

# Decision Procedures in Verification

Combinations of decision procedures (3)

28.01.2013

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# Until now:

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## Logical Theories: generalities

- Theory of Uninterpreted Function Symbols
- Decision procedures for numeric domains

Difference logic

Linear arithmetic: Fourier-Motzkin

- Combinations of decision procedures

Definitions

The Nelson/Oppen Procedure

DPLL(T) (idea)

# Motivation

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## Question:

how to check satisfiability (modulo theories) for sets of clauses?

# SAT Modulo Theories (SMT)

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## The “very eager” approach to SMT

### Method:

- translate problem into equisatisfiable propositional formula;
- use off-the-shelf SAT solver

- Why “eager”?

Search uses **all** theory information from the **beginning**

- Characteristics:

- + Can use best available SAT solver
- Sophisticated encodings are needed for each theory
- Sometimes translation and/or solving too slow

**Main Challenge** for alternative approaches is to combine:

- DPLL-based techniques for handling the boolean structure
- Efficient theory solvers for conjunctions of  $\mathcal{T}$ -literals

# SAT Modulo Theories (SMT)

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“Lazy” approaches to SMT: **Idea**

**Example:** consider  $\mathcal{T} = \text{UIF}$  and the following set of clauses:

$$\underbrace{f(g(a)) \not\approx f(c)}_{\neg P_1} \vee \underbrace{g(a) \approx d}_{P_2}, \quad \underbrace{g(a) \approx c}_{P_3}, \quad \underbrace{c \not\approx d}_{\neg P_4}$$

1. Send  $\{\neg P_1 \vee P_2, P_3, \neg P_4\}$  to SAT solver

SAT solver returns model  $[\neg P_1, P_3, \neg P_4]$

Theory solver says  $\neg P_1 \wedge P_3 \wedge \neg P_4$  is  $\mathcal{T}$ -inconsistent

2. Send  $\{\neg P_1 \vee P_2, P_3, \neg P_4, P_1 \vee \neg P_3 \vee P_4\}$  to SAT solver

SAT solver returns model  $[P_1, P_2, P_3, \neg P_4]$

Theory solver says  $P_1 \wedge P_2 \wedge P_3 \wedge \neg P_4$  is  $\mathcal{T}$ -inconsistent

3. Send  $\{\neg P_1 \vee P_2, P_3, \neg P_4, P_1 \vee \neg P_3 \vee P_4, \neg P_1 \vee \neg P_2 \vee \neg P_3 \vee P_4\}$  to SAT solver

SAT solver says UNSAT

# SAT Modulo Theories (SMT)

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## Optimized lazy approach

LA      • Check T-consistency only of full propositional models

OLA     • Check T-consistency of partial assignment while being built

LA      • Given a T-inconsistent assignment  $M$ , add  $\neg M$  as a clause

OLA     • Given a T-inconsistent assignment  $M$ , find an explanation  
(a small T-inconsistent subset of  $M$ ) and add it as a clause

LA      • Upon a T-inconsistency, add clause and restart

OLA     • Upon a T-inconsistency, do conflict analysis of the  
explanation and Backjump

# SAT Modulo Theories (SMT)

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## “Lazy” approaches to SMT

- Why “lazy”?

Theory information used only lazily, when checking  $\mathcal{T}$ -consistency of propositional models

- Characteristics:

- + Modular and flexible

- Theory information does not guide the search  
(only validates a posteriori)

**Tools:** CVC-Lite, ICS, MathSAT, TSAT+, Verifun, ...

# “Lazy” approaches to SMT

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Lazy theory learning:

$$M, L, M_1 \parallel F \Rightarrow \emptyset \parallel F, \neg L_1 \vee \dots \vee \neg L_n \vee \neg L \quad \text{if} \quad \left\{ \begin{array}{l} M, L, M_1 \models F \\ \{L_1, \dots, L_n\} \subseteq M \\ L_1 \wedge \dots \wedge L_n \wedge L \models_{\mathcal{T}} \perp \end{array} \right.$$

Lazy theory learning + no repetitions

$$M, L, M_1 \parallel F \Rightarrow \emptyset \parallel F, \neg L_1 \vee \dots \vee \neg L_n \vee \neg L \quad \text{if} \quad \left\{ \begin{array}{l} \{L_1, \dots, L_n\} \subseteq M \\ L_1 \wedge \dots \wedge L_n \wedge L \models_{\mathcal{T}} \perp \\ \neg L_1 \vee \dots \vee \neg L_n \vee \neg L \notin F \end{array} \right.$$



# DPLL(T) Rules

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## UnitPropagation

$M || F, C \vee L \Rightarrow M, L || F, C \vee L$       if  $M \models \neg C$ , and  $L$  undef. in  $M$

## Decide

$M || F \Rightarrow M, L^d || F$       if  $L$  occurs in  $F$ ,  $L$  undef. in  $M$

## Fail

$M || F, C \Rightarrow \text{Fail}$       if  $M \models \neg C$ , no backtrack possible

## Backjump

$M, L^d, N || F \Rightarrow M, L' || F$       if  $\left\{ \begin{array}{l} \text{there is some clause } C \vee L' \text{ s.t.:} \\ F \models C \vee L', M \models \neg C, \\ L' \text{ undefined in } M \\ L' \text{ or } \neg L' \text{ occurs in } F. \end{array} \right.$

## Restart/Learn

$M || F \Rightarrow \emptyset || F, F'$       if  $F \models F'$ ,  $F'$  obtained from  $M, F$

## TPropagation

$M || F \Rightarrow M, L || F$       if  $M \models_{\mathcal{T}} L$

# DPLL(T) Example

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Consider again same example with UIF:

$$\underbrace{f(g(a)) \not\approx f(c)}_{\neg P_1} \vee \underbrace{g(a) \approx d}_{P_2}, \quad \underbrace{g(a) \approx c}_{P_3}, \quad \underbrace{c \not\approx d}_{\neg P_4}$$

$$\emptyset \quad || \neg P_1 \vee P_2, P_3, \neg P_4 \Rightarrow (UnitPropagation)$$

$$P_3 \quad || \neg P_1 \vee P_2, P_3, \neg P_4 \Rightarrow (TPropagation)$$

$$P_3 P_1 \quad || \neg P_1 \vee P_2, P_3, \neg P_4 \Rightarrow (UnitPropagation)$$

$$P_3 P_1 P_2 \quad || \neg P_1 \vee P_2, P_3, \neg P_4 \Rightarrow (TPropagation)$$

$$P_3 P_1 P_2 P_4 \quad || \neg P_1 \vee P_2, P_3, \neg P_4 \Rightarrow fail$$

No search in this example

# Termination

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**Idea:**  $DPLL(T)$  terminates if no clause is learned infinitely many times, since only finitely many such new clauses (built over input literals) exist.

**Theorem.** There exists no infinite sequence of the form

$$\emptyset || F \Rightarrow S_1 \Rightarrow S_2 \dots$$

if no clause  $C$  is learned by **Reset & Learn/Lazy Theory Learning** infinitely many times along a sequence.

A similar termination result holds also for the  $DPLL(T)$  approach with **Theory Propagation**.

# Termination

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**Theorem.** There exist no infinite sequences of the form  $\emptyset || F \Rightarrow S_1 \Rightarrow S_2 \dots$

**Proof. (Idea)** We define a well-founded strict partial ordering  $\succ$  on states, and show that each rule application  $M || F \Rightarrow M' || F'$  is decreasing with respect to this ordering, i.e.,  $M || F \succ M' || F'$ .

Let  $M$  be of the form  $M_0, L_1, M_1, \dots, L_p, M_p$ , where  $L_1, \dots, L_p$  are all the decision literals of  $M$ . Similarly, let  $M'$  be  $M'_0, L'_1, M'_1, \dots, L'_{p'}, M'_{p'}$ .

Let  $N$  be the number of distinct atoms (propositional variables) in  $F$ .

(Note that  $p, p'$  and the length of  $M$  and  $M'$  are always smaller than or equal to  $N$ .)

# Termination

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**Theorem.** There exist no infinite sequences of the form  $\emptyset || F \Rightarrow S1 \Rightarrow \dots$

**Proof.** (continued)

Let  $m(M)$  be  $N - \text{length}(M)$  (nr. of literals missing in  $M$  for  $M$  to be total).

Define:  $M_0 L_1 M_1 \dots L_p M_p || F \succ M'_0 L'_1 M'_1 \dots L'_{p'} M'_{p'} || F'$  if

(i) there is some  $i$  with  $0 \leq i \leq p, p'$  such that

$m(M_0) = m(M'_0), \dots, m(M_{i-1}) = m(M'_{i-1}), m(M_i) > m(M'_i)$  or

(ii)  $m(M_0) = m(M'_0), \dots, m(M_p) = m(M'_{p'})$  and  $m(M) > m(M')$ .

Comparing the number of missing literals in sequences is a strict ordering (irreflexive and transitive) and it is well-founded, and hence this also holds for its lexicographic extension on tuples of sequences of bounded length.

**No learning/forgetting:** It is easy to see that all Basic DPLL rule applications are decreasing with respect to  $\succ$  if fail is added as an additional minimal element. (The rules UnitPropagate and Backjump decrease by case (i) of the definition and Decide decreases by case (ii).)

# Termination

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**Theorem.** There exist no infinite sequences of the form  $\emptyset || F \Rightarrow S1 \Rightarrow \dots$

**Note:** Combine learning with basic DPLL(T): no clause learned infinitely many times.

**Forget:** For this termination condition to be fulfilled, applying at least one rule of the Basic DPLL system between any two Learn applications does not suffice. It suffices if, in addition, no clause generated with Learning is ever forgotten.

# Soundness, Correctness, Termination

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**Lemma.** If  $\emptyset || F \Rightarrow^* M || F'$  then:

- (1) All atoms in  $M$  and all atoms in  $F'$  are atoms of  $F$ .
- (2)  $M$ : no literal more than once, no complementary literals
- (3)  $F'$  is logically equivalent to  $F$
- (4) if  $M = M_0 L_1 M_1 \dots L_n M_n$  where  $L_i$  all decision literals then  $F, L_1, \dots, L_i \models M_i$ .

**Lemma.** If  $\emptyset || F \Rightarrow^* M || F'$ , where  $M || F'$  is a final state wrt the Basic DPLL system and Lazy Theory Learning, then:

- (1) All literals of  $F'$  are defined in  $M$
- (2) There is no clause  $C$  in  $F'$  such that  $M \models \neg C$
- (3)  $M$  is a model of  $F$ .

# Soundness, Correctness, Termination

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**Lemma.** If  $\emptyset || F \Rightarrow^* M || F'$ , where  $M || F'$  is a final state wrt the Basic DPLL system and Lazy Theory Learning, then  $M$  is a  $\mathcal{T}$ -model of  $F$ .

**Theorem.** The Lazy Theory learning DPLL system provides a decision procedure for the satisfiability in  $\mathcal{T}$  of CNF formulae  $F$ , that is:

1.  $\emptyset || F \Rightarrow^* fail$  if, and only if,  $F$  is unsatisfiable in  $\mathcal{T}$ .
2.  $\emptyset || F \Rightarrow^* M || F'$ , where  $M || F'$  is a final state wrt the Basic DPLL system and Lazy Theory Learning, if, and only if,  $F$  is satisfiable in  $\mathcal{T}$ .

## Proof

(1) If  $\emptyset || F \Rightarrow^* fail$  then there exists state  $M || F'$  with  $\emptyset || F \Rightarrow^* M || F' \Rightarrow fail$ , there is no decision literal in  $M$  and  $M \models \neg C$  for some clause  $C$  in  $F$ . By the construction of  $M$ ,  $F \models M$ , so  $F \models \neg C$ . Thus  $F$  is unsatisfiable.

To prove the converse, if  $\emptyset || F \not\Rightarrow^* fail$  then by there must be a state  $M || F'$  such that  $\emptyset || F \Rightarrow^* M || F'$ . Then  $M \models F$ , so  $F$  is satisfiable.



# Soundness, Correctness, Termination

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**Lemma.** If  $\emptyset || F \Rightarrow^* M || F'$ , where  $M || F'$  is a final state wrt the Basic DPLL system and Lazy Theory Learning, then  $M$  is a  $\mathcal{T}$ -model of  $F$ .

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**Proof**

2. If  $\emptyset || F \Rightarrow^* M || F$  then  $F$  is satisfiable. Conversely, if  $\emptyset || F \not\Rightarrow^* M || F$  then  $\emptyset || F \Rightarrow^* fail$ , so  $F$  is unsatisfiable.

# Termination, Soundness and Completeness

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DPLL( $\mathcal{T}$ ) with (eager) theory propagation

**Lemma.** If  $\emptyset || F \Rightarrow M || F$  then  $M$  is  $\mathcal{T}$ -consistent.

**Proof.** This property is true initially, and all rules preserve it, by the fact that  $M \models_{\mathcal{T}} L$  if, and only if,  $M \cup \neg L$  is  $\mathcal{T}$ -inconsistent: the rules only add literals to  $M$  that are undefined in  $M$ , and **Theory Propagate** adds all literals  $L$  of  $F$  that are theory consequences of  $M$ , before any literal  $\neg L$  making it  $\mathcal{T}$ -inconsistent can be added to  $M$  by any of the other rules.

# Termination, Soundness and Completeness

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## DPLL( $\mathcal{T}$ ) with (eager) theory propagation

**Definition.** A DPLL( $\mathcal{T}$ ) procedure with Eager Theory Propagation for  $\mathcal{T}$  is any procedure taking an input CNF  $F$  and computing a sequence  $\emptyset || F \Rightarrow^* S$  where  $S$  is a final state wrt. **Theory Propagate** and the Basic DPLL system.

**Theorem** The DPLL system with eager theory propagation provides a decision procedure for the satisfiability in  $\mathcal{T}$  of CNF formulae  $F$ , that is:

1.  $\emptyset || F \Rightarrow^* \text{fail}$  if, and only if,  $F$  is unsatisfiable in  $\mathcal{T}$ .
2.  $\emptyset || F \Rightarrow^* M || F'$ , where  $M || F'$  is a final state wrt the Basic DPLL system and **Theory Propagate**, if, and only if,  $F$  is satisfiable in  $\mathcal{T}$ .
3. If  $\emptyset || F \Rightarrow M || F'$ , where  $M || F'$  is a final state wrt the Basic DPLL system and Theory Propagate, then  $M$  is a  $\mathcal{T}$ -model of  $F$ .

# Literature

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Full proofs and further details can be found in:

Robert Nieuwenhuis, Albert Oliveras and Cesare Tinelli:

“Solving SAT and SAT Modulo Theories: From an Abstract Davis-Putnam-Logemann-Loveland Procedure to DPLL(T)”

Journal of the ACM, Vol. 53, No. 6, November 2006, pp.937-977.

# SMT tools

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## SAT problems

Given: conjunction  $\phi$  of prop. clauses

Task: check if  $\phi$  satisfiable

Method: DPLL

- deterministic choices first
  - unit resolution
  - pure literal assignment
- case distinction (splitting)
- heuristics
  - selection criteria for splitting
  - backtracking
  - conflict-driven learning

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## SMT problems

Given: conjunction  $\phi$  of clauses

Task: check if  $\phi \models_{\mathcal{T}} \perp$

Method: DPLL( $\mathcal{T}$ )

- Boolean assignment found using DPLL
- ... and checked for  $\mathcal{T}$ -satisfiability
- the assignment can be partial and checked before splitting
- usual heuristics are used:
  - non-chronological backtracking
  - learning

# SMT tools

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## SAT problems

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## SMT problems

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- ... and checked for  $\mathcal{T}$ -satisfiability
- the assignment can be partial and checked before splitting
- usual heuristics are used:
  - non-chronological backtracking
  - learning

Systems implementing such specialized satisfiability problems: Yices, Barcelogic Tools, CVC lite, haRVey, Math-SAT,... are called (S)atisfiability (M)odulo (T)heory solvers.

# Satisfiability of formulae with quantifiers

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In many applications we are interested in testing the satisfiability of formulae containing (universally quantified) variables.

## Examples

- check satisfiability of formulae in the Bernays-Schönfinkel class
- check whether a set of (universally quantified) Horn clauses entails a ground clause
- check whether a property is consequence of a set of axioms

**Example 1:**  $f : \mathbb{Z} \rightarrow \mathbb{Z}$  is monotonely increasing  
and  $g : \mathbb{Z} \rightarrow \mathbb{Z}$  is defined by  $g(x) = f(x + x)$   
then  $g$  is also monotonely increasing

**Example 2:** If array  $a$  is increasingly sorted, and  
 $x$  is inserted before the first position  $i$  with  $a[i] > x$   
then the array remains increasingly sorted.



# A theory of arrays

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We consider the theory of arrays in a many-sorted setting.

## Syntax:

- Sorts: Elem (elements), Array (arrays) and Index (indices, here integers).
- Function symbols: read, write.

$$a(\text{read}) = \text{Array} \times \text{Index} \rightarrow \text{Element}$$

$$a(\text{write}) = \text{Array} \times \text{Index} \times \text{Element} \rightarrow \text{Array}$$

# Theories of arrays

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We consider the theory of arrays in a many-sorted setting.

**Theory of arrays**  $\mathcal{T}_{arrays}$ :

- $\mathcal{T}_i$  (theory of indices): Presburger arithmetic
- $\mathcal{T}_e$  (theory of elements): arbitrary
- Axioms for read, write

$$\begin{aligned} read(write(a, i, e), i) &\approx e \\ j \not\approx i \vee read(write(a, i, e), j) &= read(a, j). \end{aligned}$$

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$$\begin{aligned} read(write(a, i, e), i) &\approx e \\ j \not\approx i \vee read(write(a, i, e), j) &= read(a, j). \end{aligned}$$

**Fact:** Undecidable in general.

**Goal:** Identify a fragment of the theory of arrays which is decidable.

# A decidable fragment

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- **Index guard** a positive Boolean combination of atoms of the form  $t \leq u$  or  $t = u$  where  $t$  and  $u$  are either a variable or a ground term of sort Index

**Example:**  $(x \leq 3 \vee x \approx y) \wedge y \leq z$  is an index guard

$x \leq c - 1$  (where  $c$  is a constant) is an index guard

**Example:**  $x + 1 \leq y$ ,  $x + 3 \leq y - 1$ ,  $x + x \leq 2$  are not index guards.

- **Array property formula** [Bradley,Manna,Sipma'06]

$(\forall i)(\varphi_I(i) \rightarrow \varphi_V(i))$ , where:

$\varphi_I$ : index guard

$\varphi_V$ : formula in which any universally quantified  $i$  occurs in a direct array read; no nestings

**Example:**  $c \leq x \leq y \leq d \rightarrow a(x) \leq a(y)$  is an array property formula

**Example:**  $x < y \rightarrow a(x) < a(y)$  is not an array property formula

# Decision Procedure

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(Rules should be read from top to bottom)

**Step 1:** Put  $F$  in NNF.

**Step 2:** Apply the following rule exhaustively to remove writes:

$$\frac{F[\text{write}(a, i, v)]}{F[a'] \wedge a'[i] = v \wedge (\forall j. j \neq i \rightarrow a[j] = a'[j])} \quad \text{for fresh } a' \text{ (write)}$$

Given a formula  $F$  containing an occurrence of a write term  $\text{write}(a, i, v)$ , we can substitute every occurrence of  $\text{write}(a, i, v)$  with a fresh variable  $a'$  and explain the relationship between  $a'$  and  $a$ .

# Decision Procedure

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**Step 3** Apply the following rule exhaustively to remove existential quantification:

$$\frac{F[\exists i. G[i]]}{F[G[j]]} \text{ for fresh } j \text{ (exists)}$$

Existential quantification can arise during Step 1 if the given formula contains a negated array property.

# Decision Procedure

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**Steps 4-6** accomplish the reduction of universal quantification to finite conjunction.

The main idea is to select a set of symbolic index terms on which to instantiate all universal quantifiers.

# Theories of arrays

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**Step 4** From the output  $F3$  of **Step 3**, construct the index set  $\mathcal{I}$ :

$$\begin{aligned}\mathcal{I} = & \{\lambda\} \cup \\ & \{t \mid \cdot[t] \in F3 \text{ such that } t \text{ is not a universally quantified variable}\} \cup \\ & \{t \mid t \text{ occurs as an } \textit{evar} \text{ in the parsing of index guards}\}\end{aligned}$$

(*evar* is any constant or unquantified variable.)

This index set is the finite set of indices that need to be examined. It includes all terms  $t$  that occur in some  $\textit{read}(a, t)$  anywhere in  $F$  (unless it is a universally quantified variable) and all terms  $t$  that are compared to a universally quantified variable in some index guard.

$\lambda$  is a fresh constant that represents all other index positions that are not explicitly in  $\mathcal{I}$ .



# Theories of arrays

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**Step 5** Apply the following rule exhaustively to remove universal quantification:

$$\frac{H[\forall \bar{i}. F[\bar{i}] \rightarrow G[\bar{i}]]}{H \left[ \bigwedge_{\bar{i} \in \mathcal{I}^n} (F[\bar{i}] \rightarrow G[\bar{i}]) \right]} \quad (\text{forall})$$

where  $n$  is the size of the list of quantified variables  $\bar{i}$ .

This is the key step.

It replaces universal quantification with finite conjunction over the index set. The notation  $\bar{i} \in \mathcal{I}^n$  means that the variables  $\bar{i}$  range over all  $n$ -tuples of terms in  $\mathcal{I}$ .

# Theories of arrays

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**Step 6:** From the output  $F5$  of [Step 5](#), construct

$$F6 : \quad F5 \wedge \bigwedge_{i \in \mathcal{I} \setminus \{\lambda\}} \lambda \neq i$$

The new conjuncts assert that the variable  $\lambda$  introduced in [Step 4](#) is unique: it does not equal any other index mentioned in  $F5$ .

**Step 7:** Decide the TA-satisfiability of  $F6$  using the decision procedure for the quantifier free fragment.

# Example

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Consider the array property formula

$$F : \text{write}(a, l, v)[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\forall i. i \neq l \rightarrow a[i] = b[i])$$

It contains one array property,

$$\forall i. i \neq l \rightarrow a[i] = b[i]$$

$$\text{index guard: } i \neq l \quad \text{value constraint: } a[i] = b[i]$$

Step 1: The formula is already in NNF.

Step 2: We rewrite F as:

$$\begin{aligned} F2 : \quad & a'[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\forall i. i \neq l \rightarrow a[i] = b[i]) \\ & \wedge a'[l] = v \wedge (\forall j. j \neq l \rightarrow a[j] = a'[j]). \end{aligned}$$

# Example

---

Consider the array property formula

$$F : \text{write}(a, l, v)[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\forall i. i \neq l \rightarrow a[i] = b[i])$$

Step 2: We rewrite  $F$  as:

$$F2 : \quad a'[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\forall i. i \neq l \rightarrow a[i] = b[i]) \\ \wedge a'[l] = v \wedge (\forall j. j \neq l \rightarrow a[j] = a'[j]).$$

Step 3:  $F2$  does not contain any existential quantifiers  $\mapsto F3 = F2$ .

Step 4: The index set is

$$\mathcal{I} = \{\lambda\} \cup \{k\} \cup \{l\} = \{\lambda, k, l\}$$

# Example

---

Consider the array property formula

$$F : \text{write}(a, l, v)[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\forall i. i \neq l \rightarrow a[i] = b[i])$$

Step 3:

$$F3 : \quad a'[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\forall i. i \neq l \rightarrow a[i] = b[i]) \\ \wedge a'[l] = v \wedge (\forall j. j \neq l \rightarrow a[j] = a'[j]).$$

$$\text{Step 4: } \mathcal{I} = \{\lambda\} \cup \{k\} \cup \{l\} = \{\lambda, k, l\}$$

Step 5: we replace universal quantification as follows:

$$F5 : \quad a'[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge \bigwedge_{i \in \mathcal{I}} (i \neq l \rightarrow a[i] = b[i]) \\ \wedge a'[l] = v \wedge \bigwedge_{i \in \mathcal{I}} (i \neq l \rightarrow a[i] = a'[i]).$$

# Example

---

Consider the array property formula

$$F : \text{write}(a, l, v)[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\forall i. i \neq l \rightarrow a[i] = b[i])$$

$$\mathcal{I} = \{\lambda\} \cup \{k\} \cup \{l\} = \{\lambda, k, l\}$$

Step 5 (continued) Expanding produces:

$$\begin{aligned} F5' : \quad & a'[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\lambda \neq l \rightarrow a[\lambda] = b[\lambda]) \\ & \wedge (k \neq l \rightarrow a[k] = b[k]) \wedge (l \neq l \rightarrow a[l] = b[l]) \\ & \wedge a'[l] = v \wedge (\lambda \neq l \rightarrow a[\lambda] = a'[\lambda]) \wedge (k \neq l \rightarrow a[k] = a'[k]) \\ & \wedge (l \neq l \rightarrow a[l] = a'[l]). \end{aligned}$$

# Example

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Consider the array property formula

$$F : \text{write}(a, l, v)[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\forall i. i \neq l \rightarrow a[i] = b[i])$$

$$\mathcal{I} = \{\lambda\} \cup \{k\} \cup \{l\} = \{\lambda, k, l\}$$

Step 5 (continued): Simplifying produces

$$\begin{aligned} F''5 : \quad & a'[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\lambda \neq l \rightarrow a[\lambda] = b[\lambda]) \\ & \wedge (k \neq l \rightarrow a[k] = b[k]) \\ & \wedge a'[l] = v \wedge (\lambda \neq l \rightarrow a[\lambda] = a'[\lambda]) \\ & \wedge (k \neq l \rightarrow a[k] = a'[k]). \end{aligned}$$

# Example

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Consider the array property formula

$$F : \text{write}(a, l, v)[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\forall i. i \neq l \rightarrow a[i] = b[i])$$

Step 6 distinguishes  $\lambda$  from other members of  $l$ :

$$\begin{aligned} F6 : \quad & a'[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\lambda \neq l \rightarrow a[\lambda] = b[\lambda]) \\ & \wedge (k \neq l \rightarrow a[k] = b[k]) \\ & \wedge a'[l] = v \wedge (\lambda \neq l \rightarrow a[\lambda] = a'[\lambda]) \\ & \wedge (k \neq l \rightarrow a[k] = a'[k]) \wedge \lambda \neq k \wedge \lambda \neq l. \end{aligned}$$



# Example

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Consider the array property formula

$$F : \text{write}(a, l, v)[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \wedge (\forall i. i \neq l \rightarrow a[i] = b[i])$$

Step 6 Simplifying, we have

$$\begin{aligned} F'6 : \quad & a'[k] = b[k] \wedge b[k] \neq v \wedge a[k] = v \\ & \wedge a[\lambda] = b[\lambda] \wedge (k \neq l \rightarrow a[k] = b[k]) \\ & \wedge a'[l] = v \wedge a[\lambda] = a'[\lambda] \\ & \wedge (k \neq l \rightarrow a[k] = a'[k]) \wedge \lambda \neq k \wedge \lambda \neq l. \end{aligned}$$

There are two cases to consider.

- (1) If  $k=l$ , then  $a'[l]=v$  and  $a'[k]=b[k]$  imply  $b[k]=v$ , yet  $b[k] \neq v$ .
- (2) If  $k \neq l$ , then  $a[k]=v$  and  $a[k]=b[k]$  imply  $b[k]=v$ , but again  $b[k] \neq v$ .

Hence,  $F'6$  is TA-unsatisfiable, indicating that  $F$  is TA-unsatisfiable.