The Quest for Efficient Boolean Satisfiability Solvers

Sharad Malik Princeton University



A Brief History of SAT Solvers

Sharad Malik Princeton University



SAT in a Nutshell



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

F = (a + b)(a' + b' + 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, 151-158

Problem Representation

- Conjunctive Normal Form
 - F = (a + b)(a' + b' + c)
 - Simple representation (more efficient data structures)
- Logic circuit representation
 - Circuits have structural and direction information
- Circuit CNF conversion is straightforward





Why Bother?

- Core computational engine for major applications
 - Al
 - Knowledge base deduction
 - Automatic theorem proving
 - EDA
 - Testing and Verification
 - Logic synthesis
 - FPGA routing
 - Path delay analysis
 - And more...



The Timeline



1869: William Stanley Jevons: Logic Machine [Gent & Walsh, SAT2000]

> Pure Logic and other Minor Works – Available at amazon.com!

The Timeline



1960: Davis Putnam Resolution Based ≈10 variables

Resolution



Resolution of a pair of clauses with exactly ONE incompatible variable



Davis Putnam Algorithm



M .Davis, H. Putnam, "A computing procedure for quantification theory", *J. of ACM*, Vol. 7, pp. 201-214, 1960 (335 citations in citeseer)

- Iteratively select a variable for resolution till no more variables are left.
- Can discard all original clauses after each iteration.



Potential memory explosion problem!



The Timeline







The Timeline

1962 Davis Logemann Loveland Depth First Search ≈ 10 var ¹⁹⁶⁰ DP ≈ 10 var

pprox 10 var

DLL Algorithm



• Davis, Logemann and Loveland

M. Davis, G. Logemann and D. Loveland, "A Machine Program for Theorem-Proving", *Communications of ACM*, Vol. 5, No. 7, pp. 394-397, 1962 (231 citations)

- Basic framework for many modern SAT solvers
- Also known as DPLL for historical reasons



- (a' + b + c)
- (a + c + d)
- (a + c + d')
- (a + c' + d)
- (a + c' + d')
- (b' + c' + d)
- (a' + b + c')
- (a' + b' + c)





(a' + b + c) (a + c + d) (a + c + d') (a + c' + d) (a + c' + d') (b' + c' + d) (a' + b + c') (a' + b' + c)



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Conflict!



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Implications and Boolean Constraint Propagation

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

a = T, b = T, c is unassigned

Satisfied Literal Unsatisfied Literal Unassigned Literal

- The unassigned literal is implied because of the unit clause.
- Boolean Constraint Propagation (BCP)
 - Iteratively apply the unit clause rule until there is no unit clause available.
- Workhorse of DLL based algorithms.



Features of DLL

- Eliminates 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 (1300 clauses)


The Timeline



1986 Binary Decision Diagrams (BDDs) ≈100 var



Using BDDs to Solve SAT



- R. Bryant. "Graph-based algorithms for Boolean function manipulation". *IEEE Trans. on Computers*, C-35, 8:677-691, 1986. (1189 citations)
- Store the function in a Directed Acyclic Graph (DAG) representation. Compacted form of the function decision tree.
- Reduction rules guarantee canonicity under fixed variable order.
- Provides for Boolean function manipulation.
- Overkill for SAT.

The Timeline



1992 GSAT Local Search ≈300 Var



Local Search (GSAT, WSAT)



B. Selman, H. Levesque, and D. Mitchell. "A new method for solving hard satisfiability problems". *Proc. AAAI*, 1992. (354 citations)

- Hill climbing algorithm for local search
- Make short local moves
- Probabilistically accept moves that worsen the cost function to enable exits from local minima
- Incomplete SAT solvers
 - Geared towards satisfiable instances, cannot prove unsatisfiability



The Timeline





EDA Drivers (ATPG, Equivalence Checking) start the push for practically useable algorithms! Deemphasize random/synthetic benchmarks.

The Timeline



1996 Stålmarck's Algorithm ≈1000 Var





M. Sheeran and G. Stålmarck "A tutorial on Stålmarck's proof procedure", *Proc. FMCAD*, 1998 (10 citations)

- Algorithm:
 - Using triplets to represent formula
 - Closer to a circuit representation
 - Branch on variable relationships besides on variables
 - Ability to add new variables on the fly
 - Breadth first search over all possible trees in increasing depth



• Try both sides of a branch to find forced decisions (relationships between variables)

• Try both sides of a branch to find forced decisions





• Try both side of a branch to find forced decisions







• Try both sides of a branch to find forced decisions

(a + b) (a' + c) (a' + b) (a + d)

$$a=0 \Rightarrow b=1, d=1$$

$$a=1 \Rightarrow b=1, c=1$$

- Repeat for all variables
- Repeat for all pairs, triples,... till either SAT or UNSAT is proved

The Timeline





GRASP



• Marques-Silva and Sakallah [SS96,SS99]

J. P. Marques-Silva and K. A. Sakallah, "GRASP -- A New Search Algorithm for Satisfiability," Proc. ICCAD 1996. (49 citations)

J. P. Marques-Silva and Karem A. Sakallah, "GRASP: A Search Algorithm for Propositional Satisfiability", *IEEE Trans. Computers*, C-48, 5:506-521, 1999. (19 citations)

- Incorporates conflict driven learning and non-chronological backtracking
- Practical SAT instances can be solved in reasonable time
- Bayardo and Schrag's ReISAT also proposed conflict driven learning [BS97]

R. J. Bayardo Jr. and R. C. Schrag "Using CSP look-back techniques to solve real world SAT instances." *Proc. AAAI*, pp. 203-208, 1997(124 citations)



















































What's the big deal?



Significantly prune the search space – learned clause is useful forever!

Useful in generating future conflict clauses.

Restart

- 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
- Adds to robustness in the solver



Conflict clause: x1'+x3+x5'

SAT becomes practical!

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



The Timeline



$\begin{array}{c} 2001 \\ \text{Chaff} \\ \text{Efficient BCP and decision making} \\ \approx 10 \text{k var} \end{array}$





Large Example: Tough

- Industrial Processor Verification
 - Bounded Model Checking, 14 cycle behavior
- Statistics
 - 1 million variables
 - 10 million literals initially
 - 200 million literals including added clauses
 - 30 million literals finally
 - 4 million clauses (initially)
 - 200K clauses added
 - 1.5 million decisions
 - 3 hours run time

Chaff



 One to two orders of magnitude faster than other solvers...

> M. Moskewicz, C. Madigan, Y. Zhao, L. Zhang, S. Malik, "Chaff: Engineering an Efficient SAT Solver" *Proc. DAC* 2001. (18 citations)

- Widely Used:
 - BlackBox Al Planning

•Henry Kautz (UW)

• NuSMV – Symbolic Verification toolset

A. Cimatti, et. al. "NuSMV 2: An Open Source Tool for Symbolic Model Checking" *Proc. CAV* 2002.

- GrAnDe Automatic theorem prover
- Several industrial licenses

Chaff Philosophy

- Make the core operations fast
 - profiling driven, most time-consuming parts:
 - Boolean Constraint Propagation (BCP) and Decision
- Emphasis on coding efficiency and elegance
- Emphasis on optimizing data cache behavior
- As always, good search space pruning (i.e. conflict resolution and learning) is important



Motivating Metrics: Decisions, Instructions, Cache Performance and Run Time



| | | 1dlx_c_mc_ex_b | p_f |
|---------------------|---------------|----------------|---------------|
| | Num Variables | | 776 |
| | Num Clauses | | 3725 |
| | Num Literals | 10 | 0045 |
| | Z-Chaff | SATO | GRASP |
| # Decisions | 3166 | 3771 | 1795 |
| # Instructions | 86.6M | 630.4M | 1415.9M |
| # L1/L2 accesses | 24M / 1.7M | 188M / 79M | 416M / 153M |
| % L1/L2 misses | 4.8% / 4.6% | 36.8% / 9.7% | 32.9% / 50.3% |
| # Seconds | 0.22 | 4.41 | 11.78 |

BCP Algorithm (1/8)



- What "causes" an implication? When can it occur?
 - All literals in a clause but one are assigned to F
 - (v1 + v2 + v3): implied cases: (0 + 0 + v3) or (0 + v2 + 0) or (v1 + 0 + 0)
 - For an N-literal clause, this can only occur after N-1 of the literals have been assigned to F
 - 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.
 - Example: (v1 + v2 + v3 + v4 + v5)
 - (**v1=X + v2=X** + v3=? {i.e. X or 0 or 1} + v4=? + v5=?)
BCP Algorithm (1.1/8)

- Big Invariants
 - Each clause has two watched literals.
 - If a clause can become newly implied via any sequence of assignments, then this sequence will include an assignment of one of the watched literals to F.
 - Example again: (v1 + v2 + v3 + v4 + v5)
 - (**v1=X** + **v2=X** + v3=? + v4=? + v5=?)
- BCP consists of identifying implied clauses (and the associated implications) while maintaining the "Big Invariants"

BCP Algorithm (2/8)

• Let's illustrate this with an example:

```
v2 + v3 + v1 + v4 + v5
v1 + v2 + v3'
v1 + v2'
v1' + v4
v1'
```



BCP Algorithm (2.1/8)



• Let's illustrate this with an example:



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

BCP Algorithm (3/8)



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



BCP Algorithm (3.1/8)



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



To maintain our invariants, we must examine each clause where the assignment being processed has set a watched literal to F.

BCP Algorithm (3.2/8)



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



- To maintain our invariants, we must examine each clause where the assignment being processed has set a watched literal to F.
- 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 implied.

BCP Algorithm (3.3/8)



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



- To maintain our invariants, we must examine each clause where the assignment being processed has set a watched literal to F.
- 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 implied.
- We certainly need not process any clauses where neither watched literal changes state (in this example, where v1 is not watched).

BCP Algorithm (4/8)



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



BCP Algorithm (4.1/8)



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



For the second clause, we replace v1 with v3' as a new watched literal. Since v3' is not assigned to F, this maintains our invariants.

BCP Algorithm (4.2/8)



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



- For the second clause, we replace v1 with v3' as a new watched literal. Since v3' is not assigned to F, this maintains our invariants.
- The third clause is implied. We record the new implication of v2', and add it to the queue of assignments to process. Since the clause cannot again become newly implied, our invariants are maintained.

BCP Algorithm (5/8)



• Next, we process v2'. We only examine the first 2 clauses.



- For the first clause, we replace v2 with v4 as a new watched literal. Since v4 is not assigned to F, this maintains our invariants.
- The second clause is implied. We record the new implication of v3', and add it to the queue of assignments to process. Since the clause cannot again become newly implied, our invariants are maintained.

BCP Algorithm (6/8)



• Next, we process 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, this maintains our invariants.
- 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.

BCP Algorithm (7/8)



• 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.

BCP Algorithm (8/8)



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



- The first clause is implied. However, the implication is v4=T, which is a duplicate (since v4=T already) 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 problem is sat, and we are done.

The Timeline



1996 SATO Head/tail pointers ≈1k var



SATO



H. Zhang, M. Stickel, "An efficient algorithm for unit-propagation" *Proc. of the Fourth International Symposium on Artificial Intelligence and Mathematics,* 1996. (7 citations)

H. Zhang, "SATO: An Efficient Propositional Prover" *Proc. of International Conference on Automated Deduction*, 1997. (40 citations)

- The Invariants
 - Each clause has a head pointer and a tail pointer.
 - All literals in a clause before the head pointer and after the tail pointer have been assigned false.
 - If a clause can become newly implied via any sequence of assignments, then this sequence will include an assignment to one of the literals pointed to by the head/tail pointer.



Chaff: v1 + v2' + v4 + v5 + v8' + v10 + v12 + v15



Chaff: v1 + v2' + v4 + v5 + v8' + v10 + v12 + v15



Chaff: v1 + v2' + v4 + v5 + v8' + v10 + v12 + v15



Chaff: $v1 + v2^2 + v4 + v5 + v8^2 + v10 + v12 + v15$





Chaff: v1 + v2' + v4 + v5 + v8' + v10 + v12 + v15



Chaff: v1 + v2' + v4 + v5 + v8' + v10 + v12 + v15

Backtrack

BCP Algorithm Summary

- During forward progress: Decisions and Implications
 - Only need to examine clauses where watched literal is set to F
 - Can ignore any assignments of literals to T
 - Can ignore any assignments to non-watched literals
- During backtrack: Unwind Assignment Stack
 - Any sequence of chronological unassignments will maintain our invariants
 - So no action is required at all to unassign variables.
- Overall
 - Minimize clause access



Decision Heuristics – Conventional Wisdom



- DLIS is a relatively simple dynamic decision heuristic
 - 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 – for one implementation:
 - Must touch *every* clause that contains a literal that has been set to true.
 Often restricted to initial (not learned) clauses.
 - Maintain "sat" counters for each clause
 - When counters transition $0 \rightarrow 1$, update rankings.
 - Need to reverse the process for unassignment.
 - The total effort required for this and similar decision heuristics is *much more* than for our BCP algorithm.
- Look ahead algorithms even more compute intensive

C. Li, Anbulagan, "Look-ahead versus look-back for satisfiability problems" *Proc. of CP*, 1997. (7 citations)

Chaff Decision Heuristic - VSIDS

- Variable State Independent Decaying Sum
 - Rank variables by literal count in the initial clause database
 - Only increment counts as new clauses are added.
 - Periodically, divide all counts by a constant.
- Quasi-static:
 - Static because it doesn't depend on var state
 - Not static because it gradually changes as new clauses are added
 - Decay causes bias toward *recent* conflicts.
- Use heap to find unassigned var with the highest ranking
 - Even single linear pass though variables on each decision would dominate run-time!
- Seems to work fairly well in terms of # decisions
 - hard to compare with other heuristics because they have too much overhead

Interplay of BCP and Decision

- This is only an intuitive description ...
 - Reality depends heavily on specific instance
- Take some variable ranking (from the decision engine)
 - Assume several decisions are made
 - Say v2=T, v7=F, v9=T, v1=T (and any implications thereof)
 - Then a conflict is encountered that forces v2=F
 - The next decisions may still be v7=F, v9=T, v1=T !
 - But the BCP engine has recently processed these assignments ... so these variables are unlikely to still be watched.
 - Thus, the BCP engine *inherently does a differential update.*
 - And the Decision heuristic makes differential changes more likely to occur in practice.
- In a more general sense, the more "active" a variable is, the more likely it is to *not* be watched.

The Timeline



2002 BerkMin Emphasize clause activity ≈10k var



Post Chaff Improvements — BerkMin



- E. Goldberg, and Y. Novikov, "BerkMin: A Fast and Robust Sat-Solver", *Proc. DATE* 2002, pp. 142-149.
- Decision strategy
 - Make decisions on literals that are more recently active
 - Measure a literal's activity based on its appearance in both conflict clauses and the antecedent clauses of conflict clauses
- Clause deletion strategy
 - More aggressive than that in Chaff
 - Delete clauses not only based on their length but also on their involvement in resolving conflicts

BerkMin

• Emphasize active clauses in deciding variables





BerkMin



• Emphasize active clauses in deciding variables



Chaff measures a literal's activity only by its appearances in conflict clauses

BerkMin



• Emphasize active clauses in deciding variables



BerkMin measures a literal's activity by its appearances in clauses involved in conflicts



Utility of a Learned Clause



- Utility Metric is the number of times a clause is involved in generating a new useful (conflict generating) clause.
- Most clauses have zero utility metric.
 - They are not useful for proving unsatisfiability!
 - They shouldn't be kept in database!

Utility of a Learned Clause



• If a clause is useful, it will usually be used soon.



The Timeline







Post Chaff Improvements — 2CLS+EQ



F. Bacchus "Exploring the Computational Tradeoff of more Reasoning and Less Searching", *Proc. 5th Int. Symp. Theory and Applications of Satisfiability Testing*, pp. 7-16, 2002.

- Extensive Reasoning at each node of the search tree
 - Hyper-resolution
 - $x_1+x_2+\cdots+x_n$, x_1+y , x_2+y , \cdots , $x_{n-1}+y$ resolved as x_n+y
 - Hyper resolution detects the same set of forced literals as iteratively doing the failed literal tests
 - Equality reduction
 - If formula F contains a'+b and a+b', then replace every occurrence of a(b) with b(a) and simplify F
- Demonstrate that deduction techniques other than UP (Unit Propagation) can pay off in terms of run time.
- Scalability with increasing problem size?
Summary



- Rich history of emphasis on practical efficiency.
- Presence of drivers results in maximum progress.
- Need to account for computation cost in search space pruning.
- Need to match algorithms with underlying processing system architectures.
- Specific problem classes can benefit from specialized algorithms
 - Identification of problem classes?
 - Dynamically adapting heuristics?
- We barely understand the tip of the iceberg here much room to learn and improve.

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