

EE 231A Information Theory

Lecture 15

The Multiple Access Channel

- A. Capacity Region of Multiple Access Channel and Binary Multiplier MAC example
- B. Binary Erasure MAC example
- C. Achievability of Multiple Access Channel Capacity Region
- D. Gaussian Multiple Access Channel

Part 15A:

Capacity Region

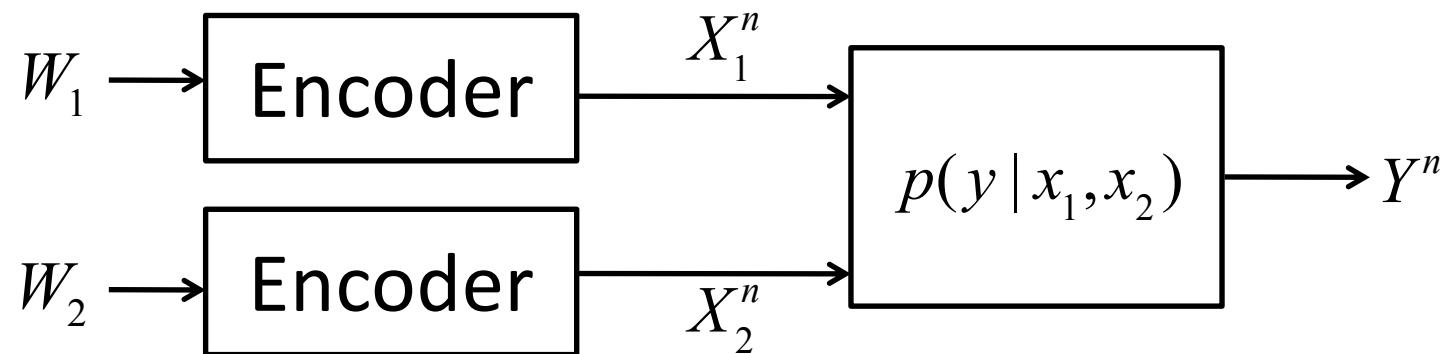
of the

Multiple Access Channel

and

Binary Multiplier MAC

Multiple Access DMC



$$\text{decode } g(Y^n) = (\hat{W}_1, \hat{W}_2)$$

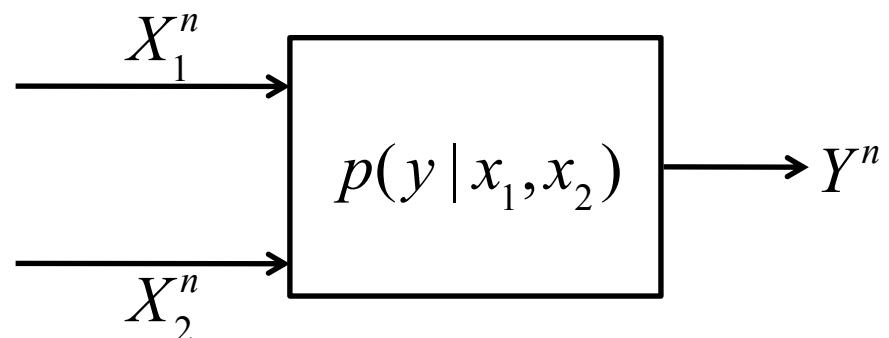
The “pentagon” inequalities

- For a fixed product distribution $p(x_1, x_2) = p(x_1)p(x_2)$, the set of achievable (R_1, R_2) rate pairs (the achievable region) is the set of all (R_1, R_2) pairs satisfying three inequalities (a “pentagon”):

$$R_1 < I(X_1; Y | X_2) = I_1$$

$$R_2 < I(X_2; Y | X_1) = I_2$$

$$R_1 + R_2 < I(X_1, X_2; Y) = I_3$$

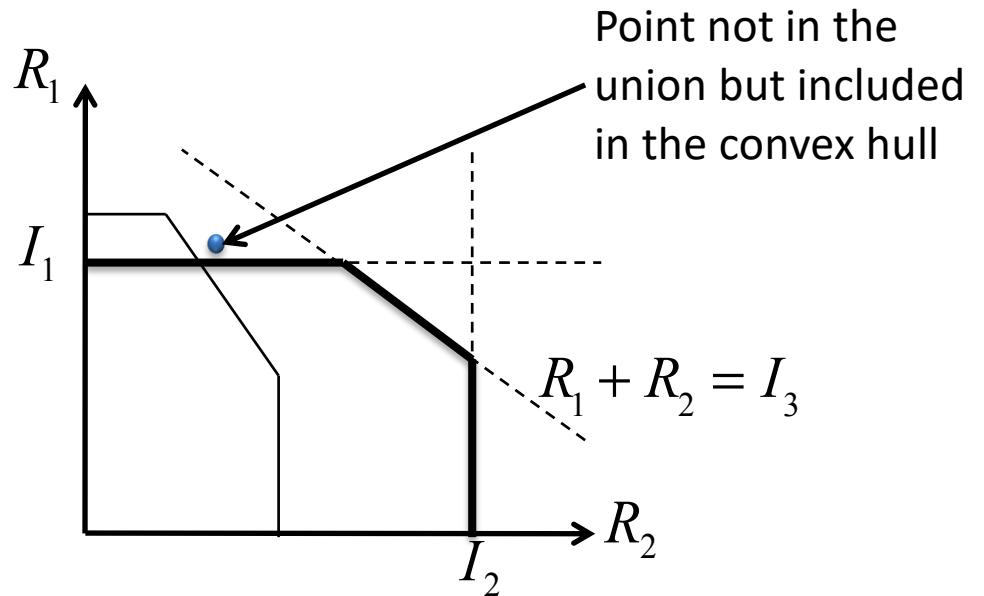


Capacity region (Theorem 15.3.1)

$$R_1 < I(X_1; Y | X_2) = I_1$$

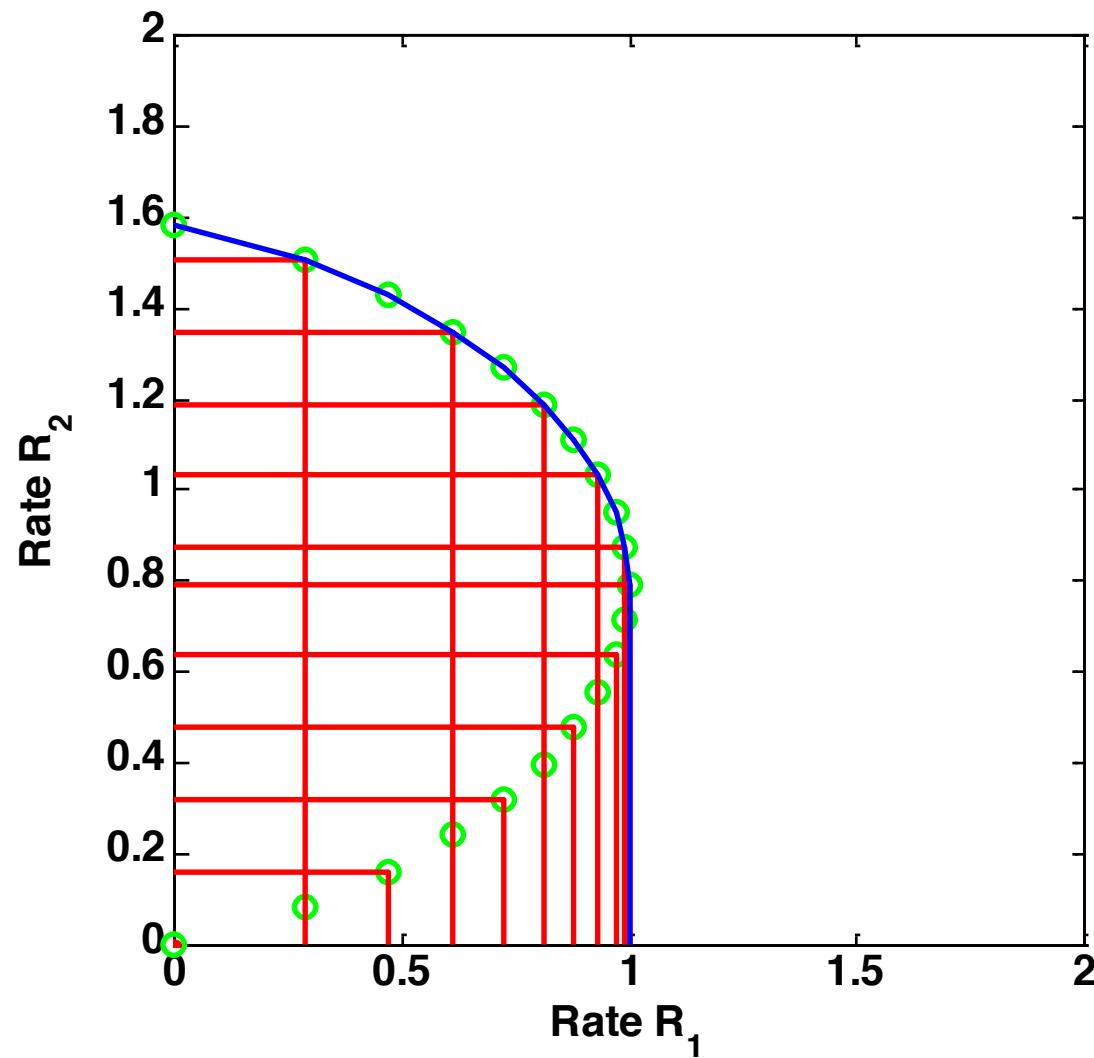
$$R_2 < I(X_2; Y | X_1) = I_2$$

$$R_1 + R_2 < I(X_1, X_2; Y) = I_3$$



- The capacity region is a union of many pentagons, possibly further increased by the closure of a **convex hull operation**.
 - Points in the convex hull are achieved by time-sharing between two points in pentagons.

Pentagons can be rectangles.



Example 1: Binary Multiplier Channel

- Binary Multiplier Channel $Y = X_1 X_2, \quad X_1, X_2 \in \{0,1\}$
- Describe all pentagons by finding the general pentagon induced by the general product distribution:

$$X_1 \sim \text{Bernoulli}(p_1)$$

$$X_2 \sim \text{Bernoulli}(p_2)$$

Computing I_1 , I_2 , and I_3

$$Y = X_1 X_2, \quad X_1, X_2 \in \{0,1\}$$

$$R_1 < I(X_1; Y | X_2) = I_1$$

$$R_2 < I(X_2; Y | X_1) = I_2$$

$$R_1 + R_2 < I(X_1, X_2; Y) = I_3$$

$$I_1 = I(X_1; Y | X_2)$$

$$= P(X_2 = 1)I(X_1; Y | X_2 = 1) + P(X_2 = 0)I(X_1; Y | X_2 = 0)$$

$$= p_2 H(p_1) + (1 - p_2) \cdot 0$$

$$= p_2 H(p_1)$$

$$I_2 = I(X_2; Y | X_1)$$

$$= p_1 H(p_2)$$

$$I_3 = I(X_1, X_2; Y)$$

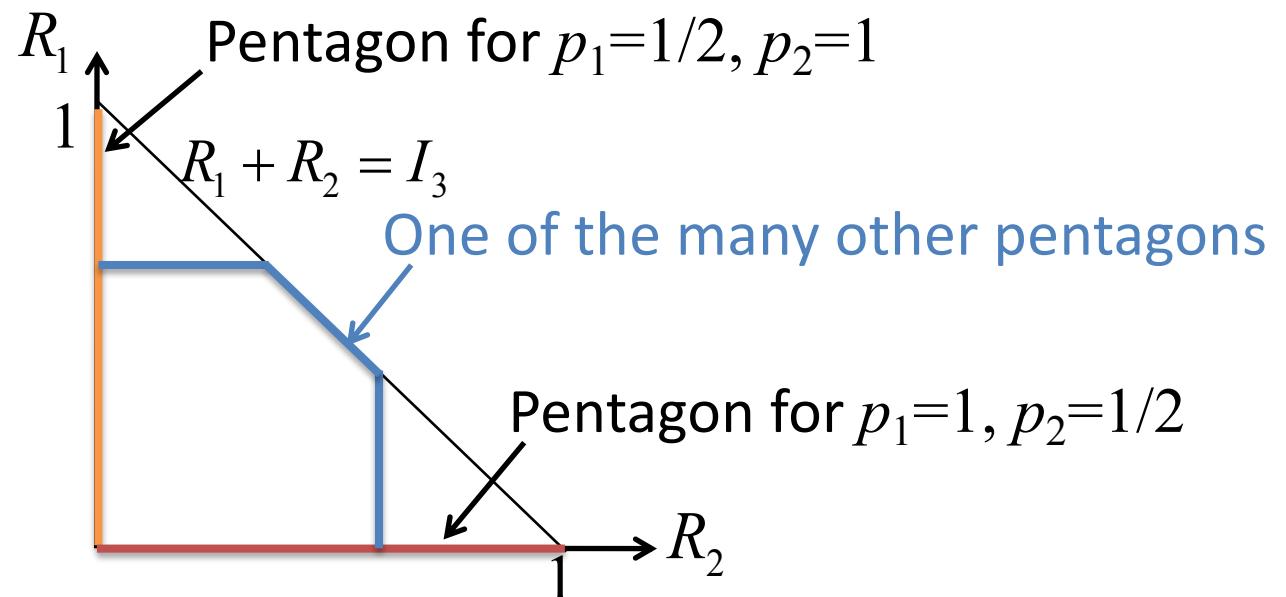
$$= H(Y) - H(Y | X_1, X_2)$$

$$= H(Y) \quad \text{where } Y = f(X_1, X_2)$$

$$= H(p_1 p_2)$$

Binary Multiplier Channel Pentagons, Rate Region

$$\begin{aligned}R_1 &< p_2 H(p_1) \\R_2 &< p_1 H(p_2) \\R_3 &< H(p_1 p_2)\end{aligned}$$



- Convex hull of these two pentagons gives the achievable region since $I_3 = I(X_1, X_2; Y) \leq H(Y) \leq 1$.
- However this rate region is also completely covered by the union of all pentagons. Convex hull is not necessary.

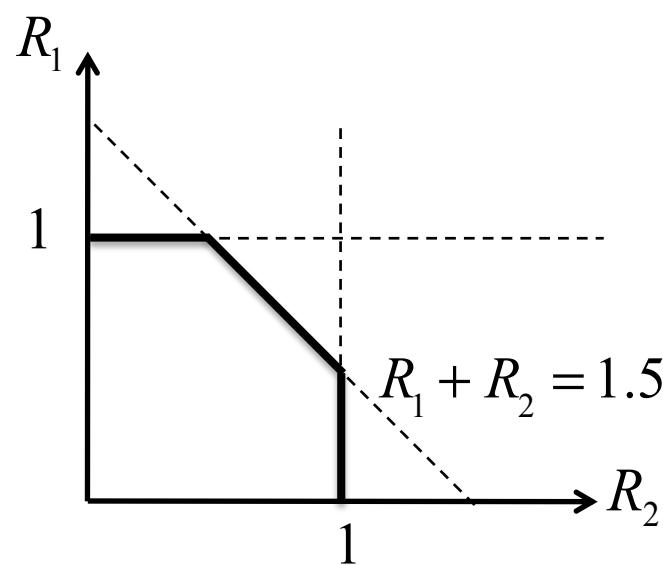
Part 15B:

Binary Erasure MAC

Example 2: Binary Erasure MAC

$p(X_1, X_2)$	$X_1 X_2$	Y
$(1 - p_1)(1 - p_2)$	00	0
$(1 - p_1)p_2$	01	1
$p_1(1 - p_2)$	10	1
$p_1 p_2$	11	2

$$Y = X_1 + X_2, \quad X_1, X_2 \in \{0, 1\}$$



Pentagon for $p_1 = p_2 = \frac{1}{2}$

$$\begin{aligned} I_1 &= I(X_1; Y | X_2) \\ &= 1 \end{aligned}$$

$$\begin{aligned} I_2 &= I(X_2; Y | X_1) \\ &= 1 \end{aligned}$$

$$\begin{aligned} I_3 &= I(X_1, X_2; Y) \\ &= H(Y) - H(Y | X_1, X_2) \\ &= H(Y) \\ &= 1.5 \end{aligned}$$

One pentagon that contains them all

- Now consider the general pentagon for

$$X_1 \sim \text{Bernoulli}(p_1), X_2 \sim \text{Bernoulli}(p_2)$$

$$I(X_1; Y | X_2) = H(p_1) \leq 1$$

$$I(X_2; Y | X_1) = H(p_2) \leq 1$$

$$I(X_1, X_2; Y) = H(Y) \leq 1.5 \quad \text{Need to show this}$$

- So every pentagon is within the pentagon for

$$p_1 = p_2 = \frac{1}{2}.$$

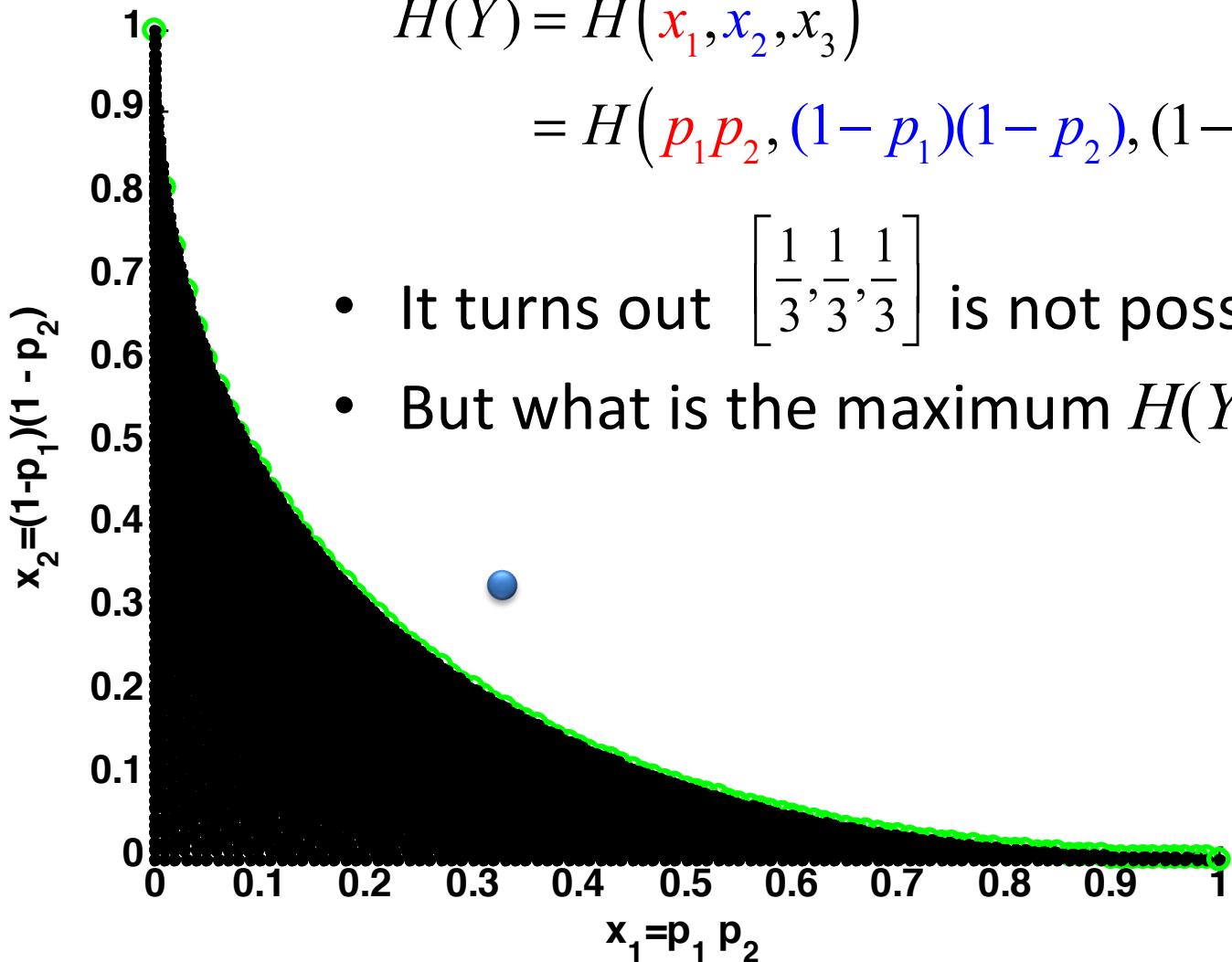
Maximum Possible $H(Y)$

- $H(Y) \leq \log 3$, but is this achievable?

$$H(Y) = H(x_1, x_2, x_3)$$

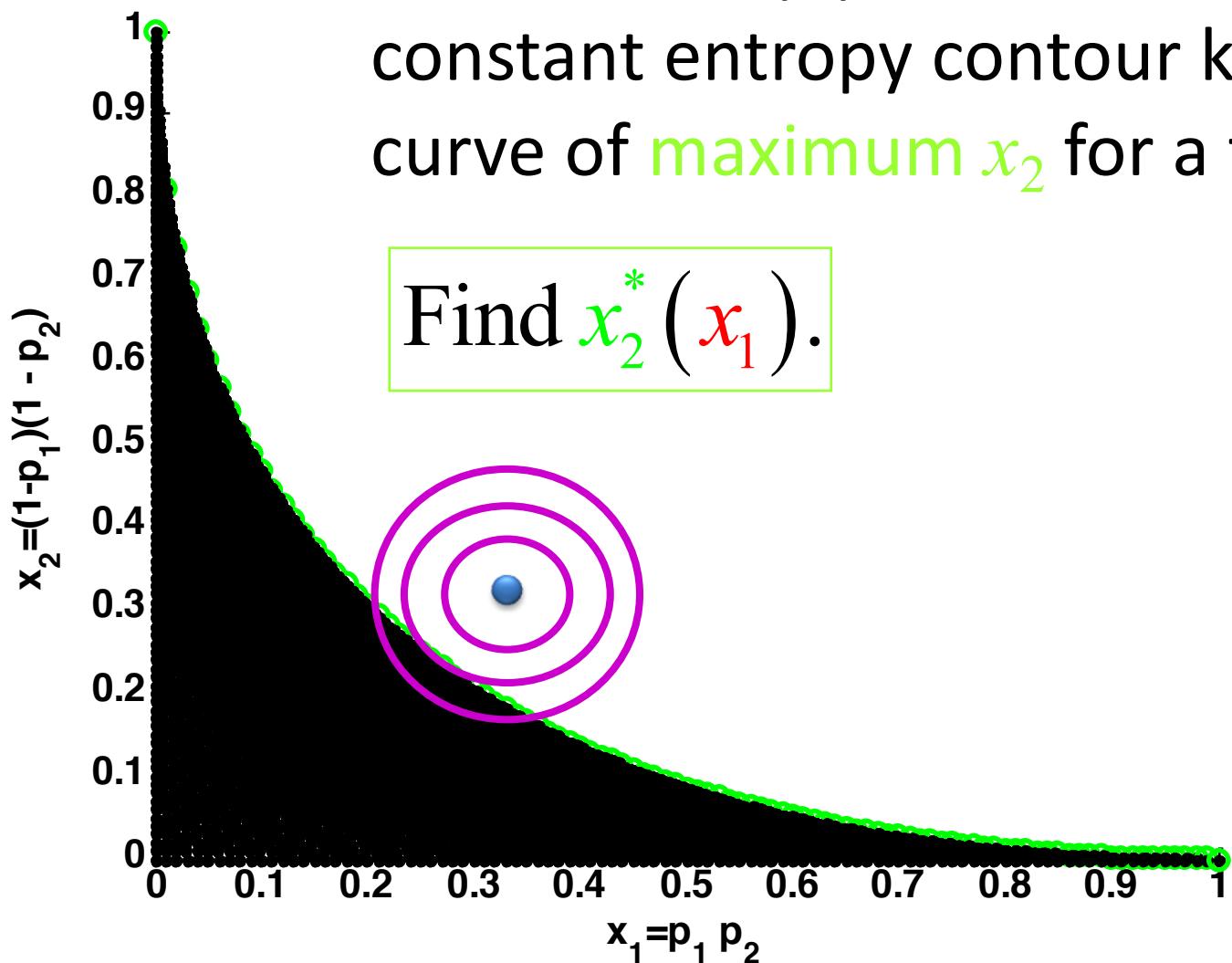
$$= H(p_1 p_2, (1-p_1)(1-p_2), (1-p_1)p_2 + p_1(1-p_2))$$

- It turns out $\left[\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right]$ is not possible.
- But what is the maximum $H(Y)$?



Contours of constant $H(Y)$

- Maximum $H(Y)$ will occur when a constant entropy contour kisses the green curve of **maximum x_2** for a fixed x_1 .



Find $x_2^*(x_1)$.

Finding the border of (x_1, x_2)

$$H(Y) = H(x_1, \color{blue}{x_2}, x_3)$$

$$= H(\color{red}{p_1 p_2}, (1-p_1)(1-p_2), (1-p_1)p_2 + p_1(1-p_2))$$

$$\color{blue}{x_2} = (1-p_1) \left(1 - \frac{\color{red}{x_1}}{p_1} \right)$$

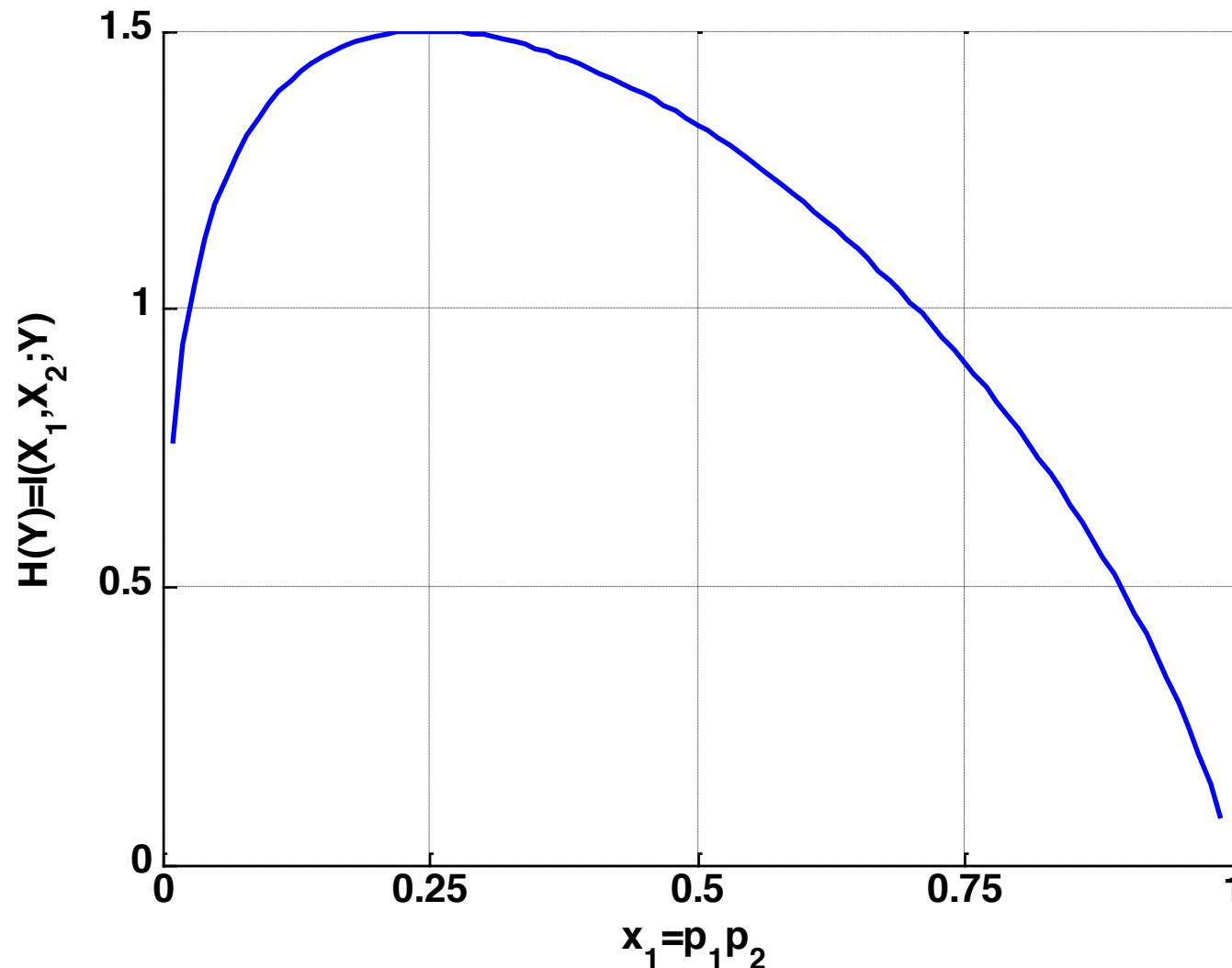
$$= 1 - p_1 - \frac{\color{red}{x_1}}{p_1} + \color{red}{x_1}$$

$$\frac{d\color{blue}{x_2}}{dp_1} = -1 + \frac{\color{red}{x_1}}{p_1^2}$$

Setting $\frac{d\color{blue}{x_2}}{dp_1} = 0 \Rightarrow p_1 = \sqrt{\color{red}{x_1}}$
 $\Rightarrow p_2 = \sqrt{\color{red}{x_1}}$

$$\color{green}{x_2^*} = \left(1 - \sqrt{\color{red}{x_1}} \right)^2$$

Maximum $H(Y)$ is 1.5 at $x_1=0.25$



Part 15C: Proof of Achievability of Multiple Access Capacity Region

Theorem 15.3.1

- The capacity of the two-user multiple-access channel is the closure of the convex hull of all rate pairs satisfying

$$R_1 < I(X_1; Y | X_2) = I_1$$

$$R_2 < I(X_2; Y | X_1) = I_2$$

$$R_1 + R_2 < I(X_1, X_2; Y) = I_3$$

Random Codebook Generation

- Fix $p(x_1, x_2) = p(x_1)p(x_2)$
- Codebook generation:
 - Generate 2^{nR_1} codewords X_1^n with each element i.i.d. $p(x_1)$.
 - Generate 2^{nR_2} codewords X_2^n with each element i.i.d. $p(x_2)$.

Typical Set Decoding

- Use typical set decoding as follows:
 - Decode Y^n as $W_1=i, W_2=j$ if a unique pair (i, j) exists such that
$$(X_1^n(i), X_2^n(j), Y^n) \in A_\epsilon^{(n)}(X_1^n, X_2^n, Y^n).$$
 - Otherwise declare an error.

Probability of error

- $P_e^{(n)} = \frac{1}{2^{n(R_1+R_2)}} \sum_{(W_1, W_2)} \Pr\{g(Y^n) \neq (W_1, W_2) \mid W_1, W_2 \text{ sent}\}$
- By symmetry, assume $W_1=1, W_2=1$

$$E_{ij} = \left\{ \left(X_1^n(i), X_2^n(j), Y^n \right) \in A_\epsilon^{(n)}(X_1, X_2, Y) \right\}$$

$$P_e = P\left(E_{11}^c \cup \bigcup_{i \neq 1, j=1} P(E_{i1}) \cup \bigcup_{i=1, j \neq 1} P(E_{1j}) \cup \bigcup_{i \neq 1, j \neq 1} P(E_{ij}) \right)$$

$$\leq P(E_{11}^c) + \sum_{i \neq 1, j=1} P(E_{i1}) + \sum_{i=1, j \neq 1} P(E_{1j}) + \sum_{i \neq 1, j \neq 1} P(E_{ij})$$

$$P(E_{i1})$$

$$P(E_{i1}) = P\left\{\left(X_1^n(i), X_2^n(1), Y^n\right) \in A_\epsilon^{(n)}\left(X_1, X_2, Y\right)\right\}$$

$$= \sum_{\left(X_1^n, X_2^n, Y\right) \in A_\epsilon^{(n)}} P\left(X_1^n(i)\right) P\left(X_2^n(1), Y\right)$$

$$\leq \sum_{\left(X_1^n, X_2^n, Y\right) \in A_\epsilon^{(n)}} 2^{-n(H(X_1)-\epsilon)} 2^{-n(H(X_2, Y)-\epsilon)}$$

$$= |A_\epsilon^{(n)}| 2^{-n(H(X_1)-\epsilon)} 2^{-n(H(X_2, Y)-\epsilon)}$$

$$= 2^{n(H(X_1, X_2, Y)+\epsilon)} 2^{-n(H(X_1)-\epsilon)} 2^{-n(H(X_2, Y)-\epsilon)}$$

$$P(E_{i1}), P(E_{1j}), P(E_{ij})$$

$$\begin{aligned} P(E_{i1}) &\leq 2^{n(H(X_1, X_2, Y) + \epsilon)} 2^{-n(H(X_1) - \epsilon)} 2^{-n(H(X_2, Y) - \epsilon)} \\ &\leq 2^{-n(H(X_1) + H(X_2, Y) - H(X_1, X_2, Y) - 3\epsilon)} \\ &= 2^{-n(I(X_1; Y|X_2) - 3\epsilon)} \end{aligned}$$

Similarly $P(E_{1j}) \leq 2^{-n(I(X_2; Y|X_1) - 3\epsilon)}$

and $P(E_{ij}) \leq 2^{-n(I(X_1, X_2; Y) - 4\epsilon)}$

$$\begin{aligned} H(X_1) + H(X_2, Y) - H(X_1, X_2, Y) &= I(X_1; X_2, Y) \\ &= I(\cancel{X_1}, \cancel{X_2})^{\nearrow 0} + I(X_1; Y | X_2) \\ &= I(X_1; Y | X_2) \end{aligned}$$

Probability of error conclusion

$$P_e = P\left(E_{11}^c \cup \bigcup_{i \neq 1, j=1} P(E_{i1}) \cup \bigcup_{i=1, j \neq 1} P(E_{1j}) \cup \bigcup_{i \neq 1, j \neq 1} P(E_{ij})\right)$$

$$\leq P(E_{11}^c) + \sum_{i \neq 1, j=1} P(E_{i1}) + \sum_{i=1, j \neq 1} P(E_{1j}) + \sum_{i \neq 1, j \neq 1} P(E_{ij})$$

$$\begin{aligned} &\leq \epsilon + 2^{nR_1} 2^{-n(I(X_1;Y|X_2)-3\epsilon)} \\ &\quad + 2^{nR_2} 2^{-n(I(X_2;Y|X_1)-3\epsilon)} \\ &\quad + 2^{n(R_1+R_2)} 2^{-n(I(X_1,X_2;Y)-4\epsilon)} \end{aligned}$$

$R_1 < I(X_1;Y X_2) = I_1$
$R_2 < I(X_2;Y X_1) = I_2$
$R_1 + R_2 < I(X_1, X_2; Y) = I_3$

$$= \epsilon + 2^{n(R_1-I(X_1;Y|X_2)+3\epsilon)} + 2^{n(R_2-I(X_2;Y|X_1)+3\epsilon)} + 2^{n(R_1+R_2-I(X_1,X_2;Y)+4\epsilon)}$$

Closure of the convex hull

- What about the closure of the convex hull?
- If $(R_1, R_2) \in C$ and $(R'_1, R'_2) \in C$, then
$$(\lambda R_1 + (1-\lambda)R'_1, \lambda R_2 + (1-\lambda)R'_2) \in C$$
for $0 \leq \lambda \leq 1$ by time-sharing.
- Thus C is a convex set.

Part 15D: Gaussian Multiple Access Channel

Gaussian MAC

$$Y = \sum_{i=1}^m X_i + Z$$

- Define $C\left(\frac{P}{N}\right) = \frac{1}{2} \log\left(1 + \frac{P}{N}\right)$

$$R_i < C\left(\frac{P_i}{N}\right)$$

$$R_i + R_j < C\left(\frac{P_i + P_j}{N}\right) = C\left(\frac{2P}{N}\right) \quad \text{if } P_i = P_j$$

⋮

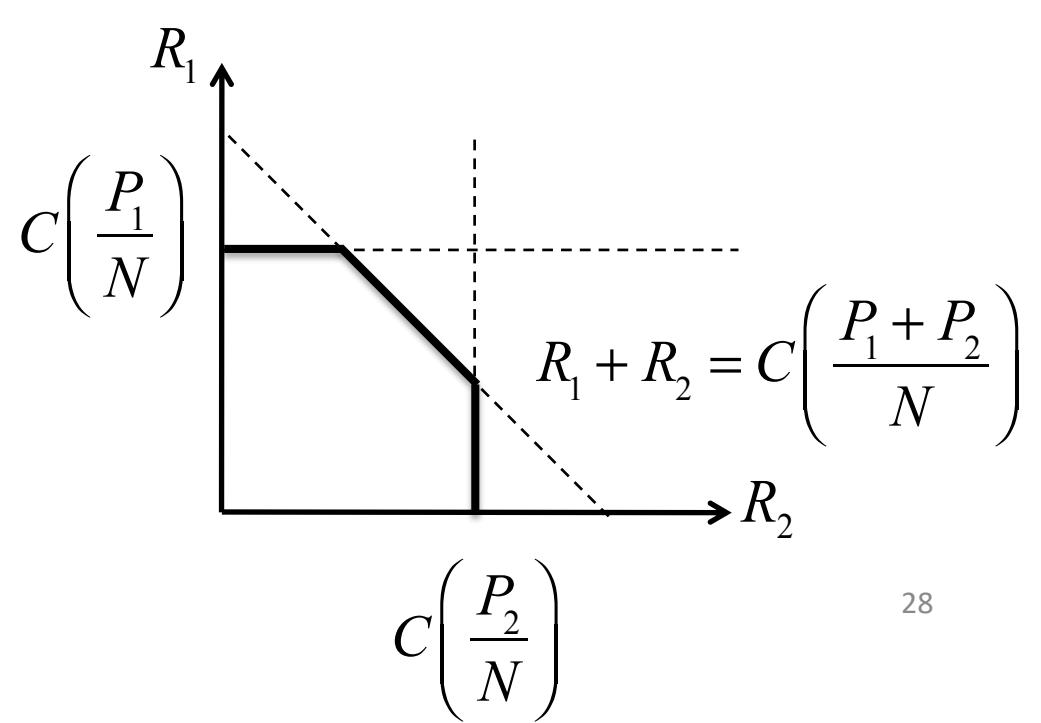
$$\sum_{i=1}^m R_i < C\left(\frac{mP}{N}\right) \quad \text{if } P_i = P_j$$

Two User Gaussian MAC

$$R_1 < C\left(\frac{P_1}{N}\right)$$

$$R_2 < C\left(\frac{P_2}{N}\right)$$

$$R_1 + R_2 < C\left(\frac{P_1 + P_2}{N}\right)$$



Limit of many users on Gaussian MAC

Note: when $R_i = R_j \quad \forall i, j$

$$C\left(\frac{mP}{N}\right) < mC\left(\frac{P}{N}\right)$$

- So the last inequality dominates and

$$C\left(\frac{mP}{N}\right) \rightarrow \infty \quad \text{as } m \rightarrow \infty$$

$$\frac{1}{m}C\left(\frac{mP}{N}\right) = \frac{1}{m} \frac{1}{2} \log\left(1 + \frac{mP}{N}\right) \approx \frac{\log_2\left(\frac{mP}{N}\right)}{2m} \rightarrow 0$$