

# To Iterate or Feedback: How to Achieve Universally Good Performance on Channels with Memory?

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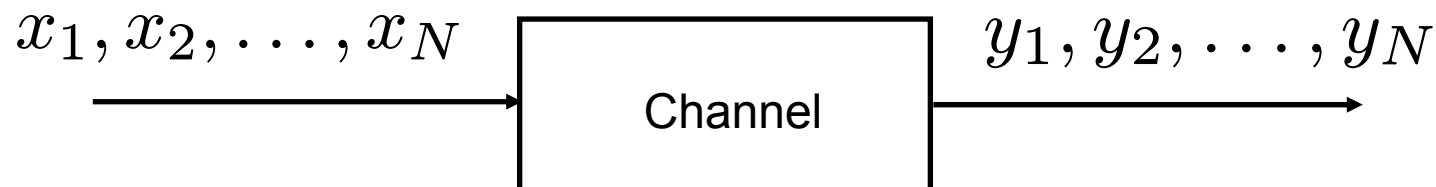
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# Definition of Channel with Memory



- Memoryless Channel:  $p(y_1, \dots, y_N | x_1, \dots, x_N) = \prod_{i=1}^N p(y_i | x_i)$
- Channel with memory: The above condition is not true
- In this talk, focus is on the case when there is a Markov structure

# Coding for Memoryless Channels

- Consider the following Memoryless Channels

Additive White Noise:  $y_k = x_k + n_k$        $x_k \in \{\pm 1\}$

BSC:  $y_k = x_k \oplus e_k$

BEC:  $y_k = x_k$  OR  $y_k = \text{Erasure}$

Fading with CSI at Rx:  $y_k = \alpha_k x_k + n_k$ ,  $\alpha_k$  is known at the receiver

- Code design for such channels is well understood
  - Low Density Parity Check (LDPC) codes perform close to capacity
  - Tools to analyze and design good code ensembles exist

# Universal LDPC Codes for Memoryless channels

- LDPC codes perform uniformly well on several memoryless channels (S.Y. Chung, Ph.D. thesis, MIT)

j	k	rate	BEC	BSC	Laplace	AWGN
3	4	0.25	0.353	0.349	0.348	0.346
3	5	0.4	0.482	0.489	0.485	0.480
3	6	0.5	0.571	0.584	0.577	0.571
3	9	2/3	0.717	0.739	0.728	0.720
3	12	0.75	0.790	0.814	0.801	0.793
4	10	0.6	0.692	0.689	0.684	0.683

- A single good LDPC code exists for the class of AWN channels

$$y_k = x_k + n_k$$

# Examples of Channels with memory

- ISI channel:  $y_k = \sum_{i=-l_2}^{l_1} h_i x_{k-i} + n_k \quad x_k \in \mathcal{X}$
- Additive correlated noise:  $y_k = x_k \oplus w_k$
- MIMO Channels:  $\mathbf{Y}_k = \mathbf{H} \mathbf{x}_k + \mathbf{n}_k$
- Fading channels without CSI:  $y_k = \alpha_k x_k + n_k$
- We want good codes for these channels (preferably universal)

# Capacity of ISI Channels with Input Constraints

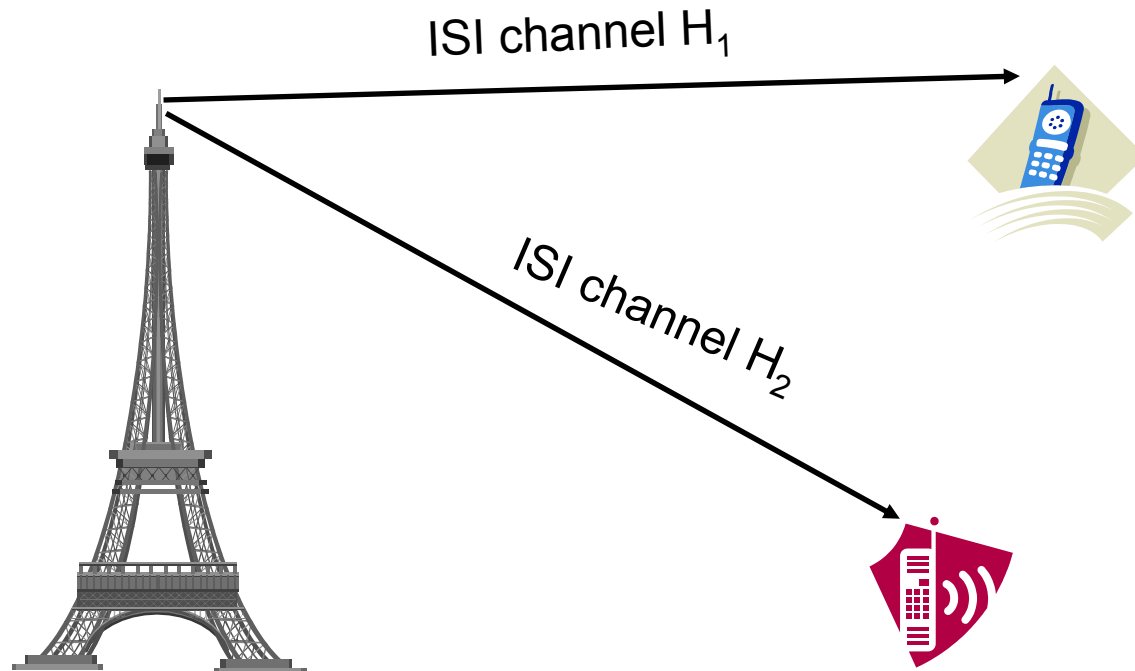
$$y_k = \sum_{i=-l_2}^{l_1} h_i x_{k-i} + n_k \quad x_k \in \{\pm 1\}$$

- Computing capacity with input constraints is quite difficult
- Good upper and lower bounds are available
- $C_{\text{i.i.d}}$  – is a lower bound assuming  $x_k$ 's are i.i.d
- Better bounds – assume  $x_k$ 's are the output of a Markov chain
- When channel is not known at transmitter,  $C_{\text{i.i.d}}$  is a meaningful measure of performance

# What is the universality we are looking for?

- Set of ISI channels  $\{H_1, H_2, \dots\}$ ,  $C_{\text{iid}}(H_i) = R$
- We want a coding scheme of rate  $R$  (or close to  $R$ ) and a receiver such that for any  $H_i$ , reliable communication is possible
- Note that the channel energy is not normalized
- In the MIMO case, for every realization of  $\mathbf{H}$  for which  $C_{\text{i.i.d}}(\mathbf{H}) > R$ , we want the code to work
- With some distribution on  $H$ , we talk of Outage capacity

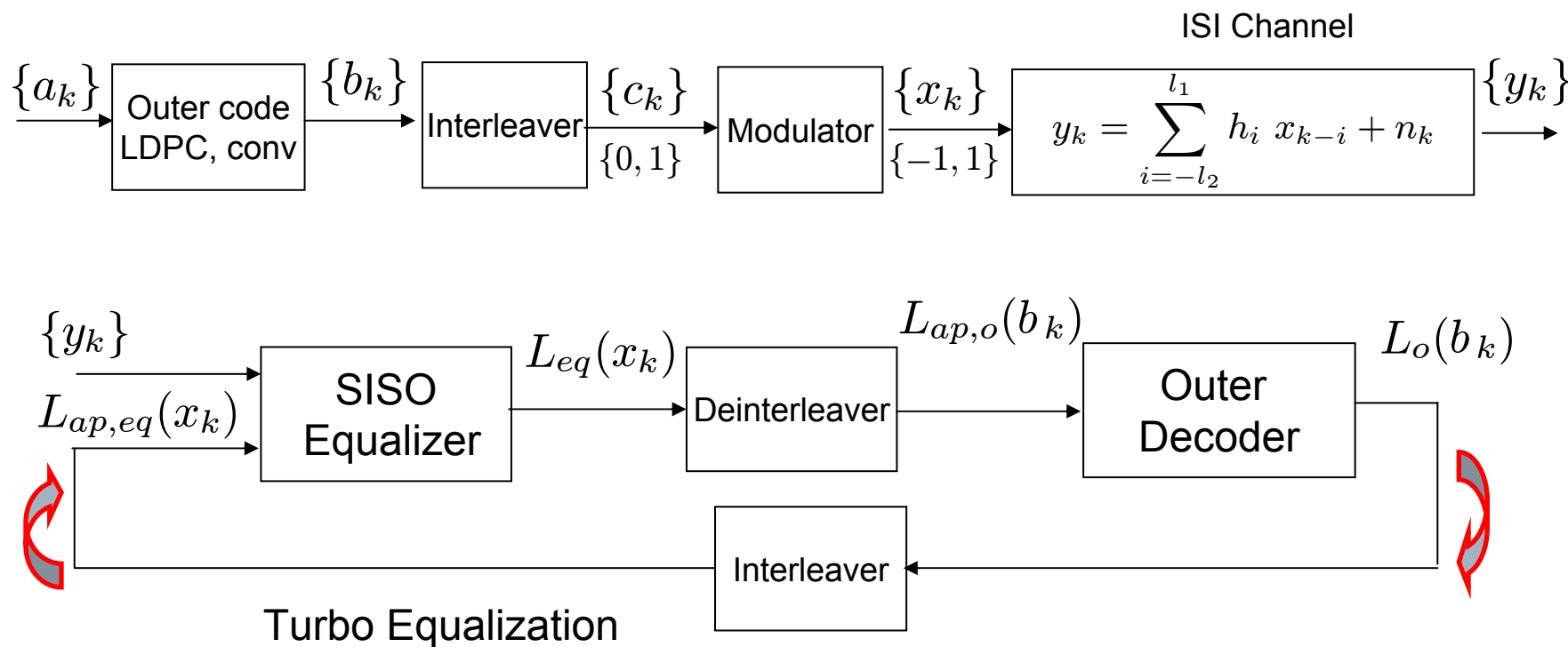
# A Possible Application



- Note that the problem is not to optimize sum rate
- In this sense, it is different from conventional broadcast problem



# How to deal with the memory? - Equalization



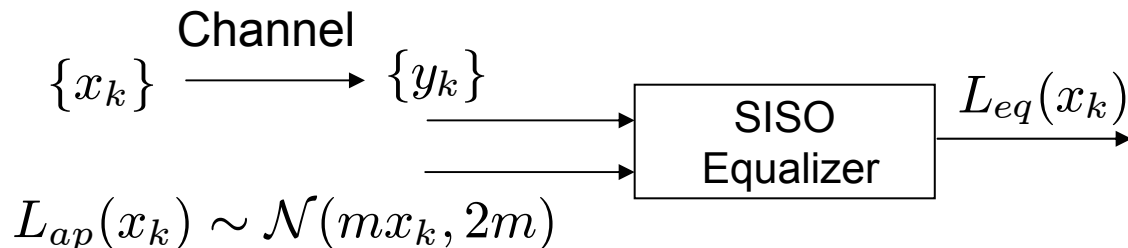
- ❑ Computationally efficient way to approximate the MAP decoder
- ❑ Code design with turbo equalization [Varnica & Kavcic '02]

# Main point of the talk

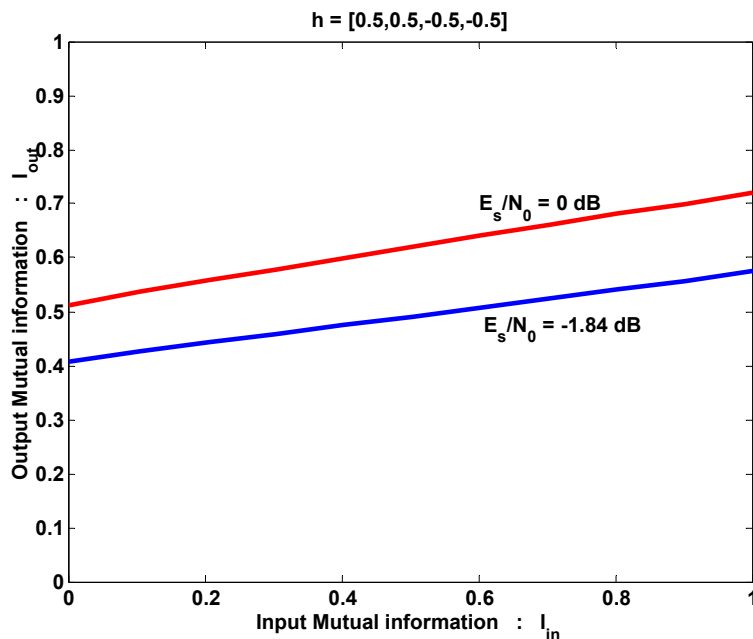
- Can Turbo equalization provide near capacity performance?
  - Yes!
  - But optimal codes are different for different channel realizations
  - Not a computationally efficient solution for near-capacity performance
  
- Coded Decision Feedback Equalization
  - Non-iterative structure that can provide close to capacity performance
  - Leads to universal codes and receiver structures
  - Complexity is lower than that of turbo equalization
  
- This principle applies to other signal processing functions as well

# Extrinsic Information Transfer (EXIT) Curves

- Proposed by Ten Brink, Similar ideas by some other researchers also

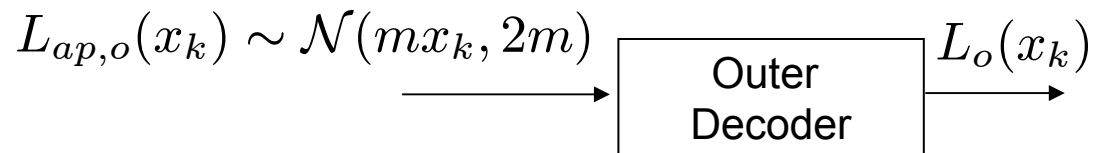


$F_{eq}$  - Plot of  $I_{in} = I(L_{ap}; X)$  vs  $I_{out} = I(L_{eq}; X)$



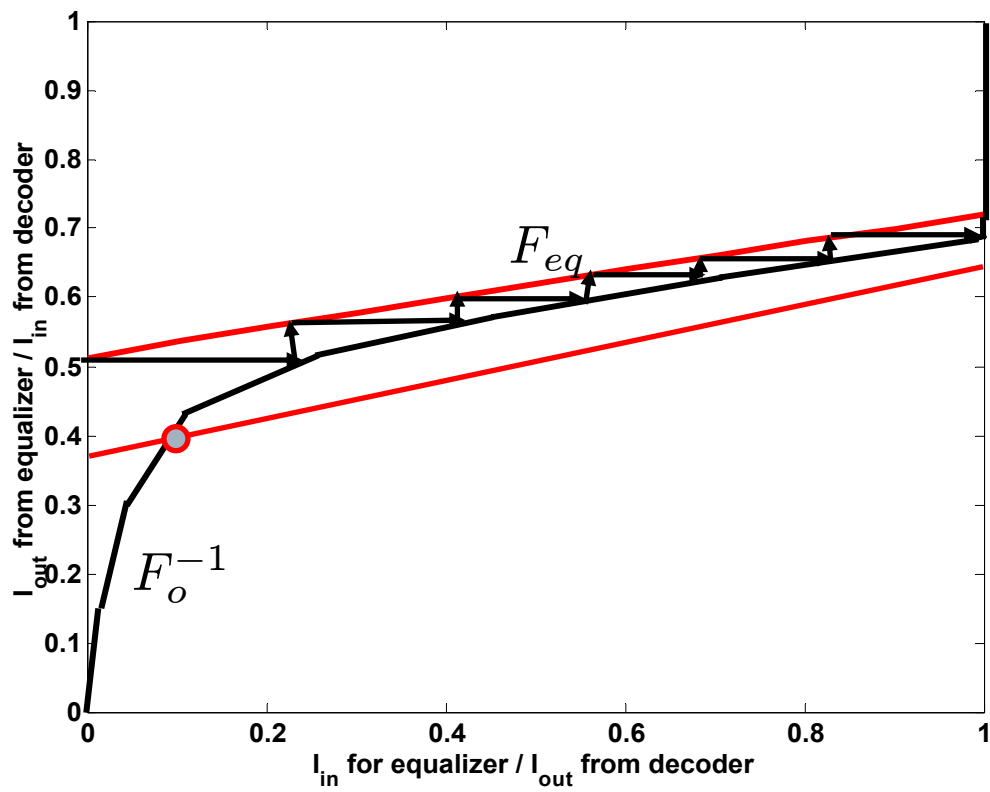
# EXIT Curve for Outer Decoder

- Similar curve for the outer decoder also



$F_o$  - Plot of  $I_{out}$  vs  $I_{in}$  - transfer function of decoder

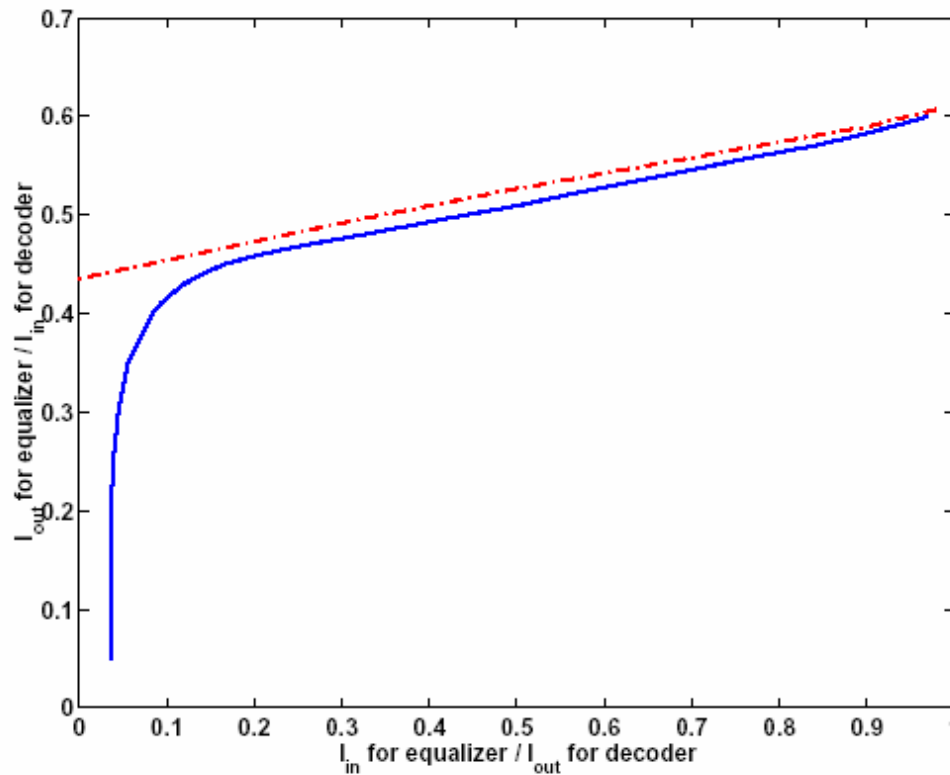
# EXIT Chart



Convergence:  $F_o^{-1} < F_{eq}$  for convergence

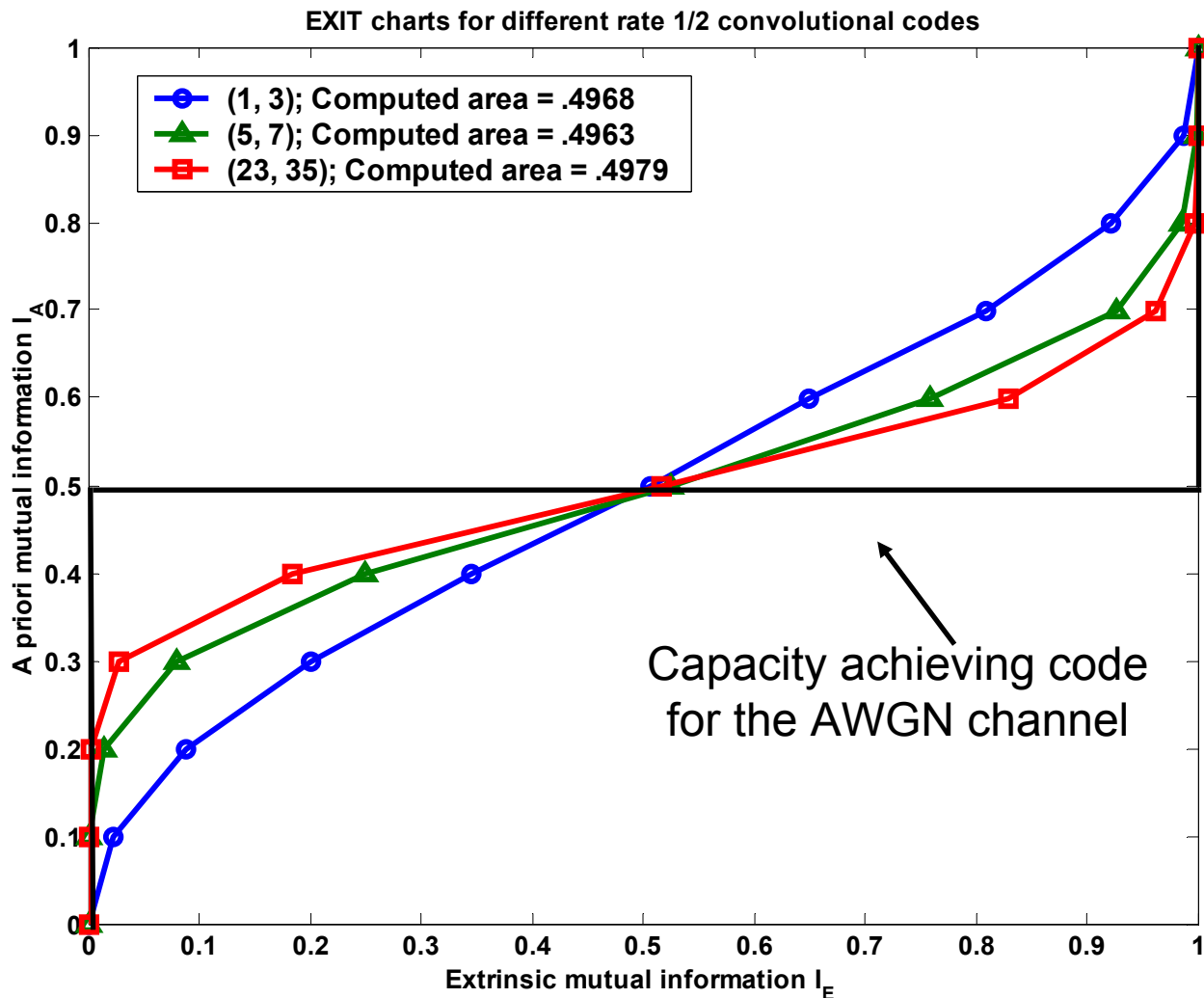
Area Property: [Ashikmin 02] If extrinsic info is from an erasure channel, area under  $F_o^{-1}(x) = R$ , Rate of the code

# Optimal Code Design with Turbo Equalization

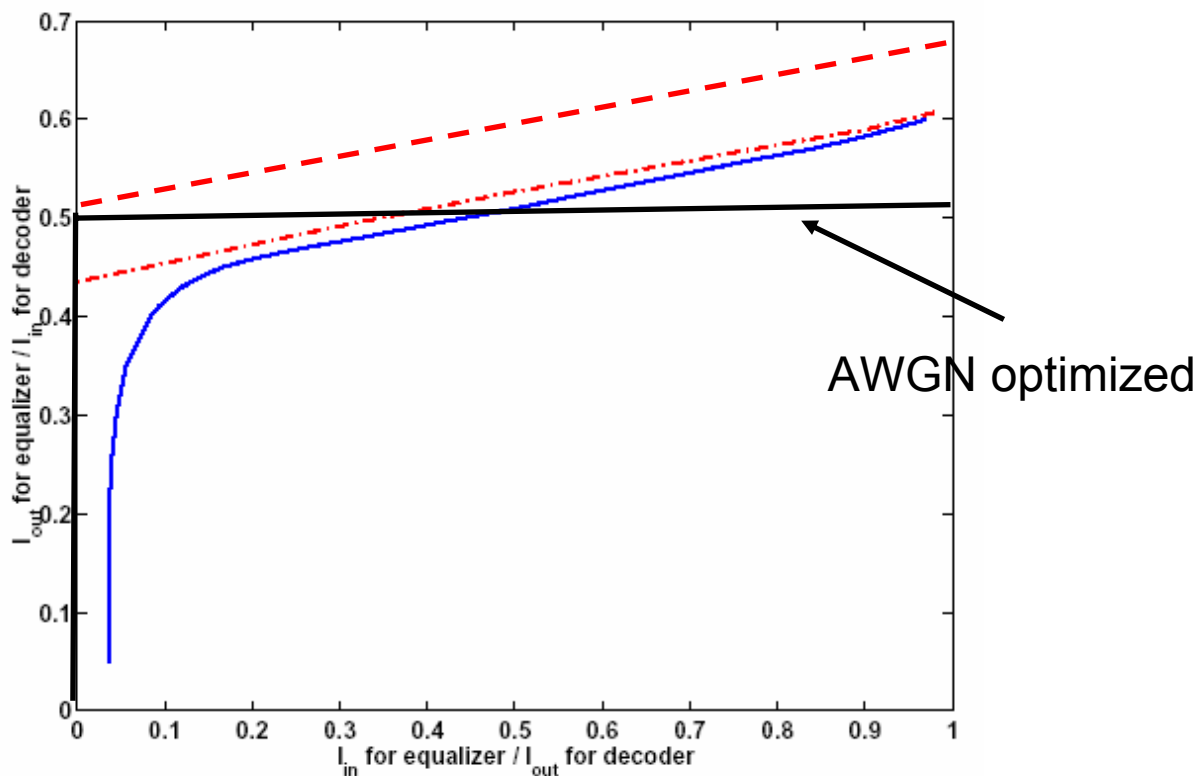


- ❑ Optimal Code:  $F_{eq} = F_o^{-1}$  - Match the EXIT Curves
- ❑ Area under  $F_{eq}$  achievable information rate
- ❑ Extension to Gaussian a priori [Our result, Allerton 2004]

# Area Property



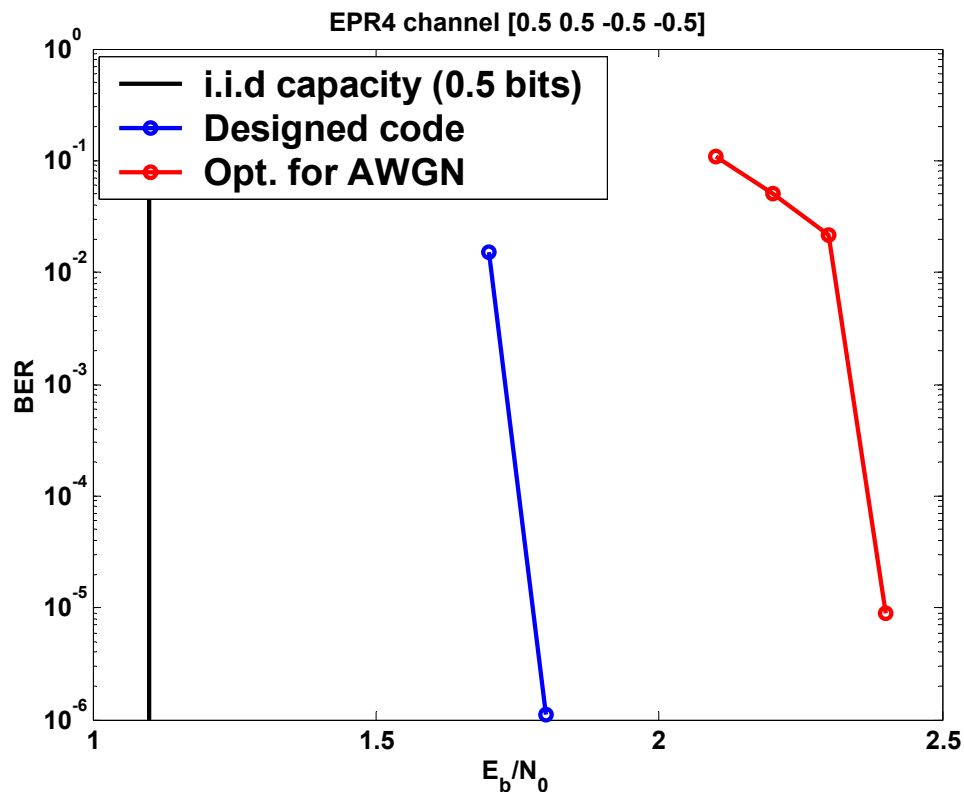
# How bad are codes optimized for other channels?



- ❑ AWGN optimized code is worse by 0.85 dB
- ❑ Or, there is a loss in rate

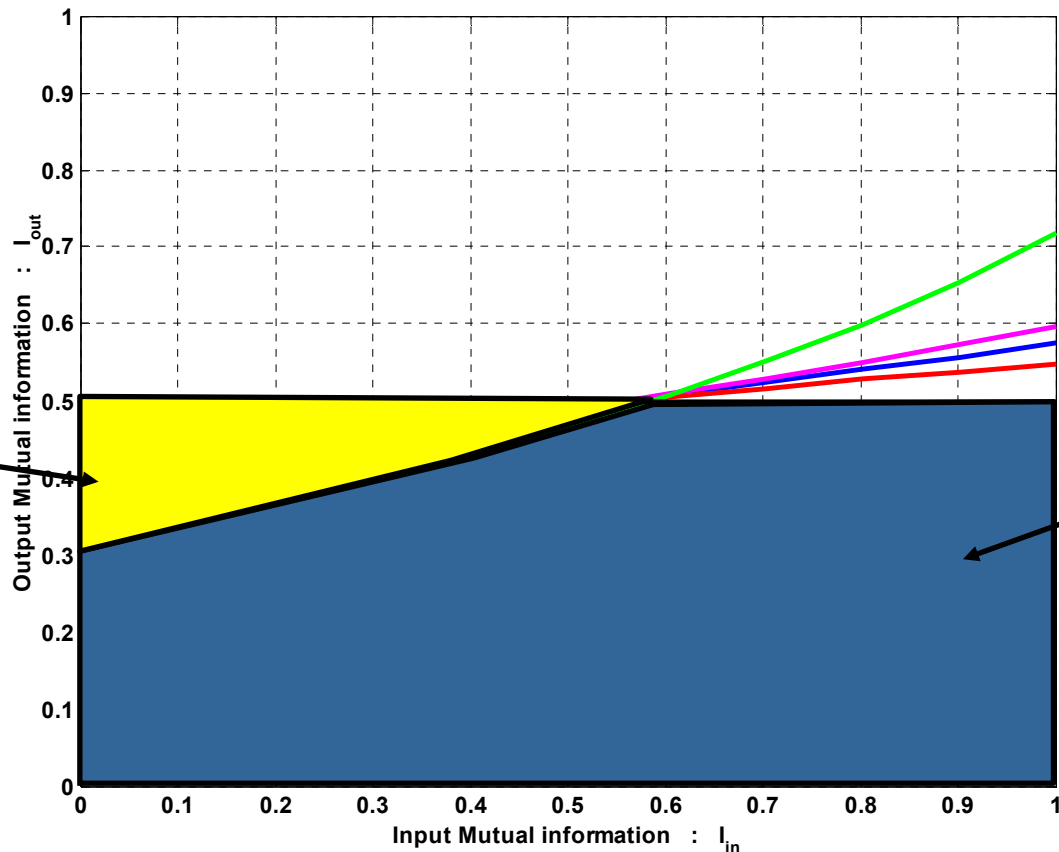


# Simulation Results



- Optimized code is better by 0.65 dB
- Fairly consistent with EXIT charts

# EXIT Charts for Different Channels



Loss in Rate

Reduce the code rate to this area

- Area under the hull is the information rate achievable
- No guarantee that matched codes exist

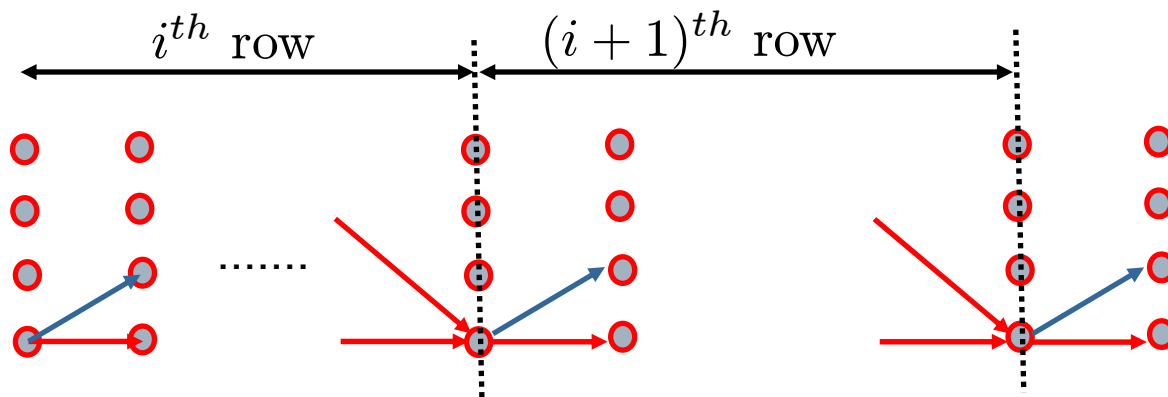


# Prior Work

- Cioffi, Dudevoir, Eyuboglu and Forney
  - MMSE-DFE is canonical
  - Transmitter precoding
  
- Varanasi and Guss
  - Optimality of DFE-MUD
  - Extension to ISI does not produce universal structure
  - Deals only with Gaussian inputs
  
- Our results are for the finite alphabet case with Markov memory

# The BCJR-DFE Receiver

- Produces  $\lambda(x_k) = P(x_k = 1 | x_{k-1}, \dots, x_{k-L}, \mathbf{Y})$  in an efficient way

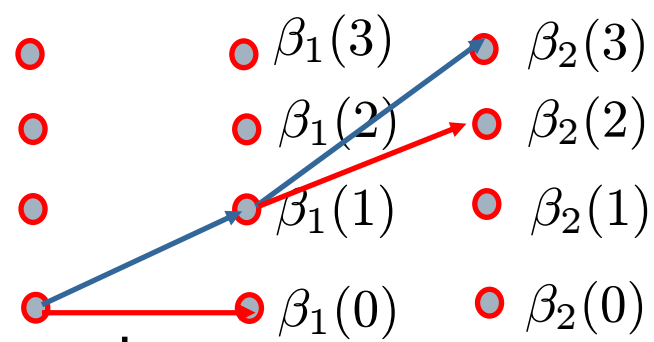


$$\begin{aligned}
 P(x_k = 1 | x_{k-1}, \dots, x_{k-L}, \mathbf{Y}) &= P(x_k | S_k, \mathbf{Y}) = P(x_k = 1 | S_k, y_{i,k}^{i,m}) \\
 &= \gamma_1(s_k, s_{k+1}, y_{i,k}) \cdot \beta_{k+1}(s)
 \end{aligned}$$

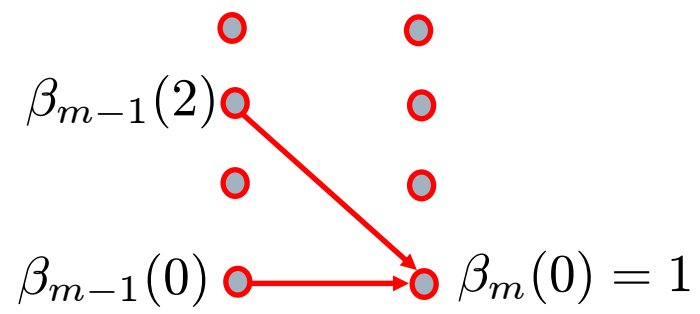
$\gamma_i(s_k, s_{k+1}, y_{i,k})$  - Branch transition probability

$$\beta_{k+1}(s) = P(y_{i,k+1}^{i,m} | S_{k+1} = s)$$

# BCJR-DFE Algorithm



.....



$\lambda(x_{1,0})$

$\lambda(x_{2,0})$

Decode

# Optimality of the Encoder/Decoder Structure

□ For  $m, n \rightarrow \infty$ , required code rate  $R = C_{i.i.d}$

$$C_{i.i.d} = I(\mathbf{X}; \mathbf{Y}) = I(x_1, x_2, \dots, x_{mn}; \mathbf{Y})$$

$$C_{i.i.d} = \sum_{k=1}^{mn} I(x_k; \mathbf{Y} | x_{k-1}, x_{k-2}, \dots, x_1)$$

$$p(x_k | \mathbf{Y}, x_{k-1}, \dots, x_1) = p(x_k | \mathbf{Y}, x_{k-1}, \dots, x_{k-L})$$

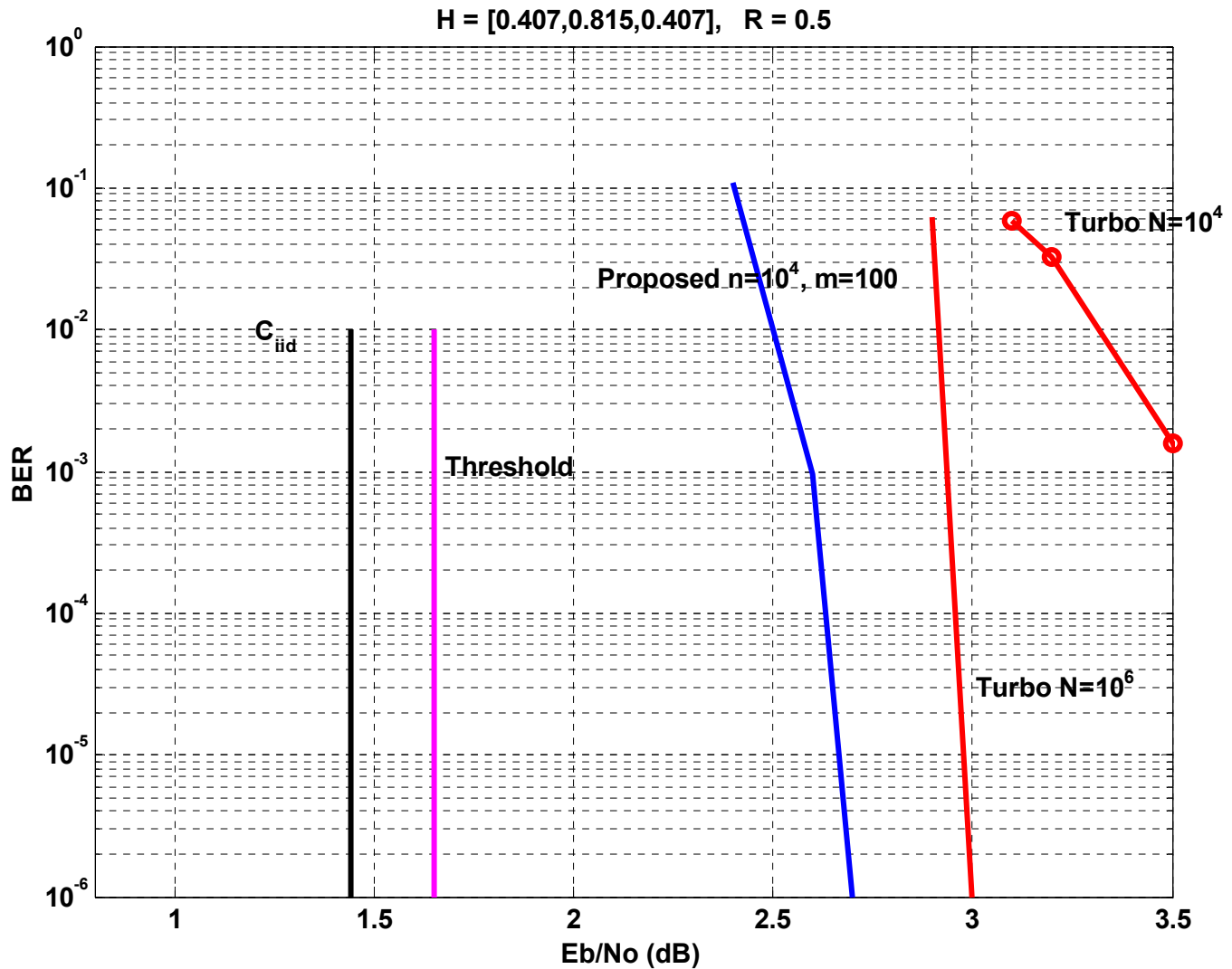
$$I(x_k; \mathbf{Y} | x_{k-1}, \dots, x_1) = I(x_k; \mathbf{Y} | x_{k-1}, \dots, x_{k-L}).$$

$$C_{i.i.d} = m \times n \times I(x_k; \mathbf{Y} | x_{k-1}, \dots, x_{k-L})$$

equivalent channel  $x_k \rightarrow \lambda(x_k) = p(x_k = 1 | \mathbf{Y}, x_{k-1}, \dots, x_{k-L})$

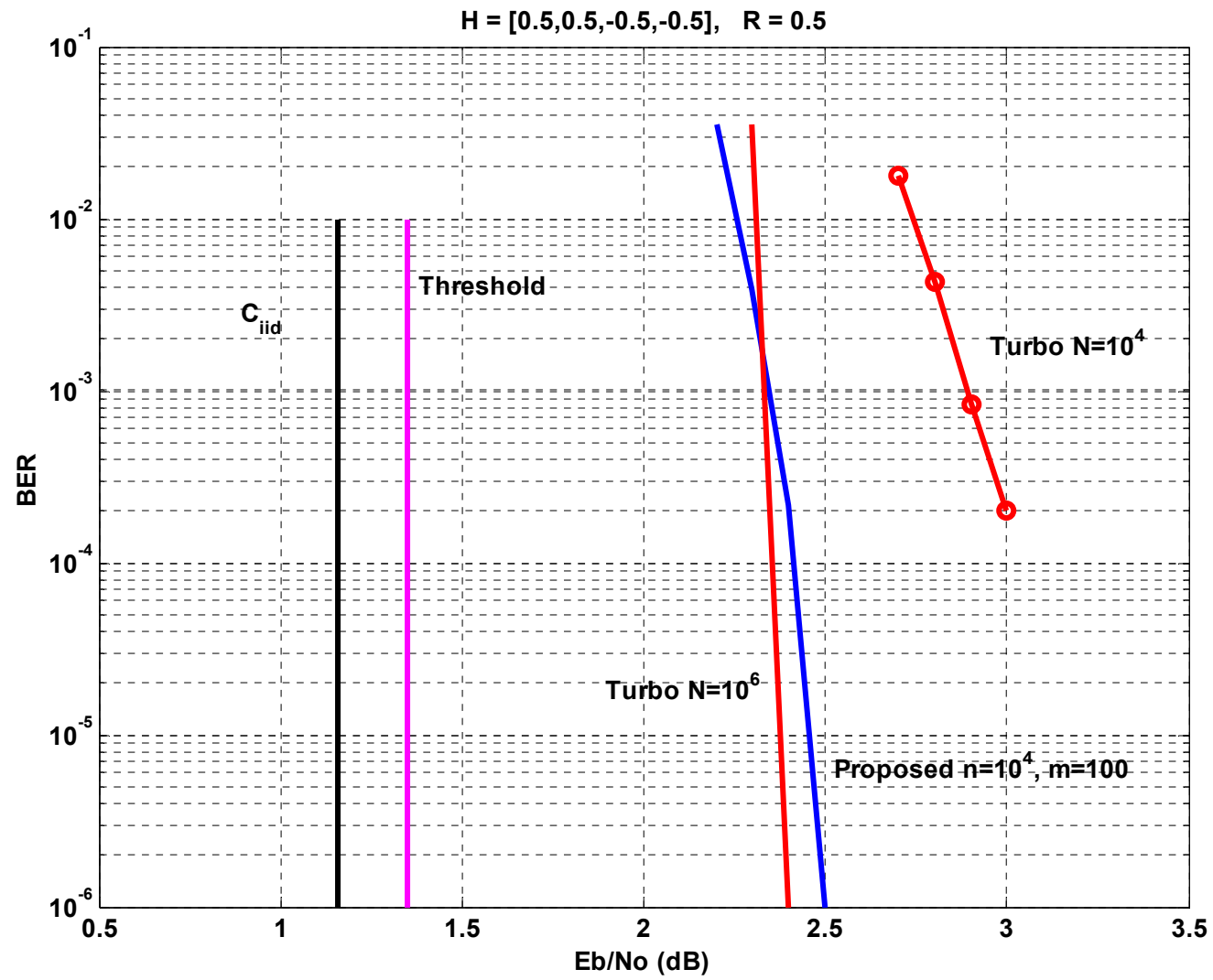
Information rate for this channel is exactly  $C_{i.i.d}$

# Results for a 3-tap channel [0.407 0.815 0.407]





# Another channel [0.5 0.5 -0.5 -0.5]



# Universality

- The equivalent channel is memoryless

$$x_k \rightarrow \lambda(x_k)$$

$$r_{i,k} = x_{i,k} + w_{i,k}, \quad w_{i,k} \text{ and } w_{j,k} \text{ are independent}$$

- Pdf of  $w_{i,k}$  depends on the channel realization
- Universal codes for the memoryless channel will work well
- The distribution is even well approximated as being Gaussian

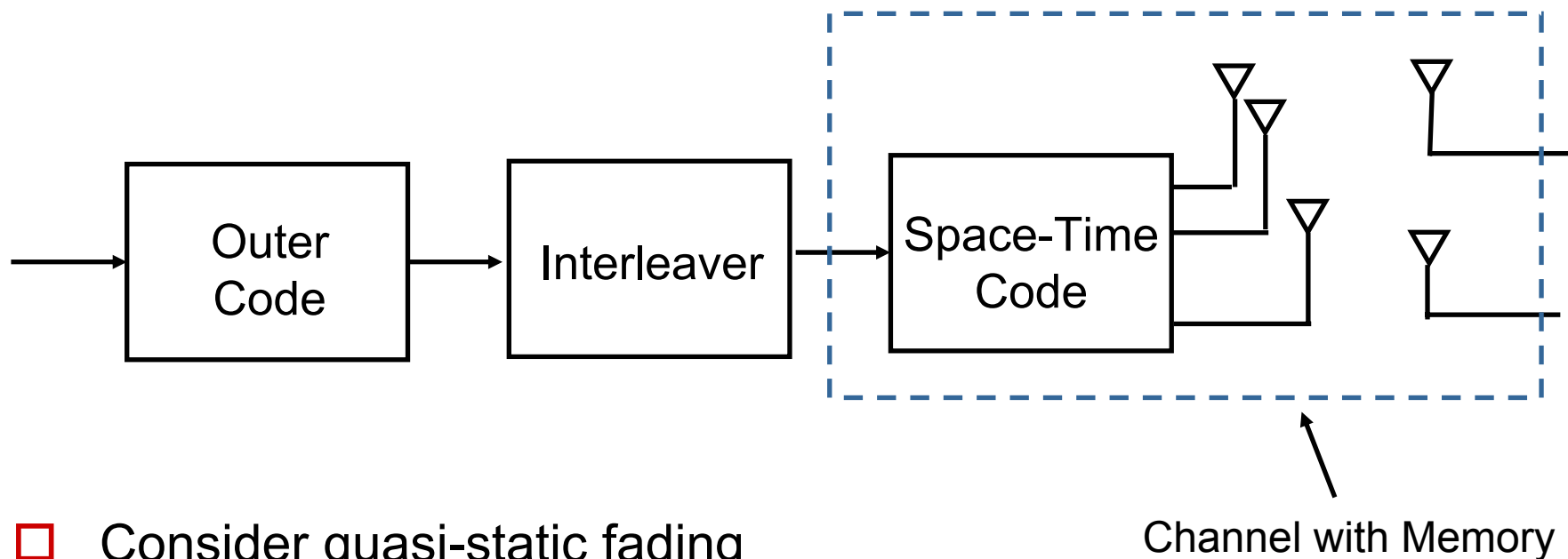
# Results to show Universality

Gap from  $C_{iid}$  dB

Channel	BCJR-DFE	Turbo Equalization
AWGN	0.2	0.2
1-D	0.2	0.6
$1+D-D^2-D^3$	0.2	1.24
$0.407+0.815D+0.407D^2$	0.21	1.6
100 realizations of 3 tap Rayleigh fading taps	0.21	

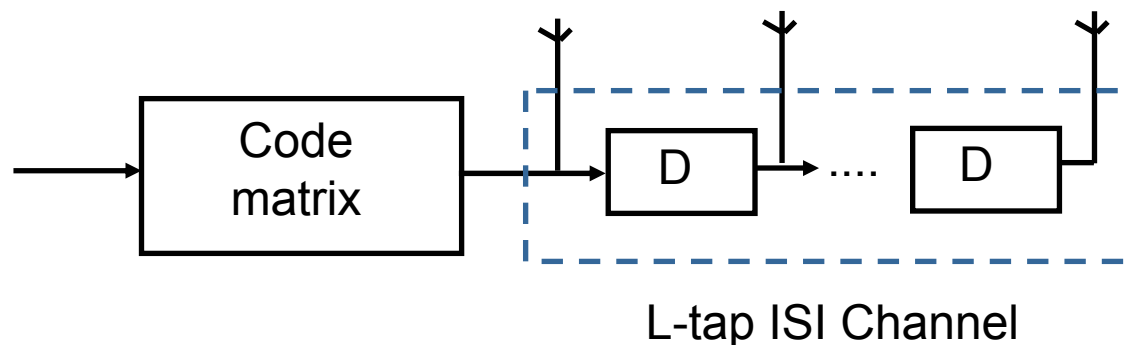
# Extension to Other Channels

# Concatenated Space-Time Codes



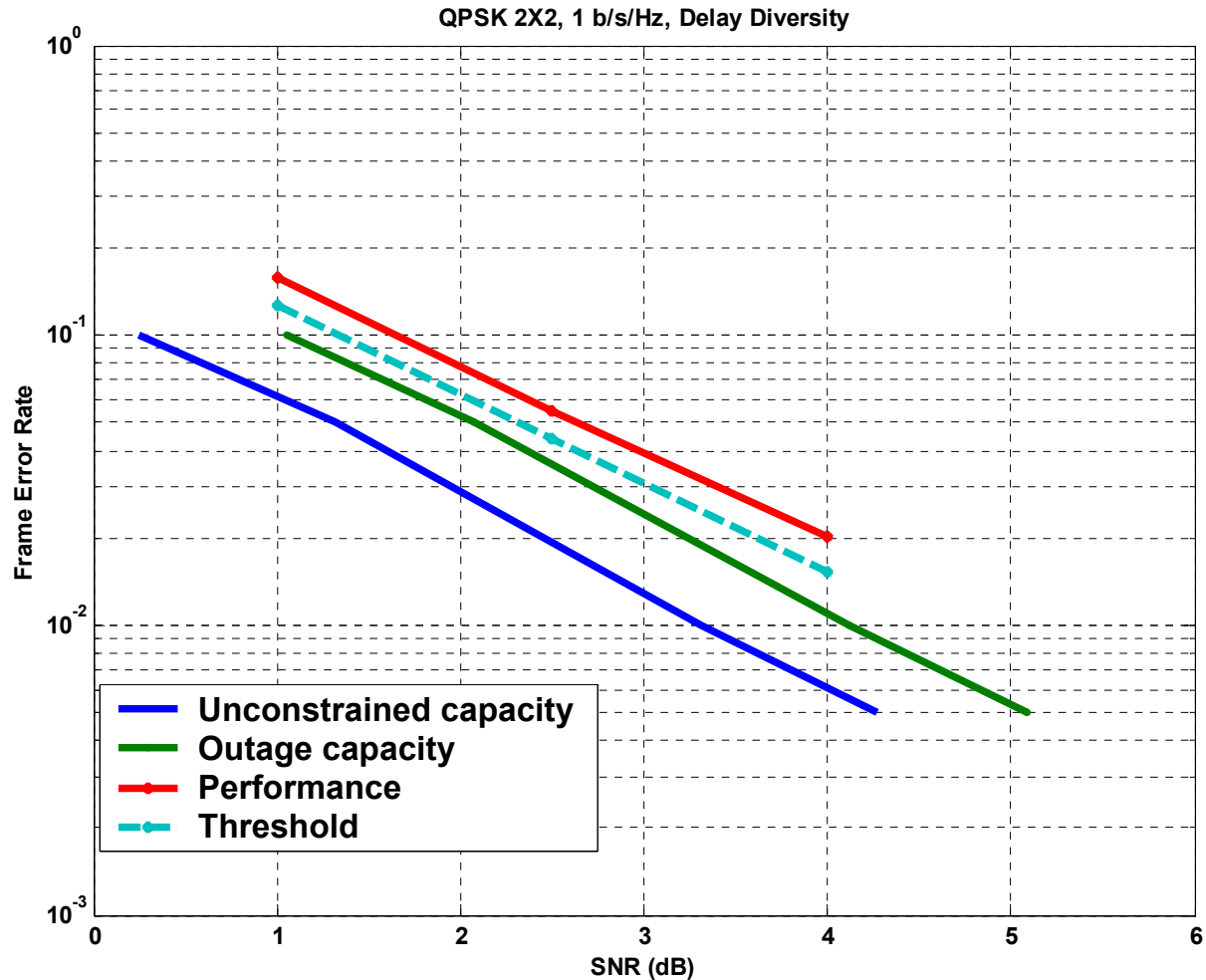
- ❑ Consider quasi-static fading
- ❑ Find a good space-time code and fix it
- ❑ How do we design good outer codes?
- ❑ How to perform joint decoding and space-time demodulation ?

# MIMO Channels – Delay Diversity

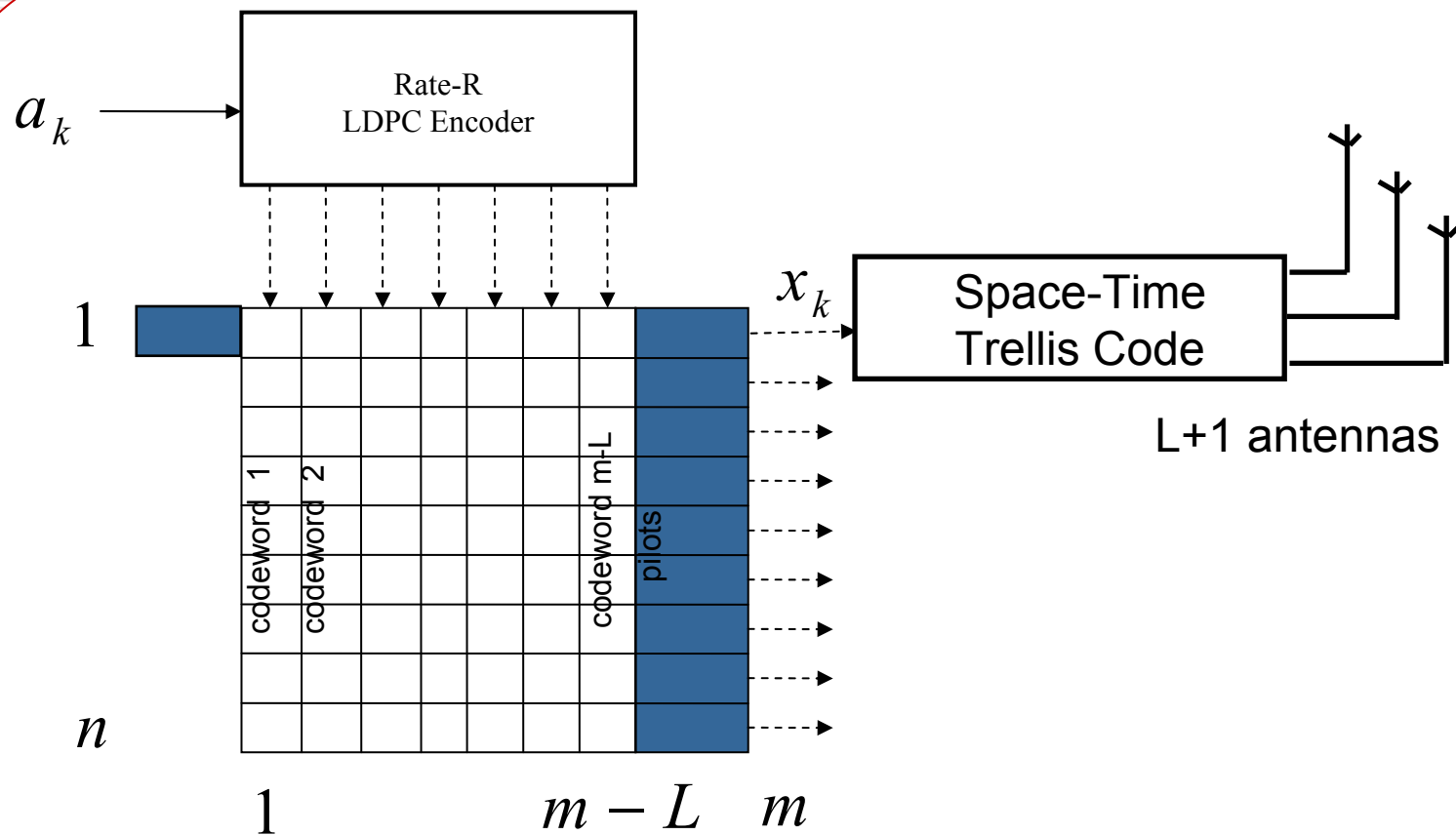


- Extension of MMSE-BLAST to finite constellations
  - MMSE part is replaced by BCJR part
  - BLAST detection is replaced by an optimum decoder
  - Silent periods are replaced by known bits to pin the state
  
- Decoded decisions prevent error propagation in D-BLAST
  
- DFE is universal whereas Turbo-BLAST is not

# Delay Diversity Results

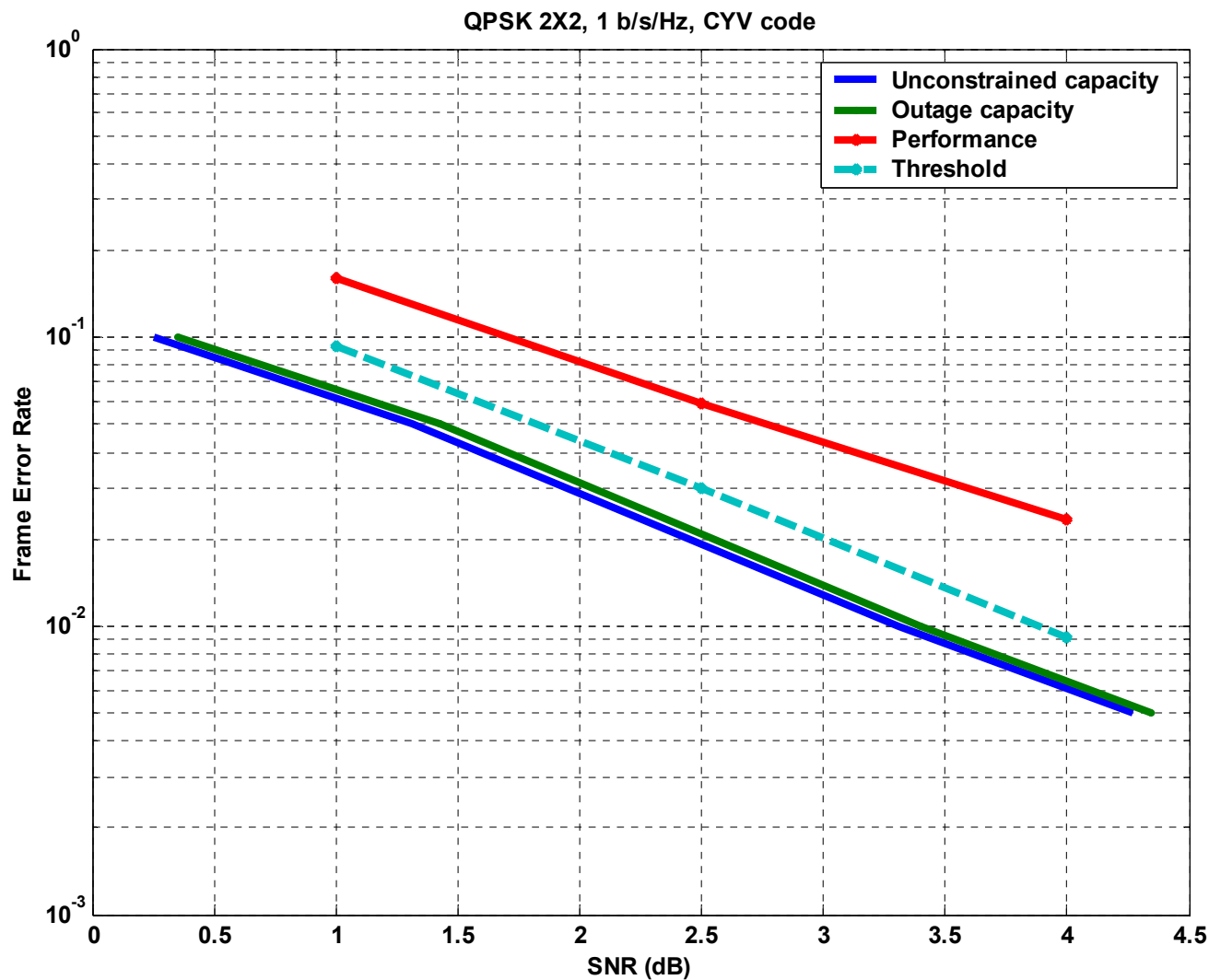


# Any Space-Time Trellis Code Can be Used





# Another QPSK Space-Time Code

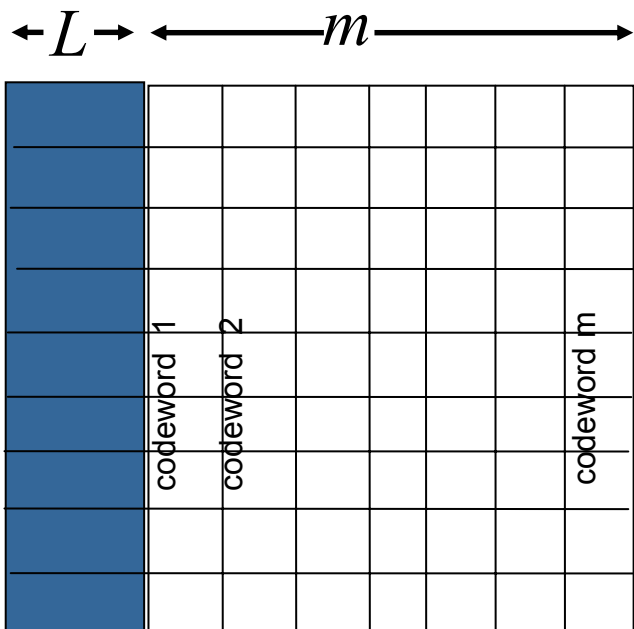


# Non Coherent Detection

$$y_k = \alpha_k x_k + n_k$$

$x_k \in \text{PSK set}$

Flat Correlated Rayleigh Fading



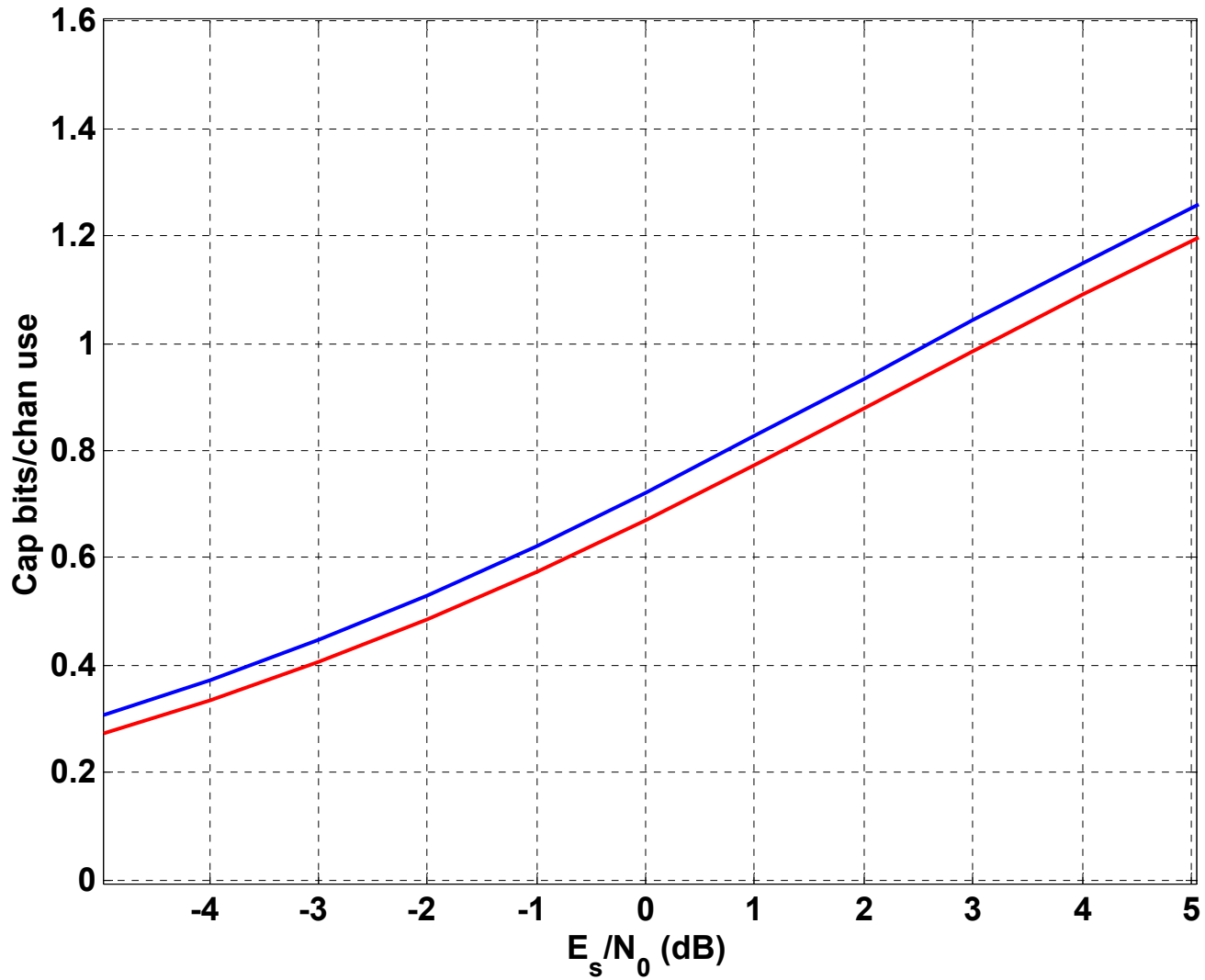
Predict  $\alpha_k$  based on past values of  $\mathbf{y}$  and  $\mathbf{x}$

Vertically channel is memoryless

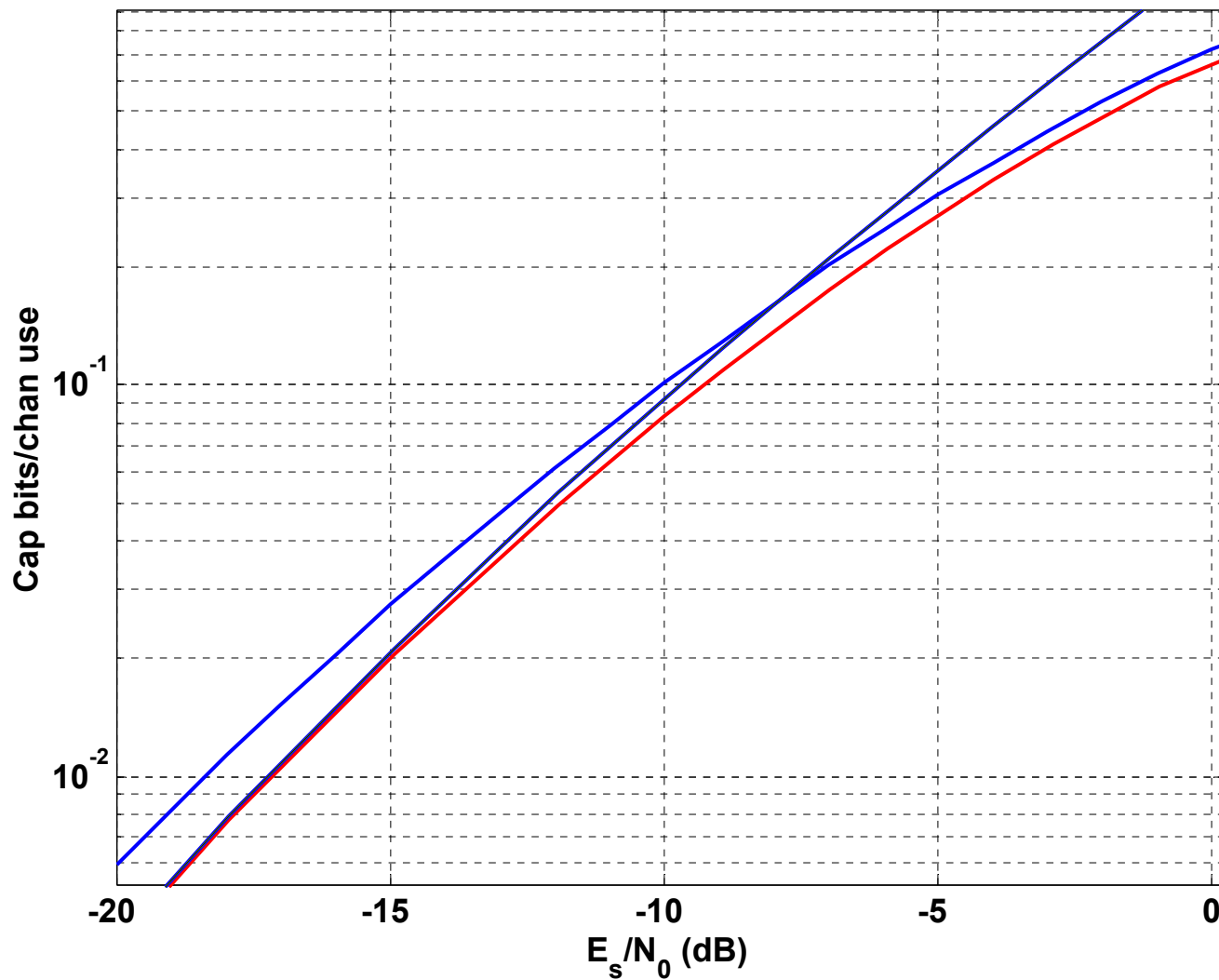
Lower bound on  $C$  that is easy to compute

$$C \leq I(x_k; y_k | \mathbf{Y}_{-\infty}^{k-1}, \mathbf{X}_{-\infty}^{k-1})$$

# Results



# Two different upper bounds and lower bound



# Gaussian Signaling at high SNR

$X_k$  is Gaussian and  $\hat{\alpha}_k$  is the estimate of  $\alpha_k$  based on the past

$$I(X_k; Y_k, \hat{\alpha}_k) \leq I(X; Y) + I(\alpha_k; \hat{\alpha}_k) \text{ [Lapidoth and Shamai]}$$

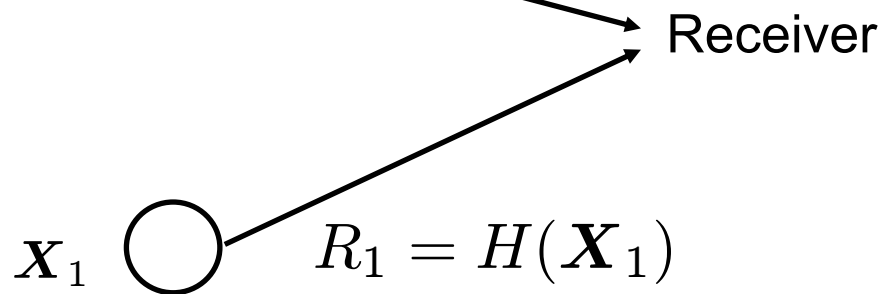
$$I(\alpha_k; \hat{\alpha}_k) = \log \frac{E[|\alpha_k|^2]}{E[|\alpha_k - E[\alpha|\hat{\alpha}_k]|^2]}$$

$I(\alpha_k; \hat{\alpha}_k)$  increases nearly logarithmically for practical SNR

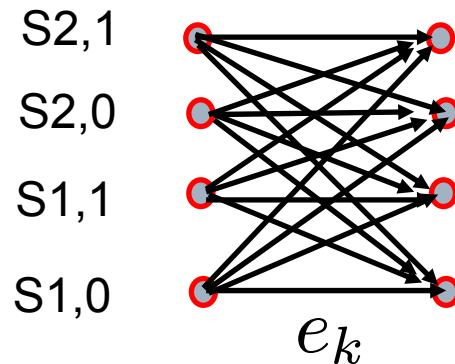
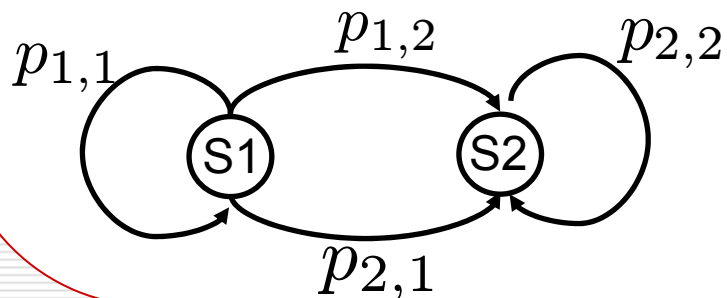
# Compression of Correlated Sources

We want to compress  $X_2$  to  $R_2$  bits

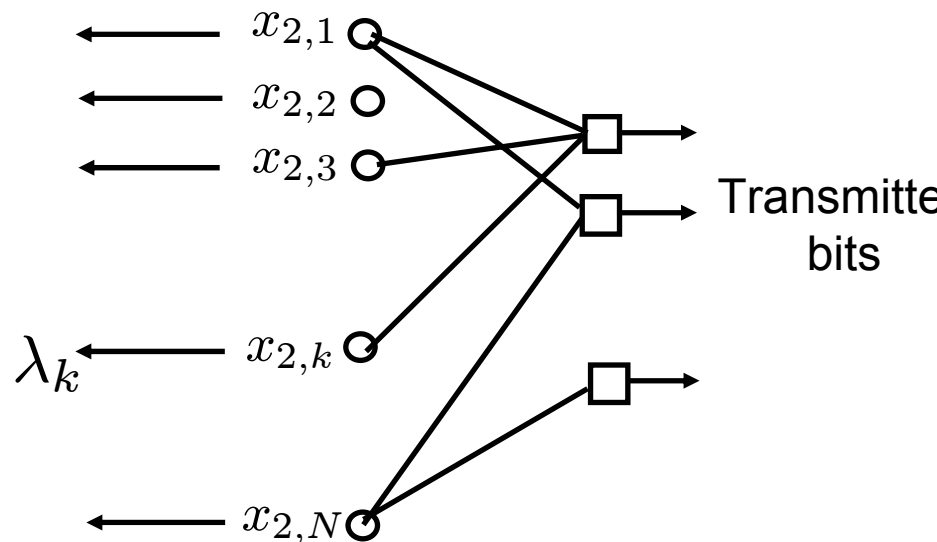
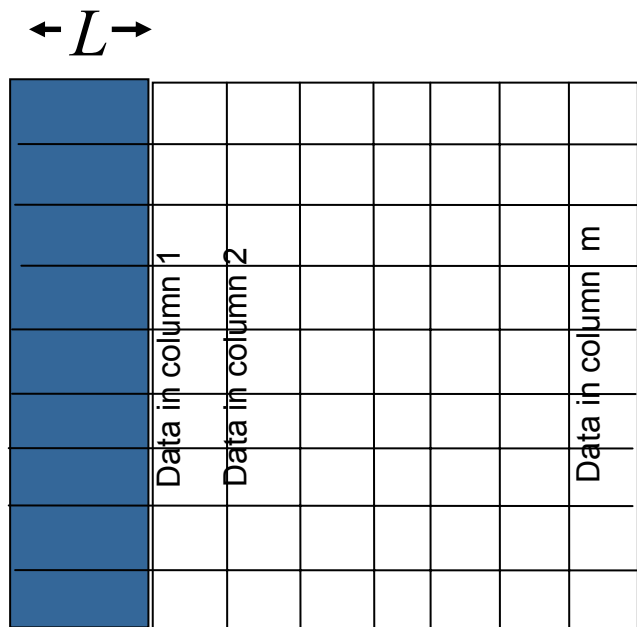
$$X_2 \quad R_2 = H(X_2|X_1) = H(e)$$



$$x_{2k} = x_{1k} \oplus e_k \quad e_k \text{ s are the output of a HMM}$$



# Compression using LDPC codes



$$\begin{aligned}
 k^{\text{th}} \text{ row: } e_{k,1}, e_{k,2}, \dots, e_{k,L} &\rightarrow \lambda_{k,L+1} = P(e_{k,L+1} | e_{k,1}, \dots, e_{k,L}) \\
 &= P(x_{2k,L+1} | e_{k,1}, \dots, e_{k,L})
 \end{aligned}$$

Compression achieved  $R = H(e_{L+1} | e_1, e_2, \dots, e_L)$

Possible to achieve  $R = \frac{1}{N} \sum_{i=1}^N H(e_i | e_1, \dots, e_{i-1}) \rightarrow H(\mathbf{e})$

# Numerical Results

- Existing Approach is Iterative [Garcia Frias et al]
  - Iterate between the HMM and the LDPC decoder
  - Makes the code design problem quite difficult
  
- Example,

Model	$H(X_2 X_1)$	$\Delta_{DF}$	$\Delta_{Turbo}$
[0.01,0.065,0.95,0.925]	0.52	0.03	0.08
[0.97,0.967,0.93,0.973]	0.45	0.03	0.11
[0.99,0.989,0.945,0.9895]	0.28	0.03	0.13

- DF structure is more universal



# Conclusion

- Capacity achieving coding schemes for channels with memory
  
- Iterative processing vs Decision Feedback Processing
  - Both can be used to achieve near capacity performance
  - DF is universal and simplifies the code design process
  - Rateless codes for channels with memory
  - Latency is a problem (it may not be insurmountable)
  
- It is not true that signal processing becomes harder as the codes get stronger
  
- Put the complexity in the decoder and use DF

# Contact Information

- ❑ I am in Room 018 until Mid May
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