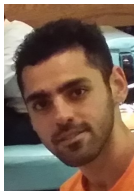
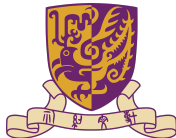


Sub-optimality of superposition coding for three or more receivers

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CISS 2018
23 Mar, 2018

The **story** of superposition coding is the **story** of broadcast channels with **degradation** (receivers or message sets)



The **story** of superposition coding is the **story** of broadcast channels with **degradation** (receivers or message sets)

It is also the **story** of auxiliary random variables



In the beginning

- ▶ Cover (1972) proposed the superposition coding achievable region for degraded broadcast channels
 - He used an **auxiliary** variable to represent the message for the weaker of two receivers.
- ▶ Bergmans (1973, Gaussian) and Gallager (1974, discrete-memoryless) established the optimality of superposition coding for the degraded broadcast channel
 - Gallager's proof of optimality of superposition coding naturally extended to a sequence of degraded receivers



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The optimality of superposition coding region was then established for

- ▶ Weaker notions of **weaker** receiver in a two-receiver broadcast channel
 - Less Noisy (Korner-Marton 75), More Capable (El Gamal 79)
- ▶ Degraded message sets (Korner-Marton 77)
 - Comparison of the sizes of the images of a set via two channels



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A (seemingly) natural "Implication": Auxiliaries (in superposition coding) **captured** the **coarser** message.



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The optimality of superposition coding was established for

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However for the following three-receiver broadcast channel setting:

- ▶ Receivers Y_2, Y_3 wish to decode message M_0
- ▶ Receiver Y_1 wishes to decode messages M_0, M_1

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Demonstrates that auxiliaries **do not** capture the coarser message.

- ▶ Associating an auxiliary with information decoded by groups of receivers improved the achievable region
 - U_{123}, U_{12}, U_{13} , and $U_1 = X$.
 - The achievable region was no longer a superposition coding region, it also involved the **other** (old) idea: **random binning**

A (seemingly) natural "Implication" (Take 2): Auxiliaries (~~in superposition coding~~) **captured** the information decoded by groups of receivers.

- ▶ Binning and Superposition both present



Consistency

The revised intuition about auxiliaries is **consistent** with Marton's achievable scheme for two-receiver broadcast channels with private messages

- ▶ The region employs three auxiliaries: U_1, U_2, U_{12}
- ▶ It is known that this region is strictly better than the one with only U_1, U_2 (even for private messages).



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Open Question

Is Marton's achievable scheme for two-receiver broadcast channels optimal?

Three-or-more receivers

A three-receiver broadcast channel with private messages would have seven auxiliaries

$$U_{123}, U_{12}, U_{13}, U_{23}, U_1, U_2, U_3$$

and a natural extension of Marton's scheme would have two layers of superposition coding and binning between random variables in each layer.

A succinct clean representation of the rate constraints is not available



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However: this region is not optimal (Padlakanda and Pradhan (2015))



Still one fundamental setting remained

Consider the setting:

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Interpretation of auxiliaries: U_{123} and $U_{12} = X$, and only superposition coding

Question: Is superposition coding optimal?

(Note: The first layer superposition coding of the previous private message setting)



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- **Open problems:**

8.1. What is the capacity region of the general 3-receiver DM-BC with one common message to all three receivers and one private message to one receiver?

8.2. Is superposition coding optimal for the general 3-receiver DM-BC with one message to all three receivers and another message to two receivers?

8.3. What is the sum-capacity of the binary skew-symmetric broadcast channel?

8.4. Is Marton's inner bound tight in general?



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3.1 SOME BASIC MATHEMATICAL PROBLEMS OF MULTIUSER SHANNON THEORY

I. Csiszár

Mathematical Institute of the
Hungarian Academy of Sciences
Budapest, Hungary

2. Image Size Characterization Problem.

The η -image size $g_W(A, \eta)$ of a set $A \subset X^n$ over a discrete memoryless channel (DMC) $(W: X \rightarrow Y)$ is the minimum cardinality of $B \subset Y^n$ such that $W^n(B|x) \geq \eta$ for each $x \in A$. The problem is to find, for a distribution P on X and DMCs $\{W_i: X \rightarrow Y_i\}$, $i = 1, \dots, k$, a single-letter characterization of the limit of the sets of all $(k+1)$ -dimensional vectors

$$\left[\frac{1}{n} \log |A|, \frac{1}{n} \log g_{W_1}(A, \eta), \dots, \frac{1}{n} \log g_{W_k}(A, \eta) \right].$$

Here $A \subset X^n$ is any set of P -typical sequences, and $0 < \eta < 1$ is fixed (the result is independent of η).

Csiszar's open problem is very closely tied to finding the capacity region



Körner had proposed a region (1984) for the image size characterization over three channels

Theorem: For every RV's T , U , and V such that

$$TUV \rightarrow S \rightarrow XYZ,$$

nonnegative numbers t , t' , and t'' , the point (r_x, r_y, r_z) with coordinates

$$\begin{aligned} r_x &\triangleq \min [H(X), H(X|T) + t, H(X|TU) + t', \\ &\quad H(X|TUV) + t''], \\ r_y &\triangleq \min [H(Y), H(Y|T) + t, H(Y|TU) + t', \\ &\quad H(Y|TUV) + t''], \\ r_z &\triangleq \min [H(Z), H(Z|T) + t, H(Z|TU) + t', \\ &\quad H(Z|TUV) + t''] \end{aligned} \tag{30}$$

is an element of $\mathcal{H}(X; Y; Z|S)$. \square



Suboptimality of superposition coding

Superposition coding is sub-optimal for the setting (N-Yazdanpanah 2017)

- ▶ Constructed a channel whose 2-letter superposition-coding region was larger than the 1-letter one



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Note: The same (counter)-example showed that Korner's region is a proper subset of $\mathcal{H}(X; Y; Z|S)$.



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Remarks

- ▶ It took us three years to get counterexamples
 - The optimization problems involved are non-convex
 - In small dimensions counter-examples lie in a set of very small "size" (random sampling does not work)
- ▶ Question: Why did we believe that superposition coding was sub-optimal?
- ▶ More generally, why do we believe certain regions are optimal while certain others are not?



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- ▶ More generally, why do we believe certain regions are optimal while certain others are not?
 - An unpublished conjecture: Local-tensorization **implies** global tensorization



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This talk: Focus on the (counter)-example

- ▶ Bounds on the capacity region
- ▶ Shed light to properties of good codes for this channel



Multilevel product broadcast erasure channel (ISIT '17)

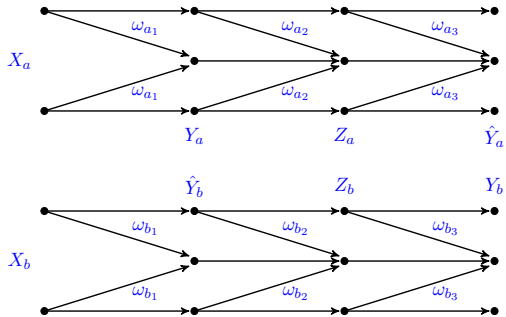
$$X_a \rightarrow Y_a : BEC(e_a), X_b \rightarrow Y_b : BEC(e_b)$$

$$X_a \rightarrow \hat{Y}_a : BEC(\hat{e}_a), X_b \rightarrow \hat{Y}_b : BEC(\hat{e}_b)$$

$$X_a \rightarrow Z_a : BEC(f_a), X_b \rightarrow Z_b : BEC(f_b)$$

$$\hat{e}_a \geq f_a \geq e_a \quad \& \quad e_b \geq f_b \geq \hat{e}_b$$

$$C_Z = (1 - f_a) + (1 - f_b)$$



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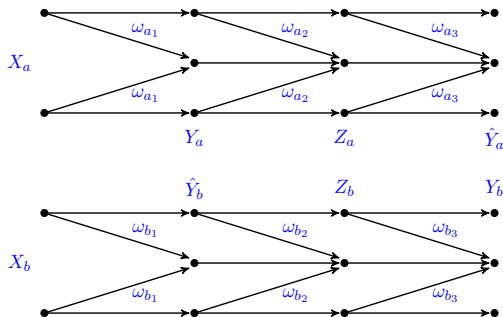
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$$X_a \rightarrow Z_a : BEC(f_a), X_b \rightarrow Z_b : BEC(f_b)$$

$$\hat{e}_a \geq f_a \geq e_a \quad \& \quad e_b \geq f_b \geq \hat{e}_b$$

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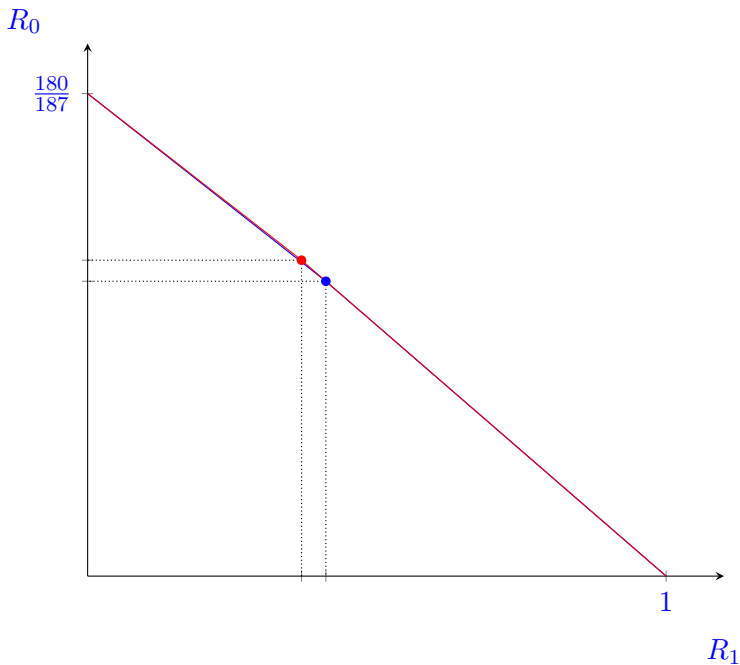
Theorem

For	$e_a = 1/2$	$\hat{e}_a = 1$	$f_a = 17/22$
	$e_b = 1/2$	$\hat{e}_b = 0$	$f_b = 9/34$

1-letter SC : $R_0 + R_1 \leq 1$ and $\frac{11}{10}R_0 + R_1 \leq \frac{18}{17} = \frac{11}{10}C_Z$

2-letter SC : $R_0 + R_1 \leq 1$ and $\frac{484}{435}R_0 + R_1 \leq \frac{528}{493} = \frac{484}{435}C_Z$

Plot

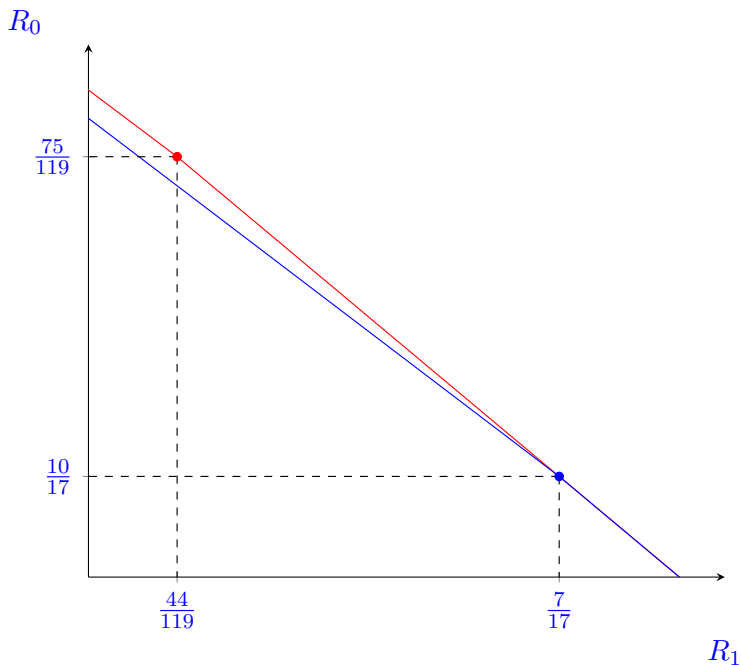


2-letter & 1-letter

SC regions



Plot



2-letter & 1-letter

SC regions



1-letter SC

The distribution that achieves the **corner-point**.

- ▶ Let U be a ternary random variable

$$\begin{cases} P(U = 0) = \frac{13}{34} & (X_a, X_b)|\{U = 0\} = (0, 0) \\ P(U = 1) = \frac{7}{34} & (X_a, X_b)|\{U = 1\} = (M, 0) \\ P(U = 2) = \frac{14}{34} & (X_a, X_b)|\{U = 2\} = (M, M) \end{cases}$$

where M is an unbiased binary random variable

- ▶ Let Q be the random variable that symmetrizes the distribution of X
- ▶ Let $\tilde{U} = (U, Q)$ and substitute (\tilde{U}, X) into SC

$$\begin{aligned} R_0 &\leq I(\tilde{U}; Z) = \frac{10}{17} \\ R_0 + R_1 &\leq I(\tilde{U}; Z) + I(X; Y|\tilde{U}) = 1 & R_0 + R_1 &\leq I(X; Y) = 1 \\ R_0 + R_1 &\leq I(\tilde{U}; Z) + I(X; \hat{Y}|\tilde{U}) = 1 & R_0 + R_1 &\leq I(X; \hat{Y}) = 1 \end{aligned}$$



2-letter SC

The distribution that achieves the **corner-point**.

$$\begin{array}{l} M_1 \text{ \& } M_2 \\ \text{two independent} \\ \text{unbiased binary r.v.} \end{array} \left\{ \begin{array}{l} P(U = 0) = \frac{20}{119} \quad (X_{a1}, X_{b1}, X_{a2}, X_{b2})|\{U = 0\} = (0, 0, 0, 0) \\ P(U = 1) = \frac{11}{119} \quad (X_{a1}, X_{b1}, X_{a2}, X_{b2})|\{U = 1\} = (M_1, 0, M_2, 0) \\ P(U = 2) = \frac{88}{119} \quad (X_{a1}, X_{b1}, X_{a2}, X_{b2})|\{U = 2\} = (M_1, M_1, M_1, 0) \end{array} \right.$$

- ▶ Let Q be the random variable that symmetrizes the distribution of X
- ▶ Let $\tilde{U} = (U, Q)$ and substitute (\tilde{U}, X) into $SC \Rightarrow (R_0, R_1) = (\frac{75}{119}, \frac{44}{119})$



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- ▶ Let Q be the random variable that symmetrizes the distribution of X
- ▶ Let $\tilde{U} = (U, Q)$ and substitute (\tilde{U}, X) into $SC \Rightarrow (R_0, R_1) = (\frac{75}{119}, \frac{44}{119})$

Observation

- ▶ A **linear** code achieves the 2-letter region

A natural question

Let $\mathbf{M} = (M_1, \dots, M_m)$ be mutually independent unbiased bits.

Let $X_a^n = \mathbf{A}\mathbf{M}$ and $X_b^n = \mathbf{B}\mathbf{M}$, where \mathbf{A}, \mathbf{B} are $n \times m$ matrices.

What is the rate region achieved by such a linear coding scheme.

(variables are $m, \mathbf{A}, \mathbf{B}$).



Routine Idea: Intersection of the two capacity regions (ignoring one of the users)

Theorem

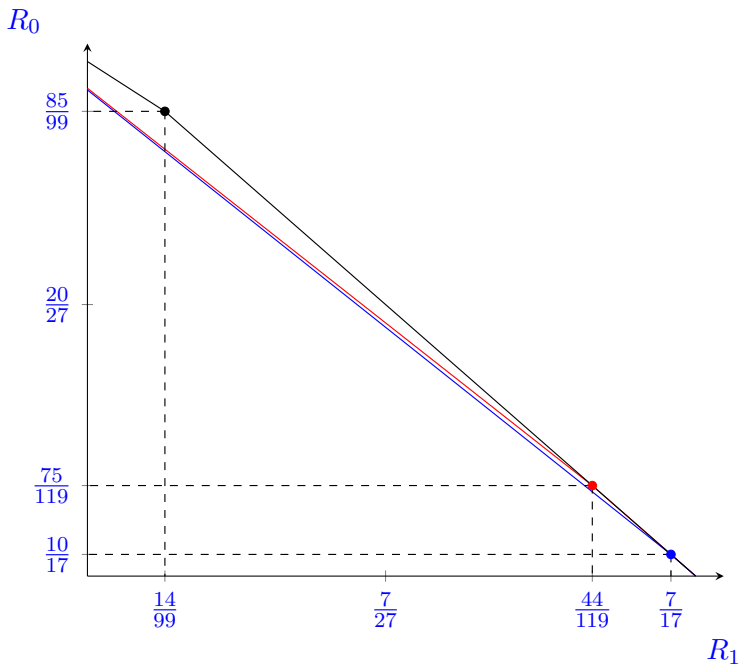
$$\begin{array}{llll} \text{For} & e_a = 1/2 & \hat{e}_a = 1 & f_a = 17/22 \\ & e_b = 1/2 & \hat{e}_b = 0 & f_b = 9/34 \end{array}$$

$$\text{ignoring } \hat{Y} : \quad R_0 + R_1 \leq 1 \quad \text{and} \quad \frac{11}{5}R_0 + R_1 \leq \frac{11}{5}C_Z$$

$$\text{ignoring } Y : \quad R_0 + R_1 \leq 1 \quad \text{and} \quad \frac{34}{25}R_0 + R_1 \leq \frac{34}{25}C_Z$$



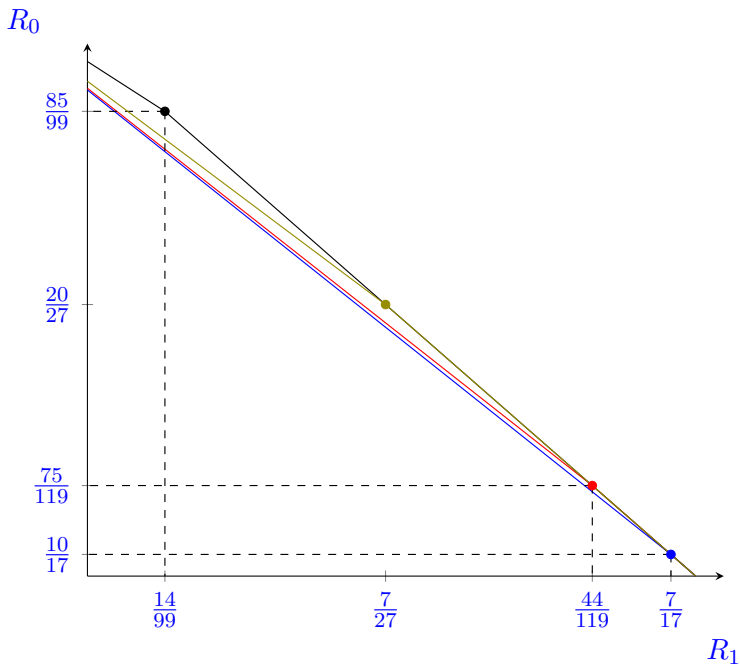
Plot



2-letter & 1-letter SC
Trad. Outer Bound



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Trad. Outer Bound
New Outer Bound



Idea for new outer-bound

From limiting n -letter **inner bound** that goes to capacity:



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From limiting n -letter **inner bound** that goes to capacity:

Theorem (Concentration of mutual information over memoryless product erasure channel)

Consider a product erasure channel, $W_a(y_a|x_a) \otimes W_b(y_b|x_b)$, mapping X_a, X_b to Y_a, Y_b with erasure probabilities ϵ_a, ϵ_b , respectively. Then

$$I(X_a^n, X_b^n; Y_a^n, Y_b^n) = \mathcal{H}(\lfloor n(1 - \epsilon_a) \rfloor, \lfloor n(1 - \epsilon_b) \rfloor) + O\left(\sqrt{n \log n}\right),$$

where

$$\mathcal{H}_n(k, l) = \frac{1}{\binom{n}{k} \binom{n}{l}} \sum_{S, T \subseteq [n]: |S|=k, |T|=l} H(X_{aS}, X_{bT}).$$



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Using (essentially) sub-modularity of entropy, we can establish that

$$\limsup_n \max_{p(x_a^n, x_b^n)} \frac{1}{n} \left(\frac{85}{160} \mathcal{H}\left(\frac{n}{2}, \frac{n}{2}\right) + \frac{75}{160} \mathcal{H}(0, n) - \frac{187}{160} \mathcal{H}\left(\frac{5n}{22}, \frac{25n}{34}\right) \right) \leq 0.$$



Outer bound continued

Theorem (Outer bound)

Any achievable rate pair (R_0, R_1) must satisfy the constraints.

$$R_0 + R_1 \leq 1 \quad \text{and} \quad \frac{187}{160}R_0 + R_1 \leq \frac{18}{16}.$$



Outer bound continued

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Achievability

If there is a non-trivial collection (X_a^n, X_b^n) such that

$$\begin{aligned}\mathcal{H}_n\left(\frac{n}{2}, \frac{n}{2}\right) &= \mathcal{H}_n\left(\frac{n}{2}, \frac{25n}{34}\right) + o(n), \\ \frac{5}{11}\mathcal{H}_n\left(\frac{n}{2}, \frac{25n}{34}\right) + \frac{6}{11}\mathcal{H}_n\left(0, \frac{25n}{34}\right) &= \mathcal{H}\left(\frac{5n}{22}, \frac{25n}{34}\right) + o(n), \\ \frac{8}{17}\mathcal{H}_n(0, n) + \frac{9}{17}\mathcal{H}_n\left(0, \frac{n}{2}\right) &= \mathcal{H}_n\left(0, \frac{25n}{34}\right) + o(n), \\ \frac{17}{25}\mathcal{H}_n\left(0, \frac{25n}{34}\right) &= \mathcal{H}_n\left(0, \frac{n}{2}\right) + o(n),\end{aligned}$$

then there are non-trivial points of the outer bound that are achievable.



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Suggests: "MDS-like" (linear increase followed by flat region) code-construction for

X_a^n, X_b^n



Observations

- ▶ Sub-optimality of superposition coding region
- ▶ Sub-optimality of Korner's image-size characterization
- ▶ Linear code achieves 2-letter inner bound
- ▶ A new (explicit) outer bound from limiting n -letter inner bound
- ▶ Outer bound yields insights into structure of good codes

