

# Capacity of a POST Channel with and without Feedback

Workshop On Coding and Information Theory

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based on joint work with:  
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# directed information

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

# directed information

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

compare to:

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

## *Directed Information*

[Massey90] inspired by [Marko 73]

$$I(X^n \rightarrow Y^n) \triangleq H(Y^n) - H(Y^n || X^n)$$

$$I(X^n; Y^n) \triangleq H(Y^n) - H(Y^n | X^n)$$

## *Causal Conditioning*

[Kramer98]

$$H(Y^n || X^n) \triangleq E[-\log P(Y^n || X^n)]$$

$$H(Y^n | X^n) \triangleq E[-\log P(Y^n | X^n)]$$

$$P(y^n || x^n) \triangleq \prod_{i=1}^n P(y_i | x^i, y^{i-1})$$

$$P(y^n || x^{n-1}) \triangleq \prod_{i=1}^n P(y_i | x^{i-1}, y^{i-1})$$

# causal conditioning

$$P(y^n || x^n) \triangleq \prod_{i=1}^n P(y_i | x^i, y^{i-1})$$

$$P(y^n || x^{n-1}) \triangleq \prod_{i=1}^n P(y_i | x^{i-1}, y^{i-1})$$

$$p(x^n, y^n) = p(x^n || y^{n-1})p(y^n || x^n)$$

# why directed information?

consider, e.g.:

BSC(1/2)

$$X_i = Y_{i-1}$$

$$I(X^n; Y^n) = ?$$

$$I(X^n \rightarrow Y^n) = ?$$

# optimization

$$\max_{p(x^n || y^{n-1})} I(X^n \rightarrow Y^n)$$

note:

- concavity of the function
- convexity of the set

# on capacity

(under conditions)

$$C = \lim_{n \rightarrow \infty} \max \frac{1}{n} I(X^n \rightarrow Y^n)$$

[Massey 1990], [Kramer 1998], [Chen and Berger 2005],  
[Tatikonda and Mitter 2010], [Kim 2010], [Permuter, Goldsmith and W. 2010]



# Finite State Channels

$$P(y_i, s_i | x^i, s^{i-1}, y^{i-1}) = P(y_i, s_i | x_i, s_{i-1})$$

# explicit computations

- memoryless channels
- mod-additive channels
- Gaussian with and without FB
- trapdoor with FB
- Ising with FB
- some more

POST channel  
Previous Output is the State

$$S_i = Y_{i-1}$$

$\Leftrightarrow$

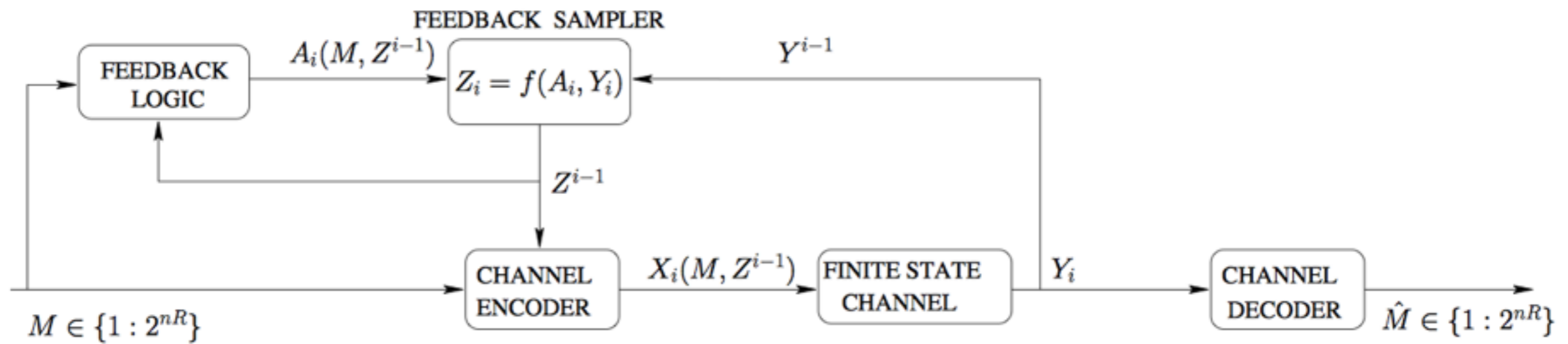
$$p(y_i | x^i, y^{i-1}) = p(y_i | x_i, y_{i-1})$$

[ T. Berger, 2002 Shannon lecture  
“living information theory” ]

# motivation

- simple
- good model
- to feed or not

## “To Feed or Not to Feed Back”



# questions for the POST channel

- $C_{FB}$
- $C_{NFB}$
- $C_{FB} > C_{NFB}$ ?

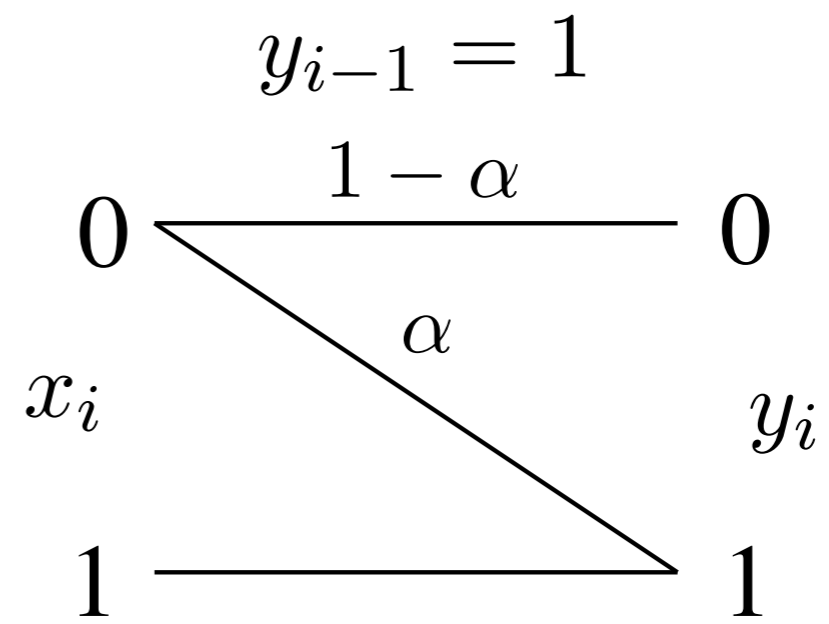
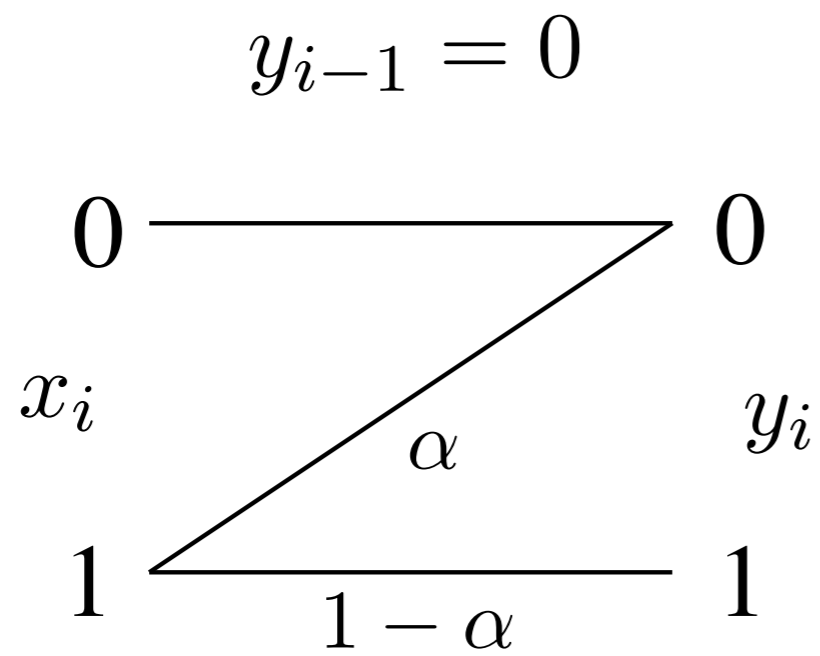
# feedback capacity of the POST channel

[Chen and Berger 2005]:

$$C_{FB} = \max_{p(x_1|y_0)} I(X_1; Y_1 | Y_0)$$

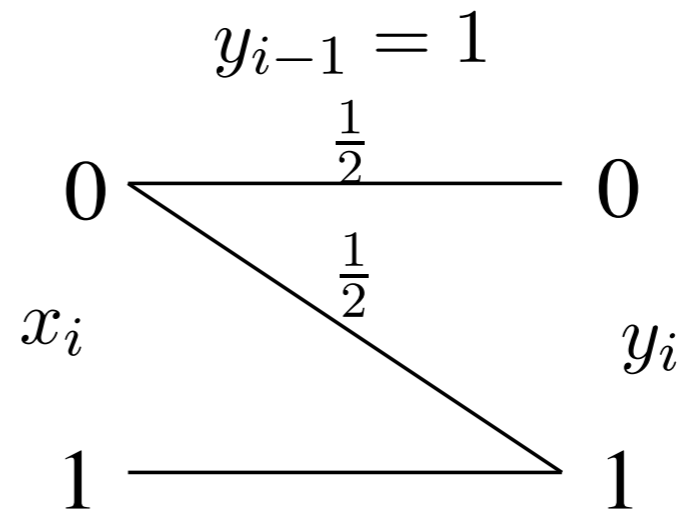
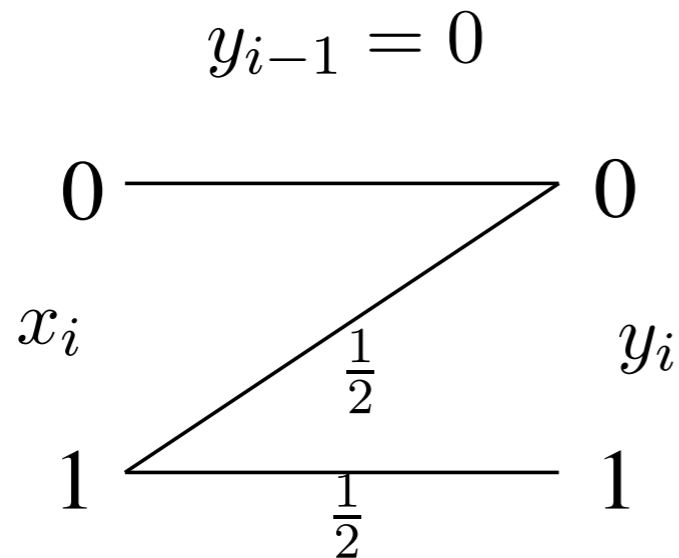
(under benign conditions)

# POST( $\alpha$ ) channel





(simple) **POST**( $\alpha = \frac{1}{2}$ )



alternatively:

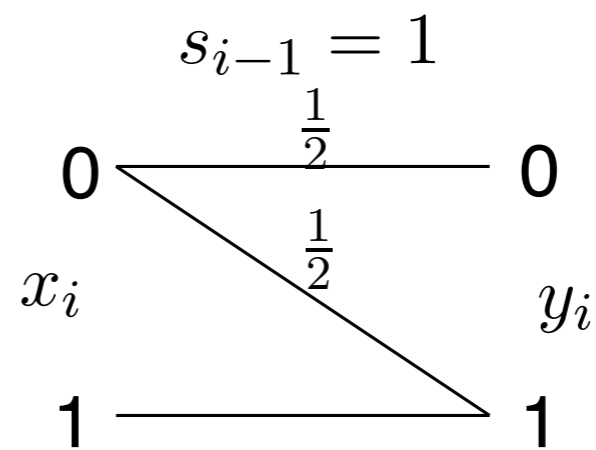
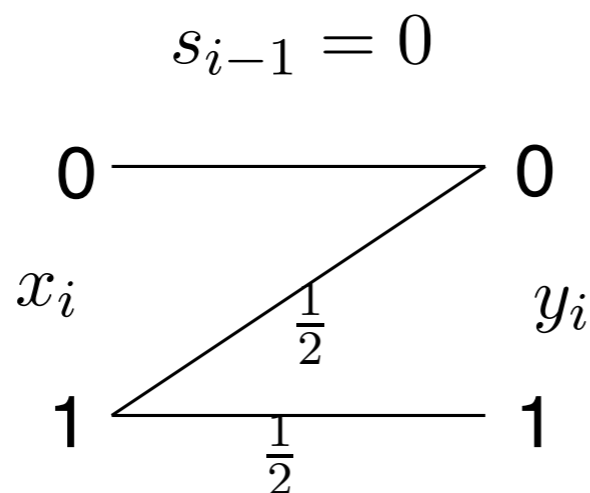
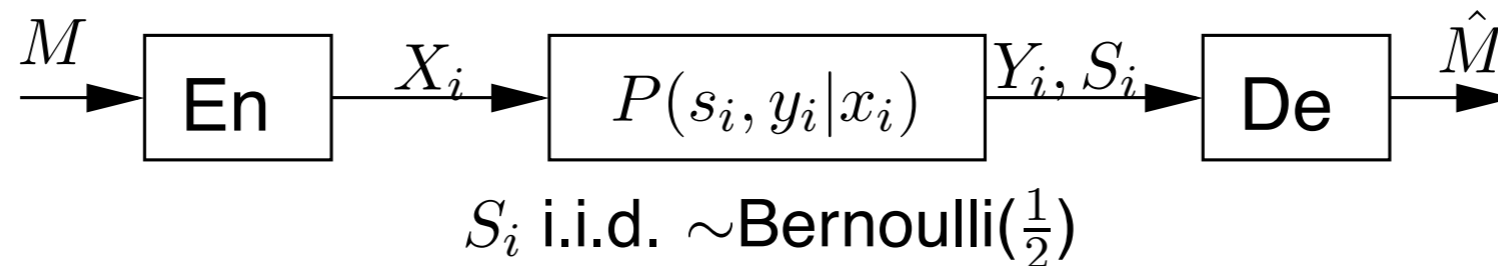
if  $X_i = Y_{i-1}$ ,

otherwise,

$$Y_i = X_i$$

$$Y_i \sim \text{Bernouli}\left(\frac{1}{2}\right)$$

# intuition?



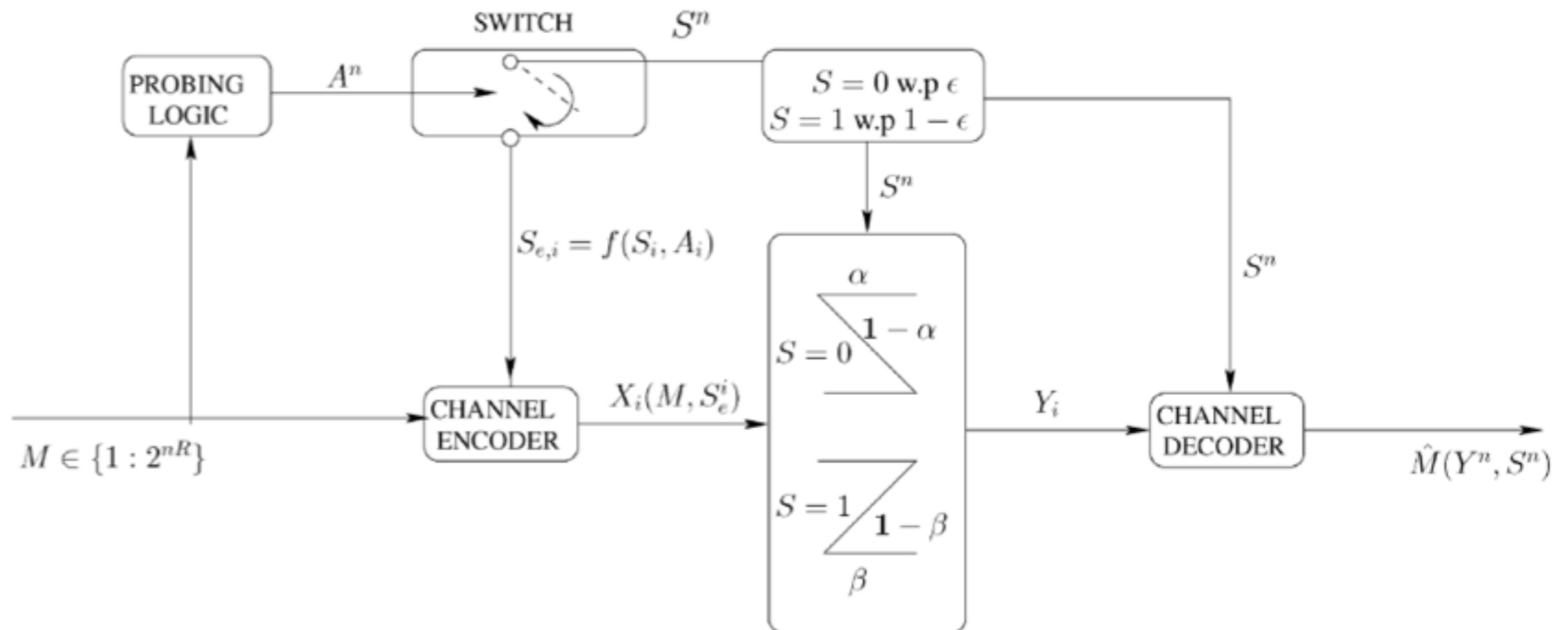
- Regular capacity

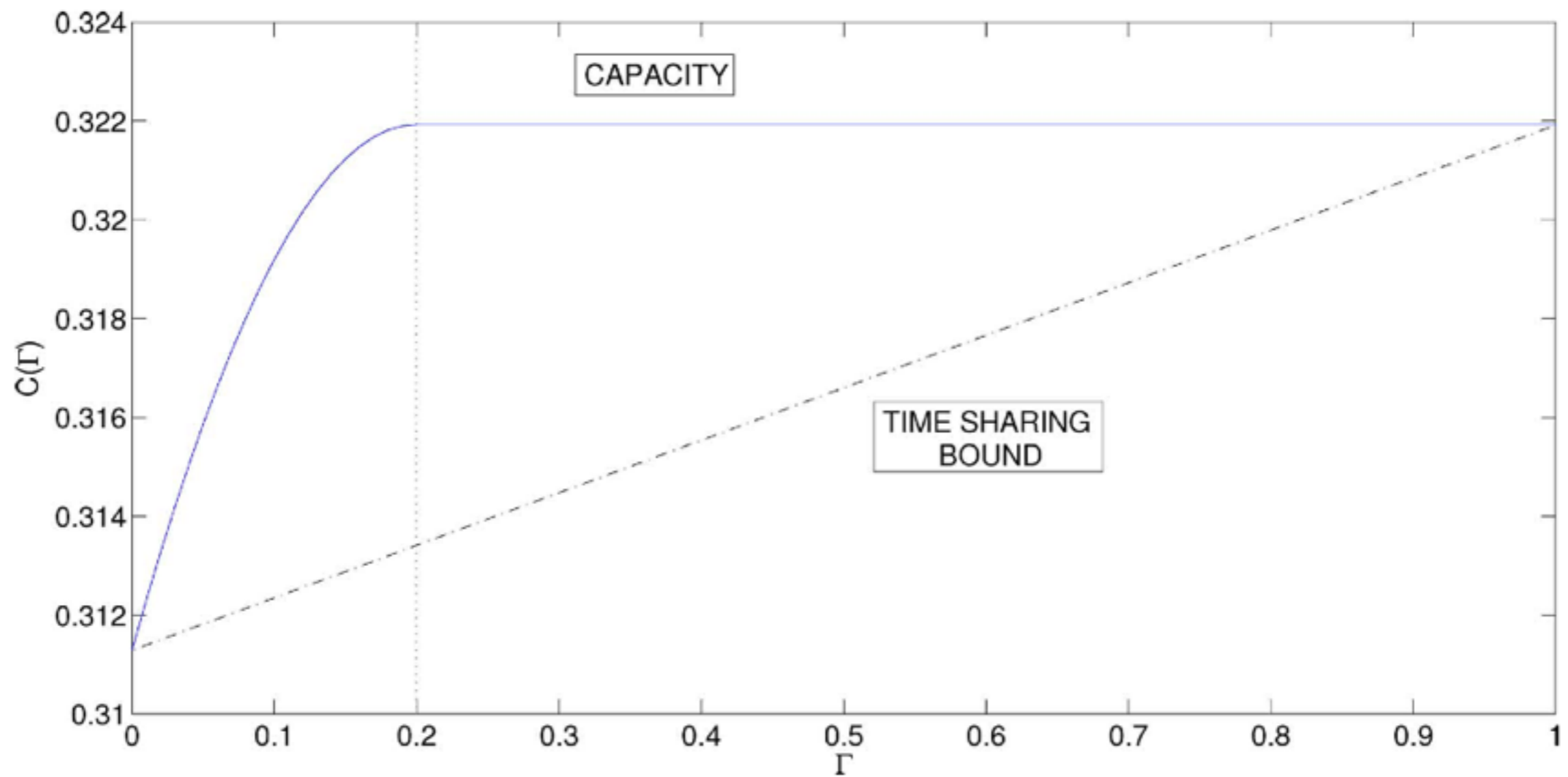
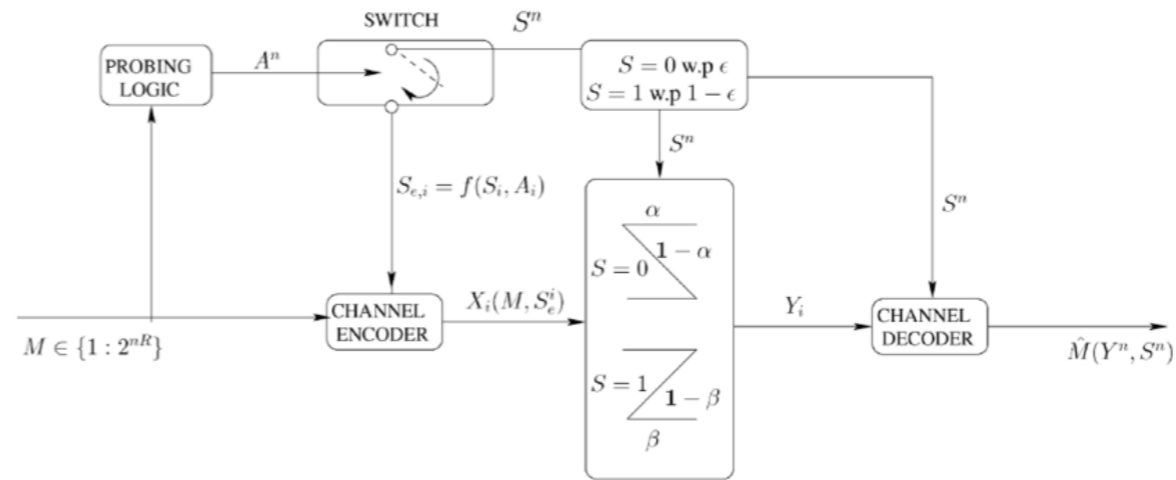
$$C = \max_{P(x)} I(X; Y, S) = H_b\left(\frac{1}{4}\right) - \frac{1}{2} = 0.3111$$

- Feedback capacity is the capacity of the  $Z$  channel

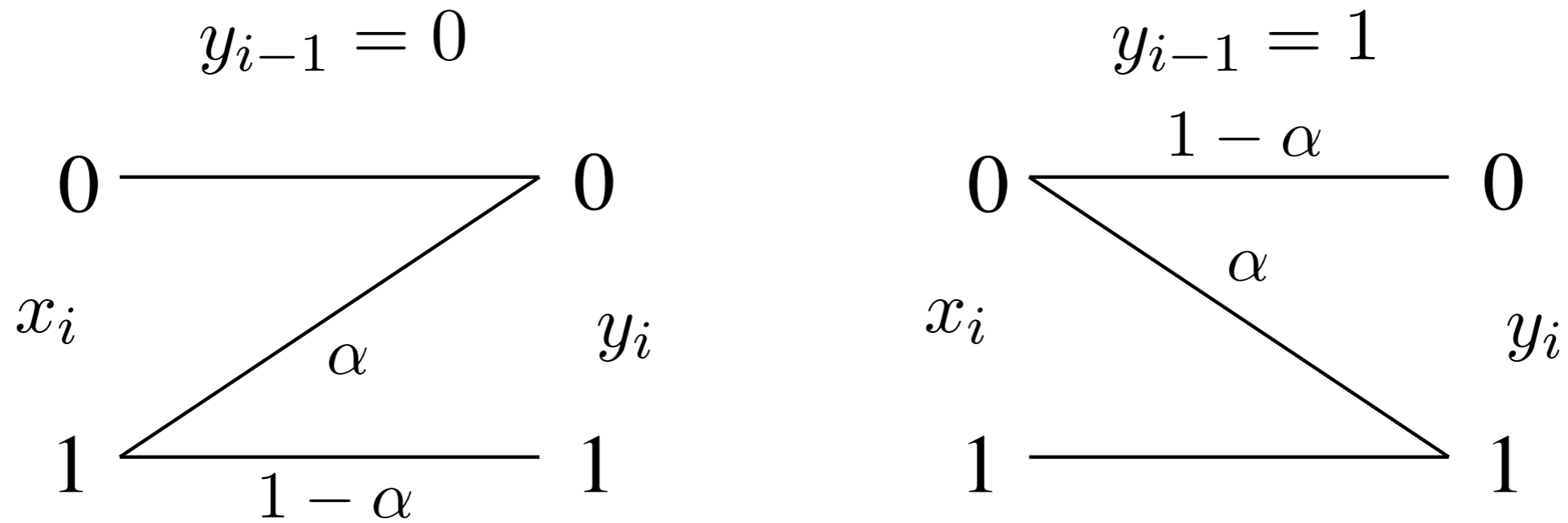
$$C_{fb} = -\log_2 0.8 = 0.3219$$

# channel probing ([Asnani, Permuter and W. 2011])





# POST( $\alpha$ ) channel



- $C_{FB}$
- $C_{NFB}$
- $C_{FB} > C_{NFB}$ ?

## Theorem

Feedback does not increase the capacity of the  $\text{POST}(\alpha)$  channel.

# main idea

show:

$$\max_{P(x^n || y^{n-1})} I(X^n \rightarrow Y^n) = \max_{P(x^n)} I(X^n \rightarrow Y^n)$$

# Necessary and sufficient for $\max I(X^n \rightarrow Y^n)$

## Theorem

A set of necessary and sufficient conditions for an input probability  $P(x^n || y^{n-1})$  to maximize  $I(X^n \rightarrow Y^n)$  is that for some numbers  $\beta_{y^{n-1}}$

$$\sum_{y^n} p(y^n || x^n) \log \frac{p(y^n || x^n)}{ep(y^n)} = \beta_{y^{n-1}}, \quad \forall x^n, y^{n-1}, \text{ if } p(x^n || y^{n-1}) > 0$$

$$\sum_{y^n} p(y^n || x^n) \log \frac{p(y^n || x^n)}{ep(y^n)} \leq \beta_{y^{n-1}}, \quad \forall x^n, y^{n-1}, \text{ if } p(x^n || y^{n-1}) = 0$$

where  $p(y^n) = \sum_{x^n} p(y^n || x^n) p(x^n || y^{n-1})$ . The solution of the optimization is

$$\max_{P(x^n || y^{n-1})} I(X^n \rightarrow Y^n) = \sum_{y^{n-1}} \beta_{y^{n-1}} + 1.$$



# the case $n=1$

[Gallager 1968]

## 4.5 Finding Channel Capacity for a Discrete Memoryless Channel

**Theorem 4.5.1.** A set of necessary and sufficient conditions on an input probability vector  $\mathbf{Q} = [Q(0), \dots, Q(K-1)]$  to achieve capacity on a discrete memoryless channel with transition probabilities  $P(j|k)$  is that for some number  $C$ ,

$$I(x = k; Y) = C; \quad \text{all } k \text{ with } Q(k) > 0 \quad (4.5.1)$$

$$I(x = k; Y) \leq C; \quad \text{all } k \text{ with } Q(k) = 0 \quad (4.5.2)$$

in which  $I(x = k; Y)$  is the mutual information for input  $k$  averaged over the outputs,

$$I(x = k; Y) = \sum_j P(j|k) \log \frac{P(j|k)}{\sum_i Q(i)P(j|i)} \quad (4.5.3)$$

# Necessary and sufficient for $\max I(X^n \rightarrow Y^n)$

## Theorem

A set of necessary and sufficient conditions for an input probability  $P(x^n || y^{n-1})$  to maximize  $I(X^n \rightarrow Y^n)$  is that for some numbers  $\beta_{y^{n-1}}$

$$\sum_{y_n} p(y^n || x^n) \log \frac{p(y^n || x^n)}{ep(y^n)} = \beta_{y^{n-1}}, \quad \forall x^n, y^{n-1}, \text{ if } p(x^n || y^{n-1}) > 0$$

$$\sum_{y_n} p(y^n || x^n) \log \frac{p(y^n || x^n)}{ep(y^n)} \leq \beta_{y^{n-1}}, \quad \forall x^n, y^{n-1}, \text{ if } p(x^n || y^{n-1}) = 0$$

where  $p(y^n) = \sum_{x^n} p(y^n || x^n) p(x^n || y^{n-1})$ . The solution of the optimization is

$$\max_{P(x^n || y^{n-1})} I(X^n \rightarrow Y^n) = \sum_{y^{n-1}} \beta_{y^{n-1}} + 1.$$

# key tool

## Corollary

Let  $P^*(x^n || y^{n-1})$  achieve the maximum of  $\max_{P(x^n || y^{n-1})} I(X^n \rightarrow Y^n)$  and let  $P^*(y^n)$  be the induced output pmf. If there exists an input probability distribution  $P(x^n)$  such that

$$p^*(y^n) = \sum_{x^n} p(y^n || x^n) p(x^n),$$

then

$$\max_{P(x^n || y^{n-1})} I(X^n \rightarrow Y^n) = \max_{P(x^n)} I(X^n \rightarrow Y^n)$$

# route for showing $C_{NFB} = C_{FB}$

Find:

- 1  $p^*(y^n)$
- 2  $P_n^{-1}$  (where  $P_n(y^n, x^n) = P(y^n|x^n)$ )
- 3  $P_n^{-1} \cdot p^*(y^n)$

is  $P_n^{-1} \cdot p^*(y^n) \geq 0$ ?

if yes  $\forall n \Rightarrow C_{NFB} = C_{FB}$

# specifically, for POST

- simple structure and evolution of:
  - optimal output distribution
  - channel matrix
  - its inverse

## Simple POST channel

Binary symmetric Markov  $\{Y\}_{i \geq 1}$  with transition probability 0.2 can be described recursively

$$P_0(y^n) = \begin{bmatrix} 0.8P_0(y^{n-1}) \\ 0.2P_1(y^{n-1}) \end{bmatrix} \quad P_1(y^n) = \begin{bmatrix} 0.2P_0(y^{n-1}) \\ 0.8P_1(y^{n-1}) \end{bmatrix},$$

where  $P_0(y^0) = P_1(y^0) = 1$ .

(matrix of the) **Simple POST channel**

$P(Y_1|X_1, s_0 = 0)$

	$X_1$	0	1
$Y_1$			
0		1	$\frac{1}{2}$
1		0	$\frac{1}{2}$

$P(Y_1|X_1, s_0 = 1)$

	$X_1$	0	1
$Y_1$			
0		$\frac{1}{2}$	0
1		$\frac{1}{2}$	1

$$P_{n,0} = \begin{bmatrix} 1 \cdot P_{n-1,0} & \frac{1}{2} \cdot P_{n-1,0} \\ 0 \cdot P_{n-1,1} & \frac{1}{2} \cdot P_{n-1,1} \end{bmatrix} \quad P_{n,1} = \begin{bmatrix} \frac{1}{2} \cdot P_{n-1,0} & 0 \cdot P_{n-1,0} \\ \frac{1}{2} \cdot P_{n-1,1} & 1 \cdot P_{n-1,1} \end{bmatrix}$$

$$P_{n,0}^{-1} = \begin{bmatrix} 1 \cdot P_{n-1,0} & \alpha \cdot P_{n-1,0} \\ 0 \cdot P_{n-1,1} & \bar{\alpha} \cdot P_{n-1,1} \end{bmatrix}^{-1} = \begin{bmatrix} P_{n-1,0}^{-1} & -\frac{\alpha}{\bar{\alpha}} P_{n-1,1}^{-1} \\ 0 & \frac{1}{\bar{\alpha}} P_{n-1,1}^{-1} \end{bmatrix}$$

$$P_{n,1}^{-1} = \begin{bmatrix} \bar{\alpha} \cdot P_{n-1,0} & 0 \cdot P_{n-1,0} \\ \alpha \cdot P_{n-1,1} & 1 \cdot P_{n-1,1} \end{bmatrix}^{-1} = \begin{bmatrix} \frac{1}{\bar{\alpha}} P_{n-1,0}^{-1} & 0 \\ -\frac{\alpha}{\bar{\alpha}} P_{n-1,0}^{-1} & P_{n-1,1}^{-1} \end{bmatrix}$$

using:

$$P_0(x^n) = P_{n,0}^{-1}P_0(y^n), \quad P_1(x^n) = P_{n,1}^{-1}P_1(y^n)$$

we obtain:

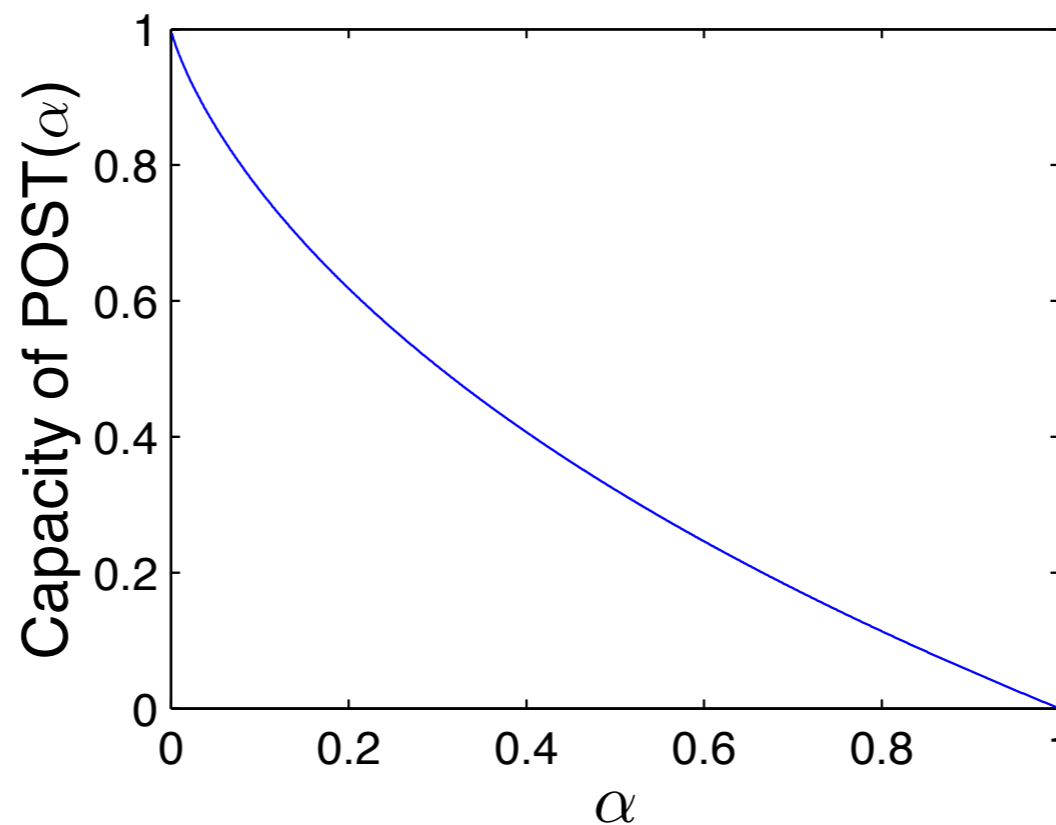
$$P_0(x^n) = \begin{bmatrix} 0.8P_0(x^{n-1}) - 0.2P_1(x^{n-1}) \\ 0.4P_1(x^{n-1}) \end{bmatrix}$$
$$P_1(x^n) = \begin{bmatrix} 0.4P_0(x^{n-1}) \\ 0.8P_1(x^{n-1}) - 0.2P_0(x^{n-1}) \end{bmatrix}$$



## Feedback does not increase capacity of POST( $\alpha$ )

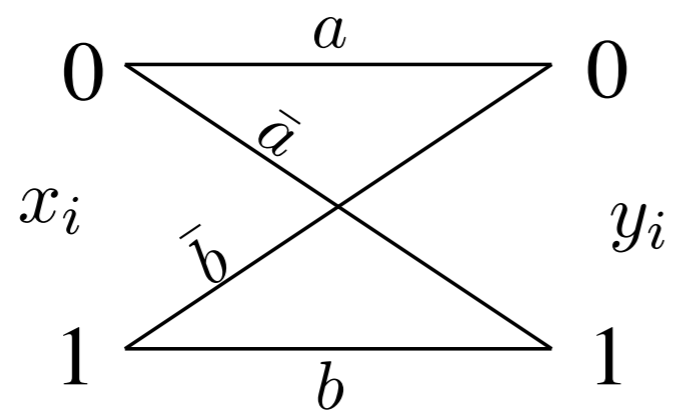
The feedback and the non-feedback capacity of POST( $\alpha$ ) channel is the same as of the memoryless  $Z$  channel with parameter  $\alpha$ , which is  $C = -\log_2 c$  where

$$c = (1 + \bar{\alpha}\alpha^{\frac{\alpha}{\bar{\alpha}}})^{-1}$$

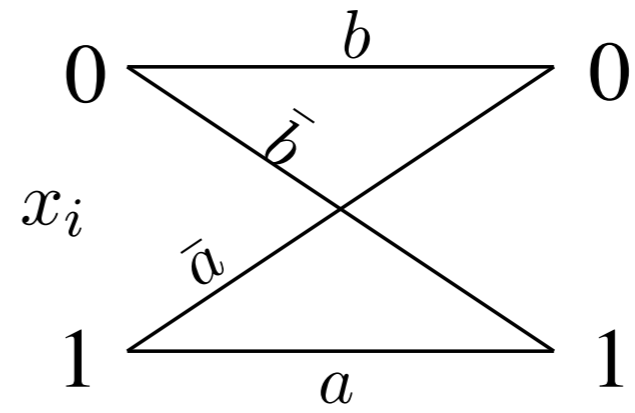


# POST( $a, b$ ) channel

$$y_{i-1} = 0$$



$$y_{i-1} = 1$$



# the input distribution

$$P_0(x^n) = \frac{1}{(a+b-1)(\gamma+1)} \begin{bmatrix} b\gamma P_0(x^{n-1}) - \bar{b}P_1(x^{n-1}) \\ -\bar{a}\gamma P_0(x^{n-1}) + aP_1(x^{n-1}) \end{bmatrix}$$

$$P_1(x^n) = \frac{1}{(a+b-1)(\gamma+1)} \begin{bmatrix} aP_0(x^{n-1}) - \bar{a}\gamma P_1(x^{n-1}) \\ -\bar{b}P_0(x^{n-1}) + b\gamma P_1(x^{n-1}) \end{bmatrix}$$

$$\gamma = 2 \frac{H(b) - H(a)}{a+b-1}.$$

In order to prove that  $P(x^n)$  is valid we needed:

- $\gamma \geq \frac{\bar{b}}{b}$
- $\gamma \leq \frac{a}{\bar{a}}$
- $\gamma \geq \frac{a}{b}$  for  $a \geq \bar{b}$
- $\gamma^2 \leq \frac{a^2}{b\bar{a}}$  for  $a \geq \bar{b}$
- $\frac{\gamma(\bar{a}+b)}{2\bar{b}} \geq 1$  for  $a \geq \bar{b}$  and  $a\bar{a} \leq b\bar{b}$
- $\gamma^2(\bar{a} + b)^2 - 4a\bar{b} \geq 0$
- $\gamma(\bar{a} + b) - \sqrt{\gamma^2(\bar{a} + b)^2 - 4a\bar{b}} \leq 2\bar{b}$ , for  $a \geq \bar{b}$  and  $a\bar{a} \leq b\bar{b}$

where

$$\gamma = 2 \frac{H(b) - H(a)}{a + b - 1}.$$

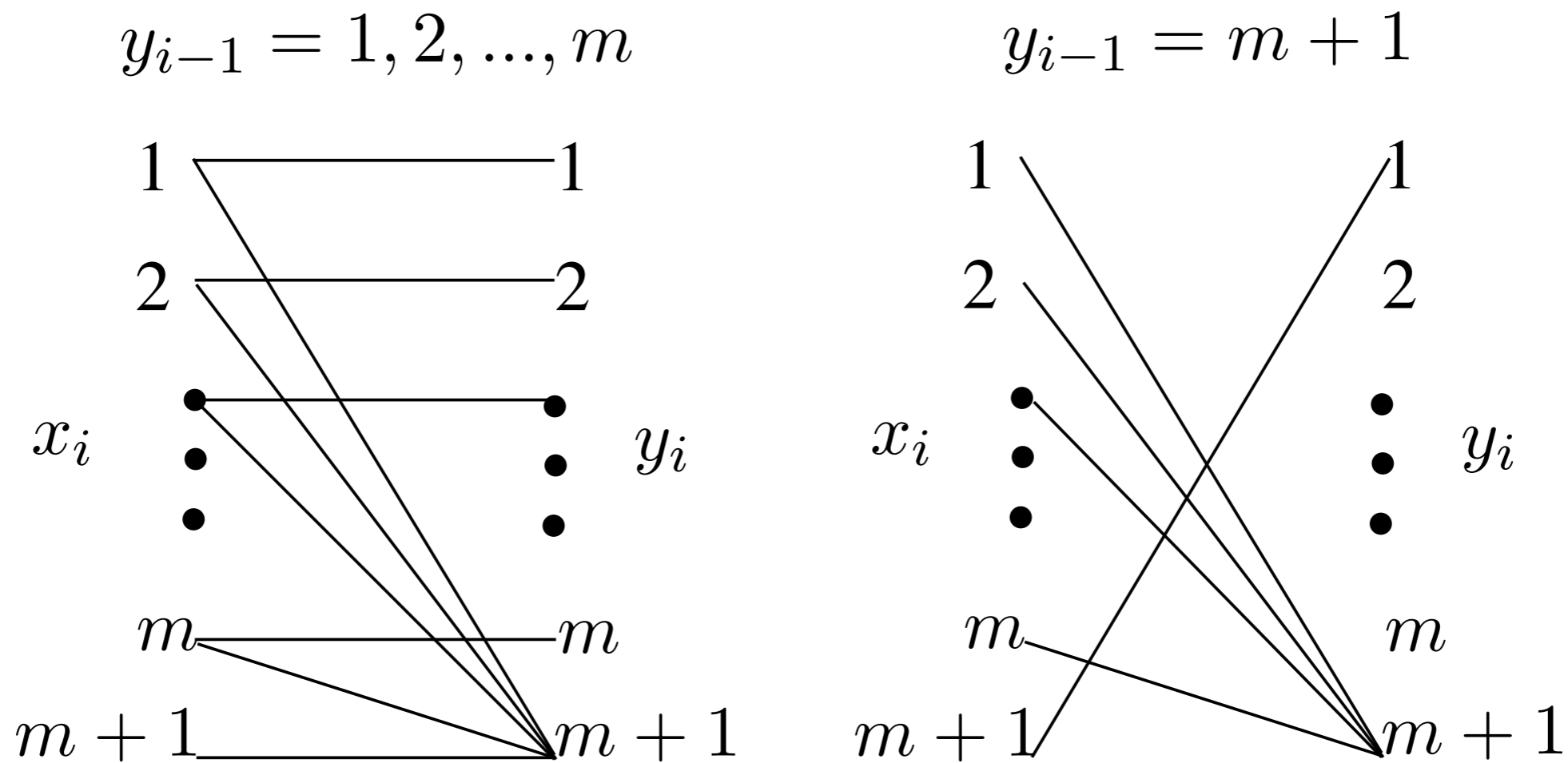
## Feedback does not increase capacity of a POST( $a, b$ ) channel

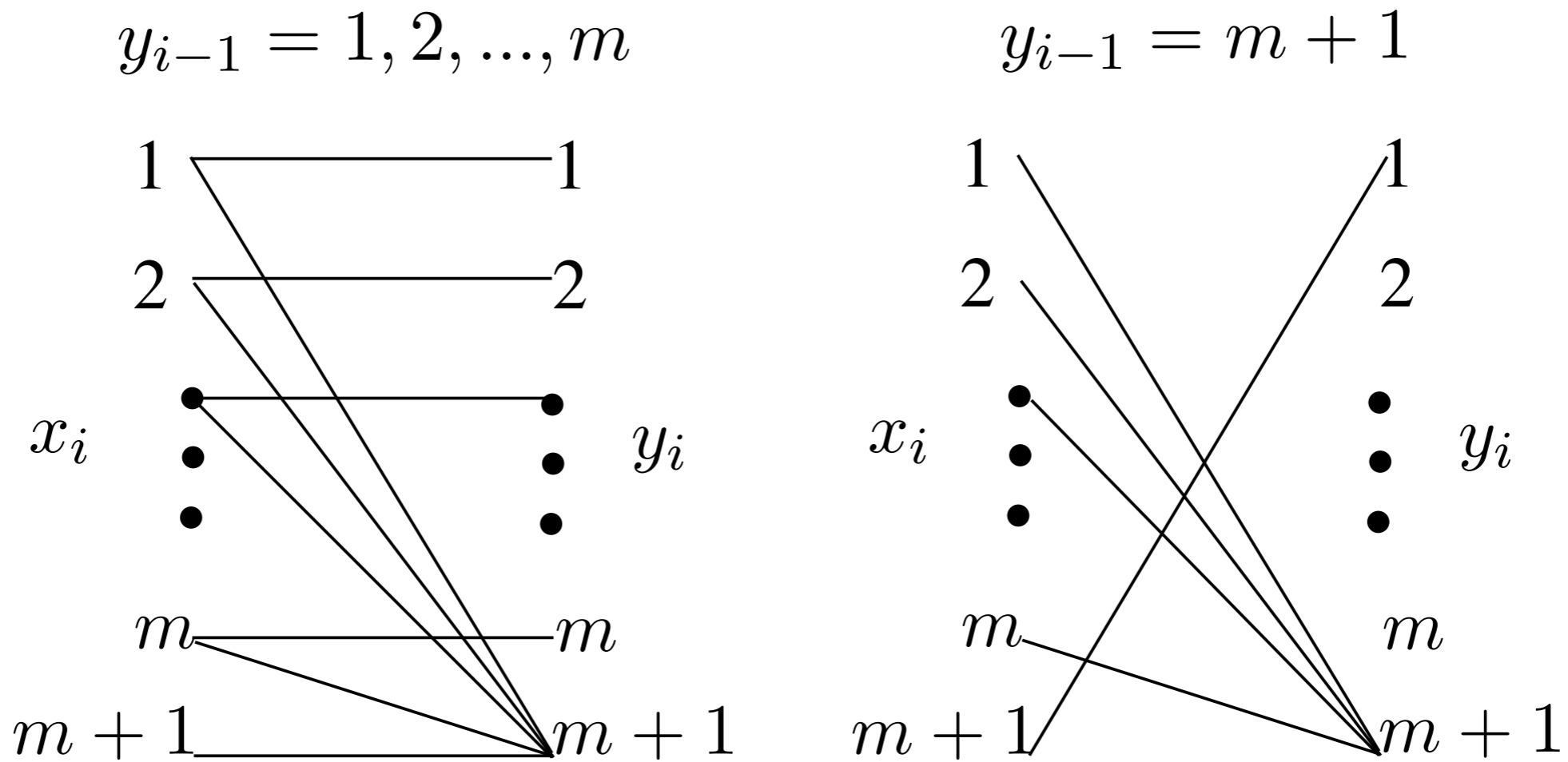
The feedback and the non-feedback capacity of POST( $a, b$ ) channel is the same as of a binary DMC channel with parameters  $(a, b)$ , which is given by

$$C = \log \left[ 2^{\frac{\bar{a}H_b(b) - bH_b(a)}{a+b-1}} + 2^{\frac{\bar{b}H_b(a) - aH_b(b)}{a+b-1}} \right].$$

**can feedback help for a general POST channel?**

# large alphabet





	upper bound on capacity	lower bound on $C_{fb}$
$m$	$\frac{1}{6} \max_{s_0} \max_{P(x^6)} I(X^6; Y^6   s_0)$	$R = \frac{\log_2 m}{3}$
$2^9$	<b>2.5376</b>	<b>3.0000</b>



# is the sufficient condition necessary?

- we didn't know
- numerics in binary case were inconclusive

# condition is necessary and sufficient

## Theorem:

Let  $C_{FB} = \max_{p(x_1|y_0)} I(X_1; Y_1|Y_0)$  and let  $p^*(y_0)$  be induced by  $p^*(x_1|y_0)$ .

$C_{FB} = C_{NFB}$  if and only if  $\forall n$ :

Can induce  $p_{fb}^*(y^n)$  without feedback and without knowledge of  $Y_0 \sim p^*(y_0)$ .

# implication

$\exists$  binary POST channels with  $C_{FB} > C_{NFB}$

# conclusions

- $C_{FB} = C_{NFB} \Leftrightarrow$  can induce  $p_{fb}^*(y^n)$  without feedback  $\forall n$
- $C_{FB} = C_{NFB}$  for binary POST( $a, b$ ) channels
- $\exists$  POST channels with  $C_{FB} > C_{NFB}$

# questions

- crisper necessary and sufficient condition for  $C_{FB} = C_{NFB}$
- capacity achieving schemes