Decision Procedures An Algorithmic Point of View

1

SAT

(slides from http://www.decision-procedures.org/)

Daniel Kroening and Ofer Strichman

Next: Deciding Propositional Formulas

SAT solvers

Binary Decision Diagrams

A Basic SAT algorithm

Given ϕ in CNF: (x,y,z),(-x,y),(-y,z),(-x,-y,-z)



SAT made some progress...



4



Basic Backtracking Search

- Organize the search in the form of a decision tree
 - □ Each node corresponds to a decision
 - □ Definition: Decision Level (DL) is the depth of the node in the decision tree.
 - □ Notation: x=v@d $x \in \{0,1\}$ is assigned to *v* at decision level *d*

Backtracking Search in Action

$$\begin{array}{c}
\omega_{1} = (x_{2} \lor x_{3}) \\
\omega_{2} = (-x_{1} \lor -x_{4}) \\
\omega_{3} = (-x_{2} \lor x_{4}) \\
\end{array}$$

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\end{array}$$

$$\begin{array}{c}
\omega_{3} = (-x_{2} \lor x_{4}) \\
\omega_{3} = 1 @ 1 \Rightarrow x_{4} = 0 @ 1 \Rightarrow x_{2} = 0 @ 1 \\
\Rightarrow x_{3} = 1 @ 1 \\
\Rightarrow x_{3} = 1 @ 1 \\
\end{array}$$

No backtrack in this example, regardless of the decision!

An algorithmic point of view

Backtracking Search in Action



Status of a clause

- A clause can be
 - □ Satisfied: at least one literal is satisfied
 - □ Unsatisfied: all literals are assigned but non are satisfied
 - Unit: all but one literals are assigned but none are satisfied
 Unresolved: all other cases

• Example:
$$C = (x_1 \lor x_2 \lor x_3)$$

x_1	x_2	x_3	C	
1	0		Satisfied	
0	0	0	Unsatisfied	
0	0		Unit	
	0		Unresolved	

Decision heuristics - DLIS

<u>DLIS</u> (Dynamic Largest Individual Sum) – choose the assignment that increases the most the number of satisfied clauses

■ For a given variable *x*:

- \Box C_{xp} # unresolved clauses in which x appears positively
- \Box C_{xn} # unresolved clauses in which x appears negatively
- \Box Let x be the literal for which C_{xp} is maximal
- \Box Let y be the literal for which C_{yn} is maximal
- □ If $C_{xp} > C_{yn}$ choose x and assign it TRUE
- \Box Otherwise choose *y* and assign it FALSE
- Requires l (#literals) queries for each decision.

Decision heuristics - JW

Jeroslow-Wang method

Compute for every clause ω and every variable l (in each phase):

$$\bullet \quad J(l) := \sum_{l \in \omega, \omega \in \varphi} 2^{-|\omega|}$$

- Choose a variable l that maximizes J(l).
- This gives an exponentially higher weight to literals in shorter clauses.

Pause... ||

- We will see other (more advanced) decision Heuristics soon.
- These heuristics are integrated with a mechanism called Learning with Conflict-Clauses, which we will learn next.

Implication graphs and learning: option #1

Current truth assignment: $\{x_9=0@1, x_{10}=0@3, x_{11}=0@3, x_{12}=1@2, x_{13}=1@2\}$ Current decision assignment: $\{x_1=1@6\}$





We learn the *conflict clause* ω_{10} : $(\neg x_1 \lor x_9 \lor x_{11} \lor x_{10})$

Decision Procedures An algorithmic point of view

Implication graph, flipped assignment option #1

$$\begin{split}
\omega_1 &= (\neg x_1 \lor x_2) \\
\omega_2 &= (\neg x_1 \lor x_3 \lor x_9) \\
\omega_3 &= (\neg x_2 \lor \neg x_3 \lor x_4) \\
\omega_4 &= (\neg x_4 \lor x_5 \lor x_{10}) \\
\omega_5 &= (\neg x_4 \lor x_6 \lor x_{11}) \\
\omega_6 &= (\neg x_5 \lor x_6) \\
\omega_7 &= (x_1 \lor x_7 \lor \neg x_{12}) \\
\omega_8 &= (x_1 \lor x_8) \\
\omega_9 &= (\neg x_7 \lor \neg x_8 \lor \neg x_{13}) \\
\omega_{10} &: (\neg x_1 \lor x_9 \lor x_{11} \lor x_{10})
\end{split}$$



where should we backtrack to now ?

An algorithmic point of view

Non-chronological backtracking



Non-chronological backtracking

- So the rule is: backtrack to the largest decision level in the conflict clause.
- This works for both the initial conflict and the conflict after the flip.
- Q: What if the flipped assignment works?A: Change the decision retroactively.

Non-chronological Backtracking



Decision Procedures An algorithmic point of view

More Conflict Clauses

- Def: A Conflict Clause is any clause implied by the formula
- Let *L* be a set of literals labeling nodes that form a cut in the implication graph, separating the conflict node from the roots.
- Claim: $\bigvee_{l \in L} \neg l$ is a Conflict Clause.



1. $(x_{10} \lor \neg x_1 \lor x_9 \lor x_{11})$ 2. $(x_{10} \lor \neg x_4 \lor x_{11})$ 3. $(x_{10} \lor \neg x_2 \lor \neg x_3 \lor x_{11})$

Decision Procedures An algorithmic point of view

Conflict clauses

- How many clauses should we add ?
- If not all, then which ones ?
 - \Box Shorter ones ?
 - □ Check their influence on the backtracking level ?
 - \Box The most "influential"?

Conflict clauses

- Def: An Asserting Clause is a Conflict Clause with a single literal from the current decision level.
 Backtracking (to the right level) makes it a Unit clause.
- Asserting clauses are those that force an immediate change in the search path.
- Modern solvers only consider Asserting Clauses.

Unique Implication Points (UIP's)

- Definition: A Unique Implication Point (UIP) is an internal node in the Implication Graph that all paths from the decision to the conflict node go through it.
- The First-UIP is the closest UIP to the conflict.



Conflict-driven backtracking (option #2)



- Conflict clause: $(x_{10} \lor \neg x_4 \lor x_{11})$
- With standard Non-Chronological Backtracking we backtracked to DL = 6.
- Conflict-driven Backtrack: backtrack to the second highest decision level in the clause (without erasing it).
- In this case, to DL = 3.

• Q: why?

Conflict-driven Non-chronological Backtracking



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Conflict-Driven Backtracking

- So the rule is: backtrack to the second highest decision level *dl*, but do not erase it.
- This way the literal with the currently highest decision level will be implied in DL = dl.
- Q: what if the conflict clause has a single literal ?
 □ For example, from (x∨ ¬y) ∧ (x ∨ y) and decision x=0, we learn the conflict clause (x).

Conflict clauses and Resolution

The Binary-resolution is a sound inference rule:

$$\frac{(a_1 \lor \ldots \lor a_n \lor \beta) \qquad (b_1 \lor \ldots \lor b_m \lor (\neg \beta))}{(a_1 \lor \ldots \lor a_n \lor b_1 \lor \ldots \lor b_m)}$$
(Binary Resolution)

• Example:

$$\frac{(x_1 \lor x_2) \qquad (\neg x_1 \lor x_3 \lor x_4)}{(x_2 \lor x_3 \lor x_4)}$$

Decision Procedures An algorithmic point of view

Conflict clauses and resolution

Consider the following example:



• Conflict clause:
$$c_5$$
: $(x_2 \lor \neg x_4 \lor x_{10})$

Decision Procedures An algorithmic point of view

Conflict clauses and resolution

Conflict clause: c_5 : $(x_2 \lor \neg x_4 \lor x_{10})$

$$c_1 = (\neg x_4 \lor x_2 \lor x_5)$$

$$c_2 = (\neg x_4 \lor x_{10} \lor x_6)$$

$$c_3 = (\neg x_5 \lor \neg x_6 \lor \neg x_7)$$

$$c_4 = (\neg x_6 \lor x_7)$$

$$\vdots \qquad \vdots$$

Resolution order: x_4, x_5, x_6, x_7 T1 = Res(c₄, c₃, x₇) = ($\neg x_5 \lor \neg x_6$)
T2 = Res(T1, c₂, x₆) = ($\neg x_4 \lor \neg x_5 \lor X_{10}$)
T3 = Res(T2, c₁, x₅) = ($x_2 \lor \neg x_4 \lor x_{10}$)



Finding the conflict clause:



Applied to our example:

nam	e <i>cl</i>	lit	var	ante
c_4	$(\neg x_6 \lor x_7)$	x_7	x_7	c_3
	$(\neg x_5 \lor \neg x_6)$	$\neg x_6$	x_6	c_2
	$(\neg x_4 \lor x_{10} \lor \cdot$	$\neg x_5) \neg x_5$	x_5	c_1
c_5	$(\neg x_4 \lor x_2 \lor x_2$	10)		

The Resolution-Graph keeps track of the "inference relation"



The resolution graph

What is it good for ? Example: for computing an **Unsatisfiable core**



[Picture Borrowed from Zhang, Malik SAT'03]

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Resolution graph: example



Decision heuristics - VSIDS

<u>VSIDS</u> (Variable State Independent Decaying Sum)

Each variable in each polarity has a counter initialized to 0.
 When a clause is added, the counters are updated.
 The unassigned variable with the highest counter is chosen.
 Periodically, all the counters are divided by a constant.

(Implemented in Chaff)

Decision heuristics – VSIDS (cont'd)

Chaff holds a list of unassigned variables sorted by the counter value.

Updates are needed only when adding conflict clauses.

Thus - decision is made in constant time.

Decision heuristics <u>VSIDS (cont'd)</u>

VSIDS is a 'quasi-static' strategy:

- *static* because it doesn't depend on current assignment

- *dynamic* because it gradually changes. Variables that appear in recent conflicts have higher priority.

This strategy is a *conflict-driven* decision strategy.

"..employing this strategy dramatically (i.e. an order of magnitude) improved performance ... "

Decision Heuristics - Berkmin

- Keep conflict clauses in a stack
- Choose the first unresolved clause in the stack

 If there is no such clause, use VSIDS
- Choose from this clause a variable + value according to some scoring (e.g. VSIDS)

This gives absolute priority to conflicts.

Berkmin heuristic



tailfirst conflict clause



End of SAT (for now)

Beginning of Binary Decision Diagrams

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