Information inequalities in network information theoretic settings

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NETWORK INFORMATION THEORY

Information theoretic study of multi-user communication scenarios

- Extension of point-to-point communication theory [Shannon '48]
- Started with Shannon's two-way channel ['62]
 - Model of a telephone conversation
 - A very hard model (unfortunately)
 - feedback
 - interference
 - message cognition
- Other simpler channels were proposed and studied
 - Multiple access channel [Shannon '61, van-der-Muelen '71]
 - Broadcast channel [Cover '72]
 - Interference channel [Ahlswede '74]
 - Relay channel [van-der-Muelen '71]

• Most fundamental problems remain open

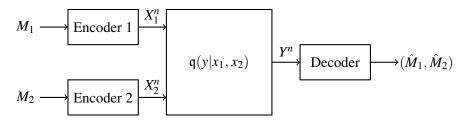


Figure : Discrete memoryless multiple access channel (MAC)

Goal: Compute *Capacity Region* or set of achievable rates?

One of the few settings where the capacity region is established

The capacity region is the set of rate pairs (R_1, R_2) such that they satisfy

$$R_{1} \leq I(X_{1}; Y|X_{2}, Q)$$

$$R_{2} \leq I(X_{2}; Y|X_{1}, Q)$$

$$R_{1} + R_{2} \leq I(X_{1}, X_{2}; Y|Q)$$

for some pmf $p(q)p(x_1|q)p(x_2|q)$, with $|Q| \leq 2$.

Remarks

- Q: convexification random variable (time-sharing)
- The region corresponds to a general cut-set outer bound

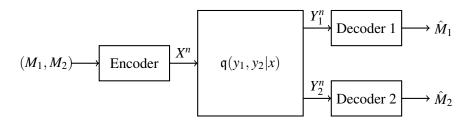


Figure : Discrete memoryless broadcast channel

Goal: Compute Capacity Region or set of achievable rates?

Capacity region: open; established in several classes of channels (including additive Gaussian noise model)

BEST KNOWN ACHIEVABLE REGION (M-IB) [MARTON '79]

The union of rate pairs (R_1, R_2) satisfying

$$R_{1} \leq I(U, W; Y_{1})$$

$$R_{2} \leq I(V, W; Y_{2})$$

$$R_{1} + R_{2} \leq \min\{I(W; Y_{1}), I(W; Y_{2})\} + I(U; Y_{1}|W)$$

$$+ I(V; Y_{2}|W) - I(U; V|W)$$

over all $(U, V, W) \rightarrow X \rightarrow (Y_1, Y_2)$ is achievable.

Remarks

- U,V,W: *auxiliary* random variables
- Suffices: $|W| \le |X| + 4$, $|U| \le |X|$, $|V| \le |X|$ [Gohari-Anantharam '11]
- New: $|W| \le |X| + 4$, $|U| + |V| \le |X| + 1$ [Gohari-Nair-Anantharam '13]
- Not known if it is optimal or not
- Factorization conjecture [Gohari-Nair-Anantharam '12]: If true, the above region is capacity

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The union of rate pairs (R_1, R_2) satisfying

$$\begin{aligned} R_1 &\leq I(U; Y_1) \\ R_2 &\leq I(V; Y_2) \\ R_1 + R_2 &\leq \min\{I(U; Y_1) + I(X; Y_2|U), I(V; Y_2) + I(X; Y_1|V)\} \end{aligned}$$

over all $(U, V) \rightarrow X \rightarrow (Y_1, Y_2)$ forms an outer bound.

Remarks

- Suffices: $|U| \le |X| + 1, |V| \le |X| + 1$
- Strictly improves on Körner-Marton outer bound [Nair-El Gamal '07]
- Differs from M-IB [Nair-Wang '08, Gohari-Anantharam '11]
- Known to be strictly suboptimal [Geng-Gohari-Nair-Yu '11]

INTERFERENCE CHANNEL (AHLSWEDE 1974)

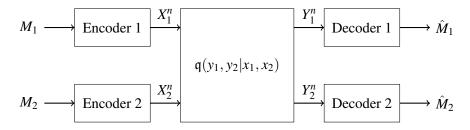


Figure : Discrete memoryless interference channel

Goal: Compute *Capacity Region* or set of achievable rates? *Capacity region*: open problem; known for some (limited) classes Not known even in the *scalar Gaussian* case

• Known in strong-interference regime [Sato '78, Han-Kobayashi '81]

BEST INNER BOUND (HK-IB) [HAN-KOBAYASHI '81]

A rate pair (R_1, R_2) is achievable if

$$\begin{split} R_1 &\leq I(X_1; Y_1 | V, Q) \\ R_2 &\leq I(X_2; Y_2 | U, Q) \\ R_1 + R_2 &\leq I(X_1, V; Y_1 | Q) + I(X_2; Y_2 | U, V, Q) \\ R_1 + R_2 &\leq I(X_2, U; Y_2 | Q) + I(X_1; Y_1 | U, V, Q) \\ R_1 + R_2 &\leq I(X_1, V; Y_1 | U, Q) + I(X_2, U; Y_2 | V, Q) \\ 2R_1 + R_2 &\leq I(X_1, V; Y_1 | Q) + I(X_1; Y_1 | U, V, Q) + I(X_2, U; Y_2 | V, Q) \\ R_1 + 2R_2 &\leq I(X_2, U; Y_2 | Q) + I(X_2; Y_2 | U, V, Q) + I(X_1, V; Y_1 | U, Q) \end{split}$$

for some pmf $p(q)p(u_1, x_1|q)p(v, x_2|q)$, where $|U_1| \le |X_1| + 4$, $|V| \le |X_2| + 4$, and $|Q| \le 6$.

Remarks

- Q: coded time-division (more than convexification)
- Not known if it is optimal or not
- Above representation is due to [Chong-Motani-Garg-El Gamal '08]

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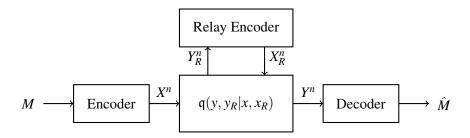


Figure : Point-to-point system with relay

Remarks

- Capacity is unknown; very few special cases known
- Cut set outer bound: $C \leq \max_{p(x_1, x_2)} \min\{I(X_1, X_2; Y_3), I(X_1; Y_2, Y_3 | X_2)\}$
- Cut set outer bound is strictly sub-optimal [Aleksic,Razagi,Yu '09]

COMMENTS ON THE PROBLEMS

One approach [Avestimehr-Diggavi-Tse, ...]

- These problems have been open for long time
- May be simple expressions for capacity regions do not exist
- Why not settle for appromations

A counter argument [this talk]

- Before we give up, at the very least, we need to
 - Determine if Marton's inner bound is optimal/sub-optimal for the broadcast channel
 - Determine if Han-Kobayashi inner bound is optimal/sub-optimal for the interference channel

Information inequalities play a central role to answer these two questions

- Inequalities to prove optimality of inner bounds (*factorization inequalities*)
- Inequalities to compute extreme points of inner bounds (*extremal inequalities*)

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ON OPTIMALITY OF INNER BOUNDS

Converses to coding theorems

Illustrative example: Degraded broadcast channel

$$X \longrightarrow \qquad \mathfrak{q}(y_1|x) \xrightarrow{Y_1} \mathfrak{q}(y_2|y_1) \longrightarrow Y_2$$

Figure : Degraded broadcast channel

The capacity region is the set of rate pairs (R_1, R_2) such that they satisfy

$$R_1 \le I(X_1; Y_1 | U)$$

$$R_2 \le I(U; Y_2)$$

for some pmf p(u)p(x|u), with $|U| \le |X|$. [Cover '72, Gallager '74]

GALLAGER'S CONVERSE

Given any sequence of codebooks with $P_e^{(n)} \rightarrow 0$, observe that

$$nR_{2} = H(M_{2})$$
 (message is uniformly distributed)

$$= I(M_{2}; Y_{2}^{n}) + H(M_{2}|Y^{n})$$

$$\leq I(M_{2}; Y_{2}^{n}) + nR_{2}P_{e}^{(n)}$$
 (Fano's inequality)

$$= \sum_{i=1}^{n} I(M_{2}; Y_{2i}|Y_{21}^{i-1}) + nR_{2}P_{e}^{(n)}$$
 (chain rule)

$$\leq \sum_{i=1}^{n} I(M_{2}, Y_{21}^{i-1}; Y_{2i}) + nR_{2}P_{e}^{(n)}$$

$$= \sum_{i=1}^{n} I(U_{i}; Y_{2i}) + nR_{2}P_{e}^{(n)}$$
 where $U_{i} := (M_{2}, Y_{21}^{i-1})$

$$= n \left(I(U_{Q}; Y_{2}|Q) + R_{2}P_{e}^{(n)} \right)$$
 where $Q \sim$ uniform $\{1, \dots, n\}$

$$\leq n \left(I(U^{(n)}; Y_{2}) + R_{2}P_{e}^{(n)} \right)$$
 where $U^{(n)} := (Q, U_{Q})$

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GALLAGER'S CONVERSE -CTD

Similarly

$$nR_{1} = H(M_{1}) = I(M_{1}; Y_{1}^{n} | M_{2}) + H(M_{1} | M_{2}, Y_{1}^{n})$$

$$\leq \sum_{i=1}^{n} I(M_{1}; Y_{1i} | M_{2}, Y_{11}^{i-1}) + nR_{1}P_{e}^{(n)} \qquad \text{(Fano's inequality)}$$

$$\leq \sum_{i=1}^{n} I(X_{i}; Y_{1i} | M_{2}, Y_{11}^{i-1}) + nR_{1}P_{e}^{(n)} \qquad (M_{1}, Y_{11}^{i-1}) \to X_{i} \to Y_{1i}$$

$$= \sum_{i=1}^{n} I(X_{i}; Y_{1i} | M_{2}, Y_{11}^{i-1}, Y_{21}^{i-1}) + nR_{1}P_{e}^{(n)} \qquad Y_{21}^{i-1} \to Y_{11}^{i-1} \to (M_{2}, X_{i})$$

$$\leq \sum_{i=1}^{n} I(X_{i}; Y_{1i} | M_{2}, Y_{21}^{i-1}) + nR_{1}P_{e}^{(n)} \qquad (M_{1}, Y_{11}^{i-1}, Y_{21}^{i-1}) \to X_{i} \to Y_{1i}$$

$$= \sum_{i=1}^{n} I(X_{i}; Y_{1i} | M_{2}, Y_{21}^{i-1}) + nR_{1}P_{e}^{(n)} \qquad (M_{1}, Y_{11}^{i-1}, Y_{21}^{i-1}) \to X_{i} \to Y_{1i}$$

CONTINUED..

n

$$nR_{1} = \sum_{i=1}^{n} I(X_{i}; Y_{1i}|U_{i}) + nR_{1}P_{e}^{(n)}$$

= $n\Big(I(X; Y_{1}|Q, U_{Q}) + R_{1}P_{e}^{(n)}\Big)$ Recall $Q \sim$ uniform $\{1, \dots, n\}$
= $n\Big(I(X; Y_{1}|U^{(n)}) + R_{2}P_{e}^{(n)}\Big)$

By Caratheodory's theorem there is $(\tilde{U}^{(n)}, X)$ with $|\tilde{U}^{(n)}| \leq |X|$ such that

$$I(X; Y_1|U^{(n)}) = I(X; Y_1|\tilde{U}^{(n)})$$
 and $I(U^{(n)}; Y_2) = I(\tilde{U}^{(n)}; Y_2).$

Letting $n \to \infty$, we see that $\exists (U, X)$ such that

$$R_2 \le I(U; Y_2)$$
$$R_1 \le I(X; Y_1|U) \quad \Box$$

COMMENTS ON GALLAGERS PROOF AND MORE GENERALLY

- Absolutely brilliant argument
 - Explicitly constructs an auxiliary U
 - Every converse and outer bound known today follows this approach
- However
 - It is doing more than it should
 - Another proof may exhibit existence of auxiliary U without identifying it
 - The auxiliary *U* constructed in the converse has no relation to the *U actually employed* in the coding phase
- Maybe
 - The problem is in the use of auxiliaries in the representation
 - Seek other representations
 - These might lead to different converse arguments
- This leads us to information inequalities [this talk]
 - Factorization inequalities prove optimality directly
 - Extremal inequality To simplify inner bounds

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REVISIT THE CAPACITY REGION FOR DEGRADED BROADCAST CHANNEL

The capacity region is the set of rate pairs (R_1, R_2) such that they satisfy

$$R_1 \le I(X_1; Y_1 | U)$$

$$R_2 \le I(U; Y_2)$$

for some pmf p(u)p(x|u), with $|U| \le |X|$. [Cover '72, Gallager '74]

Seek: $C_{\lambda} := \max_{(R_1, R_2) \in C} R_1 + \lambda R_2$, for $\lambda \ge 1$ (supporting hyperplanes) *Elementary manipulations*

$$C_{\lambda} = \max_{p(u,x)} \lambda I(U; Y_{2}) + I(X; Y_{1}|U)$$
 (corner point)
$$= \max_{p(u,x)} \lambda I(X; Y_{2}) + I(X; Y_{1}|U) - \lambda I(X; Y_{2}|U)$$
 $U \to X \to Y_{2}$
$$= \max_{p(x)} \left(\lambda I(X; Y_{2}) + \max_{p(u|x)} (I(X; Y_{1}|U) - \lambda I(X; Y_{2}|U)) \right)$$

$$= \max_{p(x)} \left(\lambda I(X; Y_{2}) + \mathfrak{C}[I(X; Y_{1}) - \lambda I(X; Y_{2})] \right)$$

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UPPER CONCAVE ENVELOPES

Given a function f(x), its upper concave envelope is

$$\mathfrak{C}[f] := \inf\{g(x) : g(x) \ge f(x), g(x) \text{ is concave}\}.$$

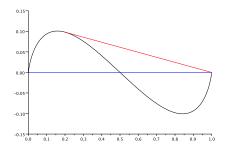


Figure : Illustration of concave envelope

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Inequalities in Network Info. Theory - EII

AN OBSERVATION

Fix a broadcast channel $q(y_1, y_2|x)$.

Can treat $I(X; Y_1) - \lambda I(X; Y_2)$ as a function of p(x).

For $\lambda \geq 1$ define

$$T_{\lambda}(X) := \mathfrak{C}[I(X;Y_1) - \lambda I(X;Y_2)]$$

It is easy to see that

$$\max_{p(u|x)} I(X;Y_1|U) - \lambda I(X;Y_2|U) = \mathfrak{C}[I(X;Y_1) - \lambda I(X;Y_2)]$$

Define $C_{\lambda}(\mathfrak{q}) = \max_{p(x)} \left(\lambda I(X; Y_2) + T_{\lambda}(X) \right)$

Capacity region of degraded broadcast channel can be expressed as

$$igcap_{\lambda\geq 1}\{(R_1,R_2)\subset \mathbb{R}^2_+: R_1+\lambda R_2\leq C_\lambda(\mathfrak{q})\}$$

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$$\bigcap_{\lambda \ge 1} \{ (R_1, R_2) \subset \mathbb{R}^2_+ : R_1 + \lambda R_2 \le C_\lambda(\mathfrak{q}) \}$$

A CONVERSATION

Alice: Hey, we have a new characterization for the superposition coding region without auxiliaries

$$igcap_{\lambda\geq 1}\{(R_1,R_2)\subset \mathbb{R}^2_+: R_1+\lambda R_2\leq C_\lambda(\mathfrak{q})\}$$

Bob: But is it really different? Isn't your concave envelope hiding the auxiliaries

Alice: But the only auxliaries that show up are the ones that are "extremal". i.e. the only interesting ones are the ones that help compute the upper concave envelope

Bob: So what? Does this buy you anything: Can we get new results? Can we simplify old proofs?

Alice: Yes, when we apply this to other settings. Focusing on extremal auxiliaries has yielded new results. Greatly simplifies old proofs.

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DEGRADED BINARY SYMMETRIC BROADCAST CHANNEL

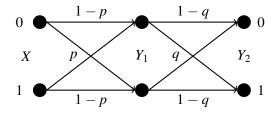


Figure : Degraded BSC broadcast channel

To compute the union of rate pairs (R_1, R_2) such that they satisfy

 $R_1 \le I(X_1; Y_1 | U)$ $R_2 \le I(U; Y_2)$

for some pmf p(u)p(x|u), with $|U| \le |X|$, it suffices to consider $U \mapsto X \sim BSC(s)$ [Cover '72, Wyner-Ziv '73]

Proof is non-trivial; uses Mrs. Gerber's lemma (convexity of $h(p * h^{-1}(x))$)

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USING CONCAVE ENVELOPES

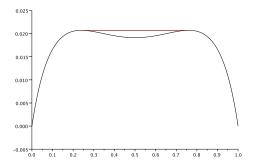


Figure : Illustration of $I(X; Y_1) - \lambda I(X; Y_2)$

Immediate that a global maximum exists when $P(X = 0) = \frac{1}{2}$ and P(X = 0|U = 0) = s, P(X = 0|U = 1) = 1 - s, i.e, $U \mapsto X \sim BSC(s)$

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Inequalities in Network Info. Theory - EII

OPTIMALITY USING CONCAVE ENVELOPE REPRESENTATION

Consider two degraded broadcast channels $q_1(y_{11}, y_{21}|x_1)$ and $q_2(y_{12}, y_{22}|x_2)$ Form a product broadcast channel: $q_1(y_{11}, y_{21}|x_1) \otimes q_2(y_{12}, y_{22}|x_2)$

$$T_{\lambda}(X_1, X_2) := \mathfrak{C}[I(X_1, X_2; Y_{11}, Y_{12}) - \lambda I(X_1; X_2; Y_{21}, Y_{22})]$$

 $T_{\lambda}(X_1, X_2)$: function of $p(x_1, x_2)$

Claim: If the following factorization inequality holds

$$T_{\lambda}(X_1, X_2) \le T_{\lambda}(X_1) + T_{\lambda}(X_2)$$

then one has optimality of the region

$$igcap_{\lambda\geq 1}\{(R_1,R_2)\subset \mathbb{R}^2_+: R_1+\lambda R_2\leq C_\lambda(\mathfrak{q})\},$$

where $C_{\lambda}(\mathfrak{q}) = \max_{p(x)} \left(\lambda I(X; Y_2) + T_{\lambda}(X) \right)$

REMARKS

- The above claim is much stronger than what is needed for proving optimality
- Note that I(X; Y) has a similar behaviour, i.e. for $q_1(y_1|x_1) \otimes q_2(y_2|x_2)$

$$I(X_1, X_2; Y_1, Y_2)$$

= $I(X_1, X_2; Y_1) + I(X_1, X_2; Y_2|Y_1)$
= $I(X_1; Y_1) + I(X_2; Y_2|Y_1)$ $(Y_2 \to X_2 \to X_1 \to Y_1)$
 $\leq I(X_1; Y_1) + I(X_2; Y_2)$

Proof of Claim: If the factorization inequality holds

$$C_{\lambda}(\mathfrak{q} \otimes \cdots \otimes \mathfrak{q}) = \max_{p(x^{n})} \lambda I(X_{1}^{n}; Y_{21}^{n}) + T_{\lambda}(X^{n})$$

$$\leq \max_{p(x^{n})} \sum_{i=1}^{n} (\lambda I(X_{i}; Y_{2i}) + T_{\lambda}(X_{i})) \qquad \text{(By assumption)}$$

$$\leq n \max_{p(x)} \lambda I(X; Y_{2}) + T_{\lambda}(X) = nC_{\lambda}(\mathfrak{q})$$

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PROOF - CTD

To complete the argument, observe

$$nR_{1} + n\lambda R_{2}$$

$$\leq \lambda I(M_{1}, M_{2}; Y_{21}^{n}) + I(M_{1}; Y_{11}^{n} | M_{2}) - \lambda I(M_{1}; Y_{21}^{n} | M_{2}) + n\epsilon_{n} \quad \text{(Fano)}$$

$$\stackrel{(a)}{\leq} \lambda I(X^{n}; Y_{21}^{n}) + \mathfrak{C}[I(X^{n}; Y_{11}^{n}) - \lambda I(X_{1}^{n}; Y_{21}^{n})]$$

$$\leq \max_{p(x^{n})} \lambda I(X_{1}^{n}; Y_{21}^{n}) + T_{\lambda}(X^{n})$$

where the (a) follows from $(M_1, M_2) \rightarrow X^n \rightarrow (Y_{11}^n, Y_{21}^n)$

Important: The proof I demonstrated is a generic proof

- If one can demonstrate an appropriate *factorization inequality* then
 - Optimality follows directly from Fano's inequality
 - Optimality of the *n*-letter form (known in many cases)
- † If a weaker *factorization inequality* does not hold then
 - Inner bound is strictly sub-optimal

PROOF - CTD

To complete the argument, observe

$$nR_{1} + n\lambda R_{2}$$

$$\leq \lambda I(M_{1}, M_{2}; Y_{21}^{n}) + I(M_{1}; Y_{11}^{n} | M_{2}) - \lambda I(M_{1}; Y_{21}^{n} | M_{2}) + n\epsilon_{n} \quad \text{(Fano)}$$

$$\stackrel{(a)}{\leq} \lambda I(X^{n}; Y_{21}^{n}) + \mathfrak{C}[I(X^{n}; Y_{11}^{n}) - \lambda I(X_{1}^{n}; Y_{21}^{n})]$$

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where the (a) follows from $(M_1, M_2) \rightarrow X^n \rightarrow (Y_{11}^n, Y_{21}^n)$

Important: The proof I demonstrated is a generic proof

- If one can demonstrate an appropriate *factorization inequality* then
 - Optimality follows directly from Fano's inequality
 - Optimality of the *n*-letter form (known in many cases)
- † If a weaker *factorization inequality* does not hold then
 - Inner bound is strictly sub-optimal

FACTORIZATION INEQUALITIES: INTRODUCTION

Inequalities that factor over product channels

Triivial Example:

• $I(X_1, X_2; Y_1, Y_2) \le I(X_1; Y_1) + I(X_2; Y_2)$ when $Y_1 \to X_1 \to X_2 \to Y_2$ is Markov.

Non-trivial example:

For any product broadcast channel $q_1(y_{11}, y_{21}|x_1) \otimes q_2(y_{12}, y_{22}|x_2)$ and $\lambda \ge 1$

 $T_{\lambda}(X_1, X_2) \leq T_{\lambda}(X_1) + T_{\lambda}(X_2),$

where $T_{\lambda}(X_1, X_2) = \mathfrak{C}[I(X_1, X_2; Y_{11}, Y_{12}) - \lambda I(X_1, X_2; Y_{21}, Y_{22})]$

- Stronger than what I needed earlier (no degradedness assumption)
- Implies the (known) capacity region for *degraded message sets* (two receivers)

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OTHER FACTORIZATION INEQUALITIES

Define for a broadcast channel

 $S(X) := \mathfrak{C}[I(X;Y_1) - I(X;Y_2) + \mathfrak{C}[I(X;Y_2) - I(X;Y_1)]]$

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• This yields the secrecy capacity (known result)

For an interference channel, † define

 $R(X_1; X_2) := \mathfrak{C} \big[I(X_1; Y_1 | X_2) - I(X_1; Y_2 | X_2) \big]$

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A FACTORIZATION INEQUALITY - CONJECTURE

For a three-receiver broadcast channel $q(y_1, y_2, y_3 | x), \mu \in [0, 1], \lambda \ge 1$

$$T_{\mu,\lambda}(X) := \mathfrak{C}\big[\mu I(X;Y_1) + (1-\mu)I(X;Y_2) - \lambda I(X;Y_3)\big]$$

For any product broadcast channel $q_1(y_{11}, y_{21}, y_{31}|x_1) \otimes q_2(y_{12}, y_{22}, y_{32}|x_2)$

$$T_{\mu,\lambda}(X_1, X_2) \le T_{\mu,\lambda}(X_1) + T_{\mu,\lambda}(X_2)$$
 (Conjecture)

- Know this holds when $\mu \in \{0, 1\}$
- If true, would imply the capacity region of three receiver broadcast channel with two degraded message sets
 - Problem has been open since the 70s
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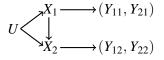
PROOF OF A FACTORIZATION INEQUALITY

Claim: For any product broadcast channel $q_1(y_{11}, y_{21}|x_1) \otimes q_2(y_{12}, y_{22}|x_2)$ and $\lambda \ge 1$

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where $T_{\lambda}(X_1, X_2) = \mathfrak{C}[I(X_1, X_2; Y_{11}, Y_{12}) - \lambda I(X_1, X_2; Y_{21}, Y_{22})]$

Note: This is equivalent to showing that for any U such that



there exists $U_1 \rightarrow X_1 \rightarrow (Y_{11}, Y_{21})$ and $U_2 \rightarrow X_2 \rightarrow (Y_{12}, Y_{22})$ such that

$$I(X_1, X_2; Y_{11}, Y_{12}|U) - \lambda I(X_1, X_2; Y_{21}, Y_{22}|U)$$

$$\leq I(X_1; Y_{11}|U_1) - \lambda I(X_1; Y_{21}|U_1) + I(X_1; Y_{12}|U_2) - \lambda I(X_1; Y_{22}|U_2)$$

PROOF CTD..

$$\begin{split} I(X_1, X_2; Y_{11}, Y_{12}|U) &- \lambda I(X_1, X_2; Y_{21}, Y_{22}|U) \\ &= I(X_1; Y_{11}|U) - \lambda I(X_1; Y_{21}|U, Y_{22}) & \text{using the Markov} \\ &+ I(X_2; Y_{12}|U, Y_{11}) - \lambda I(X_2; Y_{22}|U) & \text{structure} \\ &= I(X_1; Y_{11}|U, Y_{22}) - \lambda I(X_1; Y_{21}|U, Y_{22}) & \text{subtract/add term} \\ &+ I(X_2; Y_{12}|U, Y_{11}) - \lambda I(X_2; Y_{22}|U, Y_{11}) & I(Y_{11}; Y_{22}|U) \\ &= I(X_1; Y_{11}|U_1) - \lambda I(X_1; Y_{21}|U_1) & U_2 := (U, Y_{22}) \\ &+ I(X_2; Y_{12}|U_2) - \lambda I(X_2; Y_{22}|U_1) & U_1 := (U, Y_{11}) \end{split}$$

Note that $(U, Y_{22}) \to X_1 \to (Y_{11}, Y_{21})$ and $(U, Y_{11}) \to X_2 \to (Y_{12}, Y_{22})$ hold as desired

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REMARKS ABOUT FACTORIZATION INEQUALITIES

- Most proofs of factorization inequalities are motivated by converses/outer bounds
 - Though factorization implies optimality, not the other way
- One recent converse motivated by factorization
 - Capacity region of reversely semi-deterministic broadcast channel [Geng-Gohari-Nair-Yu '11]
 - Show strictly sub-optimality of outer bound for broadcast channels

For most problems where one has good inner bounds one can conjecture a factorization inequality

• Easily check (numerically) if these inequalities hold for small cardinalities

Need: A method for proving these factorization inequalities, an understanding of them, and a more formal and precise statement

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EXTREMAL INEQUALITIES: INTRODUCTION

Recall: Factorization inequalities are inequalities that factor over product channels

Extremal inequalities: Inequalities that compute the *extremal* auxiliary random variables

Usual example: Entropy power inequality (EPI) and its variants

• More on this later

Another example: Mrs. Gerber's Lemma ("discrete analogue" of EPI)

Begin with: A couple of discrete extremal inequalities

• Motivated by Marton's inner bound for broadcast channels

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MARTON'S INNER BOUND - BROADCAST CHANNEL

The union of rate pairs (R_1, R_2) satisfying

 $R_{1} \leq I(U, W; Y_{1})$ $R_{2} \leq I(V, W; Y_{2})$ $R_{1} + R_{2} \leq \min\{I(W; Y_{1}), I(W; Y_{2})\} + I(U; Y_{1}|W)$ $+I(V; Y_{2}|W) - I(U; V|W)$

over all $(U, V, W) \rightarrow X \rightarrow (Y_1, Y_2)$ is achievable.

 $T(X) := \max_{p(uv|x)} I(U; Y_1) + I(V; Y_2) - I(U; V)$

Conjecture [Gohari-Nair-Anantharam'12]: For any $\lambda \in [0, 1]$ the following function factorizes

$$\mathfrak{C}[-\lambda H(Y_1) - (1-\lambda)H(Y_2) + T(X)]$$

If true, then M-IB would be sum-rate optimal

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AN INFORMATION INEQUALITY

Recall
$$T(X) := \max_{p(uv|x)} I(U; Y_1) + I(V; Y_2) - I(U; V)$$

A 5-variable inequality [Geng-Jog-Nair-Wang '11]

For any $(U, V) \rightarrow X \rightarrow (Y_1, Y_2)$ and |X| = 2, the following holds:

 $I(U;Y_1) + I(V;Y_2) - I(U;V) \le \max\{I(X;Y_1), I(X;Y_2)\}$

In other words when |X| = 2

$$T(X) = \max\{I(X; Y_1), I(X; Y_2)\}\$$

- The inequality is false when |X| = 3
- Not quite in the framework of Shannon/non-Shannon type inequalities
 - The cardinality constraint (natural under a channel coding setting) destroys the convex cone property

Remarks about this inequality

- Conjectured for a particular binary input BC [Nair-Wang '08]
 - Motivation: to exhibit gap between inner and outer bounds for this channel
- Used perturbation analysis and obtained cardinality bounds for M-IB [Gohari-Anantharam '12]
 - They numerically verified the plausibility of this conjecture
- The conjecture was established for the channel [Jog-Nair '09] extending the perturbation techniques
- The proof was generalized for all binary input broadcast channels [Geng-Nair-Wang '10]

What about beyond sum-rate?

Is it true that when $(U, V) \rightarrow X \rightarrow (Y_1, Y_2)$, |X| = 2, and $\alpha > 1$

 $\alpha I(U;Y) + I(V;Z) - I(U;V) \le \max\{\alpha I(X;Y), I(X;Z)\}$

False (counterexample to this inequality due to Geng)

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BEYOND SUM-RATE

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- Indeed suffices to consider $|U| + |V| \le |X| + 1$ to compute

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ANOTHER FACTORIZATION CONJECTURE

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For $\lambda \in [0, 1], \alpha \geq 1$, define

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Conjecture [Gohari-Nair-Anantharam '12]: The functional $S_{\alpha,\lambda}(X)$ factorizes over product broadcast channels

Remarks

- If true, would imply that M-IB is the capacity region for a DM-BC (very huge deal)
- Know it is true when $\lambda \in \{0, 1\}$
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ENTROPY POWER INEQUALITY (EPI)

Perhaps one of the most famous information inequalities EPI: If *X*, *Y* are independent and have densities

$$2^{2h(X+Y)} \ge 2^{2h(X)} + 2^{2h(Y)}$$

Conditional EPI: If $X \rightarrow U \rightarrow Y$ is Markov, and conditioned on U (finite valued) they have densities, then

$$2^{2h(X+Y|U)} \ge 2^{2h(X|U)} + 2^{2h(Y|U)}$$

This is an extremal inequality

- Its utility has been to evaluate extremal auxiliaries
- Several variations of this inequality known and used
- Key to several converses in Gaussian noise settings (until recently)
- Proof uses perturbation ideas [Stam '58]

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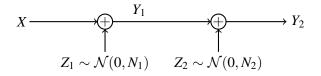
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AN ILLUSTRATION OF EPI'S USE

Consider the Gaussian degraded broadcast channel



Power constraint: $E(X^2) \leq P$.

The capacity region is the set of rate pairs (R_1, R_2) such that they satisfy

 $R_1 \le I(X_1; Y_1 | U)$ $R_2 \le I(U; Y_2)$

for some (U, X) with $E(X^2) \leq P$

How would you compute this region? (real-valued variables)

BERGMAN'S PROOF (1973)

Let

$$h(Y_1|U) = \frac{1}{2}\log(2\pi e(N_1 + aP))$$

for some $a \in [0, 1]$

By EPI

$$2^{2h(Y_2|U)} = 2^{2h(Y_1+Z_2|U)} \ge 2^{2h(Y_1|U)} + 2^{2h(Z_2|U)} = 2\pi e(N_1 + aP + N_2)$$

Thus if

$$R_1 \le I(X_1; Y_1|U) = H(Y_1|U) - h(Y_1|X) = \frac{1}{2}\log\left(1 + \frac{aP}{N_1}\right)$$

then

$$R_2 \leq I(U; Y_2) \leq \frac{1}{2} \log \left(1 + \frac{aP}{N_1 + N_2} \right)$$

Equality:
$$X = U + V, U \perp V, V \sim \mathcal{N}(0, aP), U \sim \mathcal{N}(0, (1 - a)P)$$

C. Nair (CUHK)

BEYOND THE SCOPE OF THIS TALK

Factorization inequalities used to replace EPI in converses [Geng-Nair '12]

- Recover known converse proofs without using EPI (and in a much simpler way)
- In addition solved an *important* open problem of 2-receiver vector Gaussian broadcast channel with private and common messages

Basic idea: Use factorization inequalities to deduce that if X is a maximizer to a relevant optimization problem, X_1, X_2 i.i.d. $\sim X$ then

• Both
$$\frac{X_1+X_2}{\sqrt{2}}$$
 and $\frac{X_1-X_2}{\sqrt{2}}$ are also maximizers

• Further
$$\frac{X_1+X_2}{\sqrt{2}}$$
 and $\frac{X_1-X_2}{\sqrt{2}}$ are independent

Complete the argument (optimality of Gaussian) either using

- Central limit theorem
- Gaussian characterization: If *X*, *Y* are independent and *X* + *Y*, *X Y* are independent then *X*, *Y* are i.i.d. Gaussians [Berstein '40, Skijtovic '54]

CONCLUDING REMARKS

- New representations using concave envelopes
 - Greatly simplifies some existing proofs
- Optimality using factorization inequalities
 - As a new tool to study optimality of inner bounds
- Extremal inequalities to compute extremal auxiliaries
 - Cardinality constrained inequalities are richer and messier, but essential
- Mentioned some open problems and conjectures
- Few more exciting ideas left out from this talk

WANTED: A new (\dagger) proof of the factorization of

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Thank You