

Aditya Mahajan
Yale

Acknowledgements: Demos Teneketzis (UMich), Ashutosh Nayyar (UMich), Sekhar Tatikonda (Yale), Ramji Venkataramanan (Stanford)

April 13, 2010, UC Berkeley

Stochastic control \Rightarrow Information theory

Converses in info theory

- ▶ Capacity of FSM channels
Goldsmith-Varaiya-96
- ▶ Capacity of FSM channels with f/b
Vishwanathan-99, Chen-Berger-05,
Yang-Kavcic-Tatikonda-05,
Tatikonda-Mitter-09,
Kavcic-Mandic-Huang-Ma-09
- ▶ Error exponents of FSM channels
with f/b
Como-Yüksel-Tatikonda-09
- ▶ FS-MAC with partial state info
Como-Yüksel-09



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Coding schemes

- ▶ Optimal coding schemes for channels
with feedback
Hornstein-63, Schalkwijk-Kailath-66,
Shayevitz-Feder-09, Coleman-09,
Bae-Anastasopoulos-10, Gorantla-Coleman-10



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Real-time communication

- ▶ Point-to-point communication
Witsenhausen-79, Walrand-Varaiya-82,
Teneketzis-06, Mahajan-Teneketzis-08,09
- ▶ Multi-terminal communication
Nayyar-Teneketzis-09, Mahajan-09



Stochastic control \Rightarrow Information theory

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Goldsmith

- ▶ Capacity

Vishwan

Yang-Kav

Tatikond

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Como-Yü

- ▶ FS-MAC

Como-Yü

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Kailath-66,

an-09,

rantla-Coleman-10

This talk ...

- ▶ Combination of all three

cation

tion

l-Varaiya-82,

neketzis-08,09

ation

najan-09



This talk . . .

Setup

- ▶ Discrete memoryless multiple access channel (MAC) with feedback
- ▶ Inner bounds on capacity region (achievable schemes) using block Markov superposition coding



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Setup

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Approach

- ▶ Setup block Markov superposition codes as decentralized control problem with delayed sharing of information
- ▶ A systematic study of auxiliary random variables



This talk . . .

Setup

- ▶ Discrete memoryless multiple access channel (MAC) with feedback
- ▶ Inner bounds on capacity region (achievable schemes) using block Markov superposition coding

Approach

- ▶ Setup block Markov superposition codes as decentralized control problem with delayed sharing of information
- ▶ A systematic study of auxiliary random variables

Main point

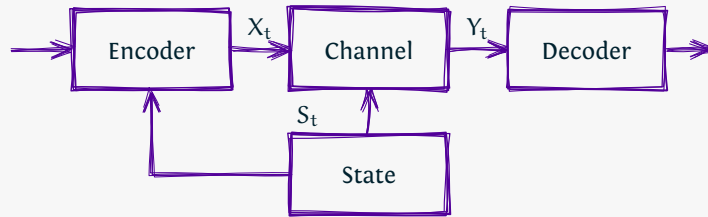
- ▶ Block Markov superposition coding schemes give rise to a specific information structure
- ▶ Stochastic control can provide insights into the design of coding schemes for a specific information structures



Are
auxiliary random variables
in info theory related to
information states in
stochastic control?

What are auxiliary random variables?

Channel coding with
causal side information

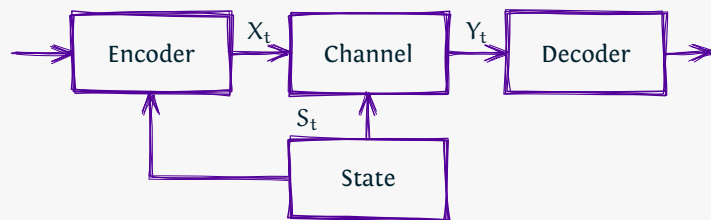


[Shannon-58]



What are auxiliary random variables?

Channel coding with
causal side information



[Shannon-58]

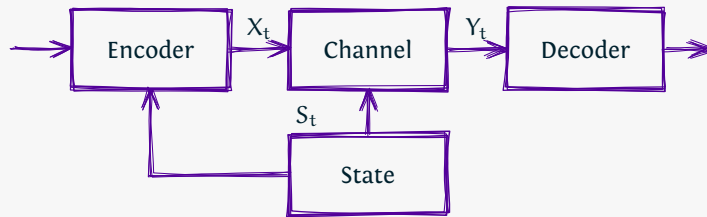
Code functions

$$C = \max_{\substack{P_F \\ F: \mathcal{S} \rightarrow \mathcal{X}}} I(F \wedge Y)$$



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[Shannon-58]

Code functions

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Auxiliary random variables

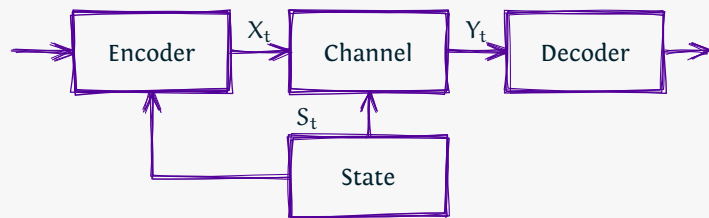
$$C = \max_{P_{\mathcal{U}}, f: \mathcal{U} \times \mathcal{S} \rightarrow \mathcal{X}} I(\mathcal{U} \wedge Y)$$

where $|\mathcal{U}| \leq |\mathcal{Y}|$



What are auxiliary random variables?

Channel coding with causal side information



[Shannon-58]

Code functions

$$C = \max_{\substack{P_F \\ F: \mathcal{S} \rightarrow \mathcal{X}}} I(F \wedge Y)$$

Auxiliary random variables

We can think of the auxiliary random variable indexing the set of all functions from \mathcal{S} to \mathcal{X} .

$$C = \max_{P_{U,f}: \mathcal{U} \times \mathcal{S} \rightarrow \mathcal{X}} I(U \wedge Y)$$

where $|\mathcal{U}| \leq |\mathcal{Y}|$

[Keshet, Steinberg, Merhav, 2007]



What are information states?

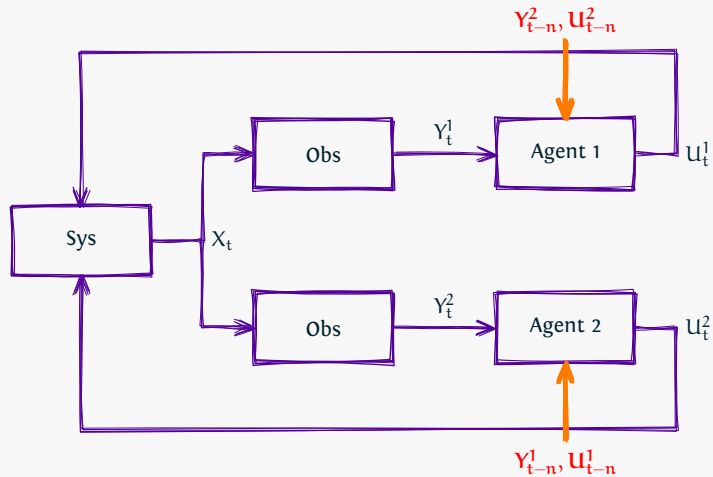
The (information) state should be
a **summary** ('compression') **of some data** (the 'past')
known to someone (an observer or a controller)
and **sufficient for some purposes**
(input-output map, optimization, dynamic programming).

[Witsenhausen-1976]



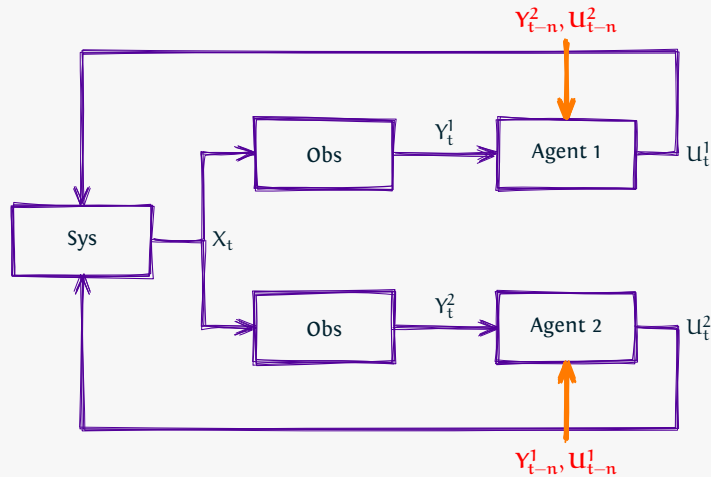
Understanding the relation

Delayed sharing
information structure



Understanding the relation

Delayed sharing information structure



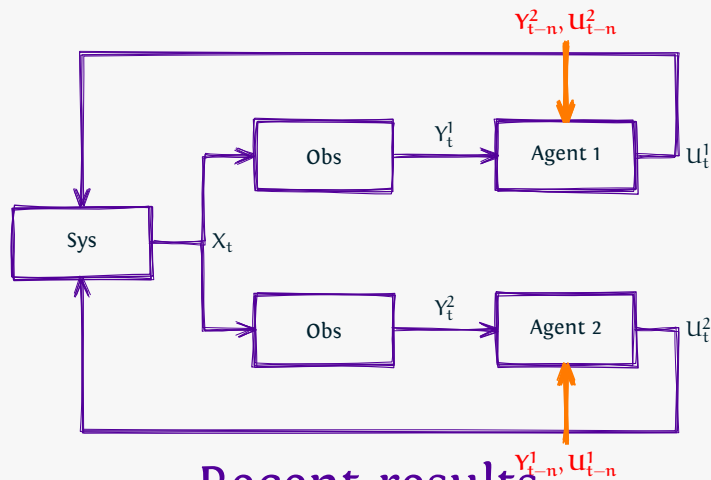
History

- ▶ A bridge between classical ($n = 0$) and non-classical ($n = \infty$) info structures
- ▶ **Witsenhausen, 1971** proposed the n -DSIS and **asserted** a structure of optimal control policies
- ▶ **Varaiya and Walrand, 1979** proved that Witsenhausen's assertion is **true** for $n = 1$ but **false** of $n > 1$



Understanding the relation

Delayed sharing information structure



Recent results

Information state: [Nayyar-Mahajan-Teneketzis-10]

$\Pr(X_{t-n} \mid \text{common info}) + n\text{-partially evaluated past control functions}$

History

- ▶ A bridge between classical ($n = 0$) and non-classical ($n = \infty$) info structures
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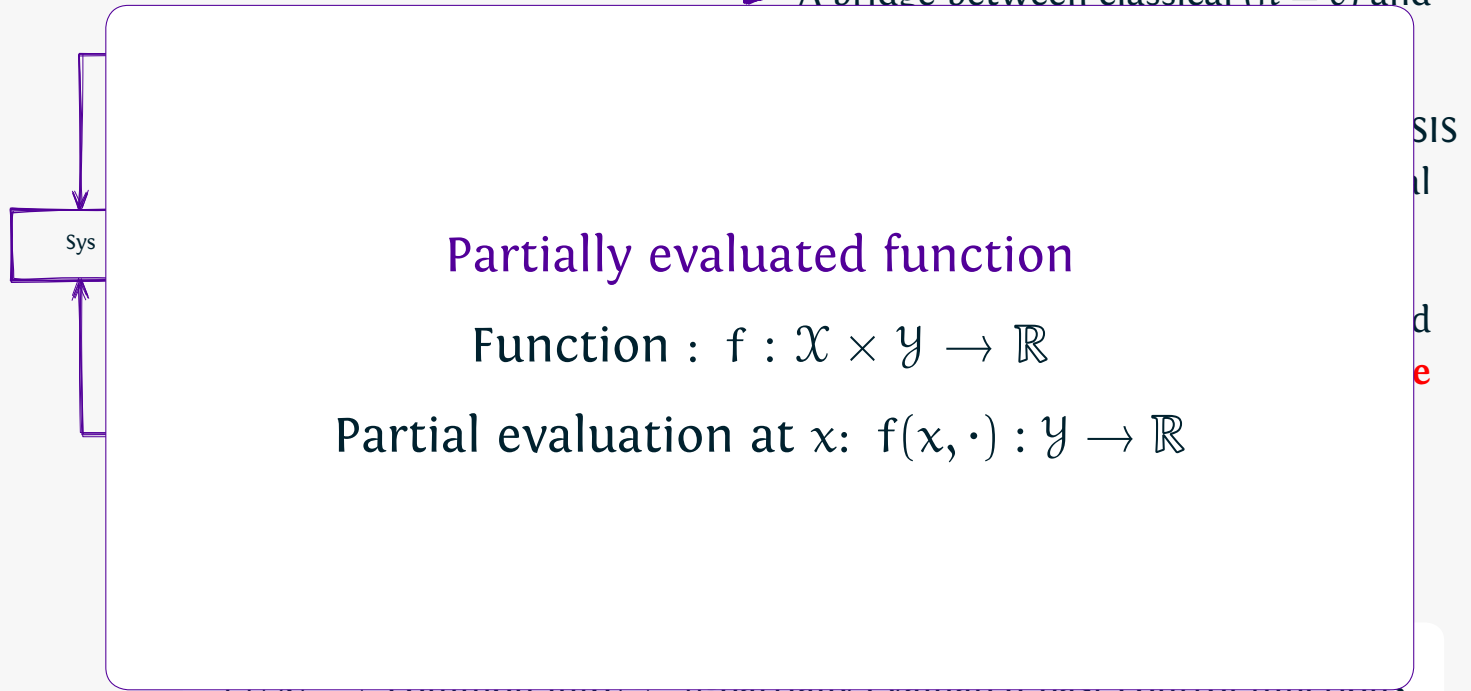


Understanding the relation

Delayed sharing
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History

► A bridge between classical ($n = 0$) and

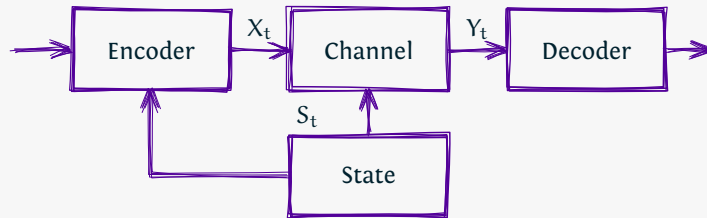


$\{f(x_{t-n} | \text{common info})\}$ is partially evaluated past control functions

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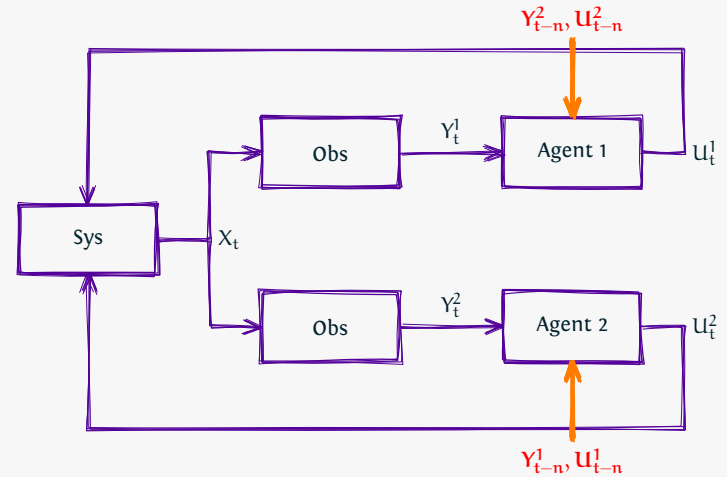
Understanding the relation

Channel coding with causal side information



Auxiliary random variable
is related to **functions**

Delayed sharing information structure



Information states are
related to **partial functions**

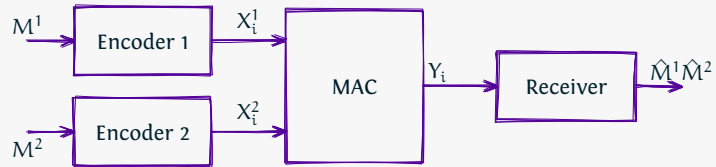


Outline

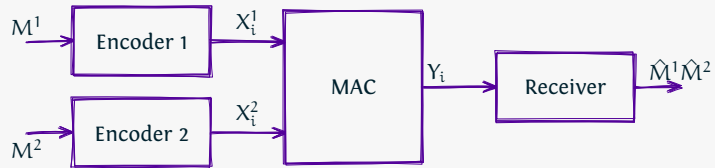
1. Overview of MAC with feedback
 - ▶ Cover-Leung scheme
 - ▶ Bross-Lapidoth scheme
 - ▶ Venkataramanan-Pradhan scheme
2. Formulation as a decentralized control problem with delayed sharing
 - ▶ delay = 1
 - ▶ delay = 2
3. Conclusion



Multiple access channel



Multiple access channel



▶ Encoder

$$X_{[1:n]}^1 = f^1(M^1), \quad X_{[1:n]}^2 = f^2(M^2)$$

▶ Channel

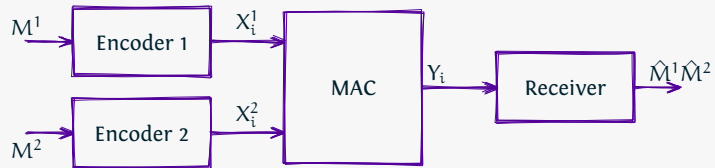
$$\mathbb{P}(Y_i | X_{[1:i]}^1 X_{[1:i]}^2) = \mathbb{P}(Y_i | X_i^1 X_i^2)$$

▶ Decoder: $\hat{M}^1 \hat{M}^2 = d(Y_{[1:n]})$

▶ Error: $\mathbb{P}(\hat{M}^1 \hat{M}^2 \neq M^1 M^2)$



Multiple access channel



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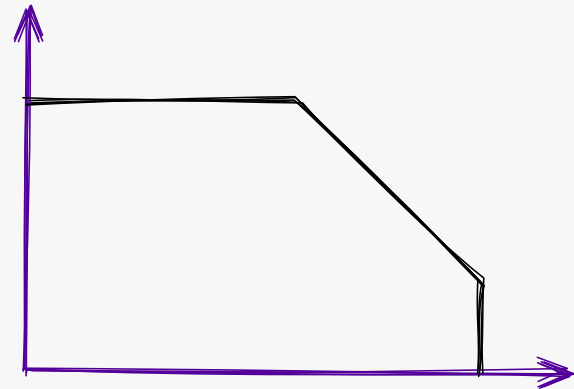
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Capacity

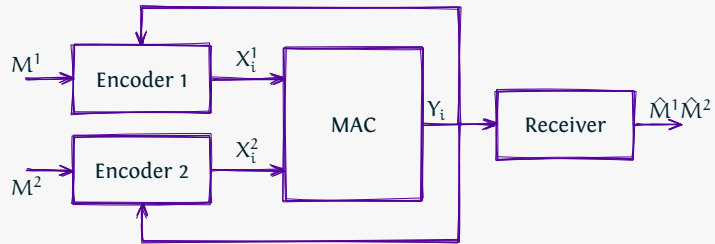
$$C = \bigcup_{P_{X^1} P_{X^2}} \left\{ (R_1, R_2) : \begin{array}{l} R_1 \leq I(X^1 \wedge Y | X^2) \\ R_2 \leq I(X^2 \wedge Y | X^1) \\ R_1 + R_2 \leq I(X^1 X^2 \wedge Y) \end{array} \right\}$$



Alshwede-71, Liao-72

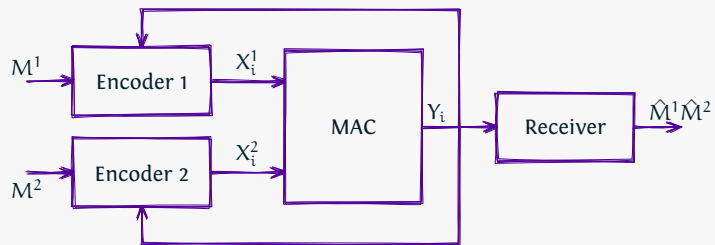


MAC with feedback



|||||

MAC with feedback



Encoder

$$X_i^1 = f^1(M^1, Y_{[1:i]}), \quad X_i^2 = f^2(M^2, Y_{[1:i]})$$

Channel

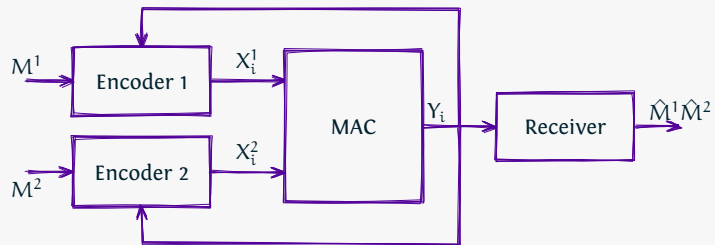
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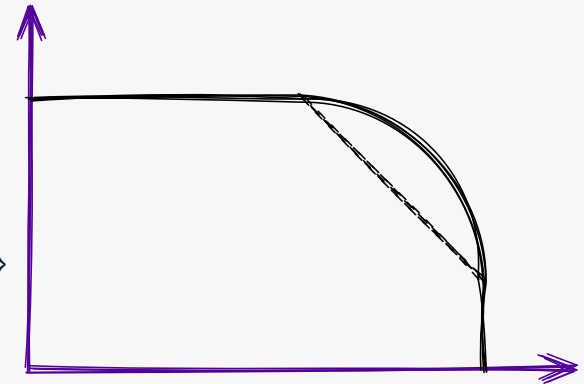
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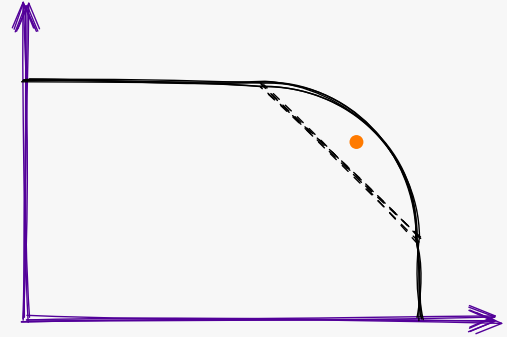
$$C = \bigcup_{\text{code trees}} \left\{ (R_1, R_2) : \begin{array}{l} R_1 \leq I_\infty(X^1 \rightarrow Y | X^2) \\ R_2 \leq I_\infty(X^2 \rightarrow Y | X^1) \\ R_1 + R_2 \leq I_\infty(X^1 X^2 \rightarrow Y) \end{array} \right\}$$



[Kramer-03]

Handwritten scribbles at the bottom of the slide.

Achievability scheme

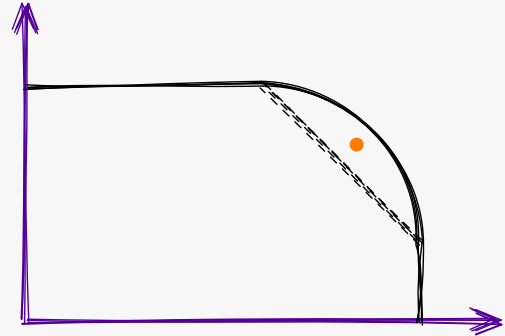


Achievability scheme

Main idea

Suppose we want to transmit at a rate point outside the no-feedback capacity region. Communicate in two phases

[Gaarder-Wolf-75]



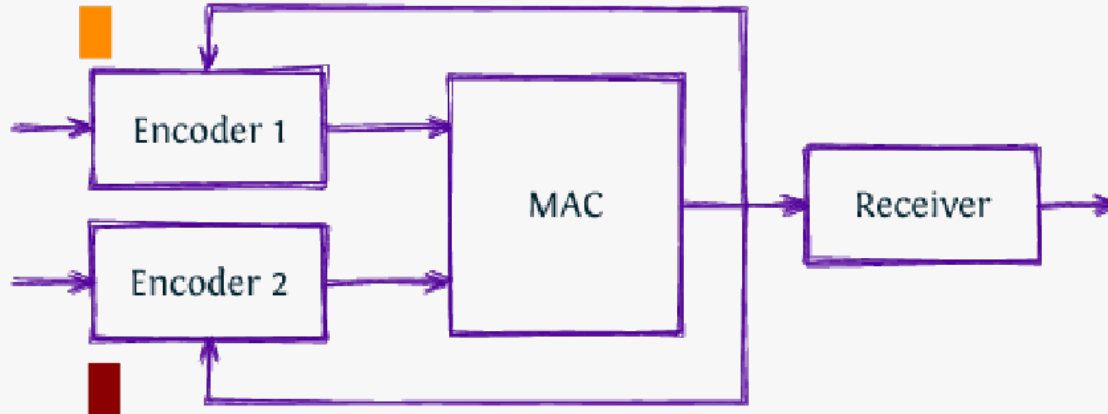
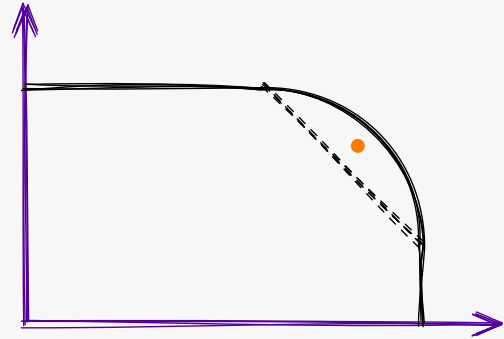
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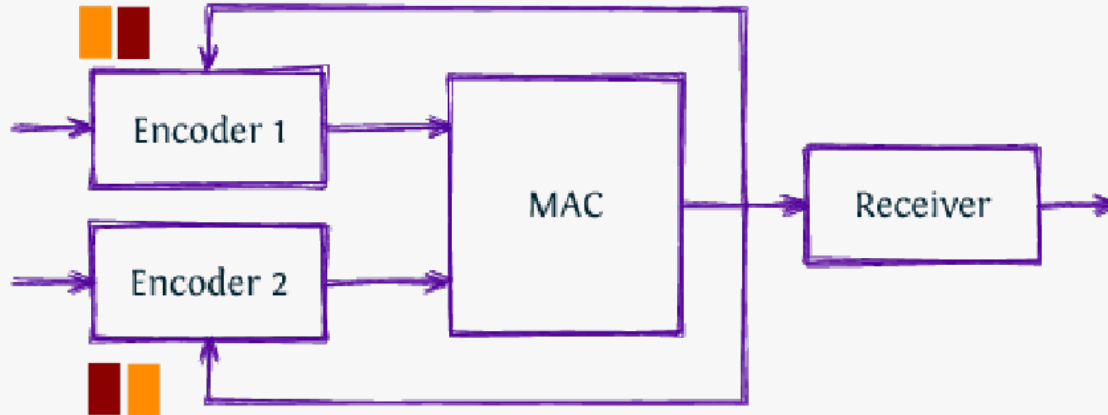
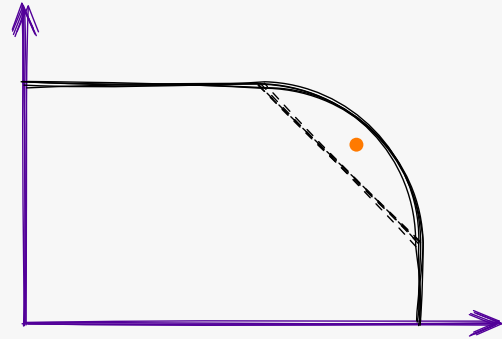
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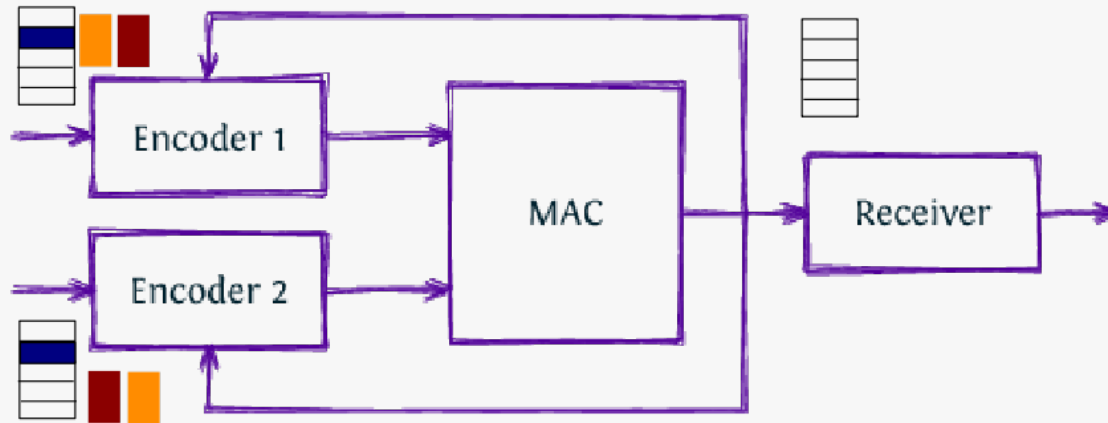
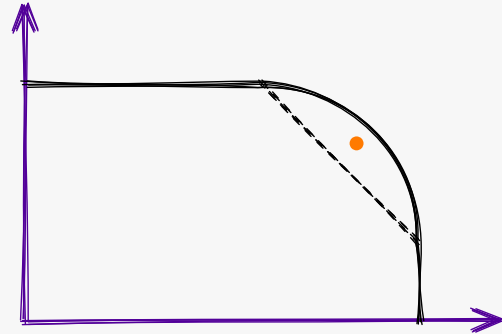
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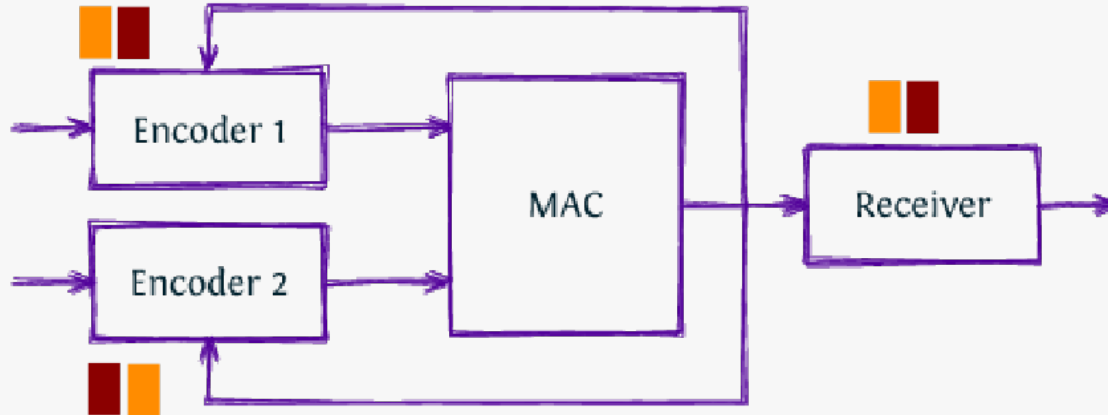
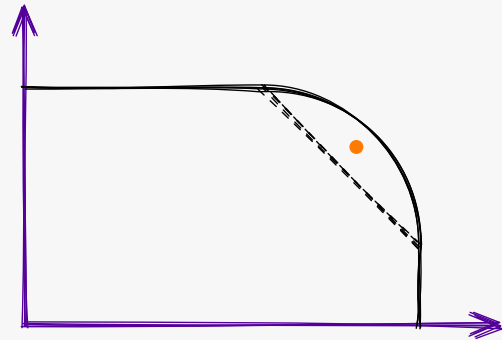
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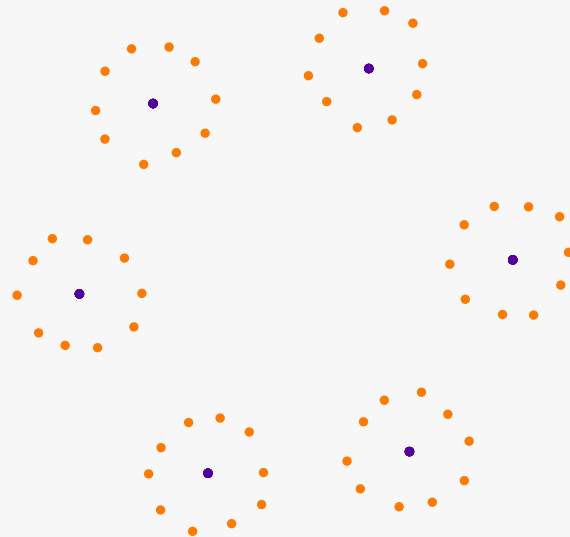
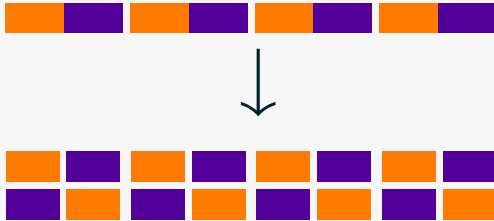
Block Markov superposition coding



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Block Markov superposition coding

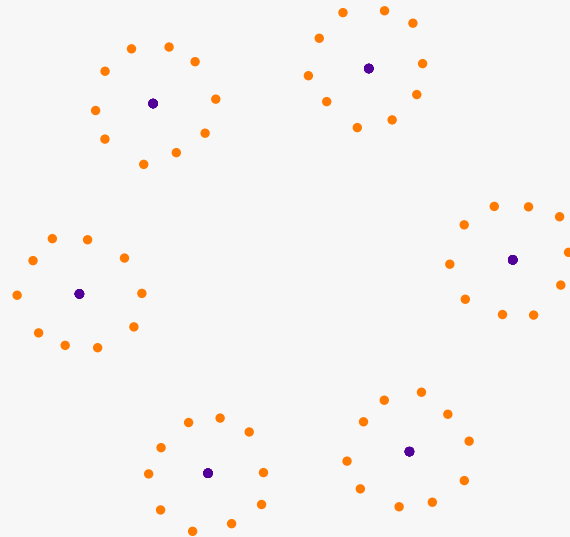
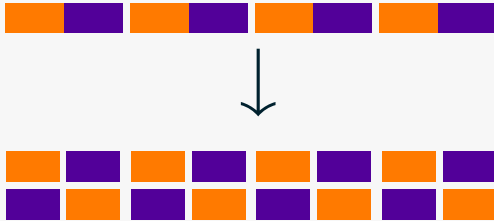
Superposition coding



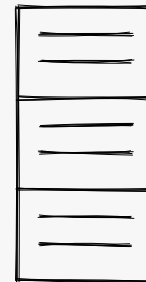
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Block Markov superposition coding

Superposition coding

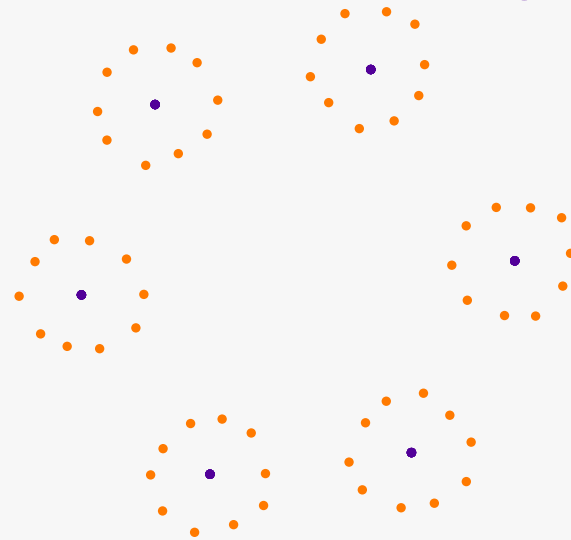
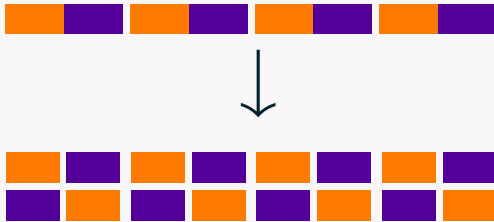


Binning



Block Markov superposition coding

Superposition coding



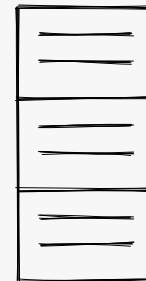
Cover-Leung region

$$R_{CL} = \bigcup_{P_U P_{X^1|U} P_{X^2|U}} \left\{ (R_1, R_2) : \begin{array}{l} R_1 \leq I(X^1 \wedge Y | U X^2) \\ R_2 \leq I(X^2 \wedge Y | U X^1) \\ R_1 + R_2 \leq I(X^1 X^2 \wedge Y) \end{array} \right\}$$

[Cover-Leung-81]

|||||

Binning



Converse

Cover-Leung region

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[Cover-Leung-81]



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[Cover-Leung-81]

CL region is tight for ...

- ▶ $H(X^1 | YX^2) = 0$ [Willems-82]
- ▶ Partial feedback [Willems-83]
- ▶ Binary erasure MAC [Willems-84]



Converse

Cover-Leung region

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[Cover-Leung-81]

CL region is tight for ...

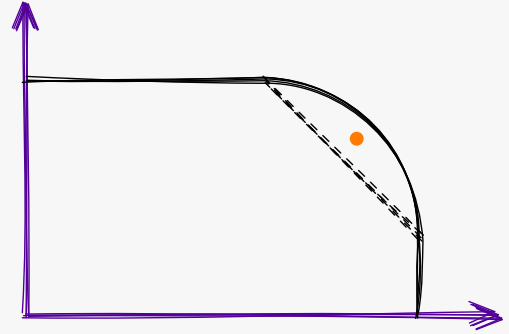
- ▶ $H(X^1|YX^2) = 0$ [Willems-82]
- ▶ Partial feedback [Willems-83]
- ▶ Binary erasure MAC [Willems-84]

CL region is **not** tight for ...

- ▶ Gaussian MAC [Ozarow-84]
- ▶ Poisson MAC [Bross-Lapidoth-05]
- ▶ Others ... [Kramer-02, Bross-Lapidoth-05, Venkataramanan-Pradhan-09]



Adding two way communication



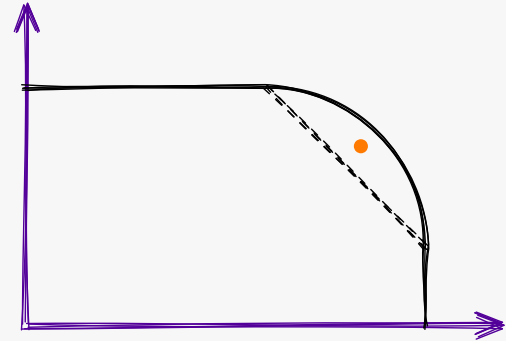
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Adding two way communication

Main idea

Each block now consists of three phases

[Bross-Lapidoth-05]



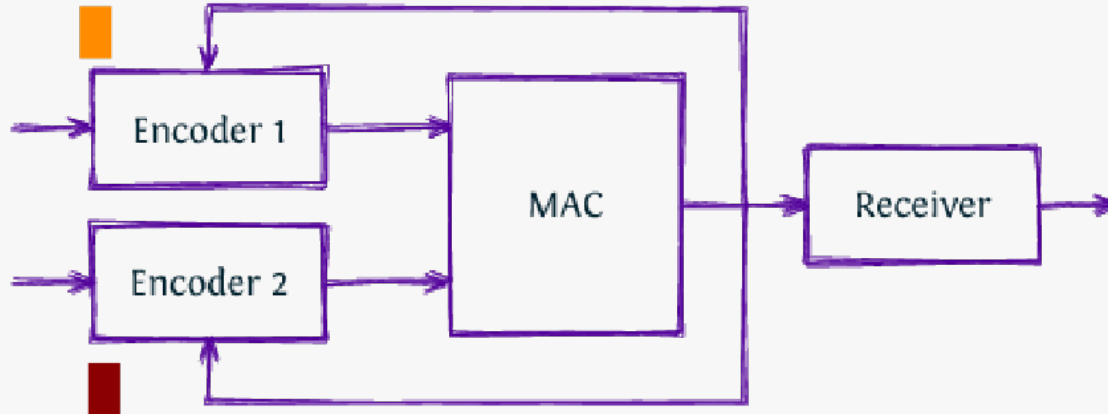
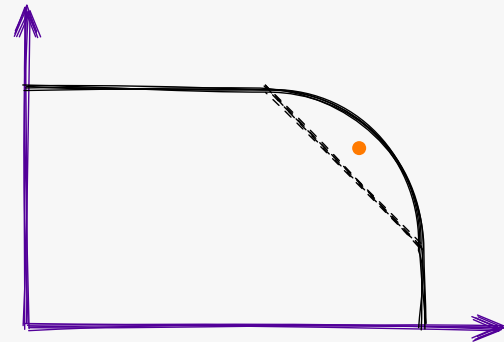
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[Bross-Lapidoth-05]



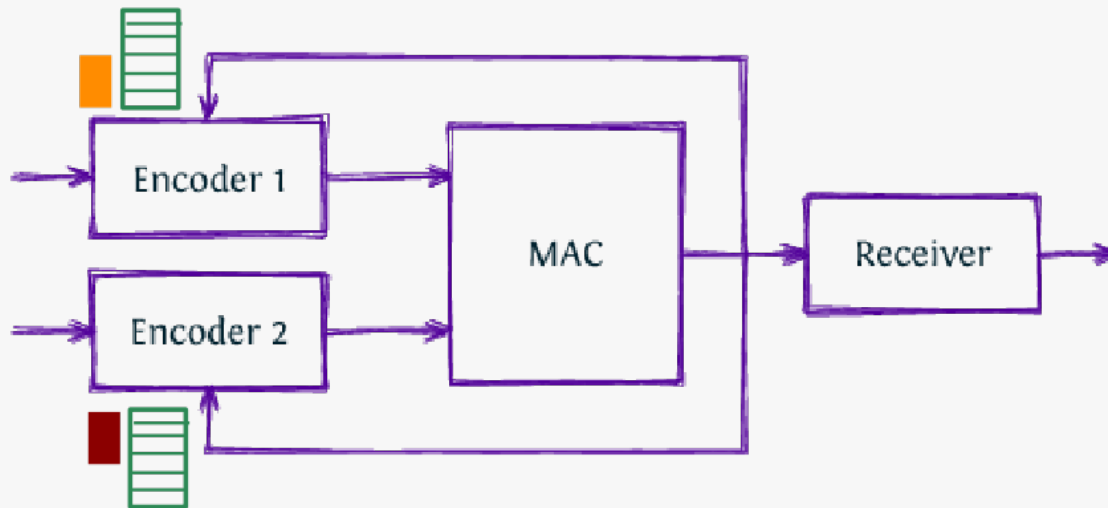
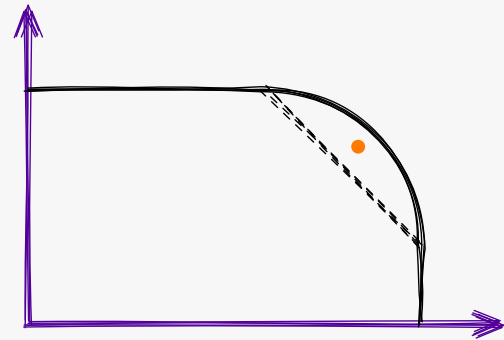
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Adding two way communication

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[Bross-Lapidoth-05]



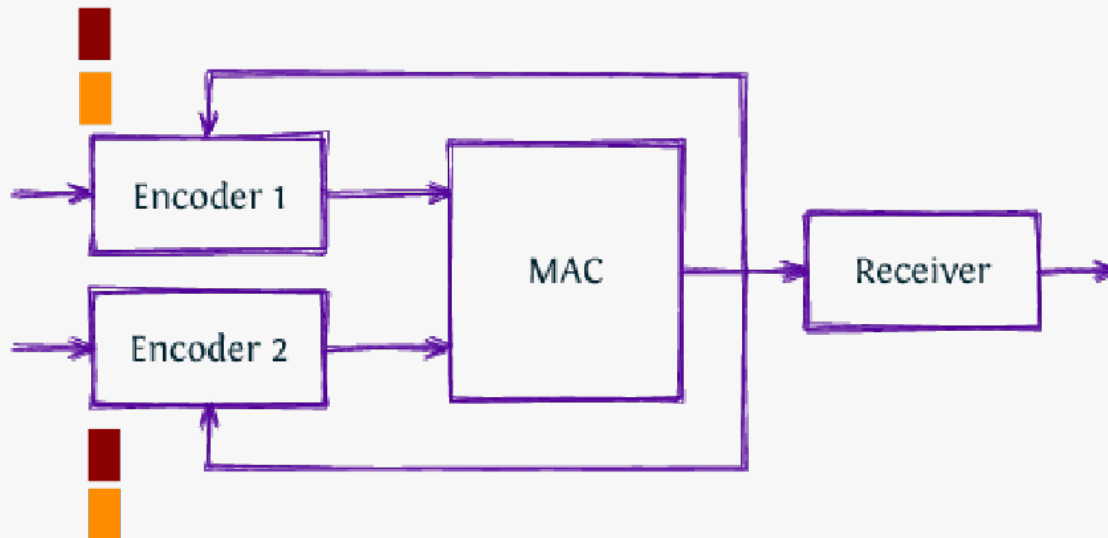
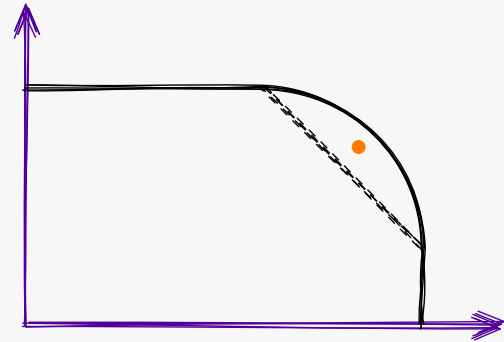
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Each block now consists of three phases

[Bross-Lapidoth-05]



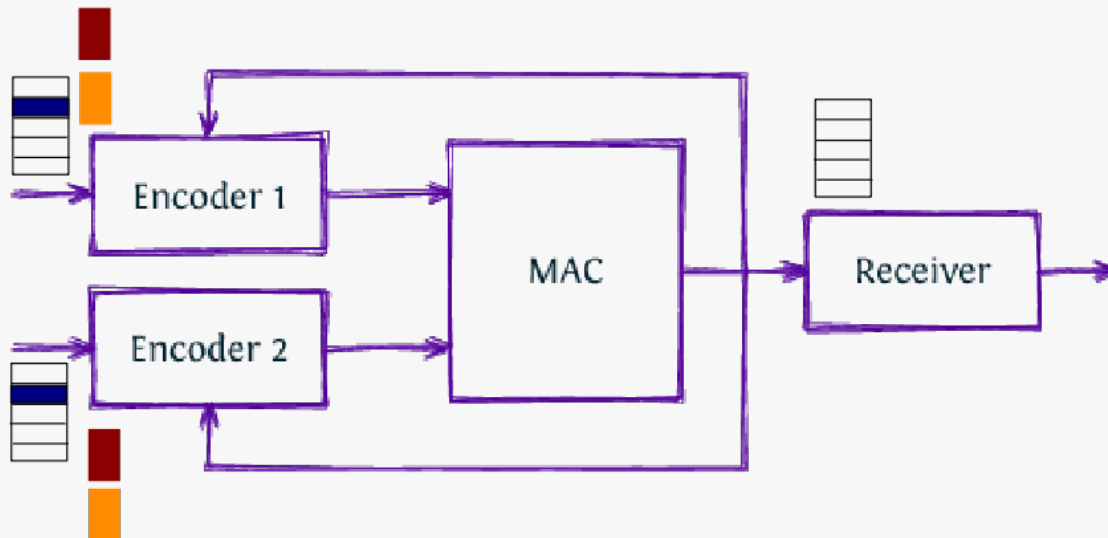
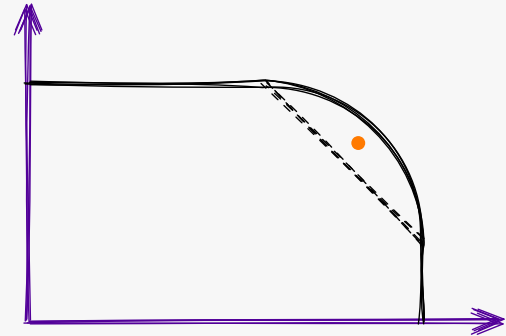
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Adding two way communication

Main idea

Each block now consists of three phases

[Bross-Lapidoth-05]



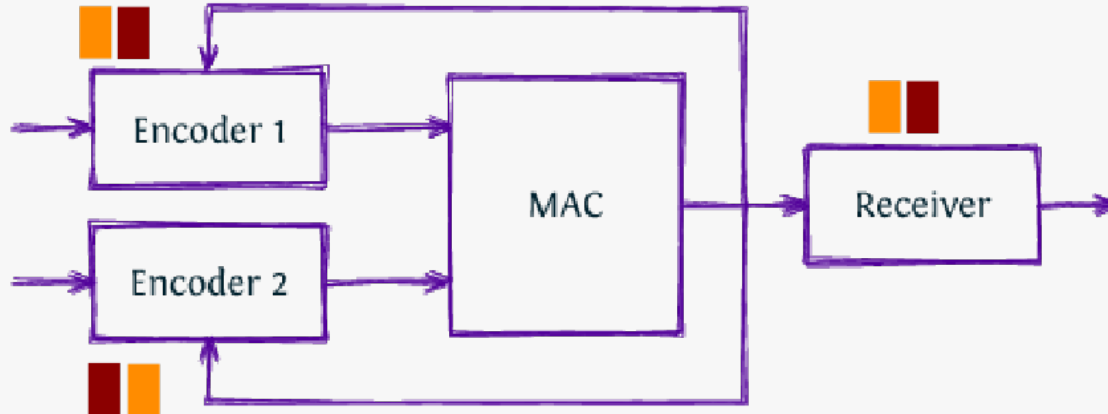
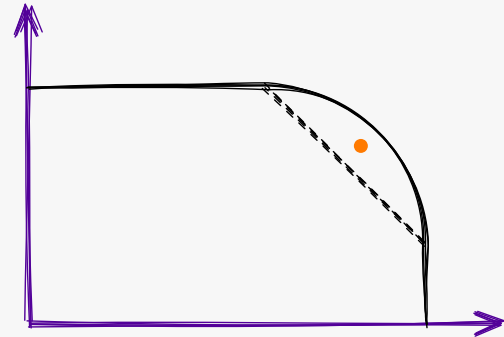
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Adding two way communication

Main idea

Each block now consists of three phases

[Bross-Lapidoth-05]



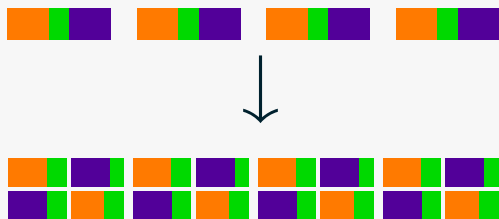
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Bross-Lapidoth Region



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Bross-Lapidoth Region



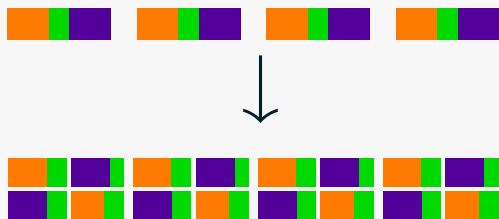
Capacity

$$R_{BL} = \bigcup_{\substack{P_U P_{X^1|U} P_{X^2|U} \\ g_1: \mathcal{X}^1 \times \mathcal{Y} \rightarrow \mathcal{V}^1, g_2: \mathcal{X}^2 \times \mathcal{Y} \rightarrow \mathcal{V}^2 \\ \text{stuff}}} \left\{ (R_1, R_2) : \begin{array}{l} R_1 \leq (1 + \eta)^{-1} I(X^1 \wedge Y | UX^2V^1) \\ R_2 \leq (1 + \eta)^{-1} I(X^2 \wedge Y | UX^1V^2) \\ R_1 + R_2 \leq (1 + \eta)^{-1} I(X^1 X^2 \wedge Y | V^1 V^2) - R_L \\ R_L \leq \text{complicated expression} \\ \eta \geq \text{complicated expression} \end{array} \right\}$$

[Bross-Lapidoth-05]

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Bross-Lapidoth Region



Comparison with CL

- ▶ BL region includes CL region
- ▶ In some cases the inclusion is strict

Capacity

$$R_{BL} = \bigcup_{\substack{P_U P_{X^1|U} P_{X^2|U} \\ g_1: \mathcal{X}^1 \times \mathcal{Y} \rightarrow \mathcal{V}^1, g_2: \mathcal{X}^2 \times \mathcal{Y} \rightarrow \mathcal{V}^2 \\ \text{stuff}}} \left\{ (R_1, R_2) : \begin{array}{l} R_1 \leq (1 + \eta)^{-1} I(X^1 \wedge Y | UX^2V^1) \\ R_2 \leq (1 + \eta)^{-1} I(X^2 \wedge Y | UX^1V^2) \\ R_1 + R_2 \leq (1 + \eta)^{-1} I(X^1 X^2 \wedge Y | V^1 V^2) - R_L \\ R_L \leq \text{complicated expression} \\ \eta \geq \text{complicated expression} \end{array} \right\}$$

[Bross-Lapidoth-05]



Block Markov superposition codes

Cover-Leung Scheme

- ▶ Decoder decodes after **one block**
- ▶ Uses **one auxiliary random variable**



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Block Markov superposition codes

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Bross-Lapidoth Scheme

- ▶ Decoder decodes after **two blocks**
- ▶ Uses **three auxiliary random variables** and **two partial functions**



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Block Markov superposition codes

Cover-Leung Scheme

- ▶ Decoder decodes after **one block**
- ▶ Uses **one auxiliary random variable**



Bross-Lapidoth Scheme

- ▶ Decoder decodes after **two blocks**
- ▶ Uses **three auxiliary random variable** and **two partial functions**



Questions

- ▶ Are these the best coding schemes that decode after one and two blocks?
- ▶ Can we improve the achievable region by using more auxiliary random variables?

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Venkataramanan-Pradhan scheme



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Venkataramanan-Pradhan scheme



Capacity

$$R_{VP} = \bigcup \left\{ \begin{array}{l} R_1 \leq I(X^1 \wedge Y | UW X^2 V^2) - [I(V^1 \wedge X^2 | Y V^2 W U) - I(U \wedge Y)]^+ \\ R_2 \leq I(X^2 \wedge Y | UW X^1 V^1) - [I(V^2 \wedge X^1 | Y V^1 W U) - I(U \wedge Y)]^+ \\ R_1 + R_2 \leq I(X^1 X^2 \wedge Y) - I(W \wedge Y | U) \end{array} \right\}$$

union over $P_U P_{V^1 V^2} P_{X^1 | U V^1} p_{X^2 | U V^2} Q_{\tilde{Y}^1 | X^1 U V^1 V^2 Y} Q_{\tilde{Y}^2 | X^2 U V^1 V^2 Y}$,

$W = (U, V^1, V^2, Y)$, stuff

[Venkataramanan-Pradhan-09]



Venkataramanan-Pradhan scheme

Comparison with earlier schemes



- ▶ VP region includes CL region
- ▶ In some cases the VP region strictly improves over BL region

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[Venkataramanan-Pradhan-09]



Questions

© Structure of optimal coding scheme

Given that we are decoding after n -blocks, can we improve performance if we use a more sophisticated coding scheme?



Questions

⑤ Structure of optimal coding scheme

Given that we are decoding after n -blocks, can we improve performance if we use a more sophisticated coding scheme?

⑤ Nature of auxiliary random variables

Are the auxiliary random variables intrinsic to the problem or are they an artifact of our solution approach?

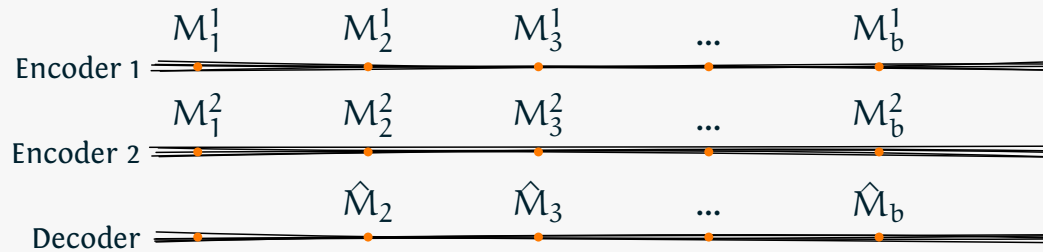


Outline

1. Overview of MAC with feedback
 - ▶ Cover-Leung scheme
 - ▶ Bross-Lapidoth scheme
 - ▶ Venkataramanan-Pradhan scheme
2. **Formulation as a decentralized control problem with delayed sharing**
 - ▶ **delay = 1**
 - ▶ **delay = 2**
3. Conclusion



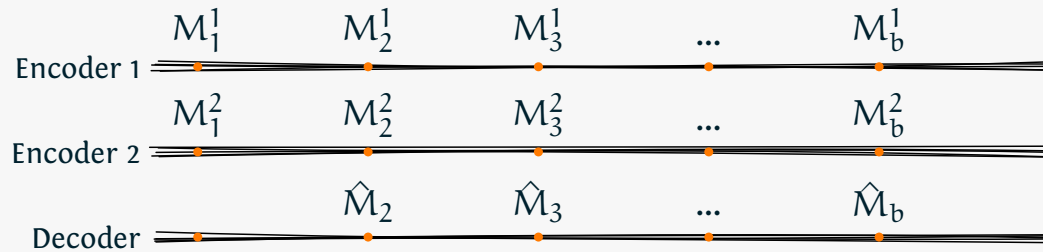
Information structure at block level (n=1)



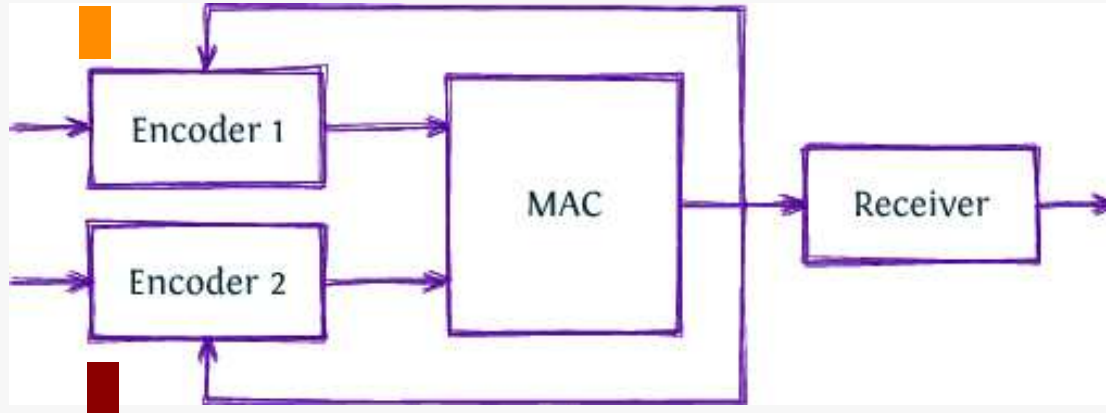
$$\hat{M}_b = M_{b-1}^1 M_{b-1}^2 \text{ with high probability}$$



Information structure at block level (n=1)

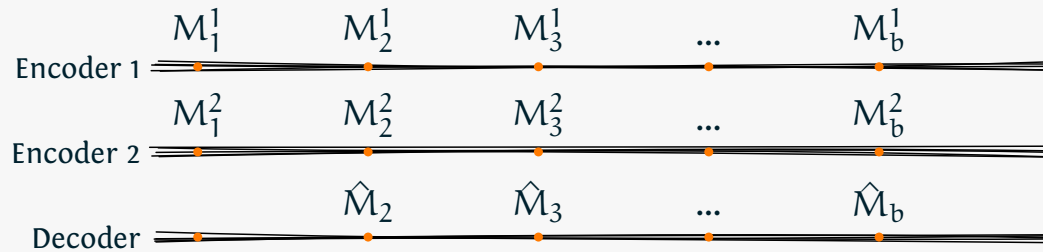


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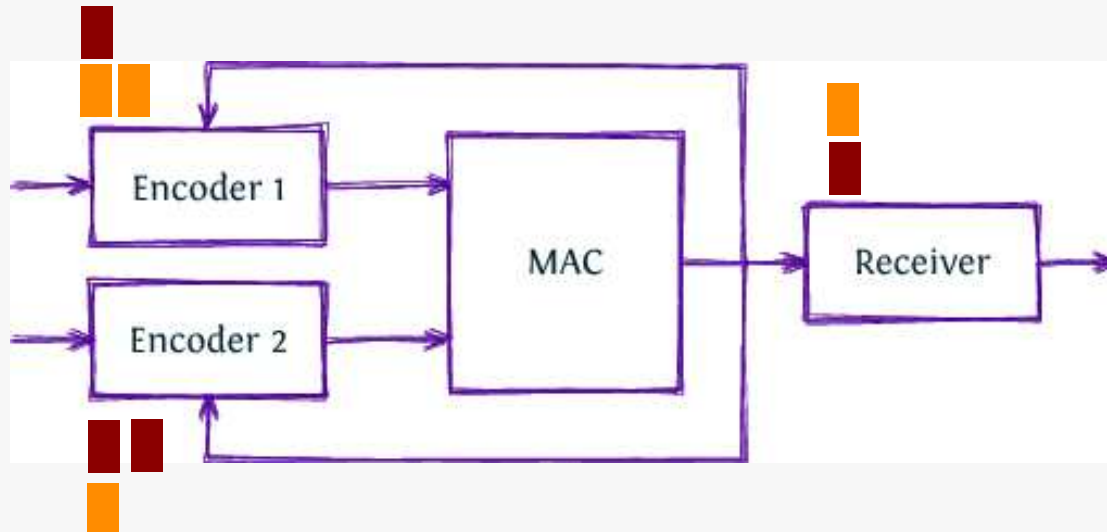


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Information structure at block level (n=1)

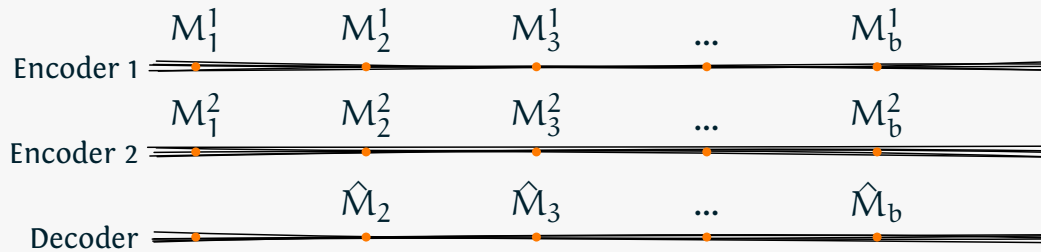


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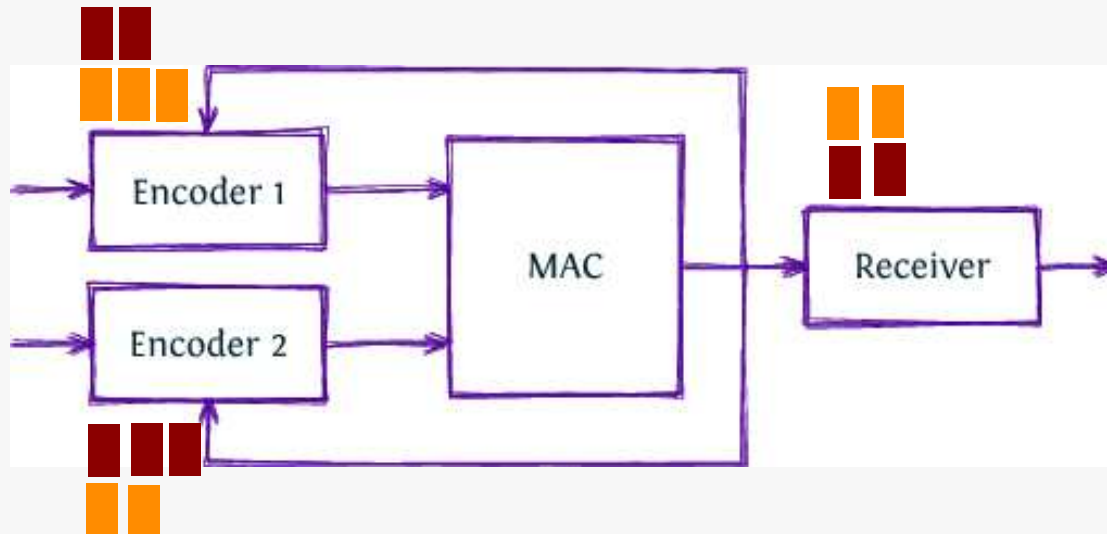


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Information structure at block level (n=1)



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Handwritten scribbles at the bottom of the slide.

Information structure at block level ($n=1$)

Encoder 1

$$\begin{aligned} &\overline{M}_{[1:b-1]}, Y_{[1:b-1]}, \\ &M_b^1 \end{aligned}$$

$$\overline{M}_{[1:b-2]}, Y_{[1:b]}$$

Decoder

$$\begin{aligned} &\overline{M}_{[1:b-1]}, Y_{[1:b-1]}, \\ &M_b^2 \end{aligned}$$

Encoder 2

|||||

Info struc at block level (general n)

Encoder 1

$$\overline{M}_{[1:b-n]}, Y_{[1:b-1]},$$

$$M^1_{[b-n+1:b]}$$

$$\overline{M}_{[1:b-n-1]}, Y_{[1:b]}$$

Decoder

$$\overline{M}_{[1:b-n]}, Y_{[1:b-1]},$$

$$M^2_{[b-n+1:b]}$$

Encoder 2

|||||

Block Markov superposition coding at block level

Setup

- ▶ **i.i.d. Messages** at each encoders
 $\overline{\mathbf{M}}_b = (\mathbf{M}_b^1, \mathbf{M}_b^2)$
- ▶ **Fixed delay decoding** $\hat{\mathbf{M}}_b$ is trying to reproduce $\overline{\mathbf{M}}_{b-n}$.



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- ▶ The encoder know $\overline{\mathbf{M}}_{b-n}$
- ▶ The decoder knows $\overline{\mathbf{M}}_{b-n-1}$



Block Markov superposition coding at block level

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Coding scheme

- ▶ **Encoder:** $X_b^i = f_b^i(\overline{M}_{[1:b-n]}, Y_{[1:b-1]}, M_{[b-n+1:b]}^i)$
- ▶ **Decoder:** $\hat{M}_b = d(\overline{M}_{[1:b-n-1]}, Y_{[1:b]})$
- ▶ **Error:** $\rho(\overline{M}_{b-n}, \hat{M}_b)$



Block Markov superposition coding at block level

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 $\overline{M}_b = (M_b^1, M_b^2)$
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- ▶ **Error:** $\rho(\overline{M}_{b-n}, \hat{M}_b)$
- ▶ **Objective:** Choose coding scheme to minimize $\sum_{b=n}^B \mathbb{E}[\rho(\overline{M}_{b-n}, \hat{M}_b)]$



Transforming information structures

Encoder 1

$$\overline{M}_{[1:b-n]}, Y_{[1:b-1]},$$

$$M^1_{[b-n+1:b]}$$

$$\overline{M}_{[1:b-n-1]}, Y_{[1:b]}$$

Decoder

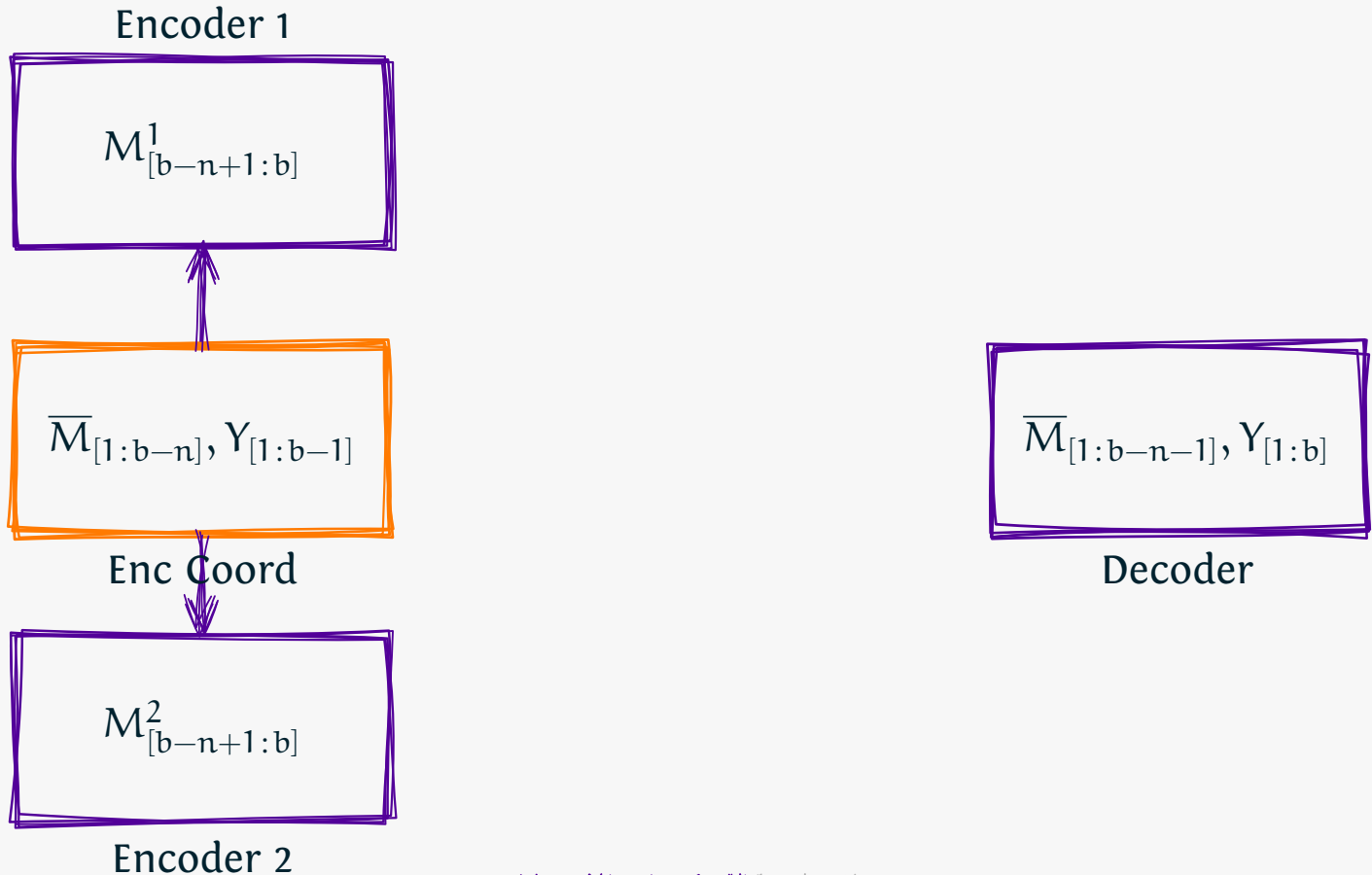
$$\overline{M}_{[1:b-n]}, Y_{[1:b-1]},$$

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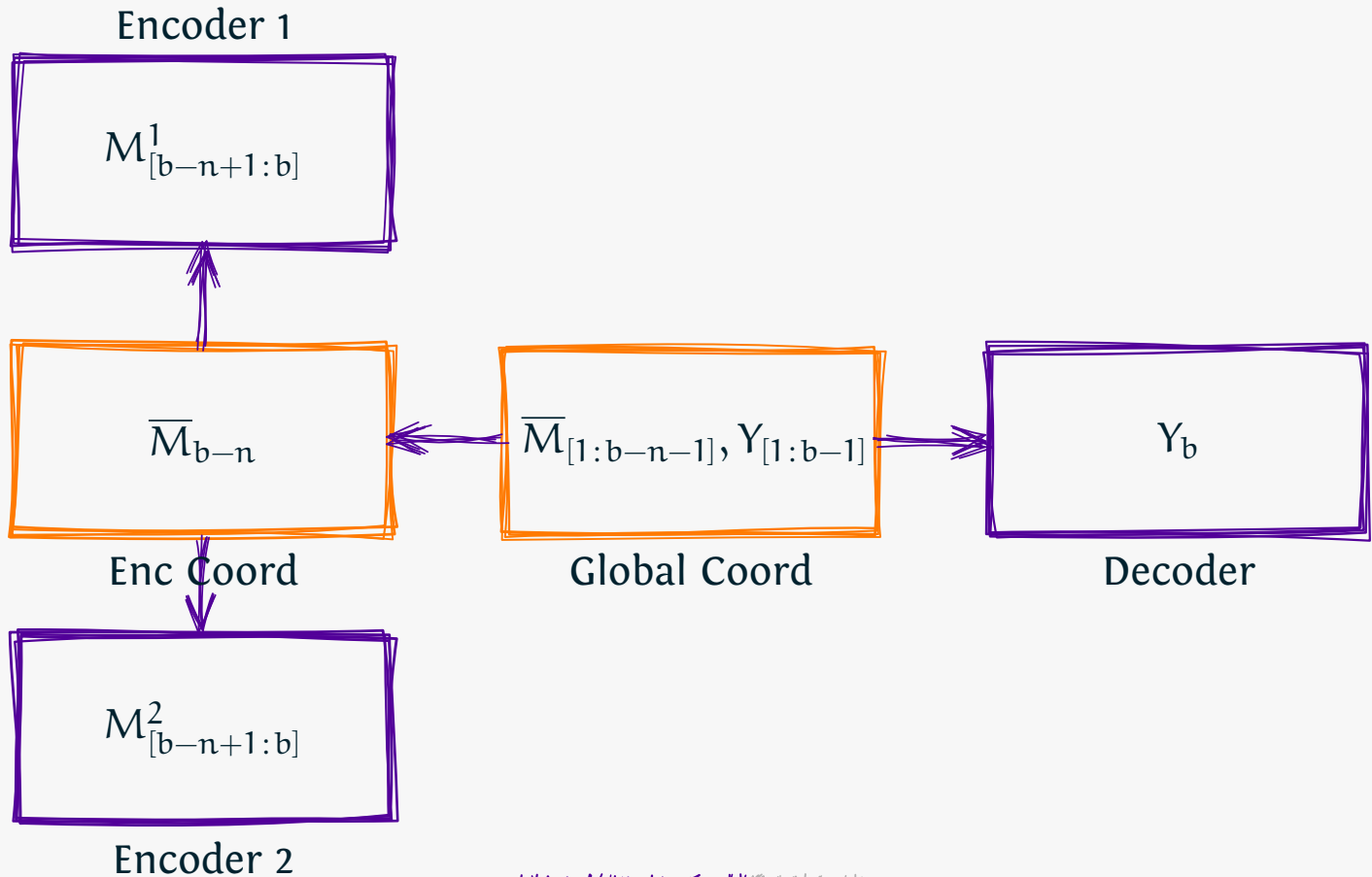
Encoder 2

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Transforming information structures

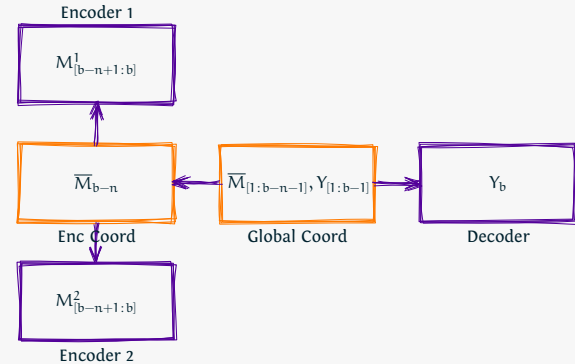
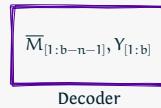
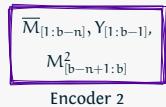
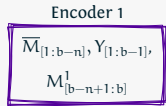


Transforming information structures

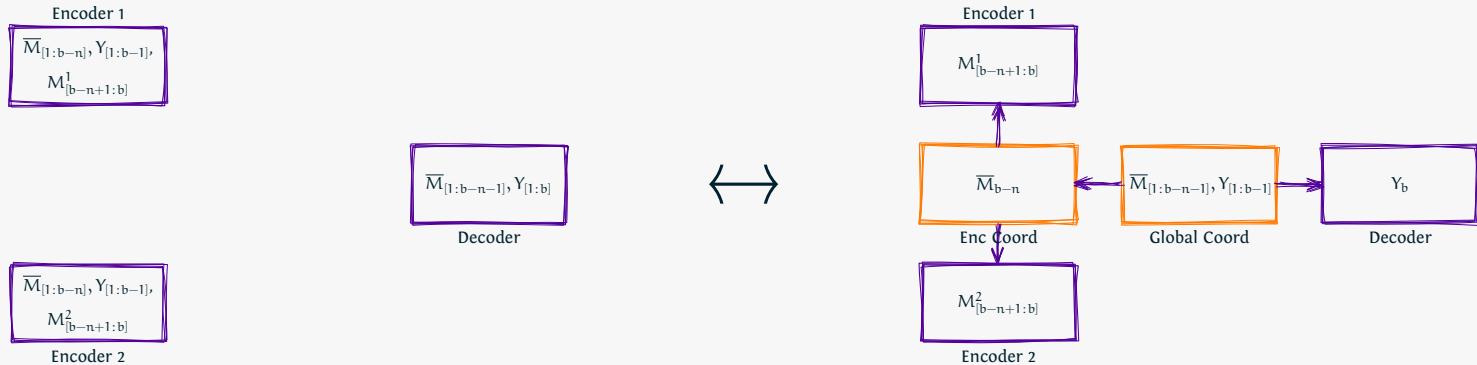


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Centralized planning in decentralized systems



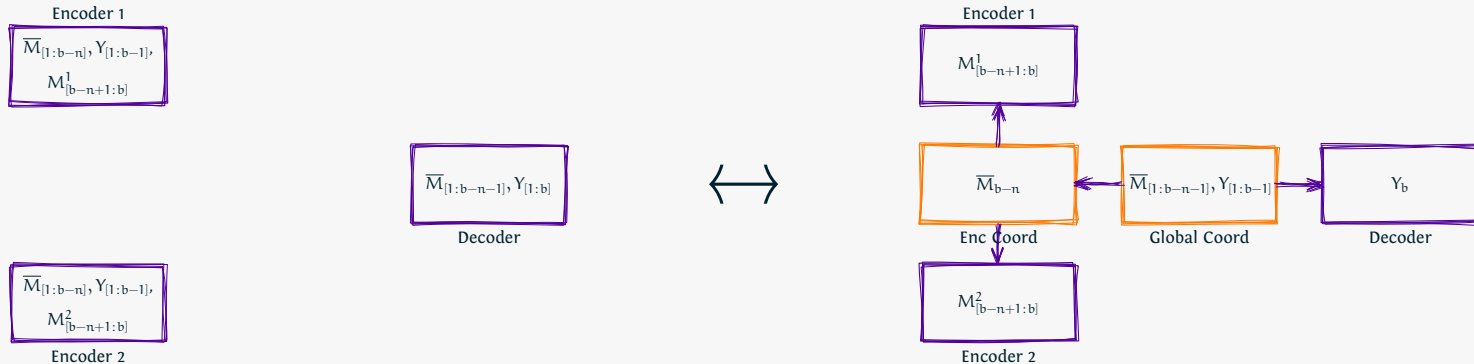
Centralized planning in decentralized systems



Hierarchical delayed sharing info structures



Centralized planning in decentralized systems

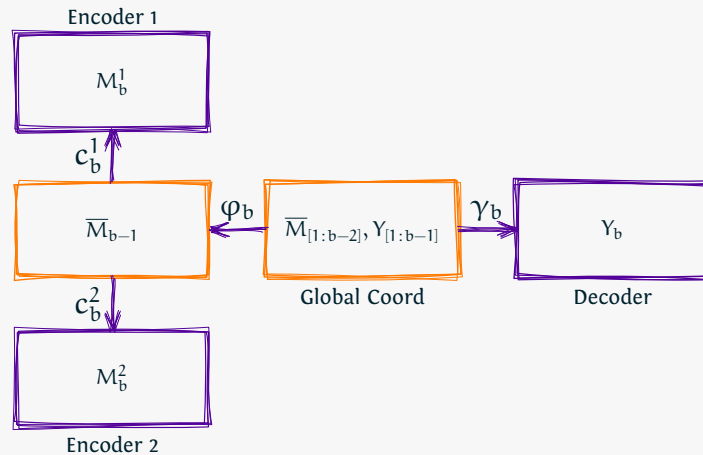


Hierarchical delayed sharing info structures

- ▶ Look at the system from global coordinator's point of view
- ▶ Show that the two optimization problems are **equivalent**
- ▶ Derive **structure** of optimal control policies **for the coordinator**

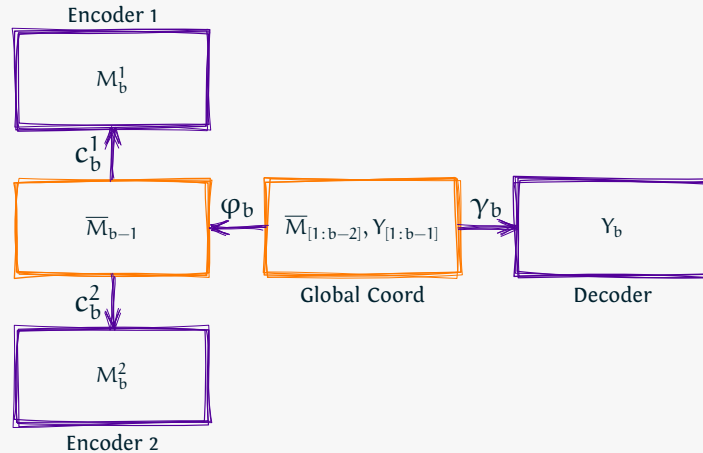


Delay $n=1$: Cover-Leung setup



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Delay n=1: Cover-Leung setup



Global coordinator

$$(\varphi_b, \gamma_b) = \psi_b(\bar{M}_{[1:b-2]}, Y_{[1:b-1]})$$

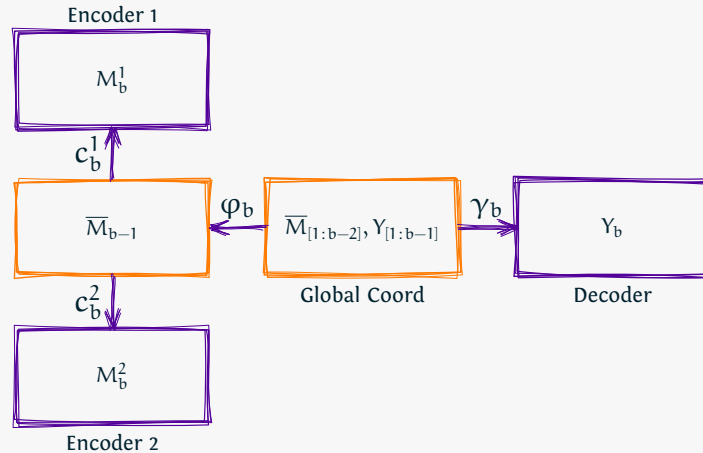
where

$$\varphi_b : \bar{\mathcal{M}} \rightarrow ((\mathcal{M}^1 \rightarrow \mathcal{X}^1), (\mathcal{M}^2 \rightarrow \mathcal{X}^2))$$

$$\gamma_b : \mathcal{Y} \rightarrow \hat{\mathcal{M}}$$



Delay n=1: Cover-Leung setup



Global coordinator

Encoder coordinator

$$(\varphi_b, \gamma_b) = \psi_b(\bar{M}_{[1:b-2]}, Y_{[1:b-1]})$$

$$(c_b^1, c_b^2) = \varphi_b(\bar{M}_{b-1})$$

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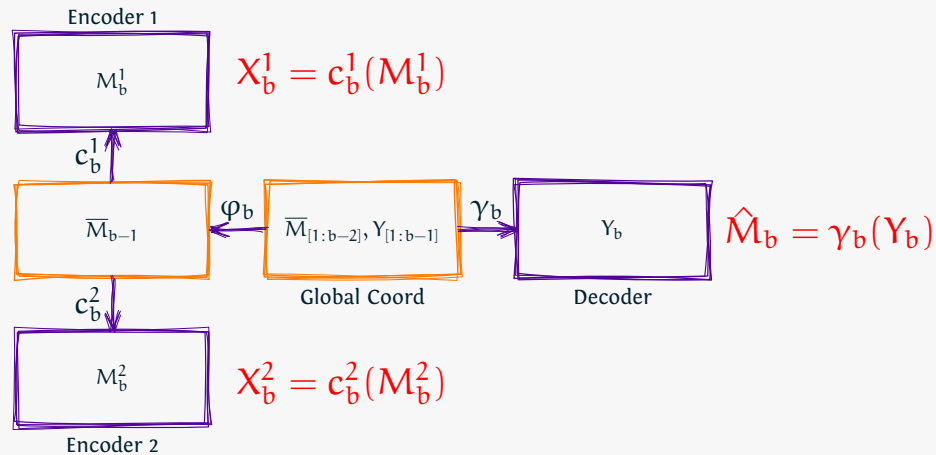
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$$c_b^i : \mathcal{M}^i \rightarrow \mathcal{X}^i$$

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Delay n=1: Cover-Leung setup



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Structure of optimal control policy

$$\left(\overline{M}_{[1:b-2]}, Y_{[1:b-1]}, \Phi_{[1:b-1]}, \Upsilon_{[1:b-1]} \right) \xrightarrow{\text{info state}} (Y_{b-1}, c_{b-1}^1, c_{b-1}^2)$$



Structure of optimal control policy

$$(\bar{M}_{[1:b-2]}, Y_{[1:b-1]}, \Phi_{[1:b-1]}, \gamma_{[1:b-1]}) \xrightarrow{\text{info state}} (Y_{b-1}, c_{b-1}^1, c_{b-1}^2)$$

Controlled Markov process

$$\begin{aligned} & \mathbb{P}(Y_b, c_b^1, c_b^2 \mid \bar{M}_{[1:b-2]}, Y_{[1:b-1]}, \Phi_{[1:b]}, \gamma_{[1:b]}) \\ &= \sum_{\bar{M}_{b-1,b}} \mathbb{P}(Y_b \mid c_b^1(M_b^1) c_b^2(M_b^2)) \times \mathbb{1}[(c_b^1, c_b^2) = \varphi_b(\bar{M}_{b-1})] \\ & \quad \times \frac{\mathbb{P}(Y_{b-1} \mid c_{b-1}^1(M_{b-1}^1) c_{b-1}^2(M_{b-1}^2)) \mathbb{P}(\bar{M}_{b-1})}{\sum_{\underline{M}_{b-1}} \mathbb{P}(Y_{b-1} \mid c_{b-1}^1(\underline{M}_{b-1}^1) c_{b-1}^2(\underline{M}_{b-1}^2)) \mathbb{P}(\underline{M}_{b-1})} \\ &= \mathbb{P}(Y_b, c_b^1, c_b^2 \mid Y_{b-1}, c_{b-1}^1, c_{b-1}^2, \varphi_b) \end{aligned}$$



Structure of optimal control policy

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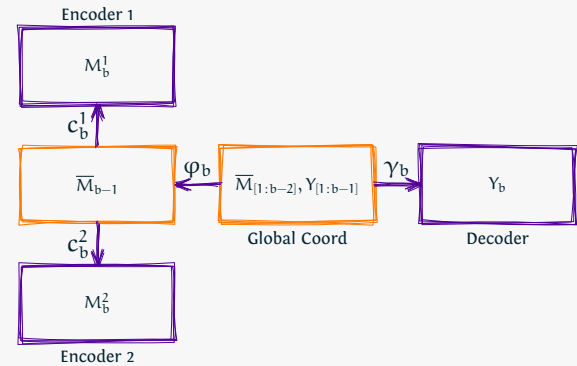
Expected cost per unit time

$$\begin{aligned} \mathbb{E}[\rho(\overline{M}_{b-1}, \hat{M}_b) \mid \overline{M}_{[1:b-2]}, Y_{[1:b-1]}, \varphi_{[1:b]}, \gamma_{[1:b]}] \\ = \mathbb{E}[\rho(\overline{M}_{b-1}, \hat{M}_b) \mid Y_{b-1}, c_{b-1}^1, c_{b-1}^2, \varphi_b, \gamma_b] \end{aligned}$$



Structure of optimal control policy

⊙ $(Y_{b-1}, c_{b-1}^1, c_{b-1}^2)$ is a controlled Markov process with control action (φ_b, γ_b) .



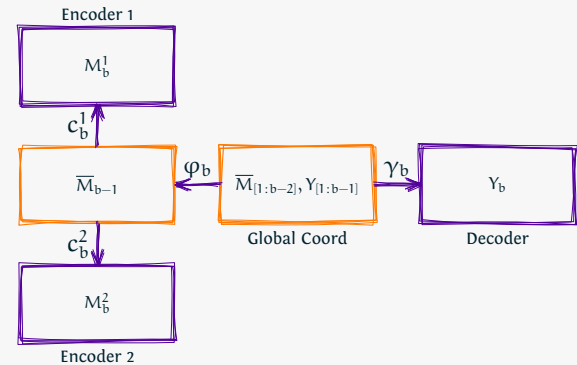
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③ **Without loss of optimality**

$$(\varphi_b, \gamma_b) = \psi_b(Y_{b-1}, c_{b-1}^1, c_{b-1}^2)$$



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Structure of optimal control policy

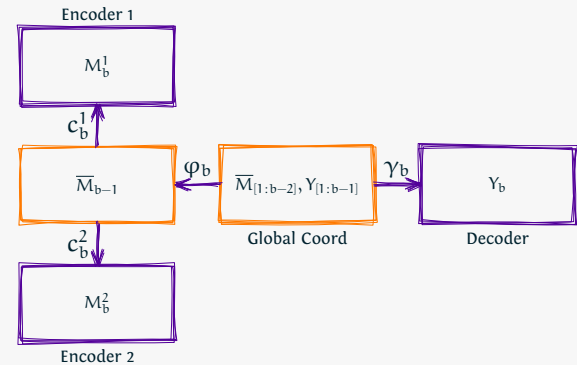
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⊙ This is equivalent to

$$X_b^i = f_b^i(\bar{M}_{b-1}, Y_{b-1}, c_{b-1}^1, c_{b-1}^2, M_b^1), \quad i = 1, 2$$



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Structure of optimal control policy

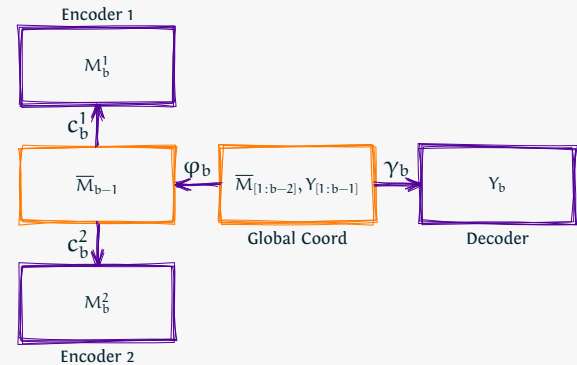
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$$(\varphi_b, \gamma_b) = \psi_b(Y_{b-1}, c_{b-1}^1, c_{b-1}^2)$$

⊙ This is equivalent to

$$\begin{aligned} X_b^i &= f_b^i(\bar{M}_{b-1}, Y_{b-1}, c_{b-1}^1, c_{b-1}^2, M_b^1), \quad i = 1, 2 \\ &= f_b^i(U_b, M_b^i) \end{aligned}$$



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② **Without loss of optimality**

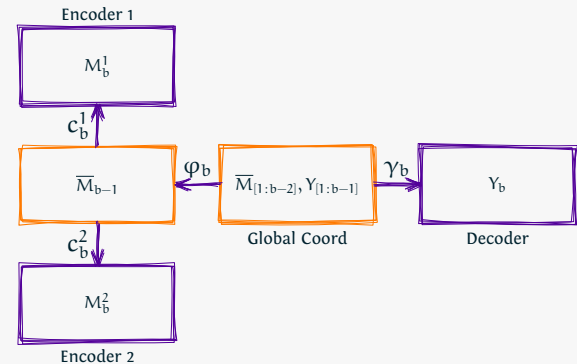
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$$\begin{aligned} X_b^i &= f_b^i(\bar{M}_{b-1}, Y_{b-1}, c_{b-1}^1, c_{b-1}^2, M_b^1), \quad i = 1, 2 \\ &= f_b^i(U_b, M_b^i) \end{aligned}$$

② Fresh information : M_b^i

② Auxiliary variable $U_b \equiv (\bar{M}_{b-1}, Y_{b-1}, c_{b-1}^1, c_{b-1}^2)$



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Structure of optimal control policy

⊙ $(Y_{b-1}, c_{b-1}^1, c_{b-1}^2)$ is a controlled Markov process with control action (φ_b, γ_b) .

⊙ **Without loss of optimality**

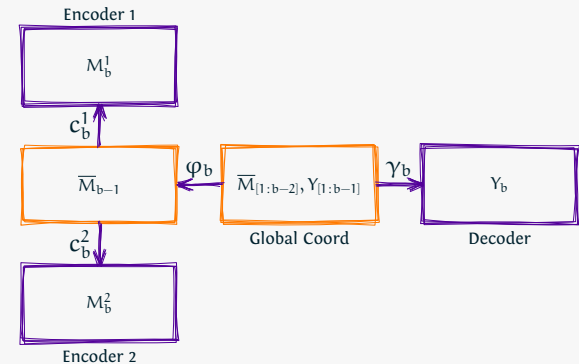
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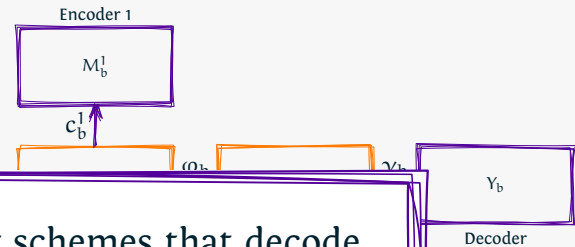
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Within the class of all block Markov coding schemes that decode after one block, Cover-Leung scheme has the optimal structure.



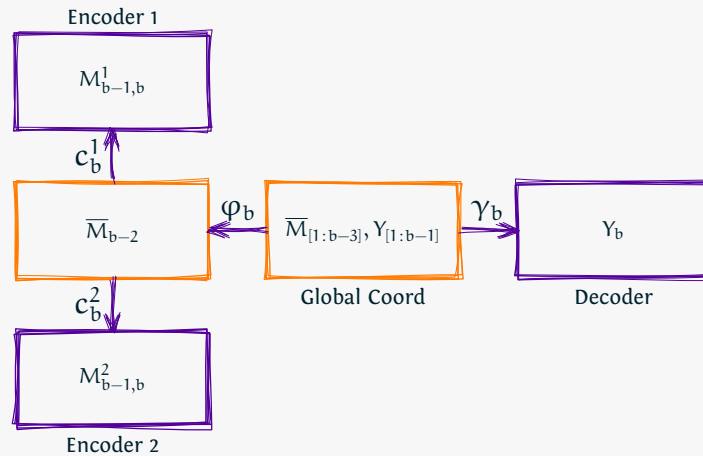
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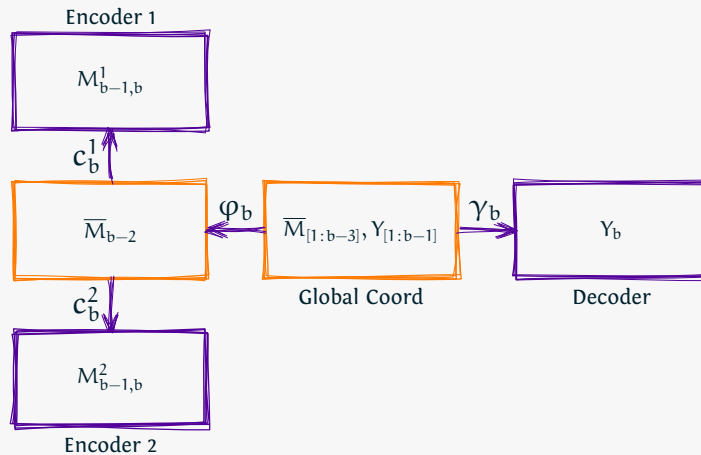


Delay $n=2$: Venkataramanan Pradhan setup



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Delay n=2: Venkataramanan Pradhan setup



Global coordinator

$$(\varphi_b, \gamma_b) = \psi_b(\bar{M}_{[1:b-3]}, Y_{[1:b-1]})$$

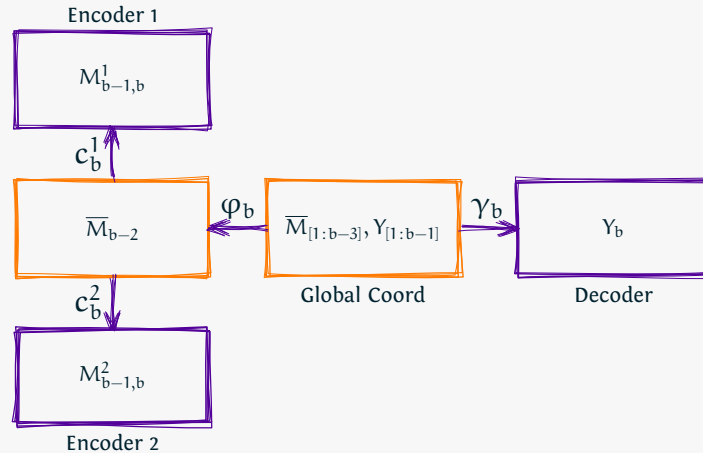
where

$$\varphi_b : \bar{\mathcal{M}} \rightarrow ((\mathcal{M}^1 \times \mathcal{M}^1 \rightarrow \mathcal{X}^1), (\mathcal{M}^2 \times \mathcal{M}^2 \rightarrow \mathcal{X}^2))$$

$$\gamma_b : \mathcal{Y} \rightarrow \hat{\mathcal{M}}$$

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Delay n=2: Venkataramanan Pradhan setup



Global coordinator

Encoder coordinator

$$(\varphi_b, \gamma_b) = \psi_b(\bar{M}_{[1:b-3]}, Y_{[1:b-1]})$$

$$(c_b^1, c_b^2) = \varphi_b(\bar{M}_{b-2})$$

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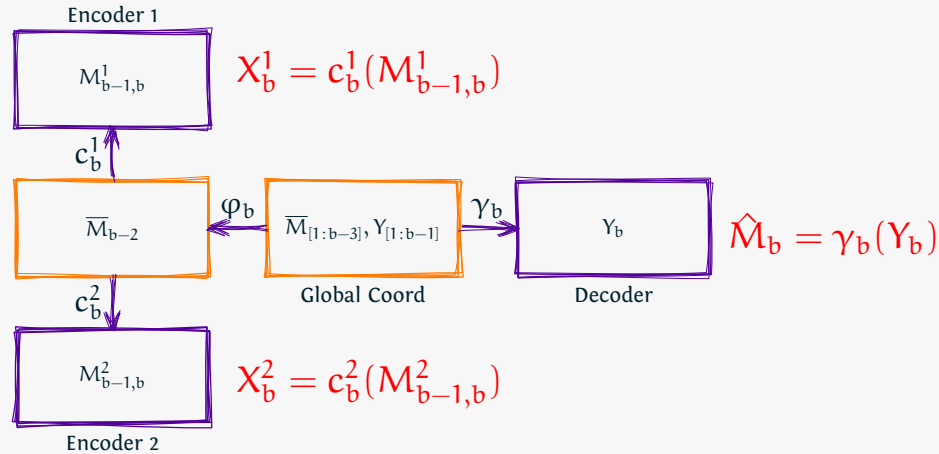
$$\varphi_b : \bar{\mathcal{M}} \rightarrow ((\mathcal{M}^1 \times \mathcal{M}^1 \rightarrow \mathcal{X}^1), (\mathcal{M}^2 \times \mathcal{M}^2 \rightarrow \mathcal{X}^2))$$

$$c_b^i : \mathcal{M}^i \times \mathcal{M}^i \rightarrow \mathcal{X}^i$$

$$\gamma_b : \mathcal{Y} \rightarrow \hat{\mathcal{M}}$$



Delay n=2: Venkataramanan Pradhan setup



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$$(\varphi_b, \gamma_b) = \psi_b(\bar{M}_{[1:b-3]}, Y_{[1:b-1]})$$

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Structure of optimal control policy

$$\left(\overline{M}_{[1:b-3]}, Y_{[1:b-1]}, \Phi_{[1:b-1]}, \Upsilon_{[1:b-1]} \right) \xrightarrow{\text{info state}} (Y_{b-2,b-1}, c_{b-1}^1, c_{b-1}^2, \hat{c}_{b-2}^1, \hat{c}_{b-2}^2)$$

where $\hat{c}_{b-2}^i(\cdot) = c_{b-2}^i(M_{b-3}^i, \cdot)$, $i = 1, 2$.



Structure of optimal control policy

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⊙ Controlled Markov process

$$\begin{aligned} & \mathbb{P}(Y_{b-1,b}, c_b^1, c_b^2, \hat{c}_{b-1}^1, \hat{c}_{b-1}^2 \mid \overline{M}_{[1:b-2]}, Y_{[1:b-1]}, \varphi_{[1:b]}, \gamma_{[1:b]}) \\ &= \mathbb{P}(Y_b, c_b^1, c_b^2, \hat{c}_{b-1}^1, \hat{c}_{b-1}^2 \mid Y_{b-2,b-1}, c_{b-1}^1, c_{b-1}^2, \hat{c}_{b-2}^1, \hat{c}_{b-2}^2, \varphi_b) \end{aligned}$$



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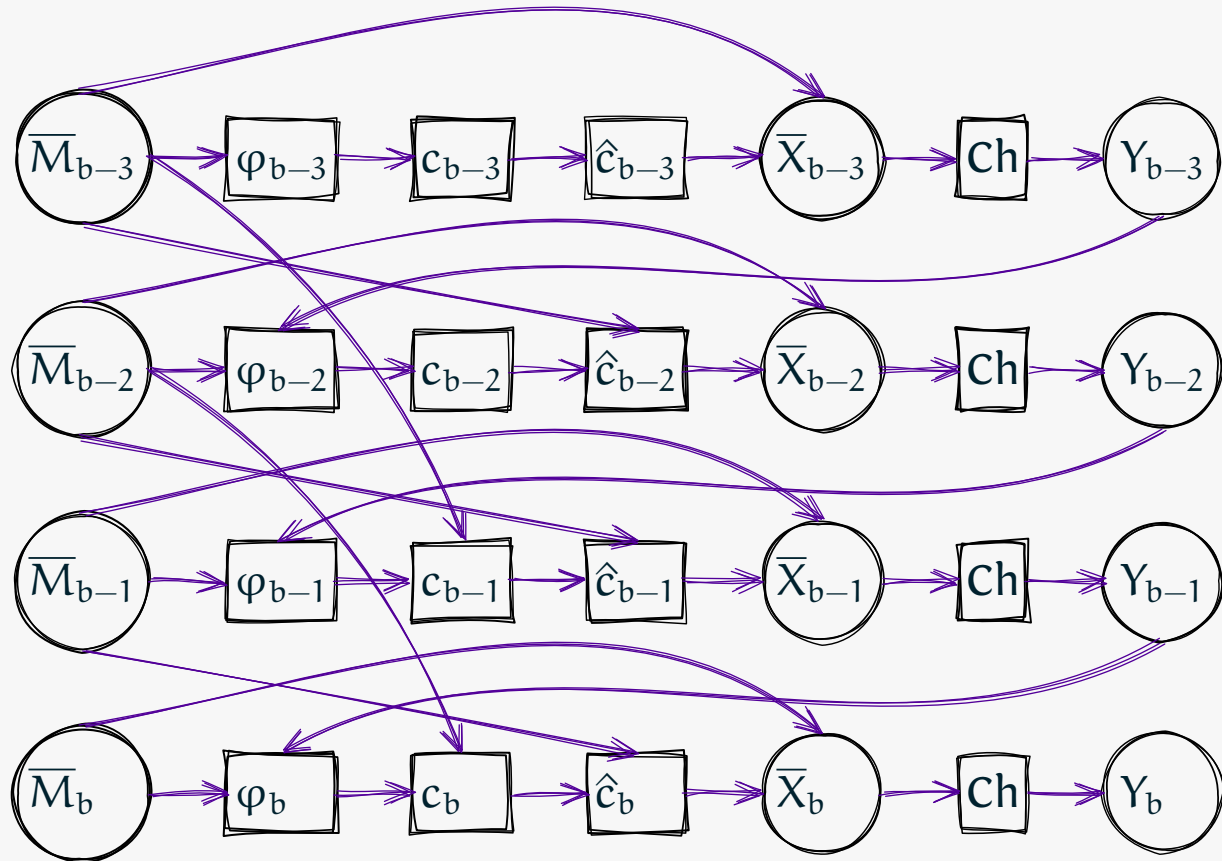
$$\begin{aligned} & \mathbb{P}(Y_{b-1,b}, c_b^1, c_b^2, \hat{c}_{b-1}^1, \hat{c}_{b-1}^2 \mid \overline{M}_{[1:b-2]}, Y_{[1:b-1]}, \varphi_{[1:b]}, \gamma_{[1:b]}) \\ &= \mathbb{P}(Y_b, c_b^1, c_b^2, \hat{c}_{b-1}^1, \hat{c}_{b-1}^2 \mid Y_{b-2,b-1}, c_{b-1}^1, c_{b-1}^2, \hat{c}_{b-2}^1, \hat{c}_{b-2}^2, \varphi_b) \end{aligned}$$

Expected cost per unit time

$$\begin{aligned} & \mathbb{E}[\rho(\overline{M}_{b-2}, \hat{M}_b) \mid \overline{M}_{[1:b-3]}, Y_{[1:b-1]}, \varphi_{[1:b]}, \gamma_{[1:b]}] \\ &= \mathbb{E}[\rho(\overline{M}_{b-2}, \hat{M}_b) \mid Y_{b-2,b-1}, c_{b-1}^1, c_{b-1}^2, \hat{c}_{b-2}^1, \hat{c}_{b-2}^2, \varphi_b, \gamma_b] \end{aligned}$$

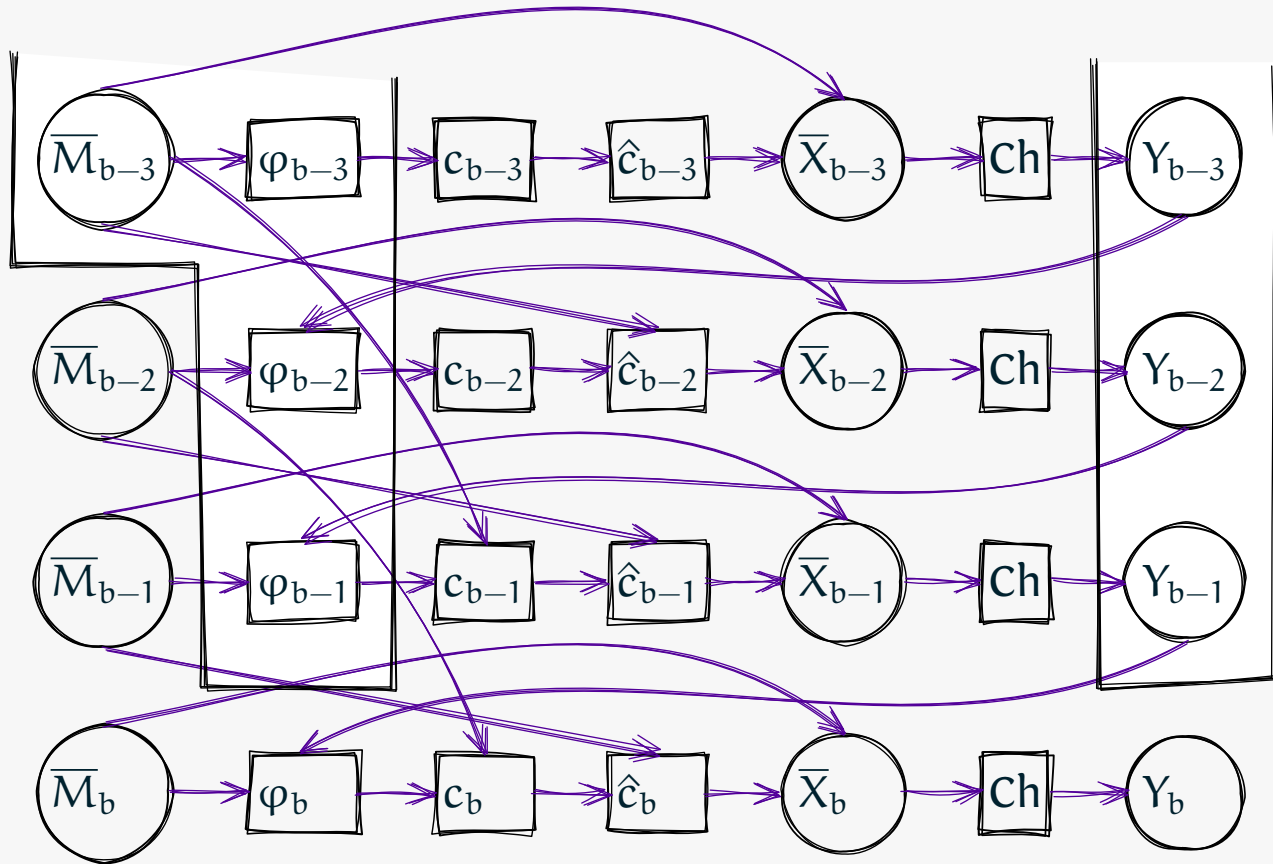


Structure of optimal control policy



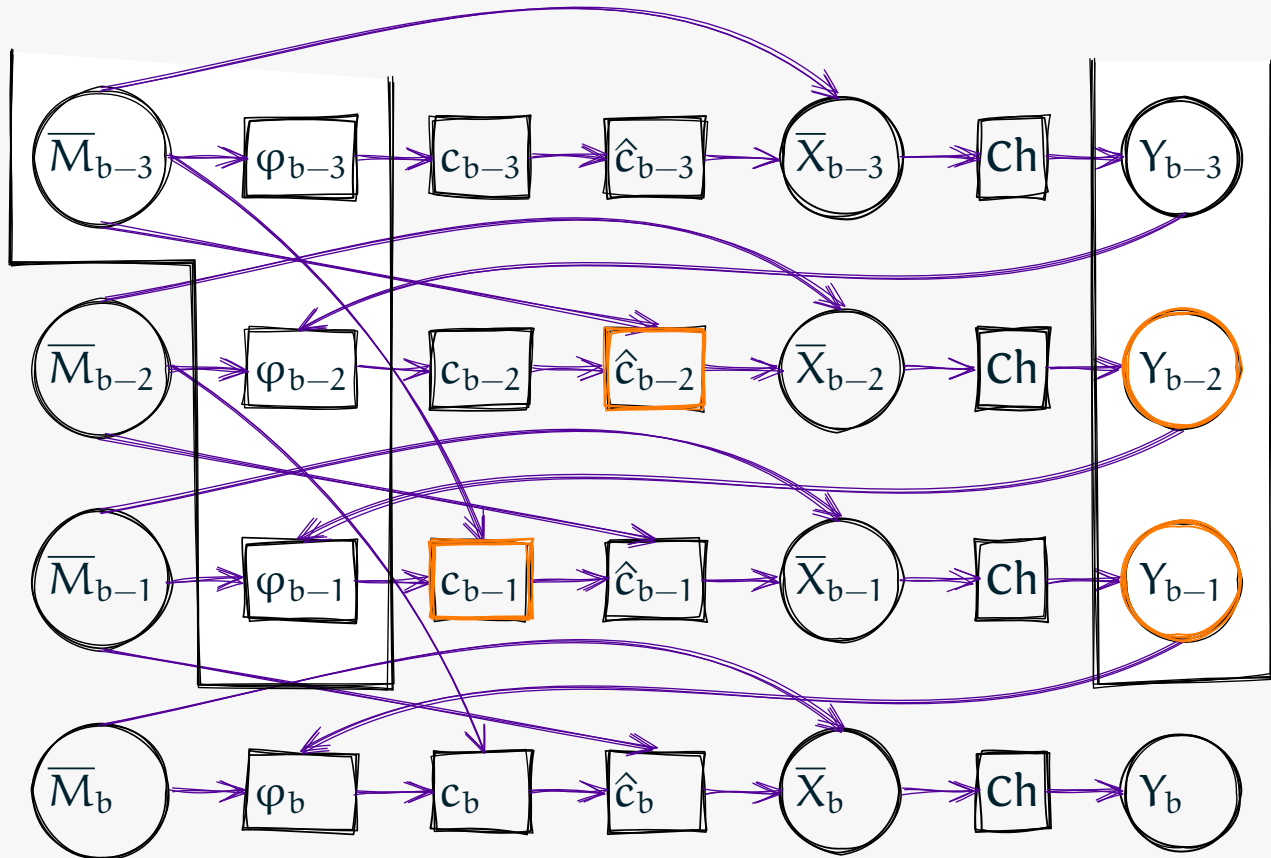
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Structure of optimal control policy



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Structure of optimal control policy



Structure of optimal control policy

⊙ $(Y_{b-2,b-1}, c_{b-1}^{1,2}, \hat{c}_{b-2}^{1,2})$ is a controlled Markov process with control action (φ_b, γ_b) .



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Within the class of all block Markov coding schemes that decode after two blocks, Venkataramanan-Pradhan scheme has the optimal structure.

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Summary

Main Results

- ▶ CL has optimal structure for delay=1
- ▶ VP has optimal structure for delay=2



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Summary

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Main point

- ▶ Auxiliary random variables are related to information states
- ▶ Systematic approaches to derive info states can be used for auxiliary random variables

Choice of info structures

- ▶ Block Markov superposition schemes enforce a specific information structure
- ▶ Stochastic control can identify whether an info structure is tractable (in many cases **without explicitly solving the system**)



Conclusion

Horses for courses

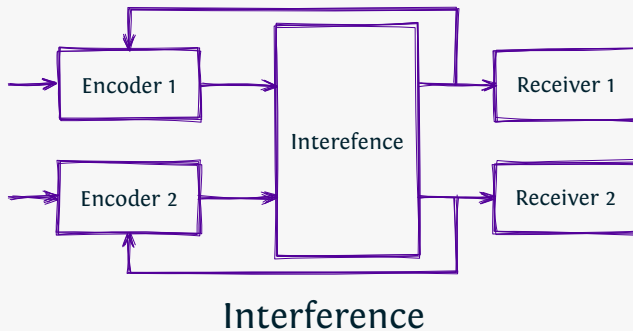
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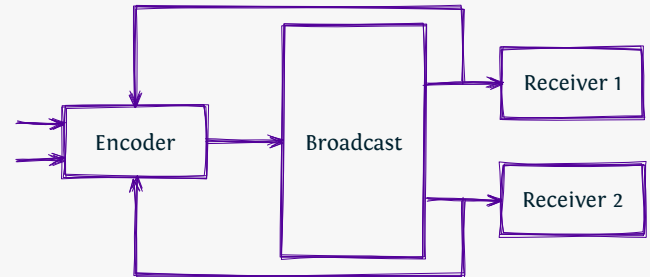
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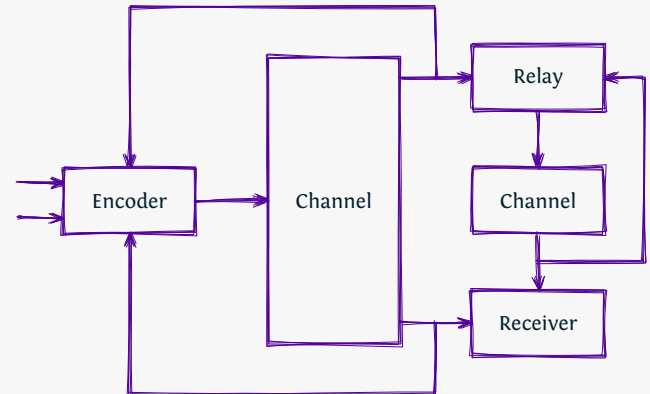
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Future directions



Broadcast



Relay

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Thank you