

# Satisfiability Modulo Theories and the SMT Competition

Tjark Weber



UPPSALA  
UNIVERSITET

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# Introduction

Satisfiability Modulo Theories =

Propositional satisfiability + background theories

## Example: Job-Shop Scheduling

Given:  $n$  jobs, each composed of  $m$  tasks of varying duration, that must be performed consecutively on  $m$  machines; a total maximum time  $max$ .

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

$$max = 8$$

Is there a [schedule](#) such that the end-time of every task is  $\leq max$ ?

## Job-Shop Scheduling: SMT Encoding

The job-shop scheduling problem has a straightforward **encoding** in propositional logic + linear integer arithmetic.

A schedule is specified by the start time  $t_{i,j}$  for the  $j$ -th task of every job  $i$ .

Precedence constraints:

$$t_{i,1} \geq 0 \wedge t_{i,2} \geq t_{i,1} + d_{i,1} \wedge t_{i,2} + d_{i,2} \leq \max \quad (\text{for } i = 1, 2, 3)$$

Resource constraints:

$$\begin{aligned} & (t_{1,j} \geq t_{2,j} + d_{2,j} \vee t_{2,j} \geq t_{1,j} + d_{1,j}) \wedge \\ & (t_{1,j} \geq t_{3,j} + d_{3,j} \vee t_{3,j} \geq t_{1,j} + d_{1,j}) \wedge \\ & (t_{2,j} \geq t_{3,j} + d_{3,j} \vee t_{3,j} \geq t_{2,j} + d_{2,j}) \quad (\text{for } j = 1, 2) \end{aligned}$$

## Job-Shop Scheduling: Solution

SMT formula encoding

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

$$\max = 8$$



Solution:

$$t_{1,1} = 5, t_{1,2} = 7$$

$$t_{2,1} = 2, t_{2,2} = 6$$

$$t_{3,1} = 0, t_{3,2} = 3$$

# Background Theories

- ▶ EUF
- ▶ Arithmetic
- ▶ Arrays
- ▶ Bit-vectors
  
- ▶ Quantifiers
- ▶ Algebraic data types
- ▶ ...

$$x = y \implies f(x) = f(y)$$

$$y < 0 \implies x + y < x$$

$$\text{select}(\text{store}(a, i, x), i) = x$$

$$2 \cdot x = x \ll 1$$

# Applications

SMT solvers are the core engine of many tools for program analysis, testing and verification.

# Dynamic Symbolic Execution

Task: To find input that can steer program execution into specific branches.



# Program Model Checking

Task: To prove/refute conjectures about the values of program variables in order to characterize a finite-state abstraction.

# Static Program Analysis

Task: To check feasibility of certain program paths.

# Program Verification

Task: To prove verification conditions that arise from claims of functional correctness.

# SMT Solver Use

We've seen what SMT solvers are good for. How do you actually interact with them?



# The SMT-LIB Language

SMT solvers provide a [textual interface](#). Most solvers support a standard language, SMT-LIB.

SMT-LIB defines

- ▶ concrete syntax for input formulas, and
- ▶ a command-based scripting language.

Solver-specific syntax is often available to extend SMT-LIB, e.g., for data types.

# SMT-LIB: Example

```
; This example illustrates basic arithmetic and
; uninterpreted functions

(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)
(push)
(assert (= x y))
(check-sat)
(pop)
(exit)
```

ask z3

Is this formula satisfiable? Click 'ask z3'!

tutorial

home

video

# SMT-LIB: Example (Result)

ask z3

```
sat
(model
  (define-fun z () Int
    0)
  (define-fun y () Int
    (- 38))
  (define-fun x () Int
    0)
  (define-fun f ((x!1 Int)) Int
    (ite (= x!1 0) (- 1)
      (ite (= x!1 (- 38)) 1
        (- 1))))
  (define-fun g ((x!1 Int) (x!2 Int)) Int
    (ite (and (= x!1 0) (= x!2 0)) 0
      0))
)
unsat
```

# Inter-Process Communication

- ▶ File-based (the basic solution)
- ▶ Stream-based (when you need online functionality)
- ▶ Web interface (mostly for quick experiments)
- ▶ In-memory API (the tightly integrated approach)
- ▶ ...



# Algorithms

So far, we have considered SMT solvers as a black box.



This view is sufficient for many applications!

## SAT: DPLL

```
 $\vartheta := \emptyset$ ; // partial Boolean valuation
while(true) {
   $\vartheta := \vartheta \cup \text{propagate}(\varphi, \vartheta)$ ; // deduce consequences
  if( $[[\varphi]]_{\vartheta} == \text{true}$ ) {
    return SATISFIABLE;
  } else if( $[[\varphi]]_{\vartheta} == \text{false}$ ) {
     $\vartheta := \text{backtrack}(\varphi, \vartheta)$ ; // try a different branch
    if( $\vartheta == \emptyset$ ) { return UNSATISFIABLE; }
  } else {
     $\vartheta := \vartheta \cup \text{decide}(\varphi, \vartheta)$ ; // branch on unassigned variable
  }
}
```

# Interfacing Theory Solvers with SAT

$\Gamma$  := **abstraction function** that maps atomic formulas to Boolean variables;

$\varphi := \Gamma(\varphi)$ ;

while(true) {

$\vartheta :=$  **dpll**( $\varphi$ );

    if( $\vartheta ==$  UNSATISFIABLE) { return UNSATISFIABLE; }

$\Theta := \Gamma^{-1}(\vartheta)$ ;

    if(**T**( $\Theta$ ) == SATISFIABLE) { return SATISFIABLE; }

$\varphi := \varphi \wedge \neg\vartheta$ ; // *theory lemma*

}

## Combining Theory Solvers

Nelson-Oppen combination method: for disjoint, stably infinite theories it is sufficient to propagate **equalities between variables**.

Example:  $x \leq y \wedge y \leq x \wedge P(f(x) - f(y)) \wedge \neg P(0)$

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$$\text{Example: } x \leq y \wedge y \leq x \wedge P(\underbrace{f(x)}_{v_1} - \underbrace{f(y)}_{v_2}) \wedge \neg P(\underbrace{0}_{v_4})$$

$v_3$

Arithmetic	EUF
$x \leq y$	$v_1 = f(x)$
$y \leq x$	$v_2 = f(y)$
$v_3 = v_1 - v_2$	$P(v_3)$
$v_4 = 0$	$\neg P(v_4)$

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$v_4 = 0$	$\neg P(v_4)$
$x = y$	$v_1 = v_2$
$v_3 = v_4$	$\perp$

# The SMT Competition

Held annually since 2005 to spur adoption of the SMT-LIB format and to spark further advances in SMT.

Roughly similar to other competitions in automated reasoning, such as CASC and SAT.

# A Virtuous Circle

