



软件分析与架构设计

# 可满足性模理论 (SMT)

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# First Order Logic (FOL)

## □ A first-order language $\mathcal{L}$ consists of:

### ➤ Logical symbols:

- Variables:  $x, y, z, \dots$
- Logical connectives:  $\neg, \wedge, \vee, \rightarrow, \leftrightarrow, \dots$
- Quantifiers:  $\forall, \exists$
- Parentheses

### ➤ Terms:

- Every variable is a term
- Every constant symbol is a term
- Every function  $f(t_1, \dots, t_n)$  is a term

### ➤ Non-logical symbols:

- Constant symbols:  $a, b, c, \dots$
- N-ary function symbols:  $f, g, h, \dots$
- N-ary predicate symbols:  $p, q, r, \dots$

### ➤ Well-formed formulas:

- Atomic formulas: Predicate  $P(t_1, \dots, t_n)$
- If  $\varphi$ , and  $\psi$  are formulas, then
  - ❖  $\neg\varphi, (\varphi \wedge \psi), (\varphi \vee \psi), (\varphi \rightarrow \psi), \dots$
- If  $\varphi$  is a formula,  $x$  is a variable, then
  - ❖  $\forall x \varphi, \exists x \varphi$

## □ Semantics (Interpretation):

- Interprets variables, functions, and predicates over domain  $D$
- A formula  $\varphi$  is **satisfiable** if  $I \models \varphi$  for **some** interpretation  $I$
- A formula  $\varphi$  is **valid** if  $I \models \varphi$  for **all** interpretations  $I$

# Satisfiability Modulo Theories (SMT)

## □ SMTs as generalizations of SAT

- SAT: reason **propositional logic**
- SMT: aims to reason **first order logic (FOL)**
  - But satisfiability of FOL is **undecidable**
  - SMT, instead, targets decidable or domain-specific fragments

## □ SMT: propositional logic + **domain-specific reasoning**

- deciding the satisfiability of a **(ground)** First Order formula with respect to a **background theory**
  - Software verification needs reasoning about **equality, arithmetic, data structures, ...**
- Example:  $g(a) = c \wedge ((f(g(a)) \neq f(c)) \vee g(a) = d) \wedge (c \neq d)$ 
  - Need Equality with Uninterpreted Functions (EUF) theory
- SMT is a **sweet spot** between expressiveness and efficiency

## □ Wide range of **applications**:

- SMT solvers as **backend “workhorse” engines** of many techniques and tools
- Model checking, Test generation, Software verification, Synthesis, ...

# SMT in Practice

□ **Beginnings of SMT: Early 2000s**

□ **GOOD NEWS:** efficient decision procedures for sets of ground literals exist for various theories of interest

□ **PROBLEM:** in practice, we need to deal with:

- arbitrary Boolean combinations of literals ( $\wedge, \vee, \neg$ )
- multiple theories: **combine decision procedures** for each theory
  - In practice, theories are **not isolated**
  - E.g., Software verifications needs **arithmetic, arrays, bitvectors, ...**
  - Formulas of the following form usually arise:

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

- quantifiers  $\forall, \exists$

□ **KEY practice:** exploit SAT solvers to improve performance

- **Eager approach** (a.k.a. "bit-blasting"): encode SMT into SAT
- **Lazy approach:** plug SAT solver with a decision procedure

# Eager Approach

□ **Methodology:** translate problem into equisatisfiable propositional formula and use off-the-shelf SAT solver

➤ Still dominant for bit-vector arithmetic

□ **Encoding SMT formula into SAT**

➤ EUF as an example:  $f(a) = c \wedge f(b) \neq c \wedge a \neq b$

➤ Introduce symbols to replace function applications

○ A for  $f(a)$ , B for  $f(b)$ , the formula becomes  $A = c \wedge B \neq c \wedge a \neq b$

○ Add constraints based on EUF axioms:  $a = b \rightarrow A = B$

○ Introduce bool variables to replace equations:

$$\diamond P_{A=c} \wedge \neg P_{B=c} \wedge \neg P_{a=b}, P_{a=b} \rightarrow P_{A=B}$$

○ Add constraints based on reflexivity: e.g.,  $P_{A=A}$  (not necessary here)

○ Add constraints for transitivity:

$$\diamond P_{A=c} \wedge P_{B=c} \rightarrow P_{A=B}, P_{A=B} \wedge P_{B=c} \rightarrow P_{A=c}, P_{A=B} \wedge P_{A=c} \rightarrow P_{B=c}, \dots$$

# Eager Approach

## □ Why called “Eager”?

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

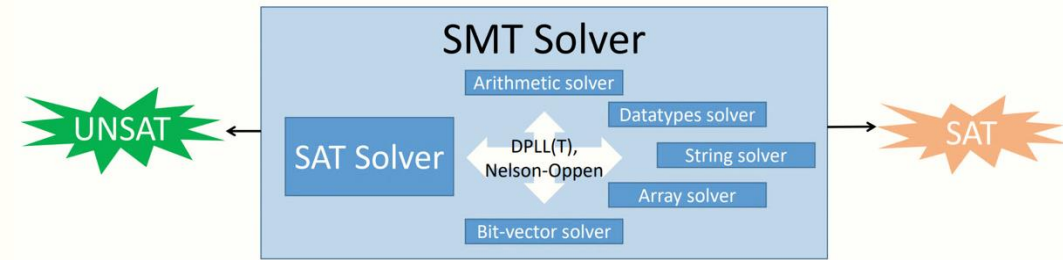
## □ Advantage:

- Can use best available SAT solver

## □ Problems of Eager Approach:

- After encoding, SAT solver cannot utilize the dedicated solution algorithms of the original theories:
  - EUF can be solved by a fixed-point algorithm that continuously merges equivalence classes
- Sophisticated encodings are needed for each theory
- Not all SMT can be encoded into SAT

# Lazy Approach



## □ Lazy SMT Solvers: CVC4, MathSat, Yices, Z3, ...

- Yices 1.0 (2006): the first efficient “general-purpose” SMT solver
- Z3 1.0 (2008): > 10000 citations, most influential tool paper at TACAS

## □ Lazy approach: Combines SAT and theory solvers

- SAT-solvers enumerate models for the boolean part
- Theory solvers check satisfiability in the theory

## □ Example 1: Arithmetic theory

$\psi$	$\triangleq$	$\psi_{\mathbb{B}}$	$\triangleq$
$c_1$	:	$\neg(2x_2 - x_3 > 2) \vee (x_1 + x_3 \leq 5)$	$\overline{A_{11}} \vee A_{12}$
$c_2$	:	$\neg(x_1 - x_3 \leq 5) \vee (x_1 - x_5 \leq 1)$	$\overline{A_{21}} \vee A_{22}$
$c_3$	:	$\neg(3x_1 - 2x_2 \leq 3) \vee \neg(x_1 - x_3 \leq 5)$	$\overline{A_{31}} \vee \overline{A_{32}}$
$c_4$	:	$\neg(3x_1 - x_3 \leq 6) \vee \neg(x_1 + x_3 \leq 5)$	$\overline{A_{41}} \vee \overline{A_{42}}$
$c_5$	:	$(x_1 + x_3 \leq 5) \vee (3x_1 - 2x_2 \leq 3)$	$A_{51} \vee A_{31}$
$c_6$	:	$(x_2 - x_4 \leq 6) \vee \neg(x_1 + x_3 \leq 5)$	$A_{61} \vee \overline{A_{62}}$
$c_7$	:	$(x_1 + x_3 \leq 5) \vee (x_3 = 3x_5 + 4) \vee \neg(x_1 - x_3 \leq 5)$	$A_{71} \vee A_{72} \vee \overline{A_{73}}$

➤ Sat-solver gives  $\{A_{12}, A_{22}, \overline{A_{31}}, \overline{A_{41}}, A_{51}, A_{61}, A_{72}\}$ ,

➤ Then the theory solver says the following is **UNSAT**

○  $\{(x_1 + x_3 \leq 5), (x_1 - x_5 \leq 1), \neg(3x_1 - 2x_2 \leq 3), \neg(3x_1 - x_3 \leq 6), (x_2 - x_4 \leq 6), (x_3 = 3x_5 + 4)\}$

➤ Add  $(A_{12} \vee A_{22} \vee A_{31} \vee A_{41} \vee A_{51} \vee A_{61} \vee A_{72})$  to  $\psi_{\mathbb{B}}$

# Lazy Approach

## □ Example 2: EUF theory

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

- SAT solver gives  $\{1, \bar{2}, \bar{4}\}$ ,
- Theory solver says the following is **T-inconsistent**:
  - $\{(g(a) = c), (f(g(a)) \neq f(c)), (c \neq d)\}$
- Add  $\bar{1} \vee \bar{2} \vee \bar{4}$  to the formula and send to SAT solver
- SAT solver gives  $\{1, 2, 3, \bar{4}\}$
- Theory solver says **T-inconsistent** to the followings:
  - $\{(g(a) = c), (f(g(a)) = f(c)), (g(a) = d), (c \neq d)\}$
- Add  $\bar{1} \vee \bar{2} \vee \bar{3} \vee \bar{4}$  to the formula
- Finally, SAT-solver declares **UNSAT** to the original formula

# Basic Lazy Approach

## □ Convert to DNF and use SAT to enumerate conjuncts

- Check one conjunction at a time
- Restrict the theory solver to conjunctions of constraints

## □ Why “lazy”?

- Theory information is used lazily when checking T-consistency of propositional models
- Theory information does not guide the search (too lazy), only validating it
  - DPLL(T) will improve it

```
1  $\psi = \text{to\_cnf}(\varphi);$ 
2 while(true){
3     res , M = check_SAT( $\psi$ );
4     if( res){
5          $M_T = \text{to\_theory}(M);$ 
6         res = check_theory( $M_T$ );
7         if(res)
8             return SAT;
9         else
10             $\psi \wedge = \neg M;$ 
11     }else
12         return UNSAT;
13 }
```

# Lazy approach: Optimizations

□ Several **optimizations for enhancing efficiency**:

~~➤ Check **T-consistency** only of full propositional models~~

➤ Check **T-consistency** of partial assignment while being built

~~➤ Given a **T-inconsistent** assignment  $M$ , add  $\neg M$  as a clause~~

➤ Given a **T-inconsistent** assignment  $M$ , identify a **T-inconsistent** subset  $M_0 \subseteq M$  and add  $\neg M_0$  as a clause

~~➤ Upon a **T-inconsistency**, add clause and restart~~

➤ Upon a **T-inconsistency**, backtrack to some point where the assignment was still **T-consistent**

➤ **T-propagation: exploited in conflict analysis**

○ Search guided by T-Solver by finding T-consequences

# DPLL(T) (also known as CDCL(T))

## □ DPLL(T) = DPLL(X) + T-Solver

- DPLL(X): Very similar to a SAT solver, enumerates Boolean models
- T-Solver: Should be incremental and backtrackable
  - Checks consistency of conjunctions of literals
  - Computes theory propagations
  - Produces explanations of inconsistency/T-propagation

## □ Example: EUF and formula $\underbrace{g(a)=c}_1 \wedge (\underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a)=d}_3) \wedge \underbrace{c \neq d}_{\bar{4}}$

- $\emptyset \parallel 1, \bar{2} \vee 3, \bar{4} \Rightarrow$  (UnitPropagate)
- $1 \parallel 1, \bar{2} \vee 3, \bar{4} \Rightarrow$  (UnitPropagate)
- $1 \bar{4} \parallel 1, \bar{2} \vee 3, \bar{4} \Rightarrow$  (T-Propagate)
- $1 \bar{4} 2 \parallel 1, \bar{2} \vee 3, \bar{4} \Rightarrow$  (T-Propagate)
- $1 \bar{4} 2 \bar{3} \parallel 1, \bar{2} \vee 3, \bar{4} \Rightarrow$  (Fail)

*UNSAT*

# DPLL(T) (also known as CDCL(T))

## □ Preprocessing

- Main goal: from arbitrary formulas to sets of clauses (CNFs)
- Also: (cheap) rewriting steps to **simplify** the input problem
  - Normalization of constraints:  $(x + x \geq 3) \rightarrow (0 \leq 2x - 3)$
  - Elimination of variables:  $(x = y) \wedge \varphi(x, y) \rightarrow \varphi(x, x)$
- **Encoding of constraints** not supported by the core engine
  - E.g., Term-ITEs:  $\varphi(\text{ite}(b, x, y)) \rightarrow \varphi(i) \wedge (b \rightarrow i = x) \wedge (\neg b \rightarrow i = y)$
  - Division:  $\varphi(x / y) \rightarrow \varphi(d) \wedge (x = d * y) \wedge (d \geq 0) \wedge (d < y)$
  - Max/Min functions:

$$\varphi(\max(x, y)) \rightarrow \varphi(m) \wedge (m \geq x) \wedge (m \geq y) \wedge ((m = x) \vee (m = y))$$

# DPLL(T) Algorithm

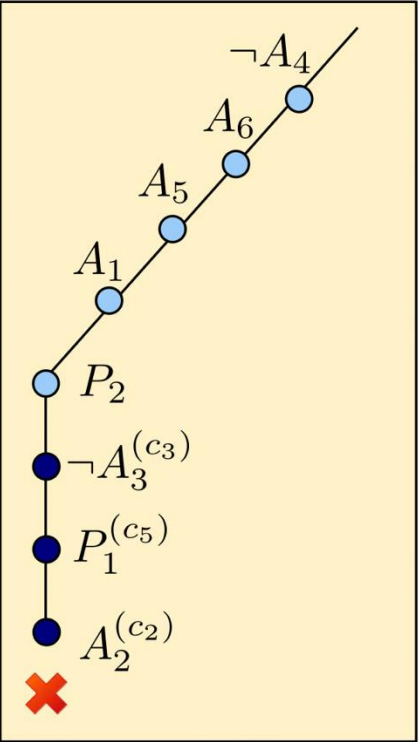
## □ T-backjumping and T-learning

- When unsat, T-solver can produce a **reason for the inconsistency**
- **T-conflict set**: inconsistent subset of the input constraints
- T-conflict clause given as input to the CDCL **conflict analysis**
  - Drives non-chronological backtracking (backjumping)
  - Can be learned by the SAT solver
- The less redundant the T-conflict set, the more search is saved
  - Ideally, should be **minimal** (irredundant)
    - ❖ Removing any element makes the set consistent
  - But for some theories might be expensive to achieve
    - ❖ Trade-off between size and cost

```
def DPLL_T():
    while True:
        conflict = False
        if unit_propagation():
            res = theory_check(not all_assigned())
            if res == False: conflict = True
            elif res == True: conflict = theory_propagation()
            elif learn_theory_lemmas(): continue
            elif not all_assigned(): decide()
        else:
            build_model()
            return SAT
    else: conflict = True
    if conflict:
        lvl, cls = conflict_analysis()
        if lvl < 0: return UNSAT
    else:
        backtrack(lvl)
        learn(cls)
```

# Example

$\varphi$	$\stackrel{\text{def}}{=}$	$\varphi^{\text{Bool}}$	$\stackrel{\text{def}}{=}$
$c_1 :$	$(2x_2 - x_3 > 2) \vee P_1$		$A_1 \vee P_1$
$c_2 :$	$\neg P_2 \vee (x_1 - x_5 \leq 1)$		$\neg P_2 \vee A_2$
$c_3 :$	$\neg(3x_1 - 2x_2 \leq 3) \vee \neg P_2$		$\neg A_3 \vee \neg P_2$
$c_4 :$	$\neg(3x_1 - x_3 \leq 6) \vee \neg P_1$		$\neg A_4 \vee \neg P_1$
$c_5 :$	$P_1 \vee (3x_1 - 2x_2 \leq 3)$		$P_1 \vee A_3$
$c_6 :$	$(x_2 - x_4 \leq 6) \vee \neg P_1$		$A_5 \vee \neg P_1$
$c_7 :$	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$



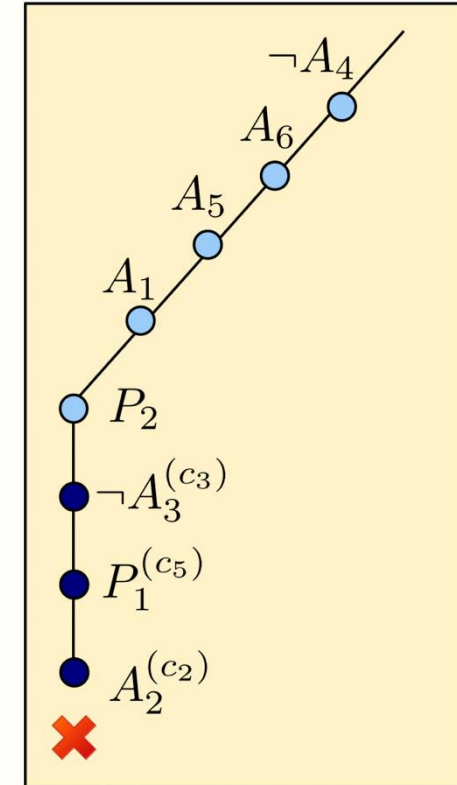
$$M = [\neg A_4, A_6, A_5, A_1, P_2, \neg A_3, P_1, A_2]$$

$$M' = \{ \neg(3x_1 - x_3 \leq 6), (x_3 = 3x_5 + 4), (x_2 - x_4 \leq 6), \neg(2x_2 - x_3 > 2), \neg(3x_1 - 2x_2 \leq 3), (x_1 - x_5 \leq 1) \}$$

T-conflict set

# Example

$\varphi$	$\stackrel{\text{def}}{=}$	$\varphi^{\text{Bool}}$	$\stackrel{\text{def}}{=}$
$c_1$	:	$(2x_2 - x_3 > 2) \vee P_1$	$A_1 \vee P_1$
$c_2$	:	$\neg P_2 \vee (x_1 - x_5 \leq 1)$	$\neg P_2 \vee A_2$
$c_3$	:	$\neg(3x_1 - 2x_2 \leq 3) \vee \neg P_2$	$\neg A_3 \vee \neg P_2$
$c_4$	:	$\neg(3x_1 - x_3 \leq 6) \vee \neg P_1$	$\neg A_4 \vee \neg P_1$
$c_5$	:	$P_1 \vee (3x_1 - 2x_2 \leq 3)$	$P_1 \vee A_3$
$c_6$	:	$(x_2 - x_4 \leq 6) \vee \neg P_1$	$A_5 \vee \neg P_1$
$c_7$	:	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$



Conflict analysis:

$$\frac{\overbrace{A_4 \vee \neg A_6 \vee \neg A_2}^{T\text{-conflict clause}} \quad \overbrace{\neg P_2 \vee A_2}^{c_2}}{A_4 \vee \neg A_6 \vee \neg P_2}$$



# Early Pruning

## □ Different strategies to call T-solver

- Eagerly, every time a new atom is assigned
- After every round of **BCP** (Boolean Constraint Propagation)
- Heuristically, based on some statistics (e.g. effectiveness, ...)

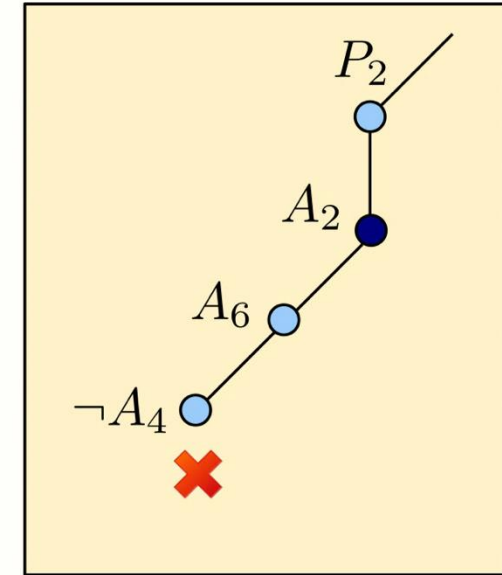
## □ No need for a conclusive answer during early pruning calls

- Can apply approximate checks
- Trade effectiveness for efficiency

Example: on linear integer arithmetic, solve only the real relaxation during early pruning calls

# Example

$\varphi$	$\stackrel{\text{def}}{=}$	$\varphi^{\text{Bool}}$	$\stackrel{\text{def}}{=}$
$c_1$	:	$(2x_2 - x_3 > 2) \vee P_1$	$A_1 \vee P_1$
$c_2$	:	$\neg P_2 \vee (x_1 - x_5 \leq 1)$	$\neg P_2 \vee A_2$
$c_3$	:	$\neg(3x_1 - 2x_2 \leq 3) \vee \neg P_2$	$\neg A_3 \vee \neg P_2$
$c_4$	:	$\neg(3x_1 - x_3 \leq 6) \vee \neg P_1$	$\neg A_4 \vee \neg P_1$
$c_5$	:	$P_1 \vee (3x_1 - 2x_2 \leq 3)$	$P_1 \vee A_3$
$c_6$	:	$(x_2 - x_4 \leq 6) \vee \neg P_1$	$A_5 \vee \neg P_1$
$c_7$	:	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$



$$M = [P_2, A_2, A_6, \neg A_4]$$

**UNSAT**

$$\text{T-conflict} = \{\neg(3x_1 - x_3 \leq 6), (x_3 = 3x_5 + 4), (x_1 - x_5 \leq 1)\}$$

# T-propagation

## □ T-solvers might support deduction of unassigned constraints

- If early pruning check on  $M$  returns sat, T-solver might also return a set of unassigned atoms  $D$  such that  $M \models_{\mathcal{T}} l$  for all  $l \in D$

## □ T-propagation: add each such $l$ to the CDCL stack

- As if BCP was applied to the (T-valid) clause  $\neg M \vee l$  (T-reason)
- But do not compute the T-reason clause explicitly yet

## □ Lazy explanation: compute T-reason clause **only if needed during conflict analysis**

- Like T-conflicts, the less redundant the better

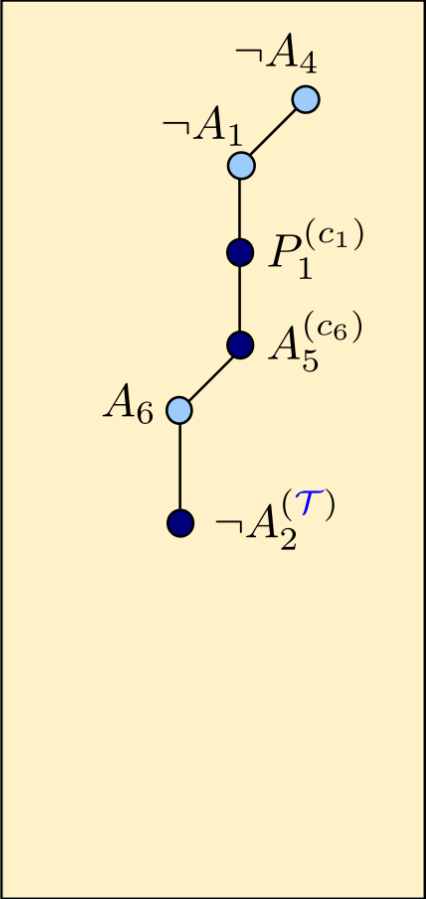
# Example

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$c_3$	:	$\neg(3x_1 - 2x_2 \leq 3) \vee \neg P_2$	$\neg A_3 \vee \neg P_2$
$c_4$	:	$\neg(3x_1 - x_3 \leq 6) \vee \neg P_1$	$\neg A_4 \vee \neg P_1$
$c_5$	:	$P_1 \vee (3x_1 - 2x_2 \leq 3)$	$P_1 \vee A_3$
$c_6$	:	$(x_2 - x_4 \leq 6) \vee \neg P_1$	$A_5 \vee \neg P_1$
$c_7$	:	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$
$c_8$	:	$P_2 \vee (2x_2 - 3x_1 \geq 5) \vee$ $(x_3 + x_5 - 4x_1 \geq 0)$	$P_2 \vee A_7 \vee A_8$

$$M = [\neg A_4, \neg A_1, P_1, A_5, A_6]$$

$$\frac{\neg(3x_1 - x_3 \leq 6) \quad (x_3 = 3x_5 + 4)}{\neg(3x_1 - 3x_5 \leq 10)}$$

$$\neg(x_1 - x_5 \leq 1) \equiv \neg A_2$$

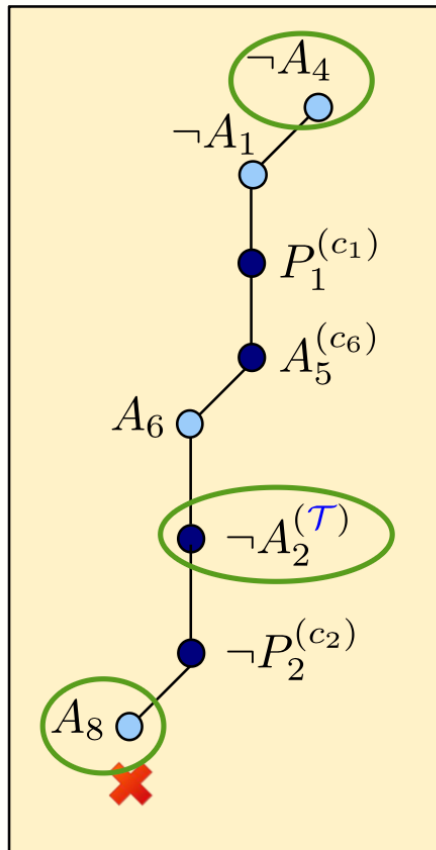


# Example

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$c_3$	$\neg(3x_1 - 2x_2 \leq 3) \vee \neg P_2$	$\neg A_3 \vee \neg P_2$	
$c_4$	$\neg(3x_1 - x_3 \leq 6) \vee \neg P_1$	$\neg A_4 \vee \neg P_1$	
$c_5$	$P_1 \vee (3x_1 - 2x_2 \leq 3)$	$P_1 \vee A_3$	
$c_6$	$(x_2 - x_4 \leq 6) \vee \neg P_1$	$A_5 \vee \neg P_1$	
$c_7$	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$	
$c_8$	$P_2 \vee (2x_2 - 3x_1 \geq 5) \vee$ $(x_3 + x_5 - 4x_1 \geq 0)$	$P_2 \vee A_7 \vee A_8$	

$$M = [\neg A_4, \neg A_1, P_1, A_5, A_6, \neg A_2, \neg P_2, A_8]$$

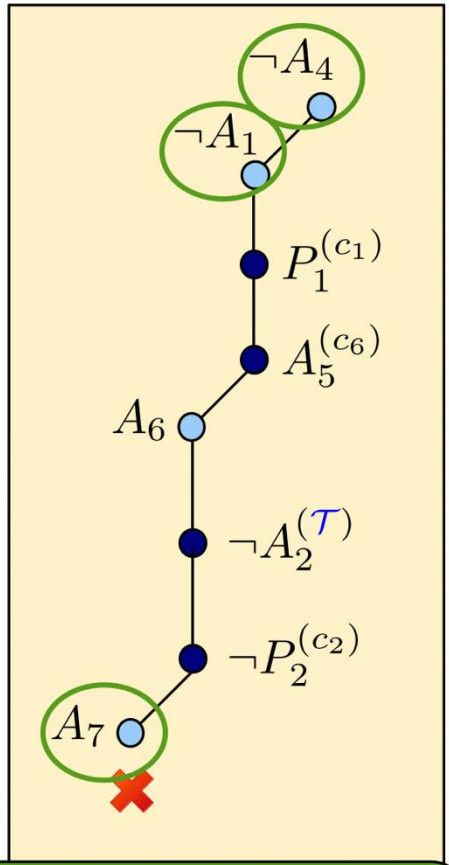
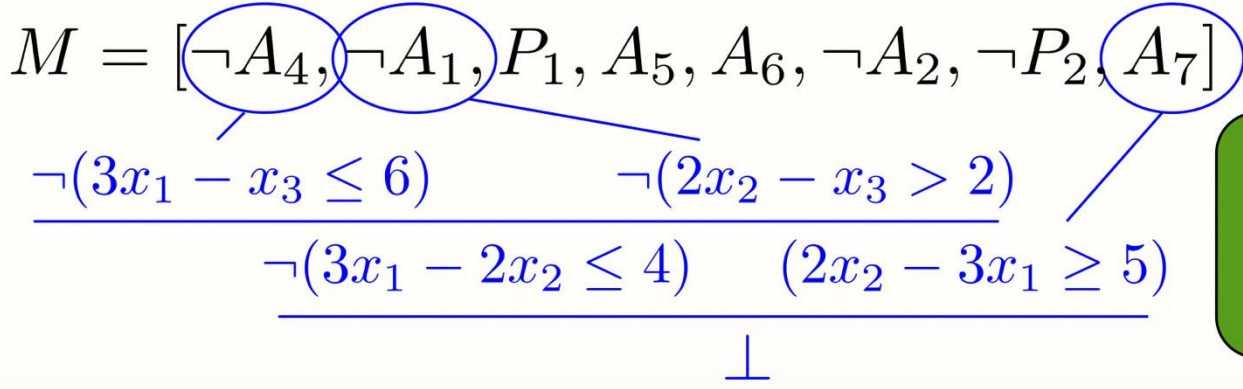
$$\frac{\frac{\neg(3x_1 - x_3 \leq 6)}{\neg(-x_3 + 3x_5 \leq 3)} \quad \frac{\neg(x_1 - x_5 \leq 1)}{(x_3 + x_5 - 4x_1 \geq 0)}}{\perp}$$



Conflict analysis →  
compute  
T-reason for  $\neg A_2$

# Example

$\varphi$	$\stackrel{\text{def}}{=}$	$\varphi^{\text{Bool}}$	$\stackrel{\text{def}}{=}$
$c_1$	:	$(2x_2 - x_3 > 2) \vee P_1$	$A_1 \vee P_1$
$c_2$	:	$\neg P_2 \vee (x_1 - x_5 \leq 1)$	$\neg P_2 \vee A_2$
$c_3$	:	$\neg(3x_1 - 2x_2 \leq 3) \vee \neg P_2$	$\neg A_3 \vee \neg P_2$
$c_4$	:	$\neg(3x_1 - x_3 \leq 6) \vee \neg P_1$	$\neg A_4 \vee \neg P_1$
$c_5$	:	$P_1 \vee (3x_1 - 2x_2 \leq 3)$	$P_1 \vee A_3$
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$c_7$	:	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$
$c_8$	:	$P_2 \vee (2x_2 - 3x_1 \geq 5) \vee$ $(x_3 + x_5 - 4x_1 \geq 0)$	$P_2 \vee A_7 \vee A_8$



$\neg A_2$  not involved in conflict analysis  $\rightarrow$  no need to compute T-reason

# Theories of Interest: EUF

## □ Equality (i.e., "=") with Uninterpreted Functions

- Used to abstract non-supported constructions,
  - e.g: Non-linear multiplication and ALUs in circuits
- Example 1:  $a * (f(b) + f(c)) = d \wedge b * (f(a) + f(c)) \neq d \wedge a = b$  is **UNSAT**
  - $a = b \Rightarrow f(a) = f(b)$  by **EUF**, no need to compute  $f$  function
- Example 2: we abstract the above formula into:
  - $h(a, g(f(b), f(c))) = d \wedge h(b, g(f(a), f(c))) \neq d \wedge a = b$
  - EUF allows to know the formula is **UNSAT** without knowing  $h, g$ , and  $f$

# Theories of Interest: Arithmetic

□ Very **useful** for **obvious reasons**

□ **Restricted** fragments support **more efficient** methods:

- Bounds:  $x \bowtie k$  with  $\bowtie \in \{<, >, \leq, \geq, =\}$
- Difference logic:  $x - y \bowtie k$ , with  $\bowtie \in \{<, >, \leq, \geq, =\}$
- UTVPI:  $\pm x \pm y \bowtie k$ , with  $\bowtie \in \{<, >, \leq, \geq, =\}$
- Linear arithmetic, e.g:  $2x - 3y + 4z \leq 5$
- Non-linear arithmetic, e.g:  $2xy + 4xz^2 - 5y \leq 10$
- Variables are either **reals** or **integers**

□ **Machine-inspired** arithmetic: **floating-point arithmetic**

# Theories of Interest: Arrays

□ Two interpreted function symbols read and write

□ Theory is axiomatized by:

➤  $\forall a \forall i \forall v \text{ (read(write(a, i, v), i) = v)}$

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

□ Sometimes **extensionality** is added:

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

□ Is the following set of literals satisfiable?

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

$a = b$

$i = j$

□ Used for:

➤ Software verification (arrays)

➤ Hardware verification (memories)

# Theories of Interest: Bit vectors

- Constants represent **vectors of bits**
- Useful both for **hardware and software verification**
- Different type of operations:
  - **String**-like operations: concatenation, extraction, ...
  - **Logical** operations: bit-wise not, or, and, ...
  - **Arithmetic** operations: add, subtract, multiply, ...
- **Example:** Assume bit-vectors have size 3. Is the formula SAT?
  - $a[0:1] \neq b[0:1] \wedge (a|b) = c \wedge c[0] = 0 \wedge a[1] + b[1] = 0$

# Combination of theories

□ Very often in practice **more than one theory** is needed

➤ Example:

$$\varphi \stackrel{\text{def}}{=} (x_1 \geq 0) \wedge (x_1 < 1) \wedge ((f(x_1) = f(0)) \rightarrow (\text{rd}(\text{wr}(P, x_2, x_3), x_2 + x_1) = x_3 + 1))$$

➤ Use theories in Arrays, EUF, Arithmetic

□ **How to build solvers for  $SMT(T_1 \cdots T_n)$  that are both efficient and modular?**

➤ Can we reuse Ti-solvers and combine them?

➤ Under what conditions?

□ **How do we go from  $DPLL(T)$  to  $DPLL(T_1 \cdots T_n)$ ?**

# The Nelson-Oppen method

□ A general technique for combining  $T_i$ -solvers

□ Requires:

- $T_i$ 's to have **disjoint signatures**, i.e. no symbols in common (except =)
- $T_i$ 's to be **stably-infinite**, i.e. every quantifier-free  $T_i$ -satisfiable formula is satisfiable in an infinite model of  $T_i$

□ How it works (for  $T_1 \cup T_2$ )

➤ Preprocessing **purification** step on the input formula  $\varphi$

- Pure formula: no atom containing symbols of different  $T_i$ 's (except =)
  - ❖ By labeling subterms

$$\begin{array}{l} : \\ \varphi \stackrel{\text{def}}{=} f(\overbrace{x+3y}^{l_1}) + 4 \leq \overbrace{g(w)}^{l_2} \mapsto \\ (f(\overbrace{l_1}^{l_3}) + 4 \leq l_2) \wedge (l_1 = x+3y) \wedge (l_2 = g(w)) \mapsto \\ \boxed{l_3} + 4 \leq \boxed{l_2} \wedge \boxed{l_1} = x+3y \wedge \boxed{l_2} = g(w) \wedge (f(\boxed{l_1}) = \boxed{l_3}) \end{array}$$

□  $T_i$ -solvers cooperate by **exchanging entailed interface equalities**

➤ i.e., equalities between shared variables

# Example

**LIA**  $(x_1 \geq 0), (x_1 \leq 1), (x_2 \geq x_6)$   
 $(x_2 \leq x_6 + 1), (x_5 = x_4 - 1)$   
 $(x_3 = 0), (x_4 = 1)$

$\models$

$(x_1 = x_3) \vee (x_1 = x_4)$

$(x_5 = x_6)$

$\models$

$(x_2 = x_3) \vee (x_2 = x_4)$

$\neg(f(x_1) = f(x_2)),$  **EUF**  
 $\neg(f(x_2) = f(x_4)),$   
 $(f(x_3) = x_5), (f(x_1) = x_6)$



$(x_1 = x_3)$

$(x_1 = x_4)$

$\models$

$(x_5 = x_6)$

No more deductions possible



$(x_2 = x_3)$

$(x_2 = x_4)$



# (Optional)项目：SMT求解器学习

## □主流SMT求解器

- [Z3](#): 微软研究院开发, 支持多种理论, Python、C++等多种编程语言, 广泛应用于软件验证、程序分析、测试生成等领域
- [Yices 2](#): 斯坦福国际研究院 (SRI) 开发, 专注于命题逻辑的SMT求解器, 在硬件设计和验证领域有广泛应用
- [CVC5](#): 斯坦福大学等开发, 在字符串理论实现上更为成熟, 处理复杂字符串操作时效率更高, 在软件安全领域发挥重要作用
- [Bitwuzla](#): 专注于位向量和数组理论
- ...

## □选择一款SMT求解器学习其使用方法, 背后原理以及支持理论

- 撰写不少于1000字的学习报告