

Satisfiability Modulo Theories and Network Verification

Nikolaj Bjørner
Microsoft Research
Formal Methods and Networks Summer School
Ithaca, June 10-14 2013

Lectures

Wednesday 2:00pm-2:45pm:

An Introduction to SMT with Z3

Thursday 11:00am-11:45am

Algorithmic underpinnings of SAT/SMT

Friday 9:00am-9:45am

Theories, Solvers and Applications

Plan

- I. Satisfiability Modulo Theories in a nutshell
- II. SMT solving in a nutshell
- III. SMT by example

Takeaways:

- Modern SMT solvers are a often good fit for program analysis tools.
 - Handle domains found in programs directly.
- The selected examples are intended to show instances where sub-tasks are reduced to SMT/Z3.

Wasn't that easy?!

Problems with bugs in your code?

Doctor Rustan's tool to the rescue

Get to know how debugging your code gets the simple look and feel of spell checking in Word.* See some of the latest and most exciting research in formal verification employed in action. This will be a hands-on tutorial, so bring your own laptop to try it for yourself.



Rustan Leino from Microsoft Research is a world leading expert in the area. Those who have seen his presentations know how programming is cool.

You don't want to miss this!

When: Tuesday March 20, 2012 at 13:15 - 15:00

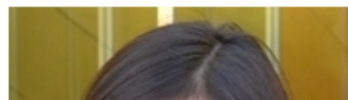
Where: E1, Osquars backe 2, KTH

<http://www.csc.kth.se/tcs/seminarsevents/rustanleino.php>

*) Your mileage may vary. Do not use when operating heavy machinery. Prolonged excitement from using programming tools may cure drowsiness. Some users report a sensation of increased and irresistible social attraction. If you experience bug withdrawal, consider collecting pet armadillidiidae.

Jean Yang

I am a fifth-year Ph.D. student in the Computer-Aided Program



- A Language for Automatically Enforcing Privacy Policies. Lezama. POPL 2012. [Paper: [pdf](#) | Slides: [pptx](#) [pdf](#) | BibTeX]
- Secure Distributed Programming with Value-Dependent Types. Pierre-Yves Strub, Karthikeyan Bharagavan, and Jean Yang. PLDI 2011. [Paper: [pdf](#) | BibTeX]
- Safe to the Last Instruction: Automated Verification of Program Transformations. Chris Hawblitzel. PLDI 2010. Best paper award. [Paper: [pdf](#) | BibTeX] This work was selected as a CACM Research Highlight (First!) by Xavier Leroy. [Full text: [html](#) [pdf](#) | Technical Report]

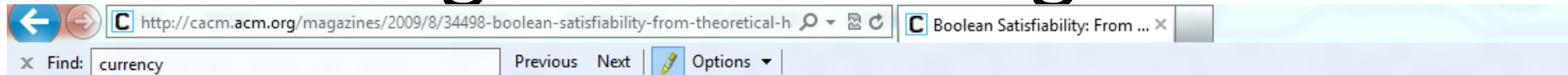


Z3 – Backed by Proof Plumbers



Leonardo de Moura, Nikolaj Bjørner, Christoph Wintersteiger

Background Reading: SAT



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Home / Magazine Archive / August 2009 (Vol. 52, No. 8) / Boolean Satisfiability: From Theoretical Hardness... / Full Text

REVIEW ARTICLES

Boolean Satisfiability: From Theoretical Hardness to Practical Success

By Sharad Malik, Lintao Zhang

Communications of the ACM, Vol. 52 No. 8, Pages 76-82

10.1145/1536616.1536637

[Comments](#)

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There are many practical situations where we need to satisfy several potentially conflicting constraints. Simple examples of this abound in daily life, for example, determining a schedule for a series of games that resolves the availability of players and venues, or finding a seating assignment at dinner consistent with various rules the host would like to impose. This also applies to applications in computing, for example, ensuring that a hardware/software system functions correctly with its overall behavior constrained by the behavior of its components and their composition, or finding a plan for a robot to reach a goal that is

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[Introduction](#)

[Boolean Satisfiability](#)

[Theoretical hardness: SAT and NP Completeness](#)

Background Reading: SMT



September 2011

Satisfiability Modulo Theories: Introduction & Applications

Leonardo de Moura
Microsoft Research
One Microsoft Way
Redmond, WA 98052
leonardo@microsoft.com

Nikolaj Bjørner
Microsoft Research
One Microsoft Way
Redmond, WA 98052
nbjorner@microsoft.com

ABSTRACT

Constraint satisfaction problems arise in many diverse surrounding software and hardware verification, type inferencing, program analysis, test-case generation, scheduling and graph problems. These areas share a common trait, they include a core component using logical formulas for describing states and transformations between them. The most well-known constraint satisfaction problem is propositional satisfiability, SAT, where the goal is to determine whether a formula over Boolean variables, formed using logical connectives can be made *true* by choosing *true/false* for its variables. Some problems are more naturally expressed using richer languages, such as arithmetic. A superset theory (of arithmetic) is then required to capture the meaning of these formulas. Solvers for such formulations are commonly called *Satisfiability Modulo Theories* (SMT)

SMT solvers have been the focus of increased recent attention thanks to technological advances and industrial applications. Yet, they draw on a combination of some of the most fundamental areas in computer science as well as discoveries from the past century of symbolic logic. They combine the problem of Boolean Satisfiability with domains, such as, those studied in convex optimization and term-manipulating symbolic systems. They involve the decision problem, completeness and incompleteness of logical theories, and finally complexity theory. In this article, we present an overview of the field of Satisfiability Modulo Theories, and some of its applications.

key driving factor [4]. An important ingredient is a common interchange format for benchmarks, called SMT-LIB [33], and the classification of benchmarks into various categories depending on which theories are required. Conversely, a growing number of applications are able to generate benchmarks in the SMT-LIB format to further inspire improving SMT solvers.

There is a relatively long tradition of using SMT solvers in select and specialized contexts. One prolific case is theorem proving systems such as ACL2 [26] and PVS [32]. These use decision procedures to discharge lemmas encountered during interactive proofs. SMT solvers have also been used for a long time in the context of program verification and *extended static checking* [21], where verification is focused on assertion checking. Recent progress in SMT solvers, however, has enabled their use in a set of diverse applications, including interactive theorem provers and extended static checkers, but also in the context of scheduling, planning, test-case generation, model-based testing and program development, static program analysis, program synthesis, and run-time analysis, among several others.

We begin by introducing a motivating application and a simple instance of it that we will use as a running example.

1.1 An SMT Application - Scheduling

Consider the classical *job shop scheduling* decision problem. In this problem, there are n jobs, each composed of m tasks of varying duration that have to be performed consecutively on m machines. The start of a new task can be delayed as long as needed in order to wait for a machine to become available, but tasks cannot be interrupted once

SAT IN A NUTSHELL

SAT in a nutshell

$(\text{Tie} \vee \text{Shirt}) \wedge (\neg \text{Tie} \vee \neg \text{Shirt}) \wedge (\neg \text{Tie} \vee \text{Shirt})$

SMT IN A NUTSHELL

Satisfiability Modulo Theories (SMT)

Is formula φ satisfiable
modulo theory T ?

SMT solvers have
specialized algorithms for T

Satisfiability Modulo Theories (SMT)

$x+2=y \Rightarrow f(\text{select}(\text{store}(a,x,3),y-2))=f(y-x+1)$

Array Theory

Arithmetic

Uninterpreted
Functions

$\text{select}(\text{store}(a,i,v),i)=v$
 $i \neq j \Rightarrow \text{select}(\text{store}(a,i,v),j)=\text{select}(a,j)$

SMT SOLVING IN A NUTSHELL

Job Shop Scheduling

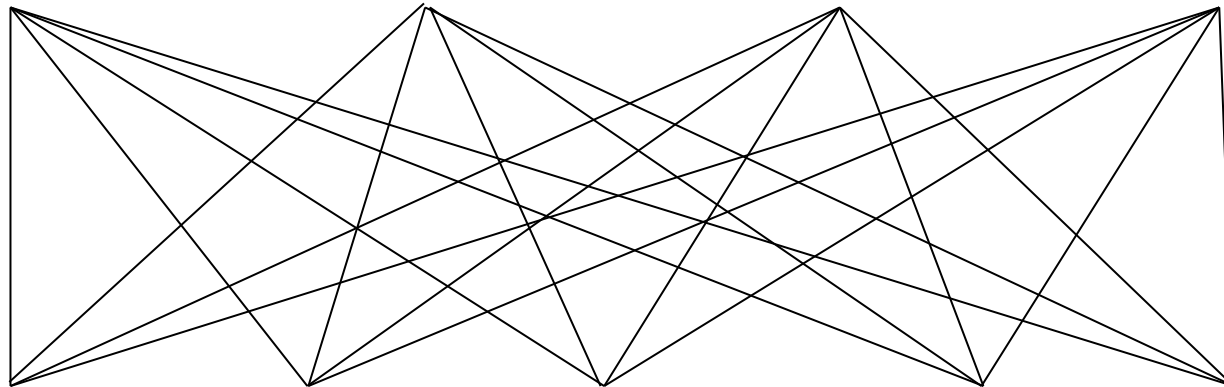
Job Shop Scheduling



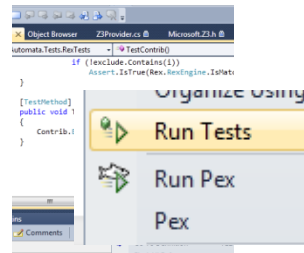
Machines

Tasks

Jobs



$P = NP?$

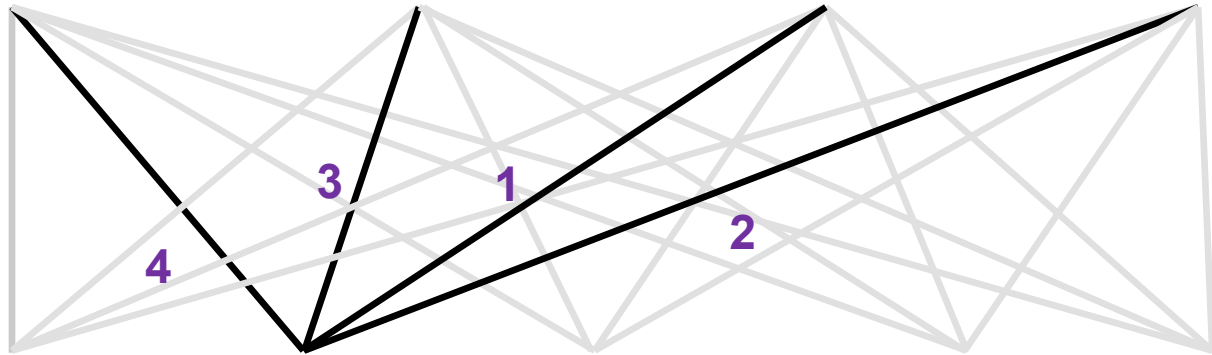


$$\zeta(s)=0 \Rightarrow s=1/2 + ir$$

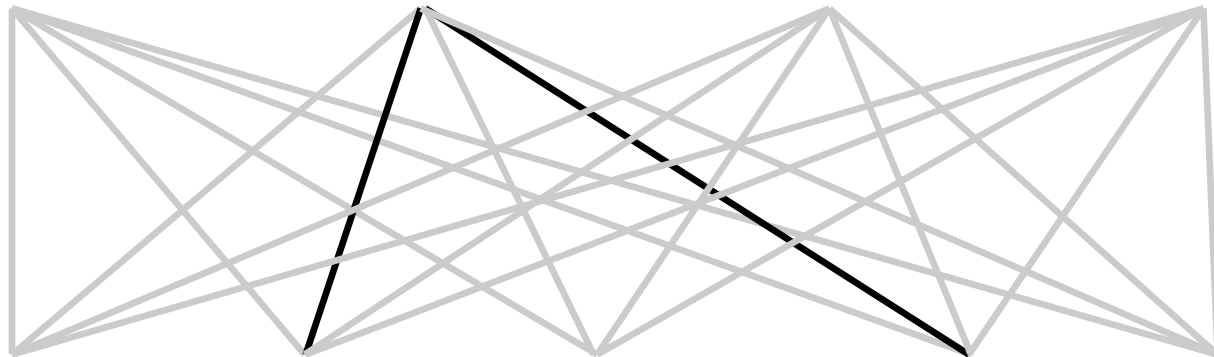
Job Shop Scheduling

Constraints:

Precedence: between two tasks of the same job



Resource: Machines execute at most one job at a time

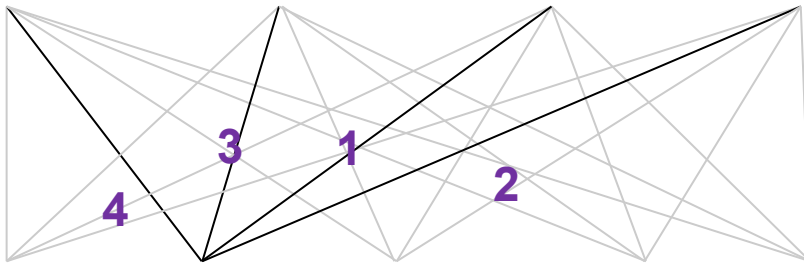


$$[start \downarrow 2, 2 .. end \downarrow 2, 2] \cap [start \downarrow 4, 2 .. end \downarrow 4, 2] = \emptyset$$

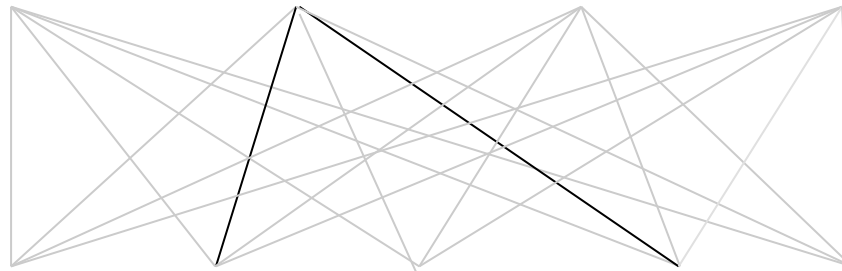
Job Shop Scheduling

Constraints:

Precedence:



Resource:



$$[start_{\downarrow 2,2} .. end_{\downarrow 2,2}] \cap [start_{\downarrow 4,2} .. end_{\downarrow 4,2}] = \emptyset$$

Encoding:

- start time of job 2 on mach 3
- duration of job 2 on mach 3

Not convex

$$t_{\downarrow 2,2} + d_{\downarrow 2,2} \leq$$

$$t_{\downarrow 4,2}$$

\vee

$$t_{\downarrow 4,2} + d_{\downarrow 4,2} \leq$$

$$t_{\downarrow 2,2}$$

Job Shop Scheduling

$d_{i,j}$	Machine 1	Machine 2
Job 1	2	1
Job 2	3	1
Job 3	2	3

$max = 8$

Solution

$t_{1,1} = 5, t_{1,2} = 7, t_{2,1} = 2,$
 $t_{2,2} = 6, t_{3,1} = 0, t_{3,2} = 3$

Encoding

$(t_{1,1} \geq 0) \wedge (t_{1,2} \geq t_{1,1} + 2) \wedge (t_{1,2} + 1 \leq 8) \wedge$
 $(t_{2,1} \geq 0) \wedge (t_{2,2} \geq t_{2,1} + 3) \wedge (t_{2,2} + 1 \leq 8) \wedge$
 $(t_{3,1} \geq 0) \wedge (t_{3,2} \geq t_{3,1} + 2) \wedge (t_{3,2} + 3 \leq 8) \wedge$
 $((t_{1,1} \geq t_{2,1} + 3) \vee (t_{2,1} \geq t_{1,1} + 2)) \wedge$
 $((t_{1,1} \geq t_{3,1} + 2) \vee (t_{3,1} \geq t_{1,1} + 2)) \wedge$
 $((t_{2,1} \geq t_{3,1} + 2) \vee (t_{3,1} \geq t_{2,1} + 3)) \wedge$
 $((t_{1,2} \geq t_{2,2} + 1) \vee (t_{2,2} \geq t_{1,2} + 1)) \wedge$
 $((t_{1,2} \geq t_{3,2} + 3) \vee (t_{3,2} \geq t_{1,2} + 1)) \wedge$
 $((t_{2,2} \geq t_{3,2} + 3) \vee (t_{3,2} \geq t_{2,2} + 1))$

Job Shop Scheduling

$$\begin{aligned}
 & (t_{1,1} \geq 0) \wedge (t_{1,2} \geq t_{1,1} + 2) \wedge (t_{1,2} + 1 \leq 8) \wedge \\
 & (t_{2,1} \geq 0) \wedge (t_{2,2} \geq t_{2,1} + 3) \wedge (t_{2,2} + 1 \leq 8) \wedge \\
 & (t_{3,1} \geq 0) \wedge (t_{3,2} \geq t_{3,1} + 2) \wedge (t_{3,2} + 3 \leq 8) \wedge \\
 & ((t_{1,1} \geq t_{2,1} + 3) \vee (t_{2,1} \geq t_{1,1} + 2)) \wedge \\
 & ((t_{1,1} \geq t_{3,1} + 2) \vee (t_{3,1} \geq t_{1,1} + 2)) \wedge \\
 & ((t_{2,1} \geq t_{3,1} + 2) \vee (t_{3,1} \geq t_{2,1} + 3)) \wedge \\
 & ((t_{1,2} \geq t_{2,2} + 1) \vee (t_{2,2} \geq t_{1,2} + 1)) \wedge \\
 & ((t_{1,2} \geq t_{3,2} + 3) \vee (t_{3,2} \geq t_{1,2} + 1)) \wedge \\
 & ((t_{2,2} \geq t_{3,2} + 3) \vee (t_{3,2} \geq t_{2,2} + 1))
 \end{aligned}$$

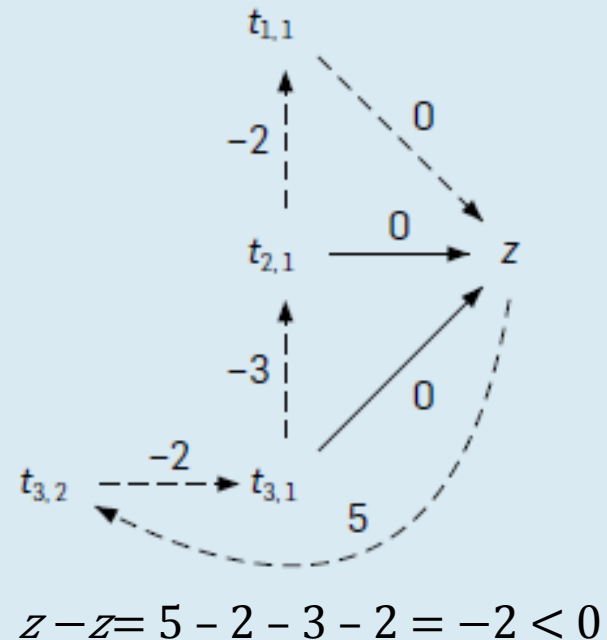
case split

case split

$$\begin{array}{rcll}
 z & - & t_{1,1} & \leq 0 \\
 z & - & t_{2,1} & \leq 0 \\
 z & - & t_{3,1} & \leq 0 \\
 t_{3,2} & - & z & \leq 5 \\
 t_{3,1} & - & t_{3,2} & \leq -2 \\
 t_{2,1} & - & t_{3,1} & \leq -3 \\
 t_{1,1} & - & t_{2,1} & \leq -2
 \end{array}$$

Efficient solvers:

- Floyd-Warshal algorithm
- Ford-Fulkerson algorithm



THEORIES

Theories

Uninterpreted functions



Is this formula satisfiable? Ask z3!

```
1 (declare-sort () A)
2 (declare-fun f (A) A)
3 (declare-const a A)
4 (assert (= a (f (f a))))
5 (assert (= a (f (f (f a)))))
6 (check-sat)
7 (get-model)
8 (echo "Adding contradiction")
9 (assert (not (= a (f a))))
10 (check-sat)
```

ask z3

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Theories **z3py**

Explore the Z3 API using Python

```

1 t11, t12, t21, t22, t31, t32 = Ints('t11 t12 t21 t22 t31 t32')
2
3 s = Solver()
4
5 s.add(And([t11 >= 0, t12 >= t11 + 2, t12 + 1 <= 8]))
6 s.add(And([t21 >= 0, t22 >= t21 + 3, t22 + 1 <= 8]))
7 s.add(And([t31 >= 0, t32 >= t31 + 2, t32 + 3 <= 8]))
8
9 s.add(Or(t11 >= t21 + 3, t21 >= t11 + 2))
10 s.add(Or(t11 >= t31 + 2, t31 >= t11 + 2))
11 s.add(Or(t21 >= t31 + 2, t31 >= t21 + 3))
12 s.add(Or(t21 >= t22 + 1, t22 >= t12 + 1))
13 s.add(Or(t12 >= t32 + 3, t32 >= t12 + 1))
14 s.add(Or(t22 >= t32 + 3, t32 >= t22 + 1))
15 |
16 print ">>", s.check()
17 print ">>", s.model()
18
19

```



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'▶' shortcut: Alt+B

>> sat

>> [t31 = 0, t21 = 4, t22 = 7, t32 = 2, t12 = 5, t11 = 2]

Uninterpreted function
Arithmetic (linear)

Theories



Explore the Z3 API using Python

```
1
2
3 x      = BitVec('x', 32)
4 powers = [ 2**i for i in range(32) ]
5 fast   = And(x != 0, x & (x - 1) == 0)
6 slow   = Or([ x == p for p in powers ])
7
8
9 prove(fast == slow)
10
11 print "buggy version..."
12
13 fast   = x & (x - 1) == 0
14
15
16 prove(fast == slow)
17
18
19
20
```



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'>' shortcut: Alt+B

proved

buggy version...

counterexample

[x = 0]

Uninterpreted functions

Arithmetic (linear)

Bit-vectors

Theories

Uninterpreted functions

Arithmetic (linear)

Bit-vectors

Algebraic data-types



Explore the Z3 API using Python

```
1 List = Datatype('List')
2 List.declare('cons', ('car', IntSort()), ('cdr', List))
3 List.declare('nil')
4 List = List.create()
5 cons = List.cons
6 car = List.car
7 cdr = List.cdr
8 nil = List.nil
9 l1 = cons(10, cons(20, nil))
10
11 print ">>", simplify(cdr(l1))
12
13 print ">>", simplify(car(l1))
14
15 print ">>", simplify(l1 == nil)
16
17
18 x, y = Ints('x y')
19 l1 = Const('l1', List)
20 l2 = Const('l2', List)
21 s = Solver()
```



tutorial

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'>' shortcut: Alt+B

Theories

Uninterpreted functions

Arithmetic (linear)

Bit-vectors

Algebraic data-types

[Arrays](#)

```
2 ; supported in Z3.
3 ; This includes Combinatory Array Logic (de Moura &
4 ;
5 (define-sort A () (Array Int Int))
6 (declare-fun x () Int)
7 (declare-fun y () Int)
8 (declare-fun z () Int)
9 (declare-fun a1 () A)
10 (declare-fun a2 () A)
11 (declare-fun a3 () A)
12 (push) ; illustrate select-store
13 (assert (= (select a1 x) x))
14 (assert (= (store a1 x y) a1))
15 (check-sat)
16 (get-model)
17 (assert (not (= x y)))
18 (check-sat)
19 (pop)
20 (define-fun all1_array () A ((as const A) 1))
21 (simplify (select all1_array x))
22 (define-sort IntSet () (Array Int Bool))
```

ask z3

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```
sat
(model
  (define-fun y () Int
    1)
  (define-fun a1 () (Array Int Int)
    (_ as-array k!0))
  (define-fun x () Int
    1)
  (define-fun k!0 ((x!1 Int)) Int
    (ite (= x!1 1) 1
```

Theories

Uninterpreted functions

Arithmetic (linear)

Bit-vectors

Algebraic data-types

Arrays


[Polynomial Arithmetic](#)

Microsoft Research

z3py



Explore the Z3 API using Python

```
1 x, y, z = Reals('x y z')
2
3 solve(x**2 + y**2 < 1, x*y > 1,
4       show=True)
5
6 solve(x**2 + y**2 < 1, x*y > 0.4,
7       show=True)
8
9 solve(x**2 + y**2 < 1, x*y > 0.4, x < 0,
10      show=True)
11
12 solve(x**5 - x - y == 0, Or(y == 1, y == -1),
13      show=True)
14
```

 **tutorial** **home** **permalink**
'>' shortcut: Alt+B

[samples](#)
[solve](#)
[simple](#)
[strategy](#)

about Z3Py - Python interface for the Z3
Z3 is a high-performance theorem prover. Z3 supports
extensional arrays, datatypes, uninterpreted functions

 Like 45  reddit this!

QUANTIFIERS

Equality-Matching

$$\blacksquare \vdash (\forall \dots) \wedge a = g(b, b) \wedge b = c \wedge f(a) \neq c \vdash (\forall x \dots) \rightarrow f(g(c, b)) \neq f(g(c, c))$$

$g(c, x)$ matches $g(b, b)$
 with substitution $[x \mapsto b]$
 modulo $b = c$

Quantifier Elimination

```
1
2 (define-fun stamp () Bool
3   (forall((x Int))
4     (=>
5       (>= x 8)
6       (exists ((u Int) (v Int))
7         (and (>= u 0) (>= v 0) (= x (+ (* 3 u) (* 5 v)))))))
8
9 (simplify stamp)
10
11 (elim-quantifiers stamp)
```

Presburger Arithmetic,
Algebraic Data-types,
Quadratic polynomials

MBQI: Model based Quantifier Instantiation

```
(set-option :mbqi true)
(declare-fun f (Int Int) Int)
(declare-const a Int)
(declare-const b Int)

(assert (forall ((x Int)) (>= (f x x) (+ x a))))

(assert (< (f a b) a))
(assert (> a 0))
(check-sat)
(get-model)

(echo "evaluating (f (+ a 10) 20)...")
(eval (f (+ a 10) 20))
```

[de Moura, Ge. CAV 2008]

[Bonachnia, Lynch, de Moura CADE 2009]

[de Moura, B. IJCAR 2010]

Superposition

■ 1. $\forall x. (x; id) = x$ 2. $\forall x. (id; x) = x$ 3. $\forall x. (id \mid x) = x$ 4.

$\forall x y z u. (x \mid y); (z \mid u) \leq (x; z) \mid (y; u) \wedge \forall p q. (p; q) \leq (p \mid q)$

5. $\forall x z u. x; (z \mid u) \leq (x; z) \mid (id; u)$ super-pose 1, 4

6. $\forall x z u. x; (z \mid u) \leq (x; z) \mid u$ super-pose 2, 5

7. $\forall x z u. x; u \leq (x; id) \mid u$ super-pose 3, 6

8. $\forall x z u. x; u \leq x \mid u$ super-pose 1, 7

[de Moura, B. IJCAR 2008]

(disabled in release 4.1)

Horn Clauses

mc(x) = x-10 **if x > 100**

mc(x) = mc(mc(x+11)) if $x \leq 100$

assert (mc(x) ≥ 91)

$$\forall X. X > 100 \rightarrow \text{mc}(X, X-10)$$
$$\forall X, Y, R. \quad X \leq 100 \wedge \text{mc}(X+11, Y) \wedge \text{mc}(Y, R) \rightarrow \text{mc}(X, R)$$
$$\forall X, R. \text{mc}(X, R) \wedge X \leq 101 \rightarrow R = 91$$

Solver finds solution for mc

[Hoder, B. SAT 2012]

MODELS, PROOFS, CORES & SIMPLIFICATION

Models

Click on a tool to load a sample then ask!

agl bek boogie code contracts concurrent revisions
dafny esm fine heapdbg poirot pex rex spec# vcc
z3

```
(define-sorts ((A (Array Int Int))))  
(declare-funs ((x Int) (y Int) (z Int)))  
(declare-funs ((a1 A) (a2 A) (a3 A)))  
(assert (= (select a1 x) x))  
(assert (= (store a1 x y) a1))  
(check-sat)  
(get-info model)
```



Logical Formula

ask z3

Is this SMT formula satisfiable?
Click 'ask Z3'! Read more or watch the
video.

```
sat  
(("model" "  
(define x 0)  
(define a1 as-array[k!0])  
(define y 0)  
(define (k!0 (x1 Int))  
(if (= x1 0) 0  
1)))")
```



Sat/Model

Proofs

Logical Formula

```
(set-logic QF_LIA)
(declare-funs ((x Int) (x1 Int)))
(declare-funs ((x3 Int) (x2 Int)))
(declare-funs ((x4 Int) (x5 Int)))
(declare-funs ((y Int) (z Int) (u Int)))
(assert (> x y))
(assert (= (- (* x 3) (* y 3)) (- z u))) proof.smt2 PROOF_MODE=2
(assert (<= 0 z))
(assert (<= 0 u))
(assert (< z 3))
(assert (< u 3))
(check-sat)
(get-proof)

ted (<= 0 u)] [rewrite (iff (<= 0 u) (<= u 0))] (<= u 0)]
in

erted (= (- (* x 3) (* y 3)) (- z u))]
ns
onotonicity
ltrans
lmonotonicity
  [rewrite (= (* x 3) (* 3 x))]
  [rewrite (= (* y 3) (* 3 y))]
  (= (- (* x 3) (* y 3)) (- (* 3 x) (* 3 y)))]
  [rewrite (= (- (* 3 x) (* 3 y)) (+ (* 3 x) (* -3 y)))]
  (= (- (* x 3) (* y 3)) (+ (* 3 x) (* -3 y)))]
  [rewrite (= (- z u) (+ z (* -1 u)))]
  (iff (= (- (* x 3) (* y 3)) (- z u))
    (= (+ (* 3 x) (* -3 y)) (+ z (* -1 u))))]
  [rewrite
    (iff (= (+ (* 3 x) (* -3 y)) (+ z (* -1 u)))
      (= (+ (* 3 x) (+ (* -3 y) (+ (* -1 z) u))) 0))]
  (iff (= (- (* x 3) (* y 3)) (- z u))
    (= (+ (* 3 x) (+ (* -3 y) (+ (* -1 z) u))) 0))]
  (= (+ (* 3 x) (+ (* -3 y) (+ (* -1 z) u))) 0)]
  [rewrite
    (iff (= (+ (* 3 x) (+ (* -3 y) (+ (* -1 z) u))) 0)
      (not (or (not (<= (+ (* 3 x) (+ (* -3 y) (+ (* -1 z) u))) 0))
        (not (<= (+ (* 3 x) (+ (* -3 y) (+ (* -1 z) u))) 0))))))]
  (not (or (not (<= (+ (* 3 x) (+ (* -3 y) (+ (* -1 z) u))) 0))
    (not (<= (+ (* 3 x) (+ (* -3 y) (+ (* -1 z) u))) 0))))]
  (<= (+ (* 3 x) (+ (* -3 y) (+ (* -1 z) u))) 0)]
Imp
  [asserted (< x y)]
  [rewrite (iff (< x y) (not (<= (+ x (* -1 y)) 0)))]
  (not (<= (+ x (* -1 y)) 0))]
Imp [asserted (< z 3)] [rewrite (iff (< z 3) (not (<= z 3)))] (not (<= z 3))]
false]
```

Unsat/Proof

Simplification

R1SE4fun

gave 48,503 answers!
Click on a tool to load a sample then ask!

agl bek boogie code contracts
concurrent revisions dafny esm fine
heapdbg poirot pex rex spec# vcc
z3

```
(declare-fun x () Real)
(declare-fun y () Real)
(simplify (>= x (+ x y)))
```

ask z3

*Is this SMT formula
satisfiable? Click 'ask*

z3'! Read more or watch the video.

(<= y 0.0)

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Logical Formula



Simplify

Cores

```
(declare-preds ((p) (q) (r) (s)))  
(set-option enable-cores)  
(assert (or p q))  
(assert (implies r s))  
(assert (implies s (iff q r)))  
(assert (or r p))  
(assert (or r s))  
(assert (not (and r q)))  
(assert (not (and s p)))  
(check-sat)  
(get-unsat-core)
```



Logical Formula

ask z3

Is this SMT formula satisfiable?
Click 'ask Z3'! Read more or watch
the video.

```
unsat  
((or p q)  
 (=> r s)  
 (or r p)  
 (or r s)  
 (not (and r q))  
 (not (and s p)))
```



Unsat. Core

TACTICS, SOLVERS

Tactics

```
(declare-const x (_ BitVec 16))
(declare-const y (_ BitVec 16))

(assert (= (bvor x y) (_ bv13 16)))
(assert (bvslt x y))

(check-sat-using (then simplify solve-eqs bit-blast sat))
(get-model)
```

Composition of tactics:

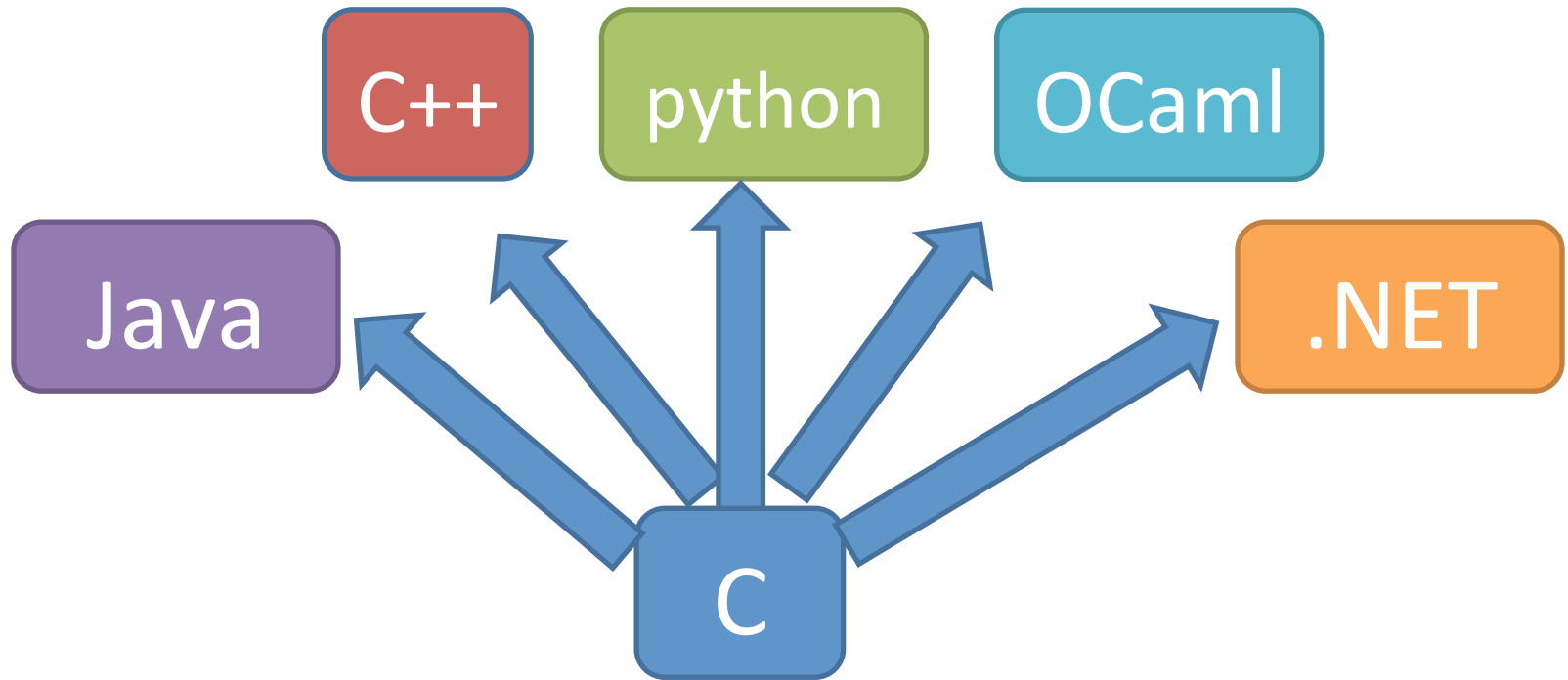
- (then t s)
- (par-then t s) applies t to the input goal and S to every subgoal produced by t in parallel.
- (or-else t s)
- (par-or t s) applies t and S in parallel until one of them succeed.
- (repeat t)
- (repeat t n)
- (try-for t ms)
- (using-params t params) Apply the given tactic using the given parameters.

Solvers

- Tactics take goals and reduce to sub-goals
- Solvers take tactics and serve as logical contexts.
 - push
 - add
 - check
 - model, core, proof
 - pop

```
bv_solver = Then(With('simplify', mul2concat=True),
                  'solve-eqs',
                  'bit-blast',
                  'aig',
                  'sat').solver()
x, y = BitVecs('x y', 16)
bv_solver.add(x*32 + y == 13, x & y < 10, y > -100)
print bv_solver.check()
m = bv_solver.model()
print m
print x*32 + y, "==", m.evaluate(x*32 + y)
print x & y, "==", m.evaluate(x & y)
```

APIS



Summary

Z3 supports several theories

- Using a default combination
- Providing custom tactics for special combinations

Z3 is more than sat/unsat

- Models, proofs, unsat cores,
- simplification, quantifier elimination are tactics

Prototype with python/smt-lib2

- Implement using smt-lib2/programmatic API