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# SMT Solvers

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## SMT Solvers

- Anything a SAT solver can do, an SMT solver can do better
- SAT solvers are used for
  - Bounded model checking, and
  - AI planning,among other things

## SAT Solving

- Find satisfying assignment to a propositional logic formula
- Formula often represented as a set of clauses
  - CNF: conjunction of disjunctions
  - Find an assignment of truth values to variable that makes at least one literal in each clause TRUE
- Example: given following 4 clauses
  - $A, B$
  - $C, D$
  - $E$
  - $\bar{A}, \bar{D}, \bar{E}$

A solution is  $A, C, \bar{D}, E$  ( $A, D, E$  is not)

- Do this when there are 1,000,000 variables and clauses

## SAT Solvers

- SAT solving is quintessential NP-complete problem
- But **now amazingly fast in practice** (most of the time)
  - Breakthroughs (starting with Chaff) since 2001
  - Sustained improvements, honed by competition
- **Has become commodity technology**
  - Can think of it as massively efficient search
- **Used in bounded model checking and in AI planning**
  - Routine to handle  $10^{300}$  states

## Bounded Model Checking (BMC)

- Is there a counterexample to property  $P$  in  $k$  steps or less?
- System specified by initiality predicate  $I$  and transition relation  $T$  on states  $S$
- Does there exist assignments to states  $s_0, \dots, s_k$  such that
$$I(s_0) \wedge T(s_0, s_1) \wedge T(s_1, s_2) \wedge \dots \wedge T(s_{k-1}, s_k) \wedge \neg(P(s_1) \wedge \dots \wedge P(s_k))$$
- Given a Boolean encoding of  $I$ ,  $T$ , and  $P$  (i.e., circuit), this is a propositional satisfiability (SAT) problem
  - Symbolic model checking uses same representation as BMC, different backend (BDDs)
- Extends from refutation to verification via  $k$ -induction
- Also generates plans (test cases)
  - counterexample to negation of property
  - Though specialized planning languages provide better frontends to the SAT solver than a model checker

## Satisfiability Modulo Theories (SMT)

- SAT can encode operations and relations on **bounded** integers (bitvector representation), and other **finite** data types and structures
- But not **unbounded** or **infinite** types (e.g., reals), or structures (e.g., queues, lists)
- And even bounded arithmetic can be **slow**
- **There are fast decision procedures for these theories**
- But they work only on **conjunctions** of clauses
- **General propositional structure requires case analysis**
  - Should use efficient search strategies of SAT solvers

**That's what an SMT solver does**

## SMT Solving

- Individual decision procedures decide **conjunctions** of formulas in their decided theories
- **Combinations** of decision procedures (using, e.g., Nelson-Oppen or Shostak methods) decide conjunctions over the **combined theories** (e.g., arithmetic plus arrays)
- **SMT allows general propositional structure**
  - e.g.,  $(x \leq y \vee y = 5) \wedge (x < 0 \vee y \leq x) \wedge x \neq y$   
... possibly continued for 1000s of terms
- Should exploit search strategies of modern SAT solvers
- So replace the **terms** by **propositional variables**
  - $(A \vee B) \wedge (C \vee D) \wedge E$
- Get a **solution from a SAT solver** (if none, we are done)
  - e.g.,  $A, D, E$

## SMT Solving by “Lemmas On Demand”

- Restore the interpretation of variables and send the conjunction to the core decision procedure
  - e.g.,  $x \leq y \wedge y \leq x \wedge x \neq y$
- If satisfiable, we are done
- If not, ask SAT solver for a new assignment—but isn't it expensive to keep doing this?
- Yes, so first, do a little bit of work to find fragments that explain the unsatisfiability, and send these back to the SAT solver as additional constraints (i.e., lemmas)
  - $A \wedge D \supset \neg E$
- Iterate to termination (e.g.,  $B, D, E: y = 5, y < x: y = 5, x = 6$ )
- This is called “lemmas on demand” or “DPLL(T)”
- it works really well: yields effective SMT solvers

## SMT Solvers

- SMT solvers are being honed by competition
- Various divisions (depending on the theories considered)
  - Equality and uninterpreted functions
  - Difference logic ( $x - y < c$ )
  - Full linear arithmetic
  - ... for integers as well as reals
  - Arrays
- Next competition at FLoC (Seattle, Summer 2006)
- SMT solvers enable infinite bounded model checking
  - And powerful backends to interactive theorem provers
  - And metric and temporal planning for AI  
(demonstrated by Martha Pollack et al using ARIIO)

## Example: Real Time

- Traditionally hard for automated analysis because continuous time excludes finite state methods
- Timed automata methods handle continuous time
  - But defeated by the case explosion when (discrete) faults are considered
- SMT solvers can handle both dimensions
  - Timeout automata, k-induction, disjunctive invariants
- E.g., Biphase Mark Protocol for asynchronous communic'n
  - Clocks at either end have different skew, rates, jitter
  - So have to encode a clock in the data stream
  - Used in CDs, Ethernet
  - Verify parameter values for reliable transmission

## Real Time: Biphase Mark (ctd)

- First verified by human-guided proof in [ACL2](#) by J Moore
- **Three different verifications** used [PVS](#)
  - One by Groote and Vaandrager used [PVS + UPPAAL](#)
  - Required **37** invariants, **4,000** proof steps, **hours** of prover time to check
- Brown and Pike recently did it with [sal-inf-bmc](#)
  - **Three** lemmas proved **automatically** with **1-induction**,
  - Statement of theorem discovered **systematically** using **disjunctive invariants** (**7** disjuncts)
  - Theorem proved **automatically** using **5-induction**
  - Verification takes **seconds** to check
- **Adapted** verification to 8-N-1 protocol (used in UARTs)
  - **Revealed a bug** in published application note

## Summary

- SAT can be extended to **MaxSAT** to deal with inconsistencies
  - e.g., to find best diagnosis, integrate learners
  - We have done this
- SMT can be extended similarly to **MaxSMT**
  - We are doing this
- SMT also can be extended to **maximize any arithmetic expression, subject to constraints**
  - We are doing this, too
- **Anything a SAT solver can do**
  - **And anything a constraint solver can do**

**An SMT solver can do better** (we think)
- See **ICS** and its descendents at [fm.csl.sri.com](http://fm.csl.sri.com)