

# Quantum Capacity for nice channels

Marius Junge

University of Illinois at Urbana-Champaign

Paris-October OSQPI 2015

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joint work with Li Gao and Nicholas LaRacurente

# Overview

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- Classical Entropy and capacity

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- The minimal entropy  $H_{\min}(\Phi) = \min_p H(\Phi(p))$  is a measure of how noisy a channel is.  $H_{\min}(id) = 0$

# Relative Entropy and mutual information

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- Let  $p^{XY}$  be a distribution on a product space and  $p^X(x) = \sum_y p(x, y)$ ,  $p^Y = \sum_x p(x, y)$  be the marginals,  $p(x|y) = p(x, y)/p^Y(y)$  the conditional probability.

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- and

$$I(X, Y)_p = H(X) + H(Y) - H(XY) = H(p^X) + H(p^Y) - H(p^{XY})$$

is the **mutual information**.



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- $H_{\min} = \lim_{p \rightarrow \infty} p \log \|\mathcal{N}^* : \ell_p \rightarrow \ell_\infty\|.$

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But only for classical channels!

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- $H_{\min}(\mathcal{N}) = \inf_{\text{tr}(\rho)=1, \rho \geq 0} H(\mathcal{N}(\rho))$ .
- Hastings (2008) showed that

$$H_{\min}(\Phi \otimes \Psi) \neq H_{\min}(\Phi) + H_{\min}(\Psi)$$

in general.



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Then  $C^{(d)}(\mathcal{N}) = \sup_{\rho = \sum_k p_k \rho_k^{A'A}} H(\mathcal{N}(\rho))$

$$+ \sum_k p_k (H(A)_{\mathcal{N}(\rho_k)} - H(AB)_{id \otimes \mathcal{N}(\rho_k)})$$

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- For  $d = 1$  we find Holevo's capacity. Shor showed (before Hastings) that the additivity of the min-entropy is equivalent to the additive of the one shot Holevo capacity.

# Classical capacity with unlimited entanglement

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- For  $d = \dim(H_{A'}) = |A'|$  we find the **classical capacity with assisted entanglement**

$$C_{EA}(\mathcal{N}) = \sup_{\rho^{A'A}} I(A, B)_{id \otimes \mathcal{N}(\rho^{A'A})}$$

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- The classical capacity with assisted entanglement is a rate (early succes in QIT: HSW, Schor, Devetak, Winter,...)



# Coherent information

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- Let  $\mathcal{N} : S_1(A') \rightarrow S_1(B)$  be a channel and  $\rho^{AB} = id \otimes \mathcal{N}(\rho^{A'A})$  the image of a bipartite state. We may write the mutual information as

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- Indeed, according to [DJKRS]

$$I_c(A)B) = c \frac{d}{dp} \|\rho^{BA}\|_{S_1(H_B; S_p(H_A))} \Big|_{p=1}$$

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- Using the contraction

$id \otimes tr_C : S_1(H_{BC}; S_p(H_A)) \rightarrow S_1(H_B; S_p(H_A))$ , we deduce by differentiation **strong super additivity**

$$H(B) - H(AB) \leq H(BC) - H(ABC) .$$

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- The regularization is general hard to calculate. By teleportation we have  $2Q(\mathcal{N}) \leq 2Q_E(\mathcal{N}) = C_{EA}(\mathcal{N})$ .



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- Our main result is that there is a nice class of channels  $\theta_f : S_1(L_2(M)) \rightarrow S_1(L_2(M))$  such that

$$\begin{aligned} \max\{\max_k \ln n_k, \tau(f \ln f)\} &\leq Q(\theta_f) \leq Q^{pot}(\theta_f) \\ &\leq \max_k \ln n_k + \tau(f \ln f) . \end{aligned}$$

Here  $f \in N$  is the symbol of a channel and  $M = \bigoplus_k M_{n_k}$  is a finite dimensional von Neumann algebra.

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Here  $f \in N$  is the symbol of a channel and  $M = \bigoplus_k M_{n_k}$  is a finite dimensional von Neumann algebra.

- For these channels we also have estimates for the (CQE) region (classical, quantum, entanglement) region of the channel, which goes back to Schor, see also Wilde.

# Motivating examples

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\* Let  $G$  be a finite group,  $\lambda(g)e_h = e_{gh}$  the left regular representation on  $\ell_2(G) = L_2(L(G))$ . Let  $f : G \rightarrow \mathbb{R}_+$  and

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- \* Then  $\theta_1$  is the conditional expectation onto right regular representation  $L(G)' = R(G) = \sum_k M_{n_k} = M$ . We have

$$\max(\max_k \ln n_k, \tau(f \ln f)) \leq Q^{(1)}(\theta_f) \leq \tau(f \ln f) + \max_k \ln n_k.$$

Both lower bounds are achievable as well as the upper bound for  $G = \mathbb{Z}_n$ .

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- Let  $G$  be a finite group,  $\lambda(g)e_h = e_{gh}$  the left regular representation on  $\ell_2(G) = L_2(L(G))$ . Let  $f : G \rightarrow \mathbb{R}_+$  and

$$\theta_f(\rho) = \frac{1}{|G|} \sum_{g \in G} f(g) \lambda(g)^* \rho \lambda(g).$$

- Then  $\theta_1$  is the conditional expectation onto right regular representation  $L(G)' = R(G) = \sum_k M_{n_k} = M$ . We have

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- ✳ For the conditional expectation  $\theta_f = \theta_1$  we have equality  $Q(\theta_f) = \max_k \ln n_k$ . Teleportation  $Q(\theta_1) \leq 2C_{EA}(\theta_1) = \ln n$  confirms that for groups the maximal dimension of an irreducible representation  $\pi$  satisfies

$$|\dim(\pi)| \leq |G|^{1/2} .$$

- ✳ According to Vershik et al, the symmetric group  $n = m!$  satisfies  $\max_k \ln n_k \geq \sqrt{n} - c\sqrt{m}$  and hence we find examples where entanglement improves the rate only by a very small amount  $\sqrt{m}$  which is small compared to the value  $\sqrt{m!}$ .



- \* For  $G = \mathbb{Z}_d^m \rtimes \mathbb{Z}_m$  the maximal dimension of an irreducible representation  $m$  is small compared to  $n = |G| = md^m$  and hence our new estimates outperform the more classical upper estimates

$$C_{EA}(\theta_f) = \ln n + \tau(f \ln f)$$

Coincidentally, this estimate is also new (but related to previous work with Ruan and Neufang).



# Schur multipliers

- ✱ Starting from a finite group we can also define the Schur multiplier

$$\theta_f([x_{gh}]) = \sum_{gh} f(g^{-1}h)x_{gh}$$

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- ✱ Schur multipliers are so-called degradable channels and those don't require regularization! Our previous group channels, however, are in general not degradable.



✳ Winter et al introduced the potential capacity

$$Q^{(pot)}(\mathcal{N}) = \sup_{\mathcal{N}'} Q^{(1)}(\mathcal{N} \otimes \mathcal{N}') - Q^{(1)}(\mathcal{N}).$$

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- ❄ Our upper estimates also hold for  $Q^{(pot)}$ .

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- Our class of channels include quantum group channels, in particular Drinfeld doubles, which have Kraus operators which are neither unitaries or projections.
- Our examples also include random unitaries given by products of the (Majorana) Clifford generators. However, in these particular cases only the estimates for  $Q^{(pot)}$  are 'new'. As of now better estimates are obtained by just adding entanglement and use teleportation.



## Tools

- (Stinespring-Kraus-Haydon/Winter) Every channel comes with a partial isometry

$$V : H_A \rightarrow H_B \otimes H_E$$

and hence a subspace  $\text{st}(\mathcal{N}) \subset H_B \otimes H_E$  of a tensor product of two Hilbert spaces.

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where  $R_d(X) = X \otimes_h R_d$  and  $C_p^d(X) = C_p \otimes_h X$ . Indeed,

$$Q^{(1)}(\mathcal{N}) = \sup_d \frac{d}{dp} \text{rc}_{p,d}(\text{st}_p(\mathcal{N}))^2 .$$

# Local comparison

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- Using similar techniques we will prove

$$\begin{aligned} \|\theta_f \otimes id(\rho)\|_{S_1(H_B; S_p(H_A))} &\leq \|f\|_p \|\theta_1 \otimes id(\rho)\|_{S_1(H_B; S_p(H_A))} \\ &\leq \|f\|_p \|\theta_f \otimes id(\rho)\|_{S_1(H_B; S_p(H_A))} . \end{aligned}$$

This allows us to prove estimates for the private capacity.

# Lower bounds

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- Lower bounds are obtained by calculating the min-cb-entropy or so-called reversed coherent information

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- An important tool for our analysis feature is [the standard form of a finite von Neumann algebra](#).



Thanks for listening