CSC2/455 Software Analysis and Improvement An Introduction to SAT/SMT Solvers

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URCS

SAT Solving

SMT Solvers

SAT Solving

SMT Solvers

So far

- So far:
 - Iterative Data Flow Analysis
 - Type Analysis
 - Region Analysis
 - Abstract Interpretation
- Today
 - Satisfiability (SAT) Solvers
 - Satisfiability Modulo Theories (SMT) Solvers
- Next:
 - Model Checking
 - Symbolic Execution
 - Hoare Logic

SAT Solving

SMT Solvers

Given a formula in propositional logic (variables, true, false, \land , \lor , \neg , and parentheses), is there an assignment to variables that makes the formula true?

- $(A \lor B \lor C) \land (\neg A \lor B)$ (conjunctive normal form, CNF)
- $(A \land B \land C) \lor (\neg A \land B)$ (disjunctive normal form, DNF)
- *A* ∧ ¬*A* (CNF)

- $(A \lor B \lor C) \land (\neg A \lor B)$
 - B = true is required in any satisfying assignment (A and C don't matter)
- $(A \land B \land C) \lor (\neg A \land B)$
 - A, B, C all true is one satisfying assignment
 - A =false and B =true is another satisfying assignment
- $A \land \neg A$ is obviously unsatisfiable

- If the maximum number of variables in a clause of a CNF formula is k, we call that problem k-SAT
- 2-SAT is solvable in polynomial time
- 3-SAT is NP-complete
 - In worst case, must explore every possible assignment of values to each variable

SAT Problem Sizes

- There are many good SAT solvers now available
 - Based on the Davis–Putnam-Logemann-Loveland (DPLL) algorithm
 - Often enhanced with Conflict-Driven Clause Learning (CDCL)
 - SAT is decidable, if untractable
- Intractability not a hindrance usually
 - Can scale to very large problems
 - Millions of clauses
 - See: The International SAT Competition
- Applied to many hardware and software verification problems
- SAT solvers return:
 - SAT: if a satisfying assignment is found (and the values that satisfy the proposition)
 - UNSAT: if no satisfying assignment exists

Prove
$$\neg (A \land B) = (\neg A \lor \neg B)$$

A	В	$A \wedge B$	$P=\neg(A \land B)$	$\neg A$	$\neg B$	$Q = \neg A \lor \neg B$	$P \iff Q$
F	F	F	Т	Т	Т	Т	Т
F	Т	F	Т	Т	F	Т	Т
Т	F	F	т	F	Т	т	Т
Т	Т	Т	F	F	F	F	Т

The statement $\neg(A \land B) = (\neg A \lor \neg B)$ is valid, it is true for all values of A and B.

Proof using a SAT solver

•
$$\neg (A \land B) \implies (\neg A \lor \neg B)$$

•
$$(\neg A \lor \neg B) \implies \neg (A \land B)$$

• Recall that $P \implies Q$ can be written as $\neg P \lor Q$

•
$$\neg \neg (A \land B) \lor (\neg A \lor \neg B)$$

•
$$\neg(\neg A \lor \neg B) \lor \neg(A \land B)$$

So we have:

•
$$R = (A \land B) \lor (\neg A \lor \neg B)$$

•
$$S = \neg(\neg A \lor \neg B) \lor \neg(A \land B)$$

- For the proof, we need both R and S to be valid
 - If R is valid, what can we say about the satisfiability of $\neg R$?

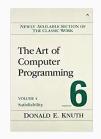
The Satisfiability of $\neg R$

A	В	$P = A \wedge B$	$\neg A$	$\neg B$	$Q = \neg A \lor \neg B$	$P \lor Q$	$\neg (P \lor Q)$
F	F	F	Т	Т	Т	Т	F
F	Т	F	Т	F	Т	Т	F
Т	F	F	F	Т	Т	Т	F
Т	Т	Т	F	F	F	Т	F

If R is valid, then $\neg R$ is unsatisfiable!

- To prove a statement using a SAT solver:
 - Convert the statement to propositional logic
 - Negate it, and check for unsatisfiability
- Interesting corollary:
 - If the formula is SAT (implying the statement is false), the values that satisfy the statement are counter-examples

• Volume 4 Facsicle 6 of The Art of Computer Programming



More details on SAT/SMT Solving (Papers)

- Weissenbacher, Subramanyan, and Malik, Boolean Satisfiability Solvers: Techniques and Extensions
- Barrett, Sebastinani, Sheshia and Tinelli, Satisfiability Modulo Theories
- Niklas Eén, Niklas Sörensson, An Extensible SAT Solver

Quantifiers, Theories

- Propositional logic can be extended with quantifiers
 - ∃, existential quantifier
 - \forall , universal quantifier
 - This is First-order Logic (FOL)
 - FOL is undecidable in general
- Both propositional logic and first-order logic are still boolean
- Can be extended with theories:
 - Arithmetic: adds numbers, operators +, -, ×, associativity, commutativity, etc.
 - Functions: adds f(x)
 - Bitvectors: model variables containing n bits (where n > 1)

- A SMT solver checks for satisfiability in a theory
 - Can think of statements as propositional logic $+\ theory$
- Allows construction of "richer" statements
 - Can formulate propositions over integers, reals, etc.
 - Can use operators like $+, -, \times$, etc.
- Example: $\forall_{x,y} x > y \implies x+1 > y+1$
 - True over integers (Z)
 - False over machine integers/bitvectors

- Some theories are decidable
 - Presburger arithmetic
- Most theories are undecidable
- However, some theories undecidable in general *are* decidable over quantifier-free fragments
- So, results of a SMT solver can be:
 - SAT
 - UNSAT
 - Don't know (or infinite loop)

SAT Solving

SMT Solvers

- Microsoft Z3
 - Used to be available online at rise4fun/Z3 $\,$
 - Available in most Linux distributions
- CVC5
- Yices
- Many more...

- Common input/output language for most SMT solvers
 - Some solvers support their own language as well
- Lisp-like
- Documented at smtlib.org
- Allows easy switching between solvers
 - We will use Z3 for the most part

Let's encode $\neg R$ from the previous example:

```
(declare-fun B () Bool)
(declare-fun A () Bool)
(assert (not (or (and A B) (not A) (not B))))
(check-sat)
```

And we run it:

\$ z3 p1.smt
unsat

```
#!/usr/bin/env python3
from z3 import *
s = Solver()
A, B = Bools('A B')
R = Or(And(A, B), Or(Not(A), Not(B)))
notR = Not(R)
s.add(notR)
print(s.check())
print(s.sexpr()) # prints the SMTLIB code
```

See: Bjørner et al., Programming Z3, for a nice introduction to programming Z3 using Python.

- The Constraint Satisfaction Problem seeks to find an assignment of values to variables subject to constraints
 - Each variable has a domain of values
 - Pick a variable, assign it a value, subject to constraints
 - If all variables can be assigned values, SAT else backtrack
- For SAT in propositional logic:
 - Two values, True and False
 - Constraint: formula must evaluate to true

- Dennis Yurichev's free book "SAT/SMT by Example" is a wonderful collection of examples
 - Minesweeper
 - Sudoku
 - Test case generation, etc.

SAT Solving

SMT Solvers

- Express program behaviour in some logic
- Express program property in that logic
- Check if the property holds (i.e. is valid)

Assertions

```
int min(int a, int b) {
    if(a < b)
        return a;
    else
        return a;
}
int x, y, r;
r = min(x, y);
assert(r == x || r == y);
assert(r <= x && r <= y);</pre>
```

- These assertions test that min always returns the minimum of x and y
- But assert executes at runtime
- We will seek to prove statically:
 - $\forall_{x,y}(\min(x,y) = x \lor \min(x,y) = y) \land (\min(x,y) \le x \land \min(x,y) \le y)$
 - over all program paths

```
from z3 import *
s = Solver()
a, b, ret = Ints('a b ret')
ret = If(a < b, a, a)
#ret = If(a < b, a, b) # correct
cond = And(Or(ret == a, ret ==b), And(ret <= a, ret <= b))
s.add(Not(cond))
print(s.sexpr())
if s.check() == sat:
    print("Incorrect. Counterexample: ", s.model())
else:
    print("Correct")</pre>
```

Output:

. . .

```
Incorrect. Counterexample: [b = 0, a = 1]
```

- If A is a source program and B is the compiled version, we would like to prove that A = B
 - This is called *translation validation*
 - What I've been doing for your submissions
- Undecidable, in general
- Not interesting only to compiler writers
 - If you take a piece of code and refactor it, did you break anything?

```
int min(int a, int b) {
    if(a < b)
        return a;
    else
        return a;
}
int x, y, r;
r = min(1, 3);
....</pre>
```

- The test case (1,3) is not sufficient to exercise all paths in the program
 - And it misses the bug!
- Can we find test cases to exercise all paths in the program?

- SMT solvers are a marvellous piece of technology
- Learn to use one!