

# Improved Semidefinite Programming Hierarchy for $h_{\text{Sep}}$ with tools from Algebraic Geometry

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# The problem $h_{\text{Sep}}$

## Definition (Separable states)

A bipartite state  $\rho \in \mathcal{D}(\mathcal{X} \otimes \mathcal{Y})$  is *separable* if

$$\rho \in \text{conv}(\{\sigma_{\mathcal{X}} \otimes \sigma_{\mathcal{Y}} : \sigma_{\mathcal{X}} \in \mathcal{D}(\mathcal{X}), \sigma_{\mathcal{Y}} \in \mathcal{D}(\mathcal{Y})\}).$$

Let  $\text{Sep} \stackrel{\text{def}}{=} \{ \text{separable states} \}$ .

## Definition ( $h_{\text{Sep}}$ )

Given a Hermitian operator  $M$  over  $\mathcal{X} \otimes \mathcal{Y}$ ,

$$h_{\text{Sep}}(M) := \max_{\rho \in \text{Sep}} \text{Tr}[M\rho].$$

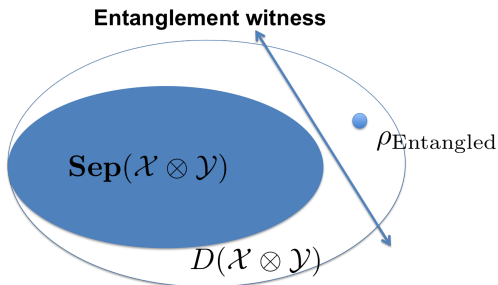
The problem  $\text{WOpt}(M, \epsilon)$  is to compute an  $\epsilon$ -additive approximation of  $h_{\text{Sep}}(M)$ .

# Entanglement Testing

## Definition (Weak Membership)

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

- ▶ By “GLS theorems” in convex optimization, can solve  $\text{WMem}(\epsilon, \|\cdot\|_1)$  efficiently with an oracle to  $\text{WOpt}(M, \epsilon)$ . (Uses ellipsoid method)
- ▶ Any  $M$  with  $h_{\text{Sep}}(M) \leq \lambda_{\max}(M)$  is an *entanglement witness*.



# Applications

- ▶ Finding the maximum acceptance probability of a QMA(2) machine  $\Leftrightarrow$  solving  $\text{WMem}(\epsilon, \|\cdot\|_1)$  for  $M$  arising from poly-time quantum circuit.
- ▶ Many more (mean-field approximations, output entropies of channels, etc.)

# Algorithms and Hardness

- ▶ Complexity is a function of *dimension*  $n$  and *accuracy*  $\epsilon$ .
- ▶ Algorithms based on Sum-of-Squares [DPS] and  $\epsilon$ -nets, [SW, BH, ...].
- ▶ **When**  $\epsilon = \Omega(1)$ :
  - ▶ Need time  $n^{\Omega(\log n)}$  assuming Exponential-Time Hypothesis [HM]. (Unconditionally for SoS hierarchy [HNW]).
  - ▶ Exists  $n^{O(\log(n)/\epsilon^2)}$  algorithm for **1-LOCC M** [BCY, BH].
- ▶ **When**  $\epsilon = 1/\text{poly}(n)$ :
  - ▶  $\text{WOpt}(\cdot, \epsilon)$  is **NP-hard**. [Gur, Ioa, Gha]
  - ▶ Exist  $(n/\sqrt{\epsilon})^{O(n)}$  algorithms based on SDPs. [NOP]
- ▶ **What about scaling with  $\epsilon$ ?**

# Why care about $\epsilon$ scaling?

- ▶ Surprising jumps in complexity in high-accuracy regime:
  - ▶ QIP jumps from PSPACE to EXP with inverse doubly exponential  $\epsilon$ . [IKW]
  - ▶ QMA equals PSPACE for inverse-exponential  $\epsilon$ . [FL]
  - ▶ QMA(2) equals NEXP for inverse-exponential  $\epsilon$ . [Per]
- ▶ Better understand “brute-force algorithms” for quantum problems
  - ▶ Ex: **no known algorithm** to approximate entangled value of non-local game to constant factor.
- ▶ Could lead to algorithms that perform **better in practice**

# Results

## Theorem (Main)

There exists an algorithm that estimates  $h_{\text{Sep}(n)}(M)$  to error  $\epsilon$  in time  $\exp(\text{poly}(n)) \text{poly} \log(1/\epsilon)$ . similar for the multi-partite case.

- ▶ One algorithm based on **quantifier elimination**—less conceptually interesting, so skip for this talk.
- ▶ Other algorithm based on **SoS hierarchy with added constraints**, inspired by [Nie, NR].
- ▶ Algorithm is **exact**: only approximation comes from numerical error in SDP solver
- ▶ As a complexity result:

$$\text{QMA}_{\log}(2)[c, s = c-1 / \exp(\exp(n))] \subseteq \text{DTIME}(\exp(\text{poly}(n))).$$

# Comparison with DPS

## Advantages

- ▶ **Primal of SDP** leads to new **monogamy relations** relative to particular observables.
- ▶ **Dual of SDP** leads to a new class of **entanglement witness**.
- ▶ Analog of the exact convergence achievable for discrete optimization, e.g., SoS for Boolean CSPs. (Note: DPS does *not* converge at any finite level)

## Disadvantages

- ▶ Unlike **[DPS]**, set of entanglement witnesses could be *non-convex*.
- ▶ Hence, cannot solve WMem directly (need to use GLS reduction).



# Proof Outline

- ▶ Treat WOpt as a constrained polynomial optimization problem.
- ▶ Use **Karush-Kuhn-Tucker conditions** to restrict to *critical set* (technique from [Nie, NR])
- ▶ **Claim 1:** For generic  $M$ , critical set consists of  $\exp(n)$ -many discrete points. (From Bertini's thm)
- ▶ Algorithm: use SoS to approximately optimize over critical points by searching for **certificates of positivity**.
- ▶ **Claim 2:** For generic  $M$ , optimum has SoS certificate of degree  $\exp(n)$ . (From Claim 1 and Gröbner basis theory)
- ▶ (Continuity argument to handle non-generic  $M$ ).

## Simplification: the symmetric real case

By simple reductions [HM], we can restrict to states of the form  $|\psi\rangle \otimes |\psi\rangle, |\psi\rangle \in \mathbb{R}^n$ .

$$h_{\text{ProdSym}(n)}(M) := \begin{aligned} & \max_{x \in \mathbb{R}^n} && f_0(x) := \sum_{i_1, i_2, j_1, j_2} M_{(i_1, i_2), (j_1, j_2)} x_{i_1} x_{i_2} x_{j_1} x_{j_2} \\ & \text{subject to} && f_1(x) := \|x\|^2 - 1 = 0. \end{aligned} \tag{1}$$

**REMARK:** this is a *polynomial optimization* problem with a homogenous degree 4 objective and a degree 2 constraints.

# 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  $f$ .

# Principle of SoS : constrained domain

## Definition (Variety)

A set  $V \subseteq \mathbb{C}^n$  is called an *algebraic variety* if

$$V = \{x \in \mathbb{C}^n : g_1(x) = \cdots = g_k(x) = 0\}.$$

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

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

**Question:** do all nonnegative polynomials on certain variety have a **SoS certificate**?

# Putinar's Positivstellensatz

## Definition (Ideal)

The *polynomial ideal*  $I$  generated by  $g_1, \dots, g_k \in \mathbb{C}[x_1, \dots, x_n]$  is

$$I = \left\{ \sum a_i g_i : a_i \in \mathbb{C}[x_1, \dots, x_n] \right\} := \langle g_1, \dots, g_k \rangle.$$

## Theorem (Putinar's Positivstellensatz)

*Under the Archimedean condition, if  $f(x) > 0$  on  $V(I) \cap \mathbb{R}^n$ , then*

$$f(x) = \sigma(x) + g(x),$$

*for some choice of SoS polynomial  $\sigma(x)$  and  $g(x) \in I$ .*

# SoS in Optimization

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

is equivalent to (under Archimedean condition)

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

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

## SoS hierarchy

- ▶ If  $\sigma(x)$  and  $b_i(x)$  can have *arbitrarily high* degrees, then the optimization problem (3) is equivalent to problem (2).
- ▶ By bounding the degrees, i.e.,  $\deg(\sigma(x))$ ,  $\deg(b_i(x)g_i(x)) \leq 2D$  for some integer  $D$ , we get **SoS hierarchy at level  $D$** .

$$\begin{aligned} \min \quad & \nu \\ \text{such that} \quad & \nu - f(x) = \sigma(x) + \sum_i b_i(x)g_i(x), \end{aligned} \quad (4)$$

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

**DPS algorithm** is SoS applied to  $h_{\text{Sep}}$

# Why is it an SDP?

## Observation

- ▶ Any  $p(x)$  (of degree  $2D$ )  $= m^T Q m$ , where  $m$  is the vector of monomials of degree up to  $2D$  and  $Q$  is the coefficients.
- ▶  $p(x)$  is a SoS iff  $Q \geq 0$ .

$$\begin{aligned} & \min_{\nu, b_{i\alpha} \in \mathbb{R}} \quad \nu \\ & \text{such that} \quad \nu A_0 - F - \sum_{i\alpha} b_{i\alpha} G_{i\alpha} \geq 0. \end{aligned} \tag{5}$$

**Complexity:**  $\text{poly}(m) \text{poly} \log(1/\epsilon)$ , where  $m = \binom{n+D}{D}$ .



# Karush-Kuhn-Tucker Conditions

For any optimization problem

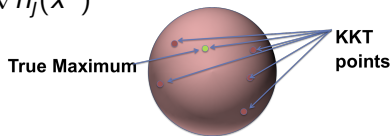
$$\max f(x) \text{ s.t. } g_i(x) \leq 0, h_j(x) = 0, \forall i, j,$$

if  $x^*$  is a *local* optimizer, then  $\exists \mu_i, \lambda_j$ ,

$$\nabla f(x^*) = \sum \mu_i \nabla g_i(x^*) + \sum \lambda_j \nabla h_j(x^*)$$

$$g_i(x^*) \leq 0, h_j(x^*) = 0,$$

$$\mu_i \geq 0, \mu_i g_i(x^*) = 0.$$



**Remark:** for convex optimization (*our case*), any global optimizer satisfies KKT.

## Our case

Recall our optimization problem is

$$\max f_0(x) \text{ s.t. } f_1(x) = 0.$$

The KKT condition is  $\nabla f_0(x) = \lambda \nabla f_1(x)$ , which is equivalent to

$$\text{rank} \begin{pmatrix} \frac{\partial f_0(x)}{\partial x_1} & \frac{\partial f_1(x)}{\partial x_1} \\ \vdots & \vdots \\ \frac{\partial f_0(x)}{\partial x_{2n}} & \frac{\partial f_1(x)}{\partial x_{2n}} \end{pmatrix} < 2.$$

$$g_{ij}(x) := \frac{\partial f_0(x)}{\partial x_i} \frac{\partial f_1(x)}{\partial x_j} - \frac{\partial f_0(x)}{\partial x_j} \frac{\partial f_1(x)}{\partial x_i} = 0, \quad \forall i, j$$

# DPS+: Optimization with KKT constraints

$$\begin{aligned} \min \quad & \nu \\ \text{such that} \quad & \nu - f_0(x) \geq 0 \\ & f_1(x) = 0 \\ \text{KKT} \quad & g_{ij}(x) = 0 \quad \forall 1 \leq i \neq j \leq 2n \end{aligned}$$

- ▶ **DPS+ hierarchy:** SoS applied to above optimization problem
- ▶ Will show finite convergence when  $D = \exp(\text{poly}(n))$ . Then  $m = \binom{n+D}{D} = \exp(\text{poly}(n))$ . Thus the final time is  $\exp(\text{poly}(n)) \text{poly} \log(1/\epsilon)$ .

# KKT Ideal

## Definition (KKT Ideal & Variety)

$$I_K = \left\{ v(x)f_1(x) + \sum h_{ij}(x)g_{ij}(x) \right\} = \langle f_1(x), g_{ij}(x) \rangle .$$

$$V(I_K) = \left\{ x \in \mathbb{C}^{2n} : \forall p(x) \in I_K, p(x) = 0 \right\}$$

## Definition (KKT Ideal to degree $m$ )

$$I_K^m = \left\{ v(x)f_1(x) + \sum h_{ij}(x)g_{ij}(x) : \deg(v(x)f_1(x)) \leq m, \right. \\ \left. \forall i, j, \deg(h_{ij}g_{ij}) \leq m \right\} .$$

# Main Technical Theorems

## Theorem (Zero-dimensionality of generic $I_K$ )

For a generic  $M$ ,  $|V(I_K)| < \infty$  and  $I_K$  is zero-dimensional.

## Theorem (Degree bound)

There exists  $m = O(\exp(\text{poly}(n)))$ , s.t. for a generic  $M$ ,  $\epsilon > 0$ ,

$$v - f_0(x) + \epsilon = \sigma(x) + g(x),$$

where  $\sigma(x)$  is SoS and  $\deg(\sigma(x)) \leq m$ ,  $g(x) \in I_K^m$ .

## Corollary (SDP solution)

For a generic  $M$ ,  $h_{\text{ProdSym}(n)}(M)$  can be estimated to error  $\epsilon$  in time  $\exp(\text{poly}(n))\text{poly} \log(1/\epsilon)$ .

# From generic to arbitrary $M$

## Observations

- ▶ Generic  $M$ s are *dense*, and value of SDP should be a continuous function of  $M$ .
- ▶ Issue: SDP might be *infeasible* for non-generic  $M$ .

## Solution

- ▶ Switch to the dual SDP: satisfies Slater's condition, i.e, strictly feasible.
- ▶ For a generic  $M$ , by strong duality,  
$$h_{\text{ProdSym}(n)}(M) = OPT_{\text{mom}}(M).$$
- ▶ For non-generic inputs  $M$ , use the continuity of the dual SDP.

# Proof of Theorem 1

Let  $\mathcal{U} = \{f_1(x) = 0\}$ ,  $\mathcal{W} = \{\forall i, j, g_{ij} = 0\}$ . then  $V(I_K) \subseteq \mathcal{U} \cap \mathcal{W}$ .  
It suffices to show  $|\mathcal{U} \cap \mathcal{W}| < \infty$ . Construct  $\mathcal{A} = \mathcal{X} \cap \mathcal{U}$  s.t.

$\mathcal{A} \cap \mathcal{W} = \emptyset$  and  $\dim(\mathcal{X}) = n - 1$ . Note  $\mathcal{W} \cap \mathcal{A} = (\mathcal{W} \cap \mathcal{U}) \cap \mathcal{X}$ .  
By Bézout's theorem, two varieties with dimension sum  $\geq n$  must intersect. Thus

$$\dim(\mathcal{W} \cap \mathcal{U}) + \dim(\mathcal{X}) = \dim(\mathcal{W} \cap \mathcal{U}) + n - 1 < n.$$

This implies  $\dim(\mathcal{W} \cap \mathcal{U}) = 0$  and thus  $|V(I_K)| < \infty$ .

## Proof of Theorem 1: construct $\mathcal{X}$

Let  $\mathcal{X} = \{f_0(x) = \mu\}$  for generic  $(\mu, M)$ .  $\dim(\mathcal{X}) = n - 1$ .

By Bertini's theorem,  $\dim(\mathcal{A}) = \dim(\mathcal{U} \cap \mathcal{X}) = n - 2$ .

The Jacobian matrix  $J_{\mathcal{A}} = \begin{pmatrix} \frac{\partial f_0}{\partial x_1} & \frac{\partial f_1}{\partial x_1} \\ \vdots & \vdots \\ \frac{\partial f_0}{\partial x_n} & \frac{\partial f_1}{\partial x_n} \end{pmatrix}$  has  $\text{rank}(J_{\mathcal{A}}) = 2$ .

$\mathcal{W}$  by definition says  $\text{rank}(J_{\mathcal{A}}) = 1$ . Thus no intersection!

**Subtleties:** varieties over **projective space**, so need to consider point at infinity



## Proof of Theorem 2

- ▶ Goal: given SoS certificate

$$\nu - f_0(x) = \sigma(x) + g(x), \sigma(x) \in \text{SoS}, g(x) \in I_K^m,$$

upper bound  $\deg(\sigma(x)), \deg(g(x))$ .

- ▶ Idea: start from any SoS certificate and lower degree of  $\sigma(x)$  using Gröbner basis.
- ▶ Then, show that remaining high-degree terms in  $g(x)$  vanish
- ▶ **Gröbner basis: set of polynomials generating  $I_K$  such that *leading term* of any poly in  $I_K$  is divisible by leading term of a basis element.**
- ▶ [MR]:

$$|V(I_K)| < \infty \implies \max \deg\{\gamma_i\} \leq D = \exp(\text{poly}(n)).$$

## Proof of Theorem 2: SoS term

- ▶ Let  $\{\gamma_i\}$  be a Gröbner basis for  $I_K$ .
- ▶ Take any SoS certificate:

$$v - f_0(x) = \sigma(x) + g(x). \text{ s.t. } \sigma(x) \text{ SoS}, g(x) \in I_K^m.$$

- ▶ Let  $\sigma(x) = \sum s_a(x)^2$ . By properties of Gröbner basis

$$s_a(x) = g_a(x) + u_a(x), \text{ s.t. } g_a(x) \in I_K, \deg(u_a(x)) \leq nD.$$

- ▶ Thus

$$v - f_0(x) = \sigma'(x) + g'(x), \deg(\sigma'(x)) \leq \exp(\text{poly}(n)), g' \in I_K.$$

## Proof of Theorem 2: Ideal term

Suffices to show  $g' \in I_K^m$ ,  $m = \exp(\text{poly}(n))$ .

- ▶ Since  $f_0(x)$  and  $\sigma(x)$  have degree at most  $m$ ,  $\deg(g'(x)) \leq m$ . Remains to show that  $g'(x)$  can be obtained as sum of degree- $m$  multiples of *original* generators  $g_{ij}$ .
- ▶ First step: decompose  $g'(x)$  in Gröbner basis:  
 $g'(x) = \sum t_k \gamma_k(x)$  with  $\deg(t_k \gamma_k(x)) \leq m$ .
- ▶ Second step: decompose Gröbner basis in terms of original generators:  $\gamma_k(x) = \sum u_{ij}(x) g_{ij}(x)$  with  $\deg(u_{ij}) \leq m$ .
- ▶ Thus

$$g'(x) = \sum t_k u_{ij} g_{ij}(x), \deg(t_k u_{ij}) \leq m, \implies g'(x) \in I_K^m.$$

# Numerical Results

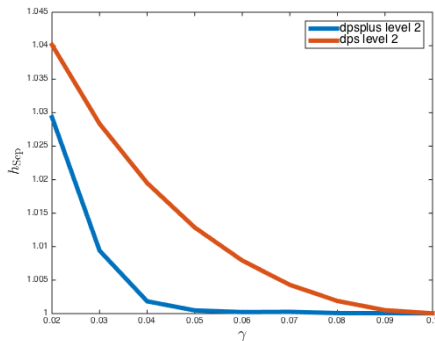
DPS+ beats DPS at very high levels, but what about **low levels** of the hierarchy?

- ▶ DPS+ beats DPS at level 2,  $n = 3$  for a family of measurements introduced by [\[DPS\]](#).

$$M_\gamma = \text{Id} - (A_\gamma \otimes \text{Id})Z(A_\gamma \otimes \text{Id})$$

$$A_\gamma = \text{diag}(1, 1/\gamma, \dots, 1/\gamma).$$

- ▶ For all  $\gamma \in [0, 1]$ ,  
 $h_{\text{Sep}}(M_\gamma) = 1$ .



# Open Questions

## DPS+

- ▶ Analyze the low levels of DPS+.
- ▶ Advantages of adding KKT conditions other than presented here.

## Nonlocal games

- ▶ What is the correct version of KKT conditions for non-commutative polynomials?
- ▶ Can we have finite convergence for the commuting-operator game value? Would imply an upper bound on commuting-operator version of MIP\* (*none known so far*)

## SoS hierarchy

- ▶ Any other applications to quantum information?

Thank you!  
Q & A