C \\ \left(\begin{array}{ccc} 1 & 0 & 0 \end{array} \right) = \Pr(\mathcal{S}^+ | \mathcal{S}^- = E)$$

$$F = \nu(10) = \begin{array}{c} A \quad B \quad C \\ \left(\begin{array}{ccc} 0 & \frac{p}{p+q(1-p)} & \frac{q(1-p)}{p+q(1-p)} \end{array} \right) \\ = \Pr(\mathcal{S}^+ | \mathcal{S}^- = F)$$

Directional ε -Machines

How to calculate $\Pr(\mathcal{S}^-, \mathcal{S}^+)$?

These MSP states give

$$\Pr(\mathcal{S}^+ | \mathcal{S}^-)$$

With $\Pr(\mathcal{S}^-)$

get

$$\Pr(\mathcal{S}^+, \mathcal{S}^-) = \Pr(\mathcal{S}^+ | \mathcal{S}^-) \Pr(\mathcal{S}^-)$$

Directional ε -Machines

Reading for next lecture: CMR articles

TBA

PRATISP

MSP

IACP, IACPLCOCS