# Modern Wireless Network Design Based on Constrained Capacity

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## Overview

- Key observations:
  - Capacity approaching binary codes are now practical.
  - M-ary modulation such as PSK, QAM, and FSK continue to be used.
  - Block space time coding is an effective way to modulate across multiple transmit antennas.
- Implications of these observations:
  - It makes sense to study point-to-point links in terms of the capacity under modulation constraints.
  - It is desirable to match binary codes with M-ary modulation.
- Overview of talk:
  - Capacity under modulation constraints.
  - Bit interleaved coded modulation (BICM).
  - BICM with iterative demodulation and EXIT charts.
  - Efficient cross-layer design of retransmission (MAC) and routing (networklayer) protocols.

# Noisy Channel Coding Theorem

- Claude Shannon, "A mathematical theory of communication," Bell Systems Technical Journal, 1948.
- Every channel has associated with it a *capacity* C.
  - Measured in bits per channel use (modulated symbol).
- The channel capacity is an upper bound on *information rate* r.
  - There exists a code of rate r < C that achieves reliable communications.
    - Reliable means an arbitrarily small error probability.

# Computing Channel Capacity

The capacity is the *mutual information* between the channel's input X and output Y maximized over all possible input distributions:

$$C = \max_{p(x)} \{I(X;Y)\}$$
$$= \max_{p(x)} \left\{ \iint p(x,y) \log_2 \frac{p(x,y)}{p(x)p(y)} dx dy \right\}$$

## Capacity of AWGN with Unconstrained Input

- Consider an AWGN channel with 1-dimensional input:
  - y = x + n
  - where n is Gaussian with variance  $N_o/2$
  - x is a signal with average energy (variance)  $E_s$
  - The capacity in this channel is:

$$C = \max_{p(x)} \{I(X;Y)\} = \frac{1}{2} \log_2 \left(\frac{2E_s}{N_o} + 1\right) = \frac{1}{2} \log_2 \left(\frac{2rE_b}{N_o} + 1\right)$$

– where  $E_b$  is the energy per (information) bit.

- This capacity is achieved by a Gaussian input x.
  - This is not a practical modulation.

Capacity of AWGN with BPSK Constrained Input



#### Capacity of AWGN w/ 1-D Signaling



### Power Efficiency of Standard Binary Channel Codes



# M-ary modulation

- μ = log<sub>2</sub> M bits are mapped to the symbol x<sub>k</sub>, which is chosen from the set S = {x<sub>1</sub>, x<sub>2</sub>, ..., x<sub>M</sub>}
  - The symbol is multidimensional.
  - 2-D Examples: QPSK, M-PSK, QAM
  - M-D Example: FSK, block space-time codes (BSTC)
- The signal  $\mathbf{y} = \mathbf{x}_k + \mathbf{n}$  is received
  - More generally (BSTC),  $\mathbf{Y} = \mathbf{H}\mathbf{X} + \mathbf{N}$
- For each signal in S, the receiver computes  $p(\mathbf{y}|\mathbf{x}_k)$ 
  - This function depends on the modulation, channel, and receiver.

Monte Carlo Approach to Computing Modulation Constrained Capacity

- Suppose we want to compute capacity of M-ary modulation
  - In each case, we cannot control input distribution.
  - The capacity is merely the mutual information between channel input and output.
  - The mutual information can be measured as the following expectation:

$$C = I(X;Y) = \mu - E \left[ \log_2 \frac{\sum_{\mathbf{x} \in S} p(\mathbf{y} \mid \mathbf{x})}{p(\mathbf{y} \mid \mathbf{x}_k)} \right]$$

This expectation can be obtained through Monte Carlo simulation.

### Simulation Block Diagram



 $C = \mu - E[\Lambda]$ 

Benefits of Monte Carlo approach: -Allows high dimensional signals to be studied.

-Can determine performance in fading.

-Can study influence of receiver design.







# BICM

- Coded modulation (CM) is required to attain the aforementioned capacity.
  - Channel coding and modulation handled jointly.
  - e.g. trellis coded modulation (Ungerboeck); coset codes (Forney)
- Most off-the-shelf capacity approaching codes are binary.
- A pragmatic system would use a binary code followed by a bitwise interleaver and an M-ary modulator.
  - Bit Interleaved Coded Modulation (BICM); Caire 1998.



# **BICM** Receiver

- Like the CM receiver, the BICM receiver calculates p(y|x<sub>k</sub>) for each signal in S.
- Furthermore, the BICM receiver needs to calculate the log-likelihood ratio of each code bit:

$$\lambda_n = \log \frac{\sum_{\mathbf{x} \in S_n^{(1)}} p(\mathbf{y} | \mathbf{x})}{\sum_{\mathbf{x} \in S_n^{(0)}} p(\mathbf{y} | \mathbf{x})}$$

- where  $S_n^{(1)}$  represents the set of symbols whose n<sup>th</sup> bit is a 1.
- and  $s_n^{(0)}$  is the set of symbols whose n<sup>th</sup> bit is a 0.

# **BICM** Capacity

The BICM capacity is then [Caire 1998]:

$$C = I(X;Y) = \mu - E\left[\sum_{n=1}^{\mu} \log_2(1+\lambda_n)\right]$$

As with CM, this can be computed using a Monte Carlo integration.
For each bit, calculate:



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# **BICM-ID**

The conventional BICM receiver assumes that all bits in a symbol are equally likely:

$$\lambda_n = \log \frac{\sum_{\mathbf{x} \in S_n^{(1)}} p(\mathbf{y} | \mathbf{x})}{\sum_{\mathbf{x} \in S_n^{(0)}} p(\mathbf{y} | \mathbf{x})}$$

■ However, if the receiver has estimates of the bit probabilities, it can use this to weight the symbol likelihoods.  $\sum p(\mathbf{y}|\mathbf{x})p(\mathbf{x} | c_n = 1)$ 

$$\lambda_n = \log \frac{\mathbf{x} \in S_n^{(1)}}{\sum_{\mathbf{x} \in S_n^{(0)}} p(\mathbf{y} | \mathbf{x}) p(\mathbf{x} | c_n = 0)}$$

- This information is obtained from decoder feedback.
  - Bit Interleaved Coded Modulation with Iterative Demodulation
  - Li and Ritcey 1999.

## Mutual Information Transfer Chart

- Now consider a receiver that has a priori information about the code bits (from a soft output decoder).
- Assume the following:
  - The a priori information is in LLR form.
  - The a priori LLR's are Gaussian distributed.
  - The LLR's have mutual information  $I_v$
- Then the mutual information I<sub>z</sub> at the output of the receiver can be measured through Monte Carlo Integration.
  - $I_z$  vs.  $I_v$  is the *Mutual Information Transfer Characteristic.*
  - ten Brink 1999.

#### Generating Random a Priori Input



#### Mutual Information Characteristic



#### **EXIT** Chart



# EXIT Chart for Space Time Block Code



## Extensions to the MAC Layer

### Hybrid-ARQ Encode data into a low-rate R<sub>M</sub> code • Implemented using rate-compatible puncturing. Break the codeword into M distinct blocks • Each block has rate $R = R_M/M$ Source begins by sending the first block. If destination does not signal with an ACK, the next block is sent. After mth transmission, effective rate is R<sub>m</sub> = R/m This continues until either the destination decodes the message or all blocks have been transmitted.

# Info Theory of Hybrid-ARQ

- Throughput of hybrid-ARQ has been studied by Caire and Tuninetti (IT 2001).
  - Let  $\gamma_m$  denote the received SNR during the m<sup>th</sup> transmission
    - $\gamma_m$  is a random.
  - Let C( $\gamma_m$ ) be the capacity of the channel with SNR  $\gamma_m$ 
    - $C(\gamma_m)$  is also random.
  - The capacity after m blocks have been transmitted is:

$$C_m = \sum_m C(\gamma_m)$$

- This is because the capacity of parallel Gaussian channels adds.
- An outage occurs after the m<sup>th</sup> block if

 $C_m < R$ 

Throughput and delay depend on the average number of blocks required to get out of an outage.

# Extensions to the Network Layer



# Generalized Hybrid-ARQ Protocol

- Source broadcasts first packet, m=1.
- Relays that can decode are added to the decoding set D.
  - The source is also in D
- The next packet is sent by a node in D.
  - The choice of which node depends on the protocol.
  - Geographic-Relaying: Pick the node in D closest to destination.
- The process continues until the destination can decode.
- We term this protocol "HARBINGER"
  - Hybrid ARq-Based INtercluster GEographic Relaying.
- Energy-latency tradeoff can be analyzed by generalizing Caire and Tuninetti's analysis.

## HARBINGER: Initialization



# HARBINGER: First Hop



# HARBINGER: Selecting the Relay for the Second Hop



contention period

# HARBINGER: Second Hop



### HARBINGER: Third Hop



### HARBINGER: Fourth Hop



### HARBINGER: Results



#### **Topology:**

Relays on straight line S-D separated by 10 m

#### Coding parameters:

Per-block rate R=1 No limit on M Code Combining

#### **Channel parameters:**

n = 3 path loss exponent  $2.4 \, \mathrm{GHz}$  $d_0 = 1$  m reference dist

Unconstrained modulation

#### Monte Carlo Integration

B. Zhao and M. C. Valenti. "Practical relay networks: A generalization of hybrid-ARQ," IEEE JSAC, Jan. 2005.

# Discussion

- Advantages.
  - Better energy-latency tradeoff than multihop.
    - Nodes can transmit with significantly lower energy.
    - System exploits momentarily good links to reduce delay.
  - No need to maintain routing tables (reactive).
- Disadvantages.
  - More receivers must listen to each broadcast.
    - Reception consumes energy.
  - Nodes within a cluster must remain quiet.
  - Longer contention period in the MAC protocol.
  - Results are intractable, must resort to simulation.
  - Requires position estimates.
  - These tradeoffs can be balanced by properly selecting the number of relays in a cluster.

# Conclusions

- Capacity analysis is a quick way to assess the impact of the modulation choice and channel model.
  - The capacity of complicated systems can be found through Monte Carlo simulation.
- Once a modulation choice is selected for the channel of interest, any off-the-shelf capacity approaching binary code can be used.
  - The interface between demodulator and decoder can be characterized by its EXIT chart.
- Capacity analysis can also be used to characterize:
  - Delay and throughput of retransmission protocols.
  - Performance of multihop routing protocols.