SAT Solver Heuristics

SAT-solver History

- Started with David-Putnam-Logemann-Loveland (DPLL) (1962)
 - Able to solve 10-15 variable problems
- Satz (Chu Min Li, 1995)
 - Able to solve some 1000 variable problems
- Chaff (Malik et al., 2001)
 - Intelligently hacked DPLL, Won the 2004 competition
 - Able to solve some 10000 variable problems
- Current state-of-the-art
 - MiniSAT and SATELITEGTI (Chalmer's university, 2004-2006)
 - Jerusat and Haifasat (Intel Haifa, 2002)
 - Ace (UCLA, 2004-2006)

MiniSAT

- MiniSat is a fast SAT solver developed by Niklas Eén an d Niklas Sörensson
 - MiniSat won all industrial categories in SAT 2005 competition
 - MiniSat won SAT-Race 2006
- MiniSat is simple and well-documented
 - Well-defined interface for general use
 - Helpful implementation documents and comments
 - Minimal but efficient heuristic
 - Main.C (344 lines)
 - Solver.C (741 lines)

Overview (1/2)

- A set of propositional variables and CNF clauses involving variables
 - $-(x_1 \vee x_1' \vee x_3) \wedge (x_2 \vee x_1' \vee x_4)$
 - $-x_1, x_2, x_3$ and x_4 are variables (true or false)
- Literals: Variable and its negation
 - $-x_1$ and x_1
- A clause is satisfied if one of the literals is true
 - x₁=true satisfies clause 1
 - x₁=false satisfies clause 2
- Solution: An assignment that satisfies all clauses

Overview (2/2)

- Unit clause is a clause in which all but one of literals is assigned to False
- Unit literal is the unassigned literal in a unit clause

$$(x_0) \land$$

 $(-x_0 \lor x_1) \land$
 $(-x_2 \lor -x_3 \lor -x_4) \land$

.....

- $-(x_0)$ is a unit clause and ' x_0 ' is a unit literal
- $(-x_0 \lor x_1)$ is a unit clause since x_0 has to be True
- (- x_2 \vee - x_3 \vee - x_4) can be a unit clause if the current assignment is that x_3 and x_4 are True
- Boolean Constrain Propagation(BCP) is the process of assigning the True value to all unit literals

DPLL Overview (1/3)

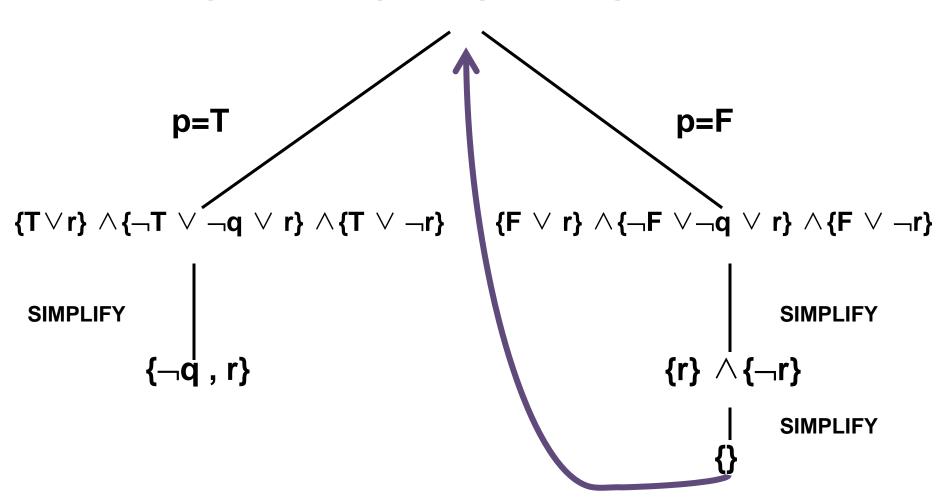
```
/* The Quest for Efficient Boolean Satisfiability Solvers
* by L.Zhang and S.Malik, Computer Aided Verification 2002 */
DPLL(a formula ', assignment) {
   necessary = deduction(', assignment);
   new_asgnment = union(necessary, assignment);
   if (is_satisfied(', new_asgnment))
         return SATISFIABLE;
   else if (is_conflicting(', new_asgnmnt))
         return UNSATISFIABLE:
   var = choose_free_variable(', new_asgnmnt);
   asgn1 = union(new_asgnmnt, assign(var, 1));
   if (DPLL(', asgn1) == SATISFIABLE)
         return SATISFIABLE;
   else {
         asgn2 = union (new_asgnmnt, assign(var,0));
         return DPLL (', asgn2);
```

Three techniques added to modern SAT solvers

- 1. Learnt clauses
- 2. Non-chronological backtracking
- 3. Restart

DPLL Overview (2/3)

$$\{p \lor r\} \land \{\neg p \lor \neg q \lor r\} \land \{p \lor \neg r\}$$



DPLL Overview (3/3)

```
/* overall structure of Minisat solve procedure */
Solve(){
   while(true){
        boolean_constraint_propagation();
        if(no_conflict){
                 if(no_unassigned_variable) return SAT;
                 make_decision();
        }else{
                 if (no_decisions_made) return UNSAT;
                 analyze_conflict();
                 undo_assignments();
                 add_conflict_clause();
```

Conflict Clause Analysis (1/10)

 A conflict happens when one clause is falsified by unit propagation

```
Assume x_4 is False (x_1 \lor x_4) \land (-x_1 \lor x_2) \land (-x_2 \lor x_3) \land (-x_3 \lor -x_2 \lor -x_1) Falsified! Omitted clauses
```

- Analyze the conflicting clause to infer a clause
 - $(-x_3 \lor -x_2 \lor -x_1)$ is conflicting clause
- The inferred clause is a new knowledge
 - A new learnt clause is added to constraints

Conflict Clause Analysis (2/10)

Learnt clauses are inferred by conflict analysis

```
(x_1 \lor x_4) \land

(-x_1 \lor x_2) \land

(-x_2 \lor x_3) \land

(-x_3 \lor -x_2 \lor -x_1) \land

omitted clauses \land
```

- They help prune future parts of the search space
 - Assigning False to x₄ is the casual of conflict
 - Adding (x₄) to constraints prohibit conflict from -x₄
- Learnt clauses actually drive backtracking

Conflict Clause Analysis (3/10)

```
/* conflict analysis algorithm */
Analyze_conflict(){
    cl = find conflicting clause();
    /* Loop until cl is falsified and one literal whose value is determined in current
    decision level is remained */
    While(!stop criterion met(cl)){
          lit = choose literal(cl); /* select the last propagated literal */
          Var = variable_of_literal(lit);
           ante = antecedent(var);
          cl = resolve(cl, ante, var);
    add clause to database(cl);
    /* backtrack level is the lowest decision level for which the learnt clause is unit
    clause */
    back_dl = clause_asserting_level(cl);
    return back dl;
                                                   Algorithm from Lintao Zhang and Sharad malik
                                                   "The Quest for Efficient Boolean Satisfiability Solvers"
```

Conflict Clause Analysis (4/10)

Example of conflict clause analysis

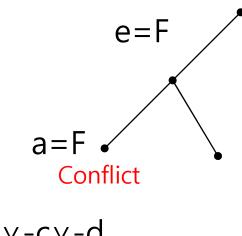
```
- a, b, c, d, e, f, g, and h: 8 variables (28 cases)
```

Satisfiable?

Unsatisfiable?

Conflict Clause Analysis (5/10)

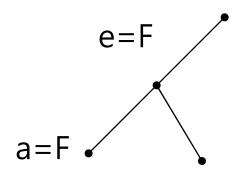
Assignments	antecedent	
e=F)	assumption	
f=F	-f∨e	
g=F DLevel=1	-g∨f	
h=F	-hvg	
a=F \	assumption	
b=T DLevel=2	b∨a∨e	
c=T	cvevfv-b	
d=T 🕽	d∨-b∨h	



Example slides are from CMU 15-414 course ppt

Conflict Clause Analysis (6/10)

Assignments	antecedent	
e=F)	assumption	
f=F Dlevel=1	-f∨e	
g=F	-g∨f	
h=F	-h∨g	
a=F \	assumption	
b=T	b∨a∨e	
c=T DLevel=2	cvevfv-b	
d=T	<mark>d</mark> v-bvh	



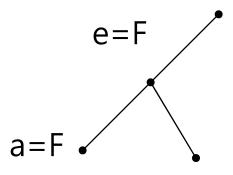
$$-b \vee -c \vee -d$$

Resolution

- Resolution is a process to generate a clause from two clauses
- Given two clauses (x \(\nabla \) y) and (-y \(\nabla \) z),
 the resolvent of these two clauses is (
 x \(\nabla \) z)
 - $-(x \lor y) \land (-y \lor z)$ is satisfiable iff $(x \lor y) \land (-y \lor z) \land (x \lor z)$ is satisfiable
 - The resolvent is redundant

Conflict Clause Analysis (7/10)

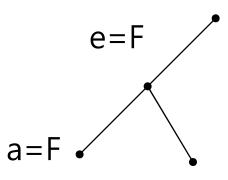
Assign	ments	antecedent
e=F)		assumption
f=F	► DLevel=1	-f∨e
g=F	- DLEVEI-1	-g∨f
h=F 🕽		-hvg
a=F ¬		assumption
b=T		b∨a∨e
c=T	➤ DLevel=2	cvevfv-b
d=T 🗸		dv-bvh



-bv-cvh
(a resolvent of
-bv-cv-d
and dv-bvh)

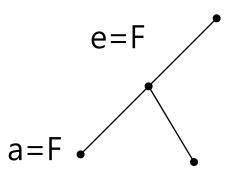
Conflict Clause Analysis (8/10)

Assign	ments	antecedent
e=F)		assumption
f=F	► DLevel=1	-f∨e
g=F	- DLEVEI-1	-g∨f
h=F 🕽		-hvg
a=F ¬		assumption
b=T		b∨a∨e
c=T	➤ DLevel=2	cvevfv-b
d=T 🔎		dv-bvh



Conflict Clause Analysis (9/10)

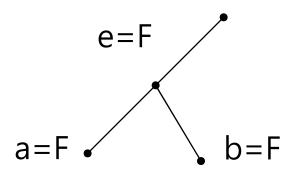
Assignments	antecedent
e=F)	assumption
f=F Dlevel=1	-f∨e
g=F	-gvf
h=F	-hvg
a=F \	assumption
b=T	b∨a∨e
c=T DLevel=2	cvevfv-b
d=T	dv-bvh



-bvevfvh learnt clause

Conflict Clause Analysis (10/10)

•	Assignments		antecedent
	e=F)		assumption
·	f=F		-fve
,	g=F	DLevel=1	-g∨f
	h=F		-hvg
New assign ment@	gn b=F /		-bvevfvh
level 1	•••		•••



-bv-cv-d

-bv-cvh

-bvevfvh

Variable State Independent Decaying Sum(VSIDS)

- Decision heuristic to determine what variable will be assigned next
- Decision is independent from the current assignment of each variable
- VSIDS makes decisions based on activity
 - Activity is a literal occurrence count with higher weight on the more recently added clauses
 - MiniSAT does not consider any polarity in VSIDS
 - Each variable, not literal has score

activity description from Lintao Zhang and Sharad malik "The Quest for Efficient Boolean Satisfiability Solvers"

VSIDS Decision Heuristic – MiniSAT style (1/8)

- Initially, the score for each variable is 0
- First make a decision e = False
 - The order between same score is unspecified.
 - MiniSAT always assigns False to variables.

Initial constraints (-fve) ^ (-gvf) ^ (bvave) ^ (cvevfv-b) ^ (-hvg) ^ (dv-bvh) ^ (-bv-cv-d) ^ (cvd)

Variable	Score	Value
а	0	
b	0	
С	0	
d	0	
е	0	F
f	0	
g	0	
h	0	

VSIDS Decision Heuristic (2/8)

f, g, h are False after BCP

```
(-fve) \( \)
(-g\times f) \( \)
(b\times a\times e) \( \)
(c\times vf\times -b) \( \)
(-h\times g) \( \)
(d\times -b\times h) \( \)
(-b\times -c\times -d) \( \)
(c\times d)
```

Variable	Score	Value
a	0	
b	0	
С	0	
d	0	
е	0	F
f	0	F
g	0	F
h	0	F

VSIDS Decision Heuristic (3/8)

a is next decision variable

```
(-fve) \( \)
(-gvf) \( \)
(bvave) \( \)
(cvevfv-b) \( \)
(-hvg) \( \)
(dv-bvh) \( \)
(-bv-cv-d) \( \)
(cvd)
```

Variable	Score	Value
а	0	F
b	0	
С	0	
d	0	
е	0	F
f	0	F
g	0	F
h	0	F

VSIDS Decision Heuristic (4/8)

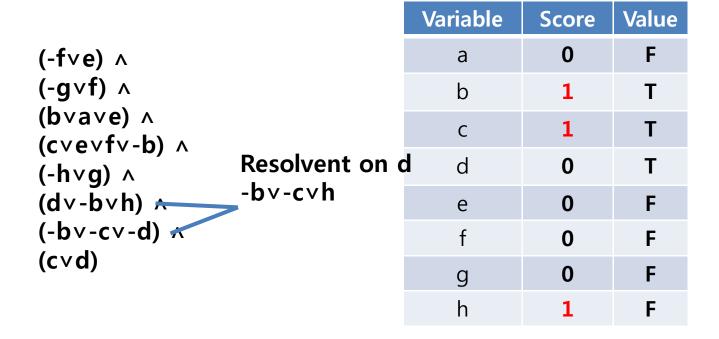
- b, c are True after BCP
- Conflict occurs on variable d
 - Start conflict analysis

```
(-fve) ^
  (-gvf) ^
  (bvave) ^
  (cvevfv-b) ^
  (-hvg) ^
  (dv-bvh) ^
  (-bv-cv-d) ^
  (cvd)
```

Variable	Score	Value
a	0	F
b	0	Т
С	0	Т
d	0	Т
е	0	F
f	0	F
g	0	F
h	0	F

VSIDS Decision Heuristic (5/8)

- The score of variable in resolvents is increased by 1
 - Even if a variable appears in resolvents two or mores increase the score just once



VSIDS Decision Heuristic (6/8)

- The end of conflict analysis
- The scores are decaying 5% for next scoring

(-fve) ^ (-gvf) ^ (bvave) ^ (cvevfv-b) ^ (-hvg) ^ (dv-bvh) ^ (-bv-cv-d) ^	Resolvents -b∨-c∨h -b∨e∨f∨h ← learnt clause
---	---

(cvd)

Variable	Score	Value
a	0	F
b	0.95	Т
С	0.95	T
d	0	Т
е	0.95	F
f	0.95	F
g	0	F
h	0.95	F

VSIDS Decision Heuristic (7/8)

- b is now False and a is True after BCP
- Next decision variable is c with 0.95 score

Variable	Score	Value
a	0	T
b	0.95	F
С	0.95	
d	0	
е	0.95	F
f	0.95	F
g	0	F
h	0.95	F

VSIDS Decision Heuristic (8/8)

Finally we find a model!

Variable	Score	Value
a	0	T
b	0.95	F
С	0.95	F
d	0	T
е	0.95	F
f	0.95	F
g	0	F
h	0.95	F