

Recently seen

- Algebraic proof systems (NS, PC, Ideal Proof System)
- Semi-Algebraic proof systems (Cutting Planes)

Two more examples \uparrow Sherali-Adams and Sum-of-Squares

- Natural optimization algorithms that correspond to these systems

Let $F = C_1 \wedge C_2 \wedge \dots \wedge C_m$ be an unsatisfiable CNF formula.

We can encode F either as polynomial equations (e.g. algebraic pfs) or linear inequalities (e.g. Cutting planes)

Defn Let \mathcal{P}, \mathcal{Q} be sets of polynomials over \mathbb{R} in variables x_1, x_2, \dots, x_n . We assume \mathcal{Q} contains the polynomials

$$\begin{aligned} &- x_i^2 - x_i^0 \\ &- x_i^0 - x_i^1 \\ &- 1 \end{aligned}$$

For any $S, T \subseteq [n]$, $S \cap T = \emptyset$, let

$$\Sigma_{S,T} := \prod_{i \in S} x_i \prod_{j \in T} (1 - x_j) \leftarrow \text{"non-negative" junta}$$

A Sherali-Adams proof of the polynomial q from \mathcal{P} and \mathcal{Q} is an expression of the form

$$p = 0 \quad r \geq 0$$

$$\sum_{p \in P} h_p p + \sum_{r \in Q} (\sum c_i \Sigma_i r) r = q$$

where each $c_i r \in \mathbb{R}_{\geq 0}$, $\Sigma_i r$ is a non-negative junta, and each h_p is any polynomial.

i.e. if $P := \{p_1, \dots, p_m\}$ and $Q := \{r_1, \dots, r_\ell\}$ then an **SA proof** is a proof of the inequality

$$q \geq 0$$

from the system

$$p_1 = 0$$

$$\vdots$$

$$p_m = 0$$

and

$$r_1 \geq 0$$

$$\vdots$$

$$r_\ell \geq 0.$$

A Sherali-Adams **refutation** is a proof of -1

- The **degree** of the refutation is

$$\max\{\deg(h_p p), \deg(\Sigma_i r)\}$$

- The **size** of the refutation is the number of monomials obtained by expanding out all polynomials **before** cancellations.

By definition, a Nullstellensatz refutation of P over \mathbb{R} is also a Sherali-Adams refutation of $(P, \{x_i^2 - x_i, x_i - x_i^2\})$

Thm [Assignment 3]

Sherali-Adams can size- and degree- simulate Resolution!

If there is a Resolution proof of size s and width w , then there is a Sherali-Adams refutation with size $\text{poly}(s, w)$ and degree w .

ex] Let $F = C_1 \wedge C_2 \wedge \dots \wedge C_m$ be an unsat CNF formula, for any clause

$$C = \bigvee_{i \in S} x_i \vee \bigvee_{j \in T} \overline{x_j}$$

let $\tilde{C} = \sum_{i \in S} x_i + \sum_{j \in T} (1 - x_j)$, and (over $\{0, 1\}$ solns)

we can encode C as

$$\tilde{C} \geq 1.$$

$x_i^2 - x_i = 0$ encoded as

$$x_i^2 - x_i \geq 0$$

$$x_i - x_i^2 \geq 0$$

A SA proof would look like

$$\sum_{i=1}^m \left(\sum_{k_i} c_{k_i} \mathcal{S}_{k_i} \right) (\tilde{C}_i - 1) + \sum_{i=1}^n \gamma_{i,1} (x_i^2 - x_i) + \gamma_{i,2} (x_i - x_i^2) + \gamma_0 = -1$$

($\gamma \geq 0$) *

where $\gamma_0, \gamma_{i,1}, \gamma_{i,2}$ are non-negative combinations of non-negative juntas $\mathcal{S}_{S,T}$.

ex) $F = x_1 \wedge (\bar{x}_1 \vee x_2) \wedge (\bar{x}_2 \vee x_3) \wedge \bar{x}_3$

\downarrow \downarrow \downarrow \downarrow
 $x_1 \geq 1$ $(1-x_1)+x_2 \geq 1$ $(1-x_2)+x_3 \geq 1$ $1-x_3 \geq 1$

along with $x_i^2 - x_i \geq 0$, $x_i - x_i^2 \geq 0$, $1 \geq 0$

$$\begin{aligned}
 & 1 \cdot (x_1 - 1) + 1 \cdot (1 - x_1 + x_2 - 1) \\
 & \quad + 1 \cdot (1 - x_2 + x_3 - 1) \\
 & \quad + 1 \cdot (1 - x_3 - 1) \\
 & = -1
 \end{aligned}$$

$P = \emptyset$
 $Q = \left\{ \begin{array}{l} x_1 - 1, \\ (1 - x_1) + x_2 - 1, \\ (1 - x_2) + x_3 - 1, \\ (1 - x_3) - 1, \\ x_i^2 - x_i, x_i - x_i^2, \\ 1 \end{array} \right\}$

Already stronger than NS! NS requires degree $\Theta(\log n)$ to refute over any field (for the formula with n variables).

What about other proof systems?

- SA is definitely stronger than NS over \mathbb{R}
- On the assignment, prove that SA can simulate Resolution
- SA vs. Polynomial Calculus over \mathbb{R} ?
Separations in both directions!
- SA vs. Cutting Planes?
 - There is a formula that requires $\Omega(n)$ -degree SA with short CP

proofs. (PHP_n^{n+1} !)

Open Problem: Find a CNF F s.t. SA has efficient proofs (size/degree) while CP requires long proofs.

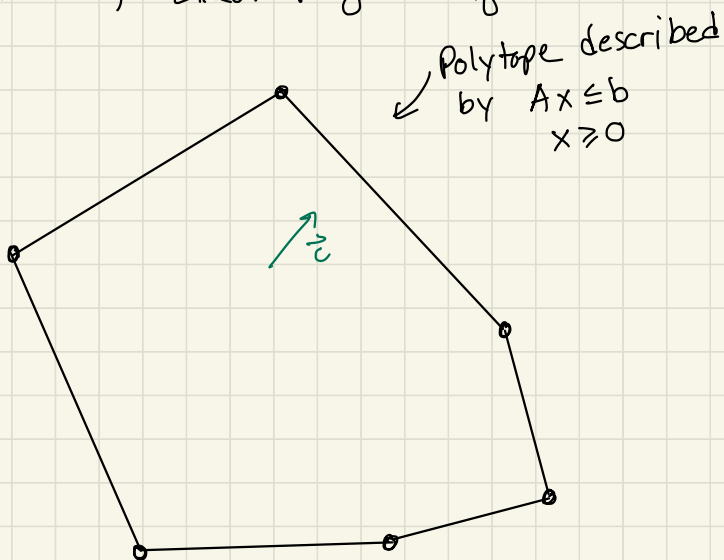
Thm Sherali-Adams requires degree $\Omega(\text{expansion}(G))$ to refute Tseig.

The Linear Programming Perspective

Recall the definitions of Linear Programming:

$$\begin{aligned} \max \quad & c \cdot x \\ \text{s.t.} \quad & Ax \leq b \\ & x \geq 0 \end{aligned}$$

Linear Programs are often used to give approximation algorithms for NP-Hard problems.



ex] Max-SAT: Given $F = C_1 \wedge \dots \wedge C_m$ CNF, find x maximizes the # of satisfied clauses.

Integer LP for Max-SAT

$$S, T \subseteq [n], S \cap T = \emptyset$$

$$C = \bigvee_{i \in S} x_i \vee \bigvee_{j \in T} \overline{x_j}$$

$$\max \sum_{i=1}^m c_i$$

$$\text{s.t. } \tilde{C}_i \geq c_i \quad \text{for all } i=1 \dots m$$

$$\Rightarrow \tilde{C} := \sum_{i \in S} x_i + \sum_{j \in T} (1 - x_j)$$

$$0 \leq x_i \leq 1 \quad \text{for all } i=1 \dots n$$

$$0 \leq c_i \leq 1 \quad \text{for all } i=1 \dots n$$

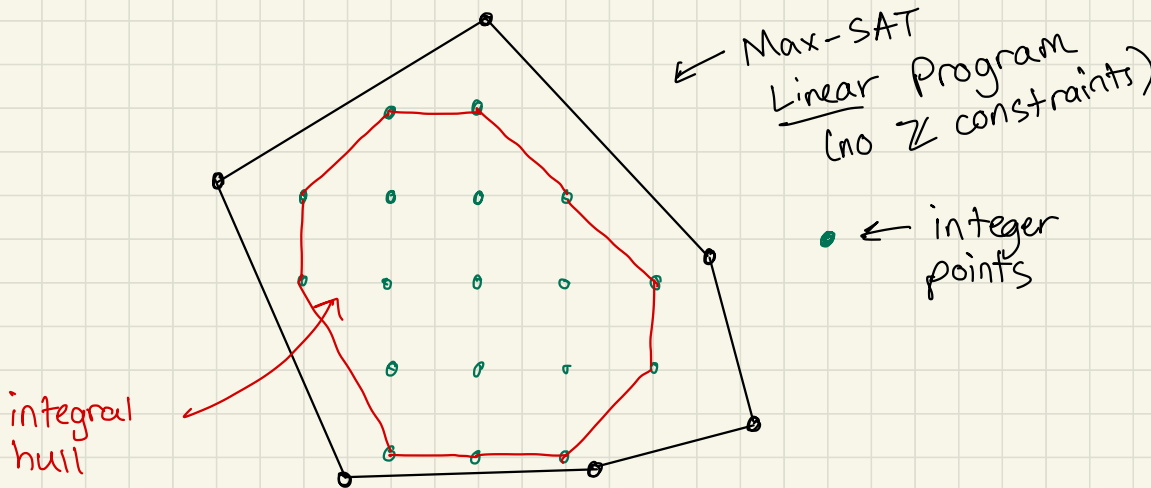
$$x_i, c_i \in \mathbb{Z}$$

C sat

iff

$$\tilde{C} \geq 1$$

Obtain the LP relaxation by removing \mathbb{Z} constraint.



Sherali-Adams can be used to systematically add new constraints