

SMT Solver Z3

Testing, Quality Assurance, and Maintenance
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Satisfiability Modulo Theory (SMT)

Satisfiability is the problem of determining whether a formula F has a model

- if F is **propositional**, a model is a truth assignment to Boolean variables
- if F is **first-order formula**, a model assigns values to variables and interpretation to all the function and predicate symbols

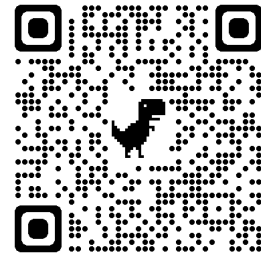
SAT Solvers

- check satisfiability of propositional formulas

SMT Solvers

- check satisfiability of formulas in a **decidable** first-order theory (e.g., linear arithmetic, uninterpreted functions, array theory, bit-vectors)

(Optional) Background Reading: SMT



Satisfiability Modulo Theories: Introduction & Applications

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ABSTRACT

Constraint satisfaction problems arise in many diverse areas including software and hardware verification, type inference, program analysis, test-case generation, scheduling, and graph problems. These areas share a common trait, they include a core component using logical theories 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 propositional 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)

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.

... delayed as long as needed in order to wait for a machine to become available, but tasks cannot be interrupted once

September 2011

Example

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

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Arithmetic

Example

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Array theory

Example

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

Uninterpreted function

Example

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

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

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

then, the formula is **unsatisfiable**

Example 2

$$x \geq 0 \wedge f(x) \geq 0 \wedge y \geq 0 \wedge f(y) \geq 0 \wedge x \neq y$$

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This formula is **satisfiable**

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$$x \geq 0 \wedge f(x) \geq 0 \wedge y \geq 0 \wedge f(y) \geq 0 \wedge x \neq y$$

This formula is **satisfiable**:

Example model:

$$x \rightarrow 1$$

$$y \rightarrow 2$$

$$f(1) \rightarrow 0$$

$$f(2) \rightarrow 1$$

$$f(\dots) \rightarrow 0$$

SMT-LIB: <http://smt-lib.org>

International initiative for facilitating research and development in SMT
Provides rigorous definition of syntax and semantics for theories

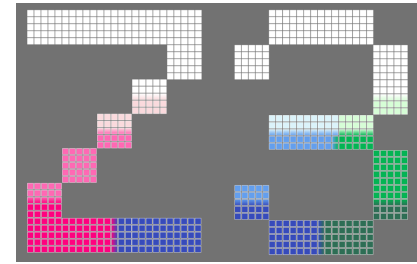
SMT-LIB syntax

- based on s-expressions (LISP-like)
 - <https://en.wikipedia.org/wiki/S-expression>
- common syntax for interpreted functions of different theories
 - e.g. `(and (= x y) (<= (* 2 x) z))`
- commands to interact with the solver
 - `(declare-fun ...)` declares a constant/function symbol
 - `(assert p)` conjoins formula p to the current context
 - `(check-sat)` checks satisfiability of the current context
 - `(get-model)` prints current model (if the context is satisfiable)
- see examples at <http://rise4fun.com/z3>

SMT-LIB Syntax

```
(declare-fun x () Int)
(declare-fun y () Int)
(declare-fun z () Int)
(assert (>= (* 2 x) (+ y z)))
(declare-fun f (Int) Int)
(declare-fun g (Int Int) Int)
(assert (< (f x) (g x x)))
(assert (> (f y) (g x x)))
(check-sat)
(get-model)
```


SMT Example in Z3



🏠 > Freeform Editing

Freeform Editing

Run Z3 on SMTLIB on the web!

```
4  (assert (>= (* z x) (+ y z)))
5
6  (declare-fun f (Int) Int)
7  (declare-fun g (Int Int) Int)
8  (assert (< (f x) (g x x)))
9  (assert (> (f y) (g x x)))
10 (check-sat)
11 (get-model)
12
13 (push)
14 (assert (= x y))
15 (check-sat)
16 (pop)
17 (exit)
```

Run

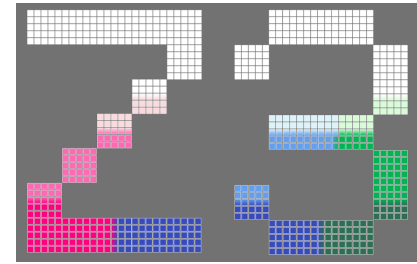
<https://microsoft.github.io/z3guide/playground/Freeform%20Editing>



Example

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

And now in Z3...



🏠 > Freeform Editing

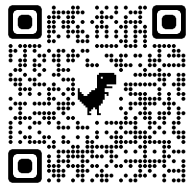
Freeform Editing

Run Z3 on SMTLIB on the web!

```
1 (set-option :produce-proofs true)
2 (declare-fun b () Int)
3 (declare-fun c () Int)
4 (declare-fun a () (Array Int Int))
5 (declare-fun f (Int) Int)
6 (assert (= (+ b 2) c))
7 (assert (not (= (f (select (store a b 3) (- c 2))) (f (+ (- c b) 1)))))
8 (check-sat)
9 (get-proof)
```

Run

<https://microsoft.github.io/z3guide/playground/Freeform%20Editing>



```

import z3

def main():
    b, c = z3.Ints("b c")
    a = z3.Array("a", z3.IntSort(), z3.IntSort())
    f = z3.Function("f", z3.IntSort(), z3.IntSort())
    solver = z3.Solver()
    solver.add(c == b + z3.IntVal(2))
    lhs = f(z3.Store(a, b, 3)[c - 2])
    rhs = f(c - b + 1)
    solver.add(lhs != rhs)
    res = solver.check()
    if res == z3.sat:
        print("sat")
    elif res == z3.unsat:
        print("unsat")
    else:
        print("unknown")

if __name__ == "__main__":
    main()

```

z3 python package

```
import z3
```

create constants

```
def main():
```

```
    b, c = z3.Ints("b c")
```

```
    a = z3.Array("a", z3.IntSort(), z3.IntSort())
```

```
    f = z3.Function("f", z3.IntSort(), z3.IntSort())
```

```
    solver = z3.Solver()
```

SMT solver

```
    solver.add(c == b + z3.IntVal(2))
```

```
    lhs = f(z3.Store(a, b, 3)[c - 2])
```

```
    rhs = f(c - b + 1)
```

```
    solver.add(lhs != rhs)
```

create constraints
and add to solver

```
    res = solver.check()
```

```
    if res == z3.sat:
```

```
        print("sat")
```

```
    elif res == z3.unsat:
```

```
        print("unsat")
```

```
    else:
```

```
        print("unknown")
```

run solver. can
take long time.

result is: sat,
unsat, unknown

```
if __name__ == "__main__":  
    main()
```

Useful Z3Py Functions

All these functions are under python package `z3`

Create constants and values

- `Int(name)` – an integer constant with a given name
- `FreshInt(name)` – unique constant starting with name
- `IntVal(v)`, `BoolVal(v)` – integer and boolean values

Arithmetic functions and predicates

- `+`, `-`, `/`, `<`, `<=`, `>`, `>=`, `==`, etc.
- `Distinct(a, b, ...)` – the arguments are distinct (expands to many disequalities)

Propositional operators

- `And`, `Or`, `Not`

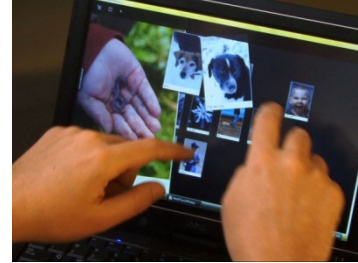
Methods of the `z3.Solver` class

- `add(fml)` – add formula fml to the solver
- `check()` – returns `z3.sat`, `z3.unsat`, or `z3.unknown` (on failure to solve)
- `model()` – model if the result is sat

Methods of `z3.Model` class

- `eval(fml)` – returns the value of fml in the model

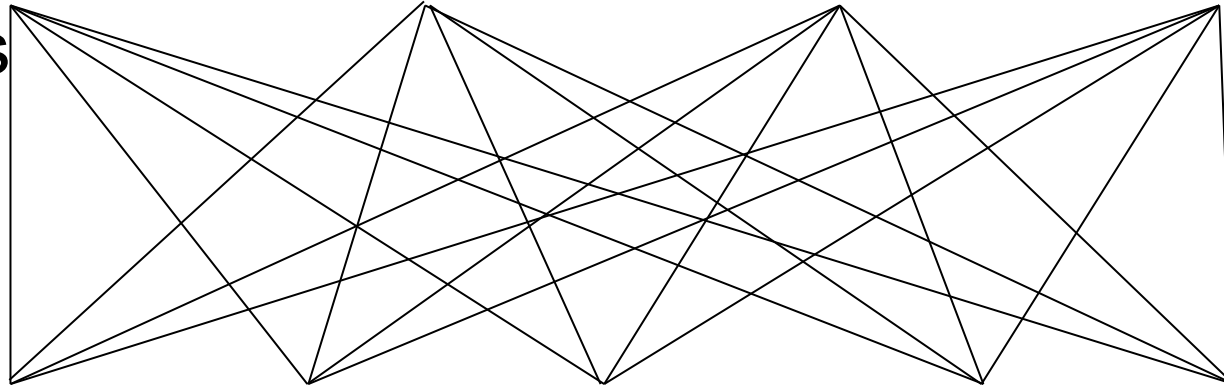
Job Shop Scheduling



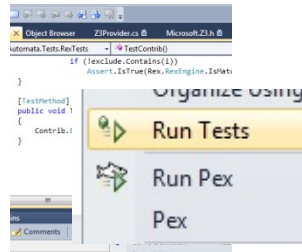
Machines

Tasks

Jobs



$P = NP?$



$$\zeta(s) = 0 \Rightarrow s = \frac{1}{2} + ir$$