

The Golden Ratio in Communication

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Blackwell's Trapdoor Channel

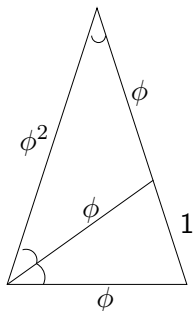
Task Assignment

Paul Cuff

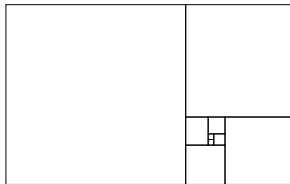
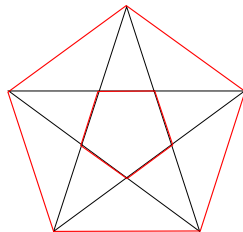
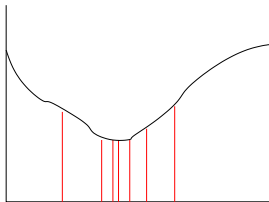
(Cover, Permuter, Weissman, Van Roy)

Stanford University

November 24, 2008

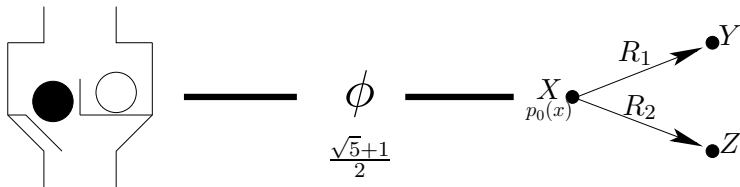


Golden Ratio



1,1,2,3,5,8,13,...

Outline - Markov Chain



The Trapdoor Channel

Introduced by David Blackwell in 1961.

[Ash 65], [Ahlswede & Kaspi 87], [Ahlswede 98], [Kobayashi 02 & 03].

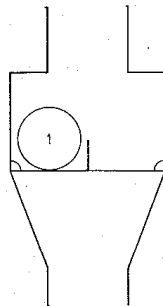
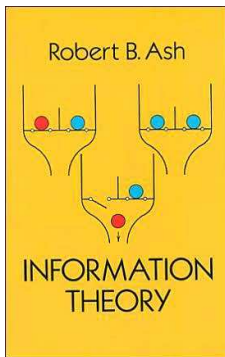
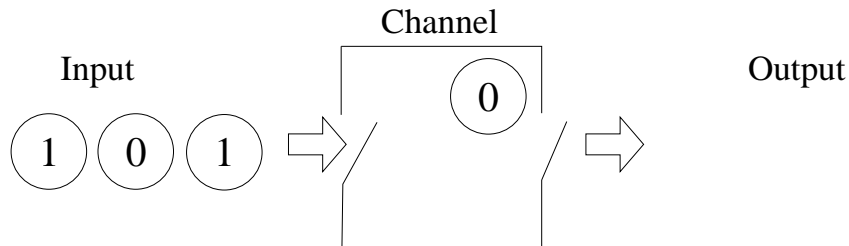


Fig. 7.1 A simple two-state channel.

A *“simple two-state channel.”* - Blackwell

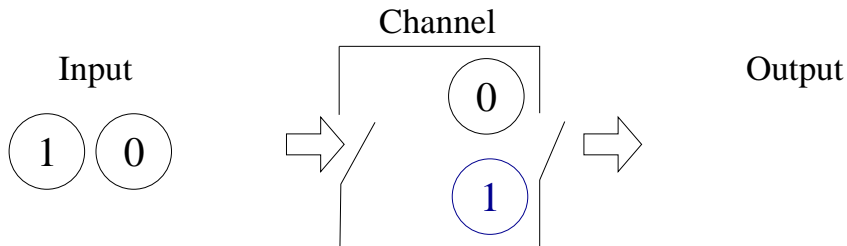
The Trapdoor Channel



$$s_t = s_{t-1} + x_t - y_t$$

$$s_0 = 0$$

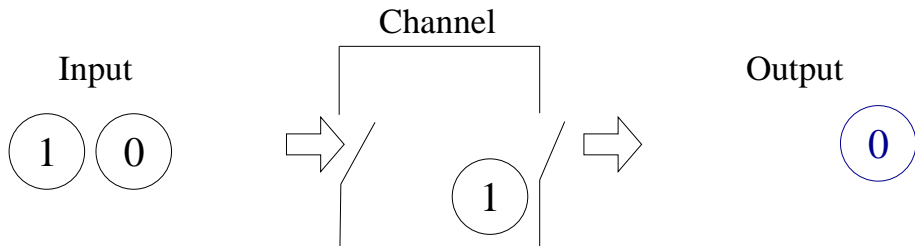
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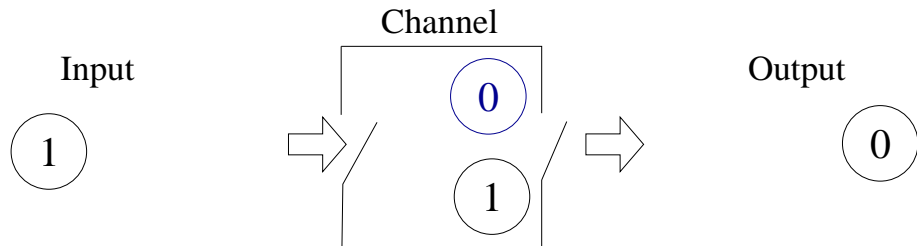


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The Trapdoor Channel



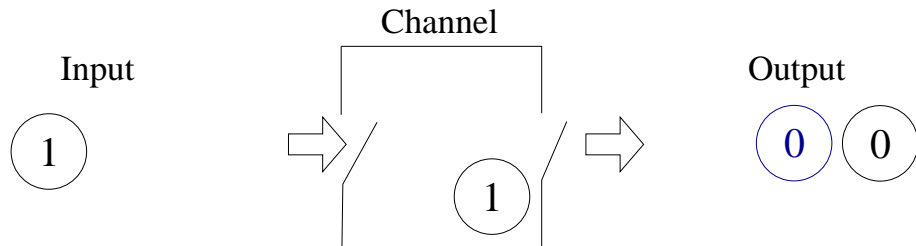
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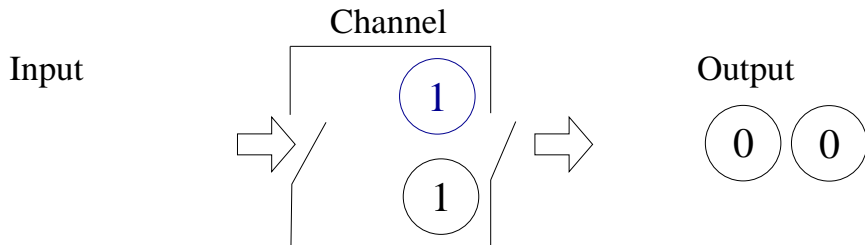
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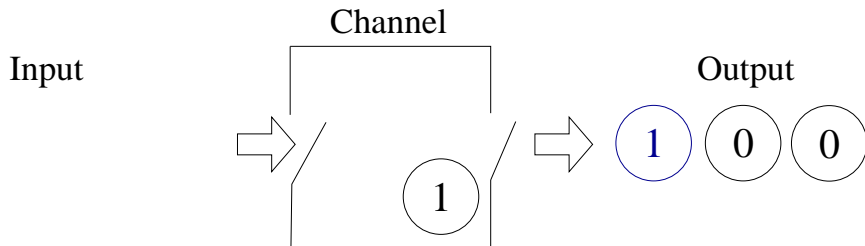
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Communication without Feedback

- Repeat each bit three time: $R = 1/3$ bit.

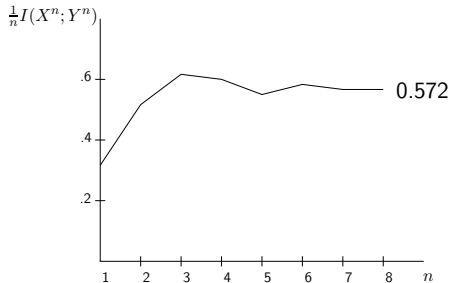
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Communication without Feedback

- Repeat each bit three time: $R = 1/3$ bit.
- Repeat each bit twice: $R = 1/2$ bit. [Ahlsweide & Kaspi 87]
- $C \approx 0.572$ bits per channel use. [Kobayashi & Morita 03]

$$C = \lim_{n \rightarrow \infty} \max_{p(x_1, \dots, x_n)} \frac{1}{n} I(X^n; Y^n).$$



Communication Setting (with feedback)

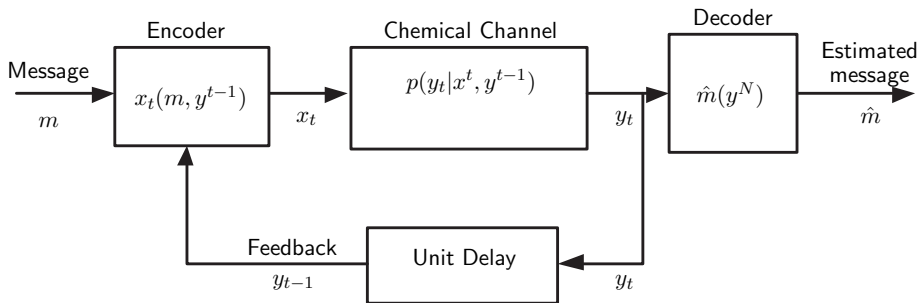


Figure: Communication with feedback

Feedback Capacity of Unifilar, Strongly Connected, FSC

- Capacity of the Trapdoor Channel

$$C = \lim_{n \rightarrow \infty} \frac{1}{n} \max_{p(x_1, \dots, x_n)} I(X^n; Y^n)$$

- Feedback capacity of the Trapdoor Channel

$$C_{FB} = \lim_{n \rightarrow \infty} \frac{1}{n} \max_{\{p(x_i | x^{i-1}, y^{i-1})\}_{i=1}^n} I(X^n \rightarrow Y^n)$$

[Permuter, Weissman & Goldsmith ISIT06]

Directed Information

Mutual Information

$$I(X^n; Y^n) = \sum_{i=1}^n I(X^n; Y_i | Y^{i-1})$$

Directed Information was defined by Massey in 1990.

$$I(X^n \rightarrow Y^n) \triangleq \sum_{i=1}^n I(X^i; Y_i | Y^{i-1})$$

Directed Information

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$$I(X^n; Y^n) = \sum_{i=1}^n I(X^{\textcolor{red}{n}}; Y_i | Y^{i-1})$$

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$$I(X^n \rightarrow Y^n) \triangleq \sum_{i=1}^n I(X^{\textcolor{red}{i}}; Y_i | Y^{i-1})$$

Intuition [Massey 05]:

$$I(X^n; Y^n) = I(X^n \rightarrow Y^n) + I(Y^{n-1} \rightarrow X^n)$$

Feedback Capacity of Unifilar, Strongly Connected, FSC

$$\begin{aligned} C_{FB} &= \lim_{N \rightarrow \infty} \frac{1}{N} \max_{\{p(x_t|x^{t-1}, y^{t-1})\}_{t=1}^N} I(X^N \rightarrow Y^N) \\ &= \lim_{N \rightarrow \infty} \frac{1}{N} \max_{\{p(x_t|x^{t-1}, y^{t-1})\}_{t=1}^N} \sum_{t=1}^N I(X^t; Y_t | Y^{t-1}) \\ &= \lim_{N \rightarrow \infty} \frac{1}{N} \max_{\{p(x_t|s_{t-1}, y^{t-1})\}_{t=1}^N} \sum_{t=1}^N I(\textcolor{red}{X}_t, \textcolor{red}{S}_{t-1}; Y_t | Y^{t-1}) \\ &= \sup_{\{p(x_t|s_{t-1}, y^{t-1})\}_{t \geq 1}} \liminf_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N I(X_t, S_{t-1}; Y_t | Y^{t-1}) \end{aligned}$$

Dynamic Programming (infinite horizon, average reward)

Variable Assignments

State: $\beta_t = p(s_t|y^t)$

Action: $u_t = p(x_t|s_{t-1})$

Disturbance: $w_t = y_{t-1}$

Dynamic Programming Requirements

State evolution:

$$\beta_t = F(\beta_{t-1}, u_t, w_t)$$

Reward function per unit time:

$$g(\beta_{t-1}, u_t) = I(X_t, S_{t-1}; Y_t | \beta_{t-1})$$

Similar dynamic programming approaches:

- Yang, Kavčič & Tatikonda - 2005
- Chen & Berger - 2005
- Mitter & Tatikonda - 2006

Dynamic Programming (infinite horizon, average reward)

Dynamic Programming Operator T

The dynamic programming operator T is given by

$$T \circ J(\beta) = \sup_{u \in \mathcal{U}} \left(g(\beta, u) + \int P_w(dw|\beta, u) J(F(\beta, u, w)) \right) .$$

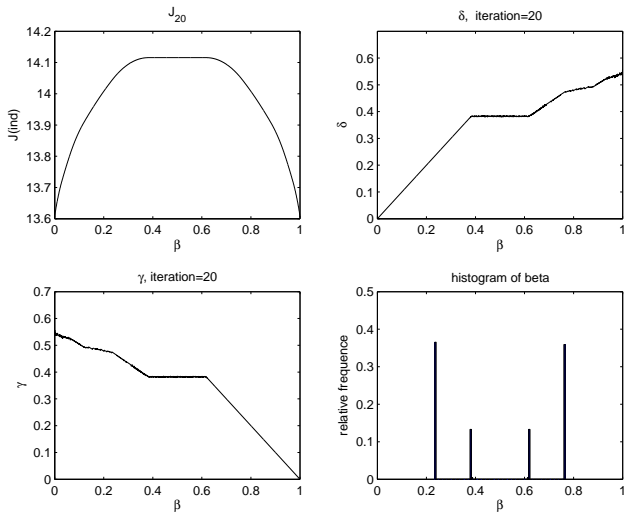
Bellman Equation

If there exist a function $J(\beta)$ and constant ρ that satisfy

$$J(\beta) = T \circ J(\beta) - \rho$$

then ρ is the optimal infinite horizon average reward.

Trapdoor Channel—20th Value iteration



Calculation and Coincidence

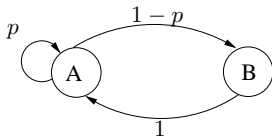
Trapdoor channel: $C_{FB} \approx 0.694$ bits

Calculation and Coincidence

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Homework Question

Entropy rate. Find the maximum entropy rate of the following two-state Markov chain:

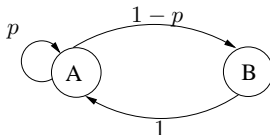


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Solution: (Golden Ratio: $\phi = \frac{\sqrt{5}+1}{2}$)

$$p^* = \phi - 1 = \frac{1}{\phi}$$

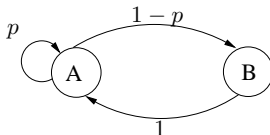
$$H(\mathcal{X}) = \log \phi = 0.6942... \text{ bits}$$

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Rename channel input: $\tilde{x}_t = x_t \oplus s_{t-1}$.

Rename channel output: $\tilde{y}_t = y_t \oplus y_{t-1}$.

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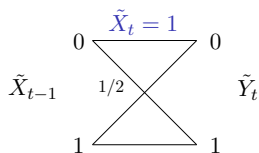
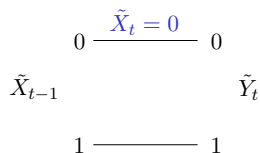
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Case 2: $\tilde{x}_t = 1$

$$\tilde{x}_t = 1 \quad \Rightarrow \quad \tilde{y}_t \sim \text{Bern}(1/2), \text{ independent of the past.}$$

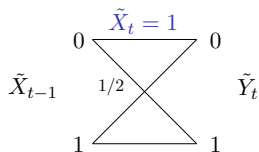
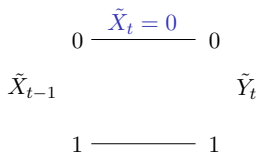
Equivalent Channel

- Two states specify whether the channel is all noise or noise-free.

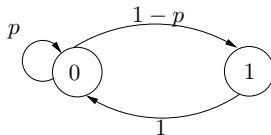


Equivalent Channel

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- Use Markov input



Decoding Example

- Transmitter promises to never use $\tilde{x} = 1$ twice in a row.
- The number of such sequences of length n is the Fibonacci sequence.

Decoding rules

$$\tilde{x}_t = 0 \quad \Rightarrow \quad \tilde{x}_{t-1} = \tilde{y}_t$$

$$\tilde{x}_t = 1 \quad \Rightarrow \quad \tilde{x}_{t-1} = 0$$

$$\begin{array}{l} \tilde{x}^n: \quad \quad \quad 0 \\ \tilde{y}^n: 1 \ 1 \ 1 \ 1 \ 0 \ 0 \ 1 \end{array}$$

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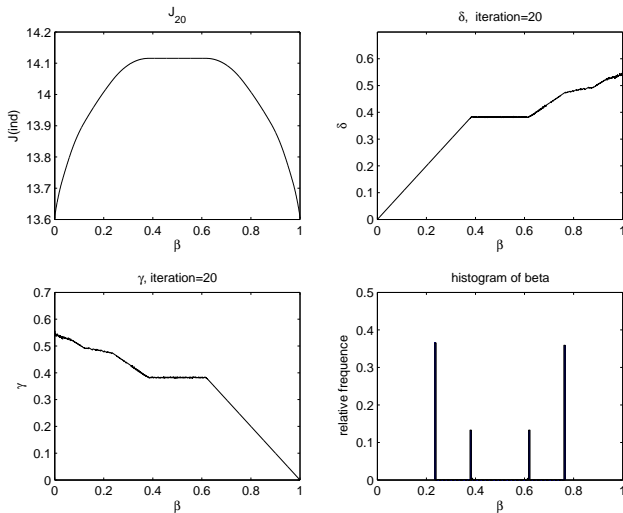
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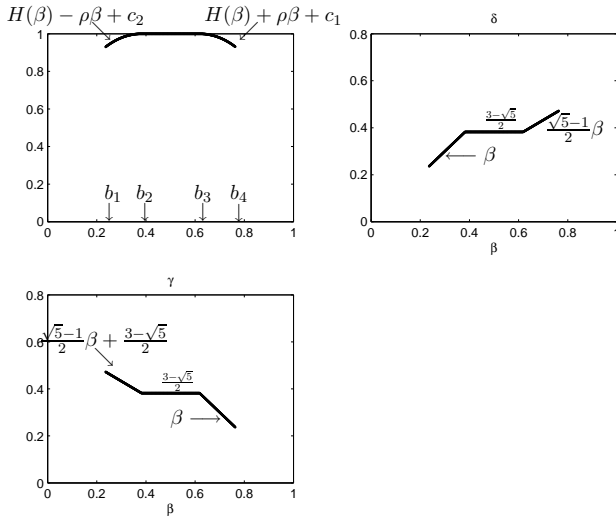
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Dynamic Programming 20th Value iteration



$$\text{Bellman Equation: } J(\beta) = T \circ J(\beta) - \rho.$$

Conjectured Solution to the Bellman Equation



Bellman Equation: $J(\beta) = T \circ J(\beta) - \rho.$

Feedback Capacity and Zero-Error Capacity

Bellman Equation is satisfied.

Trapdoor channel feedback capacity:

$$C_{FB} = \log \phi = 0.6942... \text{ bits.}$$

$$C \approx .572 \text{ bits.}$$

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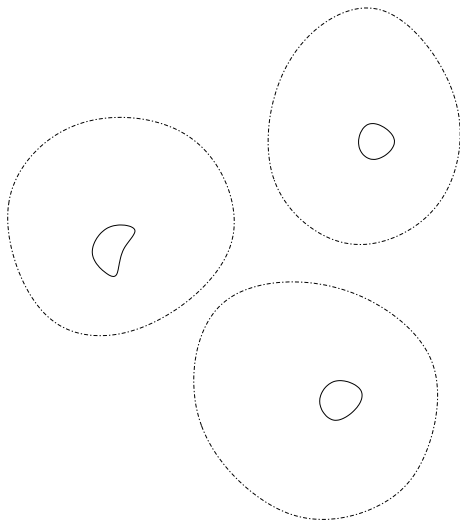
$$C_{FB} = \log \phi.$$

$$C = .5 \text{ bits.}$$

[Ahlsvede & Kaspi 87]

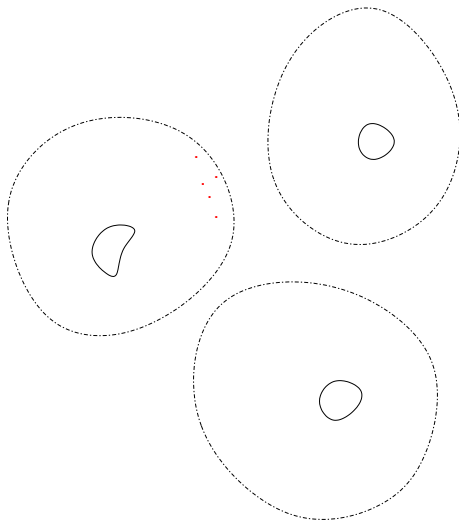
Biological Cells

Biochemical interpretation of the trapdoor channel [Berger 71]



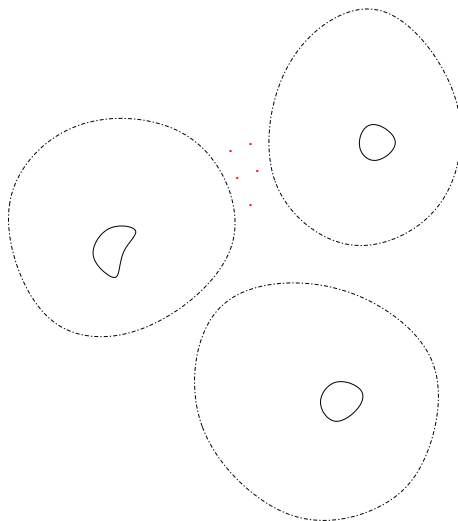
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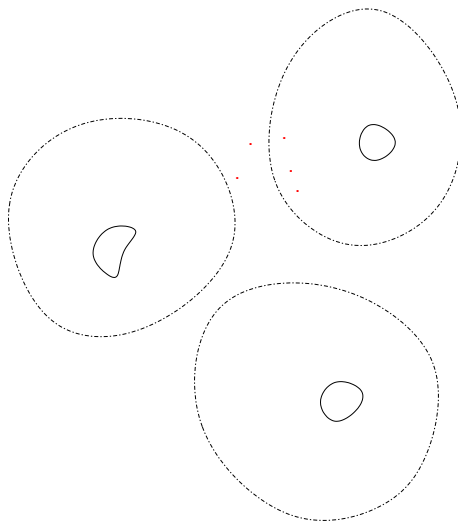
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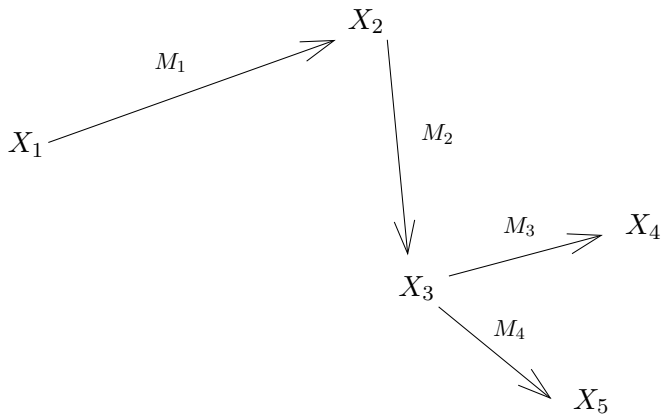


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Coordinated Action in a Network

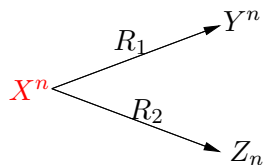


Task Assignment

Computation Tasks: $T = \{0, 1, 2\}$

Processors: X , Y , and Z

$X^n \sim iid, \text{ Unif}(\{0, 1, 2\})$



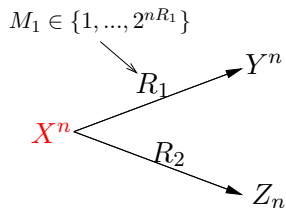
What are the required rates (R_1, R_2) ?

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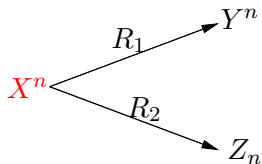
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Task Assignment: Achievable Rates

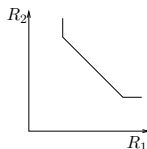
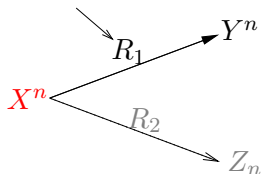


Idea 1—Describe X^n :

$$R_1 = R_2 = H(X) = \log(3).$$

A protocol tells Y and Z how to orient around X .

Task Assignment: Achievable Rates

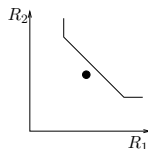
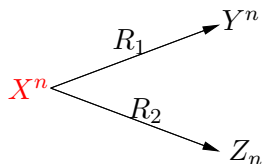


Idea 2—Minimize R_1 :

$$\begin{aligned} R_1 &\geq \min_{p(y|x): X \neq Y} I(X; Y) \\ &= H(X) - \max_{p(y|x): X \neq Y} H(Y|X) \\ &= \log 3 - \log 2. \end{aligned}$$

Describe Z^n at full rate: $R_2 = \log 3$.

Task Assignment: Achievable Rates



Idea 3—Restrict the support of Y and Z :

$$Y \in \{0, 1\}$$

$$Z \in \{0, 2\}.$$

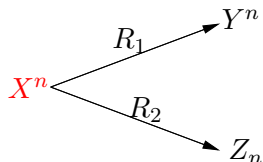
By default $Y = 1$ and $Z = 2$.

The encoder tells them when to move out of the way.

$$R_1 \geq H(Y) = H(1/3) = \log 3 - 2/3 \text{ bits},$$

$$R_2 \geq H(Z) = H(1/3) = \log 3 - 2/3 \text{ bits}.$$

Task Assignment: Achievable Rates



Achievable rates for general joint distribution $p(x, y, z)$:

$$R_1 \geq I(X; Y, U),$$

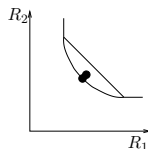
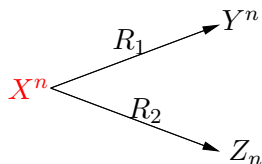
$$R_2 \geq I(X; Z, U),$$

$$R_1 + R_2 \geq I(X; Y, U) + I(X; Z, U) + I(Y; Z|X, U).$$

for some U .

[Cover, Permuter 07], [Zhang, Berger 95]

Task Assignment: Achievable Rates



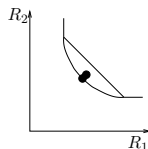
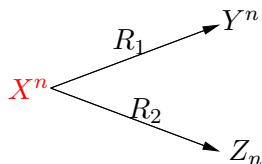
Idea 4—Restrict the support of Y and Z around \hat{X} :

First send \hat{X} to both Y and Z , where $\hat{X} = X$ with probability $\frac{\alpha}{\alpha+2}$.

$$p(\hat{x}|x=1)$$

$\frac{1}{\alpha+2}$	$\frac{\alpha}{\alpha+2}$	$\frac{1}{\alpha+2}$
0	1	2

Task Assignment: Achievable Rates



Idea 4—Restrict the support of Y and Z around \hat{X} :

$$Y \in \{\hat{X}, \hat{X} + 1\}$$

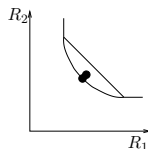
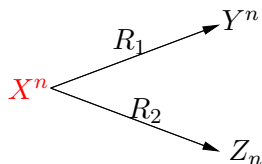
$$Z \in \{\hat{X}, \hat{X} + 2\}.$$

The encoder tells them when to move out of the way (onto \hat{X}).

$$R_1 \geq I(X; \hat{X}) + H(Y|\hat{X}),$$

$$R_2 \geq I(X; \hat{X}) + H(Z|\hat{X}).$$

Task Assignment: Achievable Rates



Idea 4—Restrict the support of Y and Z around \hat{X} :

Optimize α (compression rate of \hat{X}):

$$\alpha^* = \phi = \frac{\sqrt{5} + 1}{2}.$$

Resulting rates:

$$R_1 = R_2 = \log 3 - \log \phi.$$

Summary

1 Blackwell's Trapdoor Channel

- ▶ $C_{FB} = \log \phi$.
- ▶ Transform to equivalent channel
- ▶ Simple zero-error communication scheme
- ▶ Guess and check solution to Bellman equation

2 Three-node task assignment

- ▶ Achievable symmetric rate: $H(X) - \log \phi$
- ▶ Two phases of communication

