

# Improved Exact Solver for the Weighted Max-SAT Problem

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# Motivation

- Max-SAT is NP-hard  $\rightarrow$  no efficient algorithm known
- Applications
  - Bioinformatics  $\rightarrow$  protein structure similarity
  - Electronic markets  $\rightarrow$  combinatorial auctions
  - Sports scheduling  $\rightarrow$  break minimization
  - Probabilistic reasoning  $\rightarrow$  Most Probable Explanation (MPE)

# Max-SAT Definition

- Given a list of clauses  $C_1, \dots, C_m$
- Clause consists of disjunction of literals  $(l_1 \vee l_2 \vee \dots \vee l_k)$
- Literal is either  $x_i$  or  $\overline{x_i}$
- Find assignment of Boolean variables  $x_1, \dots, x_n$  that satisfies maximum number of clauses
- Clause is satisfied if at least one literal is assigned true

# Instantiating a variable

- Assign a value to a variable  $x_i$
- Remove literals from the clauses which have been assigned false
- Remove clauses which become satisfied

# Example

$$C_1 = (x_1 \vee x_2 \vee x_6)$$

$$C_2 = (x_2 \vee \bar{x}_6)$$

$$C_3 = (\bar{x}_1 \vee x_3)$$

$$C_4 = (\bar{x}_1 \vee x_4)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

$$C_6 = (x_1 \vee \bar{x}_2 \vee x_5)$$

$$C_7 = (x_1 \vee \bar{x}_2 \vee \bar{x}_5)$$

→ assign  $x_1 = \text{true}$

# Example

$$C_1 = (\text{true} \vee x_2 \vee x_6)$$

$$C_2 = (x_2 \vee \bar{x}_6)$$

$$C_3 = (\text{false} \vee x_3)$$

$$C_4 = (\text{false} \vee x_4)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

$$C_6 = (\text{true} \vee \bar{x}_2 \vee x_5)$$

$$C_7 = (\text{true} \vee \bar{x}_2 \vee \bar{x}_5)$$

# Example

$$C_2 = (x_2 \vee \bar{x}_6)$$

$$C_3 = (x_3)$$

$$C_4 = (x_4)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

# Branch and Bound for Max-SAT

- Search space is a binary tree with  $2^n$  nodes
- Each inner node corresponds to partial assignment
- Leaf nodes correspond to complete assignments
- Branching:
  - Select unassigned variable
  - Assign value
  - Process remaining clauses recursively
  - Assign opposite value
  - Process remaining clauses recursively



# Lower Bound

- Idea: (over)estimate the best possible value of a subtree
- If estimation is worse than best value found so far, subtree can be skipped
- For minimization problems: lower bound  $lb \leq$  minimum value in subtree
- Simplest lower bound Max-SAT: number of clauses unsatisfied by partial assignment
- Better: calculate lower bound by finding disjoint inconsistent subformulas

# Finding inconsistent subformulas by unit propagation

- Unit clauses can only be satisfied by satisfying the literal
- Propagate literals of unit clauses until empty clause is derived
- Reconstruct which clauses are needed to derive empty clause

# Example

$$C_2 = (x_2 \vee \bar{x}_3)$$

$$C_3 = (x_3)$$

$$C_4 = (x_4)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

→ Assign  $x_3 = \text{true}$

# Example

$$C_2 = (x_2 \vee \text{false})$$

$$C_3 = (\text{true})$$

$$C_4 = (x_4)$$

$$C_5 = (\text{false} \vee \bar{x}_4)$$

# Example

$$C_2 = (x_2)$$

$$C_4 = (x_4)$$

$$C_5 = (\bar{x}_4)$$

→ Assign  $x_4 = \text{true}$

# Example

$$C_2 = (x_2)$$

$$C_4 = (\text{true})$$

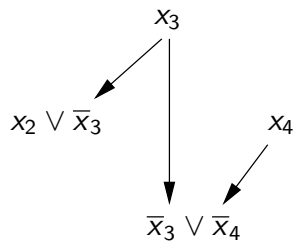
$$C_5 = (\text{false})$$

# Example

$$C_2 = (x_2)$$

$$C_5 = ()$$

# Implication graph





# One step further: failed literal detection

- Add a unit literal  $l$  to the formula
- Use unit propagation to detect inconsistent subformula  $F'$
- $S_l = F' \setminus l$  is resolution proof of  $\bar{l}$
- $l$  is called a failed literal
- If  $l$  and  $\bar{l}$  are failed literals,  $S_l \cup S_{\bar{l}}$  is inconsistent subformula

# Improved algorithm

- Use failed literal detection during unit propagation to get new unit literals
- Whenever no unit literal is left, try to find a failed literal  $l$
- If found
  - extract and store subformula  $S_l$  which is in conflict with  $l$
  - add  $\bar{l}$  to the formula and continue with unit propagation

# Difference between algorithms

- improved algorithm starts with unit propagation and uses failed literal detection in simplified formula
- after finding a failed literal  $l$ , we do not stop if  $\bar{l}$  is no failed literal, but continue in simplified formula after propagating  $\bar{l}$
- improved algorithm can be seen as a restricted sat solver (no branches, only unit clause learning) with unsatisfiable core extraction

# Example

$$C_1 = (x_1 \vee x_2 \vee x_3)$$

$$C_2 = (x_2 \vee \bar{x}_3)$$

$$C_3 = (\bar{x}_1 \vee x_3)$$

$$C_4 = (\bar{x}_1 \vee x_4)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

$$C_6 = (x_1 \vee \bar{x}_2 \vee x_5)$$

$$C_7 = (x_1 \vee \bar{x}_2 \vee \bar{x}_5)$$

→ no unit literal, starting failed literal detection

# Example

$$C_1 = (x_1 \vee x_2 \vee x_3)$$

$$C_2 = (x_2 \vee \bar{x}_3)$$

$$C_3 = (\bar{x}_1 \vee x_3)$$

$$C_4 = (\bar{x}_1 \vee x_4)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

$$C_6 = (x_1 \vee \bar{x}_2 \vee x_5)$$

$$C_7 = (x_1 \vee \bar{x}_2 \vee \bar{x}_5)$$

$$C_8 = (x_1)$$

→ adding  $x_1$  to the formula

→ assign  $x_1 = \text{true}$

# Example

$$C_2 = (x_2 \vee \bar{x}_3)$$

$$C_3 = (x_3)$$

$$C_4 = (x_4)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

→ ... continue as in the example before

# Example

$$S_{x_1} = \{(x_3), (x_4), (\bar{x}_3 \vee \bar{x}_4)\}$$

→ add  $\bar{x}_1$  to the formula

# Example

$$C_1 = (x_1 \vee x_2 \vee x_3)$$

$$C_2 = (x_2 \vee \bar{x}_3)$$

$$C_3 = (\bar{x}_1 \vee x_3)$$

$$C_4 = (\bar{x}_1 \vee x_4)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

$$C_6 = (x_1 \vee \bar{x}_2 \vee x_5)$$

$$C_7 = (x_1 \vee \bar{x}_2 \vee \bar{x}_5)$$

$$C_8 = (\bar{x}_1)$$

→ assign  $x_1 = \text{false}$



# Example

$$C_1 = (x_2 \vee x_3)$$

$$C_2 = (x_2 \vee \bar{x}_3)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

$$C_6 = (\bar{x}_2 \vee x_5)$$

$$C_7 = (\bar{x}_2 \vee \bar{x}_5)$$

→ failed literal detection, add  $x_2$  to the formula

# Example

$$C_1 = (x_2 \vee x_3)$$

$$C_2 = (x_2 \vee \bar{x}_3)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

$$C_6 = (\bar{x}_2 \vee x_5)$$

$$C_7 = (\bar{x}_2 \vee \bar{x}_5)$$

$$C_8 = (x_2)$$

→ assign  $x_2 = \text{true}$

# Example

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

$$C_6 = (x_5)$$

$$C_7 = (\bar{x}_5)$$

→ assign  $x_5 = \text{true}$

# Example

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

$$C_7 = ()$$

→ empty clause found

→  $x_2$  is failed literal

$$\rightarrow S_{x_2} = \{(\bar{x}_2 \vee x_5), (\bar{x}_2 \vee \bar{x}_5)\}$$

# Example

$$C_1 = (x_2 \vee x_3)$$

$$C_2 = (x_2 \vee \bar{x}_3)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

$$C_6 = (\bar{x}_2 \vee x_5)$$

$$C_7 = (\bar{x}_2 \vee \bar{x}_5)$$

$$C_8 = (\bar{x}_2)$$

→ add  $\bar{x}_2$  to the formula

→ assign  $x_2 = \text{false}$

# Example

$$C_1 = (x_3)$$

$$C_2 = (\bar{x}_3)$$

$$C_5 = (\bar{x}_3 \vee \bar{x}_4)$$

→ assign  $x_3 = \text{true}$

# Example

$$C_2 = ()$$

$$C_5 = (\bar{x}_4)$$

→ empty clause found

→ derive inconsistent subformula with implication graph

# Optimizations

- Assign priority order to variables for failed literal detection
- Decrease priority of variables for which a failed literal has been found
- For each variable, check only the literal which occurs in more clauses
- Sometimes more than one failed literal can be detected by only one failed literal detection  $\rightarrow$  use failed literal  $l$  for which  $S_l$  is smallest.



# Example

$$C_1 = \bar{x}_1 \vee x_2$$

$$C_2 = \bar{x}_2 \vee x_3$$

$$C_3 = \bar{x}_3 \vee x_4$$

$$C_4 = \bar{x}_3 \vee x_5$$

$$C_5 = \bar{x}_4 \vee \bar{x}_5$$

$C_3, C_4, C_5$  are resolution proof of  $x_3$

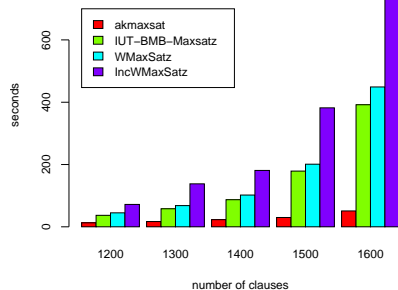
$\rightarrow x_1, x_2$  and  $x_3$  are failed literals

# Data structure

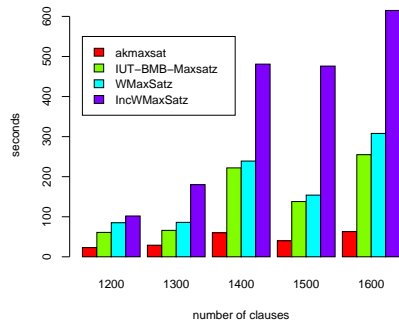
- for each literal keep list of clause pointers to clauses where literal occurs
- support lazy deletion of clause pointers
- each clause has a delete flag and deletion timestamp
- clause pointers are removed during traversal of clause pointer list when delete flag of clause is set to true
- when backtracking, it can be checked in constant time if clause pointer was deleted or not

# Comparison of runtimes

Weighted Max-2-SAT formulas with 100 variables

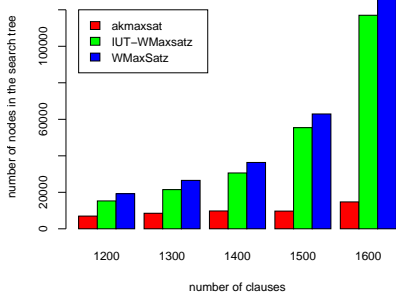


Weighted Max-2-SAT formulas with 120 variables

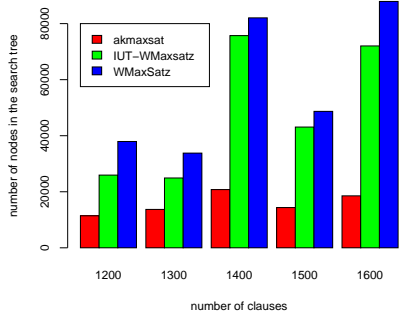


# Comparison of traversed nodes of search tree

Weighted Max-2-SAT formulas with 100 variables



Weighted Max-2-SAT formulas with 120 variables



# Conclusions

- New propagation algorithm improves lower bound
- Smaller part of the search tree needs to be traversed
- Data structure improves runtime for high clauses-to-variables ratio

# Any questions?

Thank you for your attention!