SAT Solvers

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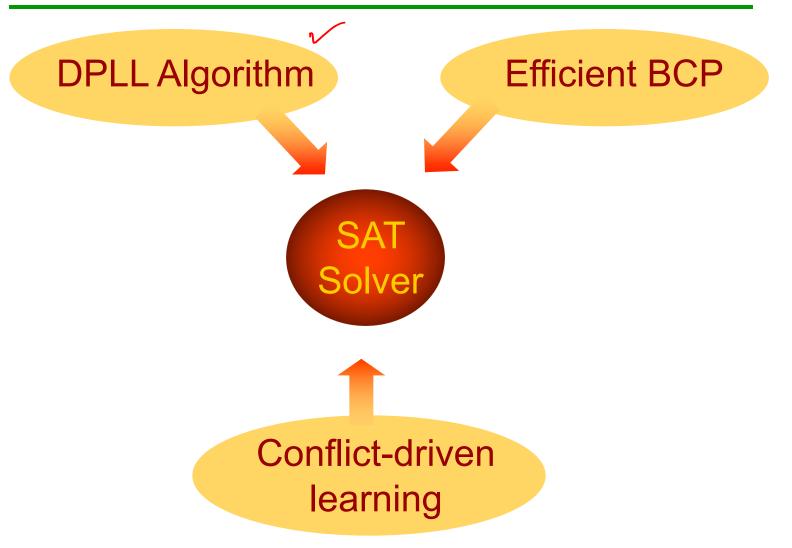
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EE-224: Digital Systems



Anatomy of a Modern SAT Solver







Stallmarck's Method (SM) in CNF

Recursive application of the branch-merge rule to each variable with the goal of identifying common conclusions

$$\varphi = (a + b)(\neg a + c)(\neg b + d)(\neg c + d)$$

Try
$$a = 0$$
: $(a = 0) \Rightarrow (b = 1) \Rightarrow (d = 1)$

$$C(a = 0) = \{a = 0, b = 1, d = 1\}$$

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$$a = 0$$
: $(a = 0) \Rightarrow (b = 1) \Rightarrow (d = 1)$ $C(a = 0) = \{a = 0, b = 1, d = 1\}$

Try $a = 1$: $C(a = 1) \Rightarrow (c = 1) \Rightarrow (d = 1)$ $C(a = 1) = \{a = 1, c = 1, d = 1\}$

$$C(a = 1) = \{a = 1, c = 1, d = 1\}$$

$$C(a = 0) \cap C(a = 1) = \{d = 1\}$$

Any assignment to variable a implies d = 1. Hence, d = 1 is a necessary assignment!

Recursion can be of arbitrary depth



Recursive Learning (RL) in CNF

Recursive evaluation of clause satisfiability requirements for identifying common assignments

$$\varphi = (a+b)(\neg a+d)(\neg b+d)$$

Try
$$a = 1$$
: $(a = 1) \Rightarrow (d = 1)$

$$C(a = 1) = \{a = 1, d = 1\}$$

 $C(b = 1) = \{b = 1, d = 1\}$

Try
$$b = 1$$
: $(b = 1) \Rightarrow (d = 1)$

$$C(b = 1) = \{b = 1, d = 1\}$$

$$C(a = 1) \cap C(b = 1) = \{d = 1\}$$

Every way of satisfying (a + b) implies d = 1. Hence, d = 1 is a necessary assignment!

Recursion can be of arbitrary depth



Recursive Learning within DP

$$\varphi = (a + b + c)(\neg a + d + e)(\neg b + d + c)$$

Implications:

$$(a = 1) \land (e = 0) \Rightarrow (d = 1)$$

$$(b=1) \land (c=0) \Rightarrow (d=1)$$

consensus

$$(b+c+e+d)$$

consensus

$$(c=0) \land ((e=0) \land (c=0)) \Rightarrow (d=1)$$

Clausal form:

$$(c + e + d)$$
 Unit clause!

Clause provides <u>explanation</u> for necessary assignment d = 1



(c + e + d)

SM vs. RL

- Both complete procedures for SAT
- Stallmarck's method:
 - hypothetic reasoning based on <u>variables</u>
- Recursive learning:
 - hypothetic reasoning based on <u>clauses</u>
- Both can be integrated into backtrack search algorithms





Local Search

- Repeat *M* times:
 - Randomly pick complete assignment
 - Repeat K times (and while exist unsatisfied clauses):
 - Flip variable that will satisfy largest number of unsat clauses

$$\varphi = (a+b)(\neg a+c)(\neg b+d)(\neg c+d)$$

$$\varphi = (a+b)(\neg a+c)(\neg b+d)(\neg c+d)$$

$$\varphi = (a+b)(\neg a+c)(\neg b+d)(\neg c+d)$$

Pick random assignment

Flip assignment ph. d

Instance is satisfied!



Comparison

- Local search is incomplete
 - If instances are known to be SAT, local search can be competitive
- Stallmarck's Method (SM) and Recursive Learning (RL) are in general slow, though robust
 - SM and RL can derive too much unnecessary information
- For most EDA applications backtrack search (DP) is currently the most promising approach!
 - Augmented with techniques for inferring new clauses/implicates (i.e. learning)!





SAT Solvers Today

Capacity:

- Formulas upto a million variables and 3-4 million clauses can be solved in few hours
- Only for structured instances e.g. derived from real-world circuits & systems

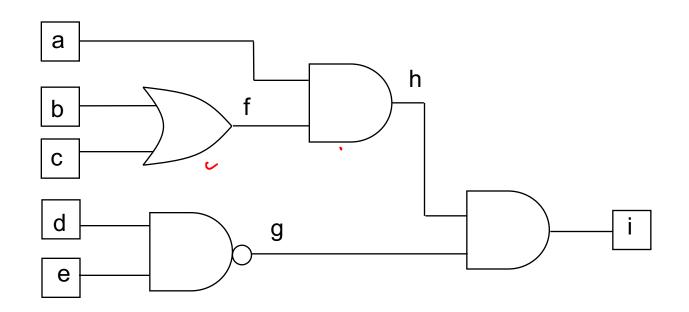
Tool offerings:

- Public domain
 - GRASP : Univ. of Michigan
 - SATO: Univ. of Iowa
 - zChaff: Princeton University
 - BerkMin: Cadence Berkeley Labs.
- Commercial
 - PROVER: Prover Technologies





Solving Circuit Problems as SAT



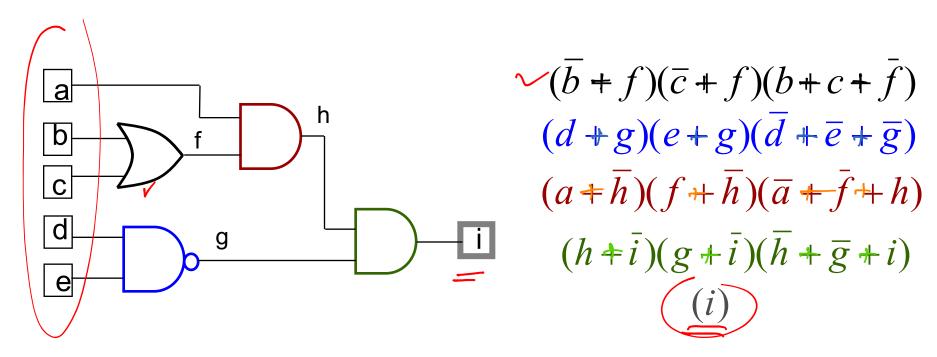
Input Vector Assignment?





Solving circuit problems as SAT

- Set of clauses representing function of each gate
- Unit literal clause asserting output to '1'



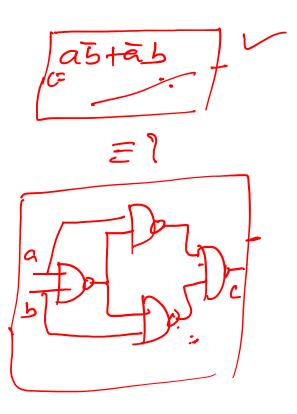


Application of SAT

EQUIVALENCE CIT EXING. VERIFICATION (SAT Exp & D1). (SAT Exp D2)
. (SAT exp XOR) = 1 If input assignment exist If does not exist



Application of SAT





Thank You



