

Nets vs hierarchies for hard optimization problems

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outline

1. separable states and operator norms
2. approximating the set of separable states
3. approximating general operator norms
4. the simple case of the simplex

entanglement and optimization

Definition: ρ is separable (i.e. not entangled) if it can be written as

$$\rho = \sum_i p_i |v_i\rangle\langle v_i| \otimes |w_i\rangle\langle w_i|$$

probability
distribution

unit vectors

$$\begin{aligned} \text{Sep} &= \text{conv}\{|v_i\rangle\langle v_i| \otimes |w_i\rangle\langle w_i|\} \\ &= \text{conv}\{\rho \otimes \sigma\} \\ &= \end{aligned}$$

Weak membership problem: Given ρ and the promise that $\rho \in \text{Sep}$ or ρ is far from Sep, determine which is the case.

$$\text{Optimization: } h_{\text{Sep}}(M) := \max \{ \text{tr}[M\rho] : \rho \in \text{Sep} \}$$

operator norms

$X:A \rightarrow B$

$$\|X\|_{A \rightarrow B} = \sup \|Xa\|_B / \|a\|_A$$

operator norm

Examples

$l_2 \rightarrow l_2$ largest singular value

$l_\infty \rightarrow l_1$ MAX-CUT = $\max\{\langle \text{vec}(X), a \otimes b \rangle : \|a\|_\infty, \|b\|_\infty \leq 1\}$

$l_1 \rightarrow l_\infty$ $\max_{i,j} |X_{i,j}| = \max\{\langle \text{vec}(X), a \otimes b \rangle : \|a\|_1, \|b\|_1 \leq 1\}$

$S_1 \rightarrow S_1$
of $X \otimes \text{id}$ channel distinguishability
(cb norm, diamond norm)

$S_1 \rightarrow S_p$ max output p-norm, min output Rényi-p entropy

$l_2 \rightarrow l_4$ hypercontractivity, small-set expansion

$S_1 \rightarrow S_\infty$ $h_{\text{sep}} = \max\{\langle \text{Choi}(X), a \otimes b \rangle : \|a\|_{S_1}, \|b\|_{S_1} \leq 1\}$

complexity of h_{Sep}

$h_{\text{Sep}}(M) \pm 0.1 \|M\|_{2 \rightarrow 2}$ at least as hard as

- planted clique [Brubaker, Vempala '09]
- 3-SAT $[\log^2(n) / \text{polyloglog}(n)]$ [H, Montanaro '10]

$h_{\text{Sep}}(M) \pm 100 h_{\text{Sep}}(M)$ at least as hard as

- small-set expansion [Barak, Brandão, H, Kelner, Steurer, Zhou '12]

$h_{\text{Sep}}(M) \pm \|M\|_{2 \rightarrow 2} / \text{poly}(n)$ at least as hard as

- 3-SAT $[n]$ [Gurvits '03], [Le Gall, Nakagawa, Nishimura '12]

complexity of $l_2 \rightarrow l_4$ norm

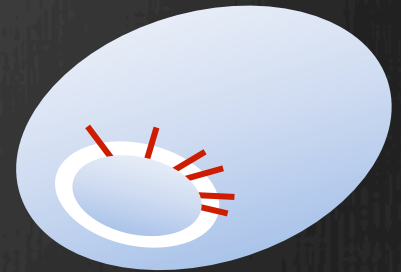
Unique Games (UG):

Given a system of linear equations: $x_i - x_j = a_{ij} \pmod k$.
Determine whether $\geq 1-\epsilon$ or $\leq \epsilon$ fraction are satisfiable.

Small-Set Expansion (SSE):

Is the minimum expansion of a set with $\leq \delta n$ vertices $\geq 1-\epsilon$ or $\leq \epsilon$?

UG \approx SSE $\leq 2 \rightarrow 4$



G = normalized adjacency matrix
 P_λ = largest projector s.t. $G \geq \lambda P$

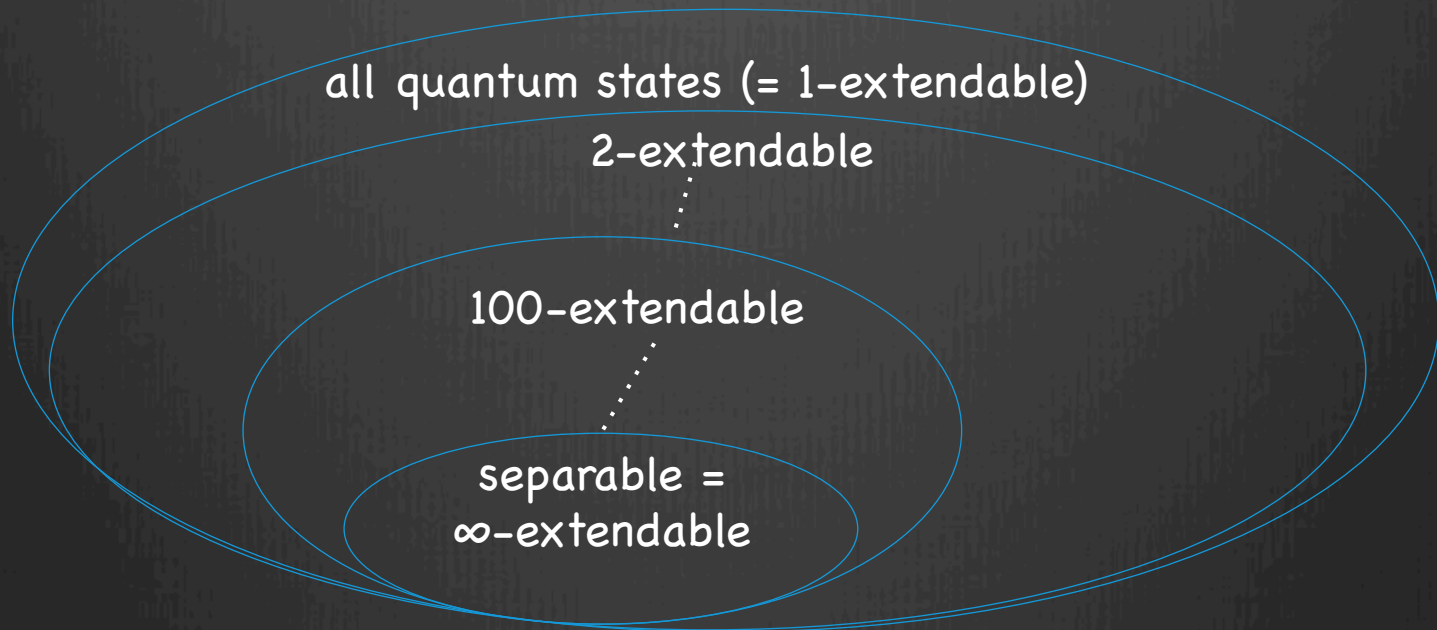
Theorem:

All sets of volume $\leq \delta$ have expansion $\geq 1 - \lambda^{O(1)}$
iff

$$\|P_\lambda\|_{2 \rightarrow 4} \leq n^{-1/4} / \delta^{O(1)}$$

A hierarchy of tests for entanglement

Definition: ρ^{AB} is **k-extendable** if there exists an extension $\rho^{AB_1 \dots B_k}$ with $\rho^{AB} = \rho^{AB_i}$ for each i .



Algorithms: Can search/optimize over k-extendable states in time $n^{O(k)}$.

Question: How close are k-extendable states to separable states?

SDP hierarchies for h_{Sep}

$$\text{Sep}(n,m) = \text{conv}\{\rho_1 \otimes \dots \otimes \rho_m : \rho_m \in \mathcal{D}_n\}$$

$$\text{SepSym}(n,m) = \text{conv}\{\rho^{\otimes m} : \rho \in \mathcal{D}_n\}$$

bipartite

doesn't match hardness

Thm: If $M = \sum_i A_i \otimes B_i$ with $\sum_i |B_i| \leq I$, each $|A_i| \leq I$, then
 $h_{\text{Sep}(n,2)}(M) \leq h_{k\text{-ext}}(M) \leq h_{\text{Sep}(n,2)}(M) + c (\log(n)/k)^{1/2}$

[Brandão, Christandl, Yard '10], [Yang '06], [Brandão, H '12], [Li, Winter '12]

multipartite

$$M = \sum_{i_1, \dots, i_m} c_{i_1, \dots, i_m} A_{i_1}^{(1)} \otimes \dots \otimes A_{i_m}^{(m)} \quad \sum_i |A_i^{(j)}| \leq I \quad |c_{i_1, \dots, i_m}| \leq 1$$

Thm:

ε -approx to $h_{\text{SepSym}(n,m)}(M)$ in time $\exp(m^2 \log^2(n)/\varepsilon^2)$.

ε -approx to $h_{\text{Sep}(n,m)}(M)$ in time $\exp(m^3 \log^2(n)/\varepsilon^2)$.

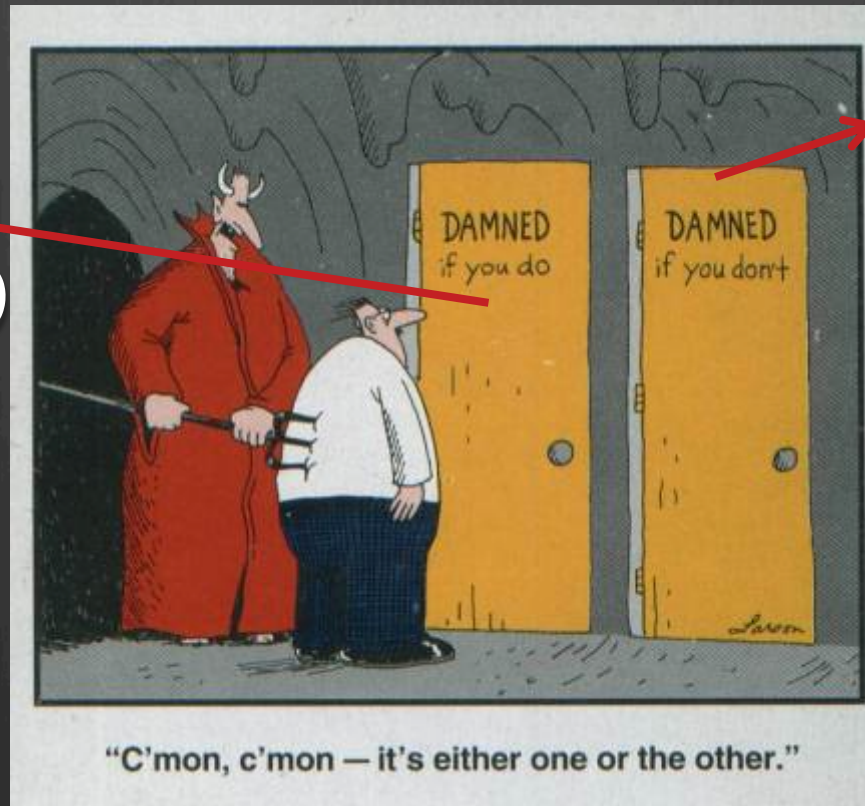
[Brandão, H '12], [Li, Smith '14]

≈ matches Chen-Drucker hardness

proof intuition

Measure extended state and get outcomes $p(a, b_1, \dots, b_k)$.
Possible because of **1-LOCC** form of M.

case 1
 $p(a, b_1) \approx p(a) \cdot p(b_1)$



case 2
 $p(a, b_2 | b_1)$
has less mutual
information

questions

- ⊗ Run-time $\exp(c \log^2(n) / \epsilon^2)$ appears in both
 - ⊗ Algorithm for M in 1-LOCC
 - ⊗ Hardness for M in SEP.

Why? Can we bridge the gap?

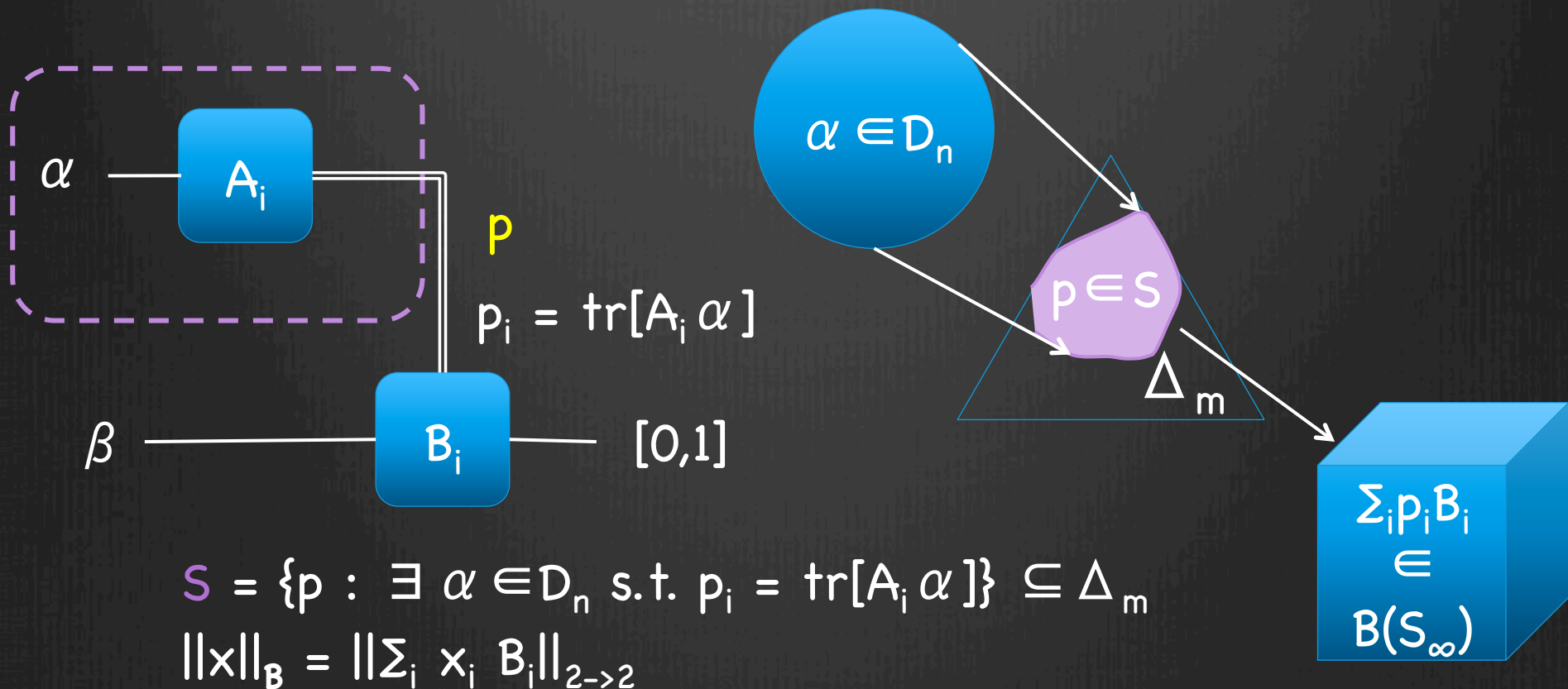
- ⊗ Can we find multiplicative approximations, or otherwise use these approaches for SSE?

net-based algorithms



$M = \sum_{i \in [m]} A_i \otimes B_i$ with $\sum_i A_i \leq I$, each $|B_i| \leq I$, $A_i \geq 0$
 Hierarchies estimate $h_{\text{sep}}(M) \pm \varepsilon$ in time $\exp(\log^2(n)/\varepsilon^2)$

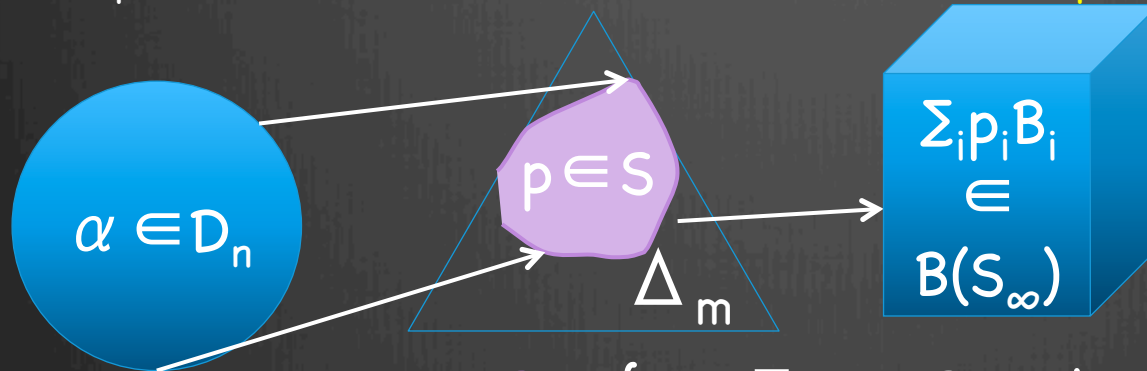
$$h_{\text{sep}}(M) = \max_{\alpha, \beta} \text{tr}[M(\alpha \otimes \beta)] = \max_{p \in \mathcal{S}} \|p\|_{\mathcal{B}}$$



net-based algorithms



$$h_{\text{Sep}}(M) = \max_{\alpha, \beta} \text{tr}[M(\alpha \otimes \beta)] = \max_{p \in S} \|p\|_B$$

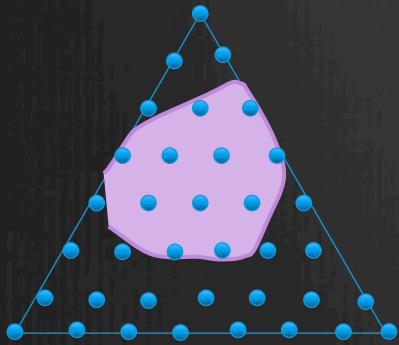


$$\|x\|_B = \|\sum_i x_i B_i\|_{2 \rightarrow 2}$$

$$S = \{p : \exists \alpha \in D_n \text{ s.t. } p_i = \text{tr}[A_i \alpha]\}$$

Lemma: $\forall p \in \Delta_m \exists q$ k -sparse (i.e. $\in \mathbb{Z}^m/k$) s.t.
 $\|p - q\|_B \leq c(\log(n)/k)^{1/2}$

Pf: matrix Chernoff [Ahlsvede-Winter]



Algorithm:

Enumerate over k -sparse q

- check whether $\exists p \in S, \|p - q\|_B \leq \epsilon$
- if so, compute $\|q\|_B$

Performance

$k \approx \log(n) / \epsilon^2, m = \text{poly}(n)$

run-time

$O(m^k) = \exp(\log^2(n) / \epsilon^2)$

nets for Banach spaces



$X:A \rightarrow B$

$$\|X\|_{A \rightarrow B} = \sup \|Xa\|_B / \|a\|_A$$

operator norm

$$\|X\|_{A \rightarrow C \rightarrow B} = \min \{ \|Z\|_{A \rightarrow C} \|Y\|_{C \rightarrow B} : X = YZ \}$$

factorization norm

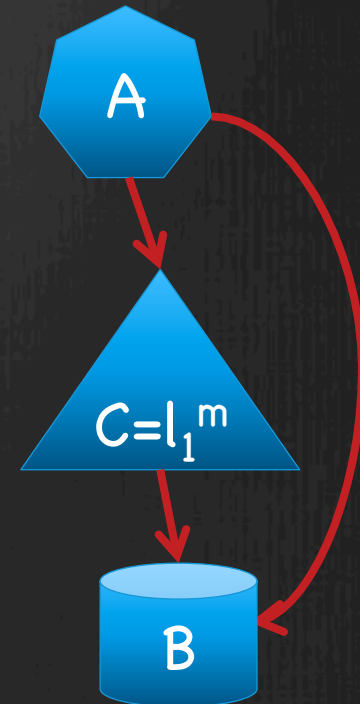
Let A, B be arbitrary. $C = \ell_1^m$

Only changes are sparsification (cannot assume $m \leq \text{poly}(n)$) and operator Chernoff for B .

Type-2 constant: $T_2(B)$ is smallest λ such that

$$\mathbb{E}_{\epsilon_1, \dots, \epsilon_n \in \{\pm 1\}} \left\| \sum_{i=1}^n \epsilon_i Z_i \right\|_B^2 \leq \lambda^2 \sum_{i=1}^n \|Z_i\|_B^2$$

result: $\|X\|_{A \rightarrow B} \pm \epsilon \|X\|_{A \rightarrow \ell_1^m \rightarrow B}$
estimated in time $\exp(T_2(B)^2 \log(m) / \epsilon^2)$



applications

$S_1 \rightarrow S_p$ norms of **entanglement-breaking** channels

$N(\rho) = \sum_i \text{tr}[A_i \rho] B_i$, where $\sum_i A_i = I$, $\|B_i\|_1 = 1$.

Can estimate $\|N\|_{1 \rightarrow p} \pm \epsilon$ in time $n^{O(c)}$ where

$$c = p / \epsilon^2 \quad \text{for } p \geq 2$$

$$c = (p / \epsilon^p)^{1/(p-1)} \quad \text{for } 1 < p < 2$$

(uses bounds on $T_2(S_p)$ from [Ball-Carlen-Lieb '94])

low-rank measurements:

$h_{\text{sep}}(\sum_i A_i \otimes B_i) \pm \epsilon$ for

$\sum_i |A_i| = 1$, $\|B_i\|_\infty \leq 1$, **rank $B_i \leq r$**

in time $n^{O(r/\epsilon^2)}$

$l_2 \rightarrow l_p$ for even $p \geq 4$

$$\|X\|_{2 \rightarrow p}^p \pm \epsilon \|X\|_{2 \rightarrow 2}^2 \|X\|_{2 \rightarrow \infty}^{p-2}$$

in time $n^{O(p/\epsilon^2)}$

Multipartite versions of 1-LOCC norm too [cf. Li-Smith '14]

ϵ -nets vs. SoS

Problem	ϵ -nets	SoS/info theory
$\max_{p \in \Delta} p^T A p$	BK '02, KLP '06	DF '80 BK '02, KLP '06
approx Nash	LMM '03	HNW '16
free games	AIM '14	BH '13
unique games	ABS '10	BRS '11
small-set expansion	ABS '10	BBHKSZ '12
h_{Sep}	SW '11 BH '15	BCY '10 BH '12 BKS '13

simplest version: polynomial optimization over the simplex

$$\Delta_n = \{p \in \mathbb{R}^n : p \geq 0, \sum_i p_i = 1\}$$

Given homogenous degree- d poly $f(p_1, \dots, p_n)$, find $\max_p f(p)$.

NP-complete: given graph G with clique number α ,
 $\max_p p^T A p = 1 - 1/\alpha$. [Motzkin-Strauss, '65]

Approximation algorithms

- Net: Enumerate over all points in $\Delta_n(k) := \Delta_n \cap \mathbb{Z}^n/k$.
- Hierarchy: $\min \lambda$ s.t. $(\sum_i p_i)^k (\lambda (\sum_i p_i)^d - f(p))$ has all nonnegative coefficients.

Thm: Each gives error $\leq (\max_p f(p) - \min_p f(p)) \exp(d) / k$
in time $n^{O(k)}$. [de Klerk, Laurent, Parrilo, '06]

sum-of-squares (SoS) proofs

Axioms:

$$g_1(x) \geq 0$$

⋮

$$g_m(x) \geq 0$$



$$f(x) \leq \lambda$$

Rules:

1. polynomial operations
 2. intermediate polys have $\text{deg} \leq k$
 3. [optional: changes LP to SDP]
- $r(x)^2 \geq 0$ for any polynomial $r(x)$

“Positivstellensatz” [Stengel '74]

hierarchies & SoS proofs

Given axioms: $\sum_i p_i = 1$ and $p_i \geq 0$
prove that $\lambda - f(p) \geq 0$.

Previous strategy:

$$\lambda (\sum_i p_i)^d - f(p) = (\sum_i p_i)^k (\lambda (\sum_i p_i)^d - f(p)) \geq 0$$

difference is divisible
by $1 - \sum_i p_i$

LHS is nonnegative sum
of products of p_i

Dual is equivalent to net enumeration for modified
objective function.

[Bomze, de Klerk '02] [de Klerk, Laurent, Sun '14]

k-extendable hierarchy

For a deg- d homogenous poly $f(p)$, define $\text{vec}(f) \in (\mathbb{R}^n)^{\otimes d}$ to be the symmetric tensor such that $f(x) = \langle \text{vec}(f), x^{\otimes d} \rangle$.

Then $\max_p f(p) = h_K(\text{vec}(f))$ for

$$K = \text{conv}\{p^{\otimes d} : p \in \Delta_n\}$$

$$h_K(y) := \max_{x \in K} \langle x, y \rangle$$

relaxation:

$q \in \Delta_{nd+k}$ symmetric (aka "exchangeable")

$$\pi = q^{(1,2,\dots,d)}$$

convergence: [Diaconis, Freedman '80]

$$\text{dist}(\pi, \text{conv}\{p^{\otimes d}\}) \leq O(d^2/k)$$

→ error $\|\text{vec}(f)\|_\infty / k$ in time $n^{O(k)}$

Nash equilibria

Non-cooperative games:

Players choose strategies $p^A \in \Delta_m$, $p^B \in \Delta_n$.

Receive values $\langle V_A, p^A \otimes p^B \rangle$ and $\langle V_B, p^A \otimes p^B \rangle$.

Nash equilibrium: neither player can improve own value
 ϵ -approximate Nash: cannot improve value by $> \epsilon$

Correlated equilibria:

Players follow joint strategy $p^{AB} \in \Delta_{mn}$.

Receive values $\langle V_A, p^{AB} \rangle$ and $\langle V_B, p^{AB} \rangle$.

Cannot improve value by unilateral change.

- Can find in $\text{poly}(m,n)$ time with LP.
- Nash equilibrium = correlated equilibrium with $p = p^A \otimes p^B$

finding (approximate) Nash eq

Known complexity:

Finding exact Nash eq. is PPAD complete.

Optimizing over exact Nash eq is NP-complete.

Algorithm for ε -approx Nash in time $\exp(\log(m)\log(n)/\varepsilon^2)$
based on enumerating over nets for Δ_m, Δ_n .

Planted clique and 3-SAT[$\log^2(n)$] reduce to optimizing
over ε -approx Nash.

[Lipton, Markakis, Mehta '03], [Hazan-Krauthgamer '11], [Braverman, Ko, Weinstein '14]

New result [HNW16]: Another algorithm for finding
 ε -approximate Nash with the same run-time.

(uses k -extendable distributions)

algorithm for approx Nash

Search over $p^{AB_1 \dots B_k} \in \Delta_{mn^k}$
such that the $A:B_i$ marginal is a correlated equilibrium
conditioned on any values for B_1, \dots, B_{i-1} .

LP, so runs in time $\text{poly}(mn^k)$

Claim: Most conditional distributions are \approx product.

Proof:

$$\log(m) \geq H(A) \geq I(A:B_1 \dots B_k) = \sum_{1 \leq i \leq k} I(A:B_i | B_{<i})$$

$$\mathbb{E}_i I(A:B_i | B_{<i}) \leq \log(m)/k =: \varepsilon^2$$

$\therefore k = \log(m)/\varepsilon^2$ suffices.

open questions

- Application to unique games, small-set expansion, etc. Which norms are the right ones here?
- Tight hardness results, e.g. for h_{Sep} .
- Explain the coincidences!