#### EE360: Multiuser Wireless Systems and Networks

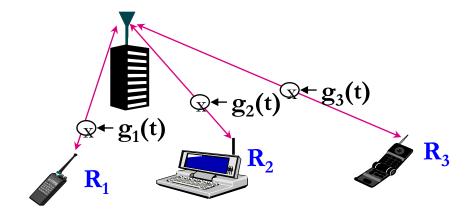
#### Lecture 3 Outline

#### Announcements

- Makeup lecture Feb 2, 5-6:15.
- Presentation schedule will be sent out later today, presentations will start 1/30.
- Next lecture: Random/Multiple Access, SS, MUD
- Capacity of Broadcast ISI Channels
- Capacity of MAC Channels
  - In AWGN
  - In Fading and ISI
- Duality between the MAC and the BC
- Capacity of MIMO Multiuser Channels

#### Review of Last Lecture

## Broadcast: One Transmitter to Many Receivers.

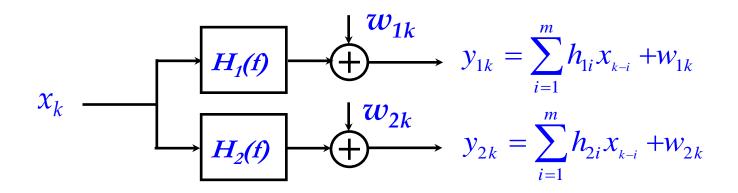


- Channel capacity region of broadcast channels
- Capacity in AWGN
  - Use superposition coding and optimal power allocation
- Capacity in fading
  - Ergodic capacity: optimally allocate resources over time
  - Outage capacity: maintain fixed rates in all states
  - Minimum rate capacity: fixed min. rate in all states, use excess rsources to optimize average rate above min.

#### **Broadcast Channels with ISI**

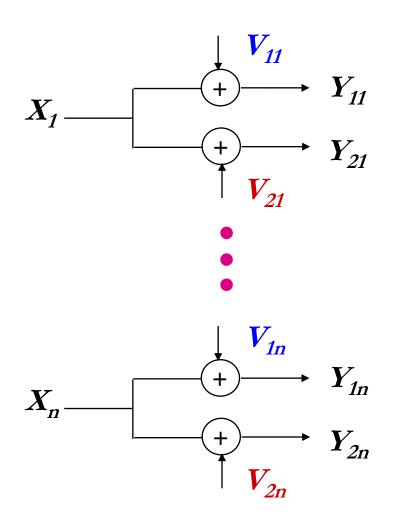
- ISI introduces memory into the channel
- The optimal coding strategy decomposes the channel into parallel broadcast channels
  - Superposition coding is applied to each subchannel.
- Power must be optimized across subchannels and between users in each subchannel.

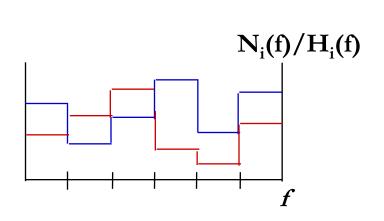
#### **Broadcast Channel Model**



- Both  $H_1$  and  $H_2$  are finite IR filters of length m.
- The  $w_{1k}$  and  $w_{2k}$  are correlated noise samples.
- For 1<k<n, we call this channel the n-block discrete Gaussian broadcast channel (n-DGBC).
- The channel capacity region is  $C=(R_1,R_2)$ .

## Equivalent Parallel Channel Model





#### Channel Decomposition

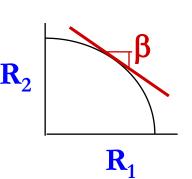
- Via a DFT, the BC with ISI approximately decomposes into *n* parallel AWGN degraded broadcast channels.
  - As n goes to infinity, this parallel model becomes exact
- The capacity region of parallel degraded broadcast channels was obtained by El-Gamal (1980)
  - Optimal power allocation obtained by Hughes-Hartogs ('75).
- The power constraint  $\sum_{i=0}^{n-1} E[x_i^2] \le nP$  on the original channel is converted by Parseval's theorem to  $\sum_{i=0}^{n-1} E[(X_i')^2] \le n^2P$  on the equivalent channel.

#### Capacity Region of Parallel Set

• Achievable Rates (no common information)

$$\begin{split} & \left\{ \boldsymbol{R}_{1} \leq .5 \sum_{j:\sigma_{1j} < \sigma_{2j}} \log \left( 1 + \frac{\alpha_{j} \boldsymbol{P}_{j}}{\sigma_{1j}} \right) + .5 \sum_{j:\sigma_{1j} \geq \sigma_{2j}} \log \left( 1 + \frac{\alpha_{j} \boldsymbol{P}_{j}}{(1 - \alpha_{j}) \boldsymbol{P}_{j} + \sigma_{1j}} \right), \\ & \boldsymbol{R}_{2} \leq .5 \sum_{j:\sigma_{1j} < \sigma_{2j}} \log \left( 1 + \frac{(1 - \alpha_{j}) \boldsymbol{P}_{j}}{\alpha_{j} \boldsymbol{P}_{j} + \sigma_{2j}} \right) + .5 \sum_{j:\sigma_{1j} \geq \sigma_{2j}} \log \left( 1 + \frac{(1 - \alpha_{j}) \boldsymbol{P}_{j}}{\sigma_{2j}} \right), \\ & 0 \leq \alpha_{j} \leq 1, \sum \boldsymbol{P}_{j} \leq \boldsymbol{n}^{2} \boldsymbol{P} \right\} \end{split}$$

- Capacity Region
  - For  $0 < \beta \le \infty$  find  $\{\alpha_i\}$ ,  $\{P_i\}$  to maximize  $R_1 + \beta R_2 + \lambda \sum P_i$
  - Let  $(R_1^*, R_2^*)_{n,\beta}$  denote the corresponding rate pair.
  - $C_n = \{(R_1^*, R_2^*)_{n,\beta} : 0 < \beta \le \infty \}, C = \liminf_{n \to \infty} \frac{1}{n} C_n$ .

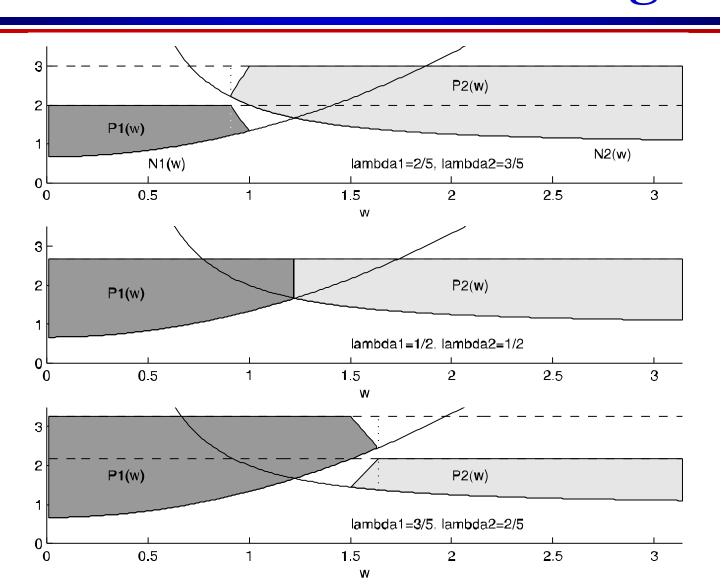


## Limiting Capacity Region

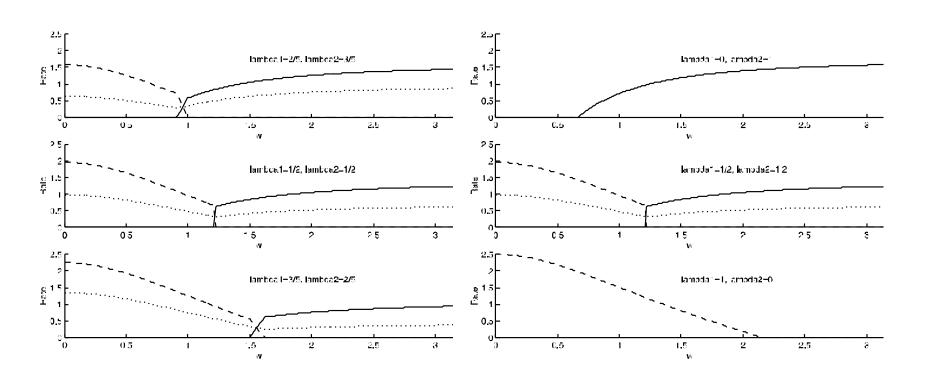
$$\begin{aligned}
& \left\{ \mathbf{R}_{1} \leq .5 \int_{f:H_{1}(f)>H_{2}(f)} \log \left( 1 + \frac{\alpha(f)\mathbf{P}(f) |\mathbf{H}_{1}(f)|^{2}}{.5N_{0}} \right) + .5 \int_{f:H_{1}(f)\leq H_{2}(f)} \log \left( 1 + \frac{\alpha_{j}\mathbf{P}_{j}}{(1-\alpha_{j})\mathbf{P}_{j} + \sigma_{1j}} \right), \\
& \mathbf{R}_{2} \leq .5 \int_{f:H_{1}(f)>H_{2}(f)} \log \left( 1 + \frac{(1-\alpha(f))\mathbf{P}(f)}{\alpha(f)\mathbf{P}(f) + .5N_{0} / |\mathbf{H}_{2}(f)|^{2}} \right) + .5 \int_{f:H_{1}(f)\leq H_{2}(f)} \log \left( 1 + \frac{(1-\alpha(f))\mathbf{P}(f) |\mathbf{H}_{2}(f)|^{2}}{.5N_{0}} \right), \end{aligned}$$

$$0 \le \alpha(f) \le 1, \qquad \int P(f) df \le P$$

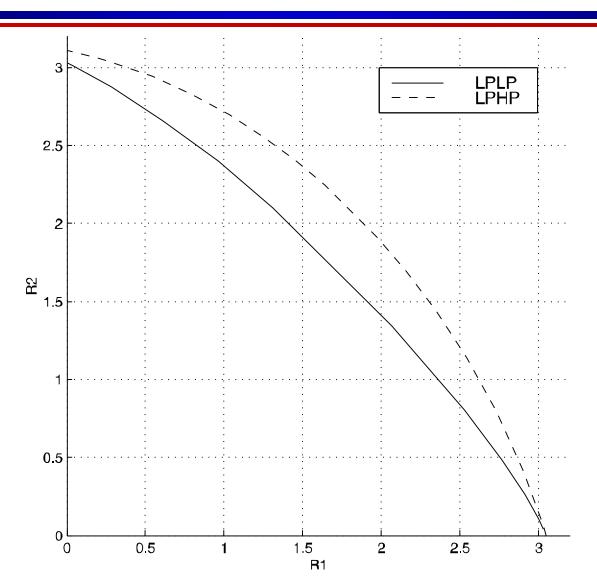
## Optimal Power Allocation: Two Level Water Filling



## Capacity vs. Frequency



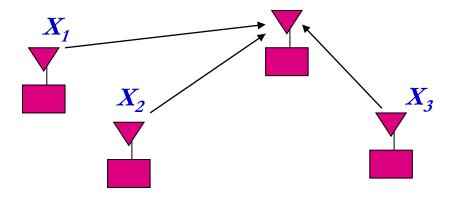
## Capacity Region



#### Multiple Access Channel

- Multiple transmitters
  - Transmitter i sends signal  $X_i$  with power  $P_i$
- Common receiver with AWGN of power  $N_0B$
- Received signal:

$$Y = \sum_{i=1}^{M} X_i + N$$



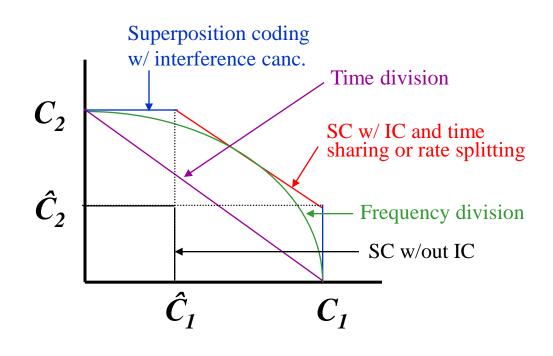
## **MAC** Capacity Region

• Closed convex hull of all  $(R_p,...,R_M)$  s.t.

$$\sum_{i \in S} R_i \le B \log \left[ 1 + \sum_{i \in S} P_i / N_0 B \right], \quad \forall S \subseteq \{1, ..., M\}$$

- For all subsets of users, rate sum equals that of 1 superuser with sum of powers from all users
- Power Allocation and Decoding Order
  - Each user has its own power (no power alloc.)
  - Decoding order depends on desired rate point

## Two-User Region



$$C_i = B \log \left[ 1 + \frac{P_i}{N_0 B} \right], i = 1, 2$$

$$\hat{C}_1 = B \log \left[ 1 + \frac{P_1}{N_0 B + P_2} \right], \qquad \hat{C}_2 = B \log \left[ 1 + \frac{P_2}{N_0 B + P_1} \right],$$

$$\hat{C}_2 = B \log \left[ 1 + \frac{P_2}{N_0 B + P_1} \right],$$

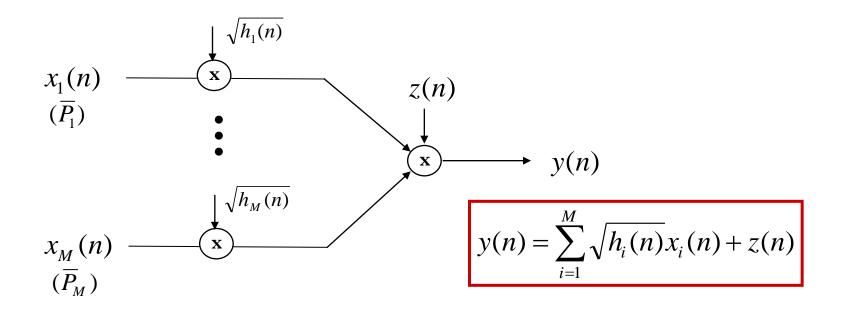
## Fading and ISI

- MAC capacity under fading and ISI determined using similar techniques as for the BC
- In fading, can define ergodic, outage, and minimum rate capacity similar as in BC case
  - Ergodic capacity obtained based on AWGN MAC given fixed fading, averaged over fading statistics
  - Outage can be declared as common, or per user
- MAC capacity with ISI obtained by converting to equivalent parallel MAC channels over frequency

#### Characteristics

- Corner points achieved by 1 user operating at his maximum rate
  - Other users operate at rate which can be decoded perfectly and subtracted out (IC)
- Time sharing connects corner points
  - Can also achieve this line via rate splitting, where one user "splits" into virtual users
- FD has rate  $R_i \leq B_i \log[1 + P_i/(N_0 B)]$
- TD is straight line connecting end points
  - With variable power, it is the same as FD
- CD without IC is box

## Fading MAC Channels



- Noise is AWGN with variance  $\sigma^2$ .
- Joint fading state (known at TX and RX):

$$h = (h_1(n), ..., h_M(n))$$

## Capacity Region\*

- Rate allocation  $R(h) \in \mathbb{R}^{M}$
- Power allocation  $P(h) \in \mathbb{R}^{M}$ 
  - Subject to power constraints:  $E_h[P(h)] \le P$
- Boundary points: R\*
  - $\exists \lambda, \mu \in \mathbb{R}^{M}$  s.t. [R(h),P(h)] solves

$$\max \mu \mathbf{R} - \lambda \mathbf{P} \quad s.t. \quad \sum_{i \in S} R_i \le .5 \log \left[ 1 + \frac{\sum_{i \in S} h_i P_i}{\sigma^2} \right], \forall S \subseteq \{1, ..., M\}$$

with 
$$E_h[R_i(h)]=R_i^*$$

## Unique Decoding Order\*

- For every boundary point R\*:
  - There is a unique decoding order that is the same for every fading state
  - Decoding order is reverse order of the priorities

$$\mu_1 \ge ... \ge \mu_M \Rightarrow Decoding order: M, M-1,...1$$

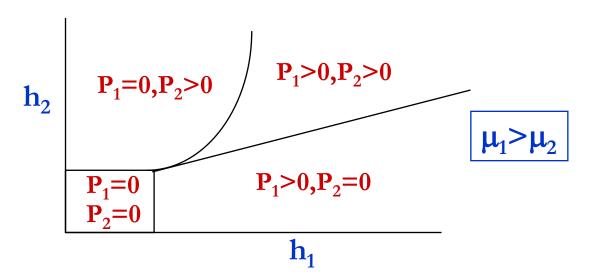
- Implications:
  - Given decoding order, only need to optimally allocate power across fading states
  - Without unique decoding order, utility functions used to get optimal rate and power allocation

# Characteristics of Optimum Power Allocation

- A user's power in a given state depends only on:
  - His channel (h<sub>ik</sub>)
  - Channels of users decoded just before (h<sub>ik-1</sub>) and just after (h<sub>ik+1</sub>)
  - Power increases with h<sub>ik</sub> and decreases with h<sub>ik-1</sub> and h<sub>ik+1</sub>
  - Power allocation is a modified waterfilling, modified to interference from active users just before and just after
- User decoded first waterfills to SIR for all active users

## Transmission Regions

- The region where no users transmit is a hypercube
  - Each user has a unique cutoff below which he does not transmit
- For highest priority user, always transmits above some h<sub>1</sub>\*
- The lowest priority user, even with a great channel, doesn't transmit if some other user has a relatively good channel



#### Two User Example

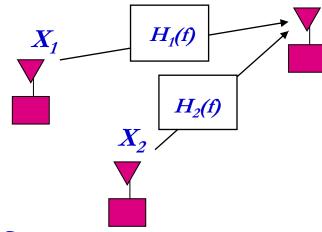
• Power allocation for  $\mu_1 > \mu_2$ 

$$P_{2}(\mathbf{h}) = \begin{cases} 0 & \frac{\mathbf{h}_{2}}{1 + \mathbf{h}_{1} \mathbf{P}_{1}(\mathbf{h})} > \frac{\lambda_{2}}{\mu_{2}} \\ \frac{\mu_{2}}{\lambda_{2}} - \frac{1 + \mathbf{h}_{1} \mathbf{P}_{1}(\mathbf{h})}{\mathbf{h}_{2}} & \frac{\mathbf{h}_{2}}{1 + \mathbf{h}_{1} \mathbf{P}_{1}(\mathbf{h})} < \frac{\lambda_{2}}{\mu_{2}} \end{cases}$$

## **Ergodic Capacity Summary**

- Rate region boundary achieved via optimal allocation of power and decoding order
- For any boundary point, decoding order is the same for all states
  - Only depends on user priorities
- Optimal power allocation obtained via Lagrangian optimization
  - Only depends on users decoded just before and after
  - Power allocation is a modified waterfilling
  - Transmission regions have cutoff and critical values

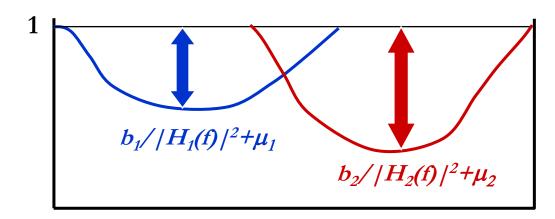
#### MAC Channel with ISI\*



- Use DFT Decomposition
- Obtain parallel MAC channels
- Must determine each user's power allocation across subchannels and decoding order
- Capacity region no longer a pentagon

#### **Optimal Power Allocation**

- Capacity region boundary: maximize  $\mu_1 R_1 + \mu_2 R_2$
- Decoding order based on priorities and channels
- Power allocation is a two-level water filling
  - Total power of both users is scaled water level
  - In non-overlapping region, best user gets all power (FD)
  - With overlap, power allocation and decoding order based on  $\lambda$ s and user channels.



## Comparison of MAC and BC

#### • Differences:

- Shared vs. individual power constraints
- Near-far effect in MAC

# $\mathbf{P}_{1}$

#### Similarities:

- Optimal BC "superposition" coding is also optimal for MAC (sum of Gaussian codewords)
- Both decoders exploit successive decoding and interference cancellation

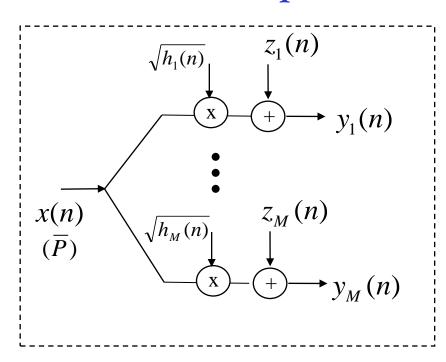
## MAC-BC Capacity Regions

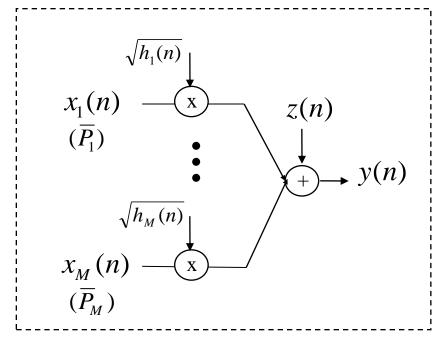
- MAC capacity region known for many cases
  - Convex optimization problem
- BC capacity region typically only known for (parallel) degraded channels
  - Formulas often not convex
- Can we find a connection between the BC and MAC capacity regions?



#### Dual Broadcast and MAC Channels

Gaussian BC and MAC with same channel gains and same noise power at each receiver



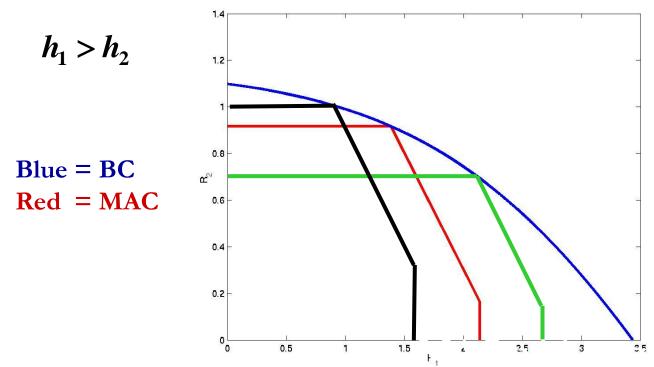


**Broadcast Channel (BC)** 

Multiple-Access Channel (MAC)

#### The BC from the MAC





$$P_1 = 0.5, P_2 = 1.5$$

$$P_1 = 1, P_2 = 1$$

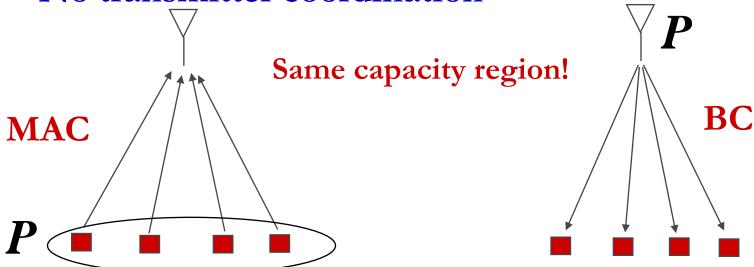
$$P_1=1.5, P_2=0.5$$

$$C_{BC}(P;h_1,h_2) = \bigcup_{0 \le P_1 \le P} C_{MAC}(P_1,P-P_1;h_1,h_2)$$

#### **Sum-Power MAC**

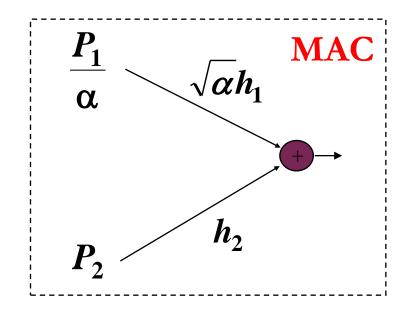
$$C_{BC}(P; h_1, h_2) = \bigcup_{0 \le P_1 \le P} C_{MAC}(P_1, P - P_1; h_1, h_2) \equiv C_{MAC}^{Sum}(P; h_1, h_2)$$

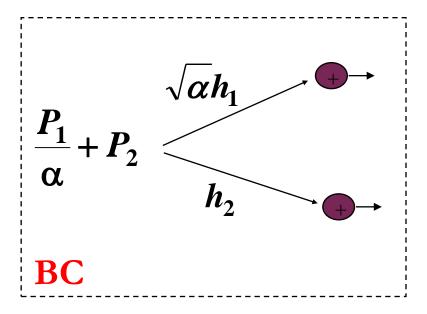
- MAC with <u>sum</u> power constraint
  - Power pooled between MAC transmitters
  - No transmitter coordination



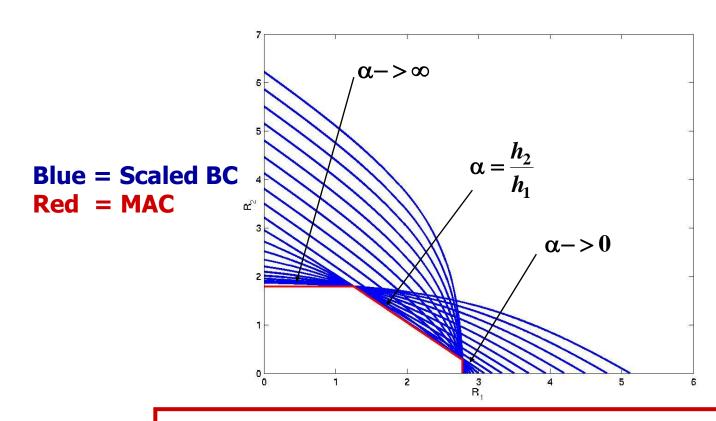
## BC to MAC: Channel Scaling

- Scale channel gain by  $\sqrt{\alpha}$ , power by  $1/\alpha$
- MAC capacity region unaffected by scaling
- Scaled MAC capacity region is a subset of the scaled BC capacity region for any α
- MAC region inside scaled BC region for any scaling





#### The BC from the MAC

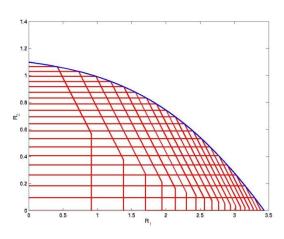


$$C_{MAC}(P_1, P_2; h_1, h_2) = \bigcap_{\alpha > 0} C_{BC}(\frac{P_1}{\alpha} + P_2; \sqrt{\alpha}h_1, h_2)$$

#### Duality: Constant AWGN Channels

BC in terms of MAC

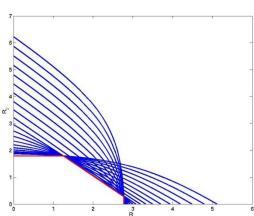
$$C_{BC}(P;h_1,h_2) = \bigcup_{0 \le P_1 \le P} C_{MAC}(P_1,P-P_1;h_1,h_2)$$



MAC in terms of BC

$$C_{MAC}(P_1, P_2; h_1, h_2) = \bigcap_{\alpha > 0} C_{BC}(\frac{P_1}{\alpha} + P_2; \alpha h_1, h_2)$$

What is the relationship between the optimal transmission strategies?



#### Transmission Strategy Transformations

Equate rates, solve for powers

$$R_1^M = \log(1 + \frac{h_1^2 P_1^M}{h_2 P_2^M + \sigma^2}) = \log(1 + \frac{h_1^2 P_1^B}{\sigma^2}) = R_1^B$$

$$R_2^M = \log(1 + \frac{h_2^2 P_2^M}{\sigma^2}) = \log(1 + \frac{h_2^2 P_2^B}{h_2^2 P_1^B + \sigma^2}) = R_2^B$$

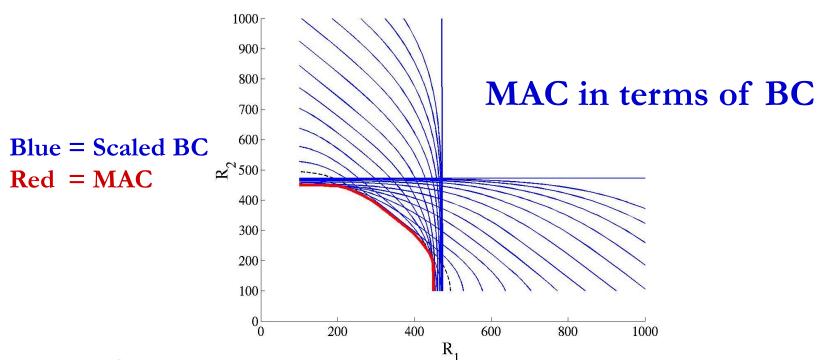
- Opposite decoding order
  - Stronger user (User 1) decoded last in BC
  - Weaker user (User 2) decoded last in MAC

#### Duality Applies to Different Fading Channel Capacities

- Ergodic (Shannon) capacity: maximum rate averaged over all fading states.
- Zero-outage capacity: maximum rate that can be maintained in all fading states.
- Outage capacity: maximum rate that can be maintained in all nonoutage fading states.
- Minimum rate capacity: Minimum rate maintained in all states, maximize average rate in excess of minimum

Explicit transformations between transmission strategies

#### Duality: Minimum Rate Capacity

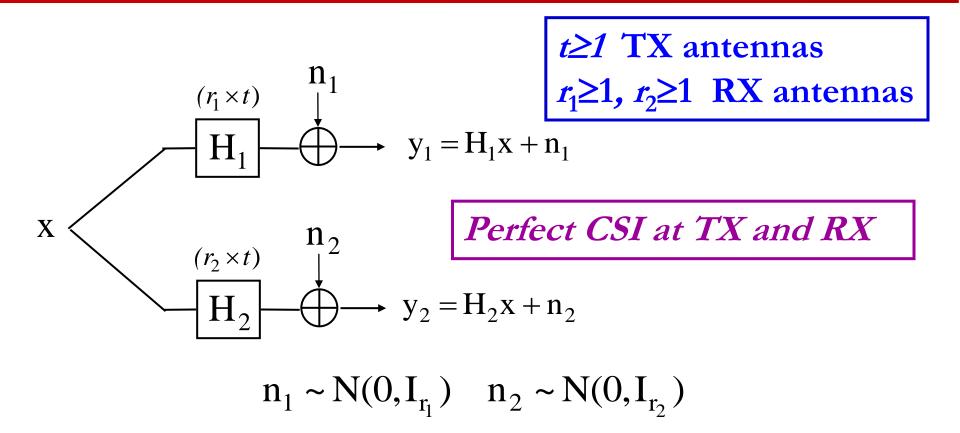


- BC region known
- MAC region can only be obtained by duality

What other capacity regions can be obtained by duality?

**Broadcast MIMO Channels** 

#### **Broadcast MIMO Channel**

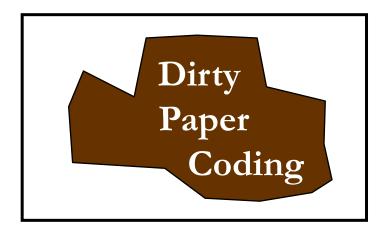


Non-degraded broadcast channel

## Dirty Paper Coding (Costa'83)

- Basic premise
  - If the interference is known, channel capacity same as if there is no interference
  - Accomplished by cleverly distributing the writing (codewords) and coloring their ink
  - Decoder must know how to read these codewords

Dirty
Paper
Coding

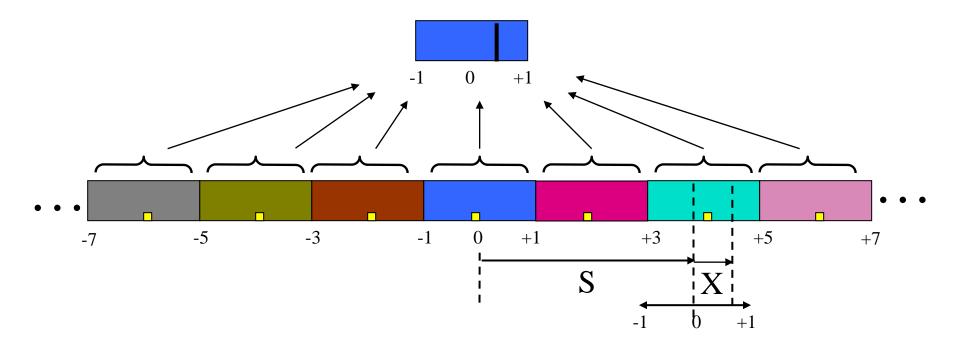


Clean Channel

**Dirty Channel** 

## Modulo Encoding/Decoding

- Received signal Y=X+S, -1≤X≤1
  - S known to transmitter, not receiver
- Modulo operation removes the interference effects
  - Set X so that  $[Y]_{[-1,1]}$  = desired message (e.g. 0.5)
  - Receiver demodulates modulo [-1,1]



## Capacity Results

- Non-degraded broadcast channel
  - Receivers not necessarily "better" or "worse" due to multiple transmit/receive antennas
  - Capacity region for general case unknown
- Pioneering work by Caire/Shamai (Allerton'00):
  - Two TX antennas/two RXs (1 antenna each)
  - Dirty paper coding/lattice precoding (achievable rate)
    - Computationally very complex
  - MIMO version of the Sato upper bound
  - Upper bound is achievable: capacity known!

# Dirty-Paper Coding (DPC) for MIMO BC

- Coding scheme:
  - Choose a codeword for user 1
  - Treat this codeword as interference to user 2
  - Pick signal for User 2 using "pre-coding"
- Receiver 2 experiences no interference:

$$\mathbf{R}_2 = \log(\det(\mathbf{I} + H_2 \Sigma_2 H_2^T))$$

• Signal for Receiver 2 interferes with Receiver 1:

$$R_1 = \log \left( \frac{\det(I + H_1(\Sigma_1 + \Sigma_2)H_1^T)}{\det(I + H_1\Sigma_2H_1^T)} \right)$$

- Encoding order can be switched
- DPC optimization highly complex

## Does DPC achieve capacity?

- DPC yields MIMO BC achievable region.
  - We call this the dirty-paper region
- Is this region the capacity region?
- We use duality, dirty paper coding, and Sato's upper bound to address this question
- First we need MIMO MAC Capacity

## MIMO MAC Capacity

MIMO MAC follows from MAC capacity formula

$$C_{MAC}(P_1,...,P_k) = \bigcup \left\{ (R_1,...,R_k) : \sum_{k \in S} R_k \le \log_2 \det \left[ I + \sum_{k \in S} H_k Q_k H_k^H \right], \right.$$

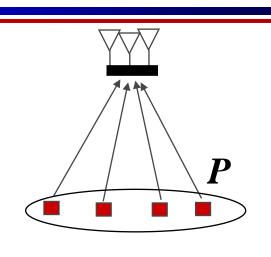
$$\forall S \subseteq \{1,...,K\} \right\}$$

- Basic idea same as single user case
  - Pick some subset of users
  - The sum of those user rates equals the capacity as if the users pooled their power
- Power Allocation and Decoding Order
  - Each user has its own power (no power alloc.)
  - Decoding order depends on desired rate point

### MIMO MAC with sum power

- MAC with sum power:
  - Transmitters code independently
  - Share power

$$C_{MAC}^{Sum}(P) = \bigcup_{0 \le P_1 \le P} C_{MAC}(P_1, P - P_1)$$



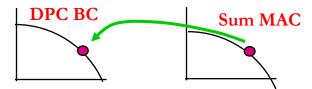
• Theorem: Dirty-paper BC region equals the dual sum-power MAC region

$$C_{BC}^{DPC}(P) = C_{MAC}^{Sum}(P)$$

#### Transformations: MAC to BC

• Show any rate achievable in sum-power MAC also achievable with DPC for BC:

$$C_{BC}^{DPC}(P) \supseteq C_{MAC}^{Sum}(P)$$

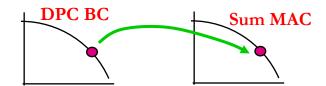


- A sum-power MAC strategy for point  $(R_1,...R_N)$  has a given input covariance matrix and encoding order
- We find the corresponding PSD covariance matrix and encoding order to achieve  $(R_1,...,R_N)$  with DPC on BC
  - The rank-preserving transform "flips the effective channel" and reverses the order
  - Side result: beamforming is optimal for BC with 1 Rx antenna at each mobile

#### Transformations: BC to MAC

• Show any rate achievable with DPC in BC also achievable in sum-power MAC:

$$C_{BC}^{DPC}(P) \subseteq C_{MAC}^{Sum}(P)$$



- We find transformation between optimal DPC strategy and optimal sum-power MAC strategy
  - "Flip the effective channel" and reverse order

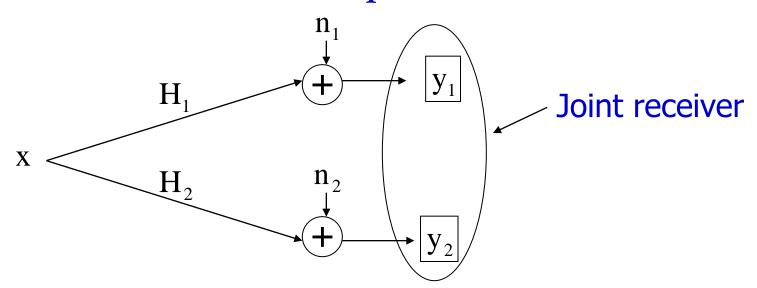
### Computing the Capacity Region

$$C_{BC}^{DPC}(P) = C_{MAC}^{Sum}(P)$$

- Hard to compute DPC region (Caire/Shamai'00)
- "Easy" to compute the MIMO MAC capacity region
  - Obtain DPC region by solving for sum-power MAC and applying the theorem
  - Fast iterative algorithms have been developed
  - Greatly simplifies calculation of the DPC region and the associated transmit strategy

# Sato Upper Bound on the BC Capacity Region

• Based on receiver cooperation



• BC sum rate capacity ≤ Cooperative capacity

$$C_{\text{BC}}^{\text{sumrate}}(P, H) \le \frac{\max}{\Sigma_x} \frac{1}{2} \log |I + H\Sigma_x H^T|$$

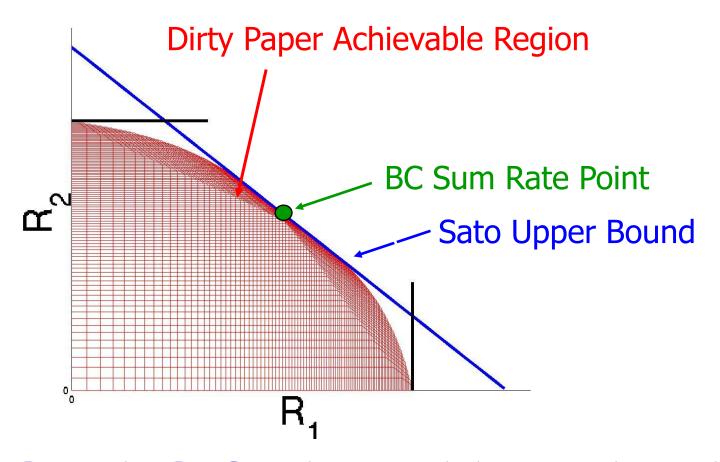
#### The Sato Bound for MIMO BC

- Introduce noise correlation between receivers
- BC capacity region unaffected
  - Only depends on noise marginals
- Tight Bound (Caire/Shamai'00)
  - Cooperative capacity with worst-case noise correlation

$$C_{\text{BC}}^{\text{sumrate}}(P, H) \le \inf_{\Sigma_{z}} \max_{\Sigma_{x}} \frac{1}{2} \log |I + \Sigma_{z}^{-1/2} H \Sigma_{x} H^{T} \Sigma_{z}^{-1/2}|$$

- Explicit formula for worst-case noise covariance
- By Lagrangian duality, cooperative BC region equals the sum-rate capacity region of MIMO MAC

## MIMO BC Capacity Bounds



Does the DPC region equal the capacity region?

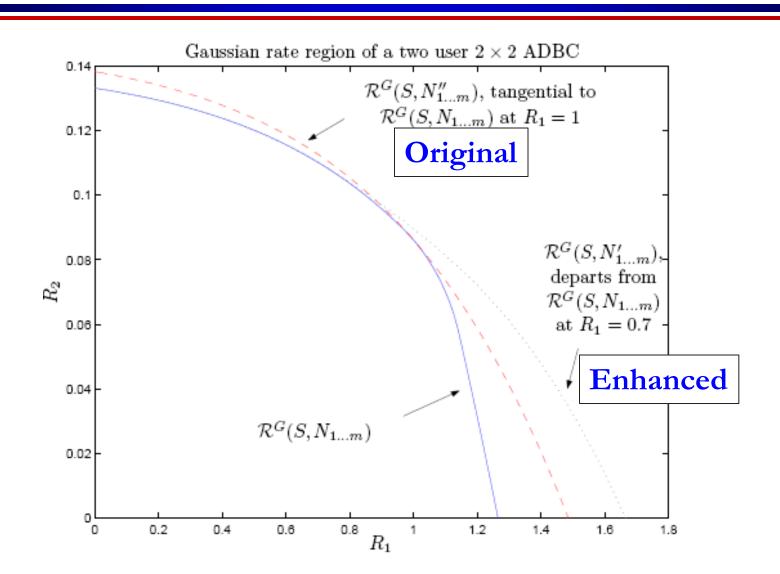
## Full Capacity Region

- DPC gives us an achievable region
- Sato bound only touches at sum-rate point
- Bergman's entropy power inequality is not a tight upper bound for nondegraded broadcast channel
- A tighter bound was needed to prove DPC optimal
  - It had been shown that if Gaussian codes optimal, DPC was optimal, but proving Gaussian optimality was open.
- Breakthrough by Weingarten, Steinberg and Shamai
  - Introduce notion of <u>enhanced channel</u>, applied Bergman's converse to it to prove DPC optimal for MIMO BC.

#### **Enhanced Channel Idea**

- The aligned and degraded BC (AMBC)
  - Unity matrix channel, noise innovations process
  - Limit of AMBC capacity equals that of MIMO BC
  - Eigenvalues of some noise covariances go to infinity
  - Total power mapped to covariance matrix constraint
- Capacity region of AMBC achieved by Gaussian superposition coding and successive decoding
  - Uses entropy power inequality on enhanced channel
  - Enhanced channel has less noise variance than original
  - Can show that a power allocation exists whereby the enhanced channel rate is inside original capacity region
- By appropriate power alignment, capacities equal

#### Illustration



#### **Main Points**

- Shannon capacity gives fundamental data rate limits for multiuser wireless channels
- Fading multiuser channels optimize at each channel instance for maximum average rate
- Outage capacity has higher (fixed) rates than with no outage.
- OFDM is near optimal for broadcast channels with ISI
- Duality connects BC and MAC channels
  - Used to obtain capacity of one from the other
- Capacity of broadcast MIMO channel obtained using duality and the notion of an enhanced channel