

# MaxSAT Resolution and SubCube Sums

Meena Mahajan



The Institute of Mathematical Sciences,  
Homi Bhabha National Institute,  
Chennai, India.

*Joint work with Yuval Filmus, Gaurav Sood, Marc Vinyals*

Satisfiability: Theory, Practice, and Beyond — Reunion  
14-17 June 2022

- The rule: From  $(C \vee x)$  and  $(D \vee \neg x)$ , infer  $C \vee D$ .
- A refutation of a propositional CNF formula  $F$ :  
A sequence  $C_1, C_2, \dots, C_t$  where  
 $C_t = \square$ , and  
for each  $i \in [t]$ , either  $C_i \in F$  or  
 $C_i$  is inferred from  $C_j, C_k$  for some  $j, k < i$ .
- Sequence  $F_0, F_1, F_2, \dots, F_t$  where  
 $F_0 = F$ , and for each  $i \in [t]$ ,  $F_i = F_{i-1} \cup \{C_i\}$ .
- The invariant: Every assignment satisfying  $F_{i-1}$  also satisfies  $F_i$ .
- $\square \in F_t$ , so no assignment satisfies  $F_t$ , so  $F_t$  unsat, so  $F_0 = F$  unsat.

# Certifying MaxSAT values

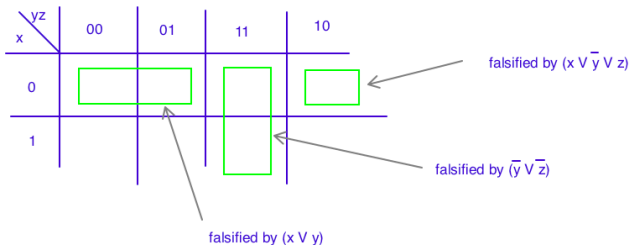
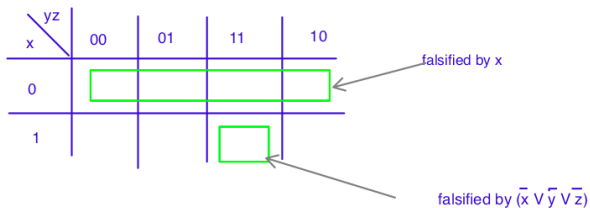
- For CNF formula  $F$ , number  $k$ ,  
Goal: show that every assignment falsifies at least  $k$  clauses.
- Produce a sequence  $F_0, F_1, \dots, F_t$  of multisets of clauses.
- Desired invariant: For every assignment  $\alpha$ , number of clauses falsified in  $F_{i-1}$  equals number of clauses falsified in  $F_i$ .

$$\text{viol}_{F_{i-1}}(\alpha) = \text{viol}_{F_i}(\alpha).$$

- Desired target:  $F_t$  has at least  $k$  copies of  $\square$ .
- Resolution does not maintain this invariant.  
The MaxSAT resolution rule, [BonetLevyManyá 2007], does.

# The MaxSat Resolution Rule

Rearrange cubes of falsifying assignments.



# The MaxSat Resolution Rule

$$\frac{\begin{array}{l} x \vee a_1 \vee \dots \vee a_s \\ \bar{x} \vee b_1 \vee \dots \vee b_t \end{array}}{a_1 \vee \dots \vee a_s \vee b_1 \vee \dots \vee b_t} \quad \begin{array}{l} (x \vee A) \\ (\bar{x} \vee B) \\ \text{(the "standard resolvent")} \end{array}$$

$$\left. \begin{array}{l} x \vee A \vee \bar{b}_1 \\ x \vee A \vee b_1 \vee \bar{b}_2 \\ \vdots \\ x \vee A \vee b_1 \vee \dots \vee b_{t-1} \vee \bar{b}_t \end{array} \right\} \text{(weakenings of } x \vee A)$$

$$\left. \begin{array}{l} \bar{x} \vee B \vee \bar{a}_1 \\ \bar{x} \vee B \vee a_1 \vee \bar{a}_2 \\ \vdots \\ \bar{x} \vee B \vee a_1 \vee \dots \vee a_{s-1} \vee \bar{a}_s \end{array} \right\} \text{(weakenings of } \bar{x} \vee B)$$

# A MaxSat Resolution derivation

- A sequence  $F_0, F_1, \dots, F_t$  of **multisets** of clauses.
- $F_0 = F$ .
- For  $i \in [t]$ ,  $F_i$  obtained from  $F_{i-1}$  by applying MaxSAT resolution rule, **replacing** the antecedents by the consequents.

# A MaxSat Resolution derivation

- A sequence  $F_0, F_1, \dots, F_t$  of **multisets** of clauses.
- $F_0 = F$ .
- For  $i \in [t]$ ,  $F_i$  obtained from  $F_{i-1}$  by applying MaxSAT resolution rule, **replacing** the antecedents by the consequents.

$$\bar{x} \vee y$$

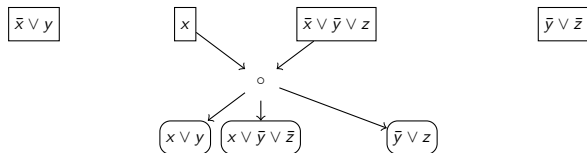
$$x$$

$$\bar{x} \vee \bar{y} \vee z$$

$$\bar{y} \vee \bar{z}$$

# A MaxSat Resolution derivation

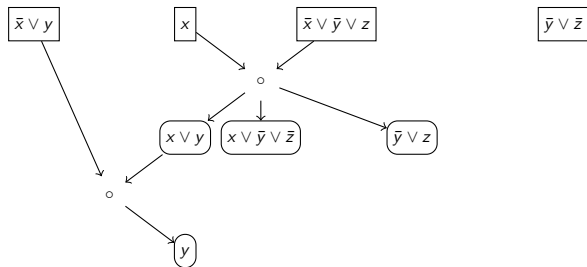
- A sequence  $F_0, F_1, \dots, F_t$  of **multisets** of clauses.
- $F_0 = F$ .
- For  $i \in [t]$ ,  $F_i$  obtained from  $F_{i-1}$  by applying MaxSAT resolution rule, **replacing** the antecedents by the consequents.





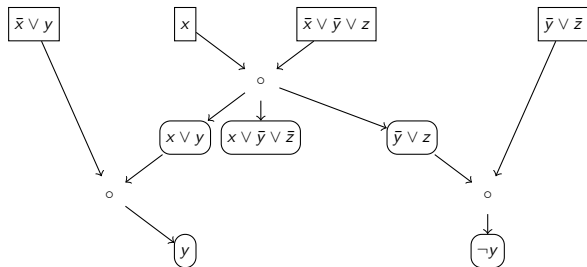
# A MaxSat Resolution derivation

- A sequence  $F_0, F_1, \dots, F_t$  of **multisets** of clauses.
- $F_0 = F$ .
- For  $i \in [t]$ ,  $F_i$  obtained from  $F_{i-1}$  by applying MaxSAT resolution rule, **replacing** the antecedents by the consequents.



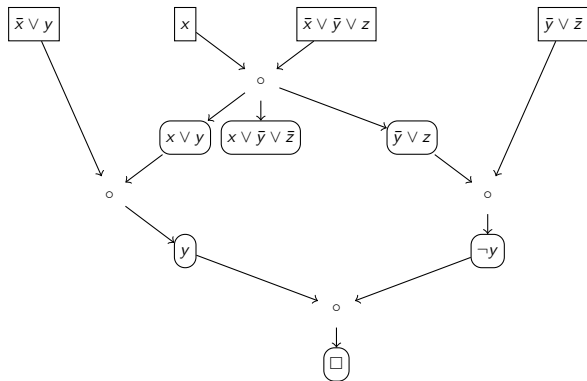
# A MaxSat Resolution derivation

- A sequence  $F_0, F_1, \dots, F_t$  of **multisets** of clauses.
- $F_0 = F$ .
- For  $i \in [t]$ ,  $F_i$  obtained from  $F_{i-1}$  by applying MaxSAT resolution rule, **replacing** the antecedents by the consequents.



# A MaxSat Resolution derivation

- A sequence  $F_0, F_1, \dots, F_t$  of **multisets** of clauses.
- $F_0 = F$ .
- For  $i \in [t]$ ,  $F_i$  obtained from  $F_{i-1}$  by applying MaxSAT resolution rule, **replacing** the antecedents by the consequents.



# The MaxSat Weakening Rule

- Recall invariant to be maintained:  
number of falsified clauses preserved.
- Usual weakening not sound. Instead:
- Replace a clause  $A$  by the two clauses  $A \vee x$  and  $A \vee \bar{x}$ .

# The MaxSat Weakening Rule

- Recall invariant to be maintained:  
number of falsified clauses preserved.
- Usual weakening not sound. Instead:
- Replace a clause  $A$  by the two clauses  $A \vee x$  and  $A \vee \bar{x}$ .
- Note: derivations are **reversible**.  
From  $F_i$  we can obtain  $F_{i-1}$  through a sequence of MaxSAT resolution and MaxSAT weakening rules.

# MaxSAT Resolution

- MaxSAT Resolution sound and complete for certifying MaxSAT value. [BonetLevyManyá 2007].
- In practice, MaxSAT solvers don't really use this rule directly.
- So why is it interesting?

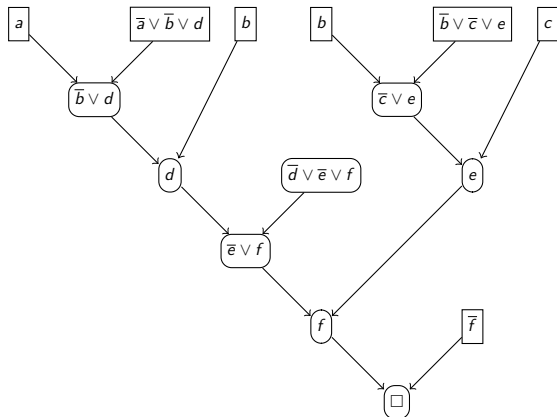
# Certifying Unsatisfiability

- Resolution can certify unsatisfiability.
- Using MaxSAT Resolution – overkill?
- Interesting things can happen if we do preprocessing.
  - Encode  $F$  into dualRailHorn  $F'$ ; then  $\max\text{SAT}(F') \geq n$ ; and  $F$  sat iff  $\max\text{SAT}(F') \leq n$ .
  - weighted DualRailMaxSAT p-simulates general Resolution.  
[Bonet,Buss,Ignatiev,Marques-Silva 2018]:
- Interesting things can happen if we allow arbitrary positive weights:  
It simulates Resolution and is equivalent to Circular Resolution.  
[BonetLevy 2020].
- Interesting things can happen if we allow negative weights and virtual creation – add  $(A, w)$  and  $(A, -w)$  to the current multiset.  
It simulates Resolution and is equivalent to Circular Resolution.  
[LarrossaRollon 2020].
- Understanding unweighted MaxSAT resolution and weakening better can help lead to more such extensions.

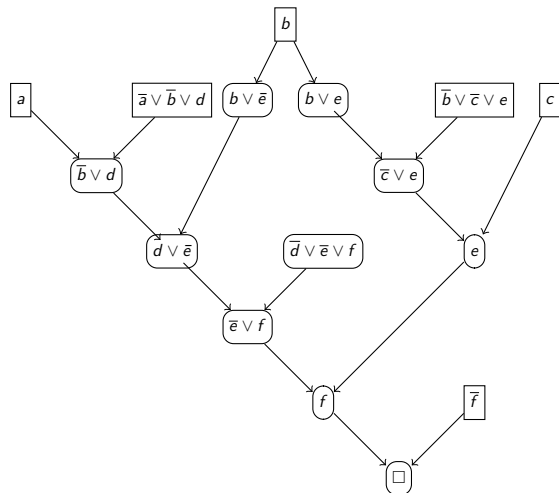
- We consider two proof systems for certifying unsatisfiability:  
MaxRes: only MaxSAT Resolution, and  
MaxResW: MaxSAT Resolution and MaxSAT Weakening.  
(All rules unweighted)
- Sound – invariant maintained at each stage
- Complete – because complete even for certifying MaxSAT value
- $p$ -simulated by Resolution: add instead of replace clauses.



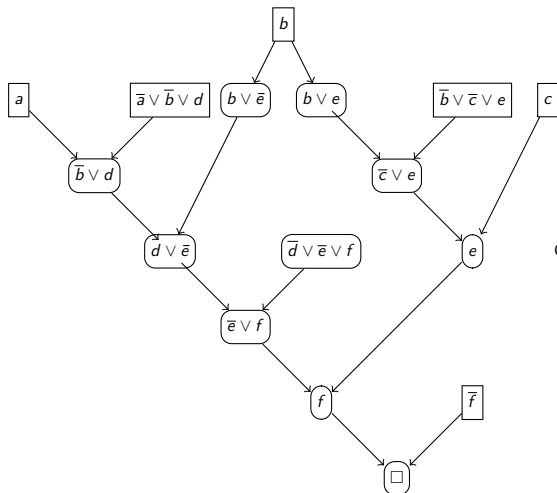
# MaxResW $p$ -simulates TreeRes



# MaxResW $p$ -simulates TreeRes



# MaxResW $\rho$ -simulates TreeRes



Open: Is the weakening rule really necessary?

# MaxResW better than TreeRes

## Theorem

*TreeRes does not simulate MaxRes*

- Pebbling formulas on single-sink DAGs: easy in TreeRes
- Compose with  $OR_2$ : hard for TreeRes on Pyramid Graphs [Ben-SassonImpagliazzoWigderson 2004].

# MaxResW better than TreeRes

## Theorem

*TreeRes does not simulate MaxRes*

- Pebbling formulas on single-sink DAGs: easy in TreeRes
- Compose with  $OR_2$ : hard for TreeRes on Pyramid Graphs [Ben-SassonImpagliazzoWigderson 2004].
- Composed formula easy in MaxRes/MaxResW?? We don't know.

# MaxResW better than TreeRes

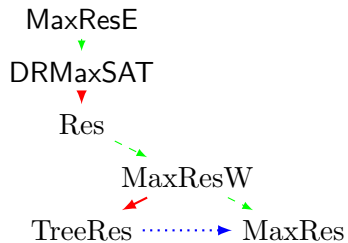
## Theorem

*TreeRes does not simulate MaxRes*

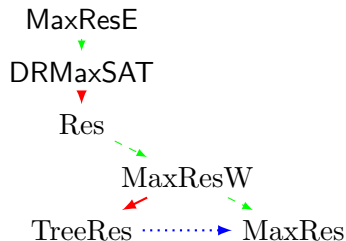
- Pebbling formulas on single-sink DAGs: easy in TreeRes
- Compose with  $OR_2$ : hard for TreeRes on Pyramid Graphs [Ben-SassonImpagliazzoWigderson 2004].
- Composed formula easy in MaxRes/MaxResW?? We don't know.
- Tweak the composed formulas — add some *hint* clauses.
- Show: now short MaxRes refutation.
- Show: still hard for TreeRes. use game, 1-query complexity, pebbling

$$\begin{aligned} \text{TreeResSz}(F \circ OR) &\geq 2^{\text{DelayScore on } F \circ OR} && \text{[PudImp 2000]} \\ &\geq 2^{DT_1(\text{Search}F)} \\ &\geq 2^{\text{peb}(G)} && \text{(if } F = \text{PebHint}(G)) \\ &\geq 2^{\text{PyramidHeight}} && \text{(for } G = \text{Pyr, [Cook 1974])} \end{aligned}$$

# Relating to Resolution



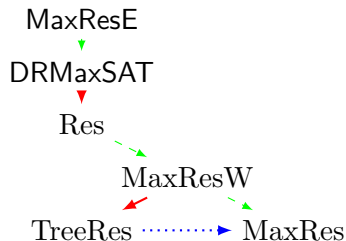
# Relating to Resolution



- Does MaxRes, or even MaxResW, simulate Res?  
We wouldn't expect this, but it seemed hard to prove.
- Need a lower bound technique that is specific to MaxResW, not inherited from Res.



# Relating to Resolution



- Does MaxRes, or even MaxResW, simulate Res?  
We wouldn't expect this, but it seemed hard to prove.
- Need a lower bound technique that is specific to MaxResW, not inherited from Res.
- Observation: MaxResW refutation  $F_0, F_1, \dots, F_t$  where  $\square \in F_t$ .  
Let  $G = F_t$  minus one copy of  $\square$ .
  - For every assignment  $\alpha$ ,  $\text{viol}_{F_0}(\alpha) = \text{viol}_{F_t}(\alpha) = 1 + \text{viol}_G(\alpha)$ .
  - $|G|$  is polynomial in  $|F|, t$ .

Showing that every such  $G$  is large gives a MaxResW lower bound.

# The SubCube Sums Proof System

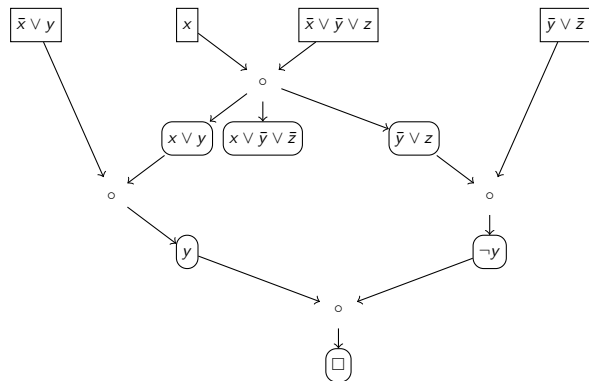
A proof that  $F$  is unsatisfiable: a multiset  $G$  of clauses such that

$$\forall \alpha : \text{viol}_F(\alpha) = 1 + \text{viol}_G(\alpha).$$

# The SubCube Sums Proof System

A proof that  $F$  is unsatisfiable: a multiset  $G$  of clauses such that

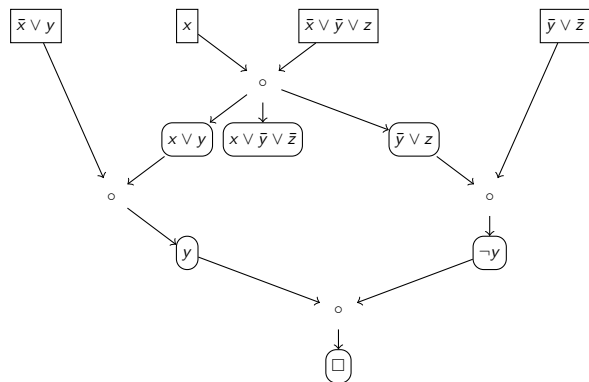
$$\forall \alpha : \text{viol}_F(\alpha) = 1 + \text{viol}_G(\alpha).$$



# The SubCube Sums Proof System

A proof that  $F$  is unsatisfiable: a multiset  $G$  of clauses such that

$$\forall \alpha : \text{viol}_F(\alpha) = 1 + \text{viol}_G(\alpha).$$



Final multiset:  
 $\{x \vee \bar{y} \vee \bar{z}, \square\}$

# The SubCube Sums Proof System (cont'd)

- A proof that  $F$  is unsatisfiable: a multiset  $G$  of clauses such that

$$\forall \alpha : \text{viol}_{F_0}(\alpha) = \text{viol}_{F_t}(\alpha) = 1 + \text{viol}_G(\alpha).$$

# The SubCube Sums Proof System (cont'd)

- A proof that  $F$  is unsatisfiable: a multiset  $G$  of clauses such that

$$\forall \alpha : \text{viol}_{F_0}(\alpha) = \text{viol}_{F_t}(\alpha) = 1 + \text{viol}_G(\alpha).$$

- Not easy to verify the proof.  
(Possible in randomized polynomial time.)
- Not important if we only use such proofs for MaxResW lower bounds.

# The SubCube Sums Proof System (cont'd)

- A proof that  $F$  is unsatisfiable: a multiset  $G$  of clauses such that

$$\forall \alpha : \text{viol}_{F_0}(\alpha) = \text{viol}_{F_t}(\alpha) = 1 + \text{viol}_G(\alpha).$$

- Not easy to verify the proof.  
(Possible in randomized polynomial time.)
- Not important if we only use such proofs for MaxResW lower bounds.
- Sound: almost by definition
- Complete: every MaxRes refutation gives a SubCubeSums proof.
- SubCubeSums lower bound  $\implies$  MaxResW lower bound,  
not Res lower bound.

# What's easy for SubCubeSums

- Everything easy for MaxResW, in particular for TreeRes.



# What's easy for SubCubeSums

- Everything easy for MaxResW, in particular for TreeRes.
- The SubsetCardinality Formulas.

A bipartite graph has a subgraph where the left-degrees are bounded above, the right-degrees are bounded below, and the sums are not the same – obviously Unsat.

On expander graphs, hard for Res and MaxResW

[MiksaNordstrom 2014].

Easy for SubCubeSums.

# What's easy for SubCubeSums

- Everything easy for MaxResW, in particular for TreeRes.
- The SubsetCardinality Formulas.  
A bipartite graph has a subgraph where the left-degrees are bounded above, the right-degrees are bounded below, and the sums are not the same – obviously Unsat.  
On expander graphs, hard for Res and MaxResW  
[MiksaNordstrom 2014].  
Easy for SubCubeSums.
- The PigeonHole Principle Formulas  $\text{PHP}_n^{n+1}$ .  
Upper bound implicit in [LarrosaRollon 2020], but proved via negative weighted MaxSAT resolution.  
We give direct combinatorial proof.

# What's hard for SubCubeSums

The Tseitin Contradictions on expander graphs.  
SubCubeSums proofs must have exponential size.

These formulas are also hard for Res.  
So not useful for separating Res from MaxResW.

# A Lower Bound Technique for SubCubeSums

## Theorem (Lifting)

*If  $d$  is the minimum width of any SubCubeSums refutation of  $F$ , then any SubCubeSums refutation of  $F \circ XOR$  has size  $\exp(\Omega(d))$ .*

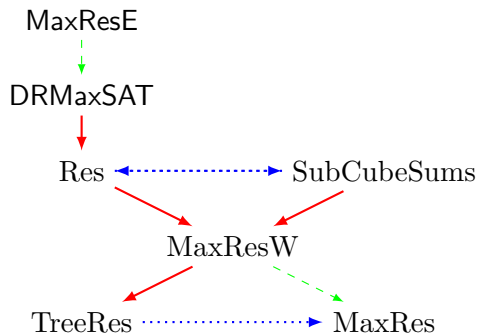
### Outline:

- $\text{viol}_F(\alpha_1 \oplus \alpha_2) = \text{viol}_{F \circ \oplus}(\alpha_1, \alpha_2)$ . Hence
- $\text{viol}_{F \circ \oplus} - 1 = ((\text{viol}_F) \circ \oplus) - 1 = (\text{viol}_F - 1) \circ \oplus$ .
- A “size-width” relation holds for SubCubeSums:  
[SubCubeSums width for  $F$ ] is  
 $O(\log[\text{SubCubeSums size for } F \circ XOR])$ .

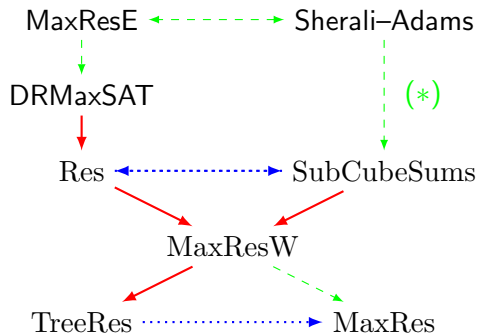
# Recent Developments

- A width lower bound for SubCubeSums; [FlemingGöösGrosserRobere 2022].
- The hard formula  $F$  is a specific kind of pebbling contradiction.  $F \circ XOR$  is easy to refute in Resolution.
- Thus, Resolution is strictly stronger than MaxResW, and incomparable with SubCubeSums.
- Another close variant of MaxRes defined and separated from Resolution; [GöösHollenderJainMaystrePiresRobereTao 2022].

# Relations between proof systems



# Relations between proof systems



# SubCubeSums: An Algebraic View

- $\text{viol}_{\mathcal{F}}(\alpha) = \text{viol}_{\mathcal{G}}(\alpha) + 1.$



# SubCubeSums: An Algebraic View

- $\text{viol}_F(\alpha) = \text{viol}_G(\alpha) + 1$ .
- Encode clauses as polynomials.  $(x \vee \neg y \vee z \rightarrow x(1-y)z)$   
 $-p_F(\alpha) + p_G(\alpha) + 1 = 0$  for all  $\alpha \in \{0, 1\}^n$ .

# SubCubeSums: An Algebraic View

- $\text{viol}_F(\alpha) = \text{viol}_G(\alpha) + 1$ .
- Encode clauses as polynomials.  $(x \vee \neg y \vee z \rightarrow x(1 - y)z)$   
 $-p_F(\alpha) + p_G(\alpha) + 1 = 0$  for all  $\alpha \in \{0, 1\}^n$ .
- Polynomials multilinear, hence this is an identity:  $-p_F + p_G + 1 \equiv 0$ .  
A restricted Sherali–Adams proof!
- Many negated literals in a clause  $\implies$  too many monomials.  
Standard approach: use twin variables  $x, \bar{x}$  and axioms  $x + \bar{x} = 1$ .
- $-p_F + p_G + 1 + (\text{a polynomial combination of Boolean axioms}) \equiv 0$ .

# Algebraic Measures for SubCubeSums

$$-P_F + P_G + 1 + \left( \begin{array}{l} \text{polynomial} \\ \text{combination} \\ \text{of Boolean} \\ \text{twin axioms} \end{array} \right) \equiv 0$$

$$-P_F + P_G + 1 + \sum_j q_j (1 - x_j - \bar{x}_j) \equiv 0$$

Degree

$$D = \max \{ D_1, D_2, 0, D_3 \}$$

$\swarrow$   $\nwarrow$   $\swarrow$   $\nwarrow$   
 $\leftarrow \text{width}(F) \quad \text{width}(G) \quad \leftarrow \max\{D_1, D_2\}$

$$= \max \{ D_1, D_2 \}$$

$\searrow$   $\swarrow$   
 $\text{degree of Conical Junta}$   
 $\text{size}$

# Monomials

$$M = M_1 + M_2 + 1 + M_3$$

$$= |F| + |G| + 1 + M_3$$

Total

$$\text{Coeff Size (Unary)} = |F| + |G| + 1 + S_3$$

# Algebraic Measures for SubCubeSums (cont'd)

$$\begin{aligned} \text{Total Unary Coeff size} &= |F| + |G| + 1 + \sum_3 \leftarrow \begin{array}{l} \text{from} \\ \text{Boolean axioms} \\ \text{part} \end{array} \\ &\quad \underbrace{\hspace{1.5cm}}_{\text{SCS (SubCubeSums) Size}} \\ &\quad \underbrace{\hspace{2.5cm}}_{\text{SCS-reduced algebraic size}} \\ &\quad \underbrace{\hspace{4.5cm}}_{\text{SCS algebraic size}} \end{aligned}$$

- For an SCS proof,  
SCS size  $\leq$  SCS reduced algebraic size  $\leq$  SCS algebraic size.
- For any unsat formula,
  - Sherali-Adams size  $\leq$  unary Sherali-Adams size  $\leq$  SCS algebraic size.
  - Sherali-Adams degree  $\leq$  SCS degree =  $\max\{\text{width}(F), \text{width}(G)\}$ .

# Take-away

- Using the MaxSAT Resolution and MaxSAT weakening rules to certify unsatisfiability:
  - no worse than TreeRes.
  - on some formulas, exponentially better.
- Key to understanding MaxSAT: rearrangements of Boolean subcubes.
- SubCubeSums proof system:
  - simulates and strictly better than MaxSAT resolution,
  - incomparable with Resolution.
  - can be viewed as a restriction of Sherali–Adams.

- Using the MaxSAT Resolution and MaxSAT weakening rules to certify unsatisfiability:
  - no worse than TreeRes.
  - on some formulas, exponentially better.
- Key to understanding MaxSAT: rearrangements of Boolean subcubes.
- SubCubeSums proof system:
  - simulates and strictly better than MaxSAT resolution,
  - incomparable with Resolution.
  - can be viewed as a restriction of Sherali–Adams.

Thank you for listening!