

Lecture 20: Proofs to algorithms and the sum-of-squares SDP.

How can we turn a proof of existence into an efficient algorithm for finding the thing?

In the worst case, this is impossible.

- some proofs are non-constructive
i.e. proof by contradiction
- other proofs are too complicated, and it's unclear how to extract an efficient (i.e. polynomial time) algorithm from it.

This class: a way to extract algorithms from structured proofs
"sum-of-squares" proofs.

will generalize several things we've seen so far:

- Goemans-Williamson max-cut
- Cheeger's inequality.

In particular, how to prove non-negativity of polynomials.

Q: how to show that

$$p(x) = x^4 + 12x^3 + 55x^2 - 94x \geq -130 \text{ for all } x?$$

One way: write

$$p(x) - 130 = \underbrace{(x-7)^2 + (x+3)^4}_{\text{sum of squares } \geq 0}.$$

$$\Rightarrow p(x) - 130 \geq 0 \Rightarrow p(x) \geq -130 \quad \forall x.$$

This is a way to certify non-negativity of a polynomial!

In 1-dimension, not such an interesting task perhaps.

But much more interesting in high dimensions.

Let $p: \mathbb{R}^n \rightarrow \mathbb{R}$ be a polynomial in n variables.

We say it has degree d if every monomial in p has total degree $\leq d$.

e.g. $x_1^2 x_2 x_3^5$ has degree 8

$x_1 x_2$ has degree 2

Def: Let $p: \mathbb{R}^n \rightarrow \mathbb{R}$ be a degree d polynomial. We say

$p(x) \geq 0$ has a sum-of-squares (sos) proof of degree k

if there exist polynomials s_1, \dots, s_t of degree $\leq k/2$

s.t. $p(x) = \sum_{i=1}^t s_i(x)^2$.

We say $p(x) \geq q(x)$ has an sos proof of deg k

if $p(x) - q(x) \geq 0$ has an sos proof of deg. k .

Fact: when $n=1$ (i.e. in 1 dimension), $p(x) \geq 0 \Leftrightarrow$

there is an sos proof of this, but this is not true for

higher dimensions.

$$M(x,y) = x^4 y^2 + x^2 y^4 - 3x^2 y^2 + 1$$

"Motzkin polynomial".

so sos is not a "complete" proof system.

i.e. it cannot prove all true statements.

examples of multivariate sos proofs:

ex 1 $\langle a, b \rangle \leq \frac{1}{2} \|a\|_2^2 + \frac{1}{2} \|b\|_2^2$ is true in deg 2 sos.

$$\frac{1}{2} \|a\|_2^2 + \frac{1}{2} \|b\|_2^2 - \langle a, b \rangle = \frac{1}{2} \|a - b\|_2^2 = \frac{1}{2} \sum_{i=1}^n (a_i - b_i)^2$$

ex 2 $\langle a, b \rangle^2 \leq \|a\|_2^2 \|b\|_2^2$ ← Cauchy Schwarz!

$$\|a\|_2^2 \cdot \|b\|_2^2 - \langle a, b \rangle^2 = \frac{1}{2} \sum_{ij} (a_i b_j - a_j b_i)^2$$

many "natural" proofs can be captured in sos, e.g.

Cauchy-Schwarz, certain AM-GM, Hölders, etc...

Sos can also capture implications

Def: Let $f_1, \dots, f_m, g_1, \dots, g_k : \mathbb{R}^n \rightarrow \mathbb{R}$ be polynomials.

We say that the axioms $\{f_i = 0\}_{i=1}^m$ and $\{g_i \geq 0\}_{i=1}^k$

sos-imply that $p \geq 0$ in deg k if there exist

polynomials s_1, \dots, s_t s.t. $\deg(s_i) \leq k/2$
 a_1, \dots, a_m $\deg(a_i f_i) \leq k$
 b_1, \dots, b_k $\deg(b_i^2 g_i) \leq k$
 $\forall i$

so that

$$(*) \quad p(x) = \sum_{i=1}^t s_i(x)^2 + \sum_{j=1}^m a_j(x) f_j(x) + \sum_{k=1}^k b_k^2(x) g_k(x).$$

Intuition: suppose x is a point satisfying

$$f_i(x) = 0 \quad \forall i, \quad g_i(x) \geq 0 \quad \forall i.$$

Then (*) implies that $p(x) \geq 0$!

Note: k can be much larger than degree of p !

For conciseness, we let $A = \{f_i = 0\} \cup \{g_i \geq 0\}$,

and if an sos proof exists, we say

$$A \vdash_k P \geq 0.$$

As before, we say $A \vdash_k P \geq q$ if $A \vdash_k P - q \geq 0$.

Fact: SoS proofs compose: if

$A \vdash_{k_1} B$ and $B \vdash_{k_2} C$, then

$$A \vdash_{\max(k_1, k_2)} C.$$

Why do we care about non-negativity of polynomials?

Can encode many hard problems!

e.g. max-cut

$$\max_{x_i \in \{+1, -1\}} \sum_{(i,j) \in E} \frac{1 - x_i x_j}{2}$$

$$\uparrow x_i^2 = 1$$

(for any k)

$$\text{If } \left\{ x_i^2 = 1 \right\}_{i=1}^n \vdash_k \sum_{(i,j) \in E} \frac{1 - x_i x_j}{2} \geq C, \text{ then}$$

we can certify that the value of the optimal cut is $\geq C$.

Can encode many interesting combinatorial optimization problems in this way!

- max-cut
- 3-SAT
- sparsest cut
- ⋮

something I've worked on in the past

Can also encode many interesting learning problems

Thm: Suppose $A = \{f_i = 0\}_{i=1}^m \cup \{g_i \geq 0\}_{i=1}^k$ SOS implies

$p \geq 0$ in deg k . $A \vdash_k p \geq 0$. Then there is an algorithm which runs in time $n^{O(k)}$ which finds this proof.

proof: Let's consider the simple case where $A = \emptyset$, i.e. we want to find an SOS proof that

$p \geq 0$, if a proof exists.

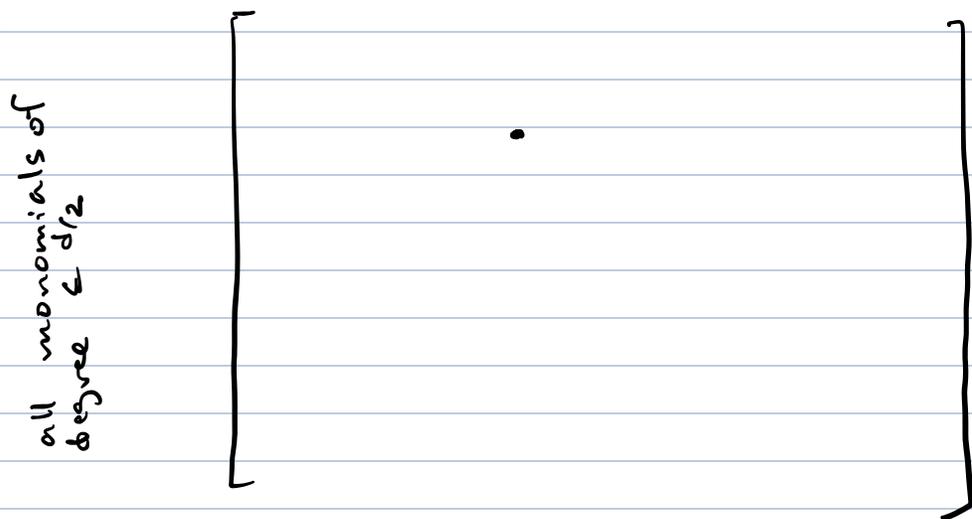
Key idea: monomial matrix.

Any degree d polynomial can be encoded as a $n^{(odd)} \times n^{(odd)}$ size matrix

A deg- m monomial is indexed by a multi-index S w/ $|S| \leq m$.

Shorthand $\left\{ \begin{array}{l} x_1^2, x_3^3, x_{10}^7 \\ \downarrow \\ x^S \end{array} \right. \leftrightarrow (2, 0, 3, \dots, 7, 0, \dots, 0)$
10th position

$\mathcal{M}_q \iff \mathcal{M}_q$
 all monomials of degree $\leq d/2$



Constraint: $\forall T:$
 $\sum_{(R,S): |R|, |S| \leq d/2, R+S=T} (M_q)_{R,S} = \text{coefficient of } x^T \text{ in } q$

Note also vv^T is PSD!

$$\forall x: x^T (vv^T) x = \langle v, x \rangle^2 \geq 0.$$

So a sum of squares corresponds to

v_1, \dots, v_t s.t.

$$M = \sum_{i=1}^t v_i v_i^T = \begin{bmatrix} v \\ \vdots \\ v \end{bmatrix} \begin{bmatrix} v^T \\ \vdots \\ v^T \end{bmatrix}$$

Claim: such a decomposition exists iff M is PSD!

pf: \Leftarrow If M is PSD:

$$M = U^T D U = \underbrace{U^T D^{1/2}}_V \underbrace{D^{1/2} U}_{V^T}$$

allowed LIC
 $D_{ii} \geq 0$

$$\Rightarrow \text{If } M = \sum v_i v_i^T$$

$$x^T M x = \sum \langle x, v_i \rangle^2 \geq 0 \Rightarrow M \succeq 0.$$

So the following SDP finds a sos proof if it exists:

find $M \succeq 0$ s.t. M satisfies the linear constraints

$$\forall T, \sum_{\substack{|R|, |S| \leq d/2 \\ R+S=T}} M_{R,S} = c_T.$$

Ph. D \rightarrow uv^T

Raghavendra's Theorem: The sos hierarchy achieves optimal approximation ratios for almost all "nice"

constraint satisfaction problems, assuming the
unique games conjecture.