

Satisfiability Modulo Theories

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Disclaimer: The literature on SMT and its applications is vast. The bibliographic references provided here are just a sample. Apologies to all authors whose work is not cited.

Introduction

A (Very) Brief History of Automated Reasoning

Philosophers have long dreamed of machines that can reason. The pursuit of this dream has occupied some of the best minds and led both to great achievements and great disappointments.



~1700
Leibniz –
mechanized
human
reasoning



1928
Hilbert
Entscheidungs-
problem



1936
Church – lambda
calculus
Turing – reduction
halting problem



1954
Davis – decision
procedure for
Presburger
arithmetic

Automated Reasoning

Automated Reasoning: A Failure?

- At the turn of the century, automated reasoning was still considered by many to be **impractical for most real-world applications**
- Interesting problems appeared to be beyond the reach of automated methods because of **decidability and complexity barriers**
- The dream of *Hilbert*'s mechanized mathematics or *Leibniz*'s calculating machine was believed by many to be simply **unattainable**

The Satisfiability Revolution

Princeton, c. 2000

- *Chaff SAT solver*: orders of magnitude faster than previous SAT solvers
- *Important observation*: many real-world problems **do not exhibit worst-case theoretical performance**

Palo Alto, c. 2001

- **Idea**: combine fast new SAT solvers with decision procedures for decidable first-order theories
- *SVC*, *CVC* solvers (Stanford); *ICS*, *Yices* solvers (SRI)
- *Satisfiability Modulo Theories* (SMT) was born

SMT solvers

SMT solvers: *general-purpose* logic engines

- Given condition X , is it possible for Y to happen
- X and Y are expressed in a *rich logical language*
 - First-order logic
 - Domain-specific reasoning
 - arithmetic, arrays, bit-vectors, data types, etc.

SMT solvers are *changing the way people solve problems*

- Instead of building a *special-purpose* solver
- *Translate* into a logical formula and use an SMT solver
- Not only easier, *often better*

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A Security Example

Django

- Widely used open-source web development platform
- A security vulnerability in Django (CVE-2013-6044) was blamed on the following function¹

```
def is_safe_url(url, host=None):  
    """  
    Return “True” if the url is a safe redirection (i.e. it doesn't  
    point to a different host).  
  
    Always returns “False” on an empty url.  
    """  
    if not url:  
        return False  
    netloc = urllib_parse.urlparse(url)[1]  
    return not netloc or netloc == host
```

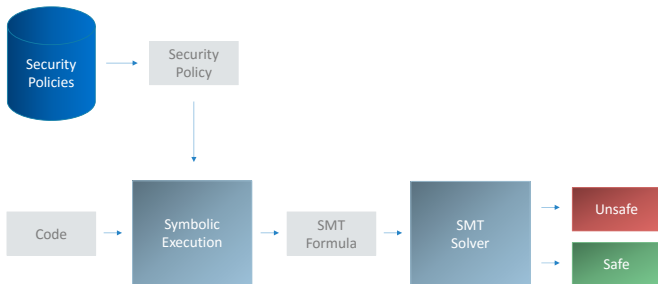
¹

<https://github.com/django/django/blob/09a5f5aabe27f63ec8d8982efa6cef9bf7b86022/django/utils/http.py#L252>

Using SMT To Find the Vulnerability

An approach for finding security vulnerabilities

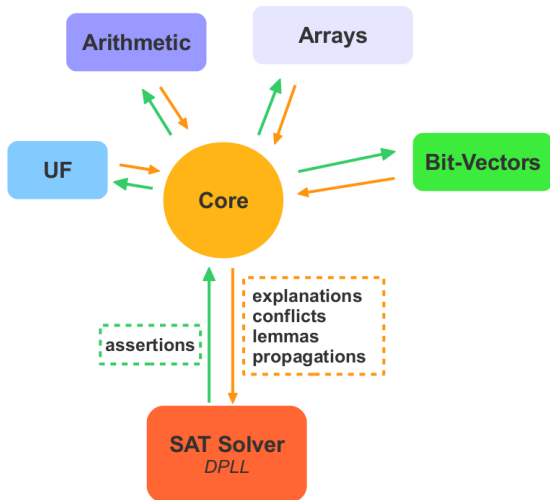
- *Symbolic execution*: generates a logical formula satisfiable iff code can violate security policy
- *SMT solver*: returns a solution or proves that none exists



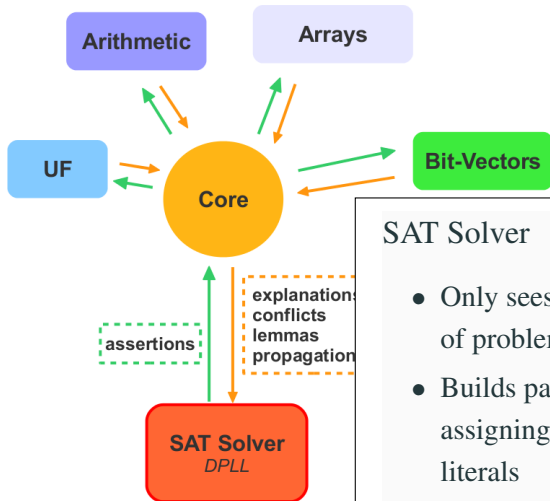
Using SMT To Find the Vulnerability

Demo: *Django XSS attack*

SMT Solvers



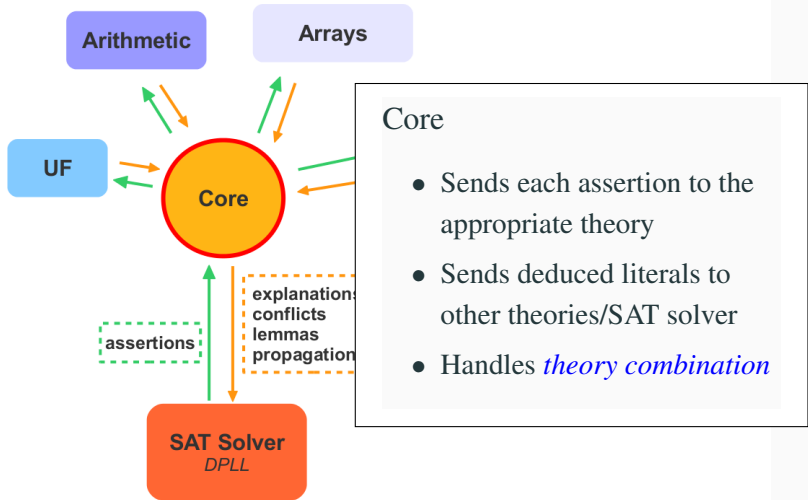
SMT Solvers



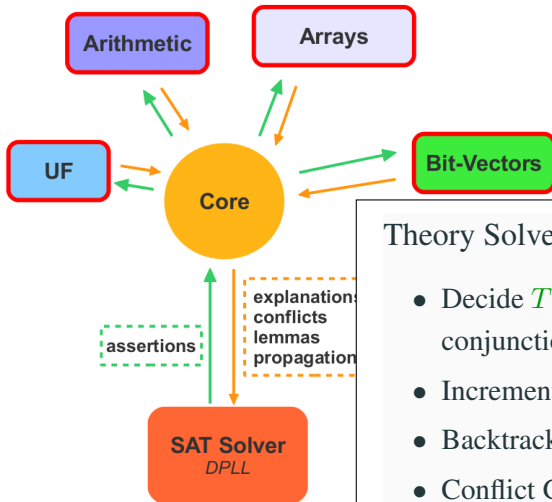
SAT Solver

- Only sees *Boolean skeleton* of problem
- Builds partial model by assigning truth values to literals
- Sends these literals to the core as *assertions*

SMT Solvers



SMT Solvers



Theory Solvers

- Decide T -satisfiability of a conjunction of theory literals
- Incremental
- Backtrackable
- Conflict Generation
- Theory Propagation

Theory Solvers

Theory Solvers

Given a theory T , a *Theory Solver* for T takes as input a set Φ of literals and determines whether Φ is T -satisfiable.

Φ is T -satisfiable iff there is some model M of T such that each formula in Φ holds in M .

Theories of Interest: UF

Equality (=) with Uninterpreted Functions [NO80, BD94, NO07]

Typically used to abstract unsupported constructs, e.g.:

- non-linear multiplication in arithmetic
- ALUs in circuits

Example: The formula

$$a * (|b| + c) = d \wedge b * (|a| + c) \neq d \wedge a = b$$

is unsatisfiable, but no arithmetic reasoning is needed

if we abstract it to

$$\text{mul}(a, \text{add}(\text{abs}(b), c)) = d \wedge \text{mul}(b, \text{add}(\text{abs}(a), c)) \neq d \wedge a = b$$

Theories of Interest: Arithmetic

Very useful, for obvious reasons

Restricted fragments (over the reals or the integers) support more efficient methods:

- Bounds: $x \bowtie k$ with $\bowtie \in \{<, >, \leq, \geq, =\}$ [BBC⁺05a]
- Difference logic: $x - y \bowtie k$, with $\bowtie \in \{<, >, \leq, \geq, =\}$ [NO05, WIGG05, CM06]
- UTVPI: $\pm x \pm y \bowtie k$, with $\bowtie \in \{<, >, \leq, \geq, =\}$ [LM05]
- Linear arithmetic, e.g: $2x - 3y + 4z \leq 5$ [DdM06]
- Non-linear arithmetic, e.g:
 $2xy + 4xz^2 - 5y \leq 10$ [BLNM⁺09, ZM10, JdM12]

Theories of Interest: Arrays

Used in software verification and hardware verification (for memories) [SBDL01, BNO⁺08a, dMB09]

Two interpreted function symbols `read` and `write`

Axiomatized by:

- $\forall a \forall i \forall v \text{ read}(\text{write}(a, i, v), i) = v$
- $\forall a \forall i \forall j \forall v \ i \neq j \rightarrow \text{read}(\text{write}(a, i, v), j) = \text{read}(a, j)$

Sometimes also with *extensionality*:

- $\forall a \forall b (\forall i \text{ read}(a, i) = \text{read}(b, i) \rightarrow a = b)$

Is the following set of literals satisfiable in this theory?

$$\text{write}(a, i, x) \neq b, \text{ read}(b, i) = y, \text{ read}(\text{write}(b, i, x), j) = y, a = b, i = j$$

Theories of Interest: Bitvectors

Useful both in hardware and software verification [BCF⁺07, BB09, HBJ⁺14]

Universe consists of (fixed-sized) vectors of bits

Different types of operations:

- *String-like*: concat, extract, ...
- *Logical*: bit-wise not, or, and, ...
- *Arithmetic*: add, subtract, multiply, ...
- *Comparison*: $<$, $>$, ...

Is this formula satisfiable over bitvectors of size 3?

$$a[1 : 0] \neq b[1 : 0] \wedge (a \mid b) = c \wedge c[0] = 0 \wedge a[1] + b[1] = 0$$

Implementing a Theory Solver: Difference Logic

We consider a simple example: *difference logic*.

In *difference logic*, we are interested in the satisfiability of a conjunction of arithmetic atoms.

Each atom is of the form $x - y \bowtie c$, where x and y are variables, c is a numeric constant, and $\bowtie \in \{=, <, \leq, >, \geq\}$.

The variables can range over either the *integers* (QF_IDL) or the *reals* (QF_RDL).

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- $x - y < c \implies x - y \leq c - 1$ (integers)

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- $x - y < c \implies x - y \leq c - 1$ (integers)
- $x - y < c \implies x - y \leq c - \delta$ (reals)

Difference Logic

Now we have a conjunction of literals, all of the form $x - y \leq c$.

From these literals, we form a weighted directed graph with a vertex for each variable.

For each literal $x - y \leq c$, there is an edge $x \xrightarrow{c} y$.

The set of literals is satisfiable iff there is no cycle for which the sum of the weights on the edges is negative.

There are a number of efficient algorithms for detecting negative cycles in graphs.

Difference Logic Example

$$x - y = 5 \wedge z - y \geq 2 \wedge z - x > 2 \wedge w - x = 2 \wedge z - w < 0$$

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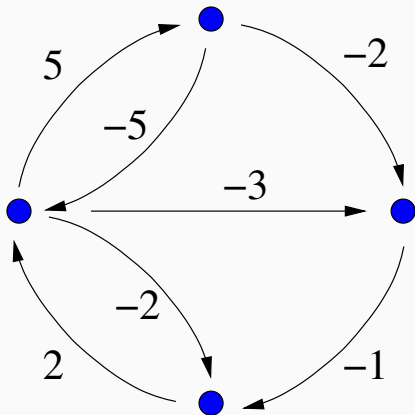
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Difference Logic Example

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$x - y = 5$	$x - y \leq 5 \wedge y - x \leq -5$
$z - y \geq 2$	$y - z \leq -2$
$z - x > 2 \quad \Rightarrow$	$x - z \leq -3$
$w - x = 2$	$w - x \leq 2 \wedge x - w \leq -2$
$z - w < 0$	$z - w \leq -1$

Difference Logic Example



DPLL(T): Combining T -Solvers with SAT

Satisfiability Modulo a Theory T

Def. A formula is *(un)satisfiable in* a theory T , or T -*(un)satisfiable*, if there is a (no) model of T that satisfies it

Note: The T -satisfiability of quantifier-free formulas is decidable iff the T -satisfiability of conjunctions/sets of literals is decidable

(Convert the formula in DNF and check if any of its disjuncts is T -sat)

Problem: In practice, dealing with Boolean combinations of literals is as hard as in propositional logic

Solution: Exploit propositional satisfiability technology

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Lifting SAT Technology to SMT

Two main approaches:

1. “Eager” [PRSS99, SSB02, SLB03, BGV01, BV02]

- translate into an equisatisfiable propositional formula
- feed it to any SAT solver

Notable systems: *UCLID*

2. “Lazy” [ACG00, dMR02, BDS02, ABC⁺02]

- abstract the input formula to a propositional one
- feed it to a (DPLL-based) SAT solver
- use a theory decision procedure to refine the formula and guide the SAT solver

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(Very) Lazy Approach for SMT – Example

$$g(a) = c \quad \wedge \quad f(g(a)) \neq f(c) \quad \vee \quad g(a) = d \quad \wedge \quad c \neq d$$

Theory *T*: Equality with Uninterpreted Functions

Simplest setting:

- Off-line SAT solver
- Non-incremental *theory solver* for conjunctions of equalities and disequalities
- Theory atoms (e.g., $g(a) = c$) abstracted to propositional atoms (e.g., 1)

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$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_2 \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_4$$

- Send $\{1, 2 \vee 3, 4\}$ to SAT solver.

- SAT solver returns model $\{1, \bar{2}, \bar{4}\}$.

Theory solver finds (concretization of) $\{1, \bar{2}, \bar{4}\}$ unsat.

- Send $\{1, 2, 3, \bar{4}\}$ to SAT solver.

- SAT solver returns model $\{3, \bar{2}, \bar{4}\}$.

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Lazy Approach – Enhancements

Several **enhancements** are possible to **increase efficiency**:

- Check T -satisfiability only of full propositional model
- Check T -satisfiability of partial assignment M as it grows
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- If M is T -unsatisfiable, identify a T -unsatisfiable subset M_0 of M and add $\neg M_0$ as a clause
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- If M is T -unsatisfiable, add $\neg M$ as a clause
- If M is T -unsatisfiable, identify a T -unsatisfiable subset M_0 of M and add $\neg M_0$ as a clause
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- If M is T -unsatisfiable, backtrack to some point where the assignment was still T -satisfiable

Lazy Approach – Enhancements

Several **enhancements** are possible to **increase efficiency**:

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Lazy Approach – Main Benefits

- Every tool **does** what it is **good** at:
 - **SAT solver** takes care of **Boolean information**
 - **Theory solver** takes care of **theory information**
- The theory solver works only with conjunctions of literals
- Modular approach:
 - SAT and theory solvers communicate via a simple API [GHN⁺04]
 - SMT for a new theory only requires new theory solver
 - An off-the-shelf SAT solver can be embedded in a lazy SMT system with few new lines of code (tens)

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An Abstract Framework for Lazy SMT

Several variants and enhancements of lazy SMT solvers exist

They can be modeled abstractly and declaratively as *transition systems*

A transition system is a binary relation over states, induced by a set of conditional transition rules

The framework can be first developed for SAT and then extended to lazy SMT [NOT06, KG07]

Advantages of Abstract Framework

An abstract framework helps one:

- skip over implementation details and unimportant control aspects
- reason formally about solvers for SAT and SMT
- model advanced features such as non-chronological backtracking, lemma learning, theory propagation, ...
- describe different strategies and prove their correctness
- compare different systems at a higher level
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The one described next is a re-elaboration of those in
[KBT⁺16, NOT06, KG07]

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The Original DPLL Procedure

- Modern SAT solvers are based on the **DPLL procedure** [DP60, DLL62]
- DPLL tries to **build** incrementally a **satisfying truth assignment** M for a CNF formula F
- M is grown by
 - **deducing** the truth value of a literal from M and F , or
 - **guessing** a truth value
- If a wrong guess for a literal leads to an inconsistency, the procedure **backtracks** and tries the opposite value

An Abstract Framework for DPLL

States:

fail or $\langle M, F \rangle$

where

- M is a sequence of literals and *decision points* • denoting a partial truth *assignment*
- F is a set of clauses denoting a CNF *formula*

Def. If $M = M_0 \bullet M_1 \bullet \dots \bullet M_n$ where each M_i contains no decision points

- M_i is *decision level* i of M
- $M^{[i]} \stackrel{\text{def}}{=} M_0 \bullet \dots \bullet M_i$

An Abstract Framework for DPLL

States:

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Initial state:

- $\langle (), F_0 \rangle$, where F_0 is to be checked for satisfiability

Expected final states:

- fail if F_0 is unsatisfiable
- $\langle M, G \rangle$ otherwise, where
 - G is equivalent to F_0 and
 - M satisfies G

Transition Rules: Notation

States treated like records:

- M denotes the truth assignment component of current state
- F denotes the formula component of current state

Transition rules in *guarded assignment form* [KG07]

$$\frac{p_1 \quad \cdots \quad p_n}{[M := e_1] \quad [F := e_2]}$$

updating M , F or both when premises p_1, \dots, p_n all hold

Transition Rules for the Original DPLL

Extending the assignment

$$\textbf{Propagate} \quad \frac{l_1 \vee \dots \vee l_n \vee l \in F \quad \bar{l}_1, \dots, \bar{l}_n \in M \quad l, \bar{l} \notin M}{M := M \, l}$$

Note: When convenient, treat M as a set

Note: Clauses are treated modulo ACI of \vee

$$\textbf{Decide} \quad \frac{l \in \text{Lit}(F) \quad l, \bar{l} \notin M}{M := M \bullet l}$$

Note: $\text{Lit}(F) \stackrel{\text{def}}{=} \{l \mid l \text{ literal of } F\} \cup \{\bar{l} \mid l \text{ literal of } F\}$

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Transition Rules for the Original DPLL

Repairing the assignment

$$\textbf{Fail} \frac{l_1 \vee \dots \vee l_n \in F \quad \bar{l}_1, \dots, \bar{l}_n \in M \quad \bullet \notin M}{\text{fail}}$$

Backtrack

$$\frac{l_1 \vee \dots \vee l_n \in F \quad \bar{l}_1, \dots, \bar{l}_n \in M \quad M = M \bullet l \ N \quad \bullet \notin N}{M := M \bar{l}}$$

Note: Last premise of **Backtrack** enforces chronological backtracking

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From DPLL to CDCL Solvers (1)

To model conflict-driven backjumping and learning, add to states a third component **C** whose value is either **no** or a *conflict clause*

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From DPLL to CDCL Solvers (2)

Replace **Backtrack** with

$$\text{Conflict} \quad \frac{C = \text{no} \quad l_1 \vee \dots \vee l_n \in F \quad \bar{l}_1, \dots, \bar{l}_n \in M}{C := l_1 \vee \dots \vee l_n}$$

$$\text{Explain} \quad \frac{C = l \vee D \quad l_1 \vee \dots \vee l_n \vee \bar{l} \in F \quad \bar{l}_1, \dots, \bar{l}_n \prec_M \bar{l}}{C := l_1 \vee \dots \vee l_n \vee D}$$

$$\text{Backjump} \quad \frac{C = l_1 \vee \dots \vee l_n \vee l \quad \text{lev } \bar{l}_1, \dots, \text{lev } \bar{l}_n \leq i < \text{lev } \bar{l}}{C := \text{no} \quad M := M^{[i]} l}$$

Maintain invariant: $F \models_p C$ and $M \models_p \neg C$ when $C \neq \text{no}$

Note: \models_p denotes propositional entailment

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Note: $l \prec_M l'$ if l occurs before l' in M
 $\text{lev } l = i$ iff l occurs in decision level i of M

Maintain invariant: $F \models_P C$ and $M \models_P \neg C$ when $C \neq \text{no}$

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From DPLL to CDCL Solvers (3)

Modify **Fail** to

$$\text{Fail} \frac{C \neq \text{no} \quad \bullet \notin M}{\text{fail}}$$

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Execution Example

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

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	F	no	
1	F	no	by Propagate
1 2	F	no	by Propagate
1 2 • 3	F	no	by Decide
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M	F	C	rule
	F	no	
1	F	no	by Propagate
1 2	F	no	by Propagate
1 2 • 3	F	no	by Decide
1 2 • 3 4	F	no	by Propagate
1 2 • 3 4 • 5	F	no	by Decide
1 2 • 3 4 • 5 $\bar{6}$	F	no	by Propagate
1 2 • 3 4 • 5 $\bar{6}$ 7	F	no	by Propagate
1 2 • 3 4 • 5 $\bar{6}$ 7	F	$\bar{2} \vee \bar{5} \vee 6 \vee \bar{7}$	by Conflict
1 2 • 3 4 • 5 $\bar{6}$ 7	F	$\bar{1} \vee \bar{2} \vee \bar{5} \vee 6$	by Explain with $\bar{1} \vee \bar{5} \vee 7$
1 2 • 3 4 • 5 $\bar{6}$ 7	F	$\bar{1} \vee \bar{2} \vee \bar{5}$	by Explain with $\bar{5} \vee \bar{6}$
1 2 $\bar{5}$	F	no	by Backjump
1 2 $\bar{5}$ • 3	F	no	by Decide
...			

Execution Example

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

M	F	C	rule
	F	no	
1	F	no	by Propagate
1 2	F	no	by Propagate
1 2 • 3	F	no	by Decide
1 2 • 3 4	F	no	by Propagate
1 2 • 3 4 • 5	F	no	by Decide
1 2 • 3 4 • 5 $\bar{6}$	F	no	by Propagate
1 2 • 3 4 • 5 $\bar{6}$ 7	F	no	by Propagate
1 2 • 3 4 • 5 $\bar{6}$ 7	F	$\bar{2} \vee \bar{5} \vee 6 \vee \bar{7}$	by Conflict
1 2 • 3 4 • 5 $\bar{6}$ 7	F	$\bar{1} \vee \bar{2} \vee \bar{5} \vee 6$	by Explain with $\bar{1} \vee \bar{5} \vee 7$
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1 2 $\bar{5}$	F	no	by Backjump
1 2 $\bar{5}$ • 3	F	no	by Decide
...			

Execution Example

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

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1 2 • 3 4 • 5 $\bar{6}$	F	no	by Propagate
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1 2 • 3 4 • 5 $\bar{6}$ 7	F	$\bar{1} \vee \bar{2} \vee \bar{5}$	by Explain with $\bar{5} \vee \bar{6}$
1 2 $\bar{5}$	F	no	by Backjump
1 2 $\bar{5}$ • 3	F	no	by Decide
...			

From DPLL to CDCL Solvers (4)

Also add

$$\textbf{Learn} \frac{F \models_p C \quad C \notin F}{F := F \cup \{C\}}$$

$$\textbf{Forget} \frac{C = \text{no} \quad F = G \cup \{C\} \quad G \models_p C}{F := G}$$

$$\textbf{Restart} \frac{}{M := M^{[0]} \quad C := \text{no}}$$

Note: Learn can be applied to **any** clause stored in **C** when **C** \neq no

Modeling Modern SAT Solvers

At the core, current CDCL SAT solvers are implementations of the transition system with rules

Propagate, Decide,

Conflict, Explain, Backjump,

Learn, Forget, Restart

Basic DPLL $\stackrel{\text{def}}{=}$

$\{ \text{Propagate, Decide, Conflict, Explain, Backjump} \}$

DPLL $\stackrel{\text{def}}{=}$ Basic DPLL + $\{ \text{Learn, Forget, Restart} \}$

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DPLL $\stackrel{\text{def}}{=}$ Basic DPLL + **{ Learn, Forget, Restart }**

The Basic DPLL System – Correctness

Some terminology:

Irreducible state: state for which no **Basic DPLL** rules apply

Execution: sequence of transitions allowed by the rules and starting with $M = \emptyset$ and $C = \text{no}$

Exhausted execution: execution ending in an irreducible state

Proposition (Soundness) For every exhausted execution starting with $F = F_0$ and ending with fail, the clause set F_0 is unsatisfiable.

Proposition (Completeness) For every exhausted execution starting with $F = F_0$ and ending with $C = \text{no}$, the clause set F_0 is satisfied by M .

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Proposition (Strong Termination) **Every** execution in Basic DPLL is finite.

Note: This is not so immediate, because of **Backjump**.

Proposition (Soundness) For every exhausted execution starting with $F = F_0$ and ending with **fail**, the clause set F_0 is unsatisfiable.

Proposition (Completeness) For every exhausted execution starting

The Basic DPLL System – Correctness

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Irreducible state: state for which no **Basic DPLL** rules apply

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Proposition (Strong Termination) **Every** execution in Basic DPLL is finite.

Lemma Every exhausted execution ends with either $C = \text{no}$ or **fail**.

Proposition (Soundness) For every exhausted execution starting with $F = F_0$ and ending with **fail**, the clause set F_0 is unsatisfiable.

Proposition (Completeness) For every exhausted execution starting

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The DPLL System – Strategies

- Applying
 - one Basic DPLL rule between each two **Learn** applications and
 - **Restart** less and less oftenensures termination

- A common basic strategy applies the rules with the following priorities:

1. If $n > 0$ conflicts have been found so far, increase n and apply **Restart**
2. If a clause is falsified by M , apply **Conflict**
3. If a literal l has $\text{Depth}(l) = \text{Depth}(M)$ and $\text{Depth}(l) > 0$, apply **Learn**
4. If $\text{Depth}(M) > 0$, apply **Learn**
5. If a literal l has $\text{Depth}(l) = 0$, apply **Learn**
6. If a literal l has $\text{Depth}(l) = 0$ and $\text{Depth}(M) = 0$, apply **Learn**
7. If a literal l has $\text{Depth}(l) = 0$ and $\text{Depth}(M) = 0$, apply **Learn**

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 5. Apply **Backjump**
 6. Apply **Propagate** to completion
 7. Apply **Decide**

From SAT to SMT

Same states and transitions but

- F contains **quantifier-free clauses** in some **theory T**
- M is a sequence of **theory literals** and decision points
- the DPLL system is augmented with rules

T -Conflict, T -Propagate, T -Explain

- maintains **invariant**: $F \models_T C$ and $M \models_p \neg C$ when $C \neq \text{no}$

Def. $F \models_T G$ iff every model of T that satisfies F satisfies G as well

SMT-level Rules

Fix a theory T

$$\textbf{\textit{T}-Conflict} \frac{C = \text{no} \quad l_1, \dots, l_n \in M \quad l_1, \dots, l_n \models_T \perp}{C := \bar{l}_1 \vee \dots \vee \bar{l}_n}$$

$$\textbf{\textit{T}-Propagate} \frac{l \in \text{Lit}(F) \quad M \models_T l \quad l, \bar{l} \notin M}{M := M \vee l}$$

$$\textbf{\textit{T}-Explain} \frac{C = l \vee D \quad \bar{l}_1, \dots, \bar{l}_n \models_T \bar{l} \quad \bar{l}_1, \dots, \bar{l}_n \prec_M \bar{l}}{C := l_1 \vee \dots \vee l_n \vee D}$$

Note: \perp = empty clause

Note: \models_T decided by theory solver

SMT-level Rules

Fix a theory T

$$\textbf{\textit{T}-Conflict} \frac{C = \text{no} \quad l_1, \dots, l_n \in M \quad l_1, \dots, l_n \models_T \perp}{C := \bar{l}_1 \vee \dots \vee \bar{l}_n}$$

$$\textbf{\textit{T}-Propagate} \frac{l \in \text{Lit}(\mathcal{F}) \quad M \models_T l \quad l, \bar{l} \notin M}{M := M \vee l}$$

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$$\textbf{\textit{T}-Propagate} \frac{l \in \text{Lit}(\mathbf{F}) \quad M \models_T l \quad l, \bar{l} \notin M}{M := M \, l}$$

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Note: \perp = empty clause

Note: \models_T decided by theory solver

Modeling the Very Lazy Theory Approach

T-Conflict is **enough** to model the **naive integration** of SAT solvers and theory solvers seen in the earlier UF example

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_2 \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_4$$

M	F	C	rule
	1, $\bar{2} \vee 3, \bar{4}$	no	
1 $\bar{4}$	1, $\bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}$	no	by Decide
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn
1 $\bar{4}$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Restart
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Propagate ⁺
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4, \bar{1} \vee 3 \vee 4$	$\bar{1} \vee 3 \vee 4$	by T-Conflict, Learn
fail			by Fail

Modeling the Very Lazy Theory Approach

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	1, $\bar{2} \vee 3, \bar{4}$	no	
1 $\bar{4}$	1, $\bar{2} \vee 3, \bar{4}$	no	by Propagate⁺
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}$	no	by Decide
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn
1 $\bar{4}$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Restart
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Propagate⁺
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4, \bar{1} \vee 3 \vee 4$	$\bar{1} \vee 3 \vee 4$	by T-Conflict, Learn
fail			by Fail

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	1, $\bar{2} \vee 3, \bar{4}$	no	
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	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Restart
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Propagate⁺
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4, \bar{1} \vee 3 \vee 4$	$\bar{1} \vee 3 \vee 4$	by T-Conflict, Learn
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$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Restart
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Propagate ⁺
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4, \bar{1} \vee 3 \vee 4$	$\bar{1} \vee 3 \vee 4$	by T-Conflict, Learn
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	1, $\bar{2} \vee 3, \bar{4}$	no	
	1, $\bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}$	no	by Decide
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn
	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Restart
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Propagate ⁺
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4, \bar{1} \vee 3 \vee 4$	$\bar{1} \vee 3 \vee 4$	by T-Conflict, Learn
fail			by Fail

Modeling the Very Lazy Theory Approach

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	1, $\bar{2} \vee 3, \bar{4}$	no	
	1, $\bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}$	no	by Decide
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn
	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Restart
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Propagate ⁺
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4, \bar{1} \vee \bar{3} \vee 4$	$\bar{1} \vee \bar{3} \vee 4$	by T-Conflict, Learn
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1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}$	no	by Decide
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn
	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Restart
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4$	no	by Propagate ⁺
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3, \bar{4}, \bar{1} \vee 2 \vee 4, \bar{1} \vee \bar{3} \vee 4$	$\bar{1} \vee \bar{3} \vee 4$	by T-Conflict, Learn
fail			by Fail

A Better Lazy Approach

The very lazy approach can be improved considerably with

- An *on-line* SAT engine,
which can accept new input clauses on the fly
- an *incremental and explicating* T -solver,
which can
 1. check the T -satisfiability of M as it is extended and
 2. identify a small T -unsatisfiable subset of M once it becomes T -unsatisfiable

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	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2$	by T-Conflict
$1 \bar{4} 2$	$1, \bar{2} \vee 3, \bar{4}$	no	by Backjump
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 3 \vee 4$	by T-Conflict
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$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2$	by T-Conflict
$\bar{1} \bar{4} 2$	$1, \bar{2} \vee 3, \bar{4}$	no	by Backjump
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee \bar{3} \vee 4$	by T-Conflict
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$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2$	by T-Conflict
$1 \bar{4} 2$	$1, \bar{2} \vee 3, \bar{4}$	no	by Backjump
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 3 \vee 4$	by T-Conflict
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fail			by Fail

Lazy Approach – Strategies

Ignoring **Restart** (for simplicity), a **common strategy** is to apply the rules using the following priorities:

1. If a clause is falsified by the current assignment **M**,
apply **Conflict**
2. If **M** is **T**-unsatisfiable, apply **T-Conflict**
3. Apply **Fail** or **Explain+Learn+Backjump** as appropriate
4. Apply **Propagate**
5. Apply **Decide**

Note: Depending on the cost of checking the **T**-satisfiability of **M**,
Step (2) can be applied with lower frequency or priority

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Theory Propagation

With ***T-Conflict*** as the **only theory rule**, the theory solver is used just to **validate** the choices of the SAT engine

With ***T-Propagate*** and ***T-Explain***, it can also be used to guide the engine's search [Tin02]

$$\textbf{\textit{T-Propagate}} \frac{l \in \text{Lit}(\mathcal{F}) \quad M \models_T l \quad l, \bar{l} \notin M}{M := M \ l}$$

$$\textbf{\textit{T-Explain}} \frac{C = l \vee D \quad \bar{l}_1, \dots, \bar{l}_n \models_T \bar{l} \quad \bar{l}_1, \dots, \bar{l}_n \prec_M \bar{l}}{C := l_1 \vee \dots \vee l_n \vee D}$$

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$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} 2$	$1, \bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1 \models_T 2$)
$1 \bar{4} 2 \bar{3}$	$1, \bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1, \bar{4} \models_T \bar{3}$)
$1 \bar{4} 2 \bar{3}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{2} \vee 3$	by Conflict
fail			by Fail

Note: *T*-propagation eliminates search altogether in this case
no applications of **Decide** are needed

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$\bar{1} \bar{4} 2 \bar{3}$	$1, \bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1, \bar{4} \models_T \bar{3}$)
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$\underline{1} \ \bar{4} \ 2 \ 3$	$1, \bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1, 4 \models_T 3$)
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Modeling Modern Lazy SMT Solvers

At the core, current lazy SMT solvers are implementations of the transition system with rules

(1) **Propagate, Decide, Conflict, Explain, Backjump, Fail**

(2) *T*-**Conflict, T-Propagate, T-Explain**

(3) **Learn, Forget, Restart**

Basic DPLL Modulo Theories $\stackrel{\text{def}}{=} (1) + (2)$

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Modeling Modern Lazy SMT Solvers

At the core, current lazy SMT solvers are implementations of the transition system with rules

(1) **Propagate, Decide, Conflict, Explain, Backjump, Fail**

(2) *T-Conflict, T-Propagate, T-Explain*

(3) **Learn, Forget, Restart**

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Correctness

Updated terminology:

Irreducible state: state to which no **Basic DPLL MT** rules apply

Execution: sequence of transitions allowed by the rules and starting with $M = \emptyset$ and $C = \text{no}$

Exhausted execution: execution ending in an irreducible state

Proposition (Soundness) For every exhausted execution starting with $F = F_0$ and ending with **fail**, the clause set F_0 is T -unsatisfiable.

Proposition (Completeness) For every exhausted execution starting with $F = F_0$ and ending with $C = \text{no}$, F_0 is T -satisfiable; specifically, M is T -satisfiable and $M \models_P F_0$.

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Proposition (Termination) Every execution in which

- (a) **Learn/Forget** are applied only **finitely many times** and
- (b) **Restart** is applied with **increased periodicity**

is finite.

Lemma Every exhausted execution ends with either $C = \text{no}$ or **fail**.

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DPLL(T) Architecture

The approach formalized so far can be implemented with a simple architecture named $\text{DPLL}(T)$ [GHN⁺04, NOT06]

$$\text{DPLL}(T) = \text{DPLL}(X) \text{ engine} + T\text{-solver}$$

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$$\text{DPLL}(T) = \text{DPLL}(X) \text{ engine} + T\text{-solver}$$

$\text{DPLL}(X)$:

- Very similar to a SAT solver, enumerates Boolean models
- Not allowed: pure literal, blocked literal detection, ...
- Required: incremental addition of clauses
- Desirable: partial model detection

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$$\text{DPLL}(T) = \text{DPLL}(X) \text{ engine} + T\text{-solver}$$

T -solver:

- Checks the T -satisfiability of conjunctions of literals
- Computes theory propagations
- Produces explanations of T -unsatisfiability/propagation
- Must be incremental and backtrackable

Reasoning by Cases in Theory Solvers

For certain theories, determining that a set M is T -unsatisfiable requires reasoning by cases.

Example: T = the theory of arrays.

$$M = \{ \underbrace{r(w(a, i, x), j) \neq x}_1, \underbrace{r(w(a, i, x), j) \neq r(a, j)}_2 \}$$

$i = j$) Then, $r(w(a, i, x), j) = x$. Contradiction with 1.

$i \neq j$) Then, $r(w(a, i, x), j) = r(a, j)$. Contradiction with 2.

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Case Splitting

A *complete* T -solver reasons by cases via (internal) case splitting and backtracking mechanisms

An alternative is to lift case splitting and backtracking from the T -solver to the SAT engine

Basic idea: encode case splits as sets of clauses and send them as needed to the SAT engine for it to split on them [BNOT06]

Possible benefits:

- All case-splitting is coordinated by the SAT engine
- Only have to implement case-splitting infrastructure in one place
- Can learn a wider class of lemmas

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Splitting on Demand

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Basic Scenario:

$$M = \{ \dots, s = \underbrace{r(w(a, i, t), j)}_{s'}, \dots \}$$

- Main SMT module: “Is M T -unsatisfiable?”
- T -solver: “I do not know yet, but it will help me if you consider these *theory lemmas*:

$$s = s' \wedge i = j \rightarrow s = t, \quad \neg s = s' \wedge i \neq j \rightarrow s = r(a, j)”$$

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Modeling Splitting on Demand

To model the generation of theory lemmas for case splits, add the rule

T -Learn

$$\frac{\models_T \exists \mathbf{v} (l_1 \vee \dots \vee l_n) \quad l_1, \dots, l_n \in L_S \quad \mathbf{v} \text{ vars not in } F}{F := F \cup \{l_1 \vee \dots \vee l_n\}}$$

where L_S is a finite set of literals dependent on the initial set of clauses (see [BNOT06] for a formal definition of L_S)

Note: For many theories with a theory solver, there exists an appropriate finite L_S for every input F

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Now we can relax the requirement on the theory solver:

When $M \models_p F$, it must *either*

- *determine whether $M \models_T \perp$ or*
- *generate a new clause by ***T-Learn*** containing at least one literal of L_S undefined in M*

The T -solver is required to determine whether $M \models_T \perp$ only if all literals in L_S are defined in M

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Example — Theory of Finite Sets

$$F : x = y \cup z \wedge y \neq \emptyset \vee x \neq z$$

M	F	rule
$x = y \cup z$	F	by Propagate ⁺
$x = y \cup z \bullet y = \emptyset$	F	by Decide
$x = y \cup z \bullet y = \emptyset \ x \neq z$	F	by Propagate
$x = y \cup z \bullet y = \emptyset \ x \neq z$	$F, (x = z \vee e \in x \vee e \in z),$ $(x = z \vee e \notin x \vee e \notin z)$	by T-Learn
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T-solver can make the following deductions at this point:

$$e \in x \cdots \Rightarrow e \in y \cup z \cdots \Rightarrow e \in y \cdots \Rightarrow e \in \emptyset \Rightarrow \perp$$

This enables an application of **T-Conflict** with clause

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$x = y \cup z \bullet y = \emptyset \ x \neq z$	$F, (x = z \vee e \in x \vee e \in z),$ $(x = z \vee e \notin x \vee e \notin z)$	by T-Learn
$x = y \cup z \bullet y = \emptyset \ x \neq z \bullet e \in x$	$F, (x = z \vee e \in x \vee e \in z),$ $(x = z \vee e \notin x \vee e \notin z)$	by Decide
$x = y \cup z \bullet y = \emptyset \ x \neq z \bullet e \in x \ e \notin z$	$F, (x = z \vee e \in x \vee e \in z),$ $(x = z \vee e \notin x \vee e \notin z)$	by Propagate

T -solver can make the following deductions at this point:

$$e \in x \cdots \Rightarrow e \in y \cup z \cdots \Rightarrow e \in y \cdots \Rightarrow e \in \emptyset \Rightarrow \perp$$

This enables an application of **T -Conflict** with clause

$$x \neq y \cup z \vee y \neq \emptyset \vee x = z \vee e \notin x \vee e \in z$$

Correctness Results

Correctness results can be extended to the new rule.

Soundness: The new ***T-Learn*** rule maintains satisfiability of the clause set.

Completeness: As long as the theory solver can decide $M \models_T \perp$ when all literals in L_S are determined, the system is still complete.

Termination: The system terminates under the same conditions as before. Roughly:

- Any lemma is (re)learned only finitely many times
- **Restart** is applied with increased periodicity

Combining Theories

Need for Combining Theories and Solvers

Recall: Many applications give rise to formulas like:

$$a \approx b + 2 \wedge A \approx \text{write}(B, a + 1, 4) \wedge \\ (\text{read}(A, b + 3) \approx 2 \vee f(a - 1) \neq f(b + 1))$$

Solving that formula requires reasoning over

- the theory of linear arithmetic (T_{LA})
- the theory of arrays (T_{A})
- the theory of uninterpreted functions (T_{UF})

Question: Given solvers for each theory, can we combine them modularly into one for $T_{\text{LA}} \cup T_{\text{A}} \cup T_{\text{UF}}$?

Under certain conditions, we can do it with the Nelson-Oppen combination method [NO79, Opp80]

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Motivating Example (Convex Case)

Consider the following set of literals over $T_{\text{LRA}} \cup T_{\text{UF}}$
(T_{LRA} , linear **real** arithmetic):

$$\begin{aligned}f(f(x) - f(y)) &= a \\f(0) &> a + 2 \\x &= y\end{aligned}$$

First step: *purify* literals so that each belongs to a single theory

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$$\begin{array}{lll}f(f(x) - f(y)) = a & \implies & f(e_1) = a \\& & e_1 = f(x) - f(y) \\& & \implies & f(e_1) = a \\& & & e_1 = e_2 - e_3 \\& & & e_2 = f(x) \\& & & e_3 = f(y)\end{array}$$

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$$\begin{array}{llll}f(0) > a + 2 & \implies & f(e_4) > a + 2 & \implies & f(e_4) = e_5 \\& & e_4 = 0 & & e_4 = 0 \\& & & & e_5 > a + 2\end{array}$$

Motivating Example (Convex Case)

Second step: exchange entailed *interface equalities*, equalities over shared constants $e_1, e_2, e_3, e_4, e_5, a$

L_1	L_2
$f(e_1) = a$	$e_2 - e_3 = e_1$
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$f(y) = e_3$	$e_5 > a + 2$
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$e_1 = e_4$	

$$L_1 \models_{\text{UF}} e_2 = e_3 \quad L_2 \models_{\text{LRA}} e_1 = e_4$$

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Third step: check for satisfiability locally

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Third step: check for satisfiability locally

Back to L_1

Back to L_2

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Third step: check for satisfiability locally

$$L_1 \models_{\text{UF}} L_2$$

$$L_2 \models_{\text{LRA}} L_1$$

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Third step: check for satisfiability locally

$$L_1 \not\models_{\text{UF}} \perp$$

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Report unsatisfiable

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$$L_1 \not\models_{\text{UF}} \perp$$

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Report **unsatisfiable**

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Third step: check for satisfiability locally

$$L_1 \not\models_{\text{UF}} \perp$$

$$L_2 \models_{\text{LRA}} \perp$$

Report **unsatisfiable**

Motivating Example (Non-convex Case)

Consider the following **unsatisfiable** set of literals over $T_{LIA} \cup T_{UF}$ (T_{LIA} , linear **integer** arithmetic):

$$\begin{aligned}1 &\leq x \leq 2 \\ f(1) &= a \\ f(2) &= f(1) + 3 \\ a &= b + 2\end{aligned}$$

First step: *purify* literals so that each belongs to a single theory

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$$\begin{aligned}f(1) = a &\implies f(e_1) = a \\ e_1 &= 1\end{aligned}$$

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First step: *purify* literals so that each belongs to a single theory

$$\begin{aligned}f(2) = f(1) + 3 &\implies e_2 = 2 \\ f(e_2) &= e_3 \\ f(e_1) &= e_4 \\ e_3 &= e_4 + 3\end{aligned}$$

Motivating Example (Non-convex Case)

Second step: exchange entailed *interface equalities* over shared constants $x, e_1, a, b, e_2, e_3, e_4$

L_1	L_2
$1 \leq x$	$f(e_1) = a$
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No more entailed equalities, but $L_1 \models_{\text{LIA}} x = e_1 \vee x = e_2$

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$e_3 = e_4 + 3$	
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Consider each case of $x = e_1 \vee x = e_2$ separately

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Case 1) $x = e_1$

Motivating Example (Non-convex Case)

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$L_2 \models_{\text{UF}} a = b$, which entails \perp when sent to L_1

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Case 2) $x = e_2$

Motivating Example (Non-convex Case)

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$L_2 \models_{\text{UF}} e_3 = b$, which entails \perp when sent to L_1

The Nelson-Oppen Method

- For $i = 1, 2$, let T_i be a first-order theory of *signature* Σ_i (set of function and predicate symbols in T_i other than $=$)
- Let $T = T_1 \cup T_2$
- Let \mathcal{C} be a finite set of *free* constants (i.e., not in $\Sigma_1 \cup \Sigma_2$)

We consider only input problems of the form

$$L_1 \cup L_2$$

where each L_i is a finite set of *ground* (i.e., variable-free) $(\Sigma_i \cup \mathcal{C})$ -literals

Note: Because of purification, there is no loss of generality in considering only ground $(\Sigma_i \cup \mathcal{C})$ -literals

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The Nelson-Oppen Method

Bare-bones, non-deterministic, non-incremental version

[Opp80, Rin96, TH96]:

Input: $L_1 \cup L_2$ with L_i finite set of ground $(\Sigma_i \cup \mathcal{C})$ -literals

Output: **sat** or **unsat**

1. Guess an *arrangement* A , i.e., a set of equalities and disequalities over \mathcal{C} such that

$$c = d \in A \text{ or } c \neq d \in A \text{ for all } c, d \in \mathcal{C}$$

2. If $L_i \cup A$ is T_i -unsatisfiable for $i = 1$ or $i = 2$, return **unsat**
3. Otherwise, return **sat**

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$$c = d \in A \text{ or } c \neq d \in A \text{ for all } c, d \in \mathcal{C}$$

2. If $L_i \cup A$ is T_i -unsatisfiable for $i = 1$ or $i = 2$, return **unsat**
3. Otherwise, return **sat**

The Nelson-Oppen Method

Bare-bones, non-deterministic, non-incremental version

[Opp80, Rin96, TH96]:

Input: $L_1 \cup L_2$ with L_i finite set of ground $(\Sigma_i \cup \mathcal{C})$ -literals

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Correctness of the NO Method

Proposition (Termination) The method is **terminating**.

(Trivially, because there is only a finite number of arrangements to guess)

Proposition (Soundness) If the method returns **unsat** for every arrangement, the input is $(T_1 \cup T_2)$ -unsatisfiable.

(Because satisfiability in $(T_1 \cup T_2)$ is always preserved)

Proposition (Completeness) If $\Sigma_1 \cap \Sigma_2 = \emptyset$ and T_1 and T_2 are stably infinite, when the method returns **sat** for some arrangement, the input is $(T_1 \cup T_2)$ -satisfiable.

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Stably Infinite Theories

Def. A theory T is *stably infinite* iff every quantifier-free T -satisfiable formula is satisfiable in an *infinite* model of T

Many interesting theories are stably infinite:

- Theories of an *infinite structure* (e.g., integer arithmetic)
- *Complete* theories with an infinite model (e.g., theory of dense linear orders, theory of lists)
- *Convex* theories (e.g., EUF, linear real arithmetic)

Def. A theory T is *convex* iff, for any set L of literals

$$L \models_T s_1 = t_1 \vee \cdots \vee s_n = t_n \implies L \models_T s_i = t_i \text{ for some } i$$

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The Nelson-Oppen method has been extended to some classes of non-stably infinite theories [TZ05, RRZ05, JB10]

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SMT Solving with *Multiple* Theories

Let T_1, \dots, T_n be theories with respective solvers S_1, \dots, S_n

How can we integrate all of them **cooperatively** into a single SMT solver for $T = T_1 \cup \dots \cup T_n$?

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Quick Solution:

1. Combine S_1, \dots, S_n with Nelson-Oppen into a theory solver for T
2. Build a DPLL(T) solver as usual

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Better Solution [Bar02, BBC⁺05b, BNOT06]:

1. Extend $\text{DPLL}(T)$ to $\text{DPLL}(T_1, \dots, T_n)$
2. **Lift Nelson-Oppen to the $\text{DPLL}(X_1, \dots, X_n)$ level**
3. Build a $\text{DPLL}(T_1, \dots, T_n)$ solver

Modeling DPLL(T_1, \dots, T_n) Abstractly

- Let $n = 2$, for simplicity
- Let T_i be of signature Σ_i for $i = 1, 2$, with $\Sigma_1 \cap \Sigma_2 = \emptyset$
- Let \mathcal{C} be a set of **free** constants
- Assume wlog that each input literal has signature $(\Sigma_1 \cup \mathcal{C})$ or $(\Sigma_2 \cup \mathcal{C})$ (**no mixed** literals)
- Let $M|_i \stackrel{\text{def}}{=} \{(\Sigma_i \cup \mathcal{C})\text{-literals of } M \text{ and their complement}\}$
- Let $I(M) \stackrel{\text{def}}{=} \{c = d \mid c, d \text{ occur in } \mathcal{C}, M|_1 \text{ and } M|_2\} \cup \{c \neq d \mid c, d \text{ occur in } \mathcal{C}, M|_1 \text{ and } M|_2\}$
(*interface literals*)

Abstract DPLL Modulo Multiple Theories

Propagate, Conflict, Explain, Backjump, Fail (unchanged)

$$\text{Decide} \frac{l \in \text{Lit}(F) \cup I(M) \quad l, \bar{l} \notin M}{M := M \bullet l}$$

Only change: decide on interface equalities as well

$$T\text{-Propagate} \frac{l \in \text{Lit}(F) \cup I(M) \quad i \in \{1, 2\} \quad M \models_{T_i} l \quad l, \bar{l} \notin M}{M := M \bullet l}$$

Only change: propagate interface equalities as well, but reason locally in each T_i

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Abstract DPLL Modulo Multiple Theories

T -Conflict

$$C = \text{no} \quad l_1, \dots, l_n \in M \quad l_1, \dots, l_n \models_{T_i} \perp \quad i \in \{1, 2\}$$

$$C := \bar{l}_1 \vee \dots \vee \bar{l}_n$$

T -Explain

$$C = l \vee D \quad \bar{l}_1, \dots, \bar{l}_n \models_{T_i} \bar{l} \quad i \in \{1, 2\} \quad \bar{l}_1, \dots, \bar{l}_n \prec_M \bar{l}$$

$$C := l_1 \vee \dots \vee l_n \vee D$$

Only change: reason locally in each T_i

I-Learn

$$\models_{T_i} l_1 \vee \dots \vee l_n \quad l_1, \dots, l_n \in M|_i \cup I(M) \quad i \in \{1, 2\}$$

$$F := F \cup \{l_1 \vee \dots \vee l_n\}$$

New rule: for entailed disjunctions of interface literals

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New rule: for entailed disjunctions of interface literals

Example — Convex Theories

$$\begin{array}{c}
 F := \underbrace{f(e_1) = a}_{5} \wedge \underbrace{f(x) = e_2}_{6} \wedge \underbrace{f(y) = e_3}_{7} \wedge \underbrace{f(e_4) = e_5}_{3} \wedge \underbrace{x = y}_{4} \wedge \\
 \underbrace{e_2 - e_3 = e_1}_{5} \wedge \underbrace{e_4 = 0}_{6} \wedge \underbrace{e_5 > a + 2}_{7} \\
 \underbrace{e_2 = e_3}_{8} \quad \underbrace{e_1 = e_4}_{9} \quad \underbrace{a = e_5}_{10}
 \end{array}$$

	M	F	C	rule
		F	no	
0 1 2 3 4 5 6 7		F	no	by Propagate ⁺
0 1 2 3 4 5 6 7 8		F	no	by T-Propagate (1, 2, 4 \models_{UF} 8)
0 1 2 3 4 5 6 7 8 9		F	no	by T-Propagate (5, 6, 8 \models_{LRA} 9)
0 1 2 3 4 5 6 7 8 9 10		F	no	by T-Propagate (0, 3, 9 \models_{UF} 10)
0 1 2 3 4 5 6 7 8 9 10		F	$\bar{7} \vee \bar{10}$	by T-Conflict (7, 10 $\models_{LRA} \perp$)
fail				by Fail

Example — Convex Theories

$$F := \underbrace{f(e_1) = a}_{\substack{0 \\ e_2 - e_3 = e_1 \\ 5}} \wedge \underbrace{f(x) = e_2}_{\substack{1 \\ e_4 = 0 \\ 6}} \wedge \underbrace{f(y) = e_3}_{\substack{2 \\ e_5 > a + 2 \\ 7}} \wedge \underbrace{f(e_4) = e_5}_{\substack{3 \\ 4}} \wedge \underbrace{x = y}_{\substack{4 \\ 5}} \wedge$$

$$\underbrace{e_2 = e_3}_{\substack{8}} \quad \underbrace{e_1 = e_4}_{\substack{9}} \quad \underbrace{a = e_5}_{\substack{10}}$$

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fail				by Fail

Example — Non-convex Theories

$$F := \underbrace{f(e_1) = a}_{4} \wedge \underbrace{1 \leq x}_{5} \wedge \underbrace{x \leq 2}_{6} \wedge \underbrace{f(x) = b}_{1} \wedge \underbrace{e_1 = 1}_{6} \wedge \underbrace{f(e_2) = e_3}_{2} \wedge \underbrace{a = b + 2}_{7} \wedge \underbrace{f(e_1) = e_4}_{3} \wedge \underbrace{e_2 = 2}_{8} \wedge \underbrace{e_3 = e_4 + 3}_{9}$$

$$\underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}$$

M	F	C	rule
	F	no	
0 ... 9	F	no	by Propagate ⁺
0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, 4 \vee 5 \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} 4 \vee 5 \vee 11 \vee 12$)
0 ... 9 10 • 11	$F, 4 \vee 5 \vee 11 \vee 12$	no	by Decide
0 ... 9 10 • 11 13	$F, 4 \vee 5 \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 • 11 13	$F, 4 \vee 5 \vee 11 \vee 12$	$7 \vee 13$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, 4 \vee 5 \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, 4 \vee 5 \vee 11 \vee 12$	no	by T-Propagate (0, 1, 13 \models_{UF} 11)
0 ... 9 10 13 11 12	$F, 4 \vee 5 \vee 11 \vee 12$	no	by Propagate (exercise)
...	by Fail
fail	by Fail

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0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, 4 \vee 5 \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} 4 \vee 5 \vee 11 \vee 12$)
0 ... 9 10 * 11	$F, 4 \vee 5 \vee 11 \vee 12$	no	by Decide
0 ... 9 10 * 11 13	$F, 4 \vee 5 \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 * 11 13	$F, 4 \vee 5 \vee 11 \vee 12$	$7 \vee 13$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, 4 \vee 5 \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, 4 \vee 5 \vee 11 \vee 12$	no	by T-Propagate (0, 1, 13 \models_{UF} 11)
0 ... 9 10 13 11 12	$F, 4 \vee 5 \vee 11 \vee 12$	no	by Propagate (exercise)
...	by Fail
fail	by Fail

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{4} \wedge \underbrace{1 \leq x}_{5} \wedge \underbrace{f(x) = b}_{6} \wedge \underbrace{x \leq 2}_{5} \wedge \underbrace{e_1 = 1}_{6} \wedge \underbrace{f(e_2) = e_3}_{7} \wedge \underbrace{a = b + 2}_{7} \wedge \underbrace{f(e_1) = e_4}_{8} \wedge \underbrace{e_2 = 2}_{8} \wedge \underbrace{e_3 = e_4 + 3}_{9} \\
 & \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{aligned}$$

M	F	C	rule
	F	no	
$0 \dots 9$	F	no	by Propagate ⁺
$0 \dots 9 \ 10$	F	no	by T -Propagate ($0, 3 \models_{UF} 10$)
$0 \dots 9 \ 10$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I -Learn ($\models_{LIA} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
$0 \dots 9 \ 10 \bullet 11$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T -Propagate ($0, 1, 11 \models_{UF} 13$)
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	$\bar{7} \vee \bar{13}$	by T -Conflict ($7, 13 \models_{UF} \perp$)
$0 \dots 9 \ 10 \ 13$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Backjump
$0 \dots 9 \ 10 \ 13 \ 11$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T -Propagate ($0, 1, \bar{13} \models_{UF} \bar{11}$)
$0 \dots 9 \ 10 \ 13 \ 11 \ 12$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Propagate (exercise)
...	by Fail
fail	by Fail

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{4} \wedge \underbrace{1 \leq x}_{5} \wedge \underbrace{f(x) = b}_{1} \wedge \underbrace{x \leq 2}_{5} \wedge \underbrace{e_1 = 1}_{6} \wedge \underbrace{f(e_2) = e_3}_{2} \wedge \underbrace{a = b + 2}_{7} \wedge \underbrace{f(e_1) = e_4}_{3} \wedge \underbrace{e_2 = 2}_{8} \wedge \underbrace{e_3 = e_4 + 3}_{9} \\
 & \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{aligned}$$

M	F	C	rule
	F	no	
0 ... 9	F	no	by Propagate ⁺
0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
0 ... 9 10 • 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	$\bar{7} \vee \bar{13}$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, $\bar{13} \models_{UF} 11$)
0 ... 9 10 13 11 12	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Propagate (exercise)
...	by Fail
fail	

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{\underbrace{1 \leq x}_4} \wedge \underbrace{x \leq 2}_{\underbrace{x \leq 2}_5} \wedge \underbrace{f(x) = b}_{\underbrace{e_1 = 1}_6} \wedge \underbrace{f(e_2) = e_3}_{\underbrace{a = b + 2}_7} \wedge \underbrace{f(e_1) = e_4}_{\underbrace{e_2 = 2}_8} \wedge \underbrace{e_3 = e_4 + 3}_{\underbrace{e_3 = e_4 + 3}_9} \\
 & \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{aligned}$$

M	F	C	rule
	F	no	
0 ... 9	F	no	by Propagate ⁺
0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
0 ... 9 10 • 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	$\bar{7} \vee \bar{13}$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, $\bar{13} \models_{UF} 11$)
0 ... 9 10 13 11 12	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Propagate
...	...	(exercise)	
fail	by Fail

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{\underbrace{1 \leq x}_4} \wedge \underbrace{f(x) = b}_{\underbrace{x \leq 2}_5} \wedge \underbrace{e_1 = 1}_{\underbrace{e_1 = 1}_6} \wedge \underbrace{f(e_2) = e_3}_{\underbrace{a = b + 2}_7} \wedge \underbrace{f(e_1) = e_4}_{\underbrace{e_2 = 2}_8} \wedge \underbrace{e_3 = e_4 + 3}_{\underbrace{e_3 = e_4 + 3}_9} \\
 & \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{aligned}$$

	M	F	C	rule
		F	no	
	0 ... 9	F	no	by Propagate ⁺
	0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
	0 ... 9 10	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
	0 ... 9 10 • 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
	0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
	0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	$\bar{7} \vee \bar{13}$	by T-Conflict (7, 13 $\models_{UF} \perp$)
	0 ... 9 10 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Backjump
	0 ... 9 10 13 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, $\bar{13} \models_{UF} 11$)
	0 ... 9 10 13 11 12	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Propagate
	(exercise)	
	fail	by Fail

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{\substack{0 \\ 1 \leq x \\ 4}} \wedge \underbrace{f(x) = b}_{\substack{1 \\ x \leq 2 \\ 5}} \wedge \underbrace{f(e_2) = e_3}_{\substack{2 \\ e_1 = 1 \\ 6}} \wedge \underbrace{f(e_1) = e_4}_{\substack{3 \\ a = b + 2 \\ 7}} \wedge \underbrace{e_2 = 2}_{\substack{8 \\ e_2 = 2 \\ 8}} \wedge \underbrace{e_3 = e_4 + 3}_{\substack{9 \\ e_3 = e_4 + 3 \\ 9}} \\
 & \underbrace{a = e_4}_{\substack{10 \\ a = e_4 \\ 10}} \quad \underbrace{x = e_1}_{\substack{11 \\ x = e_1 \\ 11}} \quad \underbrace{x = e_2}_{\substack{12 \\ x = e_2 \\ 12}} \quad \underbrace{a = b}_{\substack{13 \\ a = b \\ 13}}
 \end{aligned}$$

M	F	C	rule
	F	no	
0 ... 9	F	no	by Propagate ⁺
0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
0 ... 9 10 • 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	$\bar{7} \vee \bar{13}$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, $\bar{13} \models_{UF}$ 11)
0 ... 9 10 13 11 12	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Propagate
...	(exercise)
fail	by Fail

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{4} \wedge \underbrace{1 \leq x}_{5} \wedge \underbrace{x \leq 2}_{5} \wedge \underbrace{f(x) = b}_{1} \wedge \underbrace{e_1 = 1}_{6} \wedge \underbrace{f(e_2) = e_3}_{2} \wedge \underbrace{a = b + 2}_{7} \wedge \underbrace{f(e_1) = e_4}_{3} \wedge \underbrace{e_2 = 2}_{8} \wedge \underbrace{e_3 = e_4 + 3}_{9} \\
 & \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{aligned}$$

M	F	C	rule
	F	no	
0 ... 9	F	no	by Propagate ⁺
0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
0 ... 9 10 • 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	$\bar{7} \vee \bar{13}$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, $\bar{13} \models_{UF} \bar{11}$)
0 ... 9 10 13 11 12	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Propagate
...	(exercise)
fail	by Fail

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{\underbrace{1 \leq x}_4} \wedge \underbrace{x \leq 2}_{\underbrace{x \leq 2}_5} \wedge \underbrace{e_1 = 1}_{\underbrace{e_1 = 1}_6} \wedge \underbrace{f(e_2) = e_3}_{\underbrace{a = b + 2}_7} \wedge \underbrace{f(e_1) = e_4}_{\underbrace{e_2 = 2}_8} \wedge \underbrace{e_3 = e_4 + 3}_{\underbrace{e_3 = e_4 + 3}_9} \\
 & \underbrace{a = e_4}_{\underbrace{a = e_4}_{10}} \quad \underbrace{x = e_1}_{\underbrace{x = e_1}_{11}} \quad \underbrace{x = e_2}_{\underbrace{x = e_2}_{12}} \quad \underbrace{a = b}_{\underbrace{a = b}_{13}}
 \end{aligned}$$

M	F	C	rule
	F	no	
0 ... 9	F	no	by Propagate ⁺
0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
0 ... 9 10 • 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 • 11 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	$\bar{7} \vee \bar{13}$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, $\bar{13} \models_{UF} \bar{11}$)
0 ... 9 10 13 11 12	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Propagate
...	(exercise)
fail	by Fail

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{4} \wedge \underbrace{1 \leq x}_{5} \wedge \underbrace{x \leq 2}_{5} \wedge \underbrace{f(x) = b}_{1} \wedge \underbrace{e_1 = 1}_{6} \wedge \underbrace{f(e_2) = e_3}_{2} \wedge \underbrace{a = b + 2}_{7} \wedge \underbrace{f(e_1) = e_4}_{3} \wedge \underbrace{e_2 = 2}_{8} \wedge \underbrace{e_3 = e_4 + 3}_{9} \\
 & \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{aligned}$$

M	F	C	rule
	F	no	
0 ... 9	F	no	by Propagate ⁺
0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \overline{4} \vee \overline{5} \vee 11 \vee 12$)
0 ... 9 10 • 11	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Decide
0 ... 9 10 • 11 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 • 11 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	$\overline{7} \vee \overline{13}$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, $\overline{13} \models_{UF} \overline{11}$)
0 ... 9 10 13 11 12	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Propagate (exercise)
...	by Fail
fail	by Fail

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{\underbrace{1 \leq x}_4} \wedge \underbrace{x \leq 2}_{\underbrace{x \leq 2}_5} \wedge \underbrace{f(x) = b}_{\underbrace{e_1 = 1}_6} \wedge \underbrace{f(e_2) = e_3}_{\underbrace{a = b + 2}_7} \wedge \underbrace{f(e_1) = e_4}_{\underbrace{e_2 = 2}_8} \wedge \underbrace{e_3 = e_4 + 3}_{\underbrace{}_9} \\
 & \underbrace{a = e_4}_{\underbrace{}_{10}} \quad \underbrace{x = e_1}_{\underbrace{}_{11}} \quad \underbrace{x = e_2}_{\underbrace{}_{12}} \quad \underbrace{a = b}_{\underbrace{}_{13}}
 \end{aligned}$$

M	F	C	rule
	F	no	
0 ... 9	F	no	by Propagate ⁺
0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \overline{4} \vee \overline{5} \vee 11 \vee 12$)
0 ... 9 10 • 11	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Decide
0 ... 9 10 • 11 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 • 11 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	$\overline{7} \vee \overline{13}$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, $\overline{13} \models_{UF} \overline{11}$)
0 ... 9 10 13 11 12	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Propagate
...	...	(exercise)	
fail	by Fail

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{4} \wedge \underbrace{1 \leq x}_{5} \wedge \underbrace{f(x) = b}_{6} \wedge \underbrace{x \leq 2}_{5} \wedge \underbrace{e_1 = 1}_{6} \wedge \underbrace{f(e_2) = e_3}_{7} \wedge \underbrace{a = b + 2}_{7} \wedge \underbrace{f(e_1) = e_4}_{8} \wedge \underbrace{e_2 = 2}_{8} \wedge \underbrace{e_3 = e_4 + 3}_{9} \\
 & \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{aligned}$$

M	F	C	rule
	F	no	
0 ... 9	F	no	by Propagate ⁺
0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \overline{4} \vee \overline{5} \vee 11 \vee 12$)
0 ... 9 10 • 11	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Decide
0 ... 9 10 • 11 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 • 11 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	$\overline{7} \vee \overline{13}$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 13 \models_{UF} 11)
0 ... 9 10 13 11 12	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Propagate
...	...	(exercise)	
fail	by Fail

Example — Non-convex Theories

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{4} \wedge \underbrace{1 \leq x}_{5} \wedge \underbrace{f(x) = b}_{6} \wedge \underbrace{x \leq 2}_{5} \wedge \underbrace{e_1 = 1}_{6} \wedge \underbrace{f(e_2) = e_3}_{7} \wedge \underbrace{a = b + 2}_{7} \wedge \underbrace{f(e_1) = e_4}_{8} \wedge \underbrace{e_2 = 2}_{8} \wedge \underbrace{e_3 = e_4 + 3}_{9} \\
 & \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{aligned}$$

M	F	C	rule
	F	no	
0 ... 9	F	no	by Propagate ⁺
0 ... 9 10	F	no	by T-Propagate (0, 3 \models_{UF} 10)
0 ... 9 10	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \overline{4} \vee \overline{5} \vee 11 \vee 12$)
0 ... 9 10 • 11	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Decide
0 ... 9 10 • 11 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, 11 \models_{UF} 13)
0 ... 9 10 • 11 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	$\overline{7} \vee \overline{13}$	by T-Conflict (7, 13 $\models_{UF} \perp$)
0 ... 9 10 13	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Backjump
0 ... 9 10 13 11	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate (0, 1, $\overline{13} \models_{UF} \overline{11}$)
0 ... 9 10 13 11 12	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Propagate
...	...	(exercise)	
fail	by Fail

Applications

Some Applications of SMT

Program Analysis and Verification

- Software Model Checking² (e.g. BLAST, SLAM)
- K-Induction-Based Model Checking³ (e.g. Kind)
- Concolic or Directed Automated Random Testing⁴ (e.g. CUTE, KLEE, PEX)
- Program Verifiers (e.g. VCC,⁵ Why3⁶)
- Translation Validation for Compilers (e.g. TVOC⁷)

² Jhala and Majumdar, **Software Model Checking**, ACM Computing Surveys 2009.

³ Hagen and Tinelli, **Scaling Up the Formal Verification of Lustre Programs with SMT-Based Techniques**, FMCAD'08.

⁴ Godefroid, Klarlund, and Sen, **DART: Directed Automated Random Testing**, PLDI '05

⁵ Dahlweid, Moskal, Santen et al. **VCC: Contract-based modular verification of concurrent C**, ICSE '09.

⁶ Bobot, Filliâtre, Marché, and Paskevich, **Why3: Shepherd Your Herd of Provers**, Boogie '11.

⁷ Zuck, Pnueli, Goldberg, Barrett et al., **Translation and Run-Time Validation of Loop Transformations**, FMSD '05.

Some Applications of SMT

Non-verification Applications

- AI (e.g. Robot Task Planning⁸)
- Biology (e.g. Analysis of Synthetic Biology Models⁹)
- Databases (e.g. Checking Preservation of Database Integrity¹⁰)
- Network Analysis (e.g. Checking Security of OpenFlow Rules¹¹)
- Scheduling (e.g. Rotating Workforce Scheduling¹²)
- Security (e.g. Automatic Exploit Generation¹³)
- Synthesis (e.g. Symbolic Term Exploration¹⁴)

⁸Witsch, Skubch, et al., **Using Incomplete Satisfiability Modulo Theories to Determine Robotic Tasks**, IROS '13.

⁹Yordanov and Wintersteiger, **SMT-based analysis of Biological Computation**, NFM '13.

¹⁰Baltopoulos, Borgström, and Gordon, **Maintaining Database Integrity with Refinement Types**, ECOOP '11.

¹¹Son, Shin, Yegneswaran et al., **Model Checking Invariant Security Properties in OpenFlow**, ICC '13.

¹²Erkinger, **Rotating Workforce Scheduling as Satisfiability Modulo Theories**, Master's Thesis, TU Wien, 2013.

¹³Avgerinos, Cha, Rebert et al. **Automatic Exploit Generation**, CACM '14.

¹⁴Kneuss, Kuraj, Kuncak, and Suter, **Synthesis Modulo Recursive Functions**, OOPSLA '13.

SMT users are clamouring for more capabilities

New theories in the pipeline

- Theory of *sets with cardinality*
- Theory of *floating-point numbers*
- Theory of *separation logic*

Going forward

- There is a huge opportunity to design and implement decision procedures for new *domain-specific theories*

Scalability

Plenty of room for performance improvements

- SMT innovations continue at both the system and algorithm level
- Example: With recent breakthroughs in *arithmetic* and *bit-vector* algorithms, CVC4 can solve many problems that were previously too hard (for any solver)
- *Parallel computing* still largely untapped

Google

- Ongoing collaboration with Google with ambitious goals for using symbolic execution on Google code
- Lots of interesting research questions about how to make use of Google's *massive resources* to apply symbolic execution and SMT solving on a *massive scale*

Tool integration using proofs

Skeptical proof assistants

- Tools like *Coq* and *Isabelle/HOL* are used extensively to verify systems and algorithms, despite their lack of automation
- SMT solvers cannot currently help because these tools do not trust external results

Idea: Produce independently checkable proofs

- We are instrumenting CVC4 to produce *proof certificates* in a formal proof framework called LFSC
- One goal: *replay proofs* in tools like Coq and Isabelle/HOL
- Collaboration with Andrew Appel at Princeton to support his *verified software toolchain*

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Verifying Neural Networks

Deep Learning can do Amazing Things

- Identify images
- Recognize speech
- Drive cars and airplanes

Safety-Critical applications: Need ways to analyze safety of neural network

- Problem: case analysis explosion (SMT and LP solvers blow up)
- Culprit: activation functions, e.g. Rectified Linear Unit (ReLU)
- Solution: Develop a *custom* theory solver for neural networks

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Verifying Neural Networks

Reluplex

- Simple SMT solver for neural networks ¹⁵
- Extends simplex algorithm to reason about ReLU functions
- Result: Can prove properties of networks an order of magnitude larger than any previous method
- Dramatic reduction in case analysis required (from 2^{300} to 2^{30})

Case study: ACAS Xu

- Traffic collision avoidance algorithm for unmanned aircraft
- Neural network controller being considered by FAA
- Reluplex successfully used to prove safety properties for these networks

¹⁵Katz, Barrett, Dill, et al., **Reluplex: An Efficient SMT Solver for Verifying Deep Neural Networks**, CAV '17.

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Agile Hardware Project

`http://aha.stanford.edu`

Collaboration with Prof. Hanrahan and Horowitz

- Build an open-source hardware flow
- Make it possible to do quick and incremental iterations of hardware designs

SMT solvers being used to

- automatically verify circuit transformations
- provide automatic and incremental place-and-route

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Next steps

- Build open-source SMT-based model-checking tools
- Develop or extend SMT theories for hardware verification
- Verify interoperation of hardware module interfaces

SMT solvers

- Provide *general-purpose* logical reasoning
- Can be customized for *domain-specific* reasoning
- *Enabler for formal methods*: automatic, expressive, scalable
- No shortage of *challenging research problems*
 - with immediate practical impact

More information

SMT resources

- SMT Survey Article: available at
<http://theory.stanford.edu/~barrett/pubs/BKM14.pdf>
- SMT-LIB standards and library <http://smtlib.org>
- SMT Competition <http://smtcomp.org>
- SMT Workshop <http://smt-workshop.org>

CVC4

- Visit the CVC4 website: <http://cvc4.cs.nyu.edu>
- Contact a CVC4 team member
- We welcome questions, feedback, collaboration proposals

Suggested Readings

1. R. Nieuwenhuis, A. Oliveras, and C. Tinelli. **Solving SAT and SAT Modulo Theories: From an abstract Davis-Putnam-Logemann-Loveland procedure to DPLL(T)**. Journal of the ACM, 53(6):937-977, 2006.
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