# Lecture 15 SMT Solvers

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slides acknowledgements: Leonardo de Moura

#### Announcements: Wrapping up Projects

- Presentations
  - Apr 23 in class
    - Everyone should come. Let me know ASAP if you cannot come for some reason.
    - Good presentations
    - Pizza
  - Slides are due on Apr 18!!!
    - Dry run in my office on Apr 18 during class time
- Final report
  - Due on Apr 25
- Peer review
  - Due on Apr 28

#### This Time

- SMT solvers
  - What are they?
  - ▶ How they work?

### Many Theories

- Theory of equality
- Peano arithmetic
- Presburger arithmetic
- Linear integer arithmetic
- Reals
- Rationals
- Arrays
- Recursive data structures
- ...

#### **Combination of Theories**

- In practice, we often need a combination of theories
- Example:

```
x+2=y \rightarrow f(select(store(a,x,3),y-2)=f(y-x+1)
```

Problem: given satisfiability procedures for conjunction of literals of Theory<sub>1</sub> and Theory<sub>2</sub>, how to decide satisfiability of their combination?

## Satisfiability Modulo Theories (SMT) Solver

- Satisfiability checker with built-in support for useful theories
  - Arithmetic
  - Equality with uninterpreted functions
  - Arrays
  - ...
- Combines a SAT solver with theory solvers
- Next generation of reasoning engines
  - Automatic
  - Fast

### SMT Solvers, Library, Competition

- Solvers
  - AProve, Barcelogic, Boolector, CVC4, MathSAT5, OpenSMT, SMTInterpol, SOLONAR, STP2, veriT, Yices, Z3
- SMT-LIB
  - Standardizes various theories and input format
  - Library of benchmarks
  - http://www.smtlib.org
- **▶** SMT-COMP
  - Annual competition
  - http://www.smtcomp.org

#### **Applications**

- Test case generation
- Verifying compilers
- Software verification
- Hardware verification
- Equivalence checking
- Type checking
- Model based testing
- Scheduling and planning
- . . .

#### **Nelson-Oppen Combination Procedure**

- Initial State
  - ▶ F is a conjunction of literals over  $\Sigma_1 \cup \Sigma_2$
- Purification
  - ▶ Preserving satisfiability transform F into  $F_1 \wedge F_2$ , such that  $F_i \in \Sigma_i$
- Interaction
  - ▶ Deduce an equality x = y if  $F_1 \rightarrow x = y$ , where x and y are common (shared) variables
  - ▶ Update  $F_2 := F_2 \land x = y$
  - And vice-versa
  - Repeat until no further changes

#### Nelson-Oppen Combination Procedure

- Component procedures
  - Use individual decision procedures to decide whether F<sub>i</sub> is satisfiable
- Return
  - If both return yes, return yes
  - No, otherwise

Remark:

 $F_i \rightarrow x = y$  iff  $F_i \land x \neq y$  is not satisfiable

# Purification Example

$$f(x-1)-1=x \wedge f(y)+1=y$$

## Nelson-Oppen Procedure Example I

$$x + y = z \wedge f(z) = z \wedge f(x + y) \neq z$$

### Nelson-Oppen Procedure Example II

$$x+2=y \land f(select(store(a,x,3), y-2)) \neq f(y-x+1)$$

# Building an Efficient Solver

## Eager Approach

- Translate formula into equisatisfiable propositional formula and use off-the-shelf SAT solver
- Why "eager"?
  - Search uses all theory information from the beginning
- Can use best available SAT solver
- Sophisticated encodings are need for each theory
- Sometimes translation and/or solving too slow

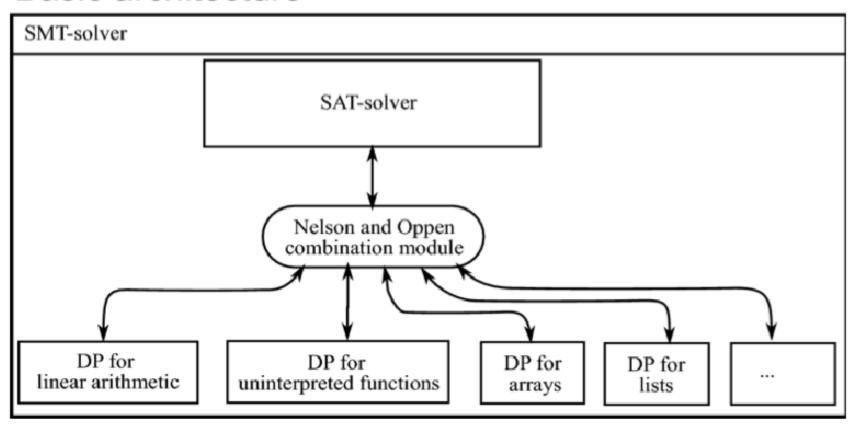
### Lazy Approach: SAT + Theories I

- Independently developed by several groups
  - CVC (Stanford)
  - ▶ ICS (SRI)
  - MathSAT (Univ. Trento, Italy)
  - Verifun (HP)
- Motivated by the breakthroughs in SAT solving
  - DPLL algorithm
  - Various optimizations and heuristics

## Lazy Approach: SAT + Theories II

- SAT solver
  - Manages the boolean structure and assigns truth values to the atoms in a formula
- Theory solvers
  - Efficiently validate (partial) assignments produced by the SAT solver
- When a theory solver detects unsatisfiability, a new clause (lemma) is created

#### Basic architecture



### Naïve Approach

- Example
  - Suppose SAT solver assigns  $\{x = y \rightarrow T, y = z \rightarrow T, f(x) = f(z) \rightarrow F\}$
  - Theory solver detects conflict
  - Lemma is created  $\neg(x = y) \lor \neg(y = z) \lor f(x) = f(z)$
- Potential problems
  - Lemmas are imprecise (not minimal)
  - Theory solver is "passive"
    - It just detects conflicts
    - There is no propagation step
  - Backtracking is expensive
    - Restart from scratch when a conflict is detected

# **Theory Solvers**

- Basic requirements
  - Deduce equalities between variables
  - Compute lemmas (conflict sets)
    - As precise as possible
- Extra desired features
  - Theory propagation
  - Incrementality
  - Backtracking

### **Equality Generation**

- Combination of theories strongly relies on the propagation of deduced equalities
- Every theory solver has to support it

#### Precise Lemmas I

- Example

  - Lemma is  $\neg a_1 \lor a_2 \lor a_3$
- An inconsistent set A is redundant if A' ⊂ A is also inconsistent
- Redundant inconsistent sets imply
  - Imprecise lemmas
  - Ineffective pruning of the search space

#### Precise Lemmas II

- Noise of a redundant set is A \ A<sub>min</sub>
- Imprecise lemma is useless in any partial assignment where an atom in the noise has a different assignment
- Example
  - Suppose a<sub>1</sub> is in the noise
  - ▶ Then  $\neg a_1 \lor a_2 \lor a_3$  is useless when  $a_1 = F$

## **Theory Propagation**

- SAT solver is assigning truth values to the atoms in a formula
- Partial assignment produced by the SAT solver may imply truth values of unassigned atoms
- Example

```
x = y \land y = z \land (f(x) \neq f(z) \lor f(x) = f(w))
Partial assignment \{x = y \rightarrow T, y = z \rightarrow T\}
implies f(x) = f(z)
```

Reduces the number of conflicts and the search space

#### Incrementality

- Theory solvers constantly receive new constraints and restart the process
  - Augmented partial assignments from SAT solver
  - Equalities coming from other theory solvers
- Do not restart from scratch
  - Reuse what you learned so far

### Efficient Backtracking

- One of the most important improvements in SAT was efficient backtracking
- Extreme (inefficient) approach in theory solvers
  - Restart from scratch on every conflict
- Efficient approach
  - Restore to a logically equivalent state
- Backtracking should be included in the design of theory solvers

# Ideal Theory Solver

- Efficient in real benchmarks
- Produces precise lemmas
- Supports theory propagation
- Incremental
- Efficient backtracking

# Dealing with Quantifiers

#### **Quantifier Instantiation**

- SMT solvers use heuristic quantifier instantiation using E-matching (matching modulo equalities)
- Divide input formula into ground and quantified portion
- Check ground portion for satisfiability
  - If SAT then extend with ground terms instantiated from the quantified part
    - Often leverage user-provided triggers
  - If UNSAT then report UNSAT
- Repeat

### Example

```
\forall x: f(g(x)) = x \{ f(g(x)) \} (trigger)

a = g(b),

b = c,

f(a) \neq c
```

#### Limitations

- Users often have to manually provide patterns
  - Automatic inference of patterns is fragile
- Bad user provided patterns
  - False positives (wrong SAT answers)
  - Nonterminating executions

## Trigger too Restrictive

```
\forall x: f(g(x)) = x \{ f(g(x)) \}

g(a) = c,

g(b) = c,

a \neq b
```

Results in false positives

# Trigger too Restrictive

More "liberal" pattern:

```
\forall x: f(g(x)) = x \{ g(x) \}

g(a) = c,

g(b) = c,

a \neq b
```

Instantiate:

$$f(g(a)) = a,$$
  
$$f(g(b)) = b$$

Implies that a=b

### **Matching Loop**

```
\forall x: f(x) = g(f(x)) \{ f(x) \}

\forall x: g(x) = f(g(x)) \{ g(x) \}

f(a) = c
```

Instantiate:

```
f(a) = g(f(a))g(f(a)) = f(g(f(a)))
```

Results in executions that do not terminate