

Renyi Generalizations of  
Conditional Mutual Information

11 APR 2014

arXiv:1403.6102

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Given is a tripartite quantum state  $\rho_{ABC}$ , w/ marginals denoted by  $\rho_A, \rho_{AB}$ , etc.

The conditional mutual information of  $A$  &  $B$  from the perspective of  $C$  is

$$I(A; B|C)_\rho = H(AC)_\rho + H(BC)_\rho - H(C)_\rho - H(ABC)_\rho$$

where  $H(F)_\sigma = -\text{Tr} \{ \sigma_F \log \sigma_F \}$

for a state  $\sigma$  on system  $F$ .

This quantity shows up everywhere in quantum information theory:

strong subadditivity, squashed entanglement, discord, etc.

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Goal of this talk: Find a compelling Renyi generalization of this quantity.

Why? 1) Renyi entropies generalize von Neumann ones and give us a refined understanding of quantum information theory (examples: min- & max-entropy, smoothed versions, strong converses, error exponents, etc.)

2) A Renyi generalization might give us a way of attacking important open questions concerning  $I(A; B|C)$  (they are fundamental questions but also have application in condensed matter phys.) & QIT

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~~Before~~ Before finding a Rényi generalization, we should determine some relevant properties that it should obey. So we turn to  $I(A; B|C)$  for this.

1) Non-negativity  
 $I(A; B|C)_p \geq 0 \quad \forall P_{ABC}$

This is the statement of strong subadditivity (Lieb & Ruskai 1973) used in nearly every coding theorem in QIT & is one reason why squashed entanglement is a sensible correlation measure.

2) Data processing under local ops on A+B

$$I(A_1 A_2; B_1 | C) \geq I(A_1; B_1 | C)$$

Why is this true? just write it out

$$\begin{aligned} \Leftrightarrow H(A_1 A_2 C) + H(B_1 | C) - H(C) \\ - H(A_1 A_2 B_1 | C) &\geq H(A_1 C) + H(B_1 | C) \\ &\quad - H(C) - H(A_1 B_1 | C) \end{aligned}$$

④

$$\Leftrightarrow H(A, A_2 | C) - H(A, A_2 | B, C) \geq H(A, C) - H(A, B, C)$$

$$\Leftrightarrow H(A, A_2 | C) + H(A, B, C) - H(A, C) - H(A, A_2 | B, C) \geq 0$$

$$\Leftrightarrow I(A_2; B_1 | A, C) \geq 0 \quad \text{by SSA}$$

symmetric argument implies

$$I(A_1; B_1, B_2 | C) \geq I(A_1; B_1 | C)$$

so we get

$$I(A_1, A_2; B_1, B_2 | C) \geq I(A_1; B_1 | C)$$

### 3) Duality for 4-party pure states

Given pure  $\Psi_{ABCD}$ , we have that  $I(A; B | C)_\Psi = I(A; B | D)_\Psi$

So if we make a Renyi generalization of CMI, we would like for it to obey properties like the above ones.

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1st bad attempt:

Let  $I'_2(A;B|c)_p =$

$$H_2(AC)_p + H_2(BC)_p - H_2(c)_p - H_2(ABC)_p$$

$$\text{where } H_2(F)_\alpha = \frac{1}{1-\alpha} \log \text{Tr} \{ \sigma_F^\alpha \}$$

$$\text{for } \alpha \in [0, \infty]$$

Observe that  $\alpha \rightarrow 1$  gives  $H(F)_p$

(need L'Hôpital's rule for this)  
+ Taylor expand about  $\alpha=1$

It is true that  $I'_2(A;B|c)_p = I'_2(A;B|D)_p$

but non-negativity + data processing

do not hold ~~in~~ in general!

(see arXiv:1212.0248 Linden, Mosonyi,  
Winter.

This paper shows that there are  
generally no constraints placed on  
linear combinations of Rényi entropies)

Since these properties are used in applications  
all the time, we should essentially discard  
 $I'_\alpha$  as a useful generalization.

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So what should we do instead?

Consider that all known Renyi generalizations of information quantities make use of the

$$\text{relative entropy } D(\rho||\sigma) = \text{Tr}\{\rho \log \rho\} - \text{Tr}\{\rho \log \sigma\}$$

$$\text{That is } H(A)_\rho = -D(\rho_A || I_A)$$

$$H(A|B)_\rho = -D(\rho_{AB} || I_A \otimes \rho_B)$$

$$I(A;B)_\rho = D(\rho_{AB} || \rho_A \otimes \rho_B)$$

So we should figure out how to write  $I(A;B|C)_\rho$  as a relative entropy.

Using the definition, we find

$$\begin{aligned} I(A;B|C) &= -\text{Tr}\{\rho_{AC} \log \rho_{AC}\} \\ &\quad -\text{Tr}\{\rho_{BC} \log \rho_{BC}\} \\ &\quad + \text{Tr}\{\rho_C \log \rho_C\} \\ &\quad + \text{Tr}\{\rho_{ABC} \log \rho_{ABC}\} \end{aligned}$$

$$= \text{Tr}\{\rho_{ABC} \log \rho_{ABC}\} - \text{Tr}\{\rho_{ABC} [\log \rho_{AC} + \log \rho_{BC} - \log \rho_C]\}$$

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$$\begin{aligned} &= \text{Tr} \left\{ \rho_{ABC} \log \rho_{ABC} \right\} \\ &\quad - \text{Tr} \left\{ \rho_{ABC} \log \left[ \exp \left( \log \rho_{AC} + \log \rho_{BC} - \log \rho_C \right) \right] \right\} \\ &= D \left( \rho_{ABC} \parallel \exp \left\{ \log \rho_{AC} + \log \rho_{BC} - \log \rho_C \right\} \right) \end{aligned}$$

This operator is interesting  
if it is non-negative &  
a fact subnormalized  
w/  $\text{Tr} \{ \cdot \} \leq 1$

(from triple matrix generalization  
of Golden-Thompson  
inequality from  
stat. mech.)

We know that the Renyi relative  
entropy is

$$D_\alpha(\rho \parallel \sigma) = \frac{1}{\alpha-1} \log \text{Tr} \left\{ \rho^\alpha \sigma^{1-\alpha} \right\}$$

there is also the "sandwiched" version

$$\tilde{D}_\alpha(\rho \parallel \sigma) = \frac{1}{\alpha-1} \log \text{Tr} \left\{ \left( \sigma^{\frac{1-\alpha}{2\alpha}} \rho \sigma^{\frac{1-\alpha}{2\alpha}} \right)^\alpha \right\}$$

For these, we have  $\lim_{\alpha \rightarrow 1} D_\alpha(\rho \parallel \sigma) = D(\rho \parallel \sigma)$

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So we might guess ~~that~~ to take  
2-CMI as

$$D_2(\rho_{ABC} \parallel \exp \{ \log \dots \} ),$$

but there is a better choice.

The Lie-Trotter product formula  
gives that

$$\exp \{ \log \rho_{AC} + \log \rho_{BC} - \log \rho_C \}$$

$$= \lim_{\alpha \rightarrow 1} \left( \rho_{AC}^{\frac{1-\alpha}{2}} \rho_C^{\frac{\alpha-1}{2}} \rho_{BC} \rho_C^{\frac{\alpha-1}{2}} \rho_{AC}^{\frac{1-\alpha}{2}} \right)^{\frac{1}{1-\alpha}}$$

this could be any operator  
ordering, but we take  
them such that resulting  
operator is positive.

So we define

$$I_2(A; B|C)_{\text{eff}} = \frac{1}{2-1} \log \text{Tr} \left\{ \rho_{ABC} \rho_{AC}^{\frac{1-\alpha}{2}} \rho_C^{\frac{\alpha-1}{2}} \rho_{BC} \rho_C^{\frac{\alpha-1}{2}} \rho_{AC}^{\frac{1-\alpha}{2}} \right\}$$

We can prove that

$$\lim_{\alpha \rightarrow 1} I_2(A; B|C)_{\text{eff}} = I(A; B|C) \quad \checkmark \text{ by L'Hopital's \& Taylor expansion}$$



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Data processing under local operations <sup>on B</sup> for  $\alpha \in [0, 2]$

For this, it suffices to prove that

$$\text{Tr} \left\{ P_{ABC} P_{AC}^{\frac{1-\alpha}{2}} P_C^{\frac{\alpha-1}{2}} P_{BC} P_C^{\frac{\alpha-1}{2}} P_{AC}^{\frac{1-\alpha}{2}} \right\} (*)$$

- 1) jointly concave in  $P_{ABC} + P_{BC}$  for  $\alpha \in [0, 1]$
- open question  $\rightarrow$  2) " " in  $P_{ABC} + P_{AC}$  "
- 3) jointly convex in  $P_{ABC} + P_{BC}$  for  $\alpha \in [1, 2]$
- open question  $\rightarrow$  4) " " in  $P_{ABC} + P_{AC}$  for  $\alpha \in [1, 2]$

Together), We know the Lieb concavity theorem, that

$$(S, R) \rightarrow \text{Tr} \{ S^\alpha X R^{1-\alpha} X^\dagger \}$$

is jointly concave in  $S + R$

for  $\alpha \in [0, 1]$

for  $S, R \geq 0$   $\dagger$   $X$  arbitrary

So set  $S = P_{ABC}$ ,  $R = P_{BC}$

$$\dagger X = ~~P_{AC}^{\frac{1-\alpha}{2}} P_C^{\frac{\alpha-1}{2}}~~ P_{AC}^{\frac{1-\alpha}{2}} P_C^{\frac{\alpha-1}{2}}$$

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Can define sandwiched Rényi CMI as

$$\tilde{I}_\alpha(A; B|C)_{\text{el}} = \tilde{D}_\alpha(P_{ABC} \| (P_{AC}^{\frac{1-\alpha}{2\alpha}} P_C^{\frac{\alpha-1}{2\alpha}} P_{BC}^{\frac{1-\alpha}{2\alpha}} P_C^{\frac{\alpha-1}{2\alpha}} P_{AC}^{\frac{1-\alpha}{2\alpha}})^{\frac{1}{\alpha}})$$

can show that this one obeys  
data processing under local ops <sup>on B</sup> for  
 $\alpha \in [\frac{1}{2}, \infty]$

data processing under local ops gives  
non-negativity (for  $\alpha$  for which  
DP holds)

Why? Take local ops to be trace out  
maps

Then

$$\begin{aligned} I_\alpha(A; B|C)_{\text{el}} &= D_\alpha(P_{ABC} \| (P_{AC}^{\frac{1-\alpha}{2\alpha}} P_C^{\frac{\alpha-1}{2\alpha}} P_{BC}^{\frac{1-\alpha}{2\alpha}} P_C^{\frac{\alpha-1}{2\alpha}} P_{AC}^{\frac{1-\alpha}{2\alpha}})^{\frac{1}{\alpha}}) \\ &\geq D_\alpha(P_{AC} \| (P_{AC}^{\frac{1-\alpha}{2\alpha}} P_C^{\frac{\alpha-1}{2\alpha}} P_C^{\frac{\alpha-1}{2\alpha}} P_{AC}^{\frac{1-\alpha}{2\alpha}})^{\frac{1}{\alpha}}) \\ &= D_\alpha(P_{AC} \| P_{AC}) = 0 \end{aligned}$$

Big conjecture: Monotonicity in  $\alpha$

I.e.,  $\forall 0 \leq \alpha \leq \beta$ , we have

$$I_\alpha(A; B|C)_{p|p} \leq I_\beta(A; B|C)_{p|p}$$

$$+ \tilde{I}_\alpha(\dots) \leq \tilde{I}_\beta(\dots)$$

lots of numerical evidence +  
a proof for  $\alpha$  in a neighborhood of 1.

Idea for proof: if derivative of  $I_\alpha$  w.r.t  $\alpha$  is non-negative, then the function is monotone increasing.

Can show that

$$\frac{d}{d\alpha} I_\alpha = \frac{(\alpha-1)^2}{2} V(A; B|C) + o((\alpha-1)^3)$$

something positive

where  $V(A; B|C) =$

$$\text{Tr} \left\{ p_{ABC} \left[ \log p_{ABC} + \log p_C - \log p_{AC} - \log p_{BC} - I(A; B|C) \right]^2 \right\}$$

$\geq 0$

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Conjecture implies that

$$I(A; B|C)_f \geq \frac{1}{4} \| P_{ABC} - P_{AC}^{1/2} P_C^{-1/2} P_{BC} P_C^{-1/2} P_{AC}^{1/2} \|$$

So it would mean that if

$$I(A; B|C)_f \leq \epsilon \quad \text{then}$$

if you have  $P_{BC}$  the

"pretty good" recovery map

$$R_{BC \rightarrow ABC}(\cdot) = P_{AC}^{1/2} P_C^{-1/2}(\cdot) P_C^{-1/2} P_{AC}^{1/2}$$

is good for recovering  $P_{ABC}$

from  $P_{BC}$  alone.

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Final point: the method given here extends beyond just CMI -

It gives a Renyi generalization for any information quantity that can be realized as a linear combination of vN entropies

w/ coefficients taken from  $\{-1, 0, +1\}$

Open questions: Solve conjecture + make use of this in applications.