

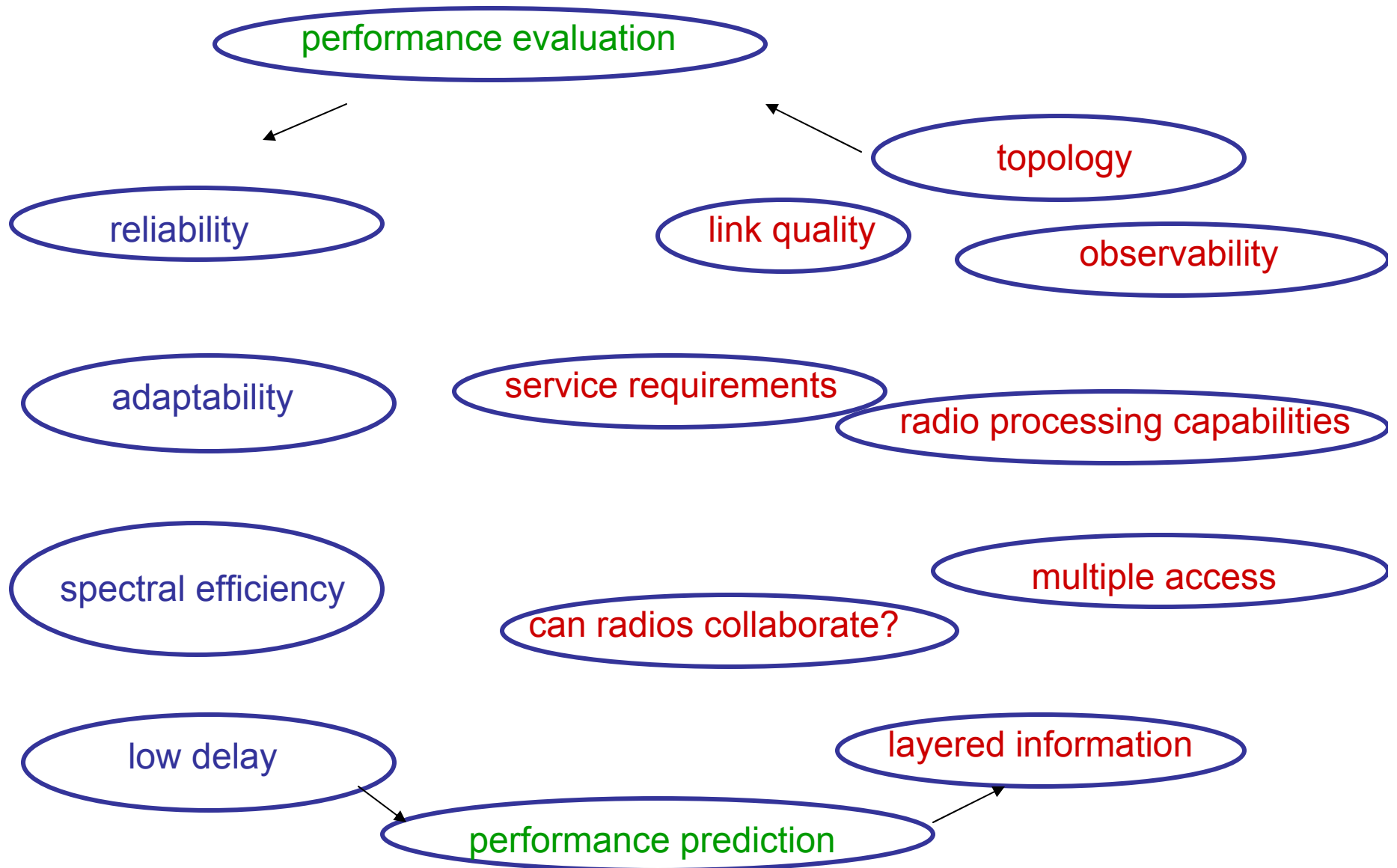
(Error Control) Coding for Wireless Networks

Predrag Spasojevic

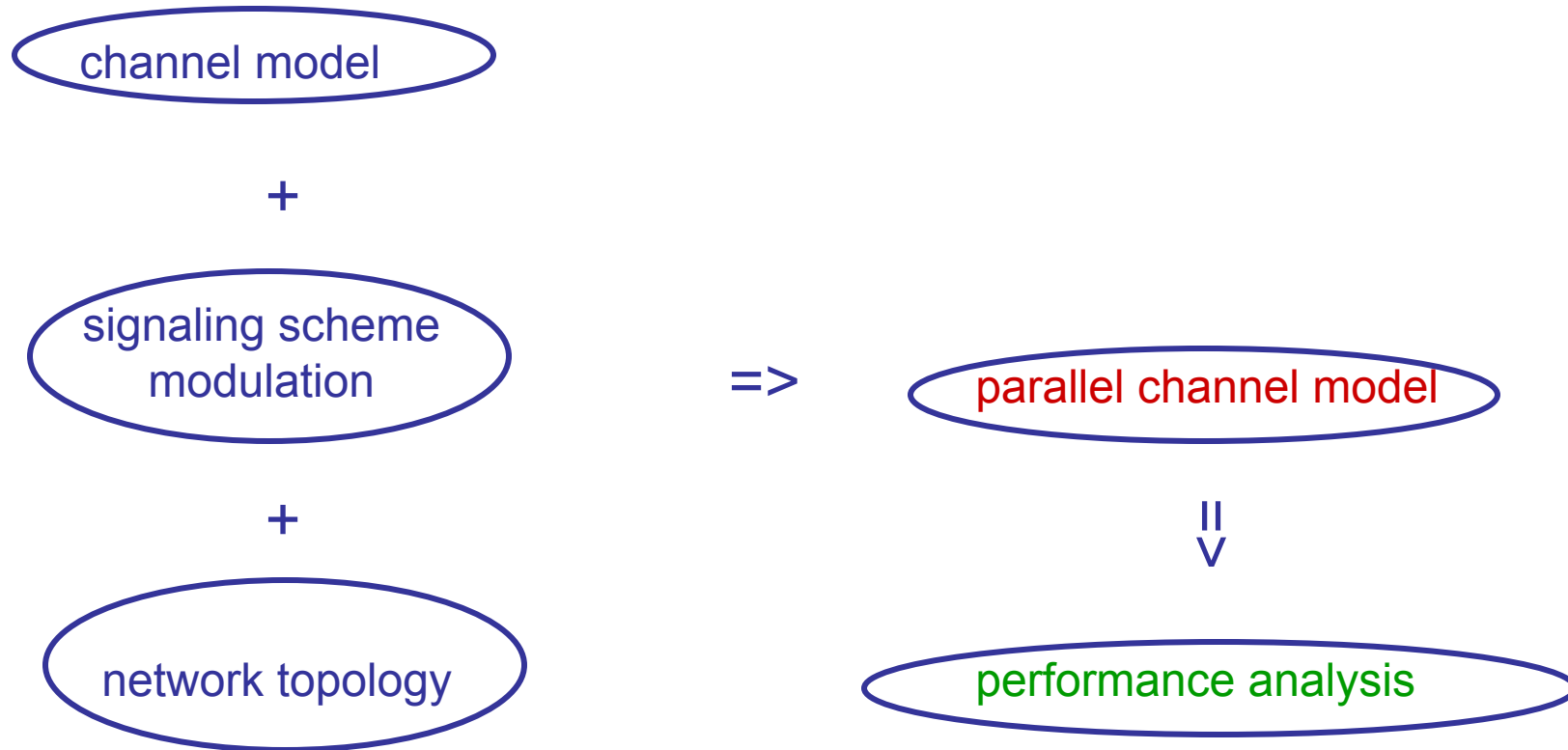
WINLAB, Rutgers University

IAB 2005

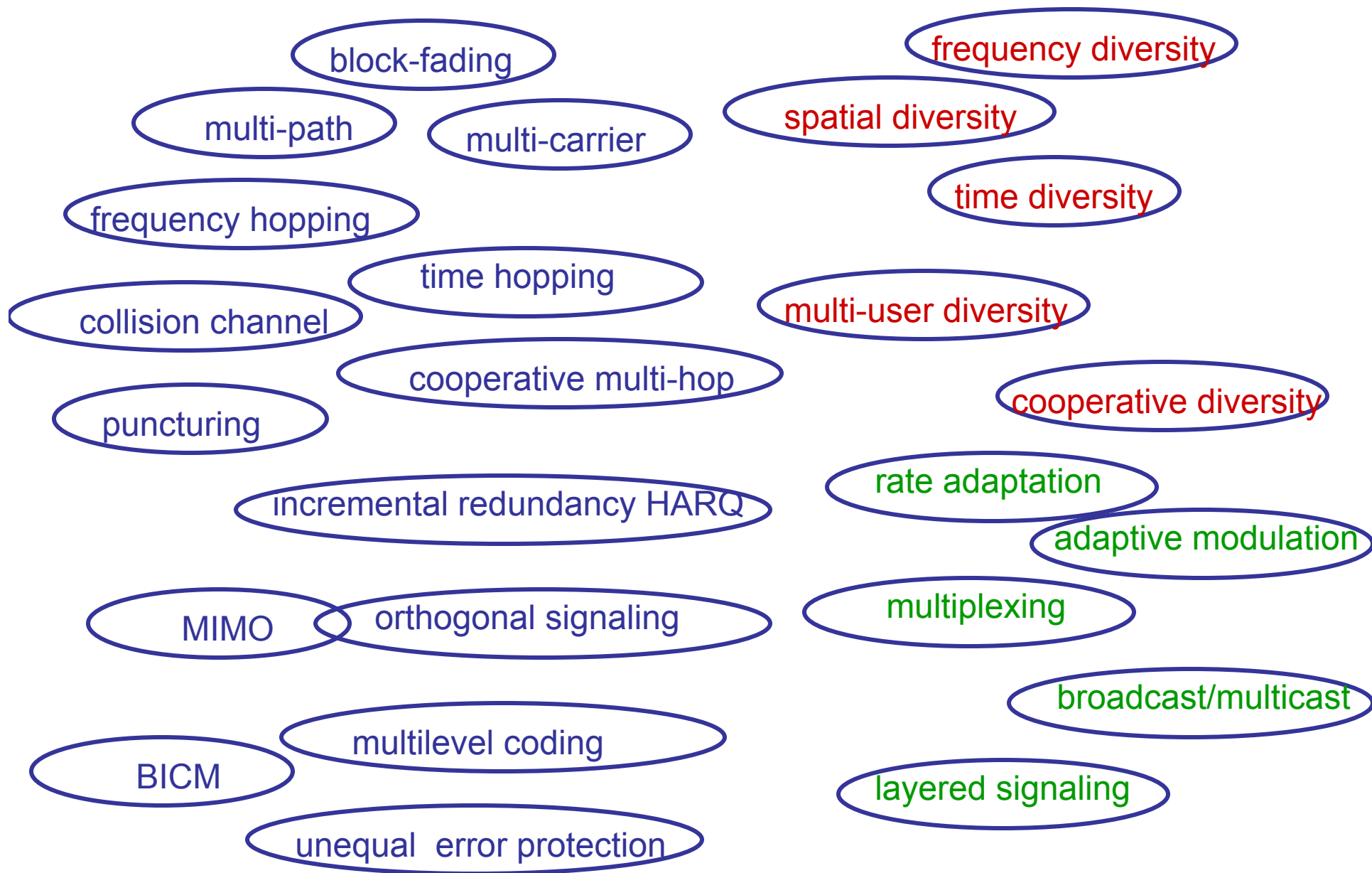
Wireless Networks: Performance Analysis



Parallel Channel Modeling



Parallel Channels: Models



Parallel Channels: Performance of Good Codes for

block-fading

puncturing

cooperative multi-hop

cooperative diversity

incremental redundancy HARQ

rate adaptation

adaptive modulation

BICM

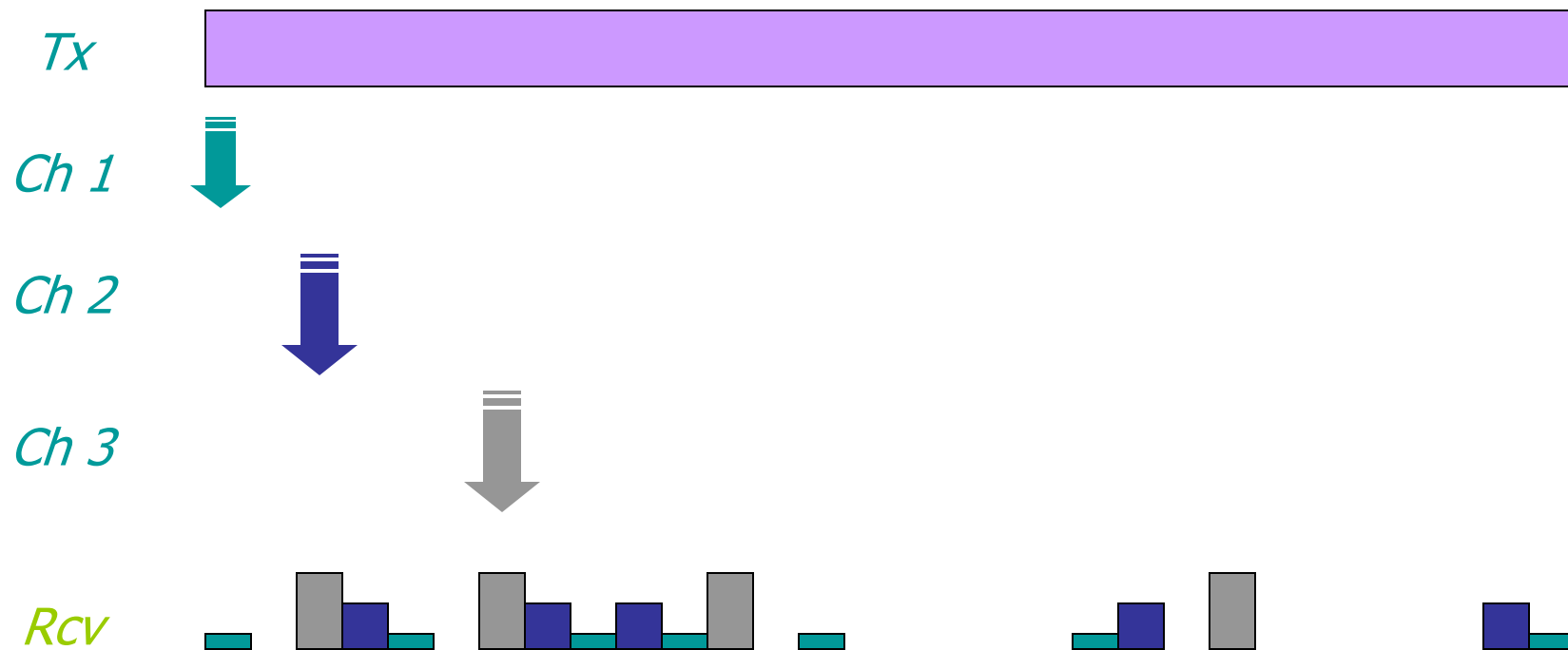
multilevel coding

broadcast/multicast

layered signaling

Codeword transmission over parallel channels

mother codeword (n bits)



On the Performance of Good Codes over Parallel Channels

Ruoheng Liu, Predrag Spasojevic
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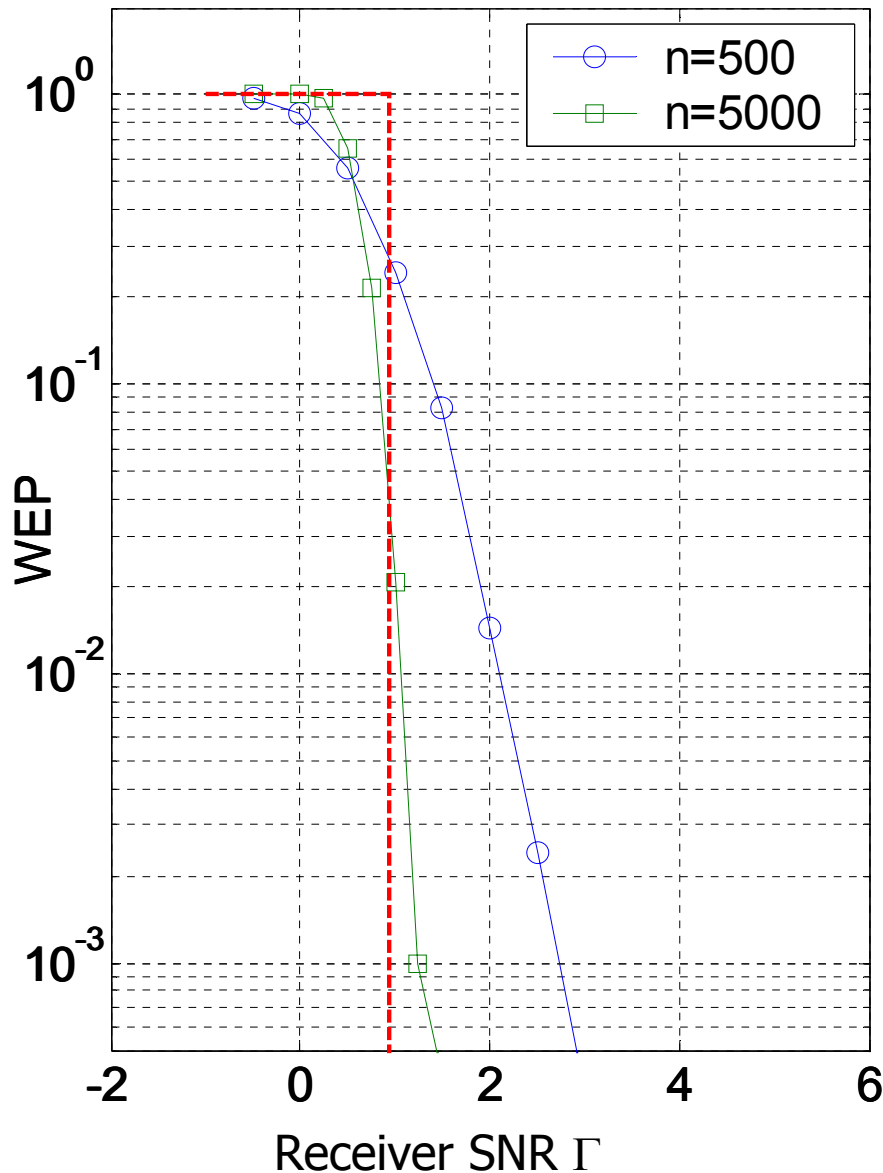
This work has been supported in part by the NSF Grant SPN-0338805.

What is a “good” code?

- good codes: MacKay 99
 - a sequence of binary linear codes
 - achieve arbitrarily small word error probability (WEP) over a noisy channel at a nonzero threshold rate.
 - include turbo codes, LDPC codes, and RA codes
- capacity achieving codes
 - good codes
 - rate threshold is equal to the channel capacity

$$X = \{X(n_i)\}_{i=1}^{\infty}$$

Threshold behavior of good codes



- an example of turbo codes
 - $R=0.7$
 - $n=500, 5000$
 - binary input AWGN channel

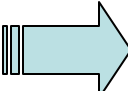
- code goodness implies there exists a threshold Γ_0

$$\lim_{n \rightarrow \infty} P_W^{X(n)}(\Gamma) = 0, \quad \Gamma > \Gamma_0$$


□ Γ : received SNR.

Code Goodness (Liu et al. 2004)

codebook design requirement
(transmitting a codeword \mathbf{x})

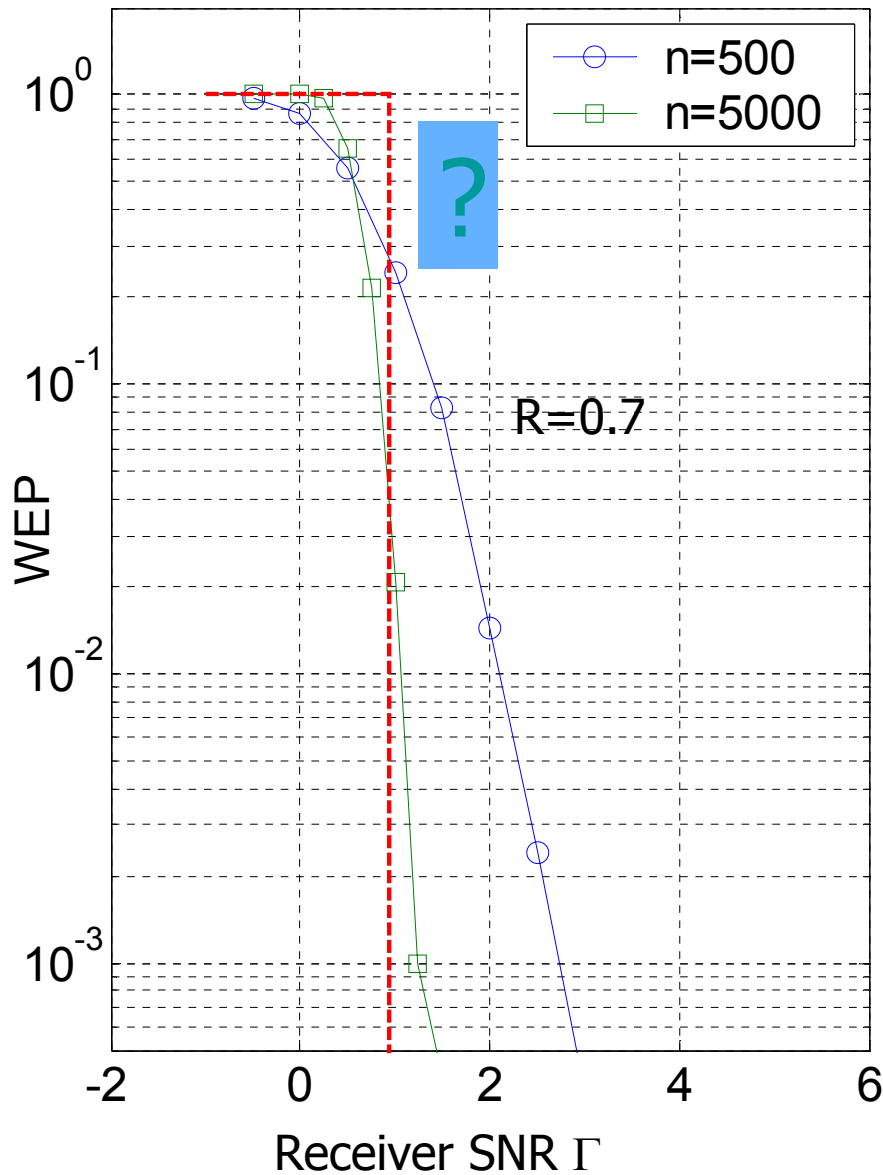
– distance from \mathbf{x} to other codewords is large 

$$d_{\min}^{X(n)} \rightarrow \infty$$

– the number of \mathbf{x} 's neighbors is small 
(low weight spectrum slope of a good code)

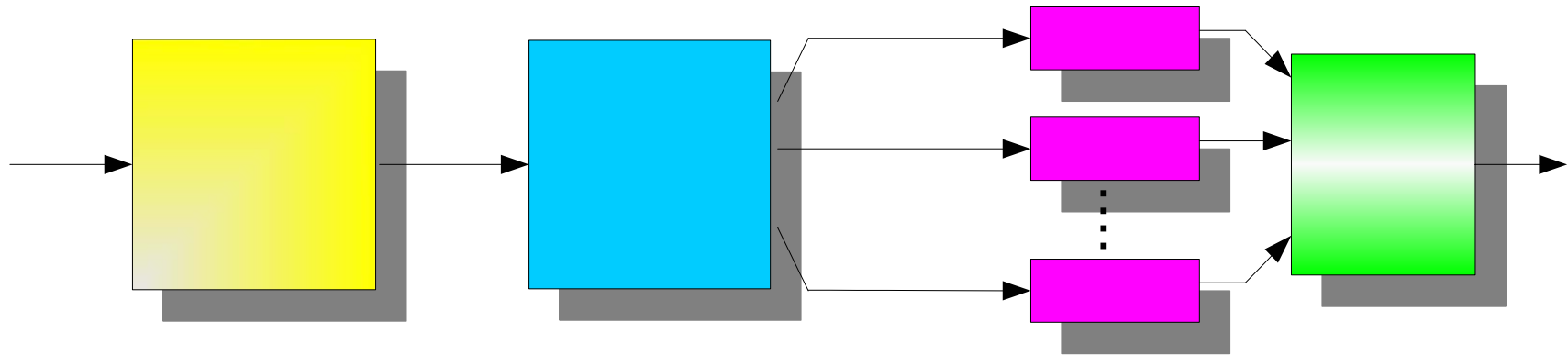
$$\limsup_{n \rightarrow \infty} S^{X(n)}(f_n) < \infty$$

Threshold calculation



- single channel
 - Richardson and Urbanke iterative decoding
 - Jin, McEliece, et. al. typical pair decoding
 - Sason and Shamai maximum likelihood (ML) decoding
 - Ashikhmin, et. al. (Exit chart)
- parallel channel model?

Parallel channel model

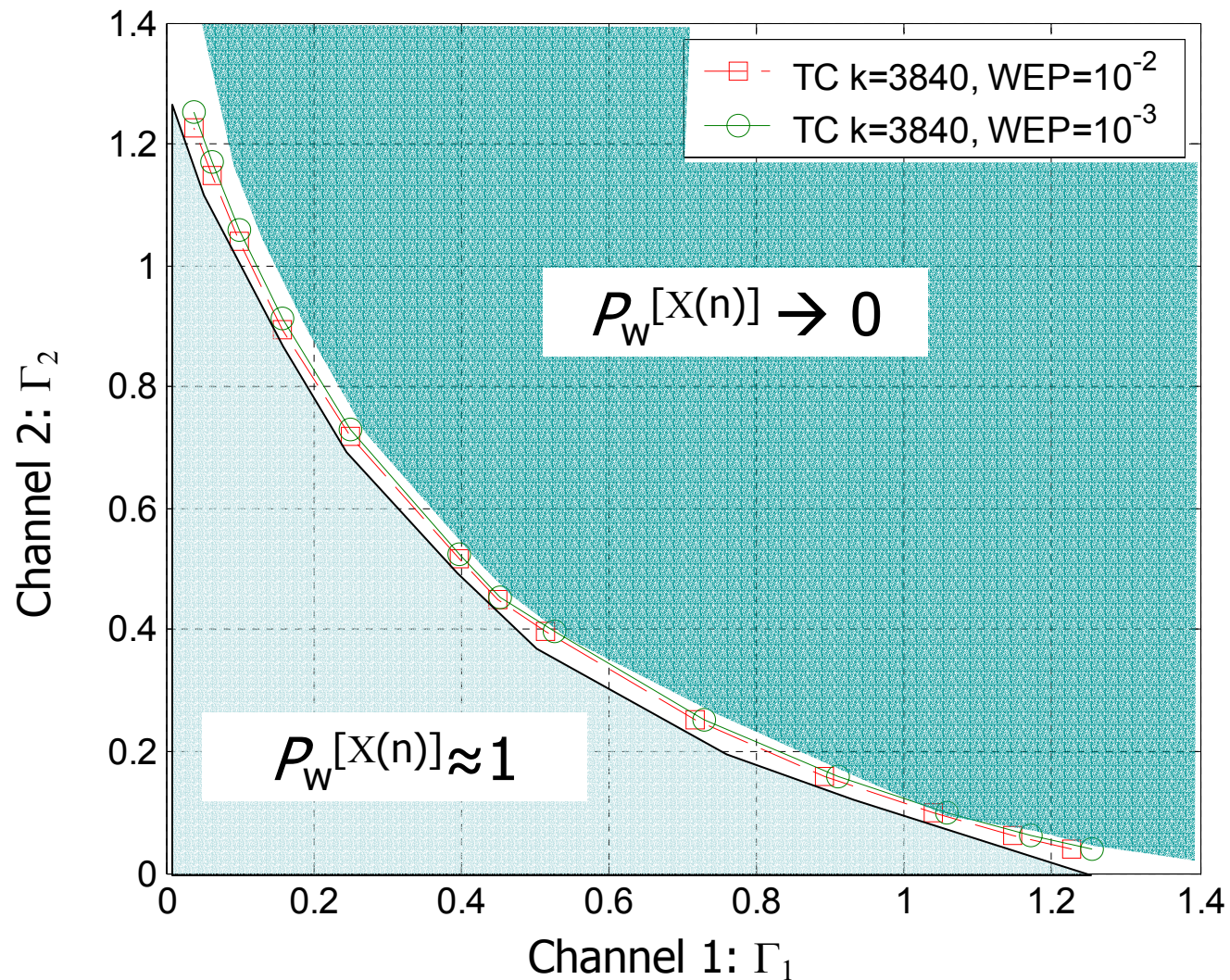


□ α_j : (asymptotic) assignment rate $\sum_{j=1}^J \alpha_j = 1$

- codeword is partitioned and transmitted over parallel channels
- under what channel conditions will the communication be reliable?

Reliable channel region

Turbo codes with $R=1/3$, two parallel AWGN channel, $\alpha_1=\alpha_2=1/2$



Union Bhattacharyya (UB) threshold

UB reliable channel region

effective
Bhattacharyya distance

if

$$-\ln \bar{\gamma} > c_0^{[X]}$$

then average ML decoding WEP $\lim_{n \rightarrow \infty} P_W^{[X(n)]} = 0$

- $c_0^{[X]} = \max_{0 \leq \delta \leq 1} \limsup_{n \rightarrow \infty} \frac{r^{[X(n)]}(\delta)}{\delta}$ UB threshold
- $r^{[X(n)]}(\delta) = \frac{\ln \bar{A}^{[X(n)]}(\lfloor \delta n \rfloor)}{n}$ normalized weight spectrum
- $\bar{\gamma} = \sum_j \alpha_j \gamma_j$ average Bhattacharyya noise parameter

Two code parameter description

if $-\ln \bar{\gamma} > c_P^{[X]}$ and $\bar{I} > R + \zeta_P^{[X]}$

then average ML decoding word error probability

$$\lim_{n \rightarrow \infty} P_W^{[X(n)]} = 0.$$

$$\bar{I} = \sum_{j=1}^J \alpha_j I_j \text{ average channel mutual information}$$

Single parameter description: simple threshold (Liu et al. 2004)

$$-\ln \bar{\gamma} > c_*^{[X]}$$

if

then the average ML decoding word error probability

$$\lim_{n \rightarrow \infty} \overline{P_W^{[X](n)}} = 0$$

$$c_*^{[X]} = \min_P \left\{ c_P^{[X]} : 1 - \exp(-c_P^{[X]}) \geq R + \zeta_P^{[X]} \right\}$$

UB threshold vs simple threshold

- an example of turbo code
($R=1/7$, $k=768$, and $J=3$ RSC encoders)

– UB threshold $c_0^{[X]} = 0.21 \Rightarrow -6.77dB$

– simple threshold $c_*^{[X]} = 0.17 \Rightarrow -7.70dB$

Puncturing and Block Fading Channel

- **punctured** simple code threshold

$$c_*^{[C_P]}(\tau) = \ln \frac{\tau}{\exp(-c_*^{[C]}) + (1 - \tau)}$$

- $1 - \tau$: punctured rate

- simple threshold bound on **block fading channel** coding

$$\begin{aligned} \overline{P}_w^{[X(n)]}(\bar{\gamma}) &= \mathbf{E}\left[P_w^{[X(n)]}(\bar{\gamma})\right] \\ &\leq P\left\{-\ln \bar{\gamma} \leq c_*^{[X]}\right\} + o(1) \end{aligned}$$

- decoding is done with full channel state information (CSI)

Adaptive Modulation for Variable-Rate Turbo-BICM

Ruoheng Liu, Jianghong Luo and Predrag Spasojevic

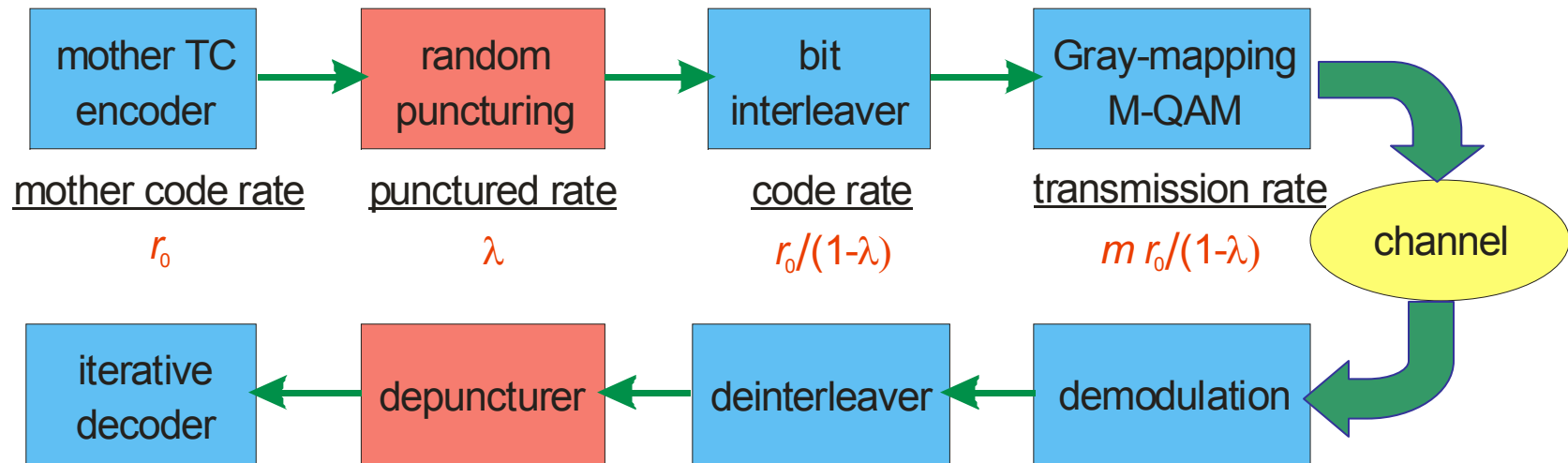
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This work has been supported in part by the NSF Grant SPN-0338805.

Motivation

- **key** requirements in **4G** or **B3G** communications systems:
 - ❖ a wide range of **data rates** according to economic and service demands
 - ❖ **QoS** for packet oriented services
- **high-speed** wireless systems: **bandwidth efficient** turbo coding scheme
 - ❖ Turbo **BICM** -- *Goff 94'* [simplicity]
 - ❖ Parallel concatenated **TCM** -- *Benedetto 95'*
 - ❖ Turbo-**TCM** -- *Robertson, 98'*
- **channel fluctuating** in the wireless propagation environment
 - ❖ communication reliability and error prediction
 - ❖ lack of closed-form expressions for error probability of turbo coded modulation

System model



Rate threshold for VR-Turbo-BICM

Theorem:

For a VR-Turbo-BICM coding scheme using a mother code ensemble $[C]$ of rate r_0 employed over an AWGN channel with a channel SNR ρ we define the rate threshold

$$r^\circ(\rho, m) = \begin{cases} 0 & \rho < \eta(m) \\ b(m) \cdot m\bar{I}(\rho, m) & \eta(m) \leq \rho \leq \zeta(m) \\ r_m^\circ [1 - \bar{\gamma}(\rho, m)] & \rho > \zeta(m) \end{cases}$$

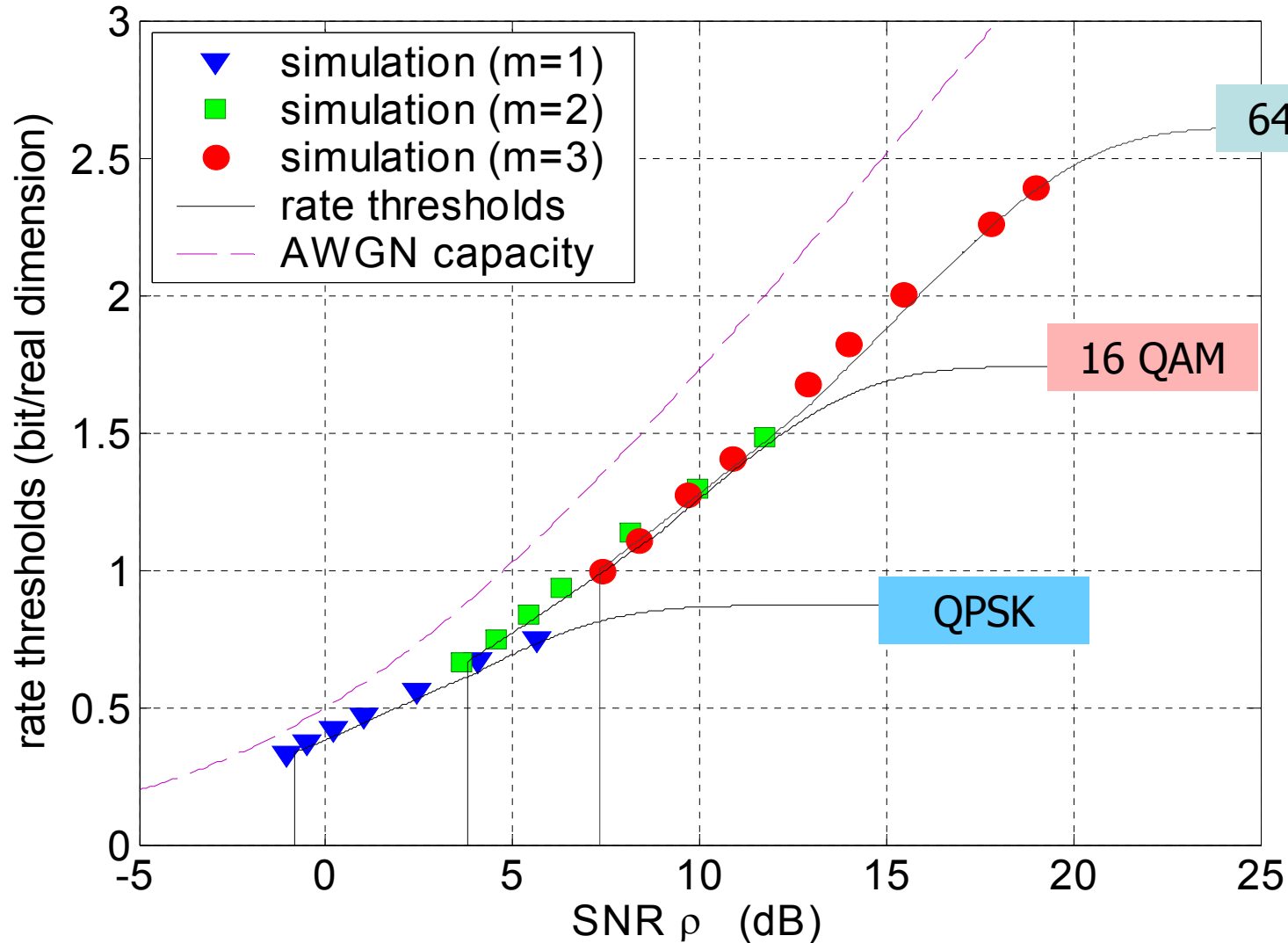
If

$$r \leq r^\circ(\rho, m)$$

then the average ML decoding word error probability approaches zero.

Rate Threshold vs SNR

$r^\circ(\rho, m)$ vs. simulation results (FER=0.01)

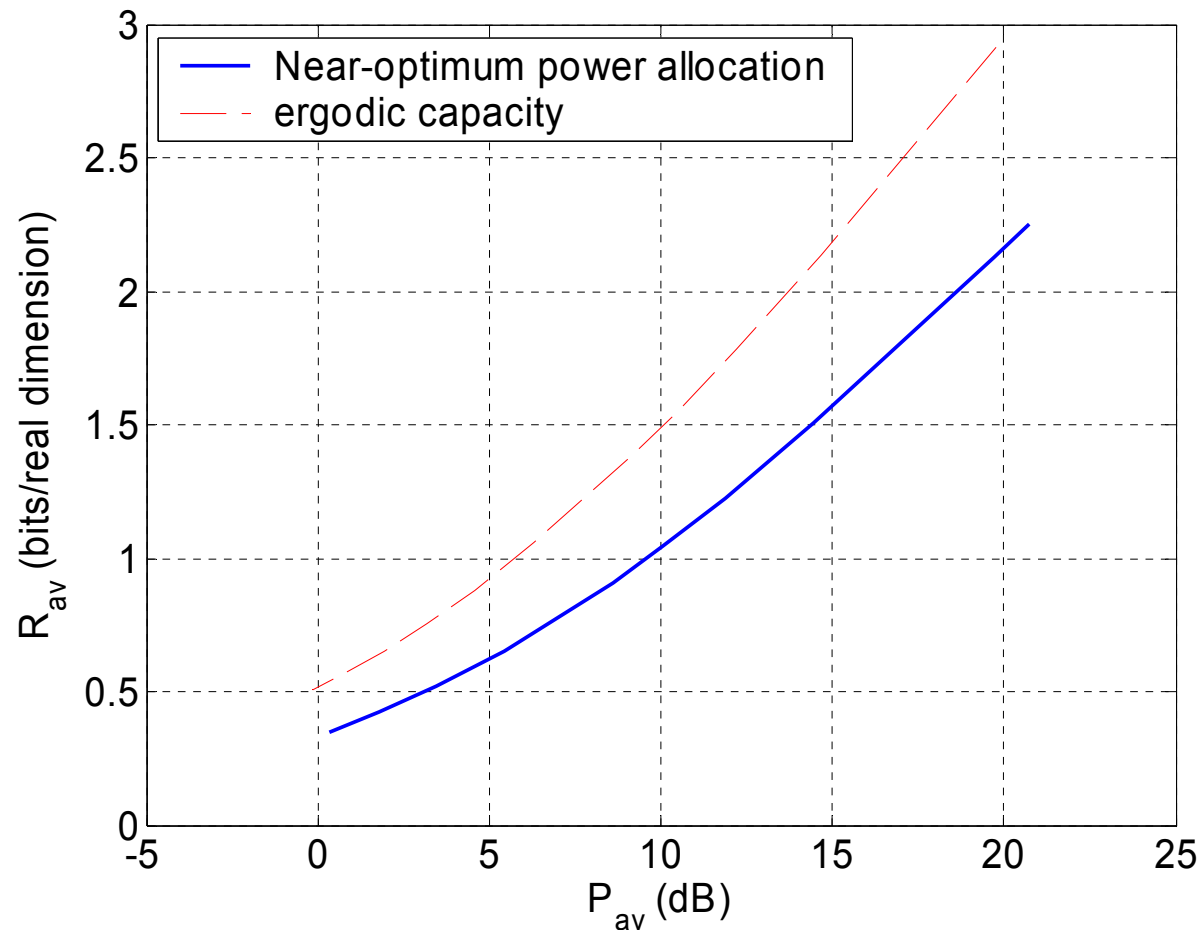


Adaptive Turbo-BICM in slow fading

Allocation Problem:

given an average power constraint P_{av} , the optimum power and modulation index maximize the expected rate threshold

$$\begin{aligned} & \max_{P(n), m(n)} E[r^\circ(P(g), m(g))] \\ & \text{subject to } E[P(g)] \leq P_{av} \\ & P(g) \geq 0, m(g) \in \{0, 1, 2, 3, \dots\}. \end{aligned}$$

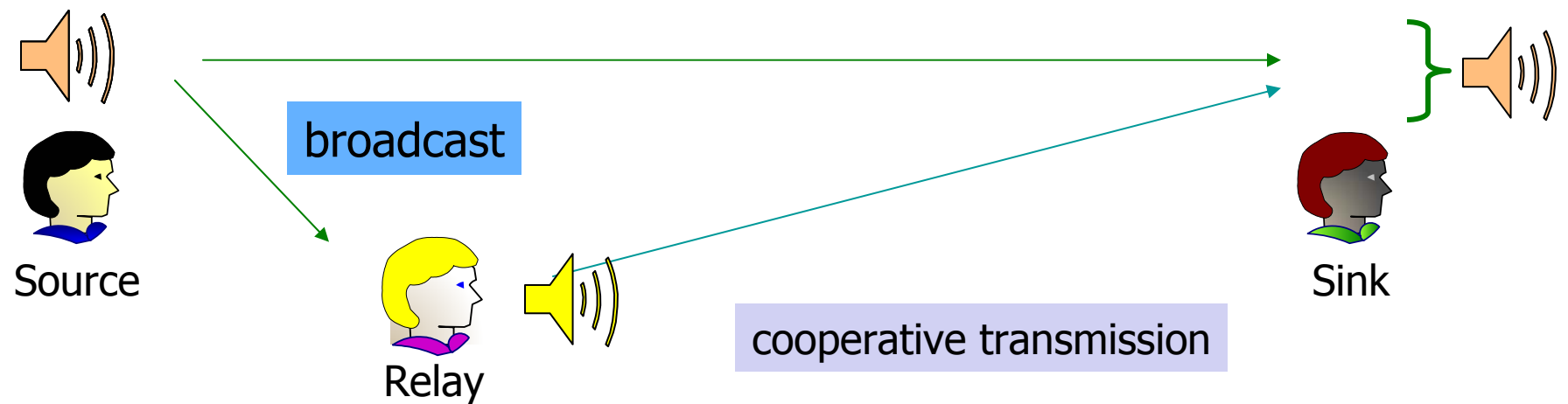
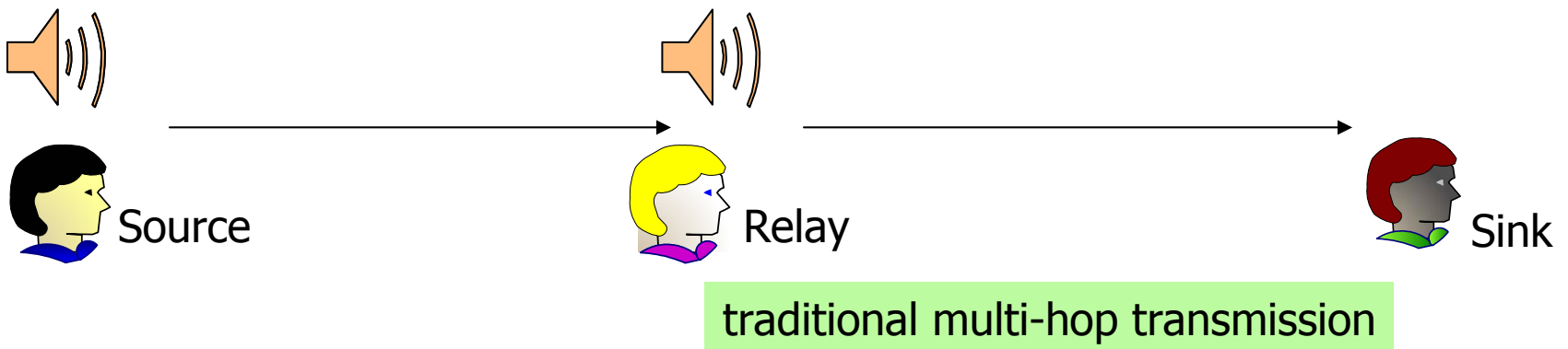
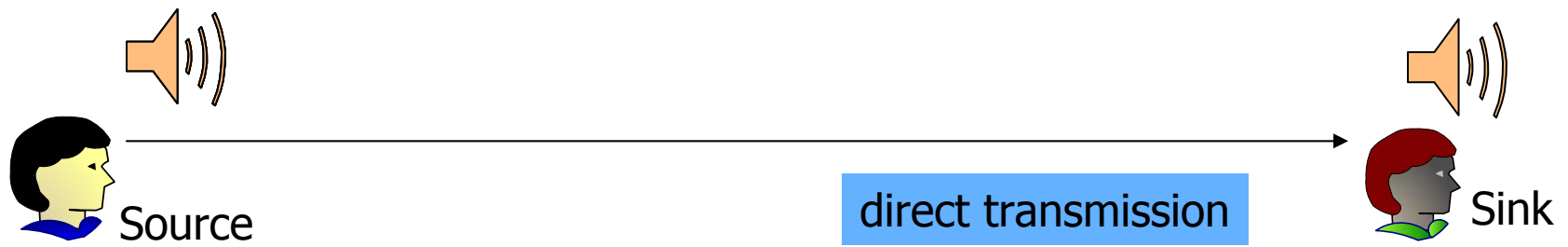


Cooperative Diversity with Incremental Redundancy (IR) Turbo Coding for Quasi Static Wireless Networks

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WINLAB, Rutgers University

Emina Soljanin
Bell Labs, Lucent

This work has been supported in part by the NSF Grant SPN-0338805.



Cooperation benefit: reliability

wireless communications

Reliable transmission

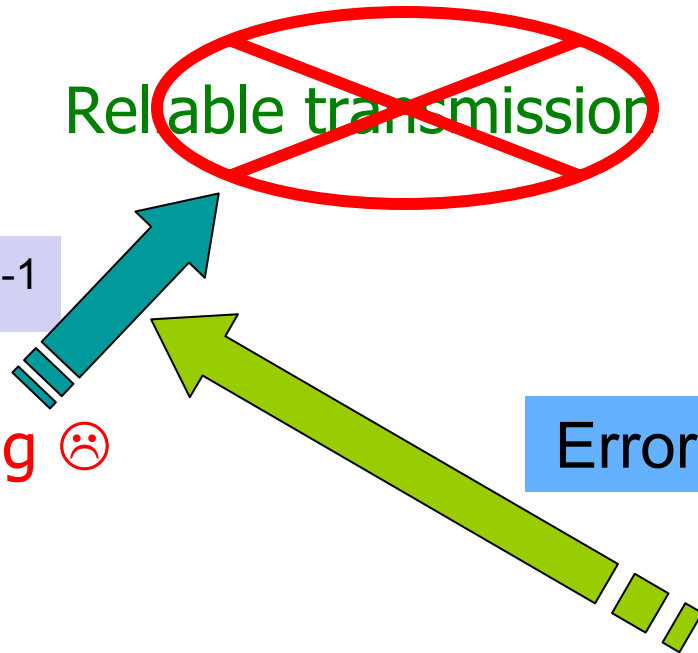


Error rate $\sim \text{SNR}^{-1}$

Deep fading ☹️

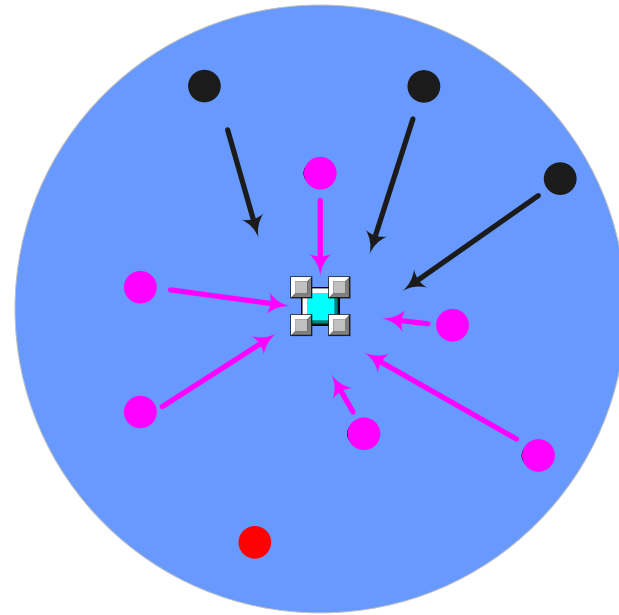
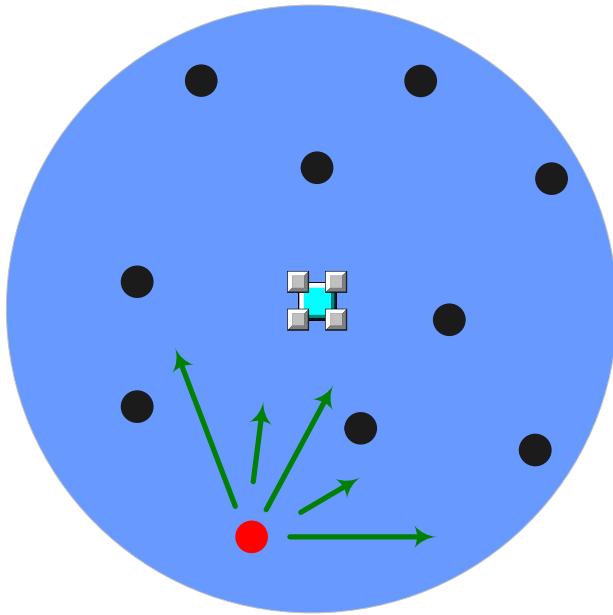
Error rate $\sim \text{SNR}^{-2}$

Diversity 😊



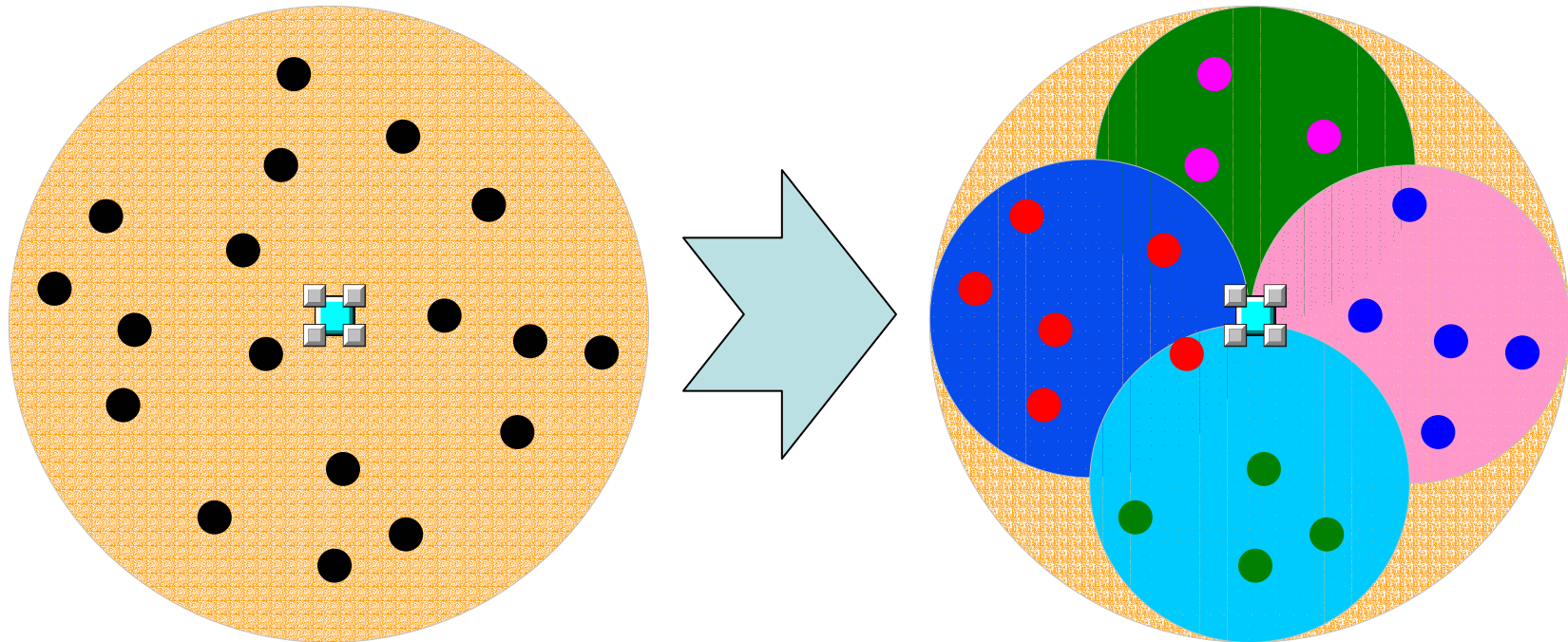
M-user Cooperation

- M users share radio channel
- orthogonal frequency-division multiple access scheme
- 2 Hops scheme (for each user)

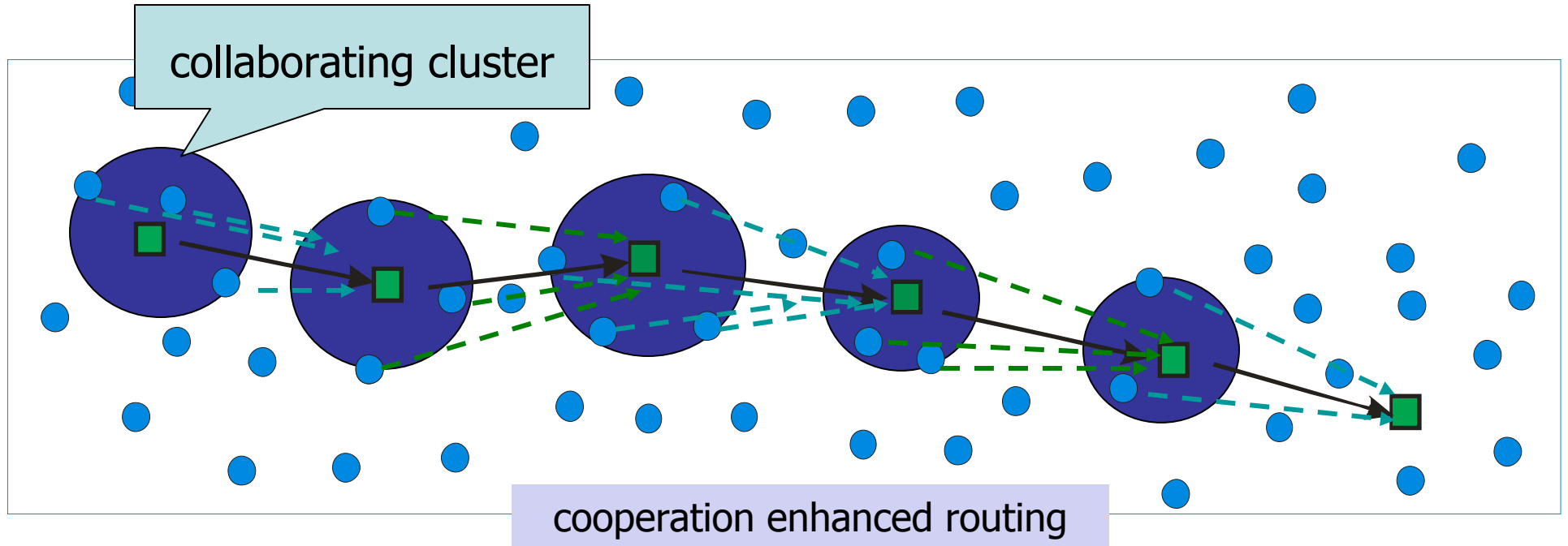


M-user Cooperation (cont.)

- Large SNR (asymptotic result)
 - Using cooperative coding each user can achieve **full (M) diversity gain**
- Medium and low SNR (how to get **benefit** ? cooperation criteria ...)
 - the user-to-destination channel quality is good (same to two user case)
 - partners are sufficiently close (cluster behavior)



Wireless cooperative routing in networks with quasi-static fading



Frame error rate of the cooperative routing scheme

- **simple threshold** upper bound

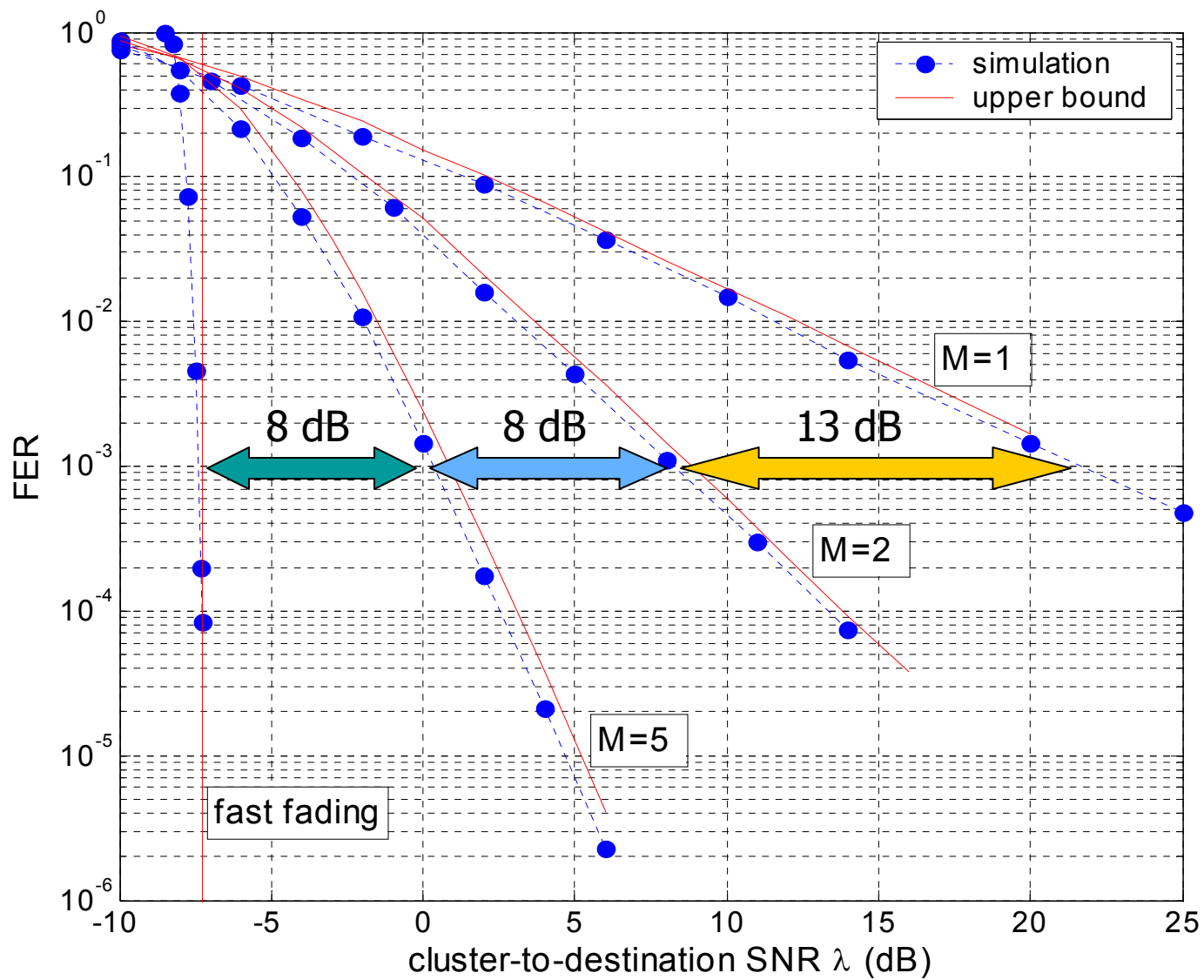
$$\text{FER}^{(M)} \leq \sum_{k=0}^{M-1} \binom{M-1}{k} P(|\mathbf{F}| = k) \cdot P\{\theta(\mathbf{F}) \leq c_*^{[\text{TC}]} \mid |\mathbf{F}| = k\}$$

– $\theta(\mathbf{F})$: effective cluster-to-destination SNR

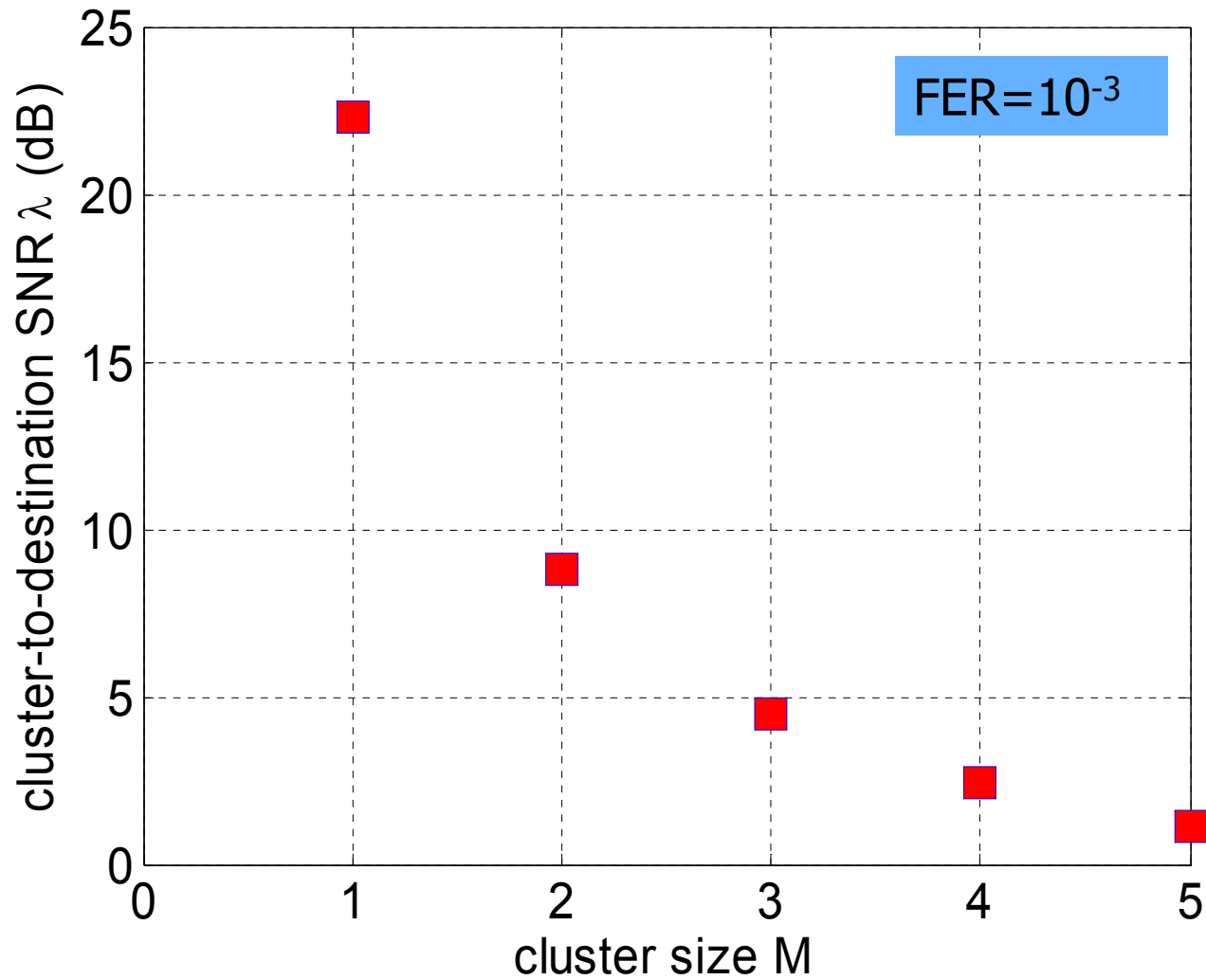
- **asymptotic** upper bound (small $c_*^{[\text{TC}]}$ and large ρ, λ)

$$\text{FER}^{(M)} \leq_{\lambda, \rho, c_*^{[\text{C}]}} \sum_{k=0}^{M-1} \binom{M-1}{k} \frac{(M \cdot c_*^{[\text{TC}]})^M}{(M-k)(k+1)!} \rho^{-M+(k+1)} \lambda^{-(k+1)}$$

Diversity Gain vs M



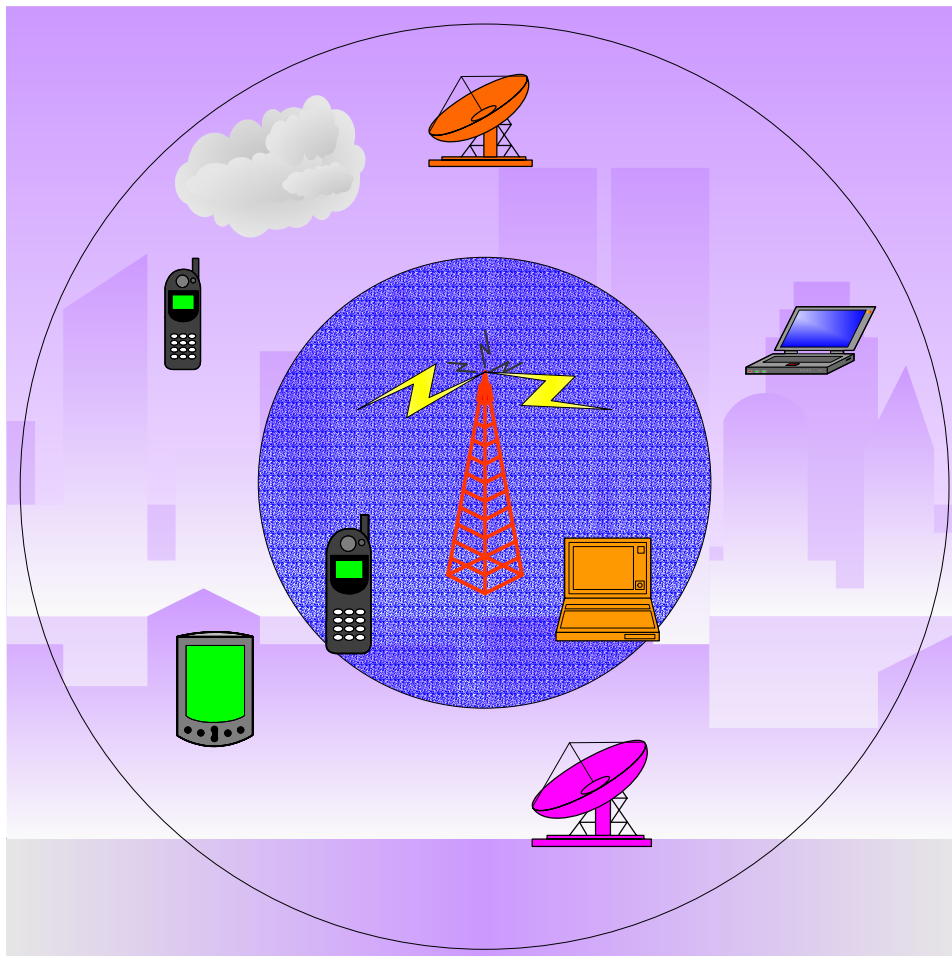
Cluster-to-destination SNR vs. M



Rate Design for Layered Broadcast using Punctured LDPC Codes and Multilevel Coding

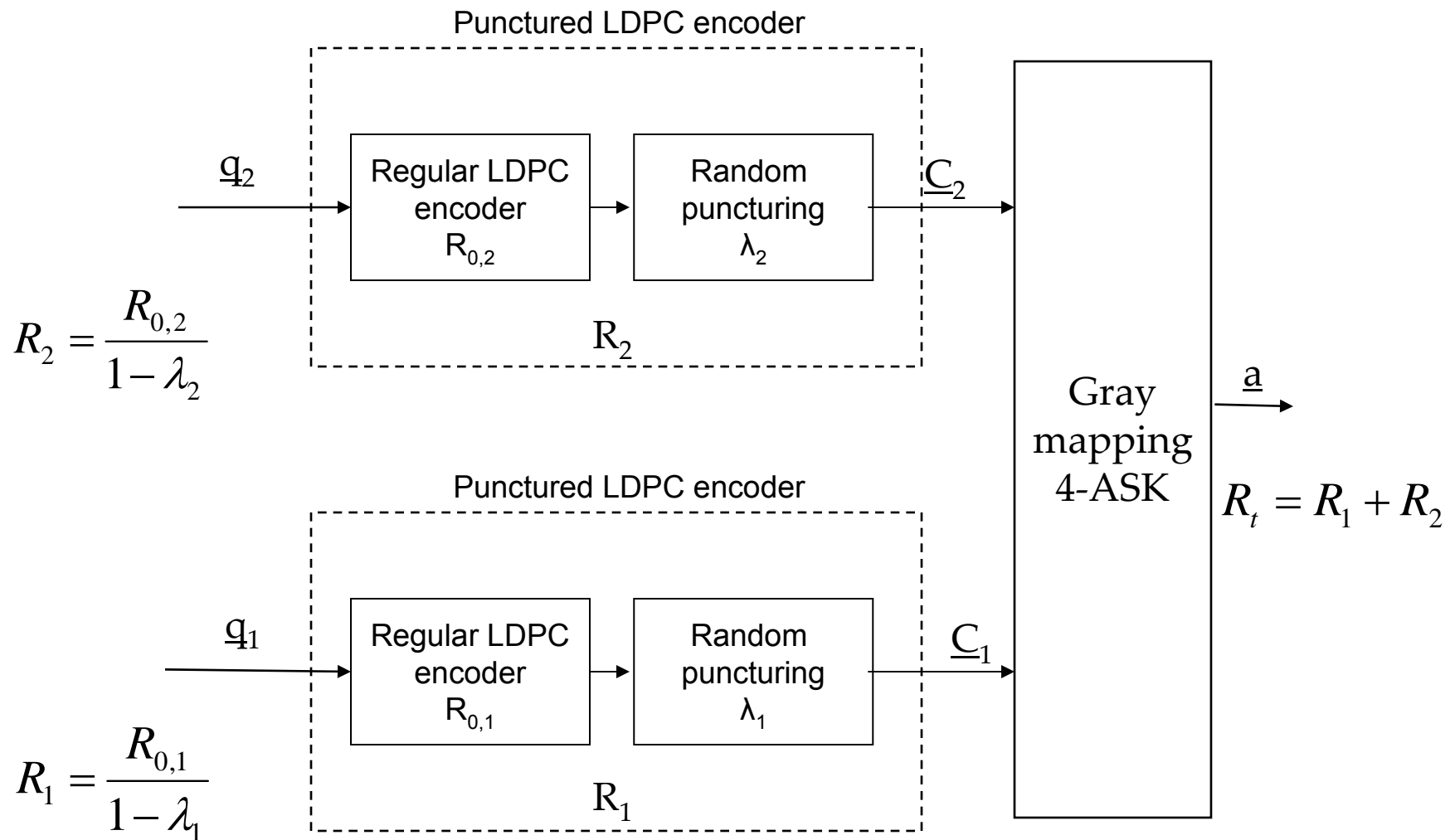
Ahmed Turk and Predrag Spasojević

Coding for Broadcast



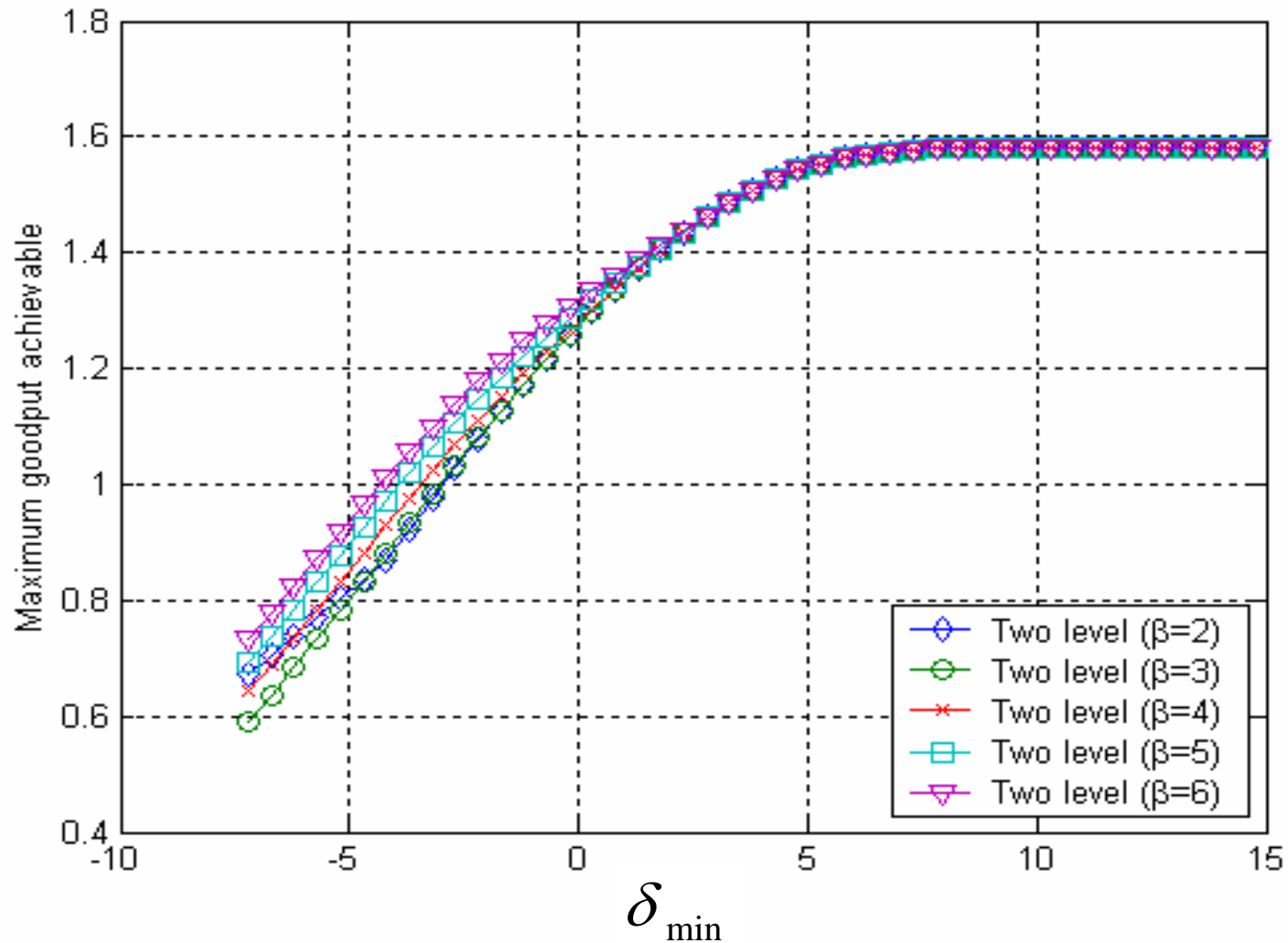
- superposition structure
 - successive refineable sources
 - hierarchical channel coding
- code rate selection
 - different channel conditions
 - unequal error protection

Multilevel Broadcast System Encoder



Goodput vs worst user SNR

(n,3,4) mother code in each level



Incremental Multi-Hop based on Punctured Turbo Codes

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Emina Soljanin

Bell Labs, Lucent

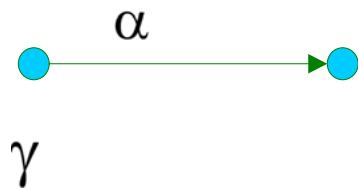
System Model



- one-dimensional multi-hop network with P nodes
- equal distance between the neighboring nodes

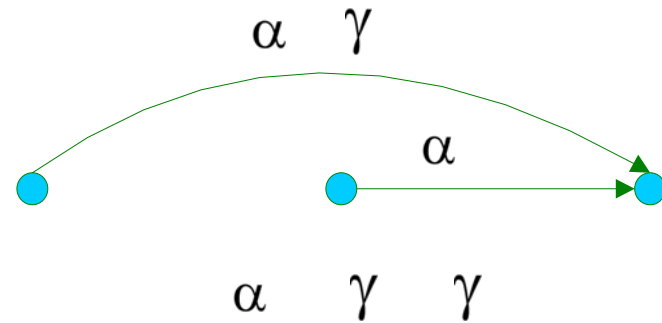
$$d_{1,2} = d_{2,3} = \dots = d_{P-1,P} = d_0$$

Performance Analysis (1)



Step 1

$$\alpha_1 > \frac{1 - \exp(-c_0^{[X]})}{1 - \gamma(1)}$$

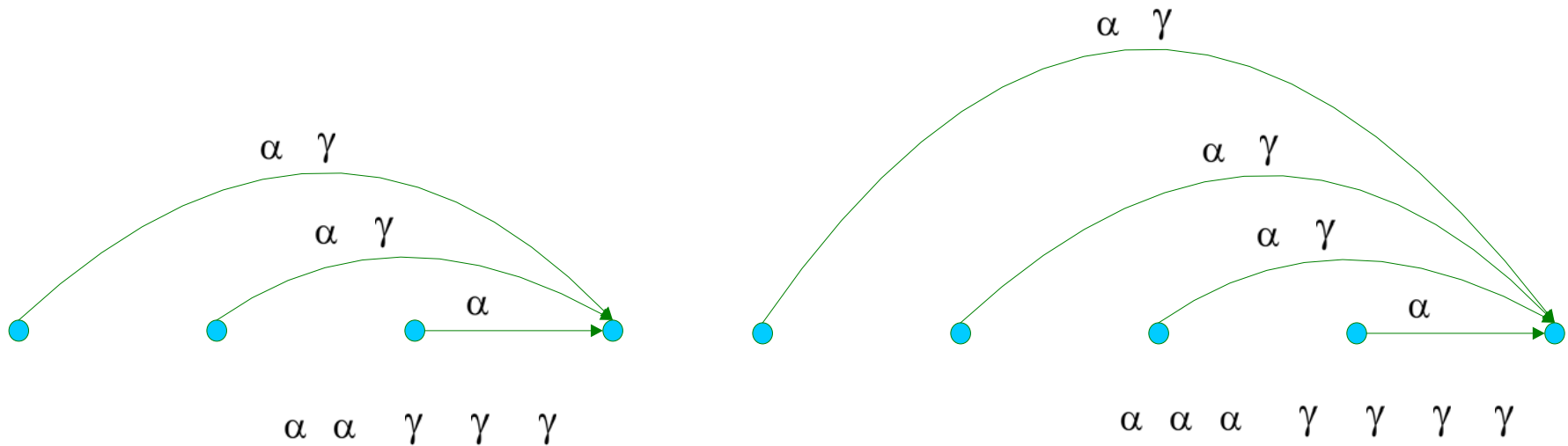


Step 2

Given: α_1

$$\alpha_2 > \frac{1 - \exp(-c_0^{[X]}) - \alpha_1 [1 - \gamma(2)]}{1 - \gamma(1)}$$

Performance Analysis (cont.)



Step 3

Step 4

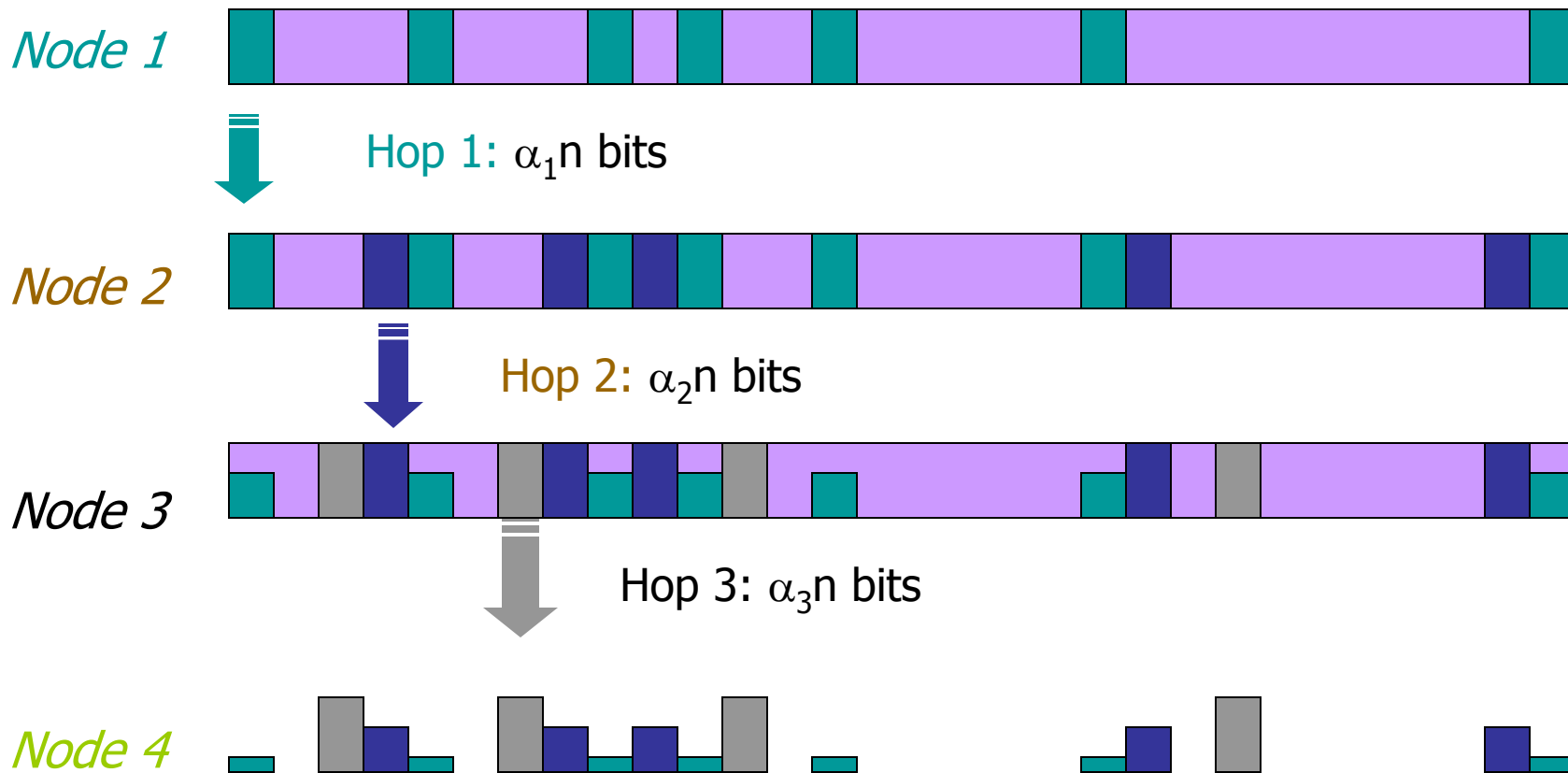
Given: $\alpha_1, \dots, \alpha_{j-1}$

$$\alpha_j > \frac{1 - \exp(-c_0^{[X]}) - \sum_{k=1}^{j-1} \alpha_k [1 - \gamma(j - k + 1)]}{1 - \gamma(1)}$$

((3))

IR multi-hop transmission scheme (2)

mother codeword (n bits)



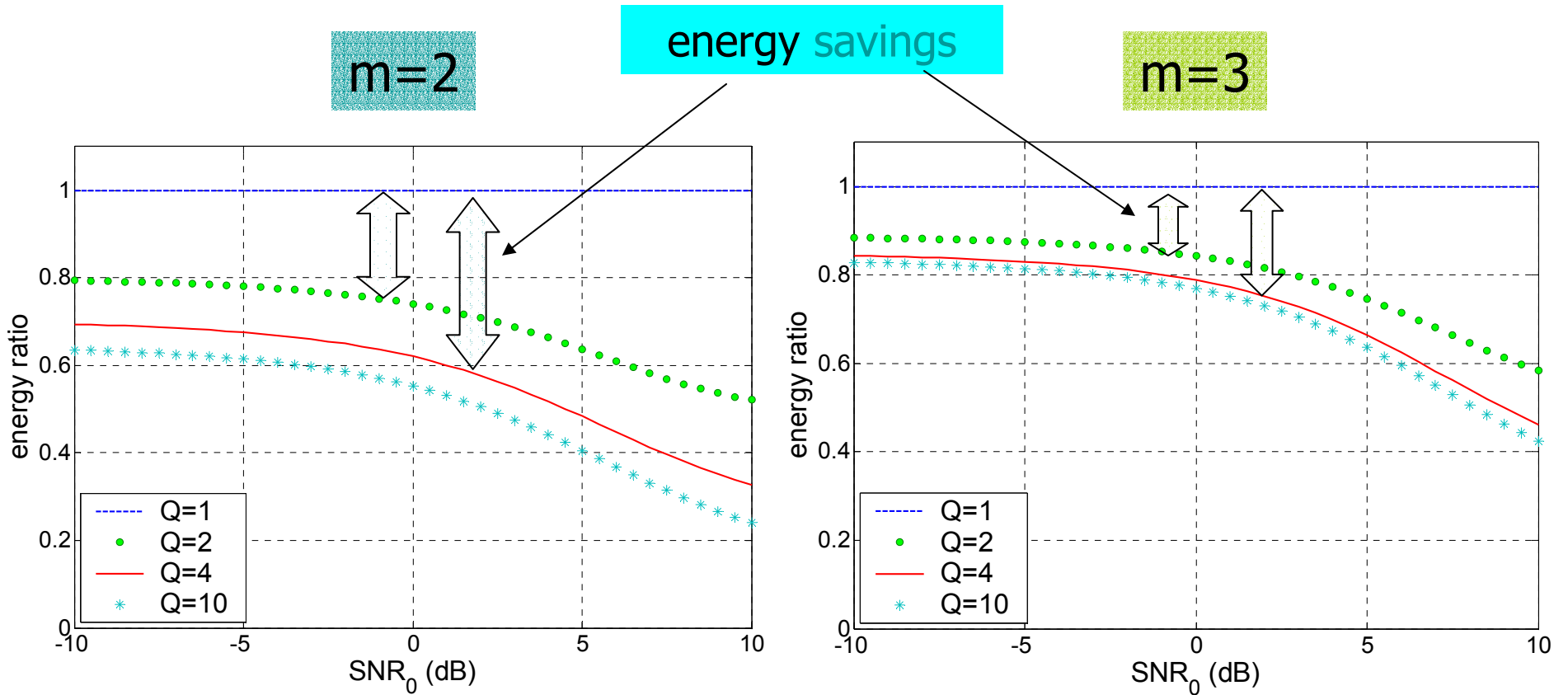
Energy Savings

- energy ratios:
$$\eta = \frac{E_{total}^{(Q)}}{E_{total}^{(1)}} = \frac{\alpha^{(Q)}}{\alpha^{(1)}} = \frac{1 - \gamma(1)}{Q - \sum_{j=1}^Q \gamma(j)}$$

- High SNR_0 :
$$\lim_{SNR_0 \rightarrow \infty} \eta = \frac{1}{Q}$$

- low SNR_0 :
$$\lim_{SNR_0 \rightarrow 0} \eta = \frac{1}{\sum_{j=1}^Q j^{-m}}$$

Traditional vs. IR Multi-hop Transmissions



Repetition Coding vs. IR Schemes

