

# The Capacity of the Trapdoor Channel with Feedback

Haim Permuter

Based on work with

Paul Cuff, Benjamin Van Roy and Tsachy Weissman

Stanford University

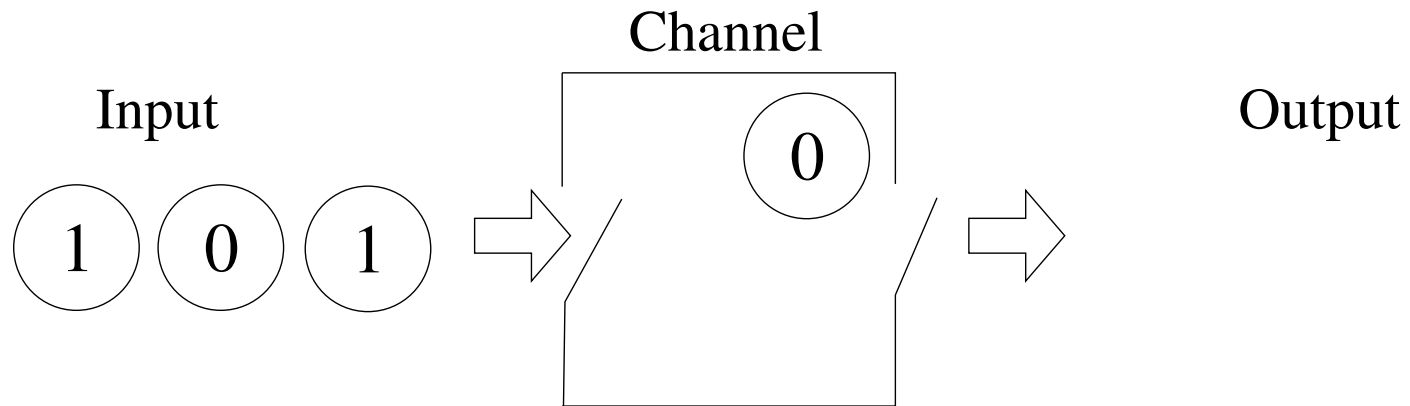
# Main Results of the Talk

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1. **capacity** of the trapdoor channel with feedback
2. **simple scheme** that achieves feedback capacity

# The trapdoor channel

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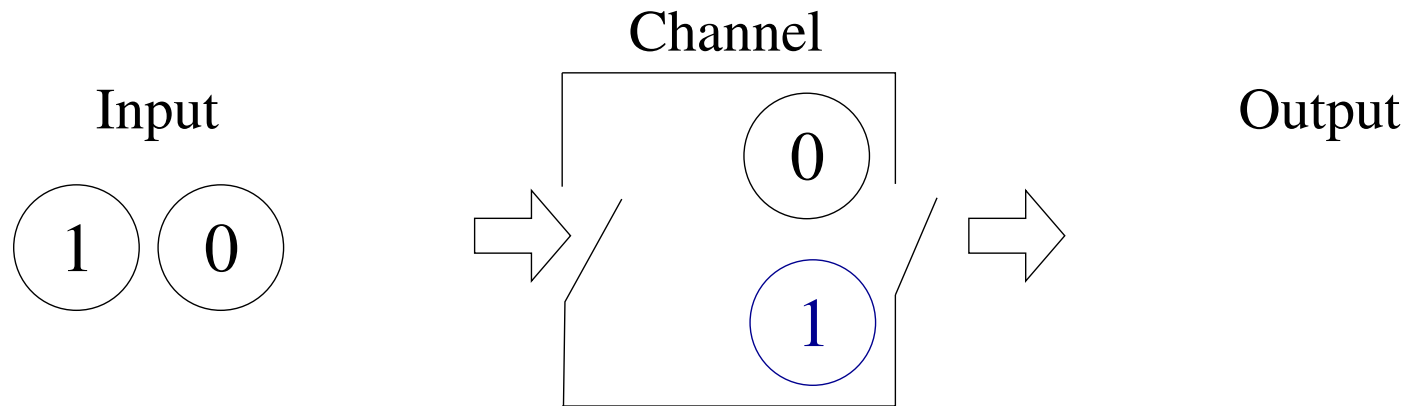


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# The trapdoor channel

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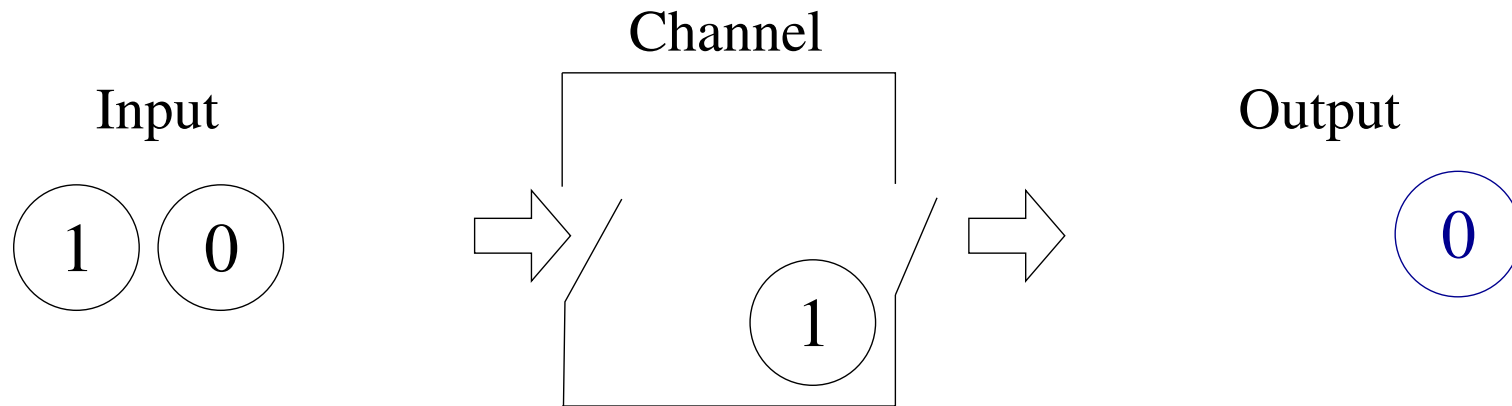


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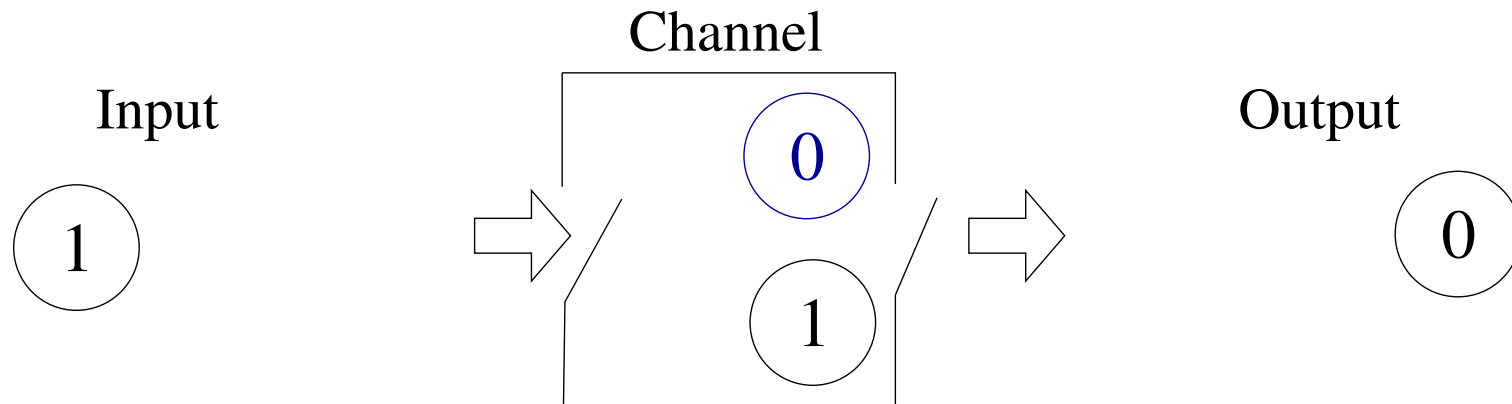
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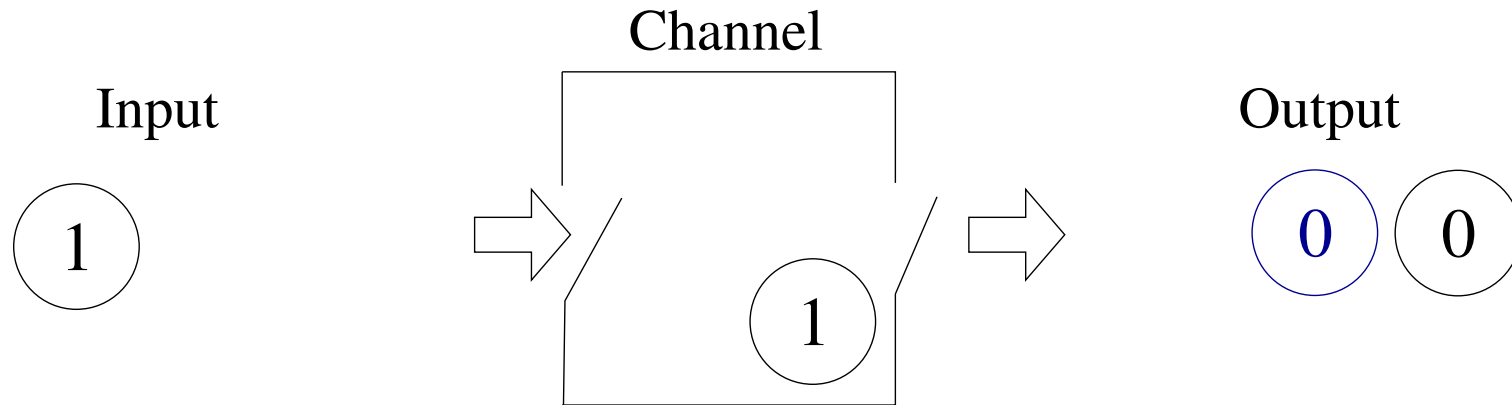
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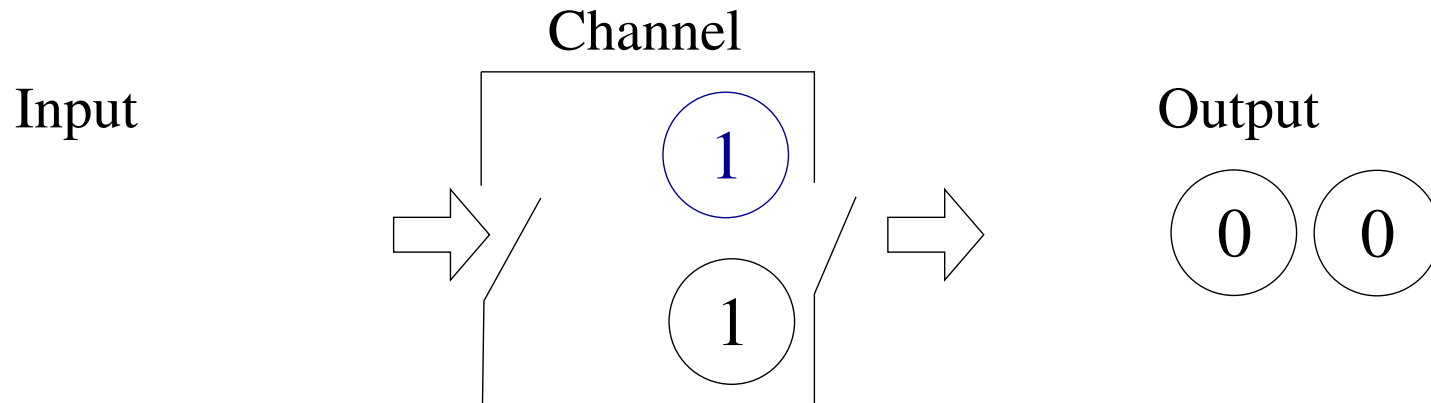
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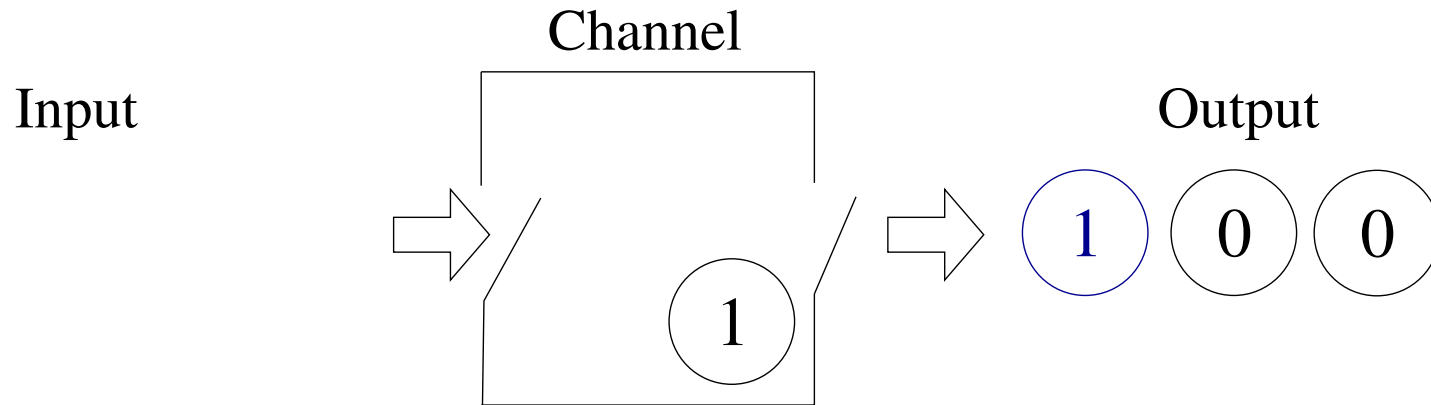
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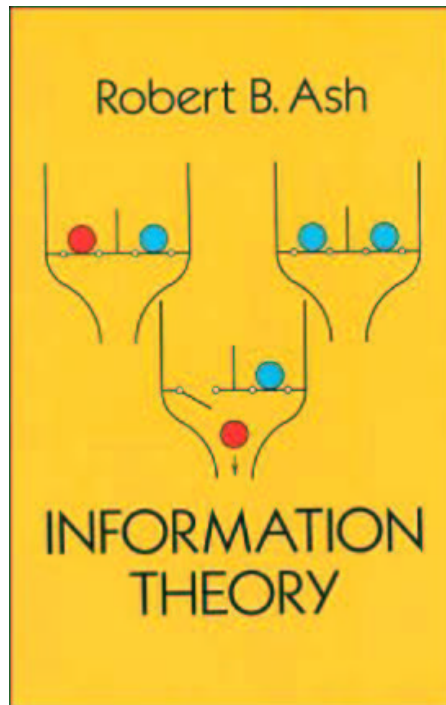
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# The trapdoor channel

Introduced by David Blackwell in 1961. [Ash65], [Ahlsvede & Kaspi 87], [Ahlsvede 98], [Kobayashi 02].



(a) Ash book

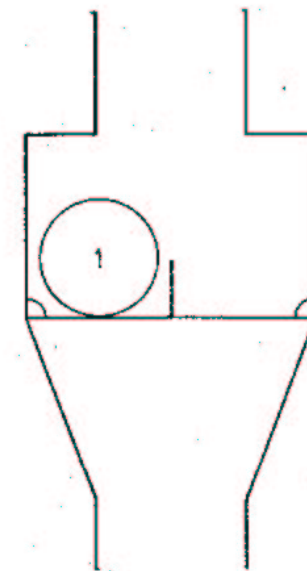


Fig. 7.1 A simple two-state channel.

(b) D. Blackwell

Another appropriate name for this channel is *chemical channel*.

# Communication setting

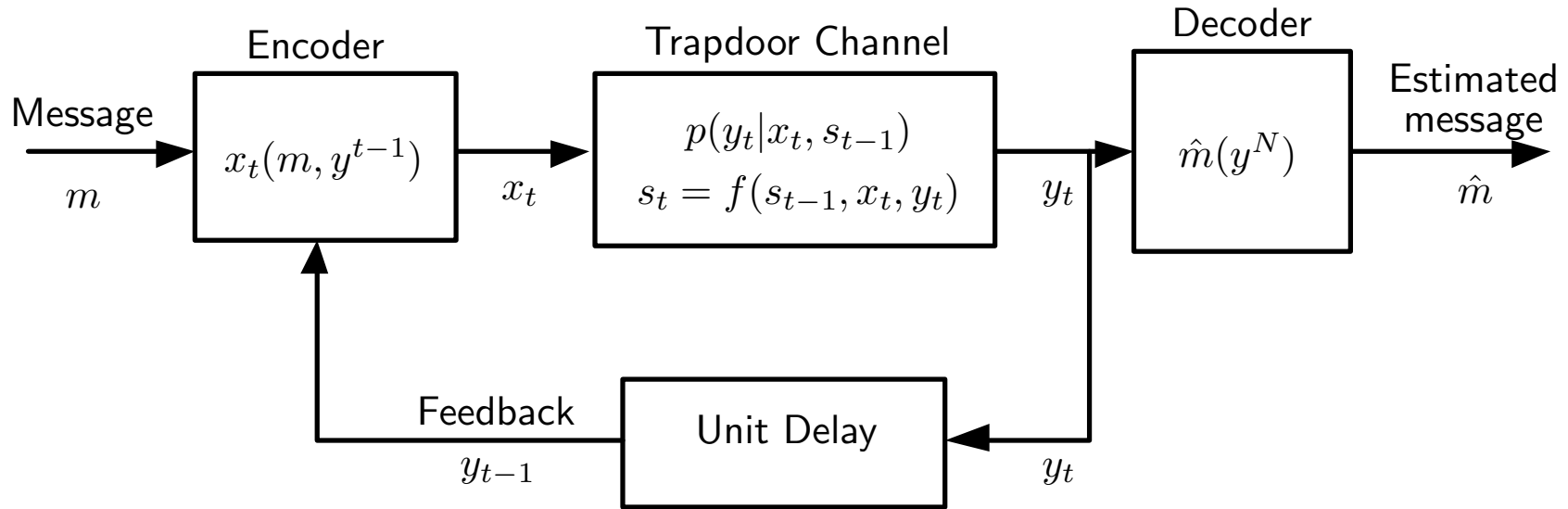


Figure 1: Unifilar FSC with feedback

Finite State Channel(FSC) property:  $p(y_i, s_i|x^i, s^{i-1}, y^{i-1}) = p(y_i, s_i|x_i, s_{i-1})$

Unifilar channel [Ziv85]:  $s_t = f(s_{t-1}, x_t, y_t)$

# Main ingredients

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1. Directed information.
2. Dynamic program average-reward.
3. Value iteration.
4. Bellman equation.
5. Homework question given by Tom Cover.

# Feedback capacity of FSC

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Lower and Upper bound

$$C_{FB} \geq \lim_{N \rightarrow \infty} \frac{1}{N} \max_{\{p(x_i|x^{i-1}, y^{i-1})\}_{i=1}^N} \min_{s_0} I(X^N \rightarrow Y^N | s_0)$$

$$C_{FB} \leq \lim_{N \rightarrow \infty} \frac{1}{N} \max_{\{p(x_i|x^{i-1}, y^{i-1})\}_{i=1}^N} \max_{s_0} I(X^N \rightarrow Y^N | s_0)$$

[Permuter, Weissman and Goldsmith ISIT06]

where

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

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In the trapdoor channel any state  $s_t$  can be reached from any state  $s_{t-1}$  with positive probability and hence we get

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

# Directed information

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*Directed Information* was defined by Massey in 1990,

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

# Directed information - intuition

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If there is no feedback

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Deterministic feedback  $k_i(y_i)$

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

# Feedback capacity

---

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

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# Feedback capacity and dynamic programming(DP)

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DP consists of of states  $\beta_{t-1}$ , actions  $u_t(\beta_{t-1})$ , and disturbance  $w_t$ .

state:

$$\beta_{t-1} = p(s_{t-1}|y^{t-1}), \quad \beta \in [0, 1]$$

action:

$$u_t = p(x_t|s_{t-1}), \quad u_t \in [0, 1] \times [0, 1]$$

disturbance:

$$w_t = y_{t-1},$$

$$\beta_t = F(\beta_{t-1}, u_t, w_t), \quad t = 1, 2, 3, \dots,$$

reward function per unit time

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

[Tatikonda00], [Yang, Kavčić and Tatikonda05]

# Dynamic programming operator, $T$

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The dynamic programming operator  $T$  is given by

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

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$$(TJ)(\beta) = \sup_{0 \leq \delta \leq \beta, 0 \leq \gamma \leq 1-\beta} \left( H \left( \frac{1}{2} + \frac{\delta - \gamma}{2} \right) + \delta + \gamma - 1 + \frac{1 + \delta - \gamma}{2} J \left( \frac{2\delta}{1 + \delta - \gamma} \right) \right. \\ \left. + \frac{1 - \delta + \gamma}{2} J \left( 1 - \frac{2\gamma}{1 - \delta + \gamma} \right) \right)$$

Properties

- Preservation of *concavity*: if  $J$  is concave then  $TJ$  is concave.
- Preservation of *continuity*: if  $J$  is continuous then  $TJ$  is continuous.
- Preservation of *symmetry*: if  $J$  is symmetric then  $TJ$  is symmetric.

# Computational study

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Executed 20 value iterations:  $J_{k+1}(\beta) = (T J_k)(\beta)$

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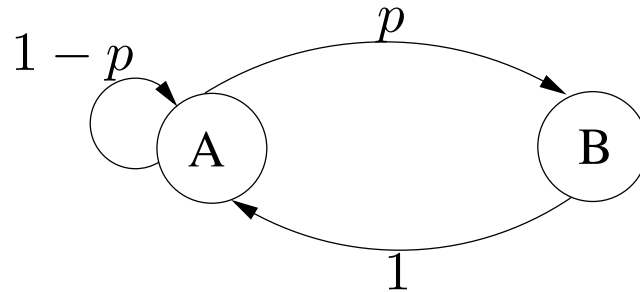
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HW question from Prof. Cover class

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



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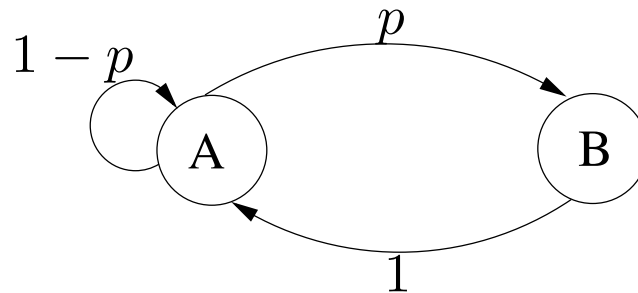
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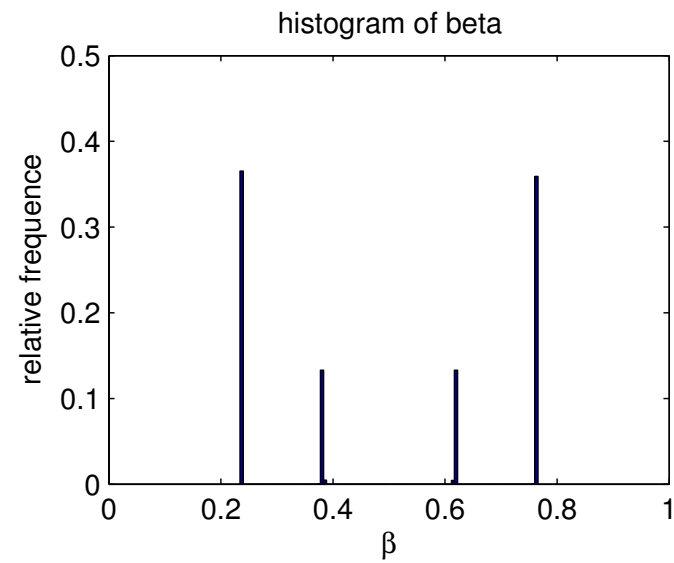
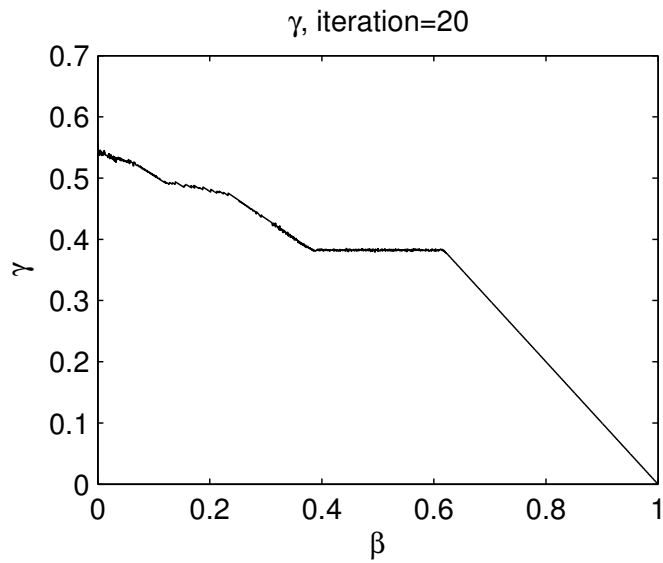
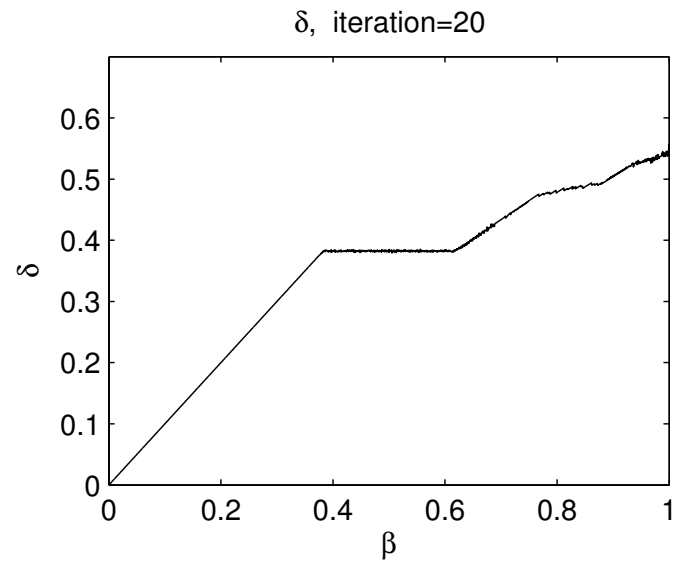
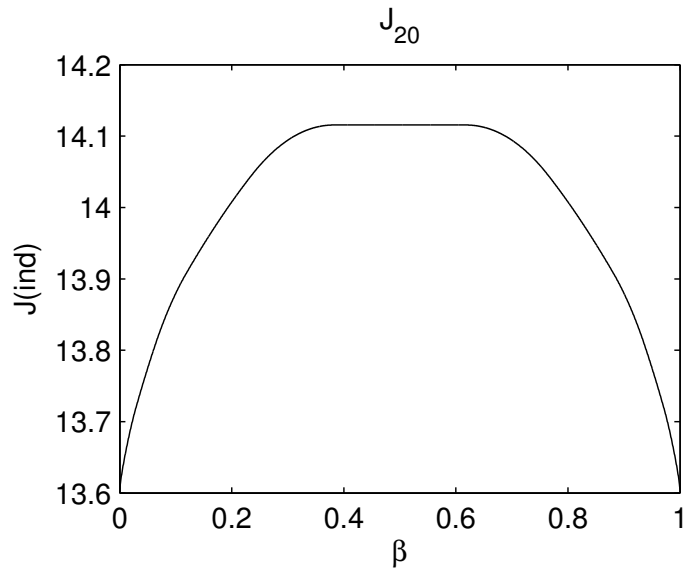
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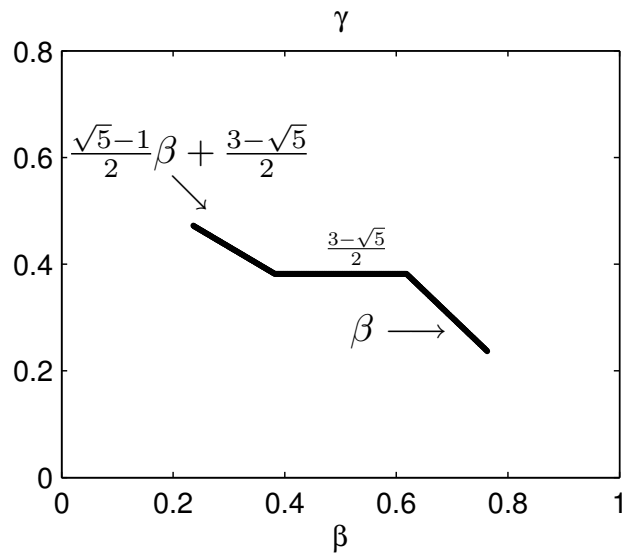
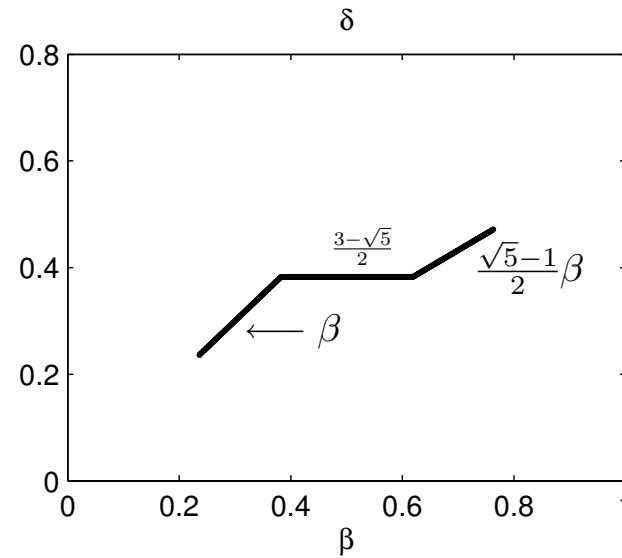
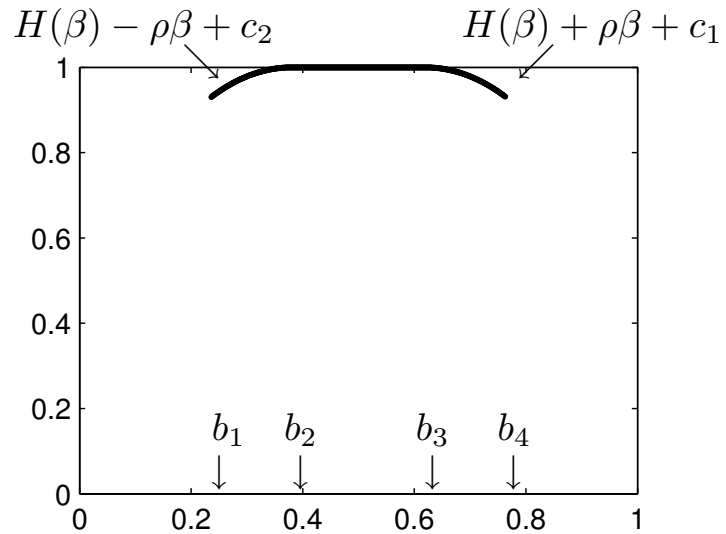
*Solution:* The entropy rate is  $\log_2 \phi = 0.6942\dots$ , where  $\phi$  is the *golden ratio*:  
 $\phi = \frac{\sqrt{5}+1}{2}$ .

# 20th Value iteration

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# Conjecture of the solution to Bellman equation



# Bellman equation

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**Theorem 1.** *If there exists  $(J(\beta), \rho)$  that satisfies*

$$J(\beta) = (TJ)(\beta) - \rho,$$

*then  $\rho$  is the optimal average reward.*

# Verifying our conjecture

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Construct *value iteration function*  $J_k(\beta)$  as follows. Let  $J_0(\beta)$  be the pointwise maximum among concave functions satisfying  $J_0(\beta) = \tilde{J}(\beta)$  for  $\beta \in [b_1, b_4]$

$$J_{k+1}(\beta) = (TJ_k)(\beta) - \tilde{\rho},$$

- concave, continuous and symmetric
- fixed point: for  $\beta \in [b_1, b_4]$ ,  $J_k(\beta) = \tilde{J}(\beta)$
- monotonically nonincreasing in  $k$
- converges uniformly to  $J^*(\beta)$

Since the sequence  $J_{k+1} = TJ_k - \tilde{\rho}\mathbf{1}$  converges uniformly and  $T$  is sup-norm continuous,  $J^* = TJ^* - \tilde{\rho}\mathbf{1}$ .

# A scheme that achieves capacity

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## Question

*Number of sequences.* To first order in the exponent, what is the number of binary sequences of length  $n$  with no two consecutive 1's?

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*Solution* The number of sequences of length  $n$  with this property, is the  $n^{\text{th}}$  Fibonacci number,  $f_n \doteq \phi^n$ .

## The scheme

Let us denote such a sequence by  $r^n$ . Map each message  $m$  to a sequence  $[r^n(m)]$ .

*encoder:*  $x_t = s_{t-1} \oplus r_t, t = 1, \dots, n$  and  $x_{n+1} = s_n$ .

*decoder:* The decoder can decode this sequence error-free!

# Conclusions

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*Thank You!*