## Lecture 17: Satisfiability Modulo Theories

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\* Course materials developed with Jonathan Aldrich and Claire Le Goues



**Carnegie Mellon University** School of Computer Science

(c) J. Aldrich, C. Le Goues, R. Padhye



#### Sometimes we need to reason about formulas

- Verification: verification condition generation turns a Hoare triple into a formula
  - Is that formula valid (i.e. always true the precondition always implies the postcondition)?
- Symbolic execution: builds path conditions as execution proceeds
   Is that path condition satisfiable (i.e. potentially true given the right inputs)?
- More applications: test generation, program synthesis, program repair, ...
- Can tools automatically reason about formula validity or satisfiability?

#### First step: reduce validity to satisfiability

- Formula validity: ∀x . F(x) is true
   (x stands for the *free variables* of F)
- Equivalent to  $\neg \exists x . F(x)$  is false
- Equivalent to ¬∃x . ¬F(x) is true
   This is asking whether ¬F(x) is satisfiable

### Satisfiability modulo theories

- Satisfiability is for Boolean formulas
   Variables, Boolean operators such as
- Verification conditions, path conditions, etc. have other elements
  - Integer, real constants and variables
  - Operations over numbers like < > + -
- We can enhance satisfiability checkers to incorporate theories
  - Presburger arithmetic can prove that 2 \* x = x + x
  - The theory of arrays can prove that assigning *x*[*y*] to 3 and then looking up *x*[*y*] yields 3

### Satisfiability (SAT) solving

- Let's start by considering Boolean formulas: variables connected with  $\land \lor \neg$
- First step: convert to conjuctive normal form (CNF)
  - A conjunction of disjunctions of (possibly negated) variables

 $(a \lor \neg b) \land (\neg a \lor c) \land (b \lor c)$ 

• If formula is not in CNF, we transform it: use De Morgan's laws, the double negative law, and the distributive laws:

$$\begin{array}{ccc} \neg (P \lor Q) & \iff & \neg P \land \neg Q \\ \neg (P \land Q) & \iff & \neg P \lor \neg Q \\ & \neg \neg P & \iff & P \\ (P \land (Q \lor R)) & \iff & ((P \land Q) \lor (P \land R)) \\ (P \lor (Q \land R)) & \iff & ((P \lor Q) \land (P \lor R)) \end{array}$$

### SAT solving goal

- Prove that a formula is *satisfiable* by giving a satisfying assignment
   A map from formula variables to Boolean values
- Example:  $X \lor Y$  is satisfiable
  - $\circ~$  A satisfying assignment is  $X\mapsto \mathtt{true}, Y\mapsto \mathtt{false}$
- Example:  $X \land \neg X$  is unsatisfiable
  - No satisfying assignment exists

#### SAT is NP-complete

- Cook-Levin theorem [1970s] proved NP-completeness
  - In NP, because can verify a satisfying assignment by evaluating the formula
  - NP-hard by reduction to polynomial-time acceptance by a nondeterministic Turing machine
- Simple solution approach: try all satisfying assignments
   Takes O(2<sup>n</sup>) time for an n-variable formula

- Developed by Davis, Putnam, Logemann, and Loveland [1961]
  - Still exponential in theory, but on many problems is much faster than trying all assignments
- Key innovation #1: *unit propagation*   $(b \lor c) \land (\bigstar) \land (\neg a \lor c \lor d) \land (\neg c \lor d) \land (\neg c \lor \neg d \lor \neg a) \land (b \lor d)$ 
  - In this example, a appears alone. It must be true.

- Developed by Davis, Putnam, Logemann, and Loveland
  - Still exponential in theory, but on many problems is much faster than trying all assignments
- Key innovation #1: *unit propagation*
- Key innovation #2: *pure literal elimination*

 $(\square \land (c \lor d) \land (\neg c \lor d) \land (\neg c \lor \neg d) \land (\square \land d))$ 

- This example is simplified from the previous slide, based on unit propagation
- Note that b appears only positively. **Setting b to true** can only help us, not hurt us!

- Developed by Davis, Putnam, Logemann, and Loveland
  - Still exponential in theory, but on many problems is much faster than trying all assignments
- Key innovation #1: *unit propagation*
- Key innovation #2: *pure literal elimination*
- When we are stuck, we guess (and backtrack later if necessary)  $(c \lor d) \land (\neg c \lor d) \land (\neg c \lor \neg d)$ 
  - Let's guess that c is true! Then we get  $(d) \land (\neg d)$
  - We apply unit propagation to set d=true. Unfortunately the result is so we failed to find a satisfying assignment
     (true) ^ (false)

- Developed by Davis, Putnam, Logemann, and Loveland
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- When we are stuck, we guess (and backtrack later if necessary)  $(c \lor d) \land (\neg c \lor d) \land (\neg c \lor \neg d)$ 
  - $\circ$  Now let's guess that c is false! Then we get (d)
  - We apply unit propagation to set d=true and the formula is satisfied

### The Full DPLL Algorithm

function DPLL( $\phi$ )

 $\text{if } \phi = \texttt{true } \texttt{then} \\$ 

return true

end if

if  $\phi$  contains a false clause then return false

end if

for all unit clauses l in  $\phi$  do

 $\phi \leftarrow \text{UNIT-PROPAGATE}(l, \phi)$ 

#### end for

for all literals l occurring pure in  $\phi$  do

 $\phi \leftarrow \text{PURE-LITERAL-ASSIGN}(l, \phi)$ end for

 $l \leftarrow \text{CHOOSE-LITERAL}(\phi)$ return DPLL $(\phi \land l) \lor \text{DPLL}(\phi \land \neg l)$ end function

Heuristic: Apply unit propagation first because it creates more units and pure literals. Pure literal assignment only removes entire clauses.

> Try both assignments of the chosen literal. If we assume ∨ is short-circuiting, then this implements backtracking.

### Practice: Applying DPLL

Show how DPLL (unit propagation, pure literal elimination, choosing a literal, backtracking) applies to the following formula:
 (a ∨ b) ∧ (a ∨ c) ∧ (¬a ∨ c) ∧ (a ∨ ¬c) ∧ (¬a ∨ ¬c) ∧ (¬d)

#### From SAT to SMT

- We'd like to check the satisfiability of formulas like  $\begin{array}{l} f(f(x) f(y)) = a & \wedge \\ f(0) = a + 2 & \wedge \end{array}$
- Includes arithmetic and the theory of unknown functions
  - E.g. we assume f is some mathematical function
- We may have solvers for each theory, but how can we combine them?
   Note that separate satisfying assignments for two theories might not be compatible!
- SMT's solution: solve theories separately, use SAT to combine them

The running example is due to Oliveras and Rodriguez-Carbonell

x = y

# Nelson-Oppen replaces expressions with variables

 $f(f(x) - f(y)) = a \land f(0) = a + 2 \land x = y$ 

#### Now we have formulas in two theories

Theory of uninterpreted functions

$$f(e1) = a$$
  

$$e2 = f(x)$$
  

$$e3 = f(y)$$
  

$$f(e4) = e5$$
  

$$x = y$$

• Theory of arithmetic

$$e1 = e2 - e3$$
$$e4 = 0$$
$$e5 = a + 2$$
$$x = y$$

Congruence closure:

for all f, x, and y, if x = y then f(x) = f(y)

Theories communicate using equalities

### Combining Theories using DPLL

- Consider the following source formula:  $x \ge 0 \land y = x + 1 \land (y > 2 \lor y < 1)$
- We can convert each subformula to a variable:  $n_1 \wedge n_2 \wedge (n_3 \vee n_4)$
- Now we solve with DPLL and get a satisfying assignment:  $p1, p2, \neg p3, p4$
- We ask the theories if this assignment is feasible
  - The theory of arithmetic says no. p1, p2, and p4 can't all be true, because p1 and p2 together imply  $y \ge 1$
- We add a clause expressing this and run DPLL again on

 $p1 \wedge p2 \wedge (p3 \vee p4) \wedge (\neg p1 \vee \neg p2 \vee \neg p4)$ 

- One satisfying assignment is  $p1, p2, p3, \neg p4$ 
  - We check this against the theories and it succeeds

#### Details on equality

- Sometimes a theory doesn't tell us an equality, but rather that one of two equalities are true
  - That's fine—we just encode this as a formula and give it to DPLL. For example:  $(e1 = e2 \lor e1 \neq e2) \land (e2 = e3 \lor e2 \neq e3)$
  - DPLL will choose which equalities are true, and we try those with other theories.

#### SMT uses a variant of DPLL called DPLL(T)

- T is for Theory
- Differences vs. plain DPLL
  - DPLL(T) doesn't use pure literal elimination
    - Variables may not be independent when they represent a formula so setting x to true can hurt you, even when x is a pure literal!
    - For example:  $(\mathbf{x} > \mathbf{10} \lor x < 3) \land (\mathbf{x} > \mathbf{10} \lor x < 9) \land (x < 7)$ 
      - Can't just set x > 10 to true, because x < 7 will be false
  - DPLL(T) supports adding clauses to the formula
    - To represent knowledge gained from theories, as mentioned above

#### How to solve arithmetic

- Approach #1: Substitution
  - If we have y = x+1, we can eliminate y by substituting it with x+1 everywhere
  - High school math!
- Approach #2: Fourier-Motzkin Elimination
  - Applies when we have inequalities rather than equalities
  - Transform all inequalities mentioning x into  $A \le x$  or  $x \le B$
  - Then eliminate X, replacing the inequalities with  $A \le B$ 
    - Detail: if there are multiple inequalities, we conjoin the cross product of them

### Modern tooling: SMT-lib

SMT-LIB THE SATISFIABILITY MODULO THEORIES LIBRARY

#### Theories

SMT-LIB logics refer to one or more theories below. Click on a theory's name to see its declaration in Version 2.x of the format.

#### ArraysEx

Functional arrays with extensionality

#### **FixedSizeBitVectors**

Bit vectors with arbitrary size

#### Core

Core theory, defining the basic Boolean operators

#### FloatingPoint

Floating point numbers

#### Ints

Integer numbers

#### Reals

Real numbers

#### Reals\_Ints

Real and integer numbers

#### Strings

Unicode character strings and regular expressions

SMT-
COMP
The International
Satisfiability Modulo
Theories (SMT)
Competition.
GitHub
Home
Introduction
Benchmark Submission
Publications
SMT-LIB
Previous Editions
SMT-COMP 2021
Rules
Benchmarks
Tools
Specs
Parallel & Cloud Tracks
Participants
Results
Slides

#### SAT Performance

Solver 🗧	Correct Score	<ul> <li>Time Score</li> </ul>	Division
cvc5	0.10587024	0.00161216	Equality+LinearArith
UltimateEliminator+MathSAT	0.08416589	0.00358447	Equality+NonLinearArith
Vampire	0.02936727	0.00424783	Equality
cvc5	0.00616228	0.00599641	QF_NonLinearIntArith
Yices2-QS	0.00553133	0.00393864	Arith
cvc5	0.00262204	0.00188031	QF_NonLinearRealArith
cvc5	0.00157784	0.00059486	QF_Equality
cvc5	0.00145186	0.0022273	QF_LinearIntArith
cvc5	0.00125507	0.00027314	Bitvec
Bitwuzla	0.00093358	0.00069896	QF_Bitvec
cvc5	0.00093134	-0.00090887	QF_Equality+NonLinearArith
cvc5	0.00055932	-0.00228172	QF_FPArith
cvc5	0.00013178	0.00015434	QF_Equality+LinearArith
Yices2	0.00010454	0.00020464	QF_Equality+Bitvec
Yices2	9.601e-05	0.00035558	QF_LinearRealArith

#### UNSAT Performance

Solver	← Correct Score	▼ Time Score	Division
cvc5	0.05633672	0.03589274	Equality+NonLinearArith
Yices2	0.02120632	0.0072311	QF_NonLinearIntArith
cvc5	0.01061534	0.04760987	Equality+LinearArith

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### Modern tooling: Z3 w/ SMT-lib

Z3 Online Demonstrator	SMT-LIB 2 Standard	L Z3 sources			
Reset       Execute         SMT-LIB 2 script       ; Variable declarations         ; (declare-fun a () Int)       ; (declare-fun b () Int)         ; (declare-fun b () Int)       ; (declare-fun c () Int)         ; constraints       ; (assert (> a 0))         ; (assert (> b 0))       ; (b 0)			<b>Output</b> z3 output sat (model (define-fun c () Int 15) (define-fun b () Int 9) (define-fun a () Int 12)		
<pre>(assert (&gt; b 0)) (assert (&gt; c 0)) (assert (= (+ (* a a) (* ; Solve (check-sat) (get-model)</pre>	b b)) (* c c)))		Summary	z3 -in -T:30 0.083 s z3-4.4.1	