

Limitations of monogamy, Tsirelson-type bounds, and other SDPs in quantum information

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SDPs in Quantum Information

Semidefinite Programmings (SDPs) admit *polynomial time* solvers and plays an important role in quantum information.

- Consistency of reduced states, Quantum conditional min-entropy, local Hamiltonians
- QIP=PSPACE, QRG=EXP,

This talk is, however, about its **limitation** in

- Separability or entanglement detection,
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Problem 1: Separability

Definition (Separable and Entangled States)

A bi-partite state $\rho \in \mathcal{D}(\mathcal{X} \otimes \mathcal{Y})$ is *separable* if \exists dist. $\{p_i\}$,

$$\rho = \sum p_i \sigma_X^i \otimes \sigma_Y^i, \text{ s.t. } \sigma_X^i \in \mathcal{D}(\mathcal{X}), \sigma_Y^i \in \mathcal{D}(\mathcal{Y}).$$

Otherwise, ρ is *entangled*. Let $\text{Sep} \stackrel{\text{def}}{=} \{ \text{separable states} \}$.

Definition (Entanglement Detection)

A **KEY** problem: given the description of $\rho \in \mathcal{D}(\mathcal{X} \otimes \mathcal{Y})$, decide

Either $\rho \in \text{Sep}$, or ρ is far away from Sep .

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Alternative Formulation

Definition (Weak Membership)

$\text{WMem}(\epsilon, \|\cdot\|)$: for any $\rho \in D(\mathcal{X} \otimes \mathcal{Y})$, decide either $\rho \in \text{Sep}$ or $\|\rho - \text{Sep}\| \geq \epsilon$.

Via standard techniques in convex optimization, equivalent to

Definition (Weak Optimization)

$\text{WOpt}(M, \epsilon)$: for any $M \in \text{Herm}(\mathcal{X} \otimes \mathcal{Y})$, estimate the value of

$$h_{\text{Sep}(d,d)}(M) := \max_{\rho \in \text{Sep}} \langle M, \rho \rangle,$$

with additive error ϵ .

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$h_{\text{Sep}(d,d)}(M)$

$$h_{\text{Sep}(d,d)}(M) := \max_{\substack{x,y \in \mathbb{C}^d \\ \|x\|_2 = \|y\|_2 = 1}} \sum_{i,j,k,l \in [d]} M_{ij,kl} x_i^* x_j y_k^* y_l. \quad (1)$$

REMARK: this is an instance of *polynomial optimization* problems with a homogenous degree 4 objective polynomial and a degree 2 constraint polynomial.

Connections

Quantum Information:

- *Mean-field* approximation in statistical quantum mechanics.
- *Positivity* test of quantum channels.
- Data hiding, Channel capacities, Privacy,
- *17 more examples* in quantum information in [HM10].

Quantum Complexity:

- Quantum Merlin-Arthur Game with Two-Provers (QMA(2)).

Classical Complexity:

- Unique Game Conjecture and Small-set Expansion.
($\ell_2 \rightarrow \ell_4$ norm)

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Heuristics

Separability Criteria:

- Positive Partial Transpose (PPT) : $\rho^{T_Y} = \rho$? [PH]
- Reduction Criteria: $I_X \otimes \rho_Y \geq \rho$? [HH]
- **FAILURE**: any such test has **arbitrarily large error**. [BS]

Doherty-Parrilo-Spedalieri (DPS) hierarchy:

- ρ is k -extendible if \exists *symmetric* $\sigma \in \mathcal{D}(X \otimes Y_1 \otimes \cdots \otimes Y_k)$,
 $\forall i, \rho = \sigma_{XY_i}$.

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Computational Hardness

reference	k	c	s	n
GNN12	2	1	$1 - \frac{1}{d \cdot \text{poly} \log(d)}$	$O(d)$
Per12	2	1	$1 - \frac{1}{\text{poly}(d)}$	$O(d)$
AB+08	$\sqrt{d} \cdot \text{poly} \log(d)$	1	0.99	$O(d)$
CD10	$\sqrt{d} \cdot \text{poly} \log(d)$	$1 - 2^{-d}$	0.99	$O(d)$
HM13	2	1	0.01	$\frac{\log^2(d)}{\text{poly} \log(d)}$

Table: Hardness results for $h_{\text{Sep}^k(d)}$ (extension of $h_{\text{Sep}(d,d)}$ to k parties.)

Hardness in the following sense: determining satisfiability of 3-SAT instances with n variables and $O(n)$ clauses can be reduced to distinguishing between $h_{\text{Sep}^k(d)} \geq c$ and $\leq s$ as above.

Computational Hardness

Exponential Time Hypothesis (ETH)

The 3-SAT problem with n variables requires $2^{\Omega(n)}$ time to solve.

- Combine with [HM13] hardness result \Rightarrow approximation of $h_{\text{Sep}(d)}$ with constant precision requires $d^{O(\log(d))}$ time.
- A matching upper bound: DPS to $O(\log(d)/\epsilon^2)$ level for 1-LOCC M : time $d^{O(\log(d)/\epsilon^2)} \rightarrow d^{O(\log(d))}$. [BYC, BH]

Question: any unconditional lower bounds for DPS or any SDPs? any matching upper bounds?

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Problem 2: Non-local Games

Non-local Game (denoted G):

- Two physically **separated** players Alice and Bob. **No** communication once the game starts.
- Sets of questions S, T and answers A, B and a distribution $\pi : S \times T \rightarrow [0, 1]$.
- Sample $(s, t) \in S \times T \sim \pi$ and ask Alice and Bob respectively. Obtain answers $a \in A, b \in B$.
- Determine **win** or **lose** by a predicate $V(a, b|s, t) \in \{0, 1\}$.

Motivation: Bell-violation (quantum **non-locality**) in a game language. Also related to **quantum multi-prover interactive proofs** with shared entanglement.

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Strategies:

- Denote by $P[a, b|s, t]$ the probability of answering (a, b) upon receiving (s, t) .
- **Quantum strategies:** share a quantum state $|\psi\rangle \in \mathcal{H}_A \otimes \mathcal{H}_B$ and answer w.r.t measurements $\{A_s^a\}$ and $\{B_t^b\}$,

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Non-local Games (cont'd)

Definition (Game Value)

$$\omega(G) = \max_P \sum_{a,b,s,t} \pi(s,t) V(a,b|s,t) P(a,b|s,t).$$

Example: CHSH game:

- $A = B = S = T = \{0, 1\}$ and $\pi(s,t) = 1/4, \forall (s,t) \in S \times T$.
- $V(a,b|s,t) = 1 \iff a \oplus b = s \wedge t$.

Question: calculate $\omega^*(G)$ for any given G . How hard is that?

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- **Classical strategies:** $\omega(\text{CHSH}) = 3/4$. **Quantum strategies:** $\omega^*(\text{CHSH}) = \cos^2(\pi/8) \approx 0.85$.
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Calculating $\omega^*(G)$ for quantum strategies

$\omega^*(G)$ for quantum strategies: an optimization problem!

$$\omega^*(G) = \lim_{d \rightarrow \infty} \max_{|\psi\rangle \in \mathbb{C}^{d \times d}} \max_{A_s^a, B_t^b} \sum_{a,b,s,t} \pi(s,t) V(a,b|s,t) \langle \psi | A_s^a \otimes B_t^b | \psi \rangle.$$

- $\omega^*(G)$ is not known to be **computable**.
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Computational Hardness

reference	k	c	s	n
KK+11	3	1	$1 - \frac{1}{\text{poly}(Q)}$	$O(Q)$
IKM09	2	1	$1 - \frac{1}{\text{poly}(Q)}$	$O(Q)$
IV12	4	1	$2^{-Q^{\Omega(1)}}$	$Q^{\Omega(1)}$
Vid13	3	1	$2^{-Q^{\Omega(1)}}$	$Q^{\Omega(1)}$

Table: Hardness results for $\omega^*(G)$ where G is a one-round k -prover interactive proof protocol with question alphabet size Q .

Hardness in the following sense: determining satisfiability of 3-SAT instances with n variables and $O(n)$ clauses can be reduced to distinguishing between $\omega^*(G) \geq c$ and $\leq s$ as above.

Result I: Unconditional Hardness for h_{Sep} ?

Will the hardness of $h_{\text{Sep}(d)}$ for const ϵ hold w/o ETH?

Theorem (Main I.1)

The DPS hierarchy (or general Sum-of-Squares SDP) requires $\Omega(\log(d))$ levels to solve $h_{\text{Sep}(d)}$ with constant precision.

Theorem (Main I.2)

*Any SDP **relaxation** that estimate $h_{\text{Sep}(d)}(M)$ with constant errors requires size $d^{\Omega(\log(d))}$.*

Remark: Match $d^{\Omega(\log(d))}$ time bound when assuming ETH.

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The DPS hierarchy (or general Sum-of-Squares SDP) requires $\Omega(\log(d))$ levels to solve $h_{\text{Sep}(d)}$ with constant precision.

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Any SDP relaxation that estimate $h_{\text{Sep}(d)}(M)$ with constant errors requires size $d^{\Omega(\log(d))}$.

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There exists a family of games $\{G_n\}$ s.t. the NPA hierarchy requires $\Omega(n)$ levels to distinguish $\omega^(G) = 1$ from $\omega^*(G) = 1 - \Omega(1/n^2)$.*

Theorem (Main II.2)

Any SDP relaxation that estimates $\omega^(G)$ with precision $O(1/n^2)$ requires size $(n/\log(n))^{\Omega(n)}$.*

Remark: Match the computational hardness of [IKM].
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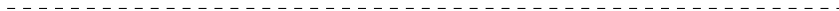
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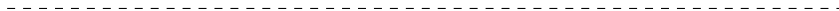
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QMA(2) vs QMA



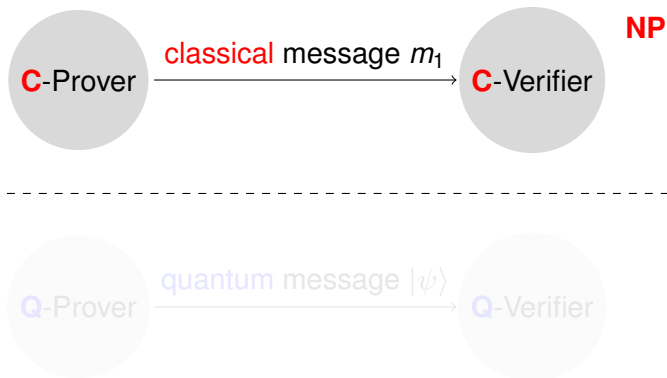
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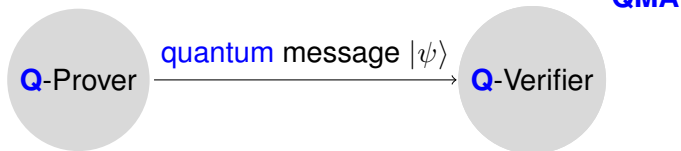
quantum message $|\psi\rangle$



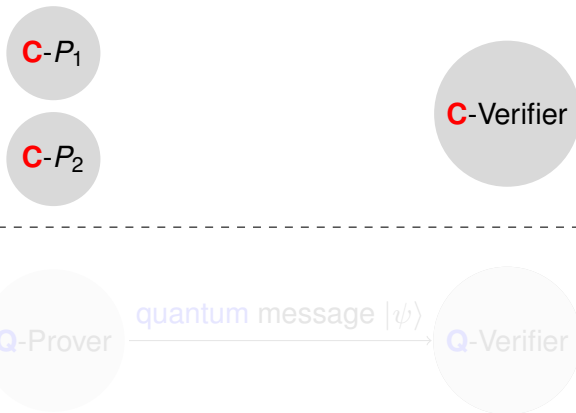
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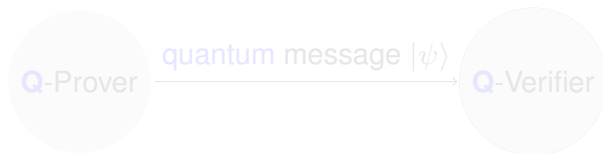
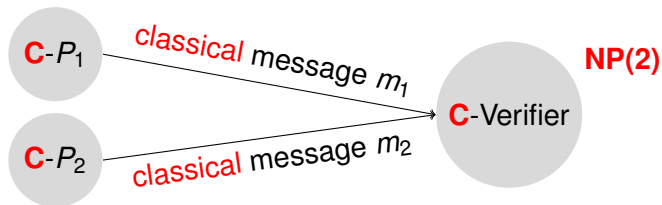
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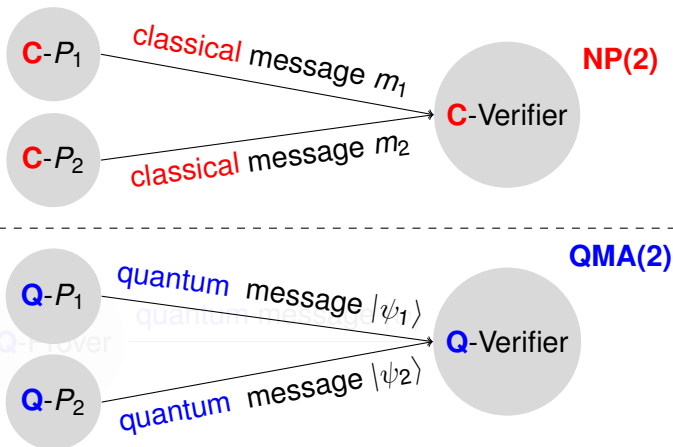
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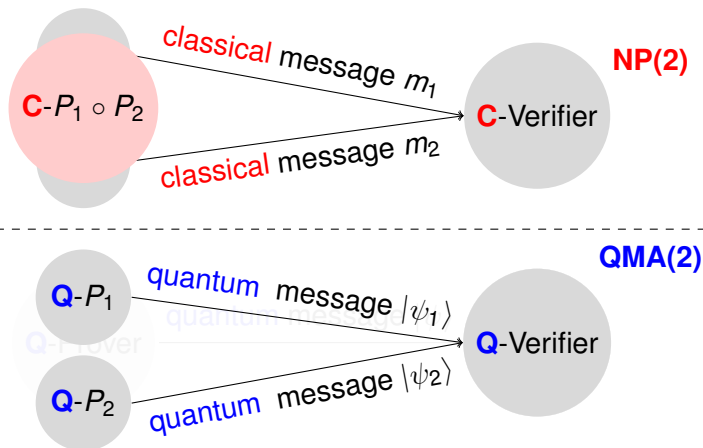
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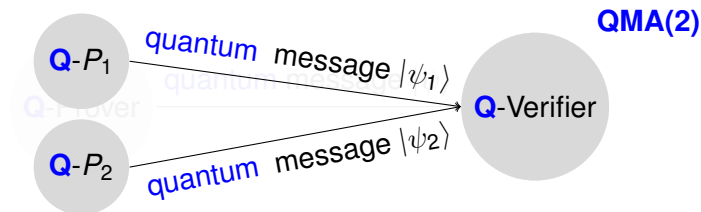
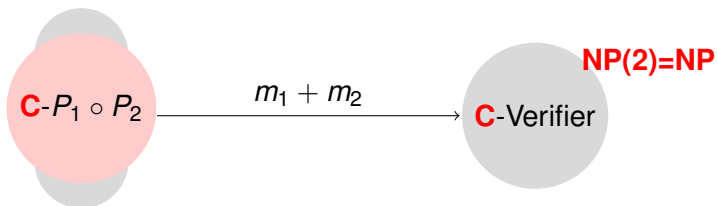
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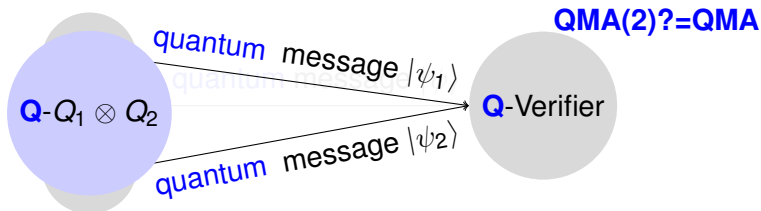
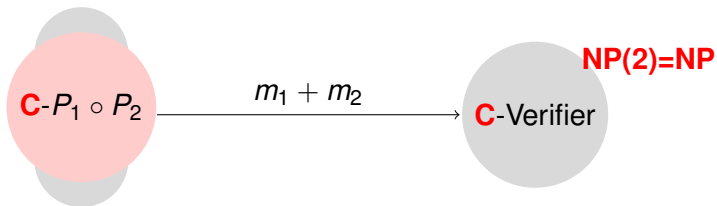
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- First study in [KMY01, KMY03]. Surprising: $\text{NP} \subseteq \text{QMA}(2)_{\log}$ [BT09] v.s. $\text{QMA}_{\log} = \text{BQP}$ [MW05].

Main open question is to improve the trivial upper bound NEXP .

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QMA(2) vs. EXP

- A recent result shows $\text{QMA}(2) \subseteq \text{EXP}$ with a logarithmic overhead [KST17].

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Hardness applies to QMA(2)

- Our explicit hard instance is a **valid** QMA(2) instance.
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Unconditional proof of Watrous's dis-entangler conjecture

- Dis-entangler: a hypothetical channel that a) its output is always ϵ -close to a separable state, and b) its image is δ -close to any separable state, both in trace distance.
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- Introduce hardness of SDPs/SoS into quantum problems.
- Deal with their limitations, such as boolean domains and commutative problems.

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Principle of Sum-of-Squares

One way to show that a polynomial $f(x)$ is *nonnegative* could be

$$f(x) = \sum a_i(x)^2 \geq 0.$$

Example

$$\begin{aligned} f(x) &= 2x^2 - 6x + 5 \\ &= (x^2 - 2x + 1) + (x^2 - 4x + 4) \\ &= (x - 1)^2 + (x - 2)^2 \geq 0. \end{aligned}$$

Such a decomposition is called a *sum of squares (SOS) certificate* for the non-negativity of f . The min degree, \deg_{SOS} .

Principle of SoS : constrained domain

Definition (Variety)

A set $V \subseteq \mathbb{C}^n$ is called an *algebraic variety* if
 $V = \{x \in \mathbb{C}^n : g_1(x) = \dots = g_k(x) = 0\}$.

Non-negativity of $f(x)$ on V could be shown by

$$f(x) = \sum a_i(x)^2 + \sum b_j(x)g_j(x) \geq 0.$$

Question: whether all nonnegative polynomials on certain variety have a **SOS certificate**? **Hilbert 17th problem!**

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SoS in Optimization

$$\begin{array}{ll} \max & f(x) \\ \text{subject to} & g_i(x) = 0 \quad \forall i \end{array} \quad (2)$$

is equivalent to (justified by *Positivstellensatz*)

$$\begin{array}{ll} \min & \nu \\ \text{such that} & \nu - f(x) = \sigma(x) + \sum_i b_i(x)g_i(x), \end{array} \quad (3)$$

where $\sigma(x)$ is SOS and $b_i(x)$ is any polynomial.

SoS relaxation: Lasserre/Parrilo Hierarchy

- If $\sigma(x), b_i(x)$ have any degrees (or $\deg_{\text{SoS}}(\nu - f)$), then problem (3) is equivalent to problem (2).
- By bounding the degrees, we get the Lasserre/Parrilo hierarchy, which is a SDP hierarchy.

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Recall $h_{\text{Sep}(d,d)}(M)$

$$h_{\text{Sep}(d,d)}(M) := \max_{\substack{x,y \in \mathbb{C}^d \\ \|x\|_2 = \|y\|_2 = 1}} \sum_{i,j,k,l \in [d]} M_{ij,kl} x_i^* x_j y_k^* y_l. \quad (5)$$

Recall: this is an instance of *polynomial optimization* problems with a homogenous degree 4 objective polynomial and a degree 2 constraint polynomial.

Its Lasserre's hierarchy is the DPS hierarchy with full symmetry.

Non-commutative (nc) SoS

Given $F, G_1, \dots, G_m \in \mathcal{R}\langle X \rangle$, define

$$F_{\max} := \sup_{\rho, X=(X_1, \dots, X_n)} \text{Tr}[\rho F(X)]$$

$$\text{subject to } \rho \geq 0, \text{Tr } \rho = 1, G_1(X) = \dots = G_m(X) = 0. \quad (6)$$

Note that the supremum here is over density operators ρ and Hermitian operators X_1, \dots, X_n that may be infinite dimensional;

ncSoS

A non-commutative SoS proof can be expressed similarly as

$$c - F = \sum_{i=1}^k P_i^\dagger P_i + \sum_{i=1}^m Q_i G_i R_i, \quad (7)$$

for $\{P_i\}, \{Q_i\}, \{R_i\} \subset \mathcal{R}\langle X \rangle$. Likewise the best degree- d ncSoS proof can be found in time $n^{O(d)} m^{O(1)}$ by SDPs.

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General SDPs

- The DPS and NPA hierarchies are just SoS and ncSoS SDP hierarchies.
- Thus, lower bounds for deg_{SoS} \Rightarrow lower bounds for DPS and NPA.
- How about general SDPs?

Lee-Raghavendra-Steurer

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- Any deg_{SoS} lower bound on $\{0, 1\}^n \Rightarrow$ a lower bound on SDP relaxations.
- SDP relaxation: $\forall x \in \{0, 1\}^n, \exists$ relaxed X^* , s.t., $f(x) = F(X^*)$. Embedding!

General SDPs

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Pseudo-distribution

Dual of the SOS cone

- Let $\Sigma_{d,2D}$ be the cone of all PSD matrices representing SOS polynomials with degree up to $2D$.
- The dual cone $\Sigma_{d,2D}^*$ is moment $M_D(x) \geq 0$, where entry (α, β) of $M_d(x)$ is $\int x^{\alpha+\beta} \mu(dx)$, $|\alpha|, |\beta| \leq d$.

Pseudo-distribution/expectation

- Moment $M_D(x)$ gives rise to *pseudo-distribution*.
Expectation on it is *pseudo-expectation*.
- Behave similar to expectation for low-degree polynomials.

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A degree- d pseudo-expectation $\tilde{\mathbb{E}}$ is an element of $\mathcal{R}[x]_d^*$ (i.e. a linear map from $\mathcal{R}[x]_d$ to \mathcal{R}) satisfying

- **Normalization.** $\tilde{\mathbb{E}}[1] = 1$.
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$$f_{\text{SoS}}^d = \max\{\tilde{\mathbb{E}}[f] : \tilde{\mathbb{E}} \text{ of degree-}d \text{ satisfying } g_1, \dots, g_m\}. \quad (8)$$

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Integrality Gaps

What constitutes an integrality gap?

- An instance Φ that has $f_{\text{opt}}(\Phi)$ is small.
- But $f_{\text{SoS}}^d(\Phi)$ is large for some $d \Rightarrow$ lower bound at level d .

Example

- 3XOR: $O(n)$ clauses on n boolean variables:
 $x_i \oplus x_j \oplus x_k = C_{ijk}$.
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Reduction from A to B

- Reduction is an instance-mapping $\phi^A \rightarrow \phi^B$.
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- Only apply to $\{0, 1\}^n \Rightarrow$ no direct application on f_{Sep} or $\omega^*(G)$.
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- Reductions R_1, \dots, R_2 lead to an SoS integrality gap at the problem **A**.
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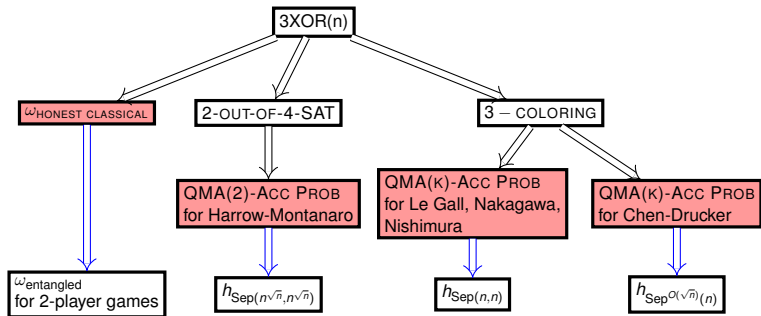
Real reductions for h_{Sep} and $\omega^*(G)$ 

Figure: All our results are derived from the integrality gaps of 3XOR.
Red nodes: problems over the boolean cube and LRS is applied.
Blue arrows are “embedding reductions”.

Reduction for h_{Sep}

$$3\text{XOR} \xrightarrow{R_1} 2\text{-OUT-OF-4-SAT-EQ} \xrightarrow{R_2} \text{QMA(2)-ACC PROB} \xrightarrow{R_3} h_{\text{Sep}}$$

- R_1 : a classical step. Low-degree & soundness similar to the degree reduction step in Dinur's proof of the PCP theorem.
- R_2 : a quantum step. Apply a modified QMA(2) protocol for 3-SAT [AB+09, HM13]. Low-degree due to the tests of the protocol. Soundness inherited from the protocol.
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Question And Answer

Thank you!
Q & A

SoS relaxation: Lasserre/Parrilo Hierarchy

- If $\sigma(x), b_i(x)$ have *any* degrees (or $\deg_{\text{SOS}}(\nu - f)$), then problem (3) is equivalent to problem (2).
- By bounding the degrees, we get the Lasserre/Parrilo hierarchy.

$$\begin{aligned} \min \quad & \nu \\ \text{such that} \quad & \nu - f(x) = \sigma(x) + \sum_i b_i(x)g_i(x), \end{aligned} \quad (9)$$

where $\sigma(x)$ is SOS and $b_i(x)$ is any polynomial and $\deg(\sigma(x)), \deg(b_i(x)g_i(x)) \leq 2D$.

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Observation

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