Peter Brown, Hamza Fawzi and Omar Fawzi

Paper 1: arXiv:2106.13692 Paper 2: arXiv this week

Mar 07 2022



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Motivation – Device-independence

Bell-nonlocality



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Nonlocal correlations are inherently random.

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- Nonlocal correlations are inherently random.
- Foundation for randomness expansion / key-distribution protocols!

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- Nonlocal correlations are inherently random.
- Foundation for randomness expansion / key-distribution protocols!
- Security and analysis relies on the *rate* (bits per round).

- Optimization of von Neumann entropies

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Asymptotic rates are given by:

Randomness expansion

$$H(AB|X = x^*, Y = y^*, E)$$

QKD

$$H(A|X = x^*, E) - H(A|X = x^*, Y = y^*, B)$$

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DI bounds

Want to compute

$$r(P) = \inf H(A|X = x^*, E)$$

where inf over all devices compatible with P.



Idea: Relax to a problem we can approximate...

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Noncommutative polynomial optimization problems

$$\begin{array}{ll} \inf & \operatorname{Tr}\left[\rho P(Z_1,\ldots,Z_n)\right] \\ \mathrm{s.t.} & \operatorname{Tr}\left[\rho Q_i(Z_1,\ldots,Z_n)\right] \geq w_i \\ & R_i(Z_1,\ldots,Z_n) \geq 0 \end{array}$$

infimum over $(\mathcal{H}, \rho, Z_1, \ldots, Z_n)$.

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Why?

(Convergent) SDP hierarchy gives lower bounds (NPA hierarchy [PNA10]).

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Goal:

Search for variational bounds on entropies with an NCPOP form.

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We actually work with the relative entropy

$$D(\rho \| \sigma) = \operatorname{Tr} \left[\rho(\log \rho - \log \sigma) \right].$$

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

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The goal

Derive something of the form

$$D(\rho \| \sigma) \leq \sum_{i=1}^{m} \sup_{Z} \operatorname{Tr} \left[\rho p_i(Z) \right] + \operatorname{Tr} \left[\sigma q_i(Z) \right]$$

with p_i and q_i some polynomials and with the RHS converging as $m \to \infty$.

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The goal

Know
$$D(\rho \| \sigma) = \sup_{(X, Y, z) \in \mathcal{F}} \operatorname{Tr} [\rho X] + \operatorname{Tr} [\sigma Y] + z$$

Derive something of the form

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$$\ln(x) = \int_0^1 \frac{x-1}{t(x-1)+1} dt \ge \sum_{i=1}^m w_i f_{t_i}(x)$$

where $f_t(x) = \frac{x-1}{t(x-1)+1}$ (RHS converges as $m \to \infty$).

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Main Result

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and RHS converges as $m \to \infty$.

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

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Theorem

The rate inf $H(A|X = x^*, Q_E)$ is never smaller than

$$c_{m} + \inf_{\text{strategies}} \sum_{i=1}^{m-1} \frac{w_{i}}{t_{i} \ln 2} \sum_{a} \operatorname{Tr} \left[\rho_{Q_{A}Q_{E}}(M_{a|x^{*}} \otimes (Z_{a,i} + Z_{a,i}^{*} + (1 - t_{i})Z_{a,i}^{*}Z_{a,i}) + t_{i}Z_{a,i}Z_{a,i}^{*}) \right]$$

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Remarks

Can now be easily relaxed to an NCPOP and solved using NPA [PNA10].

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<u>Remarks</u>

- Can now be easily relaxed to an NCPOP and solved using NPA [PNA10].
- Similar results for $H(AB|X = x, Y = y, Q_E)$ or $H(A|XQ_E)$ and others.

Results I - Recovering tight bounds for the CHSH game



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Results II - Improved DIQKD rates

Bounding inf $H(A|X = 0, Q_E) - H(A|X = 0, Y = 2, B)$



Application: squashed entanglement

The squashed entanglement [CW04] for a bipartite state ρ_{AB} is defined as

$$E(A:B) := \inf_{\operatorname{Tr}_E[\rho_{ABE}] = \rho_{AB}} I(A:B|E).$$

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Suppose ρ_{ABED} is pure, then

$$I(A:B|E) = H(A|D) + H(A|E)$$

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m-th variational lower bound

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Difficult to solve nonconvex / unbounded dimension

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SDP lower bounds via NPA hierarchy!

Results - Werner state squashed entanglement

Consider a two-qubit Werner state

$$\rho = \rho \frac{\Pi_{\rm sym}}{{\rm Tr}\left[\Pi_{\rm sym}\right]} + (1 - \rho) \frac{\Pi_{\rm asym}}{{\rm Tr}\left[\Pi_{\rm asym}\right]}$$

with $p \in [0, 1]$.

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with $p \in [0, 1]$.

Using variational lower bounds and heuristic upper bounds we find $d_A = d_B = 2$



Summary

• Technical result: convergent variational upper bounds on $D(\rho \| \sigma)$.

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<u>Outlook</u>

More efficient computations?

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- More efficient computations?
- Convergence of the numerics?

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<u>Outlook</u>

- More efficient computations?
- Convergence of the numerics?
- Other applications?

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Bibliography



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Fix some linear constraint(s) C on the joint probability distribution of the devices $p_{AB|XY}$. E.g.

$$\frac{1}{4}\sum_{xy=a\oplus b}p(ab|xy)\geq 0.8.$$

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A strategy for C is a tuple $(Q_A Q_B Q_E, \rho, \{\{M_{a|x}\}_a\}_x, \{\{N_{b|y}\}_b\}_y)$ such that

$$p(ab|xy) = \operatorname{Tr} \left[\rho(M_{a|x} \otimes N_{b|y} \otimes I_E) \right]$$

satisfies the constraints in C.

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Through the post measurement state

$$\rho_{AQ_E} = \sum_{a} |a\rangle \langle a| \otimes \operatorname{Tr}_{Q_A Q_B} \left[(M_{a|x^*} \otimes I) \rho \right] \longrightarrow H(A|X = x^*, Q_E)$$

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DI bounds

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$$r(C) = \inf H(A|X = x^*, E)$$

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 $\begin{array}{l} \mbox{Difficult to solve} \\ \mbox{nonconvex} \ / \ \mbox{unbounded dimension} \end{array}$

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Bonus results - DICKA setting (Holz inequality)



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Bonus results – Generalized CHSH ($\alpha = 1.1$)



Bonus results – Generalized CHSH ($\alpha = 1.1$)



Bonus results – Generalized CHSH ($\alpha = 0.9$)

