

Lower Bounds
Against the
Sum of Squares Algorithm

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Agenda

1. What is SoS?
2. SoS, learning, and crypto
3. Lower bounds: what we know
4. Lower bounds: what we hope for

What is SoS?

Alg. for deciding:

Does $p_1(x_1 \dots x_n) \geq 0, \dots, p_m(x_1 \dots x_n) \geq 0$

have a solution in \mathbb{R}^n ?

[Parrilo '01, Lasserre '01]

Why Polynomial Systems?

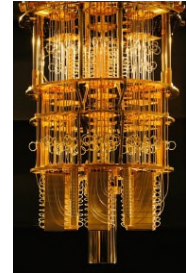
Polynomials are highly expressive



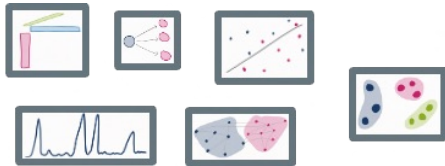
Combinatorial optimization



robotics /
optimal control



quantum info.



Statistics/learning/ avg. case algos



Crypto

What is SoS?

Degree- d SoS proof that

$$p_1(x) \geq 0 \dots p_m(x) \geq 0 : \{q_s = \sum_{S \subseteq [m]} q_{s,i}(x)^2\}$$

$$-1 = \sum_{S \subseteq [m]} q_s(x) \cdot \prod_{j \in S} p_j(x)$$

degree $\leq d$

What is SoS?

- Hilbert, ..., Krivine-Stengle:

every unsatisfiable $p_1 \succcurlyeq 0 \dots p_m \succcurlyeq 0$

has an SoS proof (!)

- Parrilo - Lasserre: can decide existence

of degree d proof in $(n \cdot m)^{O(d)}$ time

What is SoS?

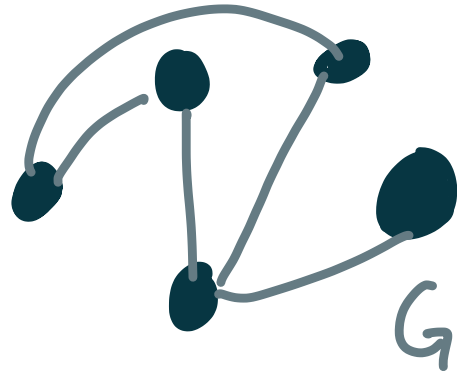
Proving power vs. running time
tradeoff

Seems to capture best known alqs
for many problems (sometimes
provably optimal)

Ex: $\text{min-cut}(G) = c$



degree $O(1)$ SoS proof
that no cut $< c$.



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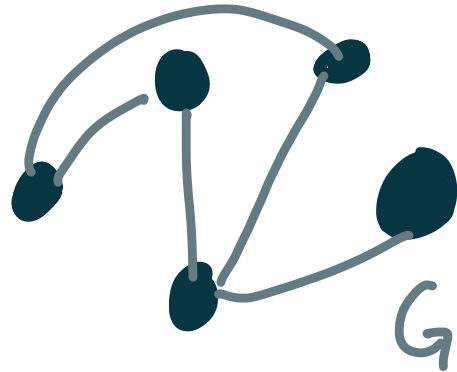
Ex: $\text{max-cut}(G) = c$

\Downarrow

degree-2 SoS proof

that no cut $> \frac{c}{0.878}$

[Goemans - Williamson '95]



Seems to capture best known alqs
for many problems (sometimes
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Ex: $\max\text{-cut}(G) = c$
 \Downarrow
degree-2 SoS proof
that no cut $> \frac{c}{0.878}$

[Goemans - Williamson '95]

Optimal under UGC
generalizes to
all CSPs!



[KKMO, Raghavendra]

SoS + Avg. Case / Stat. inference

SoS gives best known algorithms for canonical avg. case problems

- planted clique
- refuting random CSPs
- ind. set in $G(n, p)$
- etc.

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Aside: recent revolution in

SoS for high-dimensional

Statistics:

- robust stats
- latent variable models
- tensor problems, & more

Example: Planted Clique

Goal: distinguish $G(n, 1/2)$ from $G(n, 1/2)$
+ k -clique

$k \geq \Omega(\sqrt{n})$: poly-time algs known [AKS '97]

$k \ll \sqrt{n}$: conjecturally hard

many consequences, incl. private-key crypto

[Juels-Peinado '00]

SoS to detect a k -clique:

Given G , check if \exists SoS proof of:

"all cliques in G have $< k$ vertices"

vars x_1, \dots, x_n
 $x_i^2 = x_i$
 $x_i x_j = 0 \quad \forall i \neq j$
 $\sum x_i = k$

look for SoS proof
of unsatisfiability

Fact: degree-d proof exists for $k \approx \frac{\sqrt{n}}{2^d}$
 proof for $d=2$:

$x_i^2 = x_i$
 $x_i x_j = 0 \quad \forall i \neq j$
 $\sum x_i = k$

(w.h.p.)

① $k^2 = (\sum x_i)^2 = \sum x_i x_j$

$= \sum_{i \sim j} x_i x_j - \sum_{i \neq j} x_i x_j + \sum_{i \neq j} x_i x_j = x^T \overset{\text{centered adj. matrix}}{A} x + \sum_{i \neq j} x_i x_j$

② $x^T A x = \underbrace{\sqrt{n} \|x\|^2}_{\text{eigenvalue bound on } A} - \sum \langle x, a_i \rangle^2 = \sqrt{n} k - \sum \langle x, a_i \rangle^2 - \sqrt{n} \sum x_i - x_i^2 - \sqrt{n} (k - \sum x_i)$

①+②: $\sqrt{n} k - k^2 = \sum \langle x, a_i \rangle^2 + \text{pink stuff}$ ■

SoS + Crypto

Before Jain-Lin-Sahai iO construction,

Many iO candidates : Structured PRGs \Rightarrow iO

Lin-Tessaro,

Ananth-Jain-Sahai

Lin-Matt,

Agrawal

BUT: Structured PRG candidates

REPEATEDLY BROKEN

[Barak-Brakerski-Komargodski-Kothari '17, Barak-H.-Jain-Kothari-Sahai '18]

In light of algorithmic power of SoS
why lower bounds?

narrow view: good to know when SoS algs
attacks don't work

broad view: lens on hardness, esp. for
avg. case. } "SoS is optimal"
(??)

hardness in stats → source for crypto?

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Today: focus on average case

- CSPs

- Planted Clique

- Boolean vec. in random subspace

- Ind. set in sparse graphs

- Sparse PCA

- Tensor PCA

[KMOW16, BHKKMP16, GTJPR20, JPRTX21, HKPRSS'17, P'21]

Ex. Planted Clique

Thm. [BHKMP+ Peng]: There is no degree

$d = o(\log n)$ SoS refutation of

$$x_i^2 = x_i, \quad x_i x_j = 0 \text{ for } i \neq j, \quad \sum x_i = n^{\frac{1}{2}} - o(1)$$

W.h.p for $G \sim G(n, \frac{1}{2})$.

Beyond poly-time \rightarrow Thm. [BHKMP+ Peng]: There is no degree $d = o(\log n)$ SoS refutation of $x_i^2 = x_i$, $x_i x_j = 0$ for $i \sim_G j$, $\sum x_i = n^{\frac{1}{2} - o(1)}$ w.h.p for $G \sim G(n, \frac{1}{2})$.

Suggests refuting is hard (for SoS?) in poly time.

Often interpreted to mean distinguishing hard

How to prove an SoS lower bd.

$$-1 = \sum_{S \subseteq [m]} q_S(x) \prod_{j \in S} p_j(x)$$

$$\sum x_i = n^{\frac{1}{2} - \alpha(1)}$$

$$x_i^2 = x_i$$

$$x_i x_j = 0 \text{ for } i \neq j$$

Necessary & Sufficient: pseudoeexpectation

$$\tilde{\mathbb{E}} : \mathbb{R}[x]_{\leq d} \rightarrow \mathbb{R}$$

$$\tilde{\mathbb{E}} 1 = 1$$

$$\tilde{\mathbb{E}} p^2 \geq 0$$

$$\tilde{\mathbb{E}} p(x) (x_i^2 - x_i) = 0, \forall p$$

$$\tilde{\mathbb{E}} p(x) (\sum x_i - k) = 0, \forall p$$

$$\tilde{\mathbb{E}} p(x) x_i x_j = 0, \forall p$$

Need map $G \rightarrow \tilde{E}_G$

Pseudocalibration: construction of
Such a map from hard distinguishing
problem e.g. $G(n, 1/2)$ vs $G(n, 1/2)$
 H_0 + k -clique

Idea: low-degree moments of H_1
 H_1 define \tilde{E}

All known avg. case SoS l.b.'s
can (or should be) proved via
Pseudocalibration

- CSPs
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Analysis remains

Case-by-case:

- Random matrices
- random graphs,
expansion

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numerous open
problems

- understand
random matrices
w/ dependent
entries

3. Lower bounds: what we know

- a bunch of l.b.'s for avg-case refutation problems, with matching algos
- a canonical technique which we still struggle to analyze

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Things we don't understand

Refutation vs Distinguishing

For stats & crypto, usually want
hardness for distinguishing 2 distn's

SoS natively does refutation

How do we interpret SoS l.b.'s for distinguishing?

Ex. Planted Clique

Could try to refute

$$x_i^2 = x_i, \quad x_i x_j = 0 \text{ for } i \neq j, \quad \sum x_i = k,$$

$$\text{and } p(G, x) \gg 0$$

for any low-degree p s.t. $p(G, x) \gg 0$

whp for $G, x \sim G(n, 1/2) + k\text{-clique}$

Ex. Distinguishing random CSPs from
CSPs w/ planted solutions

• Usual refutation problem:

for random CSP, refute existence of
satisfying assignment

• Kothari - O'Donnell - Schramm: balanced assign.?

$$\sum x_i = 0$$

A very partial resolution:
refutation lower bounds proved using
pseudocalibration

"should not" have this problem

\tilde{E} from pseudocalibration should rule out SoS
for all low-degree refutation problems derived
from a distinguishing problem

The Role of Noise

(More) serious roadblock to using SoS to understand complexity of avg-case problems:

There are SoS lower bounds for easy problems

!!

Ex. 3XOR is easy by Gaussian elimination

BUT: degree $\Omega(n)$ SoS cannot refute random
3XOR!

SoS is inherently noise-robust

⇒ cannot accurately capture the complexity of "non-robust" problems

Can hope to use SOS lower bounds to reason about complexity of distinguishing problems when:

- all corresponding refutation problems have suitable SOS l.b.'s
- everything is a little noisy

4. Lower bounds: what we hope for

Low-Degree Conjecture: H_0 vs H_1 on $\{\pm 1\}^{\binom{n}{k}}$

H_0 product, H_1 S_n -Symmetric.

Not distinguishable via $\omega(\log n)$ moments



H_0 vs $\underbrace{\tilde{H}_1}_{\text{Noisy } H_1}$ not distinguishable in poly time

4. Lower bounds: what we hope for

SoS Low-Degree Conjecture: H_0 vs H_1 on $\{\pm 1\}^{\binom{n}{k}}$

H_0 product, H_1 S_n -Symmetric.

Not distinguishable via $\omega(\log n)$ moments



H_0 vs \tilde{H}_1 not distinguishable by $O(1)$ -deg.

SoS

A few open problems:

- SoS low degree conjecture
Special case: "Kesten-Stigum" threshold
for sparse graph models
- Non-product distributions
important for crypto applications
- Random matrices with dependent entries