EE464 Semialgebraic Lifting

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Primal and Dual Formulations So Far

Positivity of one polynomial: does there exist x such that f(x) < 0?

- ▶ Dual SDP relaxation: *f* is SOS
- Primal SDP relaxation: lifting

Semialgebraic feasibility: does there exist x such that $f_i(x) \ge 0$ and $h_j(x) = 0$ for all i, j

▶ Positivstellensatz is exact dual. Finite degree condition is an SDP: does there exist s_i, r_{ij}, t_i such that s_i, r_{ij} is SOS and

$$-1 = s_0 + \sum_i s_i f_i + \sum_{i,j} r_{ij} f_i f_j + \dots + \sum_i t_i h_i$$

Questions: what is the dual? It should give a *convex relaxation* of the primal feasible set

Valid Inequalities for the Primal

Does there exist $x \in \mathbb{R}^n$ such that $f_i(x) \geq 0$ for all $i = 1, \dots, m$

We can add a *parametrized family* of valid inequalities of the form

$$f_i(x)(a_{00} + a_{10}x + a_{01}y + a_{11}xy + \dots)^2 \ge 0$$
$$(a_{00} + a_{10}x + a_{01}y + a_{11}xy + \dots)^2 \ge 0$$

- ▶ Any vector *a* of coefficients defines a valid inequality
- The multipliers are squares; i.e., extreme rays of the SOS cone The Lagrange duality construction forms linear combinations of these, resulting in a dual with SOS multipliers

Lifting

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We can represent these multipliers as

$$a^T z = \begin{bmatrix} a_{00} & a_{10} & a_{01} & a_{11} & \dots \end{bmatrix} \begin{bmatrix} 1 \\ x \\ y \\ \vdots \end{bmatrix}$$

so an equivalent feasibility problem is: does there exist x such that

$$\begin{aligned} f_i(x)(a^T z)^2 &\geq 0 & \text{ for all } a, i \\ (a^T z)^2 &\geq 0 & \text{ for all } a \end{aligned}$$

now lift; let $Y = zz^T$, then we have

$$(a^T z)^2 = a^T Y a$$

Lifted Problem

The lifted problem is: does there exist $x \in \mathbb{R}^n$ such that

$$a^{T}(f_{i}(x)Y)a \ge 0$$
 for all a, i
 $a^{T}Ya \ge 0$ for all a
 $Y = zz^{T}$

Since Y defines a quadratic form, we have equivalently

$$\begin{aligned} f_i(x)Y \succeq 0 & \text{ for all } i \\ Y \succeq 0 \\ Y = zz^T \end{aligned}$$

Example

suppose $f(x) = x^2 + 3x + 1$; does there exist x such that f(x) < 0? Apply the lifting

$$Y = \begin{bmatrix} 1\\x\\x^2 \end{bmatrix} \begin{bmatrix} 1\\x\\x^2 \end{bmatrix}^T = \begin{bmatrix} 1 & x & x^2\\x & x^2 & x^3\\x^2 & x^3 & x^4 \end{bmatrix}$$

Then

$$f(x)Y = \begin{bmatrix} x^2 + 3x + 1 & x^3 + 3x^2 + x & x^4 + 3x^3 + x^2 \\ & x^4 + 3x^3 + x^2 & x^5 + 3x^4 + x^3 \\ & x^6 + 3x^5 + x^4 \end{bmatrix}$$
$$= \begin{bmatrix} Y_{13} + 3Y_{12} + Y_{11} & Y_{23} + 3Y_{13} + Y_{12} & Y_{33} + 3Y_{23} + Y_{13} \\ & \ddots & & & * \\ & \ddots & & & * \\ & \ddots & & & & * \\ & & & & & & * \end{bmatrix}$$

Primal SDP Relaxation

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Relaxing the constraint $Y = zz^T$, we have the SDP

$$\begin{array}{l} Y \text{ is Hankel} \\ Y_{11} = 1 \\ Z = \begin{bmatrix} Y_{13} + 3Y_{12} + Y_{11} & Y_{23} + 3Y_{13} + Y_{12} \\ Y_{23} + 3Y_{13} + Y_{12} & Y_{33} + 3Y_{23} + Y_{13} \end{bmatrix} \\ Z \succeq 0 \\ Y \succeq 0 \end{array}$$

▶ We have relaxed the valid inequality $f(x)Y \succeq 0$ to positivity of its principal 2×2 submatrix

• We can include as many monomials z as we like

SDP Dual

The SDP dual is: does there exist α, λ, P such that

$$\begin{bmatrix} -\alpha & 0 & -\lambda \\ 0 & 2\lambda & 0 \\ -\lambda & 0 & 0 \end{bmatrix} - \frac{1}{2} \begin{bmatrix} 2P_{11} & 3P_{11} + 2P_{12} & P_{11} + 6P_{12} + P_{22} \\ 0 & 2P_{12} + 3P_{22} \\ P_{22} \end{bmatrix} \succeq 0$$
$$P \succeq 0$$
$$\alpha > 0$$

To interpret this, multiply left and right by \boldsymbol{z}^T and $\boldsymbol{z},$ giving

$$\begin{aligned} -\alpha - (x^2 + 3x + 1)(P_{11} + 2P_{12}x + P_{22}x^2) \text{ is SOS} \\ (P_{11} + 2P_{12}x + P_{22}x^2) \text{ is SOS} \end{aligned}$$

that is

$$-\alpha = s_0 + s_1 f$$

Positivstellensatz and Duality

We have the Positivstellensatz refutation

$$-\alpha = s_0 + \sum_i s_i f_i$$

- ► Dual SDP relaxation: express the SOS constraints as SDP constraints
- ▶ Primal SDP relaxation: relax the lifting

$$f_i(x)Y \succeq 0 \quad \text{for all } i$$
$$Y \succeq 0$$
$$Y_{11} = 1$$
$$Y = \begin{bmatrix} 1\\x\\\vdots \end{bmatrix} \begin{bmatrix} 1\\x\\\vdots \end{bmatrix}^T$$

Convex Relaxation of Semialgebraic Sets

Given a semialgebraic set, we have the lifting

$$f_i(x)Y \succeq 0$$

$$Y \succeq 0$$

$$Y_{11} = 1$$

$$Y = \begin{bmatrix} 1 \\ x \\ \vdots \end{bmatrix} \begin{bmatrix} 1 \\ x \\ \vdots \end{bmatrix}^T$$

- ▶ Projecting the feasible set onto the space spanned by *x* gives a convex relaxation of the original semialgebraic set
- ▶ We don't need to compute the projection explicitly
- ► To tighten the relaxation, include more monomials in *Y* equivalently, increase the degree of the multipliers in the refutation

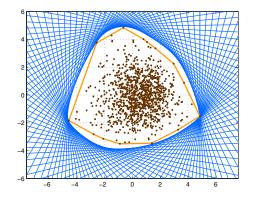
The Cut Polytope

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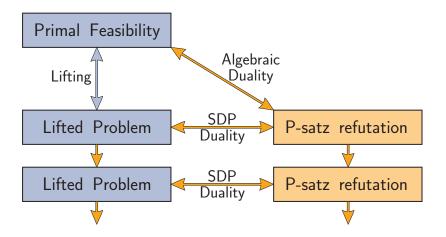
The feasible set of the MAXCUT problem is

$$C = \left\{ X \in \mathbb{S}^n \mid X = vv^T, \ v \in \{-1, 1\}^n \right\}$$

A simple SDP relaxation gives the outer approximation to its convex hull Here n = 11; the set has affine dimension 55; a projection is shown below



A General Scheme



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Distinguished Representations

We have a basic semialgebraic ${\boldsymbol S}$

$$S = \left\{ x \in \mathbb{R}^n \mid g_i(x) \ge 0 \text{ for all } i = 1, \dots, m \right\}$$

Which polynomials are non-negative on S?

- Every polynomial in $cone\{g_1, \ldots, g_m\}$ is non-negative on S
- ▶ But are there others? Recall radicality of ideals.

The Positivstellensatz gives an exact test, since $f(x) \ge 0$ for all $x \in S$ iff

$$\left\{ x \in \mathbb{R}^n \mid f(x) < 0, g_i(x) \ge 0 \right\}$$
 is empty

Distinguished Representations

If S is *compact*, then Schmüdgen showed

f(x) > 0 for all $x \in S \implies f \in \operatorname{cone}\{g_1, \dots, g_m\}$

More explicitly, this means

$$f = s_0 + \sum_i s_i g_i + \sum_{i,j} r_{ij} g_i g_j + \cdots$$

for some SOS polynomials s_i, r_{ij}, \ldots

Also notice

$$f(x) \ge 0$$
 for all $x \in S$ \Leftarrow $f \in \operatorname{cone}\{g_1, \dots, g_m\}$

Certificate of Positivity

The Positivstellensatz implies $f(x) \ge 0$ on S if and only if

$$sf = 1 + s_0 + \sum_i s_i g_i + \sum_{i,j} r_{ij} g_i g_j + \cdots$$

- ► Schmüdgen's distinguished representation implies that, to prove *strict* positivity, one can assume the multiplier *s* is a nonnegative constant
- ▶ i.e., one can *prove* positivity using *fewer axioms*. Consequently
 - proofs may become longer
 - \blacktriangleright need assumptions on S
- So we can fix the multiplier s, without theoretical loss, but this may require higher degree certificates
- Theoretical justification for optimization of polynomials over compact domains; e.g., Lyapunov stability in a basin of attraction

Reducing the Axiom Set

If there is a *single* polynomial g_k such that

$$\left\{ x \in \mathbb{R}^n \mid g_k(x) \ge 0 \right\}$$
 is compact

then Putinar's result holds:

$$f(x) > 0$$
 for all $x \in S \implies f = s_0 + \sum_i s_i g_i$ for some SOS s_i

► Stronger assumptions about *S* mean we can reduce axiom set further; we don't need to take products

Handelman Representations

Suppose that S is defined by *linear inequalities*

$$S = \left\{ x \in \mathbb{R}^n \mid b - Ax \ge 0 \right\}$$

and S is compact, with nonempty interior.

Then, if f(x) > 0, we have for $W \subset \mathbb{N}^m$

$$f = \sum_{\alpha \in W} c_{\alpha} \prod_{i=1}^{m} (b_{i} - a_{i}^{T} x)^{\alpha_{i}} \text{ for some } c_{\alpha} > 0$$

- ▶ No SOS polynomials, just constants c_{α} . Hence solvable using LP
- But proofs may be extremely long

Distinguished Representations

	Products	No products
SOS coefficients	Schmüdgen compactness	Putinar compactness++
Scalar coefficients	Handelman compactness linear inequalities	Lagrange convexity constraint qualifications

Strong duality results

Positivstellensatz requires no assumptions

▶ Tradeoffs between computation, assumptions, and proof lengths