

# **Power Control for User Cooperation**

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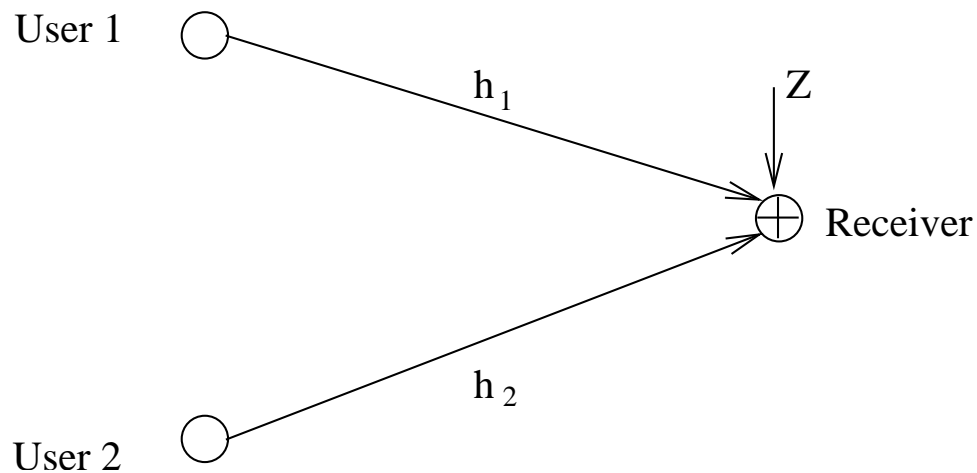
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Joint work with Onur Kaya.

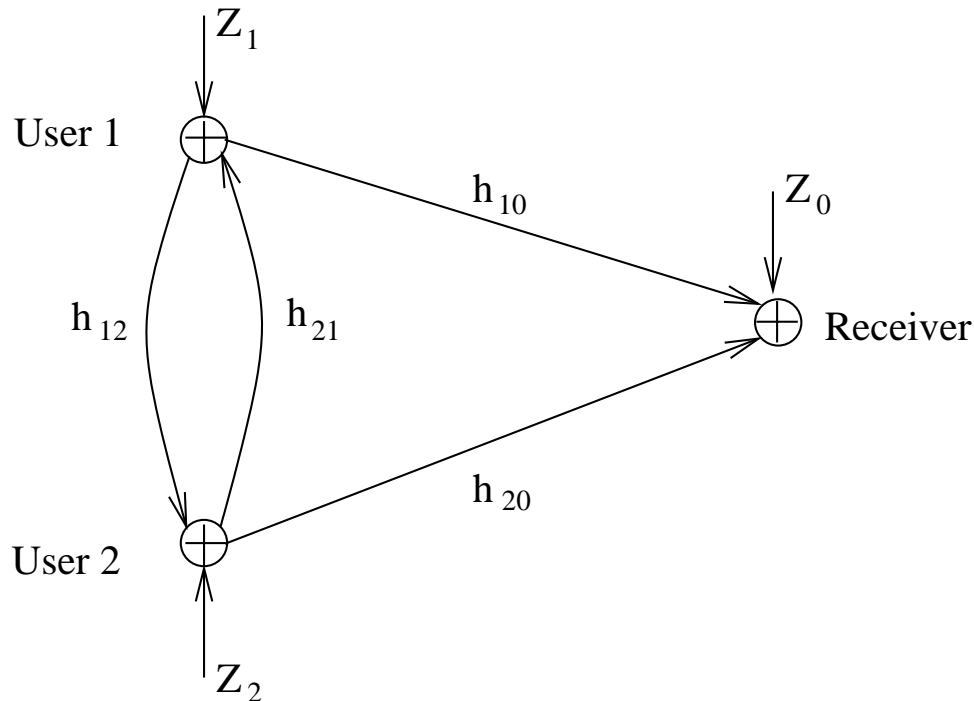
## Traditional Multiple Access



$$Y = \sqrt{h_1}X_1 + \sqrt{h_2}X_2 + Z$$

- Users talk only to the receiver, do not listen or talk to each other.
- This models the single cell of a cellular system:
  - Fixed roles for nodes (base station or mobile)
  - Strict and inflexible transmission scheme (uplink and downlink)
  - Single hop of wireless communication
- Information theoretically: multiple access channel; the capacity region is known.

## User Cooperation



$$Y_0 = \sqrt{h_{10}}X_1 + \sqrt{h_{20}}X_2 + Z_0$$

$$Y_1 = \sqrt{h_{21}}X_2 + Z_1$$

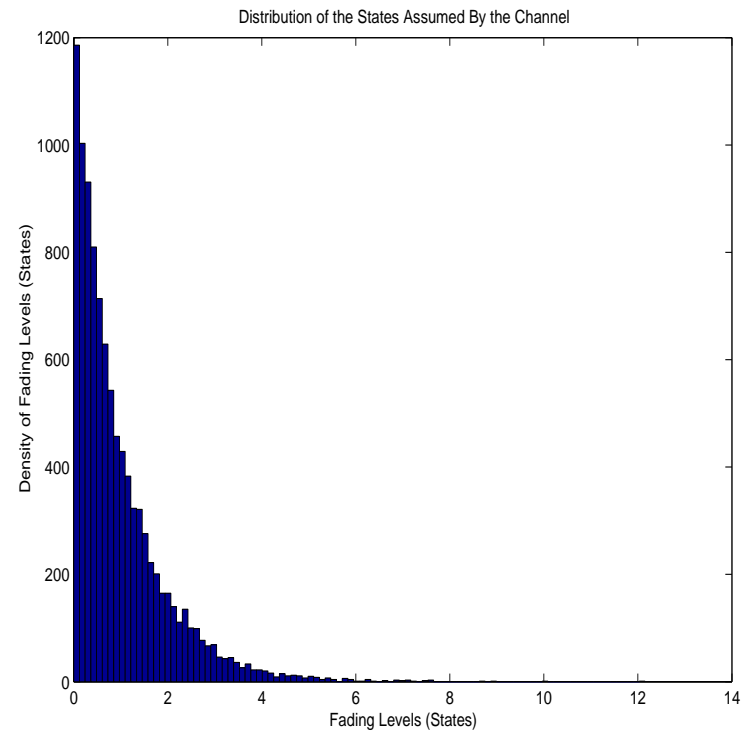
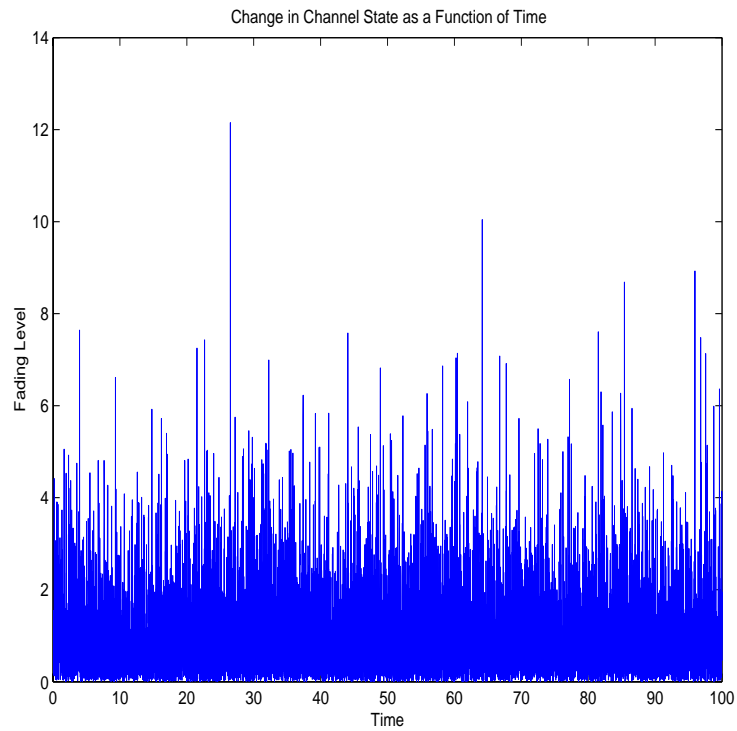
$$Y_2 = \sqrt{h_{12}}X_1 + Z_2$$

- **Interference** is **information**.
- Some versions of all transmitted signals are received by all nodes.
- Exploit overheard information to jointly design encoding, transmission, routing policies.
- Building block towards the analysis of ad-hoc and sensor networks.
- Information theoretically: multiple access channel with generalized feedback (capacity open).

# Fading

- **Fading:** random fluctuations in channel gains.

$$Y = \sqrt{h}X + Z$$

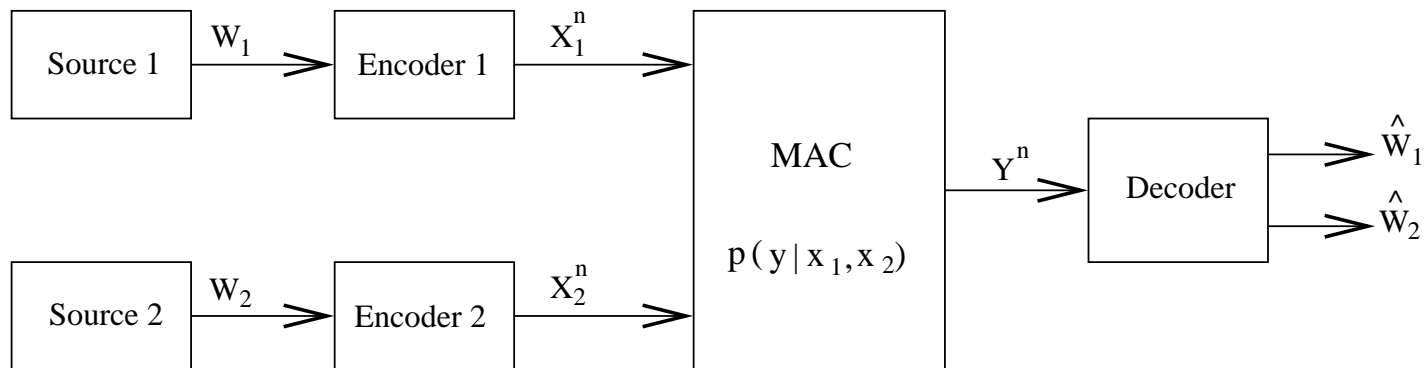


## Power Control

- **Fading:** random fluctuations in channel gains.
- If perfect channel state information (CSI) is available at transmitters
  - Dynamic resource allocation to improve quality-of-service or capacity
- Quality-of-service based
  - Provide all users with desired SIR levels
  - **Compensate** for channel fading; more power if bad channel, less if good channel
- Capacity based
  - Maximize information theoretic ergodic capacity subject to average power constraints
  - **Exploit** variations; more power if good channel, less if bad, no power if very bad
- Underlying motivation/justification:
  - Voice communication: delay intolerant; guarantee quality-of-service at all times.
  - **Data communication: delay tolerant; we may wait for the channel to be good.**

## Multiple Access Channel

- Two user multiple-access channel:



- Capacity region: closure of convex hull of all  $(R_1, R_2)$  satisfying

$$R_1 < I(X_1; Y | X_2)$$

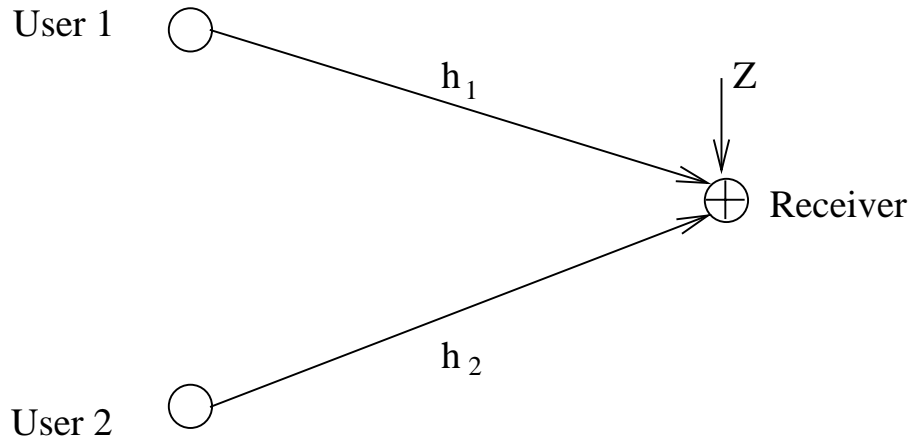
$$R_2 < I(X_2; Y | X_1)$$

$$R_1 + R_2 < I(X_1, X_2; Y)$$

for some product distribution  $p(x_1, x_2) = p(x_1)p(x_2)$ .

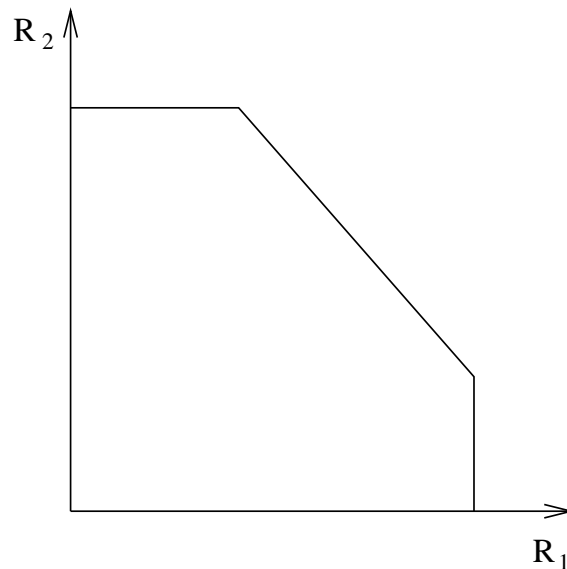
- Closure of convex hull (union) of many pentagons.

## Gaussian Multiple Access Channel



$$Y = \sqrt{h_1}X_1 + \sqrt{h_2}X_2 + Z$$

- There is one outer-most pentagon that includes all others;  $X_1 \sim N(0, P_1)$  and  $X_2 \sim N(0, P_2)$ .



$$R_1 < \frac{1}{2} \log \left( 1 + \frac{P_1 h_1}{\sigma^2} \right)$$
$$R_2 < \frac{1}{2} \log \left( 1 + \frac{P_2 h_2}{\sigma^2} \right)$$
$$R_1 + R_2 < \frac{1}{2} \log \left( 1 + \frac{P_1 h_1 + P_2 h_2}{\sigma^2} \right)$$

## Fading Single User Channel (Goldsmith-Varaiya 1994)

- Channel capacity for single user

$$C = \frac{1}{2} \log \left( 1 + \frac{p}{\sigma^2} \right)$$

- In the presence of fading, for a fixed channel state  $h$

$$C(h) = \frac{1}{2} \log \left( 1 + \frac{p(h)h}{\sigma^2} \right)$$

- Maximize the ergodic capacity, given an average power constraint

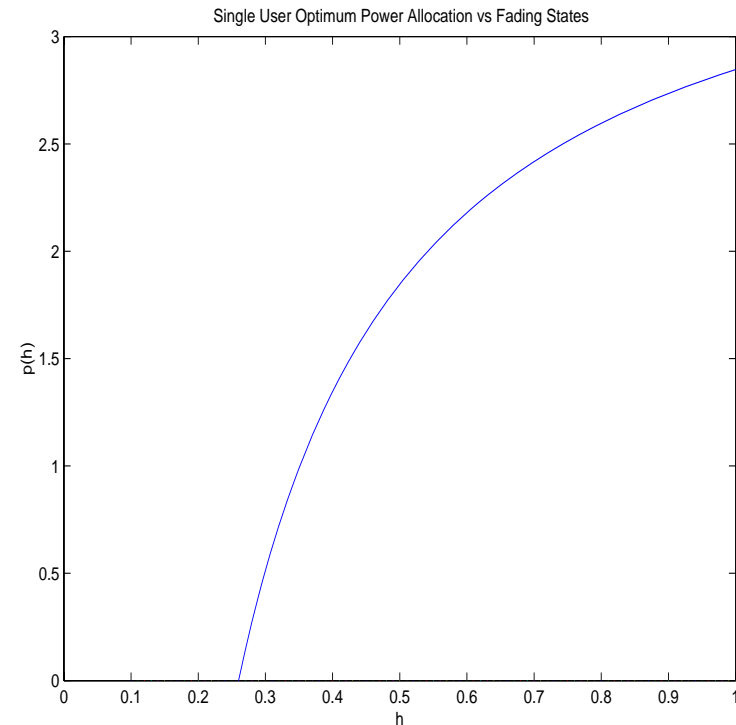
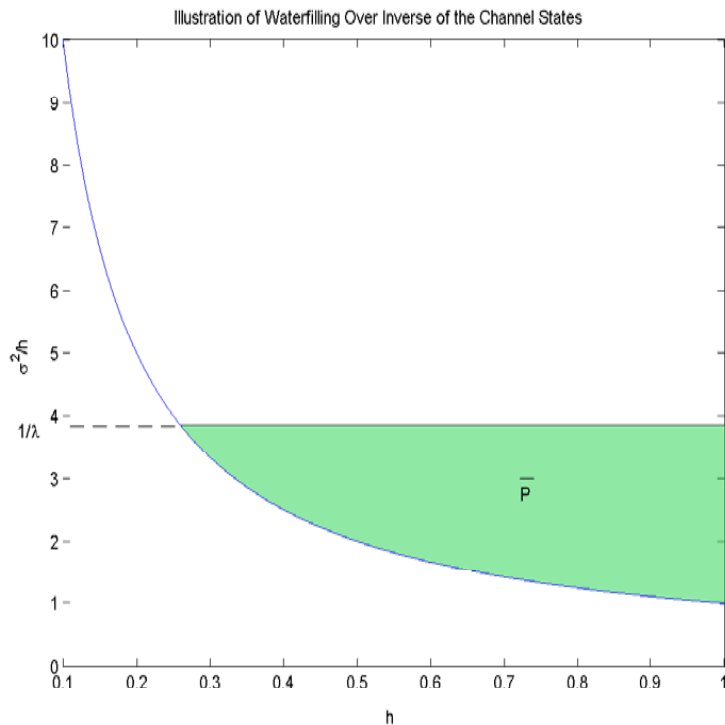
$$\begin{aligned} \max_{p(h)} \quad & E_h \left[ \log \left( 1 + \frac{p(h)h}{\sigma^2} \right) \right] \\ \text{s.t.} \quad & E_h [p(h)] \leq P \\ & p(h) \geq 0, \quad \forall h \end{aligned}$$



# Fading Single User Channel Solution: Waterfilling

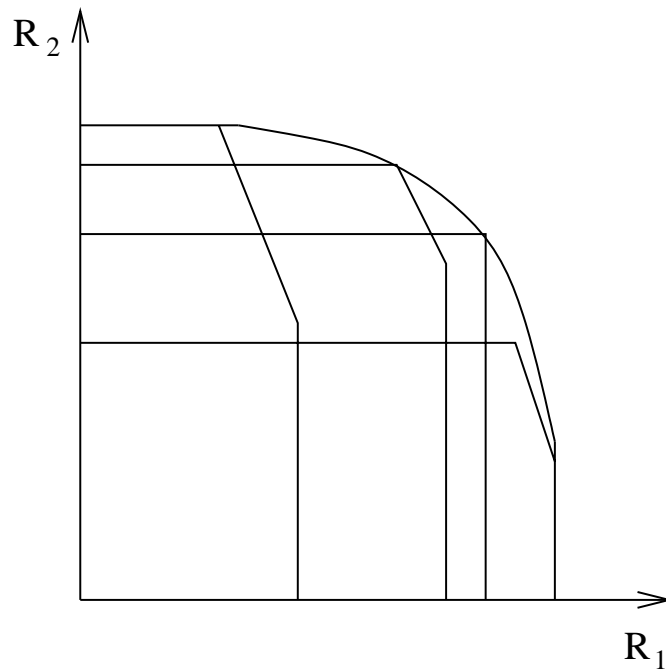
- Optimal power allocation: **waterfilling** of power over time

$$p(h) = \left( \frac{1}{\lambda} - \frac{\sigma^2}{h} \right)^+$$



- More power to better channel states; no power to very poor channel states.

## Fading Gaussian Multiple Access Channel (Tse-Hanly 1998)



Union of pentagons

$$R_1 < E \left[ \frac{1}{2} \log \left( 1 + \frac{p_1(\mathbf{h})h_1}{\sigma^2} \right) \right] \quad (\triangleq C_1)$$

$$R_2 < E \left[ \frac{1}{2} \log \left( 1 + \frac{p_2(\mathbf{h})h_2}{\sigma^2} \right) \right] \quad (\triangleq C_2)$$

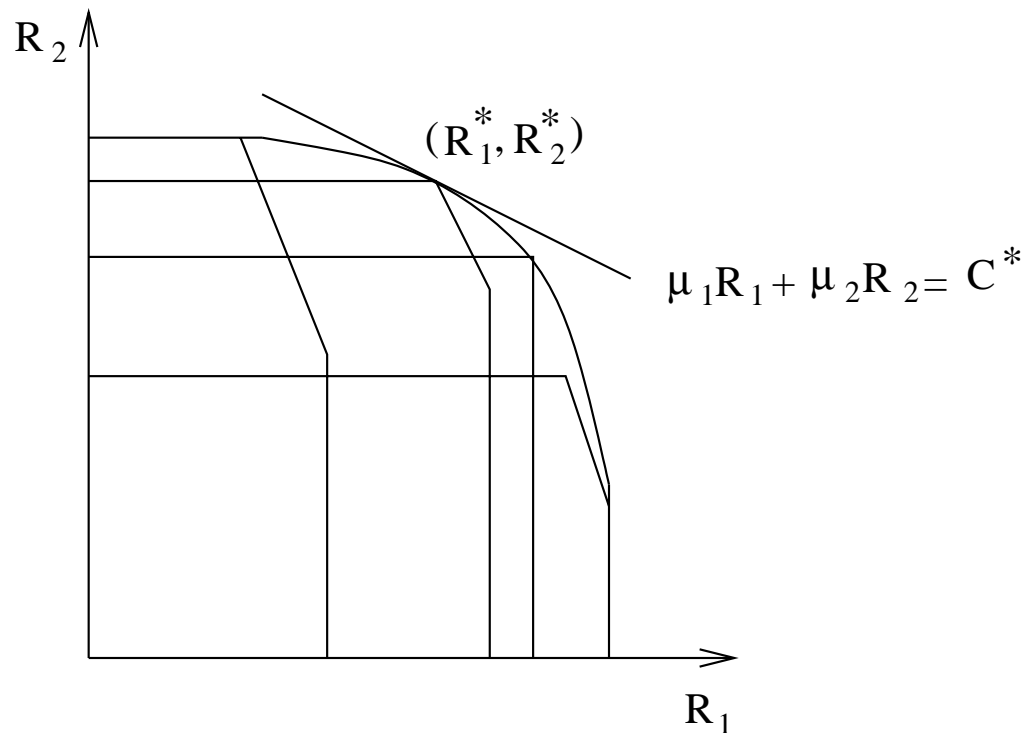
$$R_1 + R_2 < E \left[ \frac{1}{2} \log \left( 1 + \frac{p_1(\mathbf{h})h_1 + p_2(\mathbf{h})h_2}{\sigma^2} \right) \right] \quad (\triangleq C_s)$$

over all feasible power distributions

$$E [p_1(\mathbf{h})] \leq P_1$$

$$E [p_2(\mathbf{h})] \leq P_2$$

## Determining the Boundary of the Capacity Region



- The power control policy that corresponds to the rate pair  $(R_1^*, R_2^*)$  can be found by maximizing  $\mu_1 R_1 + \mu_2 R_2$  subject to the average power constraints, for some  $\mu_1, \mu_2$ .
- Any  $(R_1^*, R_2^*)$  on the curved portion of the boundary is a corner of one of the pentagons.
- If  $\mu_2 > \mu_1$  then the upper corner; if  $\mu_1 > \mu_2$  then the lower corner.

## Achieving Arbitrary Rate Tuples on the Boundary

- For given priorities  $\mu_i$ , maximize  $C_\mu \triangleq \mu_1 R_1 + \mu_2 R_2$  subject to  $E_{\mathbf{h}}[p_i(\mathbf{h})] \leq P_i$ ,  $\mathbf{R} \in \mathcal{C}$ .
- Wlog, assume  $\mu_2 > \mu_1$ . For a given power control policy, the optimum  $\mathbf{R}$  is the upper corner.
- The coordinates of the upper corner are:

$$R_2 = C_2, \quad R_1 = C_s - C_2$$

and

$$\begin{aligned} C_\mu &= \mu_1(C_s - C_2) + \mu_2 C_2 \\ &= (\mu_2 - \mu_1)C_2 + \mu_1 C_s \end{aligned}$$

- Therefore, the optimum power allocation policy is the solution to:

$$\begin{aligned} \max_{\mathbf{p}(\mathbf{h})} \quad & E_{\mathbf{h}} \left[ (\mu_2 - \mu_1) \log \left( 1 + \frac{p_2(\mathbf{h})h_2}{\sigma^2} \right) + \mu_1 \log \left( 1 + \frac{p_1(\mathbf{h})h_1 + p_2(\mathbf{h})h_2}{\sigma^2} \right) \right] \\ \text{s.t.} \quad & E_{\mathbf{h}}[p_i(\mathbf{h})] \leq P_i, \quad i = 1, 2 \\ & p_i(\mathbf{h}) \geq 0, \quad \forall \mathbf{h}, \quad i = 1, 2 \end{aligned}$$

- The objective function is concave, and the constraint set is convex in powers.

## Optimum Power Allocation - Generalized Iterative Waterfilling

- $\mathbf{p}^*(\mathbf{h})$  achieves the global maximum of  $C_\mu$  iff it satisfies the extended KKT conditions,

$$\sum_{i=1}^k \frac{(\mu_i - \mu_{i-1}) h_k}{\sum_{j=1}^i p_j(\mathbf{h}) h_j + p_k(\mathbf{h}) h_k} \leq \lambda_k, \quad \forall \mathbf{h}, k = 1, \dots, K$$

with equality at  $\mathbf{h}$ , if  $p_k(\mathbf{h}) > 0$ .

- Solution based on utilities (Tse-Hanly 1998).
- Solve the KKTs iteratively: **generalized iterative waterfilling** (Kaya-Ulukus)
  - One-user-at-a-time algorithm, converges to the optimum.
  - Gradually increase power allocated to channel states, in a “specific” order.
  - For user  $k$ , power levels can be obtained by solving  $k$ th order polynomial equations.

## Special Case: Sum Capacity

- All equal  $\mu_i$ , i.e.,  $\mu_i = \mu = 1$ .
- Maximize ergodic sum capacity, given average power constraints

$$\begin{aligned} \max_{\mathbf{p}(\mathbf{h})} \quad & E_{\mathbf{h}} \left[ \log \left( 1 + \frac{p_1(\mathbf{h})h_1 + p_2(\mathbf{h})h_2}{\sigma^2} \right) \right] \\ \text{s.t.} \quad & E_{\mathbf{h}} [p_i(\mathbf{h})] \leq P_i, \quad p_i(\mathbf{h}) \geq 0, \quad i = 1, 2 \end{aligned}$$

- KKT conditions

$$\frac{h_k}{p_1(\mathbf{h})h_1 + p_2(\mathbf{h})h_2} \leq \lambda_k, \quad \forall \mathbf{h}, k = 1, 2$$

with equality at  $\mathbf{h}$ , if  $p_k(\mathbf{h}) > 0$ .

- For both users to transmit simultaneously at channel state  $\mathbf{h} = (h_1, h_2)$ , we need

$$\frac{h_1}{h_2} = \frac{\lambda_1}{\lambda_2}$$

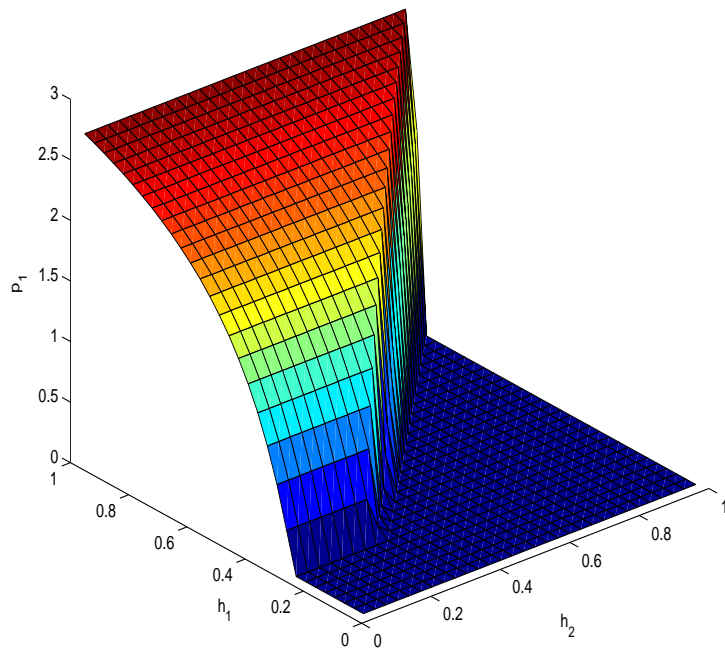
- For continuous channel gains, this is a zero-probability event.
- Only the strongest (after some scaling) user transmits at any given time.

## Sum Capacity – Closed-form Solution (Knopp-Humblet 1995)

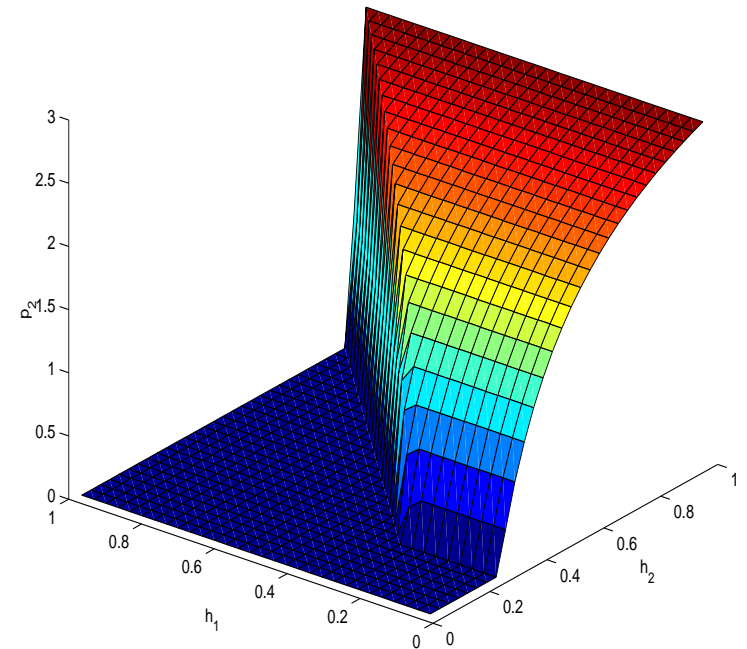
- Closed-form optimal power allocation: single-user waterfilling over favorable channel states

$$p_k(\mathbf{h}) = \begin{cases} \left( \frac{1}{\lambda_k} - \frac{\sigma^2}{h_k} \right)^+, & \text{if } h_k/\lambda_k > h_j/\lambda_j, \quad \forall j \neq k \\ 0, & \text{otherwise} \end{cases}$$

Power Distribution of User 1



Power Distribution of User 2

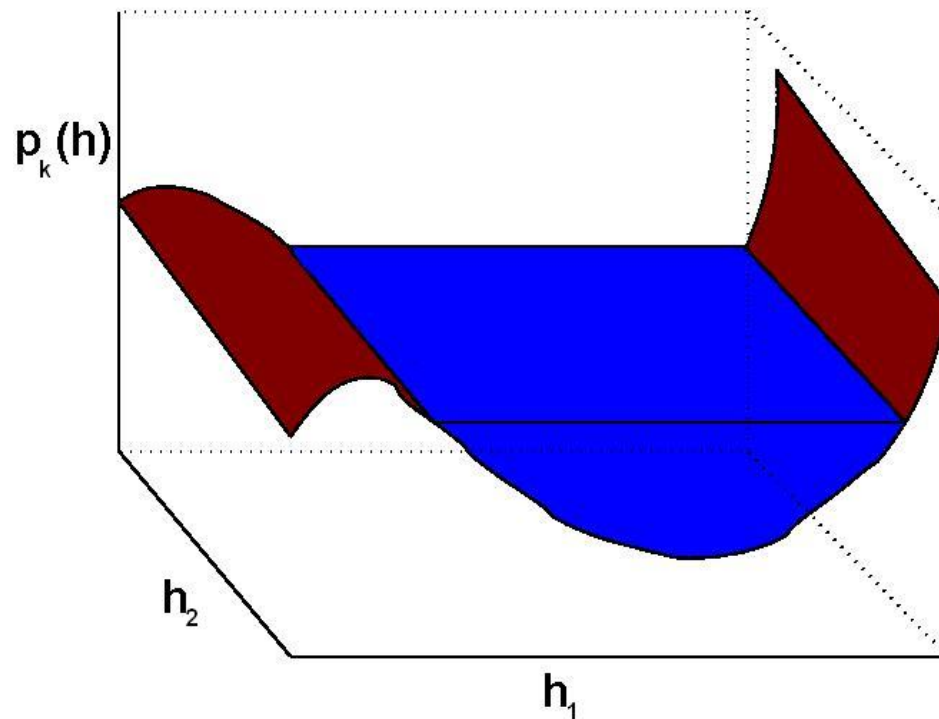


## Sum Capacity – Algorithmic Solution

- One-user-at-a-time iterative waterfilling (Kaya-Ulukus)

$$p_1(\mathbf{h}) = \left( \frac{1}{\lambda_1} - \frac{p_2(\mathbf{h})h_2 + \sigma^2}{h_1} \right)^+$$

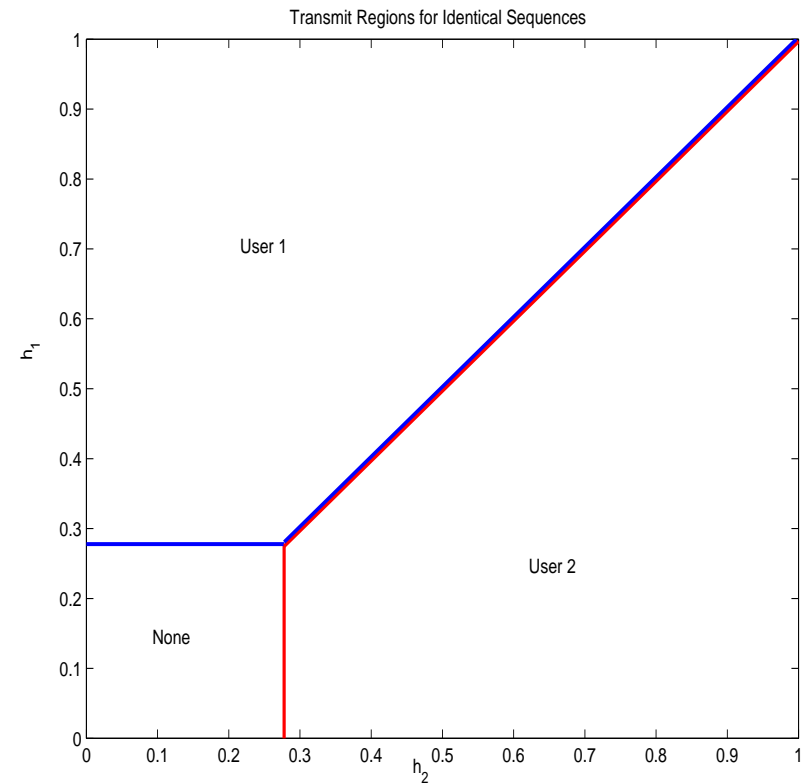
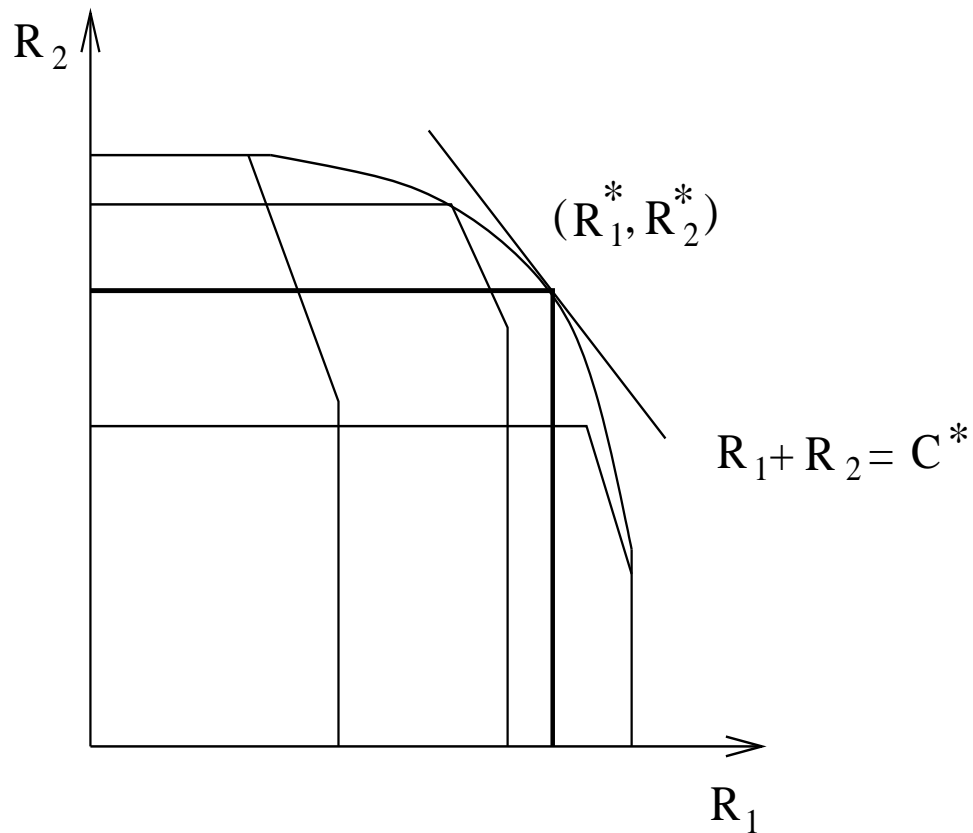
$$p_2(\mathbf{h}) = \left( \frac{1}{\lambda_2} - \frac{p_1(\mathbf{h})h_1 + \sigma^2}{h_2} \right)^+$$





## Issue of Simultaneous Transmissions Revisited

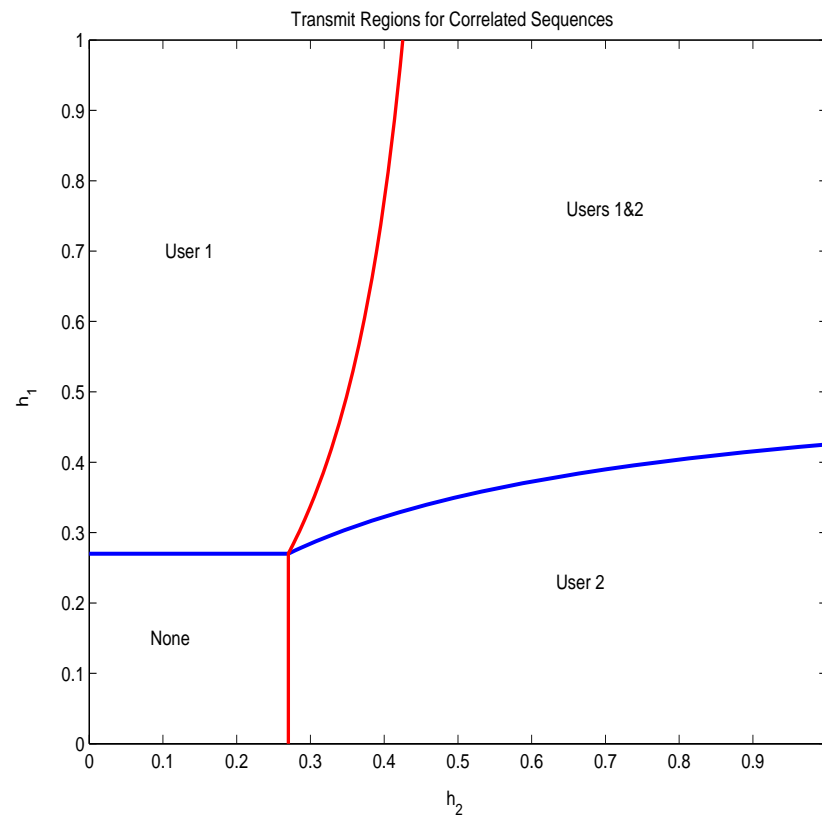
- Sum capacity achieving pentagon: a rectangle (orthogonal transmissions)



- “Only the strongest user transmits” result is specific to **sum capacity** and **scalar channel**.

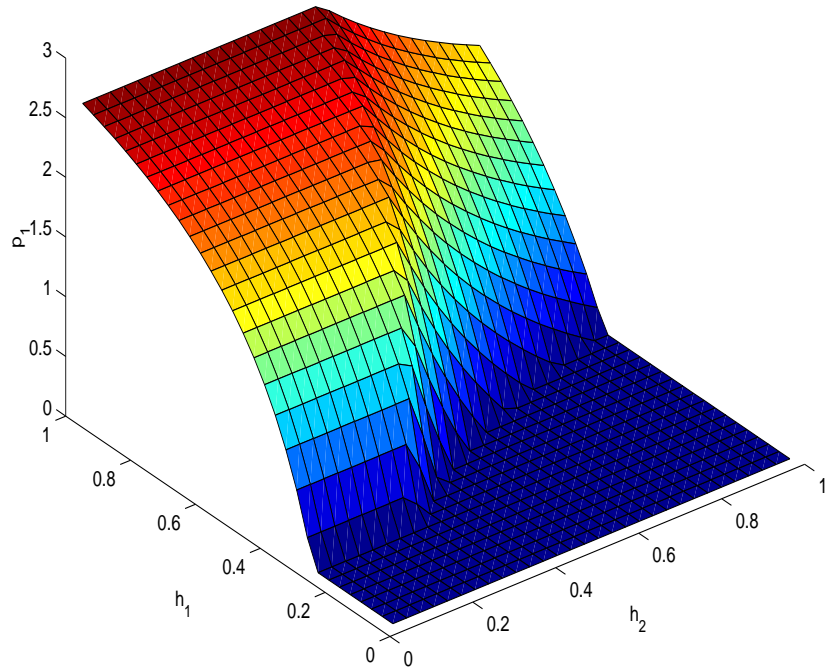
## Simultaneous Transmit Condition in CDMA (A Vector Channel)

- There exists a non-zero probability region of fading states  $\mathbf{h}$  where all  $K$  users in the system transmit simultaneously, if and only if  $\{\mathbf{s}_i; \mathbf{s}_i^\top\}_{i=1}^K$  are linearly independent.
- When  $K \leq N$ , for a set of  $K$  linearly independent signature sequences, there always exists a non-zero probability region of channel states where all  $K$  users transmit simultaneously.

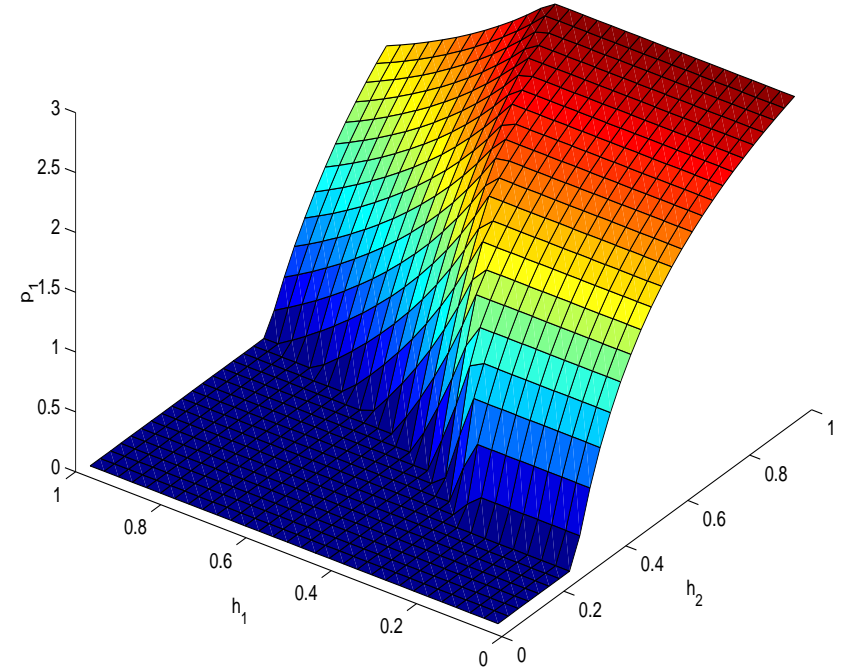


# Transmit Powers: Arbitrary Signatures (CDMA)

Power Distribution of User 1

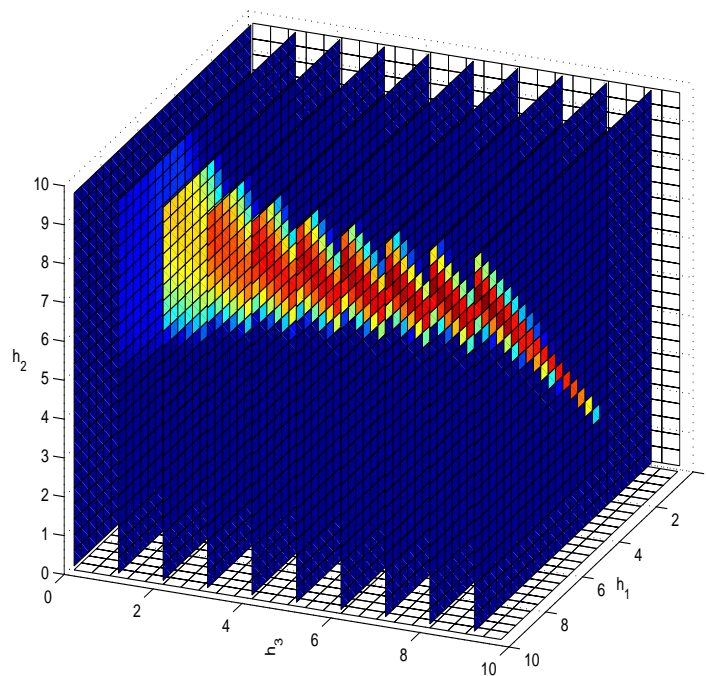


Power Distribution of User 2



## Maximum Number of Simultaneous Transmissions (CDMA)

- For a set of  $K$  signature sequences and processing gain  $N$ , the number of users that can transmit simultaneously cannot be larger than  $N(N + 1)/2$ .

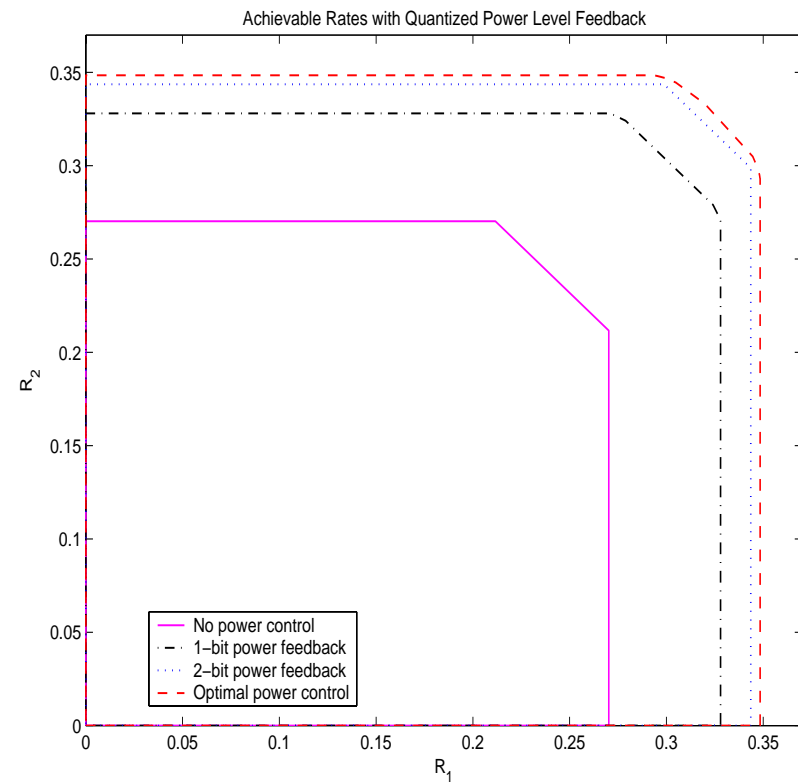
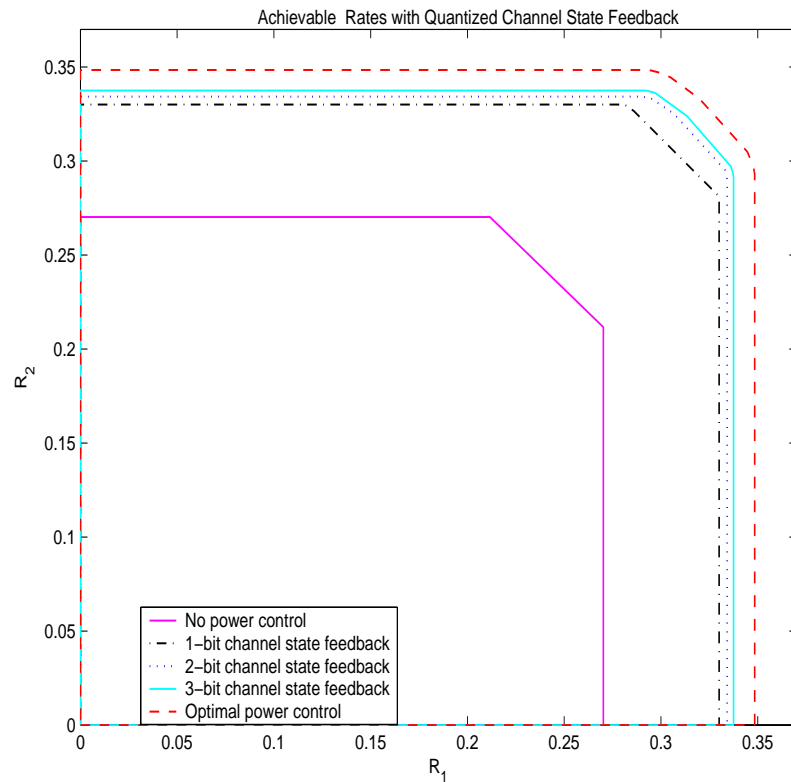


- Signature sequences  $\{\mathbf{s}_i\}_{i=1}^K$  are linearly dependent, but  $\{\mathbf{s}_i \mathbf{s}_i^\top\}_{i=1}^K$  are linearly independent.

## Power Control with Limited Feedback

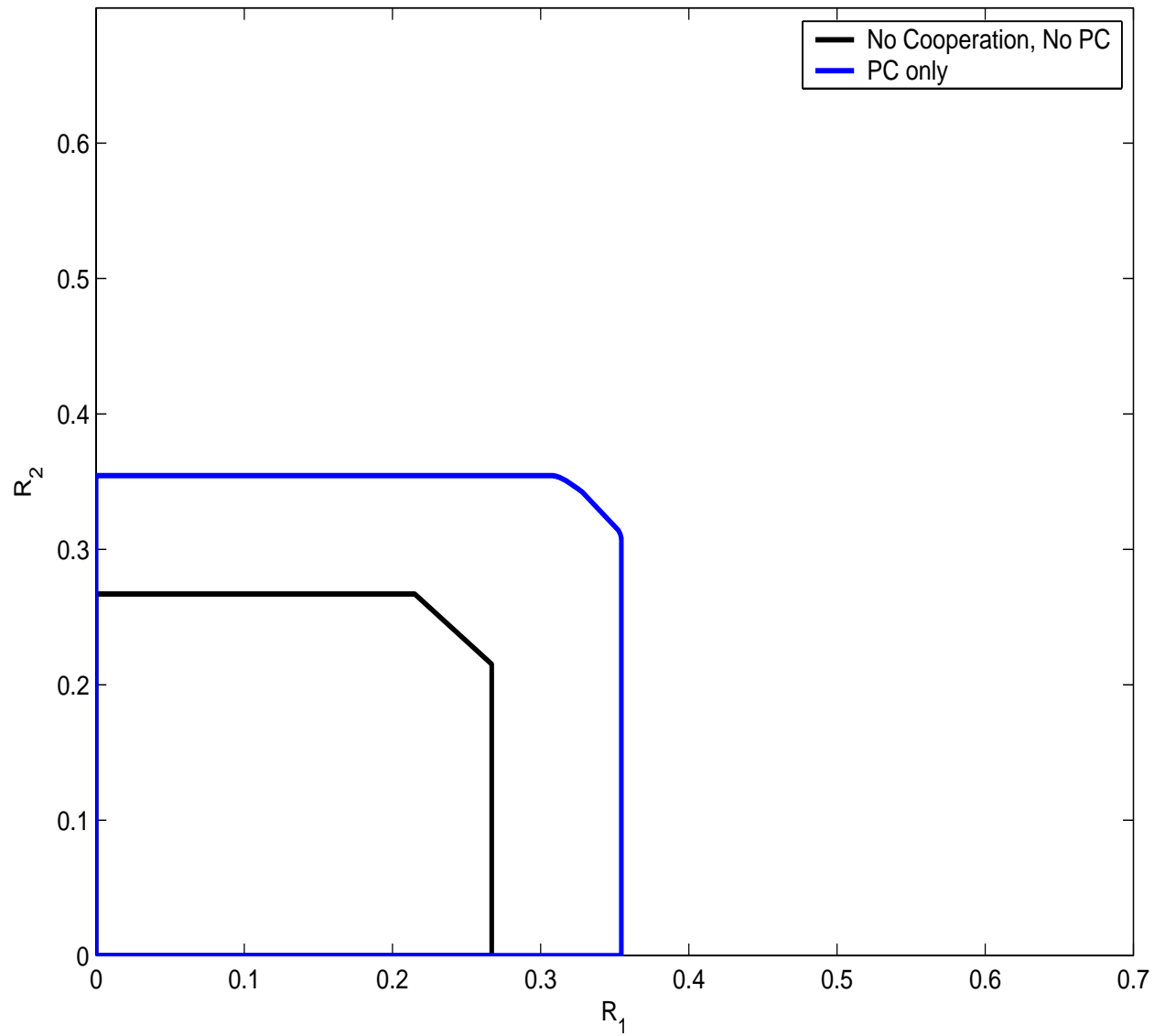
- In practice, CSI feedback from the receiver to the transmitters is very limited.
- Can formulate an optimal vector quantization problem to find the quantization bins for  $\mathbf{h}$  and quantizer output levels  $p(\mathbf{h})$  for given feedback rate constraints, difficult to solve.
- Two simple approaches to reduce the feedback information
  - Quantize the channel states to the nearest (MMSE sense) discrete levels, feed them back, design power control w.r.t. the quantized discrete random variable.
  - Obtain optimal power levels  $p(\mathbf{h})$ , find their distribution, quantize w.r.t. this induced distribution to the nearest (MMSE sense) discrete power levels, and feed back.

## Achievable Rate Regions with Limited Feedback

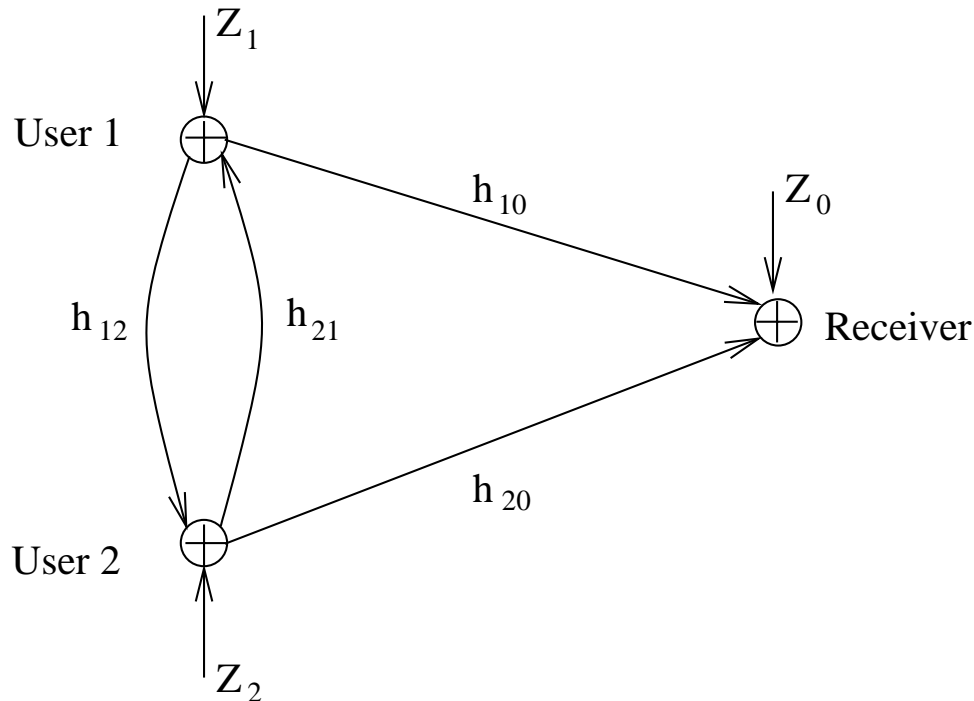


- We obtain rates very close to the capacity with perfect CSI, with very little feedback per user.

## Conclusion: Rate Region of Fading MAC with Power Control



## User Cooperation



$$Y_0 = \sqrt{h_{10}}X_1 + \sqrt{h_{20}}X_2 + Z_0$$

$$Y_1 = \sqrt{h_{21}}X_2 + Z_1$$

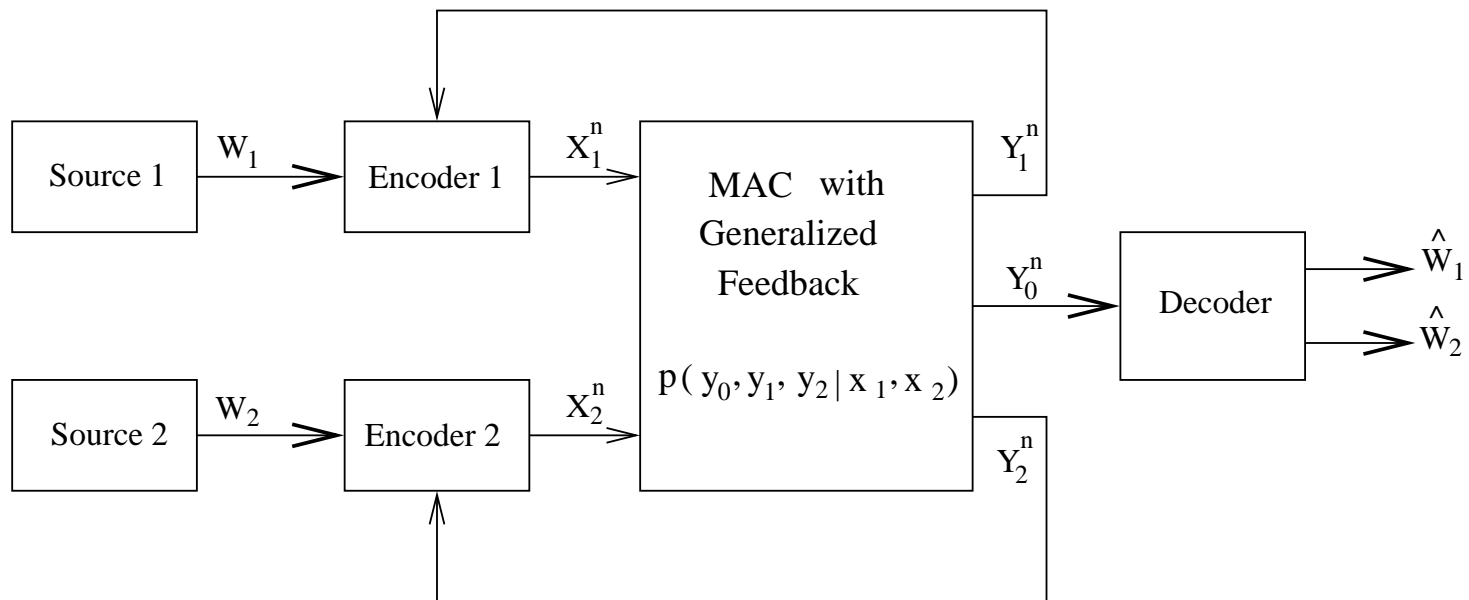
$$Y_2 = \sqrt{h_{12}}X_1 + Z_2$$

- **Interference** is **information**.
- Some versions of all transmitted signals are received by all nodes.
- Exploit overheard information to jointly design encoding, transmission, routing policies.
- Building block towards the analysis of ad-hoc and sensor networks.
- Information theoretically: multiple access channel with generalized feedback (capacity open).



## MAC with Generalized Feedback

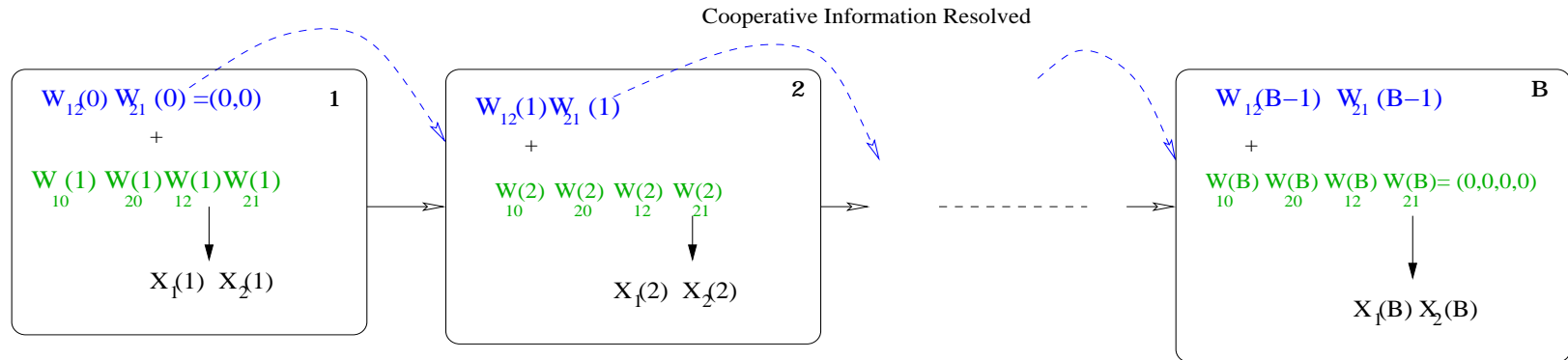
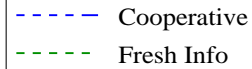
- Gaussian MAC with cooperating encoders (Sendonaris, Erkip, Aazhang 2003)
- Special case of **MAC with generalized feedback** (Willems, van der Meulen, Schalkwijk 1983)



## MAC with Generalized Feedback – Achievable Rate Region

- An achievable rate region is obtained by employing
  - **Block Markov superposition encoding**
    - \* Inject high rate fresh information to be resolved with the help of upcoming blocks.
    - \* Send resolution information for previous blocks.
  - **Backward decoding**
    - \* After receiving all blocks, decode the resolution information in the last block.
    - \* Using previously decoded resolution information, sequentially decode earlier blocks.

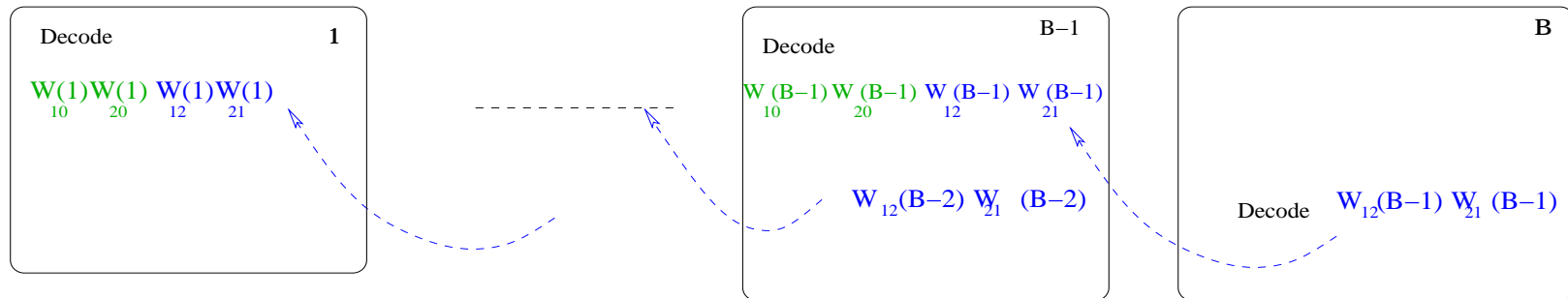
# Block Markov Superposition Encoding and Backward Decoding



Block Markov Superposition Encoding

$$R_{12} < C ( h_{12} P_{12} / h_{12} P_{10} + \sigma_1^2 )$$

$$R_{21} < C ( h_{21} P_{21} / h_{21} P_{20} + \sigma_2^2 )$$



$$R_{10} < C ( h_{10} P_{10} / \sigma_0^2 ) \quad R_{20} < C ( h_{20} P_{20} / \sigma_0^2 )$$

$$R_{10} + R_{20} < C ( h_{10} P_{10} + h_{20} P_{20} / \sigma_0^2 )$$

$$R_{10} + R_{20} + R_{12} + R_{21} < C ( h_{10} P_{10} + h_{20} P_{20} + 2 \sqrt{h_{10} h_{20} P_{10} P_{20}} / \sigma_0^2 )$$

Backwards Decoding

## Gaussian MAC with User Cooperation – No Resource Allocation

- Block Markov superposition coding (Sendonaris, Erkip, Aazhang 2003)
  - Build common information ( $X_{12}, X_{21}$ )
  - Cooperatively send ( $U$ )
  - Inject new information ( $X_{10}, X_{20}$ )

$$X_1 = \sqrt{p_{10}}X_{10} + \sqrt{p_{12}}X_{12} + \sqrt{p_{u1}}U$$

$$X_2 = \sqrt{p_{20}}X_{20} + \sqrt{p_{21}}X_{21} + \sqrt{p_{u2}}U$$

- Amplitude of the each channel's gain is assumed to be known at the corresponding receiver.
- Phases of all channel gains are assumed to be known at the receiver and the transmitters
  - Coherent combining
- **Amplitudes of the channel gains are not known at the transmitters.**

## Achievable Region with Block Markov Encoding/Backward Decoding

- Union of regions (pentagons or triangles)

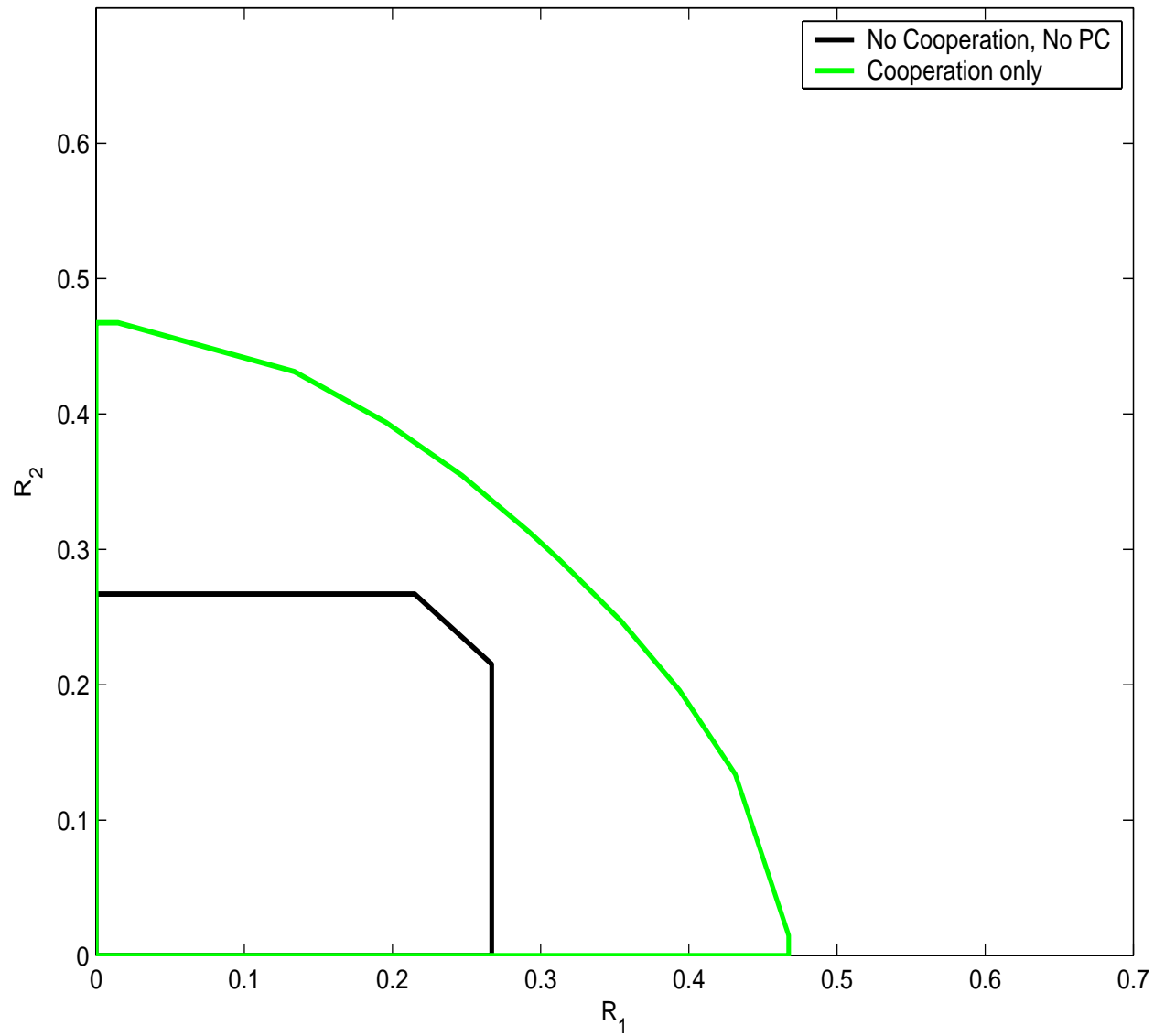
$$\begin{aligned}
 R_1 &< E \left[ \log \left( 1 + \frac{h_{12}p_{12}}{h_{12}p_{10} + \sigma_2^2} \right) + \log \left( 1 + \frac{h_{10}p_{10}}{\sigma_0^2} \right) \right] \\
 R_2 &< E \left[ \log \left( 1 + \frac{h_{21}p_{21}}{h_{21}p_{20} + \sigma_1^2} \right) + \log \left( 1 + \frac{h_{20}p_{20}}{\sigma_0^2} \right) \right] \\
 R_1 + R_2 &< \min \left\{ E \left[ \log \left( 1 + \frac{h_{10}p_1 + h_{20}p_2 + 2\sqrt{h_{10}h_{20}p_{U_1}p_{U_2}}}{\sigma_0^2} \right) \right], \right. \\
 &\quad \left. E \left[ \log \left( 1 + \frac{h_{12}p_{12}}{h_{12}p_{10} + \sigma_2^2} \right) + \log \left( 1 + \frac{h_{21}p_{21}}{h_{21}p_{20} + \sigma_1^2} \right) + \log \left( 1 + \frac{h_{10}p_{10} + h_{20}p_{20}}{\sigma_0^2} \right) \right] \right\}
 \end{aligned}$$

over all valid (**channel-independent**) power distributions

$$p_{10} + p_{12} + p_{U_1} \leq P_1$$

$$p_{20} + p_{21} + p_{U_2} \leq P_2$$

# Achievable Rate Region for MAC with User Cooperation



## Gaussian MAC with User Cooperation – Resource Allocation

- Choose transmit powers as functions of the channel state

$$X_1 = \sqrt{p_{10}(\mathbf{h})}X_{10} + \sqrt{p_{12}(\mathbf{h})}X_{12} + \sqrt{p_{u1}(\mathbf{h})}U$$

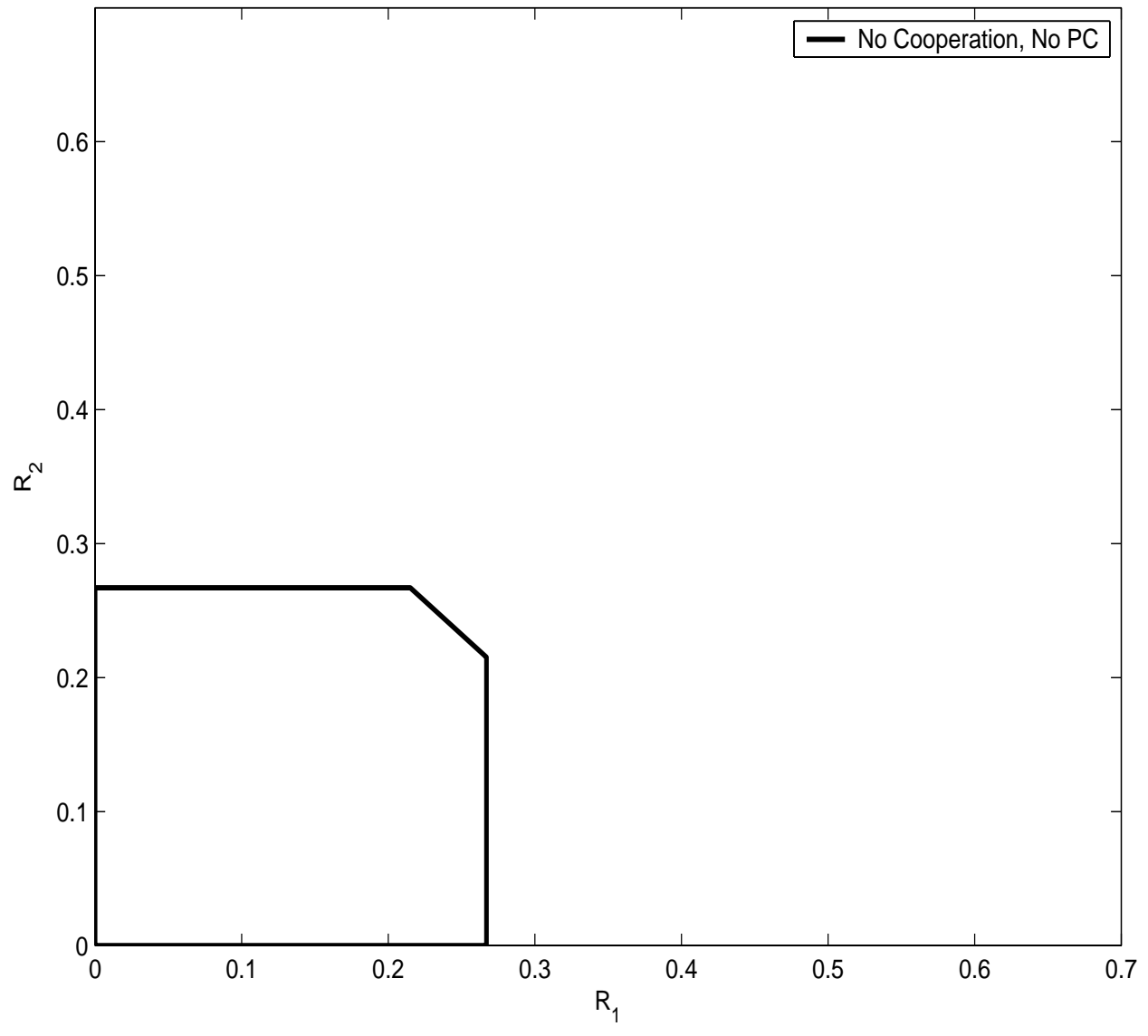
$$X_2 = \sqrt{p_{20}(\mathbf{h})}X_{20} + \sqrt{p_{21}(\mathbf{h})}X_{21} + \sqrt{p_{u2}(\mathbf{h})}U$$

- Complete channel state information at the transmitters and the receiver.
  - Only additional CSI: channel amplitudes at the transmitters.
- Transmitted codewords are modulated by channel adaptive power levels
  - Opportunistic cooperation and transmission – use available average power efficiently.
- A special case is power control only

$$X_1 = \sqrt{p_{10}(\mathbf{h})}X_{10}$$

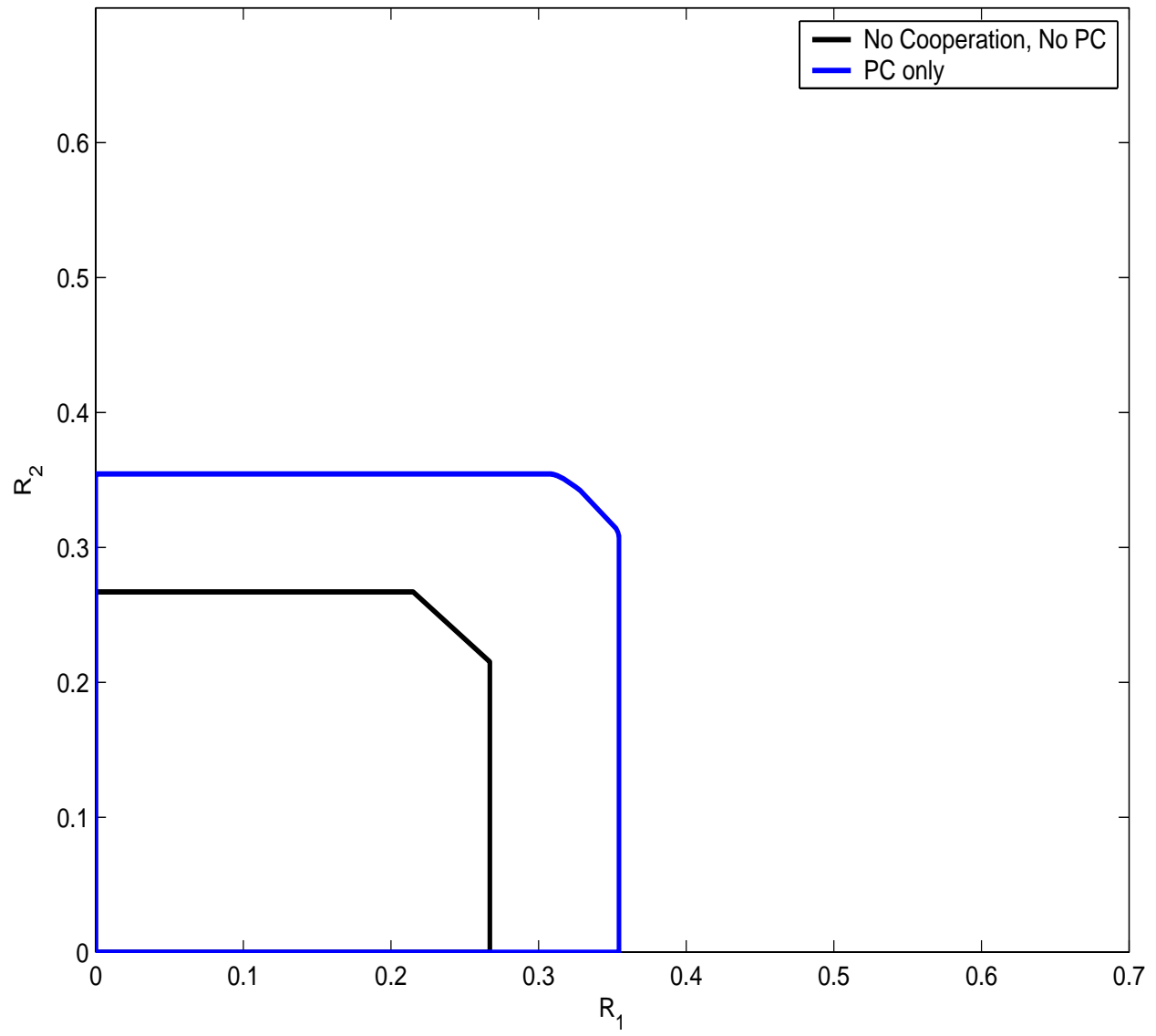
$$X_2 = \sqrt{p_{20}(\mathbf{h})}X_{20}$$

# No Cooperation, No Power Control

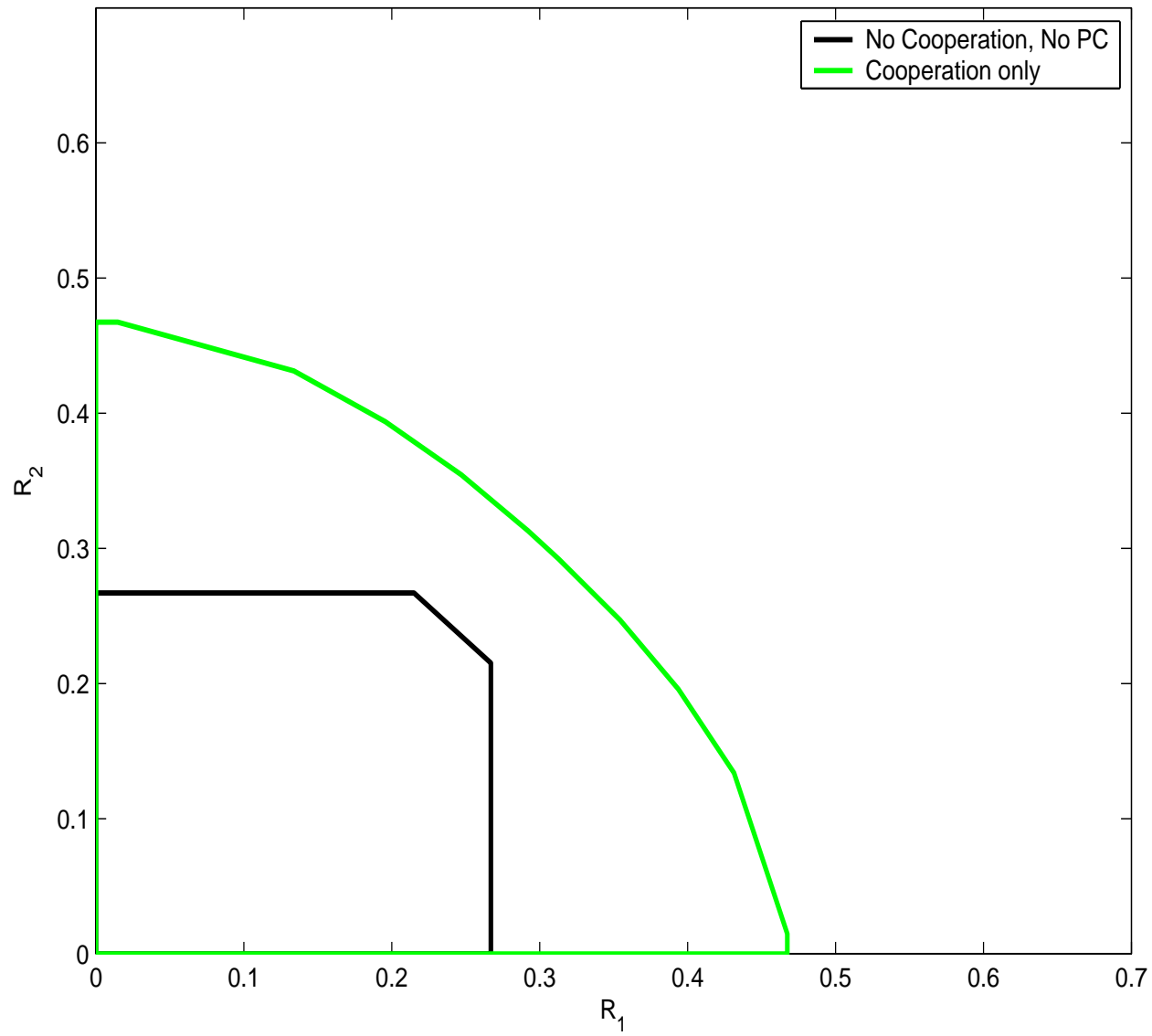




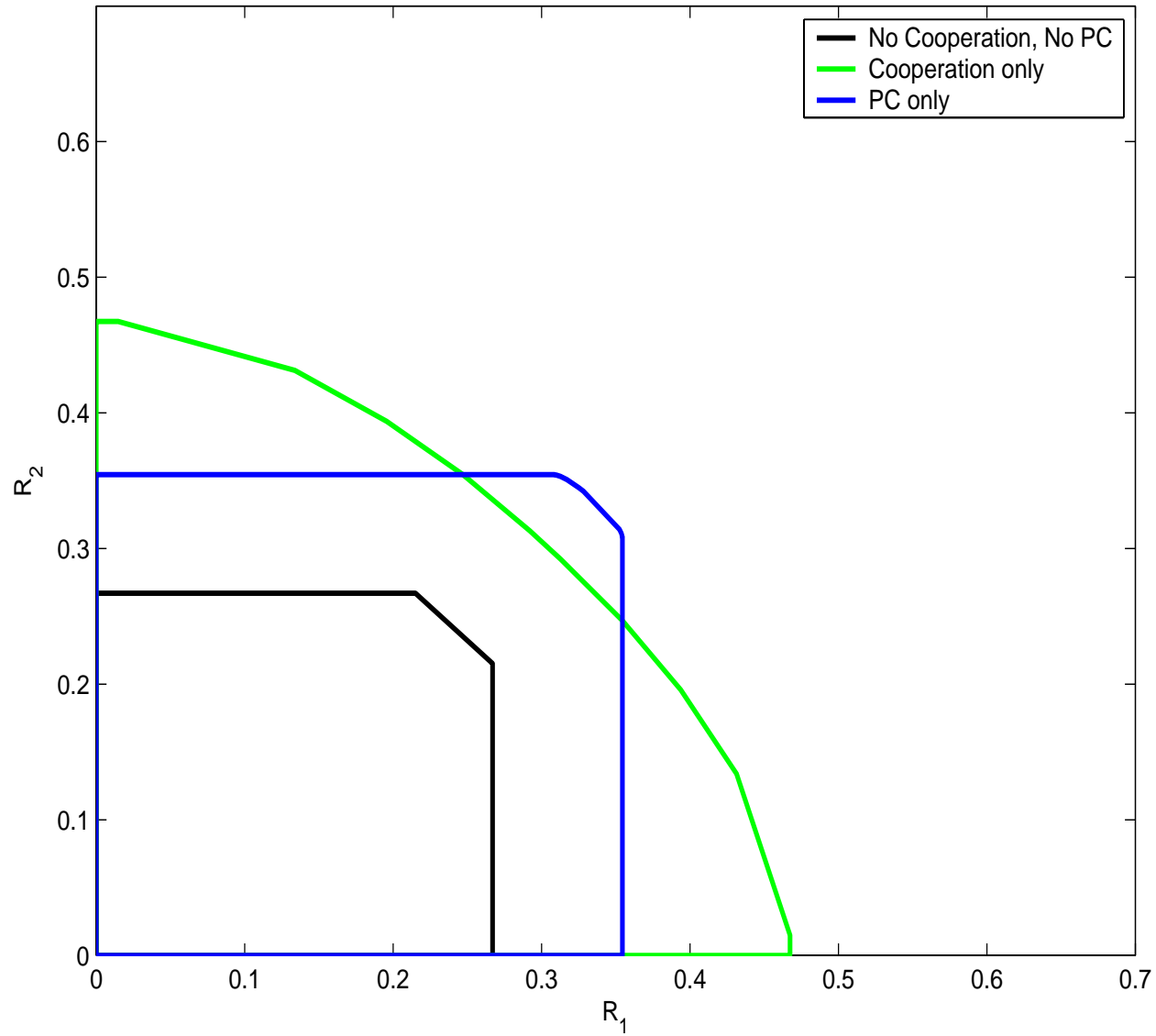
# Power Control Only



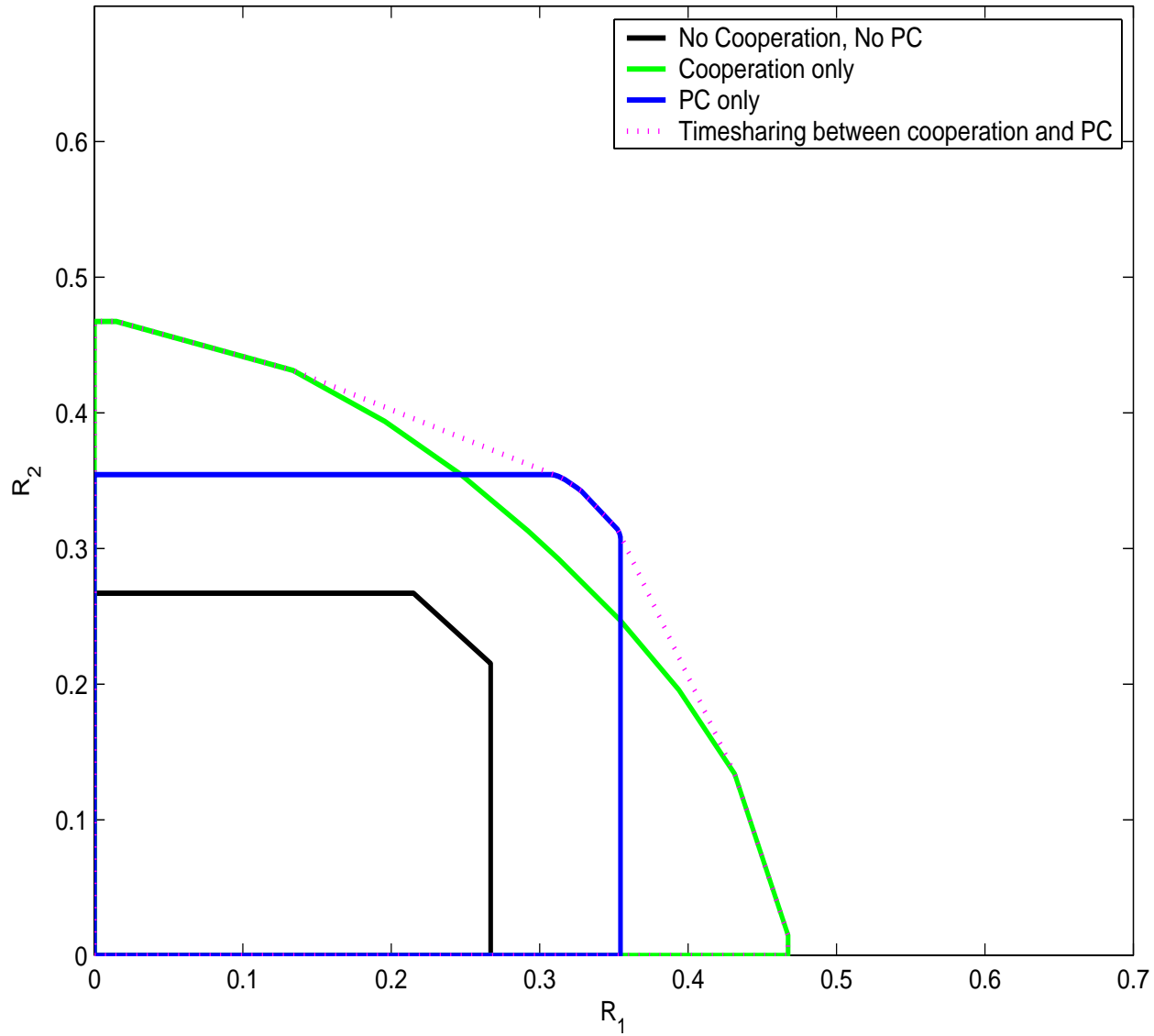
# User Cooperation Only



# Power Control Only and User Cooperation Only



# Power Control Only, User Cooperation Only and Timesharing



## Achievable Region of Rates with Power Control

- Union of regions (pentagons and triangles)

$$R_1 < E \left[ \log \left( 1 + \frac{h_{12}p_{12}(\mathbf{h})}{h_{12}p_{10}(\mathbf{h}) + \sigma_2^2} \right) + \log \left( 1 + \frac{h_{10}p_{10}(\mathbf{h})}{\sigma_0^2} \right) \right]$$

$$R_2 < E \left[ \log \left( 1 + \frac{h_{21}p_{21}(\mathbf{h})}{h_{21}p_{20}(\mathbf{h}) + \sigma_1^2} \right) + \log \left( 1 + \frac{h_{20}p_{20}(\mathbf{h})}{\sigma_0^2} \right) \right]$$

$$R_1 + R_2 < \min \left\{ E \left[ \log \left( 1 + \frac{h_{10}p_1(\mathbf{h}) + h_{20}p_2(\mathbf{h}) + 2\sqrt{h_{10}h_{20}p_{U_1}(\mathbf{h})p_{U_2}(\mathbf{h})}}{\sigma_0^2} \right) \right], \right.$$

$$\left. E \left[ \log \left( 1 + \frac{h_{12}p_{12}(\mathbf{h})}{h_{12}p_{10}(\mathbf{h}) + \sigma_2^2} \right) + \log \left( 1 + \frac{h_{21}p_{21}(\mathbf{h})}{h_{21}p_{20}(\mathbf{h}) + \sigma_1^2} \right) + \log \left( 1 + \frac{h_{10}p_{10}(\mathbf{h}) + h_{20}p_{20}(\mathbf{h})}{\sigma_0^2} \right) \right] \right\}$$

over all (**channel-dependent**) feasible power control policies

$$E[p_{10}(\mathbf{h}) + p_{12}(\mathbf{h}) + p_{U_1}(\mathbf{h})] \leq P_1$$

$$E[p_{20}(\mathbf{h}) + p_{21}(\mathbf{h}) + p_{U_2}(\mathbf{h})] \leq P_2$$

- Bounds are not concave in power vector  $\mathbf{p}(\mathbf{h}) = [p_{10}(\mathbf{h}) \ p_{12}(\mathbf{h}) \ p_{U_1}(\mathbf{h}) \ p_{20}(\mathbf{h}) \ p_{21}(\mathbf{h}) \ p_{U_2}(\mathbf{h})]$

## Properties of the Sum-Rate-Optimal Power Allocation

- Let the effective channel gains normalized by the noise powers be defined as  $s_{ij} = h_{ij}/\sigma_j^2$ . Then, for the power control policy  $\mathbf{p}^*(\mathbf{h})$  that maximizes the sum rate, we need

- $p_{10}^*(\mathbf{h}) = p_{20}^*(\mathbf{h}) = 0$ , if  $s_{12} > s_{10}$  and  $s_{21} > s_{20}$
- $p_{10}^*(\mathbf{h}) = p_{21}^*(\mathbf{h}) = 0$ , if  $s_{12} > s_{10}$  and  $s_{21} \leq s_{20}$
- $p_{12}^*(\mathbf{h}) = p_{20}^*(\mathbf{h}) = 0$ , if  $s_{12} \leq s_{10}$  and  $s_{21} > s_{20}$

$$\left. \begin{array}{l}
 p_{12}^*(\mathbf{h}) = p_{21}^*(\mathbf{h}) = 0 \\
 \text{OR} \\
 p_{10}^*(\mathbf{h}) = p_{21}^*(\mathbf{h}) = 0 \\
 \text{OR} \\
 p_{12}^*(\mathbf{h}) = p_{20}^*(\mathbf{h}) = 0
 \end{array} \right\} \text{if } s_{12} \leq s_{10} \text{ and } s_{21} \leq s_{20}$$

## Implications of the Optimal Power Allocation

- Block Markov superposition coding is simpler than originally thought.
  - Each transmitter either sends a cooperation signal or fresh information, but not both!
- The choice at each channel state “only” depends on the channel state.
  - Channel statistics, power constraints play no role in deciding which signals to transmit.
  - Except for the tiny little last case ... which usually has very insignificant probability.
- The achievable rate expressions are greatly simplified, and are now concave.
- This simplified coding policy not only maximizes the sum rate, but also the individual rate constraints on  $R_1$  and  $R_2$ , and is optimal in terms of the entire rate region.
- Concave optimization problem over a convex constraint set, but non-differentiable.

## Simplified Rate Region – Example

- Assume  $s_{12} > s_{10}$ ,  $s_{21} > s_{20}$  to illustrate the simplified rate region.

$$R_1 < E [\log (1 + s_{12}p_{12}(\mathbf{h}))]$$

$$R_2 < E [\log (1 + s_{21}p_{21}(\mathbf{h}))]$$

$$R_1 + R_2 < \min \left\{ E \left[ \log \left( 1 + s_{10}p_1(\mathbf{h}) + s_{20}p_2(\mathbf{h}) + 2\sqrt{s_{10}s_{20}p_{U_1}(\mathbf{h})p_{U_2}(\mathbf{h})} \right) \right], \right. \\ \left. E \left[ \log (1 + s_{12}p_{12}(\mathbf{h})) + \log (1 + s_{21}p_{21}(\mathbf{h})) \right] \right\}$$

- Inequalities define either a pentagon like in the traditional MAC, or a triangle.
- All bounds are concave in powers, and so is any weighted sum  $\mu_1 R_1 + \mu_2 R_2$  at the corners.
- Sum rate is not differentiable where the arguments of the min are equal.



## Rate Maximization Using Subgradient Method

- Points on the rate region boundary can be obtained by maximizing  $C_\mu = \mu_1 R_1 + \mu_2 R_2$ .
- The optimization problem for arbitrary priorities  $\mu_1$  and  $\mu_2$  is given by

$$\begin{aligned} \max_{\mathbf{p}(\mathbf{h})} \quad & \mu_1 R_1 + \mu_2 R_2 \\ \text{s.t.} \quad & E_{3,4} [p_{10}(\mathbf{h})] + E_{1,2} [p_{12}(\mathbf{h})] + E [p_{U_1}(\mathbf{h})] \leq P_1 \\ & E_{2,4} [p_{20}(\mathbf{h})] + E_{1,3} [p_{21}(\mathbf{h})] + E [p_{U_2}(\mathbf{h})] \leq P_2 \end{aligned}$$

- $\{R_1, R_2\}$  is the corner of the pentagon obtained for a given power allocation policy.
- Gradient of the objective function does not exist everywhere, find subgradient  $\mathbf{g}$  instead

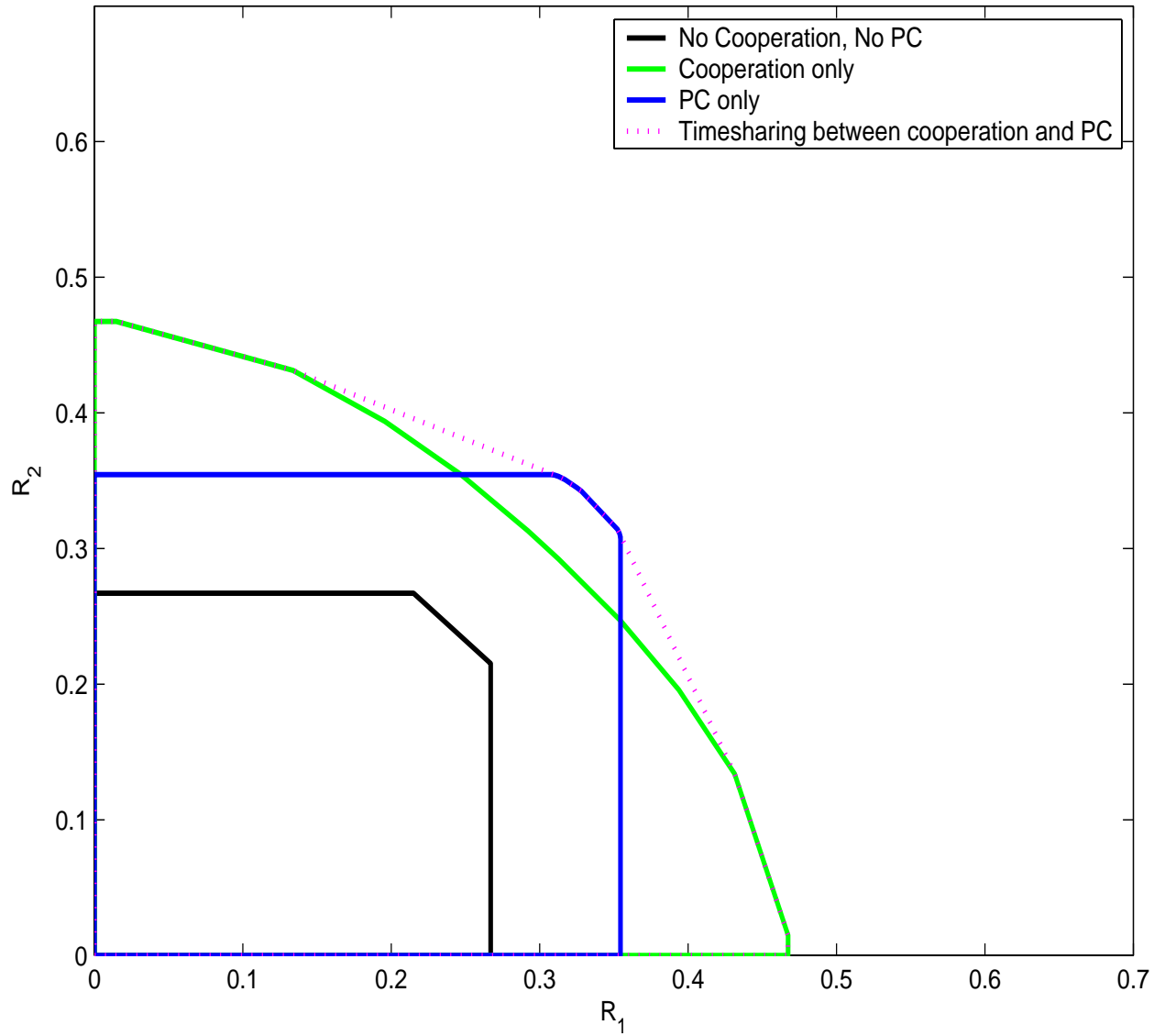
$$C_\mu(\mathbf{p}') \leq C_\mu(\mathbf{p}) + (\mathbf{p}' - \mathbf{p})^\top \mathbf{g}$$

- Use projected subgradient method to maximize  $C_\mu$

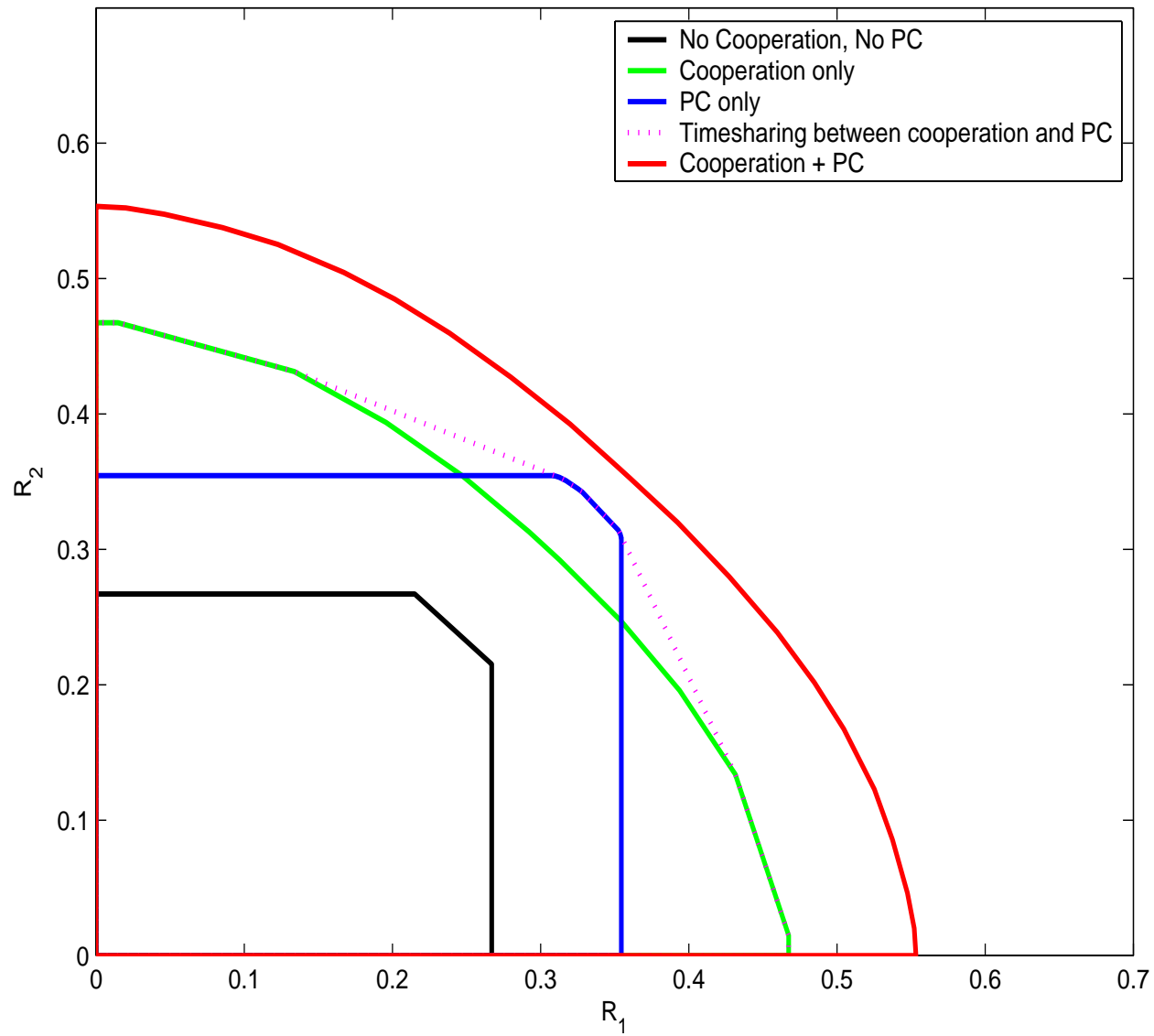
$$\mathbf{p}(k+1) = [\mathbf{p}(k) + \alpha_k \mathbf{g}_k]^+$$

- Provably converges for a diminishing stepsize  $\alpha_k$  sequence (Shor).

# Power Control Only, User Cooperation Only and Timesharing



# Achievable Rate Region for Joint Power Control and User Cooperation



## Summary

- Power control only and user cooperation only are techniques that improve capacity.
- User cooperation exploits **spatial diversity** and power control exploits **time diversity**.
- Joint power control and user cooperation exploits **both: major improvements in capacity**.
- Encoding and decoding is significantly simplified w.r.t. user cooperation only.
  - Transmitters send either cooperation or fresh information signals, but not both.
- The solution: joint power allocation, coding, medium accessing, and routing.
  - An instance of **cross-layer design**.
  - Trade-offs between complexity and capacity.
  - Value of the CSI.