Information Theory Lecture 10

- Network Information Theory (CT15); a focus on channel capacity results
 - The (two-user) multiple access channel (15.3)
 - The (two-user) broadcast channel (15.6)
 - The relay channel (15.7)
 - Some remarks on general multiterminal channels (15.10)

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Joint Typicality

- Extension of previous results to an arbitrary number of variables (most basic defs here, many additional results in CT)
- Notation
 - For any k-tuple $x_1^k = (x_1, x_2, \dots, x_k) \in \mathcal{X}_1 \times \mathcal{X}_2 \times \dots \times \mathcal{X}_k$ and subset of indices $S \subseteq \{1, 2, \dots, k\}$ let $x_S = (x_i)_{i \in S}$
 - Assume x_i ∈ Xⁿ_i, any i, and let x_S be a matrix with x_i as rows for i ∈ S. Let the |S|-tuple x_{S,j} be the jth column of x_S.
 - As in CT, $a_n \doteq 2^{n(c \pm \varepsilon)}$ means

$$\left|\frac{1}{n}\log a_n - c\right| < \varepsilon,$$

for all sufficiently large \boldsymbol{n}

• For random variables X_1^k with joint distribution $p(x_1^k)$: Generate \mathbf{X}_S via n independent copies of $\mathbf{x}_{S,j}$, $j = 1, \ldots, n$. Then,

$$\Pr(\mathbf{X}_{\mathcal{S}} = \mathbf{x}_{\mathcal{S}}) = \prod_{j=1}^{n} p(\mathbf{x}_{\mathcal{S},j}) \triangleq p(\mathbf{x}_{\mathcal{S}})$$

• For
$$S \subseteq \{1, 2, ..., k\}$$
, define the set of ε -typical *n*-sequences
$$A_{\varepsilon}^{(n)}(S) = \left\{ \mathbf{x}_{S} : \Pr(\mathbf{X}_{S'} = \mathbf{x}_{S'}) \doteq 2^{-n[H(\mathbf{X}_{S'}) \pm \varepsilon]}, \ \forall S' \subseteq S \right\}$$

• Then, for any $\varepsilon > 0$, sufficiently large n, and $\mathcal{S} \subseteq \{1, \ldots, k\}$,

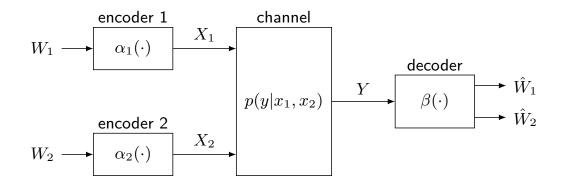
$$P(A_{\varepsilon}^{(n)}(\mathcal{S})) \ge 1 - \varepsilon$$

$$p(\mathbf{x}_{\mathcal{S}}) \doteq 2^{-n[H(\mathbf{X}_{\mathcal{S}})\pm\varepsilon]} \quad \text{if} \quad \mathbf{x}_{\mathcal{S}} \in A_{\varepsilon}^{(n)}(\mathcal{S})$$

$$|A_{\varepsilon}^{(n)}(\mathcal{S})| \doteq 2^{n[H(\mathbf{X}_{\mathcal{S}})\pm2\varepsilon]}$$

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The Multiple Access Channel



 Two "users" communicating over a common channel. (The generalization to more than two is straightforward.)

Coding:

• Memoryless pmf (or pdf):

 $p(y|x_1, x_2), x_1 \in \mathcal{X}_1, x_2 \in \mathcal{X}_2, y \in \mathcal{Y}$

- Data: $W_1 \in \mathcal{I}_1 = \{1, \dots, M_1\}$ and $W_2 \in \mathcal{I}_2 = \{1, \dots, M_2\}$
 - Assume W_1 and W_2 uniformly distributed and independent
- Encoders: $\alpha_1: \mathcal{I}_1 \to \mathcal{X}_1^n$ and $\alpha_2: \mathcal{I}_2 \to \mathcal{X}_2^n$
- Rates: $R_1 = \frac{1}{n} \log M_1$ and $R_2 = \frac{1}{n} \log M_2$
- Decoder: $\beta: \mathcal{Y}^n \to \mathcal{I}_1 \times \mathcal{I}_2$, $\beta(Y^n) = (\hat{W}_1, \hat{W}_2)$
- Error probability: $P_e^{(n)} = \Pr\left((\hat{W}_1, \hat{W}_2) \neq (W_1, W_2)\right)$

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Capacity:

We have two (or more) rates, R_1 and R_2

- \implies cannot consider one maximum achievable rate
- \implies study sets of achievable rate-pairs (R_1, R_2)
- \implies trade-off between R_1 and R_2
 - Achievable rate-pair: (R_1, R_2) is achievable if $(\alpha_1, \alpha_2, \beta)$ exists such that $P_e^{(n)} \to 0$ as $n \to \infty$
 - Capacity region:

The closure of the set of all achievable rate-pairs (R_1, R_2)

Capacity Region for the Multiple Access Channel

• Fix $\pi(x_1, x_2) = p_1(x_1)p_2(x_2)$ on \mathcal{X}_1 and \mathcal{X}_2 . Draw $\{X_1^n(i) : i \in \mathcal{I}_1\}$ and $\{X_2^n(j) : j \in \mathcal{I}_2\}$ in an i.i.d. manner according to p_1 and p_2 .

• Symmetry of codebook generation \implies

$$P_e^{(n)} = \Pr\left((\hat{W}_1, \hat{W}_2) \neq (W_1, W_2)\right)$$

= $\Pr\left\{(\hat{W}_1, \hat{W}_2) \neq (1, 1) | (W_1, W_2) = (1, 1)\right\}$

where the second "Pr" is with respect to the channel and the random codebook design.

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• Also

$$\Pr((\hat{W}_1, \hat{W}_2) \neq (1, 1)) = \Pr(\hat{W}_1 \neq 1, \hat{W}_2 \neq 1) + \Pr(\hat{W}_1 \neq 1, \hat{W}_2 = 1) + \Pr(\hat{W}_1 = 1, \hat{W}_2 \neq 1) = P_{12}^{(n)} + P_1^{(n)} + P_2^{(n)}$$

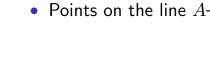
conditioned that $(W_1, W_2) = (1, 1)$ everywhere.

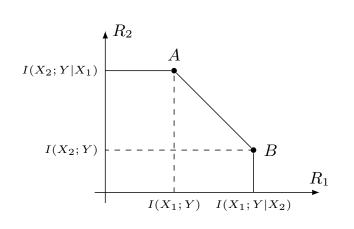
• Joint typicality decoding, declare $(\hat{W}_1, \hat{W}_2) = (1, 1)$ if

$$(X_1^n(i), X_2^n(j), Y^n) \in \mathcal{A}_{\varepsilon}^{(n)}$$

only for $i = j = 1 \Rightarrow$

$$P_{12}^{(n)} \leq 2^{n[R_1 + R_2 - I(X_1, X_2; Y) + 4\varepsilon]}$$
$$P_1^{(n)} \leq 2^{n[R_1 - I(X_1; Y | X_2) + 3\varepsilon]}$$
$$P_2^{(n)} \leq 2^{n[R_2 - I(X_2; Y | X_1) + 3\varepsilon]}$$





• Hence, for a fixed $\pi(x_1, x_2) = p_1(x_1)p_2(x_2)$ the capacity region contains at least all pairs (R_1, R_2) in the set Π defined by

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

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

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

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• The corner points

• Consider the point 'A'

$$R_{1} = I(X_{1}; Y)$$

$$R_{2} = I(X_{2}; Y|X_{1})$$

$$R_{1} + R_{2} = I(X_{1}, X_{2}; Y)$$

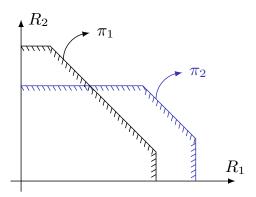
- User 1 ignores the presence of user 2 \Rightarrow $R_1 = I(X_1; Y)$
- Decode user 1's codeword \Rightarrow User 2 sees an equivalent channel with input X_2^n and output $(Y^n, X_1^n) \Rightarrow$

$$R_{2} = I(X_{2}; Y, X_{1})$$

= $I(X_{2}; Y|X_{1}) + I(X_{1}; X_{2})$
= $I(X_{2}; Y|X_{1})$

- The above can be repeated with $1 \leftrightarrow 2$ and $A \leftrightarrow B$
- Points on the line A-B can be achieved by time sharing

 Each particular choice of distribution π gives an achievable region Π; for two different π's,



Fixed π ⇒ Π is convex.
 Varying π ⇒ Π can be non-convex.
 However all rates on a line connecting two achievable rate-pairs are achievable by time-sharing.

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• The capacity region for the multiple access channel is the closure of the convex hull of the set of points defined by the three inequalities

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

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

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

over all possible product distributions $p_1(x_1)p_2(x_2)$ for (X_1, X_2) .

• **Proof**: Achievability proof based on jointly typical sequences (as shown before) and a "time-sharing variable".

Converse proof based on Fano's inequality and the independence of X_1^n and X_2^n (since they are functions of independent messages).

Example: A Gaussian Channel

• Bandlimited AWGN channel with two additive users

$$Y(t) = X_1(t) + X_2(t) + Z(t).$$

The noise Z(t) is zero-mean Gaussian with power spectral density $N_0/2$, and $X_1(t)$ and $X_2(t)$ are subject to the power constraints P_1 and P_2 , respectively. The available bandwidth is W.

• The capacity of the corresponding single-user channel (with power constraint P) is

$$W \cdot \mathsf{C}\left(\frac{P}{WN_0}\right) \quad \text{[bits/second]}$$

where

$$\mathsf{C}(x) = \log(1+x).$$

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• Time-Division Multiple-Access (TDMA):

Let user 1 use all of the bandwidth with power P_1/α a fraction $\alpha \in [0, 1]$ of time, and let user 2 use all the bandwidth with power $P_2/(1 - \alpha)$ the remaining fraction $1 - \alpha$ of time. The achievable rates then are

$$R_1 < W \cdot \alpha \operatorname{C}\left(\frac{P_1/\alpha}{WN_0}\right) \qquad R_2 < W \cdot (1-\alpha) \operatorname{C}\left(\frac{P_2/(1-\alpha)}{WN_0}\right)$$

• Frequency-Division Multiple-Access (FDMA):

Let user 1 transmit with power P_1 using a fraction α of the available bandwidth W, and let user two transmit with power P_2 the remaining fraction $(1 - \alpha)W$. The achievable rates are

$$R_1 < \alpha W \cdot \mathsf{C}\left(\frac{P_1}{\alpha W N_0}\right) \qquad R_2 < (1-\alpha)W \cdot \mathsf{C}\left(\frac{P_1}{(1-\alpha)W N_0}\right)$$

• TDMA and FDMA are equivalent from a capacity perspective!

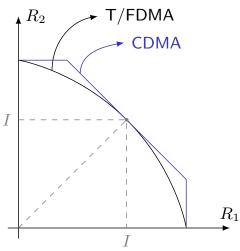
• Code-Division Multiple-Access (CDMA):

Defined, in our context, as all schemes that can be implemented to achieve the rates in the true capacity region

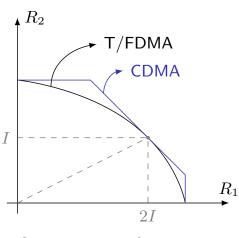
$$R_{1} \leq W \cdot \mathsf{C}\left(\frac{P_{1}}{WN_{0}}\right) = W \log\left(1 + \frac{P_{1}}{WN_{0}}\right)$$
$$R_{2} \leq W \cdot \mathsf{C}\left(\frac{P_{2}}{WN_{0}}\right) = W \log\left(1 + \frac{P_{2}}{WN_{0}}\right)$$
$$R_{1} + R_{2} \leq W \cdot \mathsf{C}\left(\frac{P_{1} + P_{2}}{WN_{0}}\right) = W \log\left(1 + \frac{P_{1} + P_{2}}{WN_{0}}\right)$$

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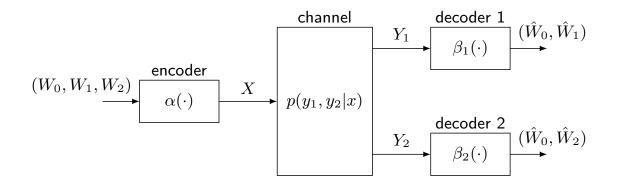
Capacity region for $P_1 = P_2$



Capacity region for $P_1 = 2P_2$

Note that T/FDMA is only optimal when $\frac{\alpha}{1-\alpha} = \frac{P_1}{P_2}$.

The Broadcast Channel

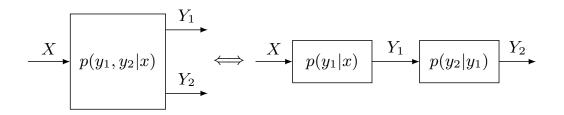


- One transmitter, several receivers
- Message W_0 is a public message for both receivers, whereas W_1 and W_2 are private messages

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The Degraded Broadcast Channel

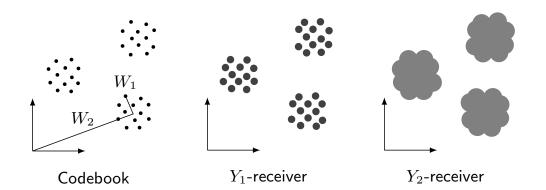


• A broadcast channel is degraded if it can be split as in the figure. That is, Y_2 is a "noisier" version of X than Y_1 ,

$$p(y_1, y_2|x) = p(y_2|y_1)p(y_1|x).$$

• The Gaussian and the binary symmetric broadcast channels are degraded (see the examples in CT).

Superposition Coding for the Degraded Broadcast Channel



- Assume there is no common information (for simplicity). Let W_2 chose a cloud of possible W_1 -codewords.
- The Y₁-receiver sees all codewords, whereas the Y₂-receiver is only able to distinguish between clouds.

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• The capacity region of the degraded broadcast channel (with no common information), is the closure of the convex hull of all rates satisfying

$$R_2 < I(U; Y_2)$$

$$R_1 < I(X; Y_1|U)$$

for some distribution $p_1(u)p_2(x|u)$.

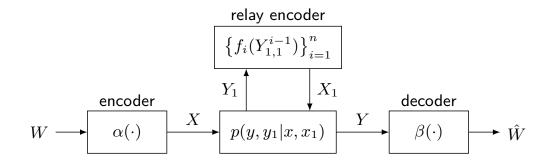
• **Proof**: Choose W_2 -codewords i.i.d. according to $p_1(u)$, and for each one, choose W_1 -codewords i.i.d. according to $p_2(x|u)$. The overall channel from W_2 to Y_2 (the clouds) can be made error-free as long as $R_2 < I(U; Y_2)$, and conditioned on W_2 the channel from W_1 to Y_1 can be made error-free as long as $R_1 < I(X; Y_1|U)$.

Converse proved in Problem 15.11 (based on Fano, as usual).

- The capacity region of the degraded broadcast channel (with common information): If the pair (R₁ = a, R₂ = b) is achievable for independent messages, as before, the triple (R₀, R₁ = a, R₂ = b - R₀) is achievable with common information at rate R₀ (as long as R₀ < b).
- Since the better receiver can decode both W_1 and W_2 , part of the W_2 -message can be made to include common information!

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The Relay Channel



- One sender, one receiver, and one intermediate node
- The problem does not define the set of relay functions
 {
 f_i(·)}ⁿ_{i=1}. The relay's strategy might be to decode the
 message, or compress its channel observation, or amplify it
 and retransmit it, or ...

- Capacity is not known in general. Some known bounds:
 - Cut-set upper bound: The relay is assumed to be co-located with the transmitter or with the receiver.

$$R \le \max_{p(x,x_1)} \min \{ I(X, X_1; Y), \ I(X; Y, Y_1 | X_1) \}$$

• Decode-and-forward lower bound:

$$R \le \max_{p(x,x_1)} \min \{ I(X,X_1;Y), \ I(X;Y_1|X_1) \}$$

Proof: Split transmission in *b* blocks.

Choose $2^{n\tilde{R}}$ codewords i.i.d. $\sim p(x_1)$, and for each one, choose 2^{nR} codewords i.i.d. $\sim p(x|x_1)$ and distribute them in $2^{n\tilde{R}}$ bins.

The relay can decode the message if $R < I(X; Y_1|X_1)$ and then it sends the bin index in the next block. The receiver can decode the bin index if $\tilde{R} < I(X_1; Y)$ and, knowing this index, it can decode the message from the previous block if $R - \tilde{R} < I(X; Y|X_1)$.

• These bounds coincide if the relay channel is degraded:

$$p(y, y_1|x, x_1) = p(y_1|x, x_1)p(y|y_1, x_1)$$

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General Multiterminal Systems

- M different nodes, each transmitting X_m and receiving Y_m . The message from node i to node j is $W_{i,j}$ with rate $R_{i,j}$. The channel between nodes is $p(y_1, \ldots, y_M | x_1, \ldots, x_M)$.
- Although significant progress in recent years, still only a few general results are known.
- One of them is El Gamal's cut-set bound: If the rates $\{R_{i,j}\}$ are achievable there exists a $p(x_1, \ldots, x_m)$ such that

$$\sum_{i \in \mathcal{S}, j \in \mathcal{S}^c} R_{i,j} < I(X^{(\mathcal{S})}; Y^{(\mathcal{S}^c)} | X^{(\mathcal{S}^c)})$$

for all $\mathcal{S} \subset \{1, \dots, M\}$.

• The source-channel separation principle:

For general multiterminal networks the source and channel codes cannot be designed separately without loss. Essentially the source code needs to know the channel to provide optimal *dependencies* between channel input variables.

• Feedback:

Feedback can increase the capacity of a multi-terminal channel, since it can help transmitters to "cooperate" to reduce interference.