Satisfiability Solving and Tools [original created by Chun-Nan Chou, Chih-Pin Tai]

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Outline

Fundamental concepts

- Algorithms for satisfiability problems
- Decision heuristics
- 📀 Restart
- SAT competitions
- A satisfiability example using MiniSat

Boolean Satisfiability (SAT)

Given a Boolean formula (propositional logic formula), find a variable assignment such that the formula evaluates to 1, or prove that no such assignment exists.

• $F = (a \lor b) \land (\bar{a} \lor \bar{b} \lor c)$

For n variables, there are 2ⁿ possible truth assignments to be checked.



- First established NP-Complete problem.
 - S. A. Cook, The complexity of theorem proving procedures, Proceedings, Third Annual ACM Symp. on the Theory of Computing, 1971.

Boolean Formula

- If a is a Boolean variable, a is also a Boolean formula.
- If g and h are Boolean formulas, then so are:
 - $(g) \lor (h)$
 - $(g) \wedge (h)$
 - 🖲 🖉
- For example:
 - Variables a and b belong to $\{0,1\}$.
 - 🝬 a is a Boolean formula.
 - * \bar{a} , $a \lor b$, $a \land b$ are Boolean formulas.

Satisfiable and Unsatisfiable

- 📀 Given a Boolean formula F
 - Unsatisfiable: for all assignemts such that F = 0.
 - Satisfiable: there exits one assignment such that F = 1.
 - Ex1: F = a is satisfiable.
 - Ex2: $F = a \land b \land (\bar{a} \lor \bar{b})$ is unsatisfiable.

Boolean Satisfiability Solvers

- Boolean SAT solvers have been very successful recent years in the verification area.
 - Cooperate with BDDs
 - Applications: equivalence checking and model checking
 - Applicable even for million-gate designs in EDA
- 📀 Most popular ones
 - MiniSat (2008 winner)
 - http://www.satcompetition.org/

Types of Boolean Satisfiability Solvers

- Conjunctive Normal Form (CNF) Based
 - A Boolean formula is represented as a CNF (i.e., Product of Sums).
 - For example:
 - $(a \lor b \lor c) \land (\bar{a} \lor \bar{b} \lor c) \land (\bar{a} \lor b \lor \bar{c})$
 - To be satisfied, all the clauses should be '1'.
- Oircuit-Based
 - A Boolean formula is represented as a circuit netlist.
 - The SAT algorithm is directly operated on the netlist.

- A conjunction of clauses, where a clause is a disjunction of literals.
- For example, a CNF formula: $(a \lor b \lor c) \land (\bar{a} \lor \bar{b} \lor c)$
 - Variables: a, b, c in this CNF formula.
 - Literals: a, b, c are literals in $(a \lor b \lor c)$. $\overline{a}, \overline{b}, c$ are literals in $(\overline{a} \lor \overline{b} \lor c)$.
 - Clauses: $(a \lor b \lor c)$, $(\bar{a} \lor \bar{b} \lor c)$ in this CNF formula.

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CNF-Based SAT Algorithms

📀 Davis-Putnam (DP), 1960.

- Explicit resolution based
- May explode in memory
- Davis-Putnam-Logemann-Loveland (DPLL), 1962.
 - Search based
 - Most successful, basis for almost all modern SAT solvers
- 📀 GRASP, 1996
 - Conflict driven learning and non-chronological backtracking
- 📀 zChaff, 2001.
 - Boolean constraint propagation (BCP) Algorithm

Davis-Putnam Algorithm

- M. Davis, H. Putnam, "A computing procedure for quantification theory", J. of ACM, 1960. (New York Univ.)
- Three satisfiability-preserving (\approx) transformations in DP:
 - Unit propagation rule
 - 🛚 Pure literal rule
 - Resolutoin rule
- By repeatedly applying these rules, eventually obtain:
 - a formula containing an empty clause indicates unsatisfiability or
 - 🌻 a formula with no clauses indicates satisfiability.

Unit Propagation Rule

- Suppose (a) is a unit clause, i.e. a clause contains only one literal.
 - Remove any instances of a from the formula.
 - Remove all clauses containing a.
- Section Example:

$$\begin{array}{l} \bullet \quad (a) \land (\overline{a} \lor b \lor c) \land (a \lor \overline{b} \lor c) \land (\overline{a} \lor \overline{c} \lor d) \\ \approx (b \lor c) \land (\overline{c} \lor d) \\ \bullet \quad (a) \land (a \lor b) \approx \quad satisfiable \end{array}$$

* (a) \wedge (ā) \approx () unsatisfiable

Pure Literal Rule

- If a literal appears only positively or only negatively, delete all clauses containing that literal.
- Strample: $(\overline{a} \lor b \lor c) \land (\overline{a} \lor \overline{b} \lor c) \land (\overline{b} \lor c \lor d) \land (\overline{a} \lor \overline{c} \lor \overline{d})$ $\approx (\overline{b} \lor c \lor d)$

Resolution Rule

- For a single pair of clauses, $(a \lor l_1 \lor \cdots \lor l_m)$ and $(\bar{a} \lor k_1 \lor \cdots \lor k_n)$, resolution on *a* forms the new clause $(l_1 \lor \cdots \lor l_m \lor k_1 \lor \cdots \lor k_n)$.
- Example: $(a \lor b) \land (\bar{a} \lor c)$ $\approx (b \lor c)$
- If a is true, then for the formula to be true, c must be true.
- If a is false, then for the formula to be true, b must be true.
- So regardless of a, for the formula to be true, $b \lor c$ must be true.

Resolution Rule (cont.)

- Choose a propositional variable p which occurs positively in at least one clause and negatively in at least one other clause.
- Let P be the set of all clauses in which p occurs positively.
- Let N be the set of all clauses in which p occurs negatively.
- Replace the clauses in P and N with those obtained by resolving each clause in P with each clause in N.

An Example

Potential memory explosion problem!

3

Image: A math a math

DPLL Algorithm

- M. Davis, G. Logemann and D. Loveland, "A Machine Program for Theorem-Proving", *Communications of ACM*, 1962. (New York Univ.)
- The basic framework for many modern SAT solvers.
 - Decision Making
 - 🔹 Unit Clause rule
 - Implication
 - Conflict Detection
 - Backtrack

DPLL Algorithm

DPLL Pseudo Code

```
Function DPLL(\Phi, A)
    A \leftarrow Unit - Propagation(\Phi, A);
    if A is inconsistent then
         return UNSAT;
    if A assigns a value to every variable then
         return SAT;
    v \leftarrow a variable not assigned a value by A;
    if DPLL(\Phi, A \cup \{ v = false \}) = SAT
         return SAT;
    else
         return DPLL(\Phi, A \cup \{ v = true \});
```

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$$\begin{array}{l} (\bar{a} \lor b \lor c) \\ (a \lor c \lor d) \\ (a \lor c \lor d) \\ (a \lor \bar{c} \lor d) \\ (a \lor \bar{c} \lor d) \\ (\bar{b} \lor \bar{c} \lor d) \\ (\bar{b} \lor \bar{c} \lor d) \\ (\bar{a} \lor b \lor \bar{c}) \\ (\bar{a} \lor \bar{b} \lor c) \end{array}$$

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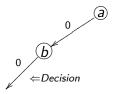
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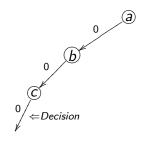


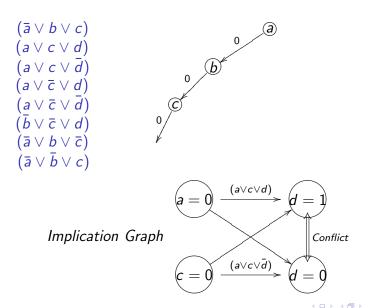
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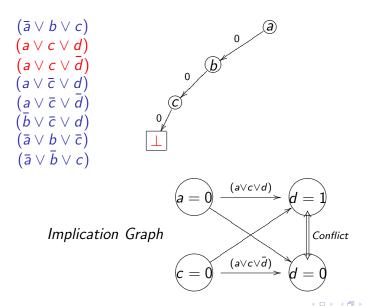
 $\begin{array}{l} (\overline{a} \lor b \lor c) \\ (a \lor c \lor d) \\ (a \lor c \lor d) \\ (a \lor \overline{c} \lor d) \\ (a \lor \overline{c} \lor d) \\ (\overline{b} \lor \overline{c} \lor d) \\ (\overline{b} \lor \overline{c} \lor d) \\ (\overline{a} \lor b \lor \overline{c}) \\ (\overline{a} \lor \overline{b} \lor c) \end{array}$





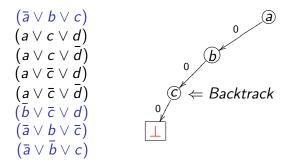


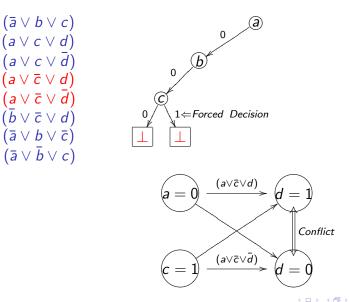




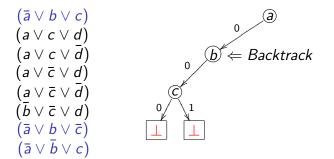
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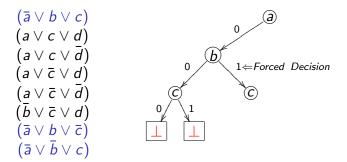
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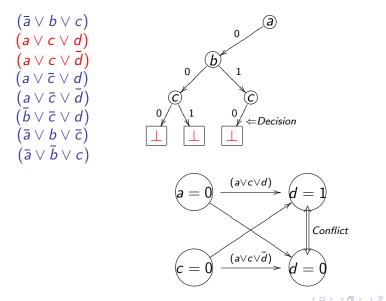




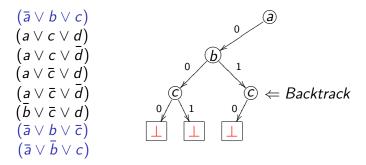
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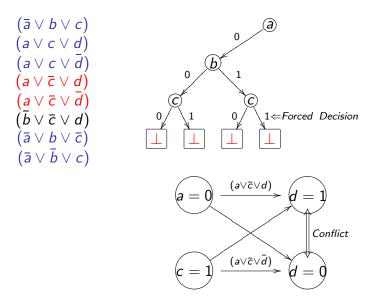




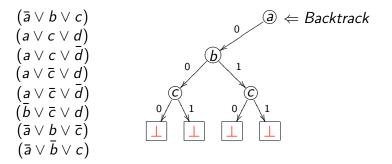


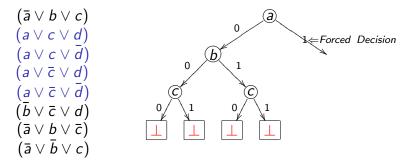
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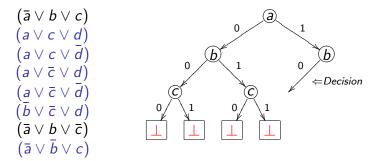




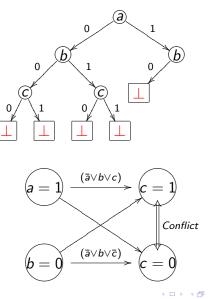
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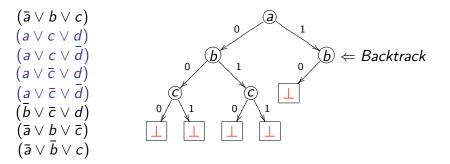


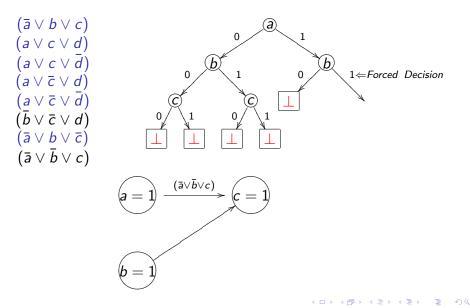


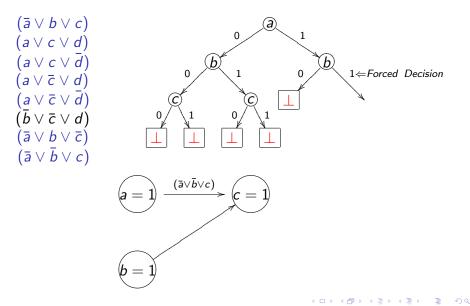


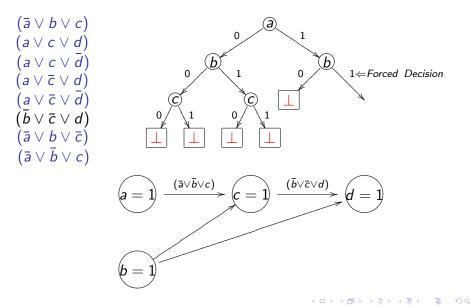


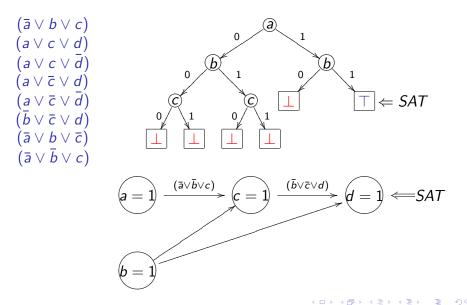












Implications and Unit Clause Rule

- Implication
 - A variable is forced to be True or False based on previous assignments.
- 📀 Unit clause rule
 - A rule for elimination of one-literal clauses
 - An unsatisfied clause is a unit clause if it has exactly one unassigned literal.

 $(a \lor \overline{b} \lor c) \land (b \lor \overline{c}) \land (\overline{a} \lor \overline{c})$ a = T, b = T, c is unassignedSatisfied Literal, Unsatisfied Literal, Unassigned Literal

The unassigned literal is implied because of the unit clause.

Boolean Constraint Propagation

- Boolean Constraint Propagation (BCP)
 - Iteratively apply the unit clause rule until there is no unit clause available.
 - a.k.a. Unit Propagation
- Workhorse of DPLL based algorithms.

Features of DPLL

- Eliminate the exponential memory requirements of DP
- Exponential time is still a problem
- Limited practical applicability largest use seen in automatic theorem proving
- Very limited size of problems are allowed
 - 32K word memory
 - Problem size limited by total size of clauses (about 1300 clauses)

GRASP

Marques-Silva and Sakallah [SS96,SS99] (Univ. of Michigan)

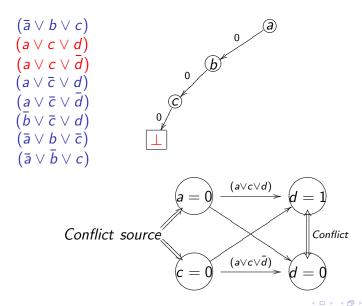
- J. P. Marques-Silva and K. A. Sakallah, "GRASP A New Search Algorithm for Satisfiability", *Proc.ICCAD*, 1996.
- J. P. Marques-Silva and Karem A. Sakallah, "GRASP: A Search Algorithm for Propositional Satisfiability", IEEE Trans. Computers, 1999.
- Incorporate conflict driven learning and non-chronological backtracking
- Practical SAT problem instances can be solved in reasonable time

SAT Improvements

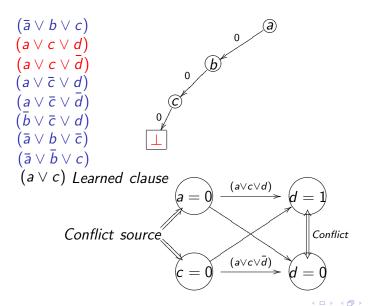
Conflict driven learning

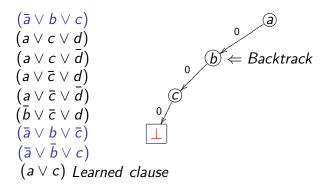
- Once we encounter a conflict, figure out the cause(s) of this conflict and prevent to see this conflict again.
- Add learned clause (conflict clause) which is the negative proposition of the conflict source.
- Non-chronological backtracking
 - After getting a learned clause from the conflict analysis, we backtrack to the "next-to-the-last" variable in the learned clause.
 - Instead of backtracking one decision at a time.

Conflict Driven Learning



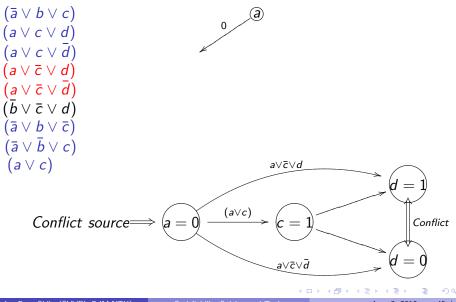
Conflict Driven Learning





• 'a' is the next-to-the-last variable in the learned clause.

Backtrack c=0 and b=0.



$$(\overline{a} \lor b \lor c)$$

$$(a \lor c \lor d)$$

$$(a \lor c \lor d)$$

$$(a \lor \overline{c} \lor d)$$

$$(\overline{a} \lor \overline{c} \lor d)$$

$$(\overline{b} \lor \overline{c} \lor d)$$

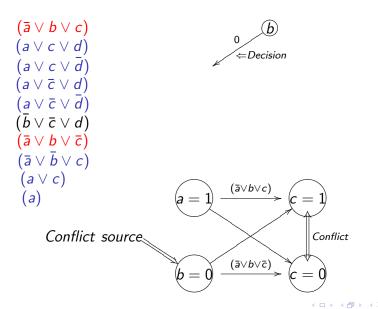
$$(\overline{a} \lor b \lor \overline{c})$$

$$(\overline{a} \lor \overline{b} \lor c)$$

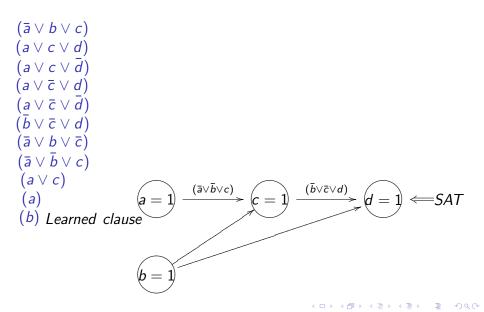
$$(a \lor c)$$

$$(a) Learned clause$$

- Since there is only one variable in the learned clause, no one is the next-to-the-last variable.
- Backtrack all decisions

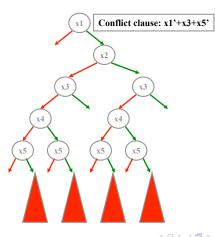


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What's the big deal?

- Significantly prune the search space because learned clause is useful forever!
- Useful in generating future conflict clauses.



Search Completeness

- With conflict driven learning, SAT search is still guaranteed to be complete.
- SAT search becomes a decision stack instead of a binary decision tree.
- When encountering a conflict, the conflict analysis does the following tasks:
 - Learned clause
 - Indicate where to backtrack

SAT Becomes Practical

- Conflict driven learning greatly increases the capacity of SAT solvers (several thousand variables) for structured problems.
- Realistic applications became plausible.
 - Usually thousands and even millions of variables
 - Typical EDA applications can make use of SAT including circuit verfication, FPGA routing and many other applications
- Research direction changes towards more efficient implementations.

zChaff

- M. Moskewicz, C. Madigan, Y. Zhao, L. Zhang, S. Malik," Chaff: Engineering an Efficient SAT Solver" *Proc. DAC 2001*. (UC Berkeley, MIT and Princeton Univ.)
- Make the core operations fast.
 - After profiling, the most time-consuming parts are Boolean Constraint Propagation (BCP) and Decision.
- As always, good search space pruning (i.e. conflict driven learning) is important.

When can BCP occur ?

All literals in a clause but one are assigned to False.

The implied cases of $(v1 \lor v2 \lor v3)$: $(0 \lor 0 \lor v3)$ or $(0 \lor v2 \lor 0)$ or $(v1 \lor 0 \lor 0)$

- ✤ For an N-literal clause, this can only occur after N − 1 of the literals have been assigned to False.
- So, (theoretically) we could completely ignore the first N-2 assignments to this clause.
- In reality, we pick two literals in each clause to "watch" and thus can ignore any assignments to the other literals in the clause.

- Heuristically start with watching two unassigned literals in each clause.
- When one of the two watched literals is assigned True, this clause becomes True.
- When one of the two watched literals is assigned False, we send the clause into an Update-Watch queue to do :
 - 1.updating (another unassigned literal exists)
 - 2.BCP(only one watched literal unassigned)
 - 3.conflict (all literals are False)

Let's illustrate this with an example:

- Green: watched literal
- Initially, we identify any two literals in each clause as the watched ones.
- Clauses of size one are a special case.

$$\begin{array}{c} v2 \lor v3 \lor v1 \lor v4 \lor v5 \\ v1 \lor v2 \lor \overline{v3} \\ \hline{v1 \lor \overline{v2}} \\ \hline{v1 \lor v4} \\ \hline{v1} \\ \hline{\end{array}} \\ \hline \end{array} \\ \hline Detect \ unit \ clause \end{array}$$

• We begin by processing the assignemt v1 = F (which is implied by the size one clause)

 $v^{2} \lor v^{3} \lor v^{1} \lor v^{4} \lor v^{5}$ $v^{1} \lor v^{2} \lor \overline{v^{3}}$ $v^{1} \lor \overline{v^{2}}$ $v^{1} \lor v^{4}$

 Examine each clause where the assignment being processed has set a watched literal to F.

```
\Rightarrow \begin{array}{c} v2 \lor v3 \lor v1 \lor v4 \lor v5 \\ \Rightarrow v1 \lor v2 \lor \overline{v3} \\ \Rightarrow v1 \lor \overline{v2} \\ \overline{v1} \lor \overline{v2} \\ \overline{v1} \lor v4 \end{array}
```

We need not process clauses where a watched literal has been set to T, because the clause is now satisfied and so can not become unit.

 $v2 \lor v3 \lor v1 \lor v4 \lor v5$ $v1 \lor v2 \lor \overline{v3}$ $\Rightarrow \quad \overline{v1} \lor \overline{v2}$ F State : (v1 = F) Pending :

We certainly need not process any clauses where neither watched literal changes state (in this example, where v1 is not watched).

 $\Rightarrow \quad v2 \lor v3 \lor v1 \lor v4 \lor v5$ $v1 \lor v2 \lor \overline{v3}$ $v1 \lor \overline{v2}$ $v1 \lor \overline{v2}$ $\overline{v1} \lor v4$

Now let's actually process the second and third clauses:

 $v^{2} \lor v^{3} \lor v^{1} \lor v^{4} \lor v^{5}$ $v^{1} \lor v^{2} \lor \overline{v^{3}}$ $v^{1} \lor \overline{v^{2}}$ $v^{1} \lor v^{4}$

• For the second clause, we replace v1 with $\overline{v3}$ as a new watched literal because $\overline{v3}$ is not assigned to F.

 $v^{2} \lor v^{3} \lor v^{1} \lor v^{4} \lor v^{5}$ $v^{1} \lor v^{2} \lor \overline{v^{3}}$ $v^{1} \lor v^{2}$ $\overline{v^{1}} \lor v^{4}$

State : (v1 = F)Pending : $v2 \lor v3 \lor v1 \lor v4 \lor v5$ $v1 \lor v2 \lor \overline{v3}$ $v1 \lor \overline{v2}$ $v1 \lor \overline{v2}$ $v1 \lor v4$

• The third clause is unit. We record the new implication of $\overline{v2}$, and add it to the queue of assignments to process.

 $v^{2} \lor v^{3} \lor v^{1} \lor v^{4} \lor v^{5}$ $v^{1} \lor v^{2} \lor \overline{v^{3}}$ $v^{1} \lor \overline{v^{2}}$ $\overline{v^{1}} \lor v^{4}$

State : (v1 = F)Pending : $v^{2} \lor v^{3} \lor v^{1} \lor v^{4} \lor v^{5}$ $v^{1} \lor v^{2} \lor \overline{v^{3}}$ $v^{1} \lor \overline{v^{2}}$ $v^{1} \lor v^{4}$

State : (v1 = F)Pending : (v2 = F)

• Next, we process $\overline{v2}$. We only examine the first two clauses.

- For the first clause, we replace v2 with v4 as a new watched literal since v4 is not assigned to F.
- * The second clause is unit. We record the new implication of $\overline{v3}$, and add it to the queue of assignments to process.

 $\begin{array}{cccc} v2 \lor v3 \lor v1 \lor v4 \lor v5 & v2 \lor v3 \lor v1 \lor v4 \lor v5 \\ v1 \lor v2 \lor \overline{v3} & \Longrightarrow & v1 \lor v2 \lor \overline{v3} \\ \hline v1 \lor \overline{v2} & & & \\ \hline v1 \lor v2 & & & \\ \hline v1 \lor v4 & & & & \\ \hline v1 \lor v4 & & & \\ \hline v1 \lor v4 & & & \\ \end{array}$

State : (v1 = F, v2 = F)Pending :

$$\frac{1}{v1} \lor v4$$
State : $(v1 = F, v2 = F)$
Pending : $(v3 = F)$

• Next, we process $\overline{v3}$. We only examine the first clause.

- For the first clause, we replace v3 with v5 as a new watched literal since v5 is not assigned to F.
- Since there are no pending assignments, and no conflict, BCP terminates and we make a decision. Both v4 and v5 are unassigned. Let's say we decide to assign v4 = T and proceed.

State :
$$(v1 = F, v2 = F, v3 = F)$$
 State : $(v1 = F, v2 = F, v3 = F)$
Pending : $v3 = F)$
Pending :

• Next, we process v4. We do nothing at all.

Since there are no pending assignments, and no conflict, BCP terminates and we make a decision. Only v5 is unassigned. Let's say we decide to assign v5 = F and proceed.

$v2 \lor v3 \lor v1 \lor v4 \lor v5$	$v2 \lor v3 \lor v1 \lor v4 \lor v5$
$v1 \lor v2 \lor \overline{v3}$	$v1 \lor v2 \lor \overline{v3}$
$v1 \vee \overline{v2}$	$v1 \vee \overline{v2}$
$\overline{v1} \vee v4$	$\overline{v1} \lor v4$

State:
$$(v1 = F, v2 = F, v3 = F, State: (v1 = F, v2 = F, v4 = T)$$

 $v3 = F, v4 = T)$

• Next, we process v5 = F. We examine the first clause.

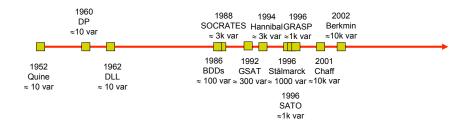
- The first clause is already satisfied by v4 so we ignore it.
- Since there are no pending assignments, and no conflict, BCP terminates and we make a decision. No variables are unassigned, so the instance is SAT, and we are done.

BCP Algorithm Summary

During forward progress: Decisions and Implications

- Only need to examine clauses where watched literal is set to F.
- Ean ignore any clauses where watched literal is set to T.
- Can ignore any assignments to non-watched literals.
- During backtrack: Unwind Assignment Stack.
 - No action is required at all to unassign variables.
 - But it is compute-intensive part in SATO.
- Overall minimize clause access.

The Timeline of the SAT Solver



Jen-Feng Shih (SVVRL @ IM.NTU)

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Outline

- Fundamental concepts
- Algorithms for satisfiability problems
- Decision heuristics
- 📀 Restart
- SAT competitions
- A satisfiability example using MiniSat

Make Decision

- Beause we want to prove that the target Boolean formula is satisfiable or not, we should start with guessing the state (true or false) of a variable until the proof is done.
 - 🔅 Random
 - Dynamic largest individual sum (DLIS)
 - Variable State Independent Decaying Sum (VSIDS)
 - BerkMin

RAND and DLIS

📀 Random

- Simply select the next decision randomly from among the unassigned variables and its value.
- Dynamic largest individual sum (DLIS)
 - Simple and intuitive: At each decision simply choose the assignment that satisfies the most unsatisfied clauses.
 - However, considerable work is required to maintain the statistics necessary for this heuristic.
 - The total effort required for this and similar decision heuristics is much more than for the BCP algorithm in zChaff.

VSIDS

- Variable State Independent Decaying Sum (VSIDS)
 - Each variable in each polarity has a counter which is initialized to zero.
 - When a new clause is added to the database, the counter associated with each literal in this clause is incremented.
 - The (unassigned) variable and polarity with the highest counter is chosen at each decision.
 - Ties are broken randomly by default configuration.
 - Periodically, all the counters are divided by a constant.

VSIDS

- VSIDS attempts to satisfy the conflict clauses but particularly attempts to satisfy recent learned clauses.
- Difficult problems generate many conflicts (and therefore many conflict clauses), the conflict clauses dominate the problem in terms of literal count.
- Since it is independent of the variable state, it has very low overhead.
- The average rum time overhead in zChaff:
 - BCP: about 80%
 - Decision: about 10%
 - Conflict analysis: about 10%

BerkMin

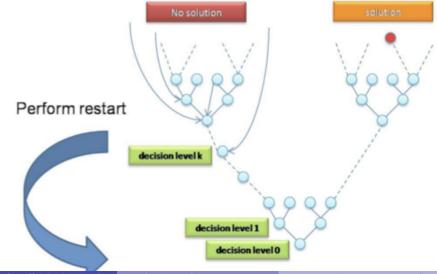
- E. Goldberg, and Y. Novikov, "BerkMin: A Fast and Robust Sat-Solver", Proc. DATE 2002. (Cadence Berkeley Labs and Academy of Sciences in Belarus)
- BerkMin tries to satisfy the most recent clause.
- The clause database is organized as a stack.
- The clauses of the original Boolean formula are located at the bottom of the stack and each new conflict clause is added to the top of the stack.
- The current top clause is the an unsatisfied clause which is the closest to the top of the stack.
- When making decision, choose the most active unassigned variable in the current top clause by using VSIDS.

Outline

- Fundamental concepts
- Algorithms for satisfiability problems
- Decision heuristics
- 📀 Restart
- SAT competitions
- A satisfiability example using MiniSat

Restart Motivation

 Best time to restart: when algorithm spends too much time under a wrong branch



Jen-Feng Shih (SVVRL @ IM.NTU)

Restart

Motivation: avoid spending too much time in "bad" branches.

- no easy-to-find satisfying assignment
- no opportunity for fast learning of strong clauses.
- All modern SAT solvers use a restart policy.
 - Following various criteria, the solver is forced to backtrack to level 0.
 - Abandon the current search tree and reconstruct a new one.
 - The clauses learned prior to the restart are still there after the restart and can help pruning the search space.
- Restarts have crucial impact on performance.
 - Helps reduce variance adds to robustness in the solver.

The basic measure for restarts

- All existing techniques use the number of conflicts learned as of the previous restart.
- The difference is only in the method of calculating the threshold.

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SAT competitions

- The international SAT Competitions http://www.satcompetition.org/
- SAT Race http://baldur.iti.uka.de/sat-race-2010/

SAT 2009 competition

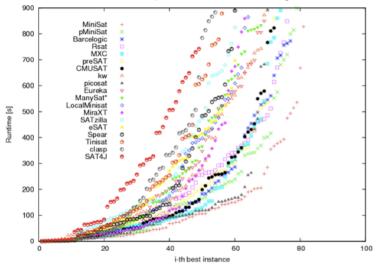
- Application
 - SAT + UNSAT: precosat > glucose > lysat
 - SAT: SATzillal > precosat > MXC
 - UNSAT: glucose > precosat > lysat
- 📀 Crafted
 - SAT + UNSAT: clasp > SATzilla2009_C > IUT_BMB_SAT
 - *SAT*: clasp > SApperloT > MXC
 - UNSAT: SATzilla2009_C > clasp > IUT_BMB_SAT
- 📀 Random
 - SAT + UNSAT: SATzilla2009₋R > March hi
 - SAT: TNM > gNovelty2+ > hybridGM3 / adapt2wsat2009++
 - UNSAT: March hi > SATzilla2009_R

SAT-Race 2010

- The Race
 - The Race itself will take place during or shortly before the SAT'08 conference.
 - Each solver will have to process 100 SAT instances.
 - Per SAT instance and solver a run-time limit of 15 minutes will be imposed.
- Execution Environment
 - Operating System: Scientific Linux 2.6.18, both 32-bit and 64-bit executables supported.
 - Processor(s): 2x Dual-Core Intel Xeon 5150, 2.66 GHz.
 - Memory: 8 GB (7 GB memory limit for solver processes enforced).
 - Cache: 4 MB L2 (shared).
 - Compilers: GCC 4.1.1, javac 1.5.0_11.

SAT-Race 2008

Results: MiniSat 2.1 > pMiniSat > Barcelogic



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Outline

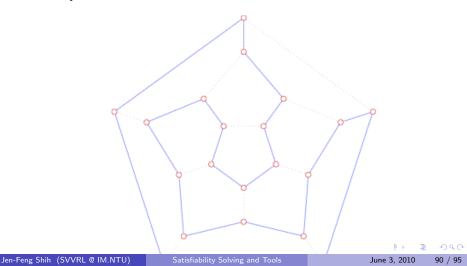
- Fundamental concepts
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The usage of the MiniSat

• Use MiniSat to find a solution to $F = (x_0 \lor x_1 \lor x_2) \land (\overline{x_1} \lor x_2)$.

Hamiltonian Cycle

Hamiltonian cycle, also called a Hamiltonian circuit, is a graph cycle (i.e., closed loop) through a graph that visits each node exactly once.



Encoding

- Encode the Hamiltonian cycle problem into SAT problem by the following way:
 - Assume that there is a path of length n which is the number of nodes.
 - And each Boolean variables x_{i,j} represent the i_{th} node in the j_{th} position of this path.
 - So there are n² Boolean variables in SAT problem by this encoding method.

Add Constraint Clauses

- First constraints: Each node only exist one position of this path.
- Second constraints: Each position of this path contains only one node.
- Third constraints: Two consecutive nodes are connected by an edge.

First Constraints

Each node only exist one position of this path

Each node is in the path:

$$(x_{i,0} \lor x_{i,1} \lor \cdots \lor x_{i,n-1}), \text{ where } 0 \leq i \leq n-1$$

Each node has only position (one hot):

$$\begin{split} & (\overline{x_{i,0}} \vee \overline{x_{i,1}}) \wedge (\overline{x_{i,0}} \vee \overline{x_{i,2}}) \wedge \dots \\ & (\overline{x_{i,0}} \vee \overline{x_{i,n-1}}) \wedge (\overline{x_{i,1}} \vee \overline{x_{i,2}}) \wedge \dots \\ & (\overline{x_{i,j}} \vee \overline{x_{i,k}}) \wedge \dots \\ & \text{where } 0 \le i \le n-1, \ 0 \le j \le n-2, \ j+1 \le k \le n-1 \end{split}$$

Second Constraints

Each position of this path contains only one node

Each position contains nodes:

$$(x_{0,i} \lor x_{1,i} \lor \cdots \lor x_{n-1,i}), \text{ where } 0 \le i \le n-1$$

Each position contains only one node (one hot):

$$\begin{split} & (\overline{x_{0,i}} \vee \overline{x_{1,i}}) \wedge (\overline{x_{0,i}} \vee \overline{x_{2,i}}) \wedge \dots \\ & (\overline{x_{0,i}} \vee \overline{x_{n-1,i}}) \wedge (\overline{x_{1,i}} \vee \overline{x_{2,i}}) \wedge \dots \\ & (\overline{x_{j,i}} \vee \overline{x_{k,i}}) \wedge \dots \\ & \text{where } 0 \leq i \leq n-1, \ 0 \leq j \leq n-2, \ j+1 \leq k \leq n-1 \end{split}$$

Third Constraints

Two consecutive nodes are connected by an edge

There is an edge between the i_{th} node and the j_{th} node:

Don't add constraint clauses into solver.

There are no edge between the i_{th} node and the j_{th} node:

$$\begin{split} & (\overline{x_{i,0}} \vee \overline{x_{j,1}}) \wedge (\overline{x_{i,1}} \vee \overline{x_{j,2}}) \wedge \dots \\ & (\overline{x_{i,n-2}} \vee \overline{x_{j,n-1}}) \\ & \text{where } 0 \leq i \leq n-1, \ 0 \leq j \leq n-1, \text{ and } i \neq j \end{split}$$