

# Proof Logging MaxCDCL and MDD-encodings

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Joint work with Bart Bogaerts and Jordi Coll

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ARTIFICIAL  
INTELLIGENCE  
RESEARCH GROUP

## OUTLINE OF THIS PRESENTATION

- ▶ What is **MaxSAT** and how to certify it?
- ▶ Proof logging the B&B solver **MaxCDCL**
- ▶ Proof logging **additional techniques** in MaxCDCL
  - ▶ **Hardening**
  - ▶ **Literal Unlocking**
- ▶ Proof logging **BDD PB-to-CNF encoding**
- ▶ Future work & Conclusions

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## THE MAXIMUM SATISFIABILITY PROBLEM

Example:

$$F = \{x_1 \vee x_2, x_2 \vee x_3, x_1 \vee \overline{x_2} \vee x_3\}$$

$$\mathcal{O} = x_1 + x_2 + x_3$$

Optimization variant of **Satisfiability Problem**.

A **MaxSAT-instance** is a tuple  $(F, \mathcal{O})$  with:

- ▶  $F$  a **propositional formula**
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A **solution** is an assignment for all variables such that:

- ▶ The formula  $F$  is **satisfied**
- ▶ No other satisfying assignment has lower **objective value**

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$$\text{Solution: } \alpha = \{x_1 \mapsto 1, x_2 \mapsto 0, x_3 \mapsto 1\}$$

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- ▶ **MaxSAT resolution** [LH05, HL06, BLM06, BLM07]
- ▶ **Tableaux reasoning** [LMS16, LCH<sup>+</sup>22, LM22]
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Solvers **specifically designed** for emitting proofs

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**No certified state-of-the-art MaxSAT solver using native proof system!**

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Four main categories:

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Different reasoning techniques!

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Only proves answer correct, not reasoning within solver!



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  - ▶ Challenge: Intricate without-loss-of-generality reasoning in the DPW encoding

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This talk:

- ▶ **Branch-and-Bound**

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This talk:

- ▶ **Branch-and-Bound** (and a little bit of **Solution-Improving** Search)



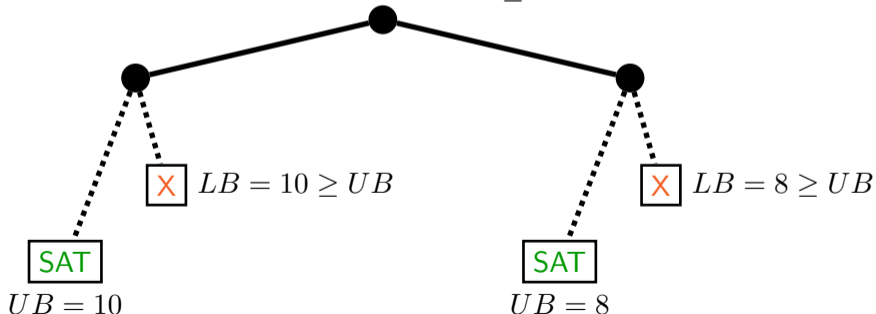
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# BRANCH AND BOUND

## Branch and Bound:

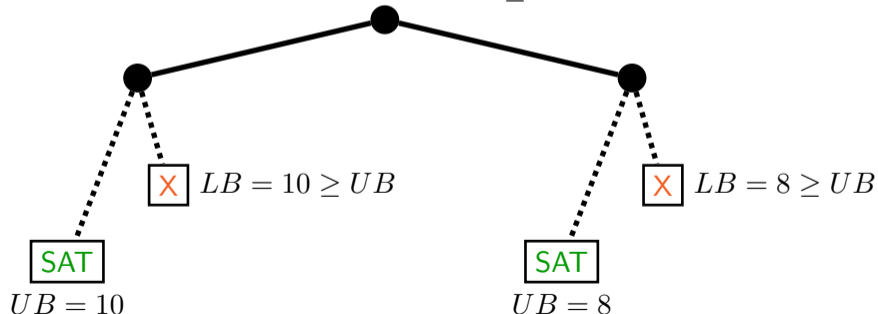
- ▶ Explore the search tree looking for optimal solutions
- ▶ Update Upper Bound  $UB$  when solution with better cost is found
- ▶ Underestimate  $LB$  of the cost at every node
- ▶ Prune branch when conflict found or when  $LB \geq UB$



# MAXCDCL AS BRANCH AND BOUND

## Branch and Bound in MaxCDCL:

- ▶ Explore the search tree (**CDCL**) looking for satisfiable assignments
- ▶ Update Upper Bound  $UB$  when solution with better cost is found
- ▶ Underestimate  $LB$  of the cost at every node **using lookahead with UP**
- ▶ Prune branch when conflict found or when  $LB \geq UB$  **and learn a clause**



# MAXCDCL AS CDCL GENERALIZATION

## MaxCDCL conflicts:

- ▶ **Hard conflict:**

- ▶ A clause is falsified

- ▶ **Soft conflict:**

- ▶ (underestimated)  $LB \geq UB$

# MAXCDCL AS CDCL GENERALIZATION

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**In both cases: conflict analysis for learning new clause (CDCL)**

## LOOKAHEAD: LB UNDERESTIMATION (UNWEIGHTED CASE)

**Lookahead with UP** for underestimating LB:

1. Assume unassigned objective literals false and apply UP until:
  - ▶ A hard clause is falsified
  - ▶ Or a not yet assigned objective literal is assigned 1
2. We have found a **local unsatisfiable core**
3. Since unweighted case: Each **disjoint** core increases the LB by 1
4. When  $LB \geq UB$ , a soft conflict is found

## SOFT CONFLICT DETECTION BY EXAMPLE (UNWEIGHTED CASE)

$$\mathcal{O}^t = y_1 + y_2 + y_3 + y_4 + y_5 + y_6 + y_7 + y_8 \quad \mathbf{UB} = 3$$

$$\mathbf{Trail: } x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p$$

## SOFT CONFLICT DETECTION BY EXAMPLE (UNWEIGHTED CASE)

$$\mathcal{O}^t = \neg y_1 + y_2 + y_3 + y_4 + y_5 + y_6 + y_7 + y_8 \quad \mathbf{UB} = 3$$

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**Find one core:**

$$x_1^d \ \overline{x_2^p} \ x_3^p \ \overline{x_4^d} \ x_5^p \ x_6^p \ x_7^p \ \overline{y_1^a} \ x_9^p \ x_{10}^p$$



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**Local core:**

$$\overline{x_2} \wedge \overline{x_4} \wedge \overline{y_1} \wedge \overline{y_4} \rightarrow \square$$

$$\overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_4 \quad (\text{Reasons} \rightarrow \text{Core})$$

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$$\mathbf{Trail:} \quad x_1^d \quad \overline{x_2^p} \quad x_3^p \quad \overline{x_4^d} \quad x_5^p \quad x_6^p \quad x_7^p$$

**Find next core:**

$$x_1^d \quad \overline{x_2^p} \quad x_3^p \quad \overline{x_4^d} \quad x_5^p \quad x_6^p \quad x_7^p \quad \overline{y_2^a} \quad \overline{y_3^a} \quad y_5^p \quad (\text{Propagate } y_5 \text{ true})$$

$$\overline{x_2^p} \quad x_7^p \quad \overline{y_2^a} \quad \overline{y_3^a} \quad y_5^p \quad (\text{Conflict analysis})$$

## SOFT CONFLICT DETECTION BY EXAMPLE (UNWEIGHTED CASE)

$$\mathcal{O}^t = \cancel{y_1} + \cancel{y_2} + \cancel{y_3} + \cancel{y_4} + \cancel{y_5} + y_6 + y_7 + y_8 \quad \mathbf{UB} = 3$$

$$\mathbf{Trail:} \quad x_1^d \quad \overline{x_2^p} \quad x_3^p \quad \overline{x_4^d} \quad x_5^p \quad x_6^p \quad x_7^p$$

**Find next core:**

$$x_1^d \quad \overline{x_2^p} \quad x_3^p \quad \overline{x_4^d} \quad x_5^p \quad x_6^p \quad x_7^p \quad \overline{y_2^a} \quad \overline{y_3^a} \quad y_5^p \quad (\text{Propagate } y_5 \text{ true})$$

$$\overline{x_2^p} \quad x_7^p \quad \overline{y_2^a} \quad \overline{y_3^a} \quad y_5^p \quad (\text{Conflict analysis})$$

**Local core:**

$$\overline{x_2} \wedge x_7 \wedge \overline{y_2} \wedge \overline{y_3} \wedge \overline{y_5} \rightarrow \square$$

$$\overline{x_2} \wedge x_7 \rightarrow y_2 \vee y_3 \vee y_5 \quad (\text{Reasons} \rightarrow \text{Core})$$

## SOFT CONFLICT DETECTION: FULL EXAMPLE (UNWEIGHTED CASE)

$$\mathcal{O} = y_1 + y_2 + y_3 + y_4 + y_5 + y_6 + y_7 + y_8 \quad \mathbf{UB} = 3$$

$$\mathbf{Trail: } x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p$$

**Found disjoint local cores**

$$\text{Core 1: } \overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_4$$

$$\text{Core 2: } \overline{x_2} \wedge x_7 \rightarrow y_2 \vee y_3 \vee y_5$$

$$\text{Core 3: } x_1 \wedge \overline{x_4} \wedge x_7 \rightarrow y_6 \vee y_7$$

## SOFT CONFLICT DETECTION: FULL EXAMPLE (UNWEIGHTED CASE)

$$\mathcal{O} = y_1 + y_2 + y_3 + y_4 + y_5 + y_6 + y_7 + y_8 \quad \mathbf{UB} = 3$$

$$\mathbf{Trail: } x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p$$

**Found disjoint local cores**

$$\mathbf{Core 1: } \overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_4$$

$$\mathbf{Core 2: } \overline{x_2} \wedge x_7 \rightarrow y_2 \vee y_3 \vee y_5$$

$$\mathbf{Core 3: } x_1 \wedge \overline{x_4} \wedge x_7 \rightarrow y_6 \vee y_7$$

$$x_1 \wedge \overline{x_2} \wedge \overline{x_4} \wedge x_7 \rightarrow (y_1 \vee y_4) \wedge (y_2 \vee y_3 \vee y_5) \wedge (y_6 \vee y_7)$$

## SOFT CONFLICT DETECTION: FULL EXAMPLE (UNWEIGHTED CASE)

$$\mathcal{O} = y_1 + y_2 + y_3 + y_4 + y_5 + y_6 + y_7 + y_8 \quad \mathbf{UB} = 3$$

$$\mathbf{Trail: } x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p$$

**Found disjoint local cores**

$$\mathbf{Core 1: } \overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_4$$

$$\mathbf{Core 2: } \overline{x_2} \wedge x_7 \rightarrow y_2 \vee y_3 \vee y_5$$

$$\mathbf{Core 3: } x_1 \wedge \overline{x_4} \wedge x_7 \rightarrow y_6 \vee y_7$$

$$x_1 \wedge \overline{x_2} \wedge \overline{x_4} \wedge x_7 \rightarrow (y_1 \vee y_4) \wedge (y_2 \vee y_3 \vee y_5) \wedge (y_6 \vee y_7)$$

$$x_1 \wedge \overline{x_2} \wedge \overline{x_4} \wedge x_7 \rightarrow \mathbf{LB} = 3 \geq 3 = \mathbf{UB}$$

## SOFT CONFLICT DETECTION: FULL EXAMPLE (UNWEIGHTED CASE)

$$\mathcal{O} = y_1 + y_2 + y_3 + y_4 + y_5 + y_6 + y_7 + y_8 \quad \mathbf{UB} = 3$$

$$\mathbf{Trail: } x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p$$

**Found disjoint local cores**

$$\mathbf{Core 1: } \overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_4$$

$$\mathbf{Core 2: } \overline{x_2} \wedge x_7 \rightarrow y_2 \vee y_3 \vee y_5$$

$$\mathbf{Core 3: } x_1 \wedge \overline{x_4} \wedge x_7 \rightarrow y_6 \vee y_7$$

$$x_1 \wedge \overline{x_2} \wedge \overline{x_4} \wedge x_7 \rightarrow (y_1 \vee y_4) \wedge (y_2 \vee y_3 \vee y_5) \wedge (y_6 \vee y_7)$$

$$x_1 \wedge \overline{x_2} \wedge \overline{x_4} \wedge x_7 \rightarrow \mathbf{LB} = 3 \geq 3 = \mathbf{UB}$$

**Soft conflict:**

$$x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p, \quad \mathbf{Conflict } \overline{x_1} \vee x_2 \vee x_4 \vee \overline{x_7} \text{ (soft conflict)}$$

## SOFT CONFLICT DETECTION (WEIGHTED CASE)

### Weighted MaxCDCL

- ▶ Weight of Local Core  $\mathcal{K}$  = smallest weight of objective literals in  $\mathcal{K}$
- ▶ Each objective literal can contribute to many cores
- ▶ The total contribution of a literal cannot exceed its weight



## SOFT CONFLICT DETECTION (WEIGHTED CASE)

### Weighted MaxCDCL

- ▶ Weight of Local Core  $\mathcal{K}$  = smallest weight of objective literals in  $\mathcal{K}$
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$$\mathcal{O}^t = 7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \quad \mathbf{UB} = 4$$

**Trail:**  $x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p$

### Found local cores

## SOFT CONFLICT DETECTION (WEIGHTED CASE)

**Weighted MaxCDCL**

- ▶ Weight of Local Core  $\mathcal{K}$  = smallest weight of objective literals in  $\mathcal{K}$
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$$\mathbf{Trail: } x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p$$

**Found local cores**

Core 1:  $\overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_2$  (weight 2)

## SOFT CONFLICT DETECTION (WEIGHTED CASE)

**Weighted MaxCDCL**

- ▶ Weight of Local Core  $\mathcal{K}$  = smallest weight of objective literals in  $\mathcal{K}$
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$$\mathcal{O}^t = 7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \quad \mathbf{UB} = 4$$

$$\mathbf{Trail: } x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p$$

**Found local cores**

Core 1:  $\overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_2$  (weight 2)

## SOFT CONFLICT DETECTION (WEIGHTED CASE)

**Weighted MaxCDCL**

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$$O^t = 7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \quad \mathbf{UB} = 4$$

$$\mathbf{Trail: } x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p$$

**Found local cores**

Core 1:  $\overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_2$  (weight 2)

Core 2:  $x_3 \wedge \overline{x_4} \rightarrow y_1 \vee y_5$  (weight 1)

## SOFT CONFLICT DETECTION (WEIGHTED CASE)

**Weighted MaxCDCL**

- ▶ Weight of Local Core  $\mathcal{K}$  = smallest weight of objective literals in  $\mathcal{K}$
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$$O^t = 7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \quad \mathbf{UB} = 4$$

$$\mathbf{Trail: } x_1^d \overline{x_2^p} x_3^p \overline{x_4^d} x_5^p x_6^p x_7^p$$

**Found local cores**

$$\text{Core 1: } \overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_2 \quad (\text{weight } 2)$$

$$\text{Core 2: } x_3 \wedge \overline{x_4} \rightarrow y_1 \vee y_5 \quad (\text{weight } 1)$$

## SOFT CONFLICT DETECTION (WEIGHTED CASE)

## Weighted MaxCDCL

- ▶ Weight of Local Core  $\mathcal{K}$  = smallest weight of objective literals in  $\mathcal{K}$
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$$O^t = 7 \bar{y}_1 + 2 y_2 + 1 y_3 + 1 y_4 + 1 \bar{y}_5 + 4 y_6 + 1 y_7 + 3 \bar{y}_8 \quad \mathbf{UB} = 4$$

$$\mathbf{Trail: } x_1^d \bar{x}_2^p x_3^p \bar{x}_4^d x_5^p x_6^p x_7^p$$

## Found local cores

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (weight 2)

Core 2:  $x_3 \wedge \bar{x}_4 \rightarrow y_1 \vee y_5$  (weight 1)

Core 3:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (weight 3)

## SOFT CONFLICT DETECTION (WEIGHTED CASE)

## Weighted MaxCDCL

- ▶ Weight of Local Core  $\mathcal{K}$  = smallest weight of objective literals in  $\mathcal{K}$
- ▶ Each objective literal can contribute to many cores
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$$O^t = 7 \bar{y}_1 + 17 \bar{y}_2 + 2y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 1y_6 + 1y_7 + 3y_8 \quad \mathbf{UB} = 4$$

$$\mathbf{Trail: } x_1^d \bar{x}_2^p x_3^p \bar{x}_4^d x_5^p x_6^p x_7^p$$

## Found local cores

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (weight 2)

~~Core 2:  $x_3 \wedge \bar{x}_4 \rightarrow y_1 \vee y_5$  (weight 1)~~

Core 3:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (weight 3)

## SOFT CONFLICT DETECTION (WEIGHTED CASE)

**Weighted MaxCDCL**

- ▶ Weight of Local Core  $\mathcal{K}$  = smallest weight of objective literals in  $\mathcal{K}$
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$$\mathbf{Trail: } x_1^d \bar{x}_2^p x_3^p \bar{x}_4^d x_5^p x_6^p x_7^p$$

**Found local cores**

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (weight 2)

Core 3:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (weight 3)



## SOFT CONFLICT DETECTION (WEIGHTED CASE)

## Weighted MaxCDCL

- ▶ Weight of Local Core  $\mathcal{K}$  = smallest weight of objective literals in  $\mathcal{K}$
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$$O^t = 7 \bar{y}_1 + 17 \bar{y}_2 + 2y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 1y_6 + 1y_7 + 3y_8 \quad \mathbf{UB} = 4$$

$$\mathbf{Trail: } x_1^d \bar{x}_2^p \ x_3^p \bar{x}_4^d \ x_5^p \ x_6^p \ x_7^p$$

## Found local cores

$$\text{Core 1: } \bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2 \quad (\text{weight 2})$$

$$\text{Core 3: } x_1 \rightarrow y_1 \vee y_6 \vee y_8 \quad (\text{weight 3})$$

$$\text{Conclusion: } x_1 \wedge \bar{x}_2 \wedge \bar{x}_4 \rightarrow LB = 5 \geq 4 = UB \quad \mathbf{Soft Conflict clause: } \bar{x}_1 \vee x_2 \vee x_4$$

## PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

# PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

---

## Found “disjoint” cores

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

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## PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

---

**Found “disjoint” cores**

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

PB:  $x_2 + x_4 + y_1 + y_2 \geq 1$

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

PB:  $\bar{x}_1 + y_1 + y_6 + y_8 \geq 1$

---

## PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

---

**Found “disjoint” cores (RUP)**

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

PB:  $x_2 + x_4 + y_1 + y_2 \geq 1$

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

PB:  $\bar{x}_1 + y_1 + y_6 + y_8 \geq 1$

---

# PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

## Found “disjoint” cores (RUP)

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

PB:  $2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2$   $\nexists$

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

PB:  $3\bar{x}_1 + 3y_1 + 3y_6 + 3y_8 \geq 3$   $\nexists$

Multiplication by their weight

# PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

## Found “disjoint” cores (RUP)

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

PB:  $2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2$   $\nexists$

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

PB:  $3\bar{x}_1 + 3y_1 + 3y_6 + 3y_8 \geq 3$   $\nexists$

Multiplication by their weight and addition:

$3\bar{x}_1 + 2x_2 + 2x_4 + 5y_1 + 2y_2 + 3y_6 + 3y_8 \geq 5$

# PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

## Found “disjoint” cores (RUP)

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

PB:  $2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2$   $\nexists$

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

PB:  $3\bar{x}_1 + 3y_1 + 3y_6 + 3y_8 \geq 3$   $\nexists$

Multiplication by their weight and addition:

$3\bar{x}_1 + 2x_2 + 2x_4 + 5y_1 + 2y_2 + 3y_6 + 3y_8 \geq 5$

## Model improving constraint

$7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \leq 3$



# PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

## Found “disjoint” cores (RUP)

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

PB:  $2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2$   $\cancel{\chi}$

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

PB:  $3\bar{x}_1 + 3y_1 + 3y_6 + 3y_8 \geq 3$   $\cancel{\chi}$

Multiplication by their weight and addition:

$3\bar{x}_1 + 2x_2 + 2x_4 + 5y_1 + 2y_2 + 3y_6 + 3y_8 \geq 5$

## Model improving constraint

$7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \leq 3$

In normalized form:

$7\bar{y}_1 + 2\bar{y}_2 + 1\bar{y}_3 + 1\bar{y}_4 + 1\bar{y}_5 + 4\bar{y}_6 + 1\bar{y}_7 + 3\bar{y}_8 \geq 20 - 3$

# PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

## Found “disjoint” cores (RUP)

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

PB:  $2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2$   $\cancel{\chi}$

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

PB:  $3\bar{x}_1 + 3y_1 + 3y_6 + 3y_8 \geq 3$   $\cancel{\chi}$

Multiplication by their weight and addition:

$3\bar{x}_1 + 2x_2 + 2x_4 + 5y_1 + 2y_2 + 3y_6 + 3y_8 \geq 5$

## Model improving constraint

$7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \leq 3$

In normalized form:

$7\bar{y}_1 + 2\bar{y}_2 + 1\bar{y}_3 + 1\bar{y}_4 + 1\bar{y}_5 + 4\bar{y}_6 + 1\bar{y}_7 + 3\bar{y}_8 \geq 20 - 3$

Weakening:

$5\bar{y}_1 + 2\bar{y}_2 + 3\bar{y}_6 + 3\bar{y}_8 \geq 13 - 3$

## PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

### Found “disjoint” cores (RUP)

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

PB:  $2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2$   $\cancel{\chi}$

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

PB:  $3\bar{x}_1 + 3y_1 + 3y_6 + 3y_8 \geq 3$   $\cancel{\chi}$

Multiplication by their weight and addition:

$3\bar{x}_1 + 2x_2 + 2x_4 + 5y_1 + 2y_2 + 3y_6 + 3y_8 \geq 5$

Addition:

$3\bar{x}_1 + 2x_2 + 2x_4 + 5y_1 + 5\bar{y}_1 + 2y_2 + 2\bar{y}_2 + 3y_6 + 3\bar{y}_6 + 3y_8 + 3\bar{y}_8 \geq 13 + 5 - 3$

### Model improving constraint

$7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \leq 3$

In normalized form:

$7\bar{y}_1 + 2\bar{y}_2 + 1\bar{y}_3 + 1\bar{y}_4 + 1\bar{y}_5 + 4\bar{y}_6 + 1\bar{y}_7 + 3\bar{y}_8 \geq 20 - 3$

Weakening:

$5\bar{y}_1 + 2\bar{y}_2 + 3\bar{y}_6 + 3\bar{y}_8 \geq 13 - 3$

## PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB** = 4

### Found “disjoint” cores (RUP)

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

PB:  $2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2$   $\chi$

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

PB:  $3\bar{x}_1 + 3y_1 + 3y_6 + 3y_8 \geq 3$   $\chi$

Multiplication by their weight and addition:

$3\bar{x}_1 + 2x_2 + 2x_4 + 5y_1 + 2y_2 + 3y_6 + 3y_8 \geq 5$

### Model improving constraint

$7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \leq 3$

In normalized form:

$7\bar{y}_1 + 2\bar{y}_2 + 1\bar{y}_3 + 1\bar{y}_4 + 1\bar{y}_5 + 4\bar{y}_6 + 1\bar{y}_7 + 3\bar{y}_8 \geq 20 - 3$

Weakening:

$5\bar{y}_1 + 2\bar{y}_2 + 3\bar{y}_6 + 3\bar{y}_8 \geq 13 - 3$

Addition:

$3\bar{x}_1 + 2x_2 + 2x_4 + \cancel{5y_1} + \cancel{5\bar{y}_1} + \cancel{2y_2} + \cancel{2\bar{y}_2} + \cancel{3y_6} + \cancel{3\bar{y}_6} + 3y_8 + \cancel{3\bar{y}_8} \geq \cancel{13} + 5 - 3$

## PROOF LOGGING SOFT CONFLICTS

To Derive:  $\bar{x}_1 + x_2 + x_4 \geq 1$     **UB = 4**

### Found “disjoint” cores (RUP)

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (2)

PB:  $2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2$   $\cancel{\chi}$

Core 2:  $x_1 \rightarrow y_1 \vee y_6 \vee y_8$  (3)

PB:  $3\bar{x}_1 + 3y_1 + 3y_6 + 3y_8 \geq 3$   $\cancel{\chi}$

Multiplication by their weight and addition:

$3\bar{x}_1 + 2x_2 + 2x_4 + 5y_1 + 2y_2 + 3y_6 + 3y_8 \geq 5$

### Model improving constraint

$7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \leq 3$

In normalized form:

$7\bar{y}_1 + 2\bar{y}_2 + 1\bar{y}_3 + 1\bar{y}_4 + 1\bar{y}_5 + 4\bar{y}_6 + 1\bar{y}_7 + 3\bar{y}_8 \geq 20 - 3$

Weakening:

$5\bar{y}_1 + 2\bar{y}_2 + 3\bar{y}_6 + 3\bar{y}_8 \geq 13 - 3$

Addition:

$3\bar{x}_1 + 2x_2 + 2x_4 + 5y_1 + 5\bar{y}_1 + 2y_2 + 2\bar{y}_2 + 3y_6 + 3\bar{y}_6 + 3y_8 + 3\bar{y}_8 \geq 13 + 5 - 3$

Division by  $5 - 3$  and Saturation:  $\bar{x}_1 + x_2 + x_4 \geq 1$

## OUTLINE OF THIS PRESENTATION

- ▶ What is **MaxSAT** and how to certify it?
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## HARDENING

$$\mathcal{O}^t = 7x_1 + 2x_2 + 1x_3 + 1x_4 + 1x_5 + 3x_6 + 1x_7 + 3x_8 \quad \mathbf{UB} = 5$$

**Found local cores**

Core 1:  $\overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_2$  (weight 2)

Core 2:  $x_3 \wedge \overline{x_4} \rightarrow y_1 \vee y_5 \vee y_6$  (weight 1)



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Conclusion:  $\overline{x_2} \wedge x_3 \wedge \overline{x_4} \wedge y_6 \rightarrow LB = 6 \geq 5 = UB$

## HARDENING

$$\mathcal{O}^t = 7 \cancel{5} 4y_1 + \cancel{2} 0y_2 + 1y_3 + 1y_4 + \cancel{1} 0y_5 + \cancel{4} 3y_6 + 1y_7 + 3y_8 \quad \mathbf{UB} = 5$$

**Found local cores**

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$$\overline{x_2} \wedge x_3 \wedge \overline{x_4} \rightarrow \overline{y_6}$$

## HARDENING

$$\mathcal{O}^t = 7x_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 3y_6 + 1y_7 + 3y_8 \quad \mathbf{UB} = 5$$

**Found local cores**

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Conclusion:  $\bar{x}_2 \wedge x_3 \wedge \bar{x}_4 \wedge y_6 \rightarrow LB = 6 \geq 5 = UB$

$$\bar{x}_2 \wedge x_3 \wedge \bar{x}_4 \rightarrow \bar{y}_6$$

Clauses Learned:  $x_2 \vee \bar{x}_3 \vee x_4 \vee \bar{y}_i \quad (i \in \{1, 6, 8\})$

## PROOF LOGGING HARDENING

To Derive:  $x_2 \vee \overline{x_3} \vee x_4 \vee y_1$

---

### Found “disjoint” cores

Core 1:  $\overline{x_2} \wedge \overline{x_4} \rightarrow y_1 \vee y_2$  (weight 2)

$$\text{PB: } x_2 + x_4 + y_1 + y_2 \geq 1$$

Core 2:  $x_3 \wedge \overline{x_4} \rightarrow y_1 \vee y_5 \vee y_6$  (weight 1)

$$\text{PB: } \overline{x_3} + x_4 + y_1 + y_6 + y_8 \geq 1$$

---

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$$\text{PB: } \overline{x_3} + x_4 + y_1 + y_6 + y_8 \geq 1$$

## Model improving constraint

$$7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \leq 3$$

In normalized form:

$$7\overline{y_1} + 2\overline{y_2} + 1\overline{y_3} + 1\overline{y_4} + 1\overline{y_5} + 4\overline{y_6} + 1\overline{y_7} + 3\overline{y_8} \geq 20 - 3$$

# PROOF LOGGING HARDENING

To Derive:  $x_2 \vee \bar{x}_3 \vee x_4 \vee y_1$

## Found “disjoint” cores

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (weight 2)

$$\text{PB: } 2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2 \quad \cancel{1}$$

Core 2:  $x_3 \wedge \bar{x}_4 \rightarrow y_1 \vee y_5 \vee y_6$  (weight 1)

$$\text{PB: } 1\bar{x}_3 + 1x_4 + 1y_1 + 1y_6 + 1y_8 \geq 1$$

## Model improving constraint

$$7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \leq 3$$

In normalized form:

$$7\bar{y}_1 + 2\bar{y}_2 + 1\bar{y}_3 + 1\bar{y}_4 + 1\bar{y}_5 + 4\bar{y}_6 + 1\bar{y}_7 + 3\bar{y}_8 \geq 20 - 3$$

Multiplying cores by their weight and addition with Model-Improving Constraint:

$$2x_2 + 1\bar{x}_3 + 3x_4 + 4\bar{y}_1 + 1\bar{y}_3 + 1\bar{y}_4 + 1\bar{y}_5 + 3\bar{y}_6 + 1\bar{y}_7 + 2\bar{y}_8 \geq 13 + 3 - 3$$

# PROOF LOGGING HARDENING

To Derive:  $x_2 \vee \bar{x}_3 \vee x_4 \vee y_1$

## Found “disjoint” cores

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (weight 2)

$$\text{PB: } 2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2 \quad \cancel{1}$$

Core 2:  $x_3 \wedge \bar{x}_4 \rightarrow y_1 \vee y_5 \vee y_6$  (weight 1)

$$\text{PB: } 1\bar{x}_3 + 1x_4 + 1y_1 + 1y_6 + 1y_8 \geq 1$$

## Model improving constraint

$$7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \leq 3$$

In normalized form:

$$7\bar{y}_1 + 2\bar{y}_2 + 1\bar{y}_3 + 1\bar{y}_4 + 1\bar{y}_5 + 4\bar{y}_6 + 1\bar{y}_7 + 3\bar{y}_8 \geq 20 - 3$$

Multiplying cores by their weight and addition with Model-Improving Constraint:

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Weakening all  $y_i$  with  $i \in \{1, 3, 4, 5, 7, 8\}$ :

$$2x_2 + 1\bar{x}_3 + 3x_4 + 3\bar{y}_6 \geq 3$$

# PROOF LOGGING HARDENING

To Derive:  $x_2 \vee \bar{x}_3 \vee x_4 \vee y_1$

## Found “disjoint” cores

Core 1:  $\bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 \vee y_2$  (weight 2)

$$\text{PB: } 2x_2 + 2x_4 + 2y_1 + 2y_2 \geq 2 \quad \cancel{1}$$

Core 2:  $x_3 \wedge \bar{x}_4 \rightarrow y_1 \vee y_5 \vee y_6$  (weight 1)

$$\text{PB: } 1\bar{x}_3 + 1x_4 + 1y_1 + 1y_6 + 1y_8 \geq 1$$

## Model improving constraint

$$7y_1 + 2y_2 + 1y_3 + 1y_4 + 1y_5 + 4y_6 + 1y_7 + 3y_8 \leq 3$$

In normalized form:

$$7\bar{y}_1 + 2\bar{y}_2 + 1\bar{y}_3 + 1\bar{y}_4 + 1\bar{y}_5 + 4\bar{y}_6 + 1\bar{y}_7 + 3\bar{y}_8 \geq 20 - 3$$

Multiplying cores by their weight and addition with Model-Improving Constraint:

$$2x_2 + 1\bar{x}_3 + 3x_4 + 4\bar{y}_1 + 1\bar{y}_3 + 1\bar{y}_4 + 1\bar{y}_5 + 3\bar{y}_6 + 1\bar{y}_7 + 2\bar{y}_8 \geq 13 + 3 - 3$$

Weakening all  $y_i$  with  $i \in \{1, 3, 4, 5, 7, 8\}$ :

$$2x_2 + 1\bar{x}_3 + 3x_4 + 3\bar{y}_6 \geq 3$$

Division by 3 and saturation:  $x_2 + \bar{x}_3 + x_4 + \bar{y}_6 \geq 1$



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## UNWEIGHTED MAXCDCL REVISITED

Unweighted MaxCDCL searches for set  $\mathcal{L}$  of tuples  $(b, L)$  such that

1. Each  $L$  is a set of objective literals
2. For each  $(b, L)$  in  $\mathcal{L}$ , it holds that  $F \wedge \alpha \models \sum_{\ell \in L} \ell \geq b$ .
3. For each pair  $(b, L)$  and  $(b', L')$  in  $\mathcal{L}$ ,  $L \cap L' = \emptyset$ .
4. The total weight exceeds the current upper bound:  $\sum_{(b,L) \in \mathcal{L}} b \geq \mathbf{UB}$ .

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$$\mathcal{O} = y_1 + y_2 + y_3 + y_4 + y_5 + y_6 + y_7 + y_8 + y_9 + \dots \quad \mathbf{UB} = 4$$

**Found disjoint local “cores”**

$$\text{Core 1: } \bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 + y_3 + y_5 + y_8 \geq 3$$

$$\text{Core 2: } x_4 \wedge \bar{x}_7 \wedge x_9 \rightarrow y_2 + y_4 + y_6 \geq 2$$

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### Found disjoint local “cores”

$$\text{Core 1: } \bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 + y_3 + y_5 + y_8 \geq 3$$

$$\text{Core 2: } x_4 \wedge \bar{x}_7 \wedge x_9 \rightarrow y_2 + y_4 + y_6 \geq 2$$

$$\bar{x}_2 \wedge \bar{x}_4 \wedge \bar{x}_7 \wedge x_9 \rightarrow LB = 5 \geq 4 = UB \quad \text{Soft conflict clause: } x_2 \vee x_4 \vee x_7 \vee \bar{x}_9$$

# LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

$$O^t = \cancel{y_1} + \cancel{y_2} + \cancel{y_3} + \cancel{y_4} + \cancel{y_5} + \cancel{y_6} + \cancel{y_7} + \cancel{y_8} + y_9 + \dots$$

**Trail:**  $x_1^d \ \overline{x_2^d} \ x_3^p \ \overline{x_4^d} \ x_5^p$

## Found disjoint local “cores”

Core 1:  $\overline{x_2} \wedge \overline{x_4} \rightarrow y_3 + y_5 + y_6 \geq 1$

Core 2:  $x_1 \wedge \overline{x_2} \rightarrow y_1 + y_2 + y_4 + y_7 + y_8 \geq 2$

# LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

$$O^t = \cancel{y_1} + \cancel{y_2} + \cancel{y_3} + \cancel{y_4} + \cancel{y_5} + \cancel{y_6} + \cancel{y_7} + \cancel{y_8} + y_9 + \dots$$

**Trail:**  $x_1^d \ \overline{x_2^d} \ x_3^p \ \overline{x_4^d} \ x_5^p \ \overline{y_9^a} \ y_1^p \ y_3^p$

## Found disjoint local “cores”

Core 1:  $\overline{x_2} \wedge \overline{x_4} \rightarrow y_3 + y_5 + y_6 \geq 1$

Core 2:  $x_1 \wedge \overline{x_2} \rightarrow y_1 + y_2 + y_4 + y_7 + y_8 \geq 2$

## LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

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Core 1:  $\overline{x_2} \wedge \overline{x_4} \rightarrow y_3 + y_5 + y_6 \geq 1$  “ $\{y_9\}$  unlocks Core 1 on  $\{y_3\}$ ”

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$$\text{Trail: } x_1^d \overline{x_2^d} x_3^p \overline{x_4^d} x_5^p \overline{y_9^a} y_1^p y_3^p \overline{y_5^a} \overline{y_6^a} y_7^p$$

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Core 2:  $x_1 \wedge \overline{x_2} \rightarrow y_1 + y_2 + y_4 + y_7 + y_8 \geq 2$  “ $\{y_9, y_5, y_6\}$  unlocks Core 2 on  $\{y_1, y_7\}$ ”



## LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

$$O^t = \cancel{y_1} + y_2 + \cancel{y_3} + y_4 + \cancel{y_5} + \cancel{y_6} + \cancel{y_7} + y_8 + y_9 + \dots$$

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## LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

$$\mathcal{O}^t = \cancel{y_1} + \cancel{y_2} + \cancel{y_3} + y_4 + \cancel{y_5} + \cancel{y_6} + \cancel{y_7} + y_8 + y_9 + \dots$$

$$\text{Trail: } x_1^d \overline{x_2^d} x_3^p \overline{x_4^d} x_5^p \overline{y_9^a} y_1^p y_3^p \overline{y_5^a} \overline{y_6^a} y_7^p \overline{y_2^a} \perp$$

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## LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

$$\mathcal{O}^t = \cancel{y_1} + \cancel{y_2} + \cancel{y_3} + y_4 + \cancel{y_5} + \cancel{y_6} + \cancel{y_7} + y_8 + y_9 + \dots$$

$$\text{Trail: } x_1^d \overline{x_2^d} x_3^p \overline{x_4^d} x_5^p \overline{y_9^a} y_1^p y_3^p \overline{y_5^a} \overline{y_6^a} y_7^p \overline{y_2^a} \perp$$

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New core:  $x_1 \wedge \overline{x_4} \rightarrow y_9 + y_5 + y_6 + y_2 \geq 1$

## LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

$$O^t = \cancel{y_1} + \cancel{y_2} + \cancel{y_3} + y_4 + \cancel{y_5} + \cancel{y_6} + \cancel{y_7} + y_8 + y_9 + \dots$$

$$\text{Trail: } x_1^d \overline{x_2^d} x_3^p \overline{x_4^d} x_5^p \overline{y_9^a} y_1^p y_3^p \overline{y_5^a} \overline{y_6^a} y_7^p \overline{y_2^a} \perp$$

## Found disjoint local “cores”

$$\text{Core 1: } \overline{x_2} \wedge \overline{x_4} \rightarrow y_3 + y_5 + y_6 \geq 1 \quad \text{“}\{y_9\} \text{ unlocks Core 1 on } \{y_3\}\text{”}$$

$$\text{Core 2: } x_1 \wedge \overline{x_2} \rightarrow y_1 + y_2 + y_4 + y_7 + y_8 \geq 2 \quad \text{“}\{y_9, y_5, y_6\} \text{ unlocks Core 2 on } \{y_1, y_7\}\text{”}$$

$$\text{New core: } x_1 \wedge \overline{x_4} \rightarrow y_9 + y_5 + y_6 + y_2 \geq 1$$

$$\text{Addition of cores: } x_1 \wedge \overline{x_2} \wedge \overline{x_4} \rightarrow y_1 + 2y_2 + y_3 + y_4 + 2y_5 + 2y_6 + y_7 + y_8 + y_9 \geq 4$$

## LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

$$O^t = \cancel{y_1} + \cancel{y_2} + \cancel{y_3} + y_4 + \cancel{y_5} + \cancel{y_6} + \cancel{y_7} + y_8 + y_9 + \dots$$

$$\text{Trail: } x_1^d \overline{x_2^d} x_3^p \overline{x_4^d} x_5^p \overline{y_9^a} y_1^p y_3^p \overline{y_5^a} \overline{y_6^a} y_7^p \overline{y_2^a} \perp$$

## Found disjoint local “cores”

$$\text{Core 1: } \overline{x_2} \wedge \overline{x_4} \rightarrow y_3 + y_5 + y_6 \geq 1 \quad \text{“}\{y_9\} \text{ unlocks Core 1 on } \{y_3\}\text{”}$$

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$$\text{Addition of cores: } \cancel{x_1 \wedge \overline{x_2} \wedge \overline{x_4} \rightarrow y_1 + 2y_2 + y_3 + y_4 + 2y_5 + 2y_6 + y_7 + y_8 + y_9 \geq 4}$$

## LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

$$O^t = \cancel{y_1} + \cancel{y_2} + \cancel{y_3} + y_4 + \cancel{y_5} + \cancel{y_6} + \cancel{y_7} + \cancel{y_8} + \cancel{y_9} + \dots$$

$$\text{Trail: } x_1^d \bar{x}_2^d x_3^p \bar{x}_4^d x_5^p \bar{y}_9^a y_1^p y_3^p \bar{y}_5^a \bar{y}_6^a y_7^p \bar{y}_2^a \perp$$

## Found disjoint local “cores”

$$\text{Core 1: } \bar{x}_2 \wedge \bar{x}_4 \rightarrow y_3 + y_5 + y_6 \geq 1 \quad \text{“}\{y_9\} \text{ unlocks Core 1 on } \{y_3\}\text{”}$$

$$\text{Core 2: } x_1 \wedge \bar{x}_2 \rightarrow y_1 + y_2 + y_4 + y_7 + y_8 \geq 2 \quad \text{“}\{y_9, y_5, y_6\} \text{ unlocks Core 2 on } \{y_1, y_7\}\text{”}$$

$$\text{New core: } x_1 \wedge \bar{x}_4 \rightarrow y_9 + y_5 + y_6 + y_2 \geq 1$$

$$\text{Addition of cores: } \cancel{x_1 \wedge \bar{x}_2 \wedge \bar{x}_4 \rightarrow y_1 + 2y_2 + y_3 + y_4 + 2y_5 + 2y_6 + y_7 + y_8 + y_9 \geq 4}$$

$$\text{Conclusion } x_1 \wedge \bar{x}_2 \wedge \bar{x}_4 \rightarrow \sum_{i=1}^9 y_i \geq 4 ?$$

## LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

$$O^t = \cancel{y_1} + \cancel{y_2} + \cancel{y_3} + y_4 + \cancel{y_5} + \cancel{y_6} + \cancel{y_7} + \cancel{y_8} + \cancel{y_9} + \dots$$

$$\text{Trail: } x_1^d \bar{x}_2^d x_3^p \bar{x}_4^d x_5^p \bar{y}_9^a y_1^p y_3^p \bar{y}_5^a \bar{y}_6^a y_7^p \bar{y}_2^a \perp$$

## Found disjoint local “cores”

$$\text{Core 1: } \bar{x}_2 \wedge \bar{x}_4 \wedge \bar{y}_9 \rightarrow y_3 + \cancel{y_5} + \cancel{y_6} \geq 1 \quad \text{“}\{y_9\} \text{ unlocks Core 1 on } \{y_3\}\text{”}$$

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$$\text{Conclusion } x_1 \wedge \bar{x}_2 \wedge \bar{x}_4 \rightarrow \sum_{i=1}^9 y_i \geq 4 ?$$

## LOOKAHEAD WITH LITERAL UNLOCKING (BY EXAMPLE)

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$$\text{Trail: } x_1^d \bar{x}_2^d x_3^p \bar{x}_4^d x_5^p \bar{y}_9^a y_1^p y_3^p \bar{y}_5^a \bar{y}_6^a y_7^p \bar{y}_2^a \perp$$

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# CERTIFYING LITERAL UNLOCKING

## Proposition

Let  $L_i |_{1 \leq i \leq k}$  and  $L$  be pairwise disjoint sets of objective literals and  $b_i |_{1 \leq i \leq k}$  natural numbers. Assume  $U_i \subseteq L_i$  with  $|U_i| = b_i$  for each  $i$  and write  $R_i$  for  $L_i \setminus U_i$ . From the constraints

$$L_i \geq b_i \quad (\forall 1 \leq i \leq k), \quad L + \sum_{j < i} R_j + \ell \geq 1 \quad (\forall 1 \leq i \leq k, \ell \in U_i), \quad L + \sum_j R_j \geq 1$$

there is a cutting planes derivation that derives

$$L + \sum_{j \geq i} U_j + \sum_j R_j \geq 1 + \sum_{j \geq i} b_j \tag{1}$$

for each  $i \in \{1, \dots, k + 1\}$ .

## OUTLINE OF THIS PRESENTATION

- ▶ What is **MaxSAT** and how to certify it?
- ▶ Proof logging the B&B solver **MaxCDCL**
- ▶ Proof logging **additional techniques** in MaxCDCL
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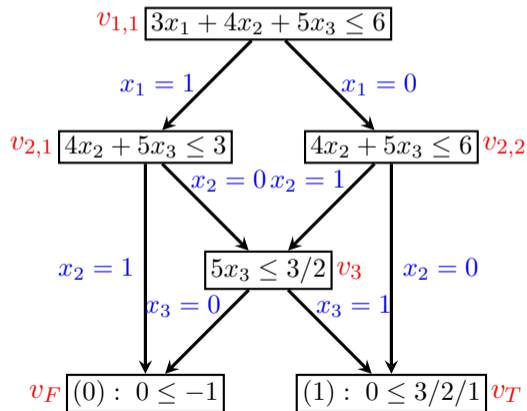
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MaxCDCL  $\cup$  Solution-Improving: MaxCDCL encodes model-improving constraint to **enhance propagation**.

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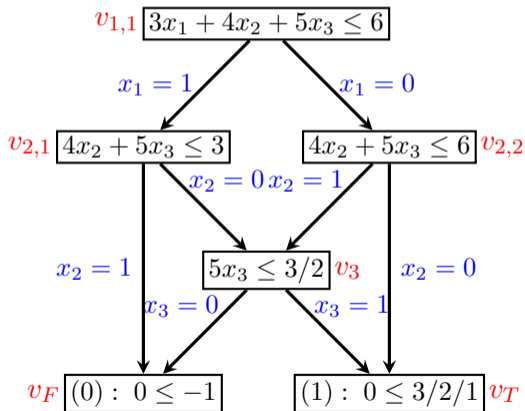


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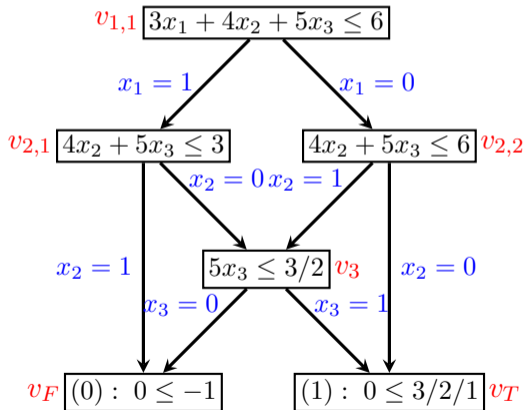


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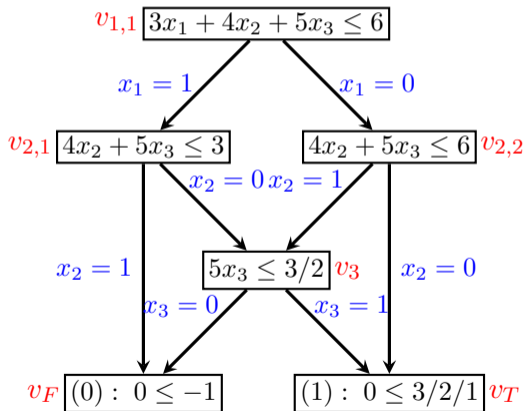


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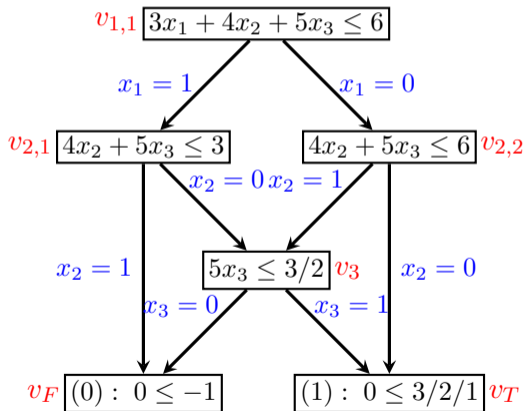


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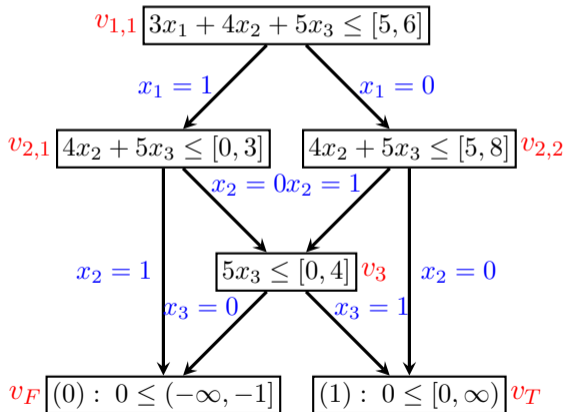


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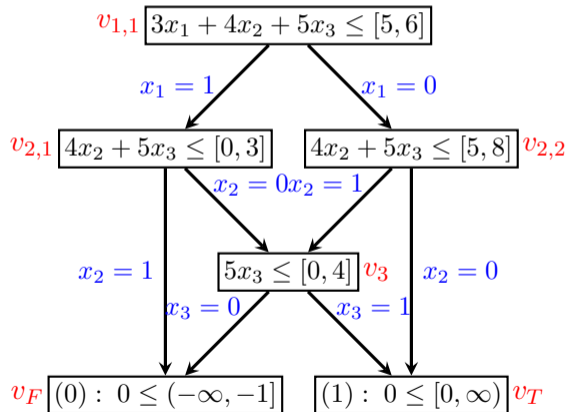


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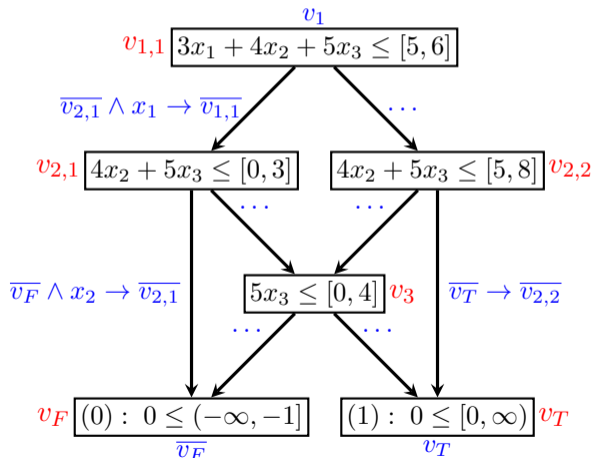
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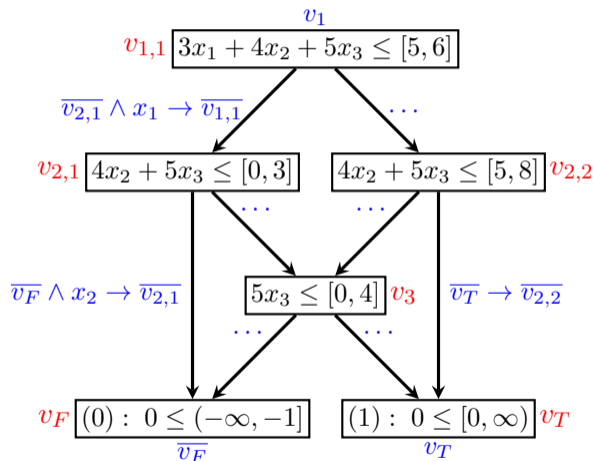
After introducing the reification variables, clauses are added to the solver.



## HOW TO CERTIFY BDDs?

Step 1: Derive reification of node variables. E.g.,

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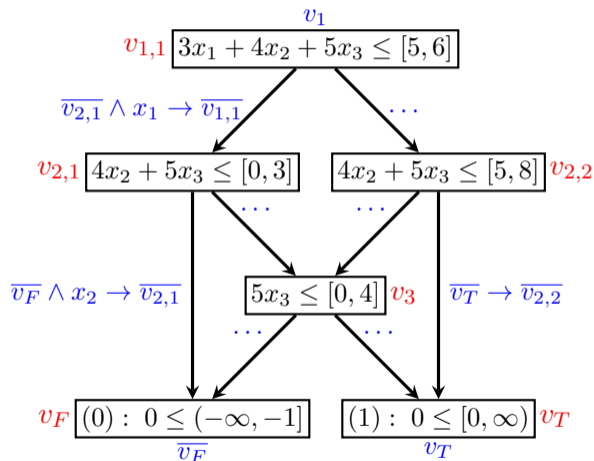
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by introducing

- ▶  $v_{2,2} \leftrightarrow 4x_2 + 5x_3 \leq 5$
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and deriving

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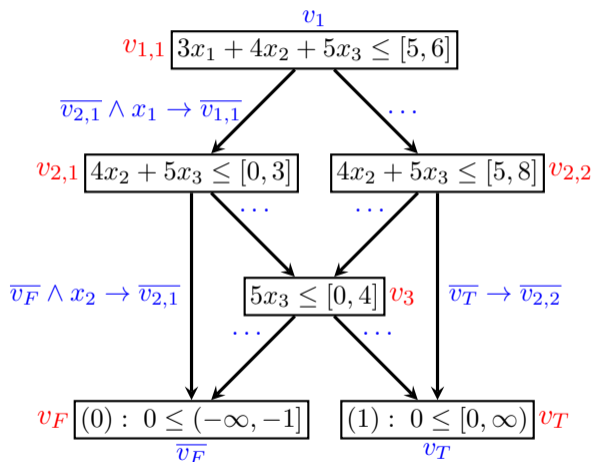
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and deriving

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Step 2: Derive clauses.

- ▶ Straight-forward cutting planes derivation.



## INTERMEZZO: PROOF BY CONTRADICTION

Remember definition of Redundance-Based Strengthening:

### Definition

A constraint  $C$  is redundant with respect to the pseudo-Boolean formula  $F$  if and only if there exists a substitution  $\omega$ , called a witness, such that

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Proof by contradiction — Take empty witness.

Condition to prove RBS becomes:

$$F \wedge \neg C \models F \wedge C$$

Only one non-trivial proof goal:

$$F \wedge \neg C \wedge \neg C \vdash 0 \geq 1$$

## INTERMEZZO: PROOF BY CASE SPLITTING

Suppose we have derived two constraints:

$$a \cdot x + \sum_i b_i l_i \geq B$$

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Luckily, possible by proof by contradiction [Van23].

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We have

▶  $v_{2,2} \rightarrow 4x_2 + 5x_3 \leq 5$

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and we want to derive

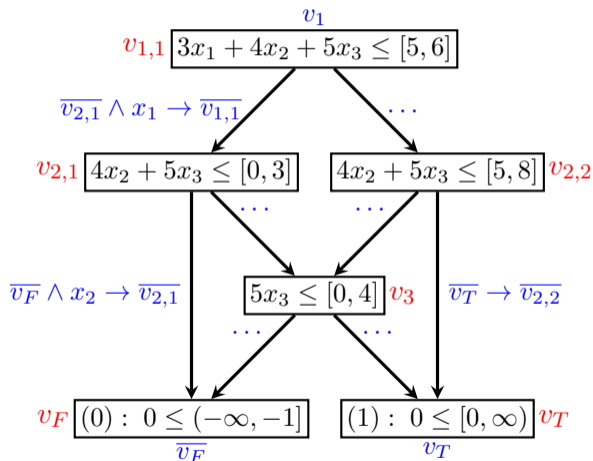
▶  $v'_{2,2} \rightarrow v_{2,2}$

If we can prove

▶  $\bar{x}_2 + \bar{v}'_{2,2} + v_{2,2} \geq 1$

▶  $x_2 + \bar{v}'_{2,2} + v_{2,2} \geq 1$

then by case splitting  $\bar{v}'_{2,2} + v_{2,2} \geq 1$  follows.



# PROVING REIFICATION OF NODE VARIABLES

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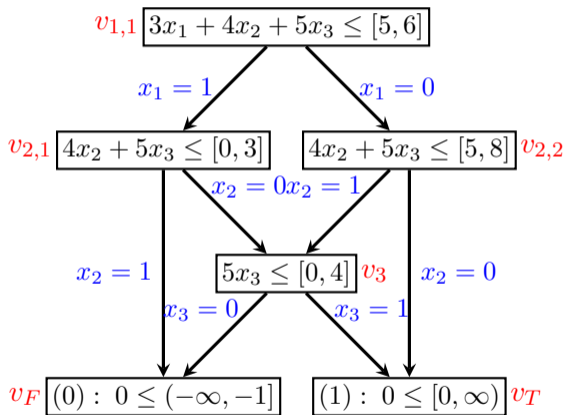
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From  $\bar{v}_{2,2} \geq 1$ :  $4x_2 + 5x_3 \geq 5 + 1$

Weakening  $x_2$ :  $5x_3 \geq 2$

Using definition of  $v_3$ :  $\bar{v}_3 \geq 1$

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Contradiction.

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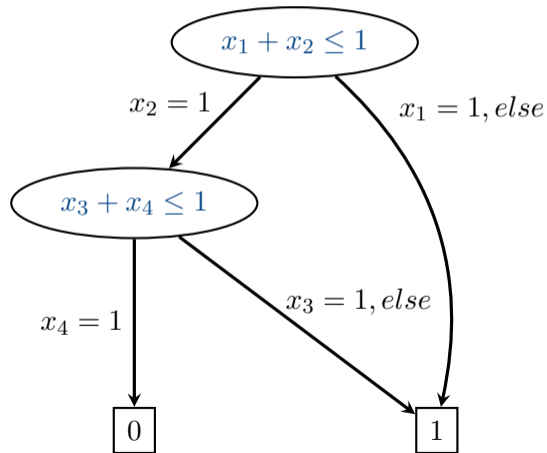
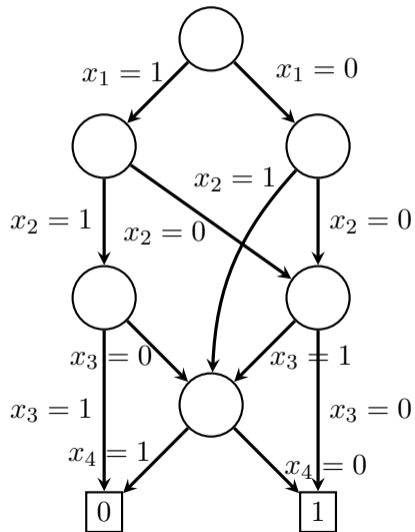
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Using definition of  $v_3$ :  $\bar{v}_3 \geq 1$

Contradiction. Same reasoning to obtain  $x_2 + \bar{v}'_{2,2} + v_{2,2} \geq 1$ .

## MULTI-VALUED DECISION DIAGRAM (MDD)



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Future work:

- ▶ **Implementation** & Experiments
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- ▶ Certified track in **MaxSAT competition?**
- ▶ Other fields of **combinatorial solving** — Interesting things happening!

## WRAPPING UP

This talk:

- ▶ MaxCDCL
  - ▶ MaxSAT solving by combining **Branch-and-Bound** and CDCL
  - ▶ Encoding the model-improving constraint using **MDD encoding**
- ▶ Proof logging is possible with **VeriPB!!**
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*Thank you for your attention!*

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## HOW TO FIND SUCH CORES?

### Definition

Let  $(b, L)$  be a cardinality constraint,  $U \subsetneq L$ , and  $L'$  a set of objective literals disjoint from  $L$ .  $L'$  unlocks  $(b, L)$  on  $U$  if  $|U| \geq b$  and  $F \wedge \alpha \wedge \bigwedge_{\ell \in L'} \bar{\ell} \models \ell'$  for each  $\ell' \in U$ .

Notation:  $(b, L)$  represents the cardinality constraint  $\sum_{\ell \in L} l \geq b$ .

### Example:

$$\mathcal{O} = y_1 + y_2 + y_3 + y_4 + y_5 + y_6 + y_7 + y_8 + y_9 + y_{10}$$

$$\text{Local Core: } y_1 + y_2 + y_3 + y_4 + y_5 + y_6 \geq 3$$

If assigning  $y_7 = y_8 = 0$  propagates literals  $y_1 \wedge y_3 \wedge y_6$ ,

then  $L' = \{y_6, y_7\}$  unlocks  $y_1 + y_2 + y_3 + y_4 + y_5 + y_6 \geq 3$  on  $U = \{y_1, y_3, y_6\}$ .

# CERTIFYING LITERAL UNLOCKING

## Proposition

Let  $L_i |_{1 \leq i \leq k}$  and  $L$  be pairwise disjoint sets of objective literals and  $b_i |_{1 \leq i \leq k}$  natural numbers. Assume  $U_i \subseteq L_i$  with  $|U_i| = b_i$  for each  $i$  and write  $R_i$  for  $L_i \setminus U_i$ . From the constraints

$$L_i \geq b_i \ (\forall 1 \leq i \leq k), \quad L + \sum_{j < i} R_j + \ell \geq 1 \ (\forall 1 \leq i \leq k, \ell \in U_i), \quad L + \sum_j R_j \geq 1$$

there is a cutting planes derivation that derives

$$L + \sum_{j \geq i} U_j + \sum_j R_j \geq 1 + \sum_{j \geq i} b_j \tag{2}$$

for each  $i \in \{1, \dots, k + 1\}$ .

## CERTIFYING LITERAL UNLOCKING

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For  $i = k + 1$  :  $L + \sum_j R_j \geq 1$ .



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For  $i$  between 1 and  $k$  (assuming already derived for  $i + 1$ ):

*Step 1.* Addition of  $L + \sum_{j < i} R_j + \ell \geq 1$  for every  $\ell \in U_i$  results in

$$b_i L + b_i \sum_{j < i} R_j + U_i \geq b_i$$

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*Step 2.* Addition with IH gives:

$$((b_{i+1} + 1) \cdot L + \sum_{j \geq i} U_j + (b_{i+1} + 1) \sum_{j < i} R_j + \sum_{j \geq i} R_j \geq 1 + \sum_{j \geq i} b_j$$

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*Step 3.* Multiplying all constraints  $L_j \geq b_j$  for  $j \geq i$  with  $b_{i+1}$  gives:

$$b_{i+1} \sum_{j \geq i} U_j + b_{i+1} \sum_{j \geq i} R_j \geq b_{i+1} \sum_{j \geq i} b_j$$

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*Step 4.* Addition of constraints from *Step 2* and *Step 3*:

$$(b_{i+1} + 1) \cdot L + (b_{i+1} + 1) \sum_j R_j + (b_{i+1} + 1) \sum_{j \geq i} R_j \geq 1 + (b_{i+1} + 1) \sum_{j \geq i} b_j$$

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*Step 5.* Dividing this by  $b_{i+1} + 1$  (and rounding the righthand-side up) yields

$$L + \sum_j R_j + \sum_{j \geq i} R_j \geq 1 + \sum_{j > i} b_j$$

## INTERMEZZO: PROOF BY CASE SPLITTING

Suppose we have derived two constraints:

$$a \cdot x + \sum_i b_i l_i \geq B$$

$$a \cdot \bar{x} + \sum_i b_i l_i \geq B$$

And we want to derive the constraint

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By contradiction. Needed: CP derivation that shows

$$(a \cdot x + \sum_i b_i l_i \geq B) \wedge (a \cdot \bar{x} + \sum_i b_i l_i \geq B) \wedge \neg(\sum_i b_i l_i \geq B) \vdash 0 \geq 1$$

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After normalization:

$$(a \cdot x + \sum_i b_i l_i \geq B) \wedge (a \cdot \bar{x} + \sum_i b_i l_i \geq B) \wedge (\sum_i b_i l_i \geq \sum_i b_i - B + 1) \vdash 0 \geq 1$$



## INTERMEZZO: PROOF BY CASE SPLITTING

To show:

$$(a \cdot x + \sum_i b_i l_i \geq B) \wedge (a \cdot \bar{x} + \sum_i b_i l_i \geq B) \wedge (\sum_i b_i \bar{l}_i \geq \sum_i b_i - B + 1) \vdash 0 \geq 1$$

Addition of  $(a \cdot x + \sum_i b_i l_i \geq B)$  with  $(\sum_i b_i \bar{l}_i \geq \sum_i b_i - B + 1)$  gives

$$a \cdot x + \sum_i b_i l_i + \sum_i b_i \bar{l}_i \geq B + \sum_i b_i - B + 1$$

which is equal to

$$a \cdot x \geq 1$$

After saturation:  $x \geq 1$ .

Similarly, addition of  $(a \cdot \bar{x} + \sum_i b_i l_i \geq B)$  and  $(\sum_i b_i \bar{l}_i \geq \sum_i b_i - B + 1)$  and saturation gives

$$\bar{x} \geq 1$$

which is clearly contradiction with  $x \geq 1$ .