

Lecture 3

On the power/limits of dynamic symbolic execution
(from “Higher-order test generation”, Patrice Godefroid PLDI 2011)

Powering up DSE with Satisfiability Modulo Theories

Thanks to Patrice Godefroid, Leonardo de Moura


Review

DSE

```

Procedure executeSymbolic( $P, I$ ) =
  initialize  $M_0$  and  $S_0$ 
  path constraint  $pc = \text{true}$ 
   $C = \text{getNextCommand}()$ 
  while ( $C \neq \text{stop}$ )
    match ( $C$ ):
      case ( $v := e$ ):
         $M = M + [\&v \mapsto \text{evalConcrete}(e)]$ 
         $S = S + [\&v \mapsto \text{evalSymbolic}(e)]$ 
      case (if  $e$  then  $C'$  else  $C''$ ):
         $b = \text{evalConcrete}(e)$ 
         $c = \text{evalSymbolic}(e)$ 
        if  $b$  then  $pc = pc \wedge c$ 
        else  $pc = pc \wedge \neg c$ 
     $C = \text{getNextCommand}()$  // end of while loop

```

```
evalSymbolic(e) =  
  match (e):  
    case v: // Program variable v  
      return S(&v)  
    case +(e1, e2): // Addition  
      f1 = evalSymbolic(e1)  
      f2 = evalSymbolic(e2)  
      if f1 and f2 are constants  
        return evalConcrete(e)  
      else  
        return createExpression('+', f1, f2)  
    etc.  
    default: // default for unhandled expression  
        
      return evalConcrete(e)
```

Soundness / Completeness

Input i

Path w

Path constraint pc_w is sound

if $\forall i \ i \models pc_w \Rightarrow \text{Path}(i) = w$

Path
Constraint

pc_w

Unsound Path Constraint \rightarrow Divergence

complex(42)
= 567

```
foo(int x, int y) {
```

```
    if(x == complex(y)) {
```

```
        if(y == 10) {
```

```
            ...
```

```

evalSymbolic(e) =
  match (e):
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      if f1 and f2 are constants
        return evalConcrete(e)
      else
        return createExpression ('+', f1 , f2)
    etc .
  default: // default for unhandled expression
     $pc = pc \wedge \bigwedge_{x_i \in e} (x_i = I_i)$ 
    return evalConcrete(e)

```

Sound Path Constraint

```
foo(int x, int y) {
```

```
    if (x == complex(y)) {
```

```
        if (y == 10) {
```

```
            ...
```


Sound Constraint Generation

lowers Coverage

```
foo(int x, int y) {
```

```
  if (x != complex(y)) {
```

```
    if (y == 10) {
```

```
      ...
```

Satisfiability Modulo Theories (SMT)

**Is formula F satisfiable
modulo theory T ?**

SMT solvers have
specialized algorithms for T

Satisfiability Modulo Theories (SMT)

$b + 2 = c$ and $f(\text{read}(\text{write}(a,b,3), c-2)) \neq f(c-b+1)$

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Arithmetic

Satisfiability Modulo Theories (SMT)

$b + 2 = c$ and $f(\text{read}(\text{write}(a, b, 3), c - 2)) \neq f(c - b + 1)$

Array Theory

Satisfiability Modulo Theories (SMT)

$b + 2 = c$ and $f(\text{read}(\text{write}(a, b, 3), c - 2)) \neq f(c - b + 1)$

Uninterpreted
Functions

Satisfiability Modulo Theories (SMT)

$$b + 2 = c \text{ and } f(\text{read}(\text{write}(a,b,3), c-2)) \neq f(c-b+1)$$

Substituting c by $b+2$

Satisfiability Modulo Theories (SMT)

$$b + 2 = c \text{ and } f(\text{read}(\text{write}(a,b,3), \textcolor{red}{b+2-2})) \neq f(\textcolor{red}{b+2-b+1})$$

Simplifying

Satisfiability Modulo Theories (SMT)

$b + 2 = c$ and $f(\text{read}(\text{write}(a,b,3), b)) \neq f(3)$

Satisfiability Modulo Theories (SMT)

$$b + 2 = c \text{ and } f(\text{read}(\text{write}(a,b,3), b)) \neq f(3)$$

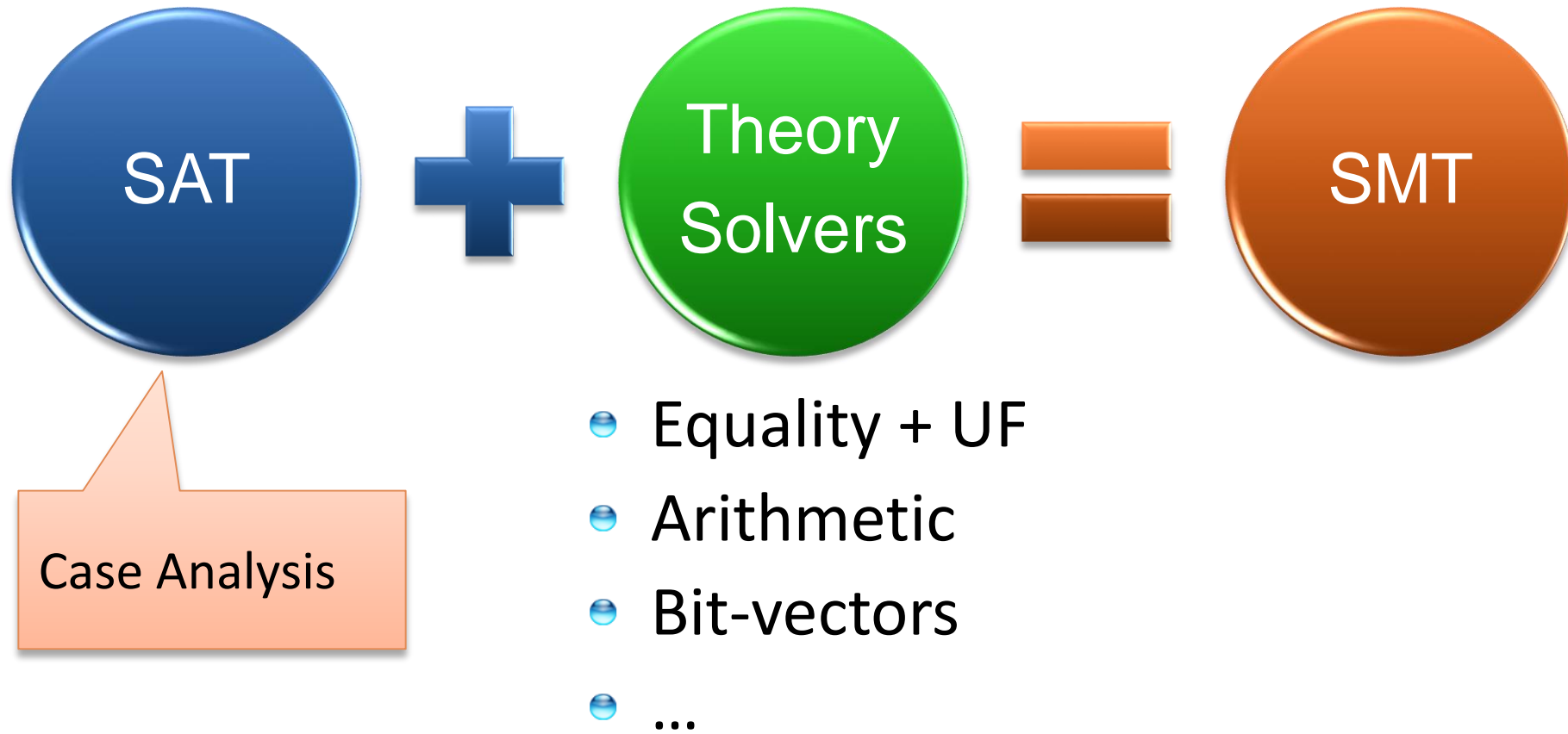
Applying array theory axiom
forall a, i, v : $\text{read}(\text{write}(a, i, v), i) = v$

Satisfiability Modulo Theories (SMT)

$b + 2 = c$ and $f(3) \neq f(3)$

Inconsistent

SMT : Basic Architecture



DPLL

The Davis–Putnam–Logemann–Loveland (DPLL) algorithm tries to find a satisfying assignment for logic formulae in conjunctive normal form (CNF).

DPLL is a complete, backtracking-based search algorithm for deciding the satisfiability of propositional logic formulae in conjunctive normal form.

DPLL (abstract view)

evolving

$M \mid F$

Partial model

Set of clauses

DPLL (abstract view)

Guessing

$$p \mid p \vee q, \neg q \vee r$$



$$p, \neg q \mid p \vee q, \neg q \vee r$$

DPLL (abstract view)

Deducing

$$p \mid p \vee q, \neg p \vee s$$

$$p, s \mid p \vee q, \neg p \vee s$$

DPLL (abstract view)

Backtracking

$p, \neg s, q \mid p \vee q, s \vee q, \neg p \vee \neg q$



$p, s \mid p \vee q, s \vee q, \neg p \vee \neg q$

SAT + Theory solvers

Basic Idea

$$x \geq 0, y = x + 1, (y > 2 \vee y < 1)$$



Abstract (aka “naming” atoms)

$$p_1, p_2, (p_3 \vee p_4) \quad \begin{array}{l} p_1 \equiv (x \geq 0), p_2 \equiv (y = x + 1), \\ p_3 \equiv (y > 2), p_4 \equiv (y < 1) \end{array}$$

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SAT
Solver

SAT + Theory solvers

Basic Idea

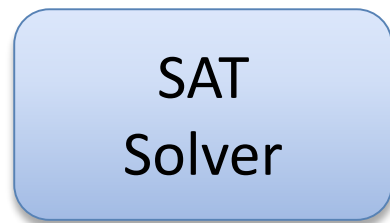
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Assignment

$p_1, p_2, \neg p_3, p_4$

SAT + Theory solvers

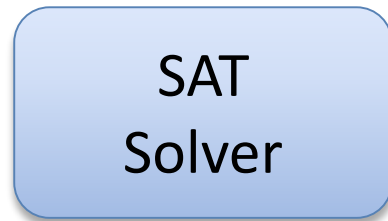
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SAT + Theory solvers

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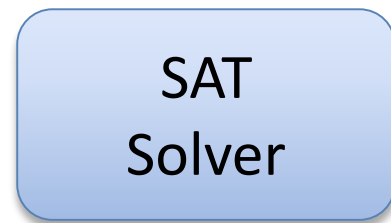
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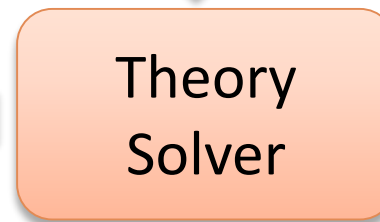


Assignment

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Unsatisfiable

$$x \geq 0, y = x + 1, y < 1$$



SAT + Theory solvers

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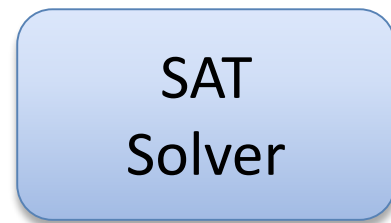
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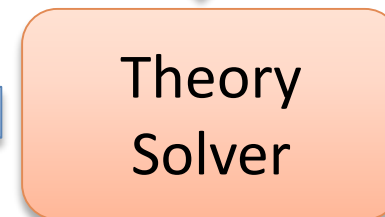


Assignment

$$p_1, p_2, \neg p_3, p_4$$



$$x \geq 0, y = x + 1, \\ \neg(y > 2), y < 1$$



Unsatisfiable

$$x \geq 0, y = x + 1, y < 1$$



New Lemma

$$\neg p_1 \vee \neg p_2 \vee \neg p_4$$



Array Theory

$$\forall a, i, v. \text{select}(\text{store}(a, i, v), i) = v$$

$$\forall a, i, j, v: i = j \vee \text{select}(\text{store}(a, i, v), j) = \text{select}(a, j)$$

Array Theory: a more familiar notation

$$\forall a, i, v. \text{select}(\text{store}(a, i, v), i) = v$$

$$\forall a, i, j, v: i = j \vee \text{select}(\text{store}(a, i, v), j) = \text{select}(a, j)$$



$$\forall a, i, v. \text{store}(a, i, v)[i] = v$$

$$\forall a, i, j, v: i = j \vee \text{store}(a, i, v)[j] = a[i]$$

Extentional Array Theory

$$\forall a, b: (\forall i: a[i] = b[i]) \Rightarrow a = b$$

Arrays are actually “maps”

- We have arrays from D to R
- D does not need to be the Integers

Models for arrays are “finite graphs”

$a = \text{store}(b, 0, 5), b = \text{store}(c, 1, 10), c[0] = 2$

$M(a) = \{ 0 \rightarrow 5, 1 \rightarrow 10, \text{else} \rightarrow 0 \}$

$M(b) = \{ 0 \rightarrow 2, 1 \rightarrow 10, \text{else} \rightarrow 0 \}$

$M(c) = \{ 0 \rightarrow 2, \text{else} \rightarrow 0 \}$

Z3 API for Arrays

- `Array(name,dom,range)`
- `Select(a,i)`
- `Update(a,i,v)`

Assignment: create a SymbolicDictionary

- Use Z3's array theory to support Python's dictionary (dict), modelling the following operations
 - Length: `__length__(self)`
 - Get: `__getitem__(self,key)`
 - Set: `__setitem__(self,key,value)`
 - Lookup: `__contains__(self,key)`
- Should support SymbolicInteger as a key and a value
- What about?
 - Delete: `__delitem__(self,key)`