
VTSA summer school 2015

Exploiting SMT for Verification of Infinite-State Systems

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Agenda



Part 1: Introduction to SMT

Part 2: Interpolation in SMT and in Verification

Part 3: SMT-based Verification with IC3

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Exploiting SMT for Verification of Infinite-State Systems

1. Introduction to SMT

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Introduction

CDCL-based SAT solvers

The DPLL(T) architecture

Some relevant T-solvers

Combination of theories

Boolean Satisfiability (SAT)

- Given a formula φ in propositional logic, with predicates (aka variables) A, B, C, \dots , find an assignment to the variables

$$\{A \mapsto \top, B \mapsto \perp, \dots\}$$

that makes the formula true, or prove that none exists

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that makes the formula true, or prove that none exists

■ Example

$$\varphi := (A \vee (B \wedge C)) \wedge (A \implies \neg C)$$

SAT, with solution (model)

$$\mu := \{A \mapsto \perp, \\ B \mapsto \top, \\ C \mapsto \top\}$$

■ Satisfiability Modulo Theories

- Given a (quantifier-free) FOL formula and a (decidable) combination of theories $\mathcal{T}_1 \cup \dots \cup \mathcal{T}_m$, is there an assignment to the free variables x_1, \dots, x_n that makes the formula true?

- Example:

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

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■ Example:

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

Linear Integer Arithmetic (LIA)

■ Satisfiability Modulo Theories

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Linear Integer Arithmetic (LIA)

Equality (EUF)

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Linear Integer Arithmetic (LIA)

Equality (EUF)

Arrays (A)

■ Satisfiability Modulo Theories

- Given a (quantifier-free) FOL formula and a (decidable) combination of theories $\mathcal{T}_1 \cup \dots \cup \mathcal{T}_m$, is there an assignment to the free variables x_1, \dots, x_n that makes the formula true?

■ Example:

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

$$\text{LIA} \models (x_1 = 0)$$

$$\text{EUF} \models f(x_1) = f(0)$$

$$\text{A} \models \text{rd}(\text{wr}(P, x_2, x_3), x_2) = x_3$$

$$\text{Bool} \models \text{rd}(\text{wr}(P, x_2, x_3), x_2 + x_1) = x_3 + 1$$

$$\text{LIA} \models \perp$$

The “early days”

- **The Simplify theorem prover [Detlefs, Nelson, Saxe]**
 - The grandfather of SMT solvers
- **Efficient decision procedures**
 - Equality logic + extensions (Congruence Closure)
 - Linear arithmetic (Simplex)
 - Theory combination (Nelson-Oppen method)
 - Quantifiers (E-matching with triggers)
- **Inefficient boolean search**

SMT: some history - 2

The SAT breakthrough

- **late '90s - early 2000: major progress in SAT solvers**
- **CDCL paradigm: Conflict-Driven Clause-Learning DPLL**
 - Grasp, (z)Chaff, Berkmin, MiniSat, ...
- combine strengths of **model search** and **proof search** in a single procedure
 - **Model search:** efficient BCP and variable selection heuristics
 - **Proof search:** conflict analysis, non-chronological backtracking, clause learning
- **Smart ideas + clever engineering “tricks”**

From SAT to SMT

- **exploit advances in SAT solving for richer logics**
 - Boolean combinations of constraints over (combinations of) background theories
- **The Eager approach** (a.k.a. “bit-blasting”)
 - Encode an SMT formula into propositional logic
 - Solve with an off-the-shelf efficient SAT solver
 - Pioneered by **UCLID**
 - Still the dominant approach for bit-vector arithmetic

SMT: some history - 4

The Lazy approach and DPLL(T) (2002 – 2004)

- **(non-trivial) combination of SAT (CDCL) and T-solvers**
 - SAT-solver enumerates models of boolean skeleton of formula
 - Theory solvers check consistency in the theory
 - Most popular approach (e.g. Barcelogic, CVC4, MathSAT, SMTInterpol, Yices, Z3, VeriT, ...)

- **Yices 1.0 (2006)**
 - The first **efficient** “general-purpose” SMT solver

- **Z3 1.0 (2008)**
 - > 1600 citations, **most influential tool** paper at TACAS

Some notation and definitions

- Signature Σ , functions f, g, h, \dots and predicates P, Q, R, \dots
- Variables x, y, z, \dots , quantifier-free formulas φ, ψ, \dots
- **Structure** $\mathcal{A} = (\mathcal{D}, \mathcal{I})$, $\mathcal{I}(f) : \mathcal{D}^n \mapsto \mathcal{D}$ $\mathcal{I}(P) : \mathcal{D}^n \mapsto \{\top, \perp\}$
- **Assignment** $\xi : \xi(x) \mapsto \mathcal{D}$
- Evaluation $\llbracket \cdot \rrbracket_{\mathcal{A}, \xi}$ $\llbracket x \rrbracket_{\mathcal{A}, \xi} \stackrel{\text{def}}{=} \xi(x)$ $\llbracket f(x) \rrbracket_{\mathcal{A}, \xi} \stackrel{\text{def}}{=} \mathcal{I}(f)(\llbracket x \rrbracket_{\mathcal{A}, \xi})$
- φ is **satisfiable** in \mathcal{A}, ξ iff $\llbracket \varphi \rrbracket_{\mathcal{A}, \xi} = \top$ (\mathcal{A}, ξ is a **model** of φ)
- φ is **valid** in \mathcal{A} ($\mathcal{A} \models \varphi$) iff satisfiable for all ξ (\mathcal{A} is a **model**)
- A **theory** T is a set of Σ -structures $\mathcal{A}_1 = (\mathcal{D}, \mathcal{I}_1) \dots \mathcal{A}_n = (\mathcal{D}, \mathcal{I}_n)$

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- A **theory** T is a set of Σ -structures $\mathcal{A}_1 = (\mathcal{D}, \mathcal{I}_1) \dots \mathcal{A}_n = (\mathcal{D}, \mathcal{I}_n)$
- **Example:** $\Sigma = \{\hat{0}, \hat{1}, \dots, \hat{+}, \hat{\cdot}, \hat{\geq}, \dots\}$ $\mathcal{D} = \mathbb{Z}$ $\mathcal{I}(\hat{n}) = n$ $\mathcal{I}(\hat{+}) = +$
 $(\hat{3}x + y \hat{\geq} \hat{5})$ is **satisfiable** $\xi \stackrel{\text{def}}{=} \{x \mapsto 1, y \mapsto 2\}$ is a model
 $(\hat{2}x + \hat{3}y \hat{\geq} 0) \wedge (-\hat{1} \hat{\geq} x) \wedge (-\hat{1} \hat{\geq} y)$ is **unsatisfiable**

Introduction

CDCL-based SAT solvers

The DPLL(T) architecture

Some relevant T-solvers

Combination of theories

CDCL-based SAT solvers

- **C**onflict-**D**riven **C**lause-**L**earning paradigm
 - Architecture of modern SAT solvers (e.g. Minisat, Lingeling, ...)
 - The “DPLL” part of DPLL(T)
- Combine efficient **model search** and **conflict analysis**
 - **Model search**
 - Stack-based representation of **partial truth assignment** (trail), extended by performing deductions and decisions
 - When all variables are assigned, return SAT with trail as model
 - **Conflict analysis**
 - When a conflict is detected, apply **boolean resolution** to generate a new implied clause that contradicts the trail
 - learn the blocking clause and use it for non-chronological backtracking
 - when the empty clause is derived, return UNSAT

The CDCL algorithm for SAT

```
CDCL(F)
  A = [], decision_level = 0
  while (true)
    if (deduce(F, A))
      if (!all_assigned(F, A))
        lit = decide(F, A)
        decision_level++
        A = A + (lit, -)
      else return SAT
    else
      lvl, cls = analyze(F, A)
      if (lvl < 0) return UNSAT
      else
        backtrack(F, A, lvl)
        learn(cls)
        decision_level = lvl
```

The CDCL algorithm for SAT

CDCL(F)

```
A = [], decision_level = 0
while (true)
  if (deduce(F, A))
    if (!all_assigned(F, A))
      lit = decide(F, A)
      decision_level++
      A = A + (lit, -)
    else return SAT
  else
    lvl, cls = analyze(F, A)
    if (lvl < 0) return UNSAT
    else
      backtrack(F, A, lvl)
      learn(cls)
      decision_level = lvl
```

Trail of
assignments
(*lit, reason*)

Model
Search

Conflict
Analysis

- Explore search space by adding elements to the trail
 - Trail encodes a set of partial assignments

$$\neg A; B; C \mapsto \{\sigma \mid \sigma(A) = \perp \wedge \sigma(B) = \sigma(C) = \top\}$$

- deductions using unit propagation
 - If a clause has one unassigned literal and all the others set to false, propagate the value of the missing one

$$\left. \begin{array}{l} \text{Trail: } \neg A; B; C \\ \text{Clause: } A \vee \neg B \vee \neg D \end{array} \right\} \neg A; B; C; \neg D$$

- All literals assigned by unit propagation have an associated reason clause in the trail
 - The unit clause that forced the assignment to the literal

CDCL: model search

- if a clause has all literals assigned to false, deduce returns false and marks the clause as conflicting
- otherwise, if no more deduction is possible, decide picks an unassigned literal to add to the trail
 - No reason is attached to the literal in this case
 - Decisions partition the trail into **decision levels**
- When **all literals** are **assigned**, SAT is returned
 - The trail is a **model** for the input CNF

Example

Input clauses

$$c_1 : \neg A_3 \vee A_2 \vee \neg A_{12}$$

$$c_2 : \neg A_1 \vee A_3 \vee A_9$$

$$c_3 : \neg A_2 \vee \neg A_3 \vee A_4$$

$$c_4 : \neg A_4 \vee A_5 \vee A_{10}$$

$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—

Example

Input clauses

$$c_1 : \neg A_3 \vee A_2 \vee \neg A_{12}$$

$$c_2 : \neg A_1 \vee A_3 \vee A_9$$

$$c_3 : \neg A_2 \vee \neg A_3 \vee A_4$$

$$c_4 : \neg A_4 \vee A_5 \vee A_{10}$$

$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—

Decide A_1

Example

Input clauses

$$c_1 : \neg A_3 \vee A_2 \vee \neg A_{12}$$

$$c_2 : \neg A_1 \vee A_3 \vee A_9$$

$$c_3 : \neg A_2 \vee \neg A_3 \vee A_4$$

$$c_4 : \neg A_4 \vee A_5 \vee A_{10}$$

$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—

Deduce

Example

Input clauses

$$c_1 : \neg A_3 \vee A_2 \vee \neg A_{12}$$

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$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2

Example

Input clauses

$$c_1 : \neg A_3 \vee A_2 \vee \neg A_{12}$$

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$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—

Decide A_{12}

Example

Input clauses

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$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Deduce

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—

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$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1

Deduce

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Deduce

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Lit	Reason
$\neg A_9$	—
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A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1
A_4	c_3

Example

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Deduce

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1
A_4	c_3
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Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
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$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Conflict!

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1
A_4	c_3
A_5	c_4
A_6	c_5

CDCL: conflict analysis

- **Goal:** backtrack from inconsistent assignment, and avoid repeating the same mistake in the future
- **Naive approach:** collect **all decisions** in the trail, and learn a blocking clause implying that at least one of them must be flipped
- **Proof-based approach:** exploit the information in the trail to generate an explanation for the conflict
 - Generate a new lemma, using boolean resolution, that blocks the current assignment and all those sharing the same reason for inconsistency

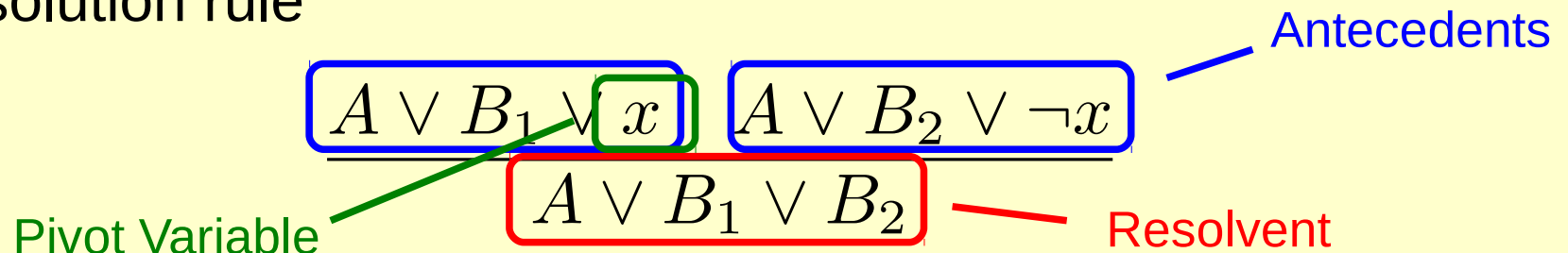
- Resolution rule

$$\frac{A \vee B_1 \vee x \quad A \vee B_2 \vee \neg x}{A \vee B_1 \vee B_2}$$

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Resolution rule



Example - "1st UIP" learning strategy

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$$\neg A_5 \vee \neg A_6$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1
A_4	c_3
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$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

$$\frac{\neg A_5 \vee \neg A_6 \quad \neg A_4 \vee \neg A_6 \vee A_{11}}{\neg A_5 \vee \neg A_4 \vee A_{11}}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
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$$c_2 : \neg A_1 \vee A_3 \vee A_9$$

$$c_3 : \neg A_2 \vee \neg A_3 \vee A_4$$

$$c_4 : \neg A_4 \vee A_5 \vee A_{10}$$

$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

$$\frac{\neg A_5 \vee \neg A_6 \quad \neg A_4 \vee \neg A_6 \vee A_{11}}{\neg A_5 \vee \neg A_4 \vee A_{11} \quad \neg A_4 \vee A_5 \vee A_{10}}$$

$$\frac{\quad}{\neg A_4 \vee A_{11} \vee A_{10}}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1
A_4	c_3
A_5	c_4
A_6	c_5

Example - "1st UIP" learning strategy

Input clauses

$$c_1 : \neg A_3 \vee A_2 \vee \neg A_{12}$$

$$c_2 : \neg A_1 \vee A_3 \vee A_9$$

$$c_3 : \neg A_2 \vee \neg A_3 \vee A_4$$

$$c_4 : \neg A_4 \vee A_5 \vee A_{10}$$

$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1
A_4	c_3
A_5	c_4
A_6	c_5

$$\begin{array}{r}
 \frac{\neg A_5 \vee \neg A_6 \quad \neg A_4 \vee \neg A_6 \vee A_{11}}{\neg A_5 \vee \neg A_4 \vee A_{11} \quad \neg A_4 \vee A_5 \vee A_{10}} \\
 \frac{\neg A_4 \vee A_{11} \vee A_{10} \quad \neg A_2 \vee \neg A_3 \vee A_4}{A_{11} \vee A_{10} \vee \neg A_2 \vee \neg A_3}
 \end{array}$$

Example - "1st UIP" learning strategy

Input clauses

$$c_1 : \neg A_3 \vee A_2 \vee \neg A_{12}$$

$$c_2 : \neg A_1 \vee A_3 \vee A_9$$

$$c_3 : \neg A_2 \vee \neg A_3 \vee A_4$$

$$c_4 : \neg A_4 \vee A_5 \vee A_{10}$$

$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1
A_4	c_3
A_5	c_4
A_6	c_5

$$\begin{array}{r}
 \frac{\neg A_5 \vee \neg A_6 \quad \neg A_4 \vee \neg A_6 \vee A_{11}}{\neg A_5 \vee \neg A_4 \vee A_{11} \quad \neg A_4 \vee A_5 \vee A_{10}} \\
 \frac{\neg A_4 \vee A_{11} \vee A_{10} \quad \neg A_2 \vee \neg A_3 \vee A_4}{\neg A_3 \vee A_2 \vee \neg A_{12} \quad A_{11} \vee A_{10} \vee \neg A_2 \vee \neg A_3} \\
 \hline
 \neg A_{12} \vee A_{11} \vee A_{10} \vee \neg A_3
 \end{array}$$

Example - "1st UIP" learning strategy

Input clauses

$$c_1 : \neg A_3 \vee A_2 \vee \neg A_{12}$$

$$c_2 : \neg A_1 \vee A_3 \vee A_9$$

$$c_3 : \neg A_2 \vee \neg A_3 \vee A_4$$

$$c_4 : \neg A_4 \vee A_5 \vee A_{10}$$

$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1
A_4	c_3
A_5	c_4
A_6	c_5

Reached level limit

$$\begin{array}{r}
 \frac{\neg A_5 \vee \neg A_6 \quad \neg A_4 \vee \neg A_6 \vee A_{11}}{\neg A_5 \vee \neg A_4 \vee A_{11} \quad \neg A_4 \vee A_5 \vee A_{10}} \\
 \frac{\neg A_4 \vee A_{11} \vee A_{10} \quad \neg A_2 \vee \neg A_3 \vee A_4}{\neg A_3 \vee A_2 \vee \neg A_{12} \quad A_{11} \vee A_{10} \vee \neg A_2 \vee \neg A_3} \\
 \hline
 \neg A_{12} \vee A_{11} \vee A_{10} \vee \neg A_3
 \end{array}$$

Example – “decision” learning strategy

Input clauses

$$c_1 : \neg A_3 \vee A_2 \vee \neg A_{12}$$

$$c_2 : \neg A_1 \vee A_3 \vee A_9$$

$$c_3 : \neg A_2 \vee \neg A_3 \vee A_4$$

$$c_4 : \neg A_4 \vee A_5 \vee A_{10}$$

$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

$$\frac{\neg A_5 \vee \neg A_6 \quad \neg A_4 \vee \neg A_6 \vee A_{11}}{\neg A_5 \vee \neg A_4 \vee A_{11} \quad \neg A_4 \vee A_5 \vee A_{10}}$$

$$\frac{\neg A_4 \vee A_{11} \vee A_{10} \quad \neg A_2 \vee \neg A_3 \vee A_4}{\neg A_3 \vee A_2 \vee \neg A_{12} \quad A_{11} \vee A_{10} \vee \neg A_2 \vee \neg A_3}$$

$$\frac{\neg A_{12} \vee A_{11} \vee A_{10} \vee \neg A_3 \quad \neg A_1 \vee A_3 \vee A_9}{\neg A_{12} \vee A_{11} \vee A_{10} \vee \neg A_1 \vee A_9}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1
A_4	c_3
A_5	c_4
A_6	c_5

Example – “decision” learning strategy

Input clauses

$$c_1 : \neg A_3 \vee A_2 \vee \neg A_{12}$$

$$c_2 : \neg A_1 \vee A_3 \vee A_9$$

$$c_3 : \neg A_2 \vee \neg A_3 \vee A_4$$

$$c_4 : \neg A_4 \vee A_5 \vee A_{10}$$

$$c_5 : \neg A_4 \vee A_6 \vee A_{11}$$

$$c_6 : \neg A_5 \vee \neg A_6$$

$$c_7 : A_1 \vee A_7 \vee \neg A_{12}$$

$$c_8 : A_1 \vee A_8$$

$$c_9 : \neg A_7 \vee \neg A_8 \vee \neg A_{13}$$

Trail

Lit	Reason
$\neg A_9$	—
$\neg A_{10}$	—
$\neg A_{11}$	—
A_{13}	—
A_1	—
A_3	c_2
A_{12}	—
A_2	c_1
A_4	c_3
A_5	c_4
A_6	c_5

$$\begin{array}{l}
 \frac{\neg A_5 \vee \neg A_6 \quad \neg A_4 \vee \neg A_6 \vee A_{11}}{\neg A_5 \vee \neg A_4 \vee A_{11} \quad \neg A_4 \vee A_5 \vee A_{10}} \\
 \frac{\neg A_4 \vee A_{11} \vee A_{10} \quad \neg A_2 \vee \neg A_3 \vee A_4}{\neg A_3 \vee A_2 \vee \neg A_{12} \quad A_{11} \vee A_{10} \vee \neg A_2 \vee \neg A_3} \\
 \frac{\neg A_{12} \vee A_{11} \vee A_{10} \vee \neg A_3 \quad \neg A_1 \vee A_3 \vee A_9}{\neg A_{12} \vee A_{11} \vee A_{10} \vee \neg A_1 \vee A_9}
 \end{array}$$

Introduction

CDCL-based SAT solvers

The DPLL(T) architecture

Some relevant T-solvers

Combination of theories

The lazy approach to SMT

- Deciding the satisfiability of φ modulo \mathcal{T} can be reduced to deciding \mathcal{T} -satisfiability of **conjunctions (sets) of constraints**
 - Can exploit efficient decision procedures for sets of constraints, existing for many important theories
- **Naive approach:** convert φ to an equivalent φ' in **disjunctive normal form** (DNF), and check each conjunction separately
- Main idea of **lazy SMT**: use an efficient SAT solver to **enumerate** conjuncts without computing the DNF explicitly

A basic approach

■ Offline lazy SMT

```
F = CNF_bool( $\varphi$ )
while true:
    res, M = check_SAT(F)
    if res == true:
        M' = to_T(M)
        res = check_T(M')
        if res == true:
            return SAT
        else:
            F += !M
    else:
        return UNSAT
```

A basic approach

■ Offline lazy SMT

```
F = CNF_bool( $\varphi$ )
while true:
    res, M = check_SAT(F)
    if res == true:
        M' = to_T(M)
        res = check_T(M')
        if res == true:
            return SAT
        else:
            F += !M
    else:
        return UNSAT
```

Boolean
reasoning

A basic approach

■ Offline lazy SMT

```
F = CNF_bool( $\varphi$ )  
while true:  
    res, M = check_SAT(F)  
    if res == true:  
        M' = to_T(M)  
        res = check_T(M')  
        if res == true:  
            return SAT  
        else:  
            F += !M  
    else:  
        return UNSAT
```

Boolean
reasoning

Theory
reasoning

A basic approach

Offline lazy SMT

```
F = CNF_bool( $\varphi$ )
while true:
    res, M = check_SAT(F)
    if res == true:
        M' = to_T(M)
        res = check_T(M')
        if res == true:
            return SAT
        else:
            F += !M
    else:
        return UNSAT
```

Boolean
reasoning

Theory
reasoning

Block bad solutions

Example

$$\begin{aligned} \varphi & \stackrel{\text{def}}{=} \\ c_1 & : (2x_2 - x_3 > 2) \vee P_1 \\ c_2 & : \neg P_2 \vee (x_1 - x_5 \leq 1) \\ c_3 & : \neg(3x_1 - 2x_2 \leq 3) \vee \neg P_2 \\ c_4 & : \neg(3x_1 - x_3 \leq 6) \vee \neg P_1 \\ c_5 & : P_1 \vee (3x_1 - 2x_2 \leq 3) \\ c_6 & : (x_2 - x_4 \leq 6) \vee \neg P_1 \\ c_7 & : P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2 \end{aligned}$$

$$\begin{aligned} \varphi^{\text{Bool}} & \stackrel{\text{def}}{=} \\ & A_1 \vee P_1 \\ & \neg P_2 \vee A_2 \\ & \neg A_3 \vee \neg P_2 \\ & \neg A_4 \vee \neg P_1 \\ & P_1 \vee A_3 \\ & A_5 \vee \neg P_1 \\ & P_1 \vee A_6 \vee \neg P_2 \end{aligned}$$

Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$

$$M = \{P_1, P_2, \neg A_1, A_2, \neg A_3, \neg A_4, A_5, A_6\}$$

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

Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$

$$M = \{P_1, P_2, \neg A_1, A_2, \neg A_3, \neg A_4, A_5, A_6\}$$

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

$$\neg(3x_1 - 3x_5 - 4 \leq 6) \mapsto \neg(x_1 - x_5 \leq 10/3) \mapsto (x_1 - x_5 > 10/3)$$

UNSAT \rightarrow add $\neg M$ and continue

- **Online** approach to lazy SMT
- **Tight integration** between a CDCL-like SAT solver (“**DPLL**”) and the decision procedure for T (“**T-solver**”), based on:
 - T -driven backjumping and learning
 - Early pruning
 - T -solver incrementality
 - T -propagation
 - Filtering of assignments to check
 - Creation of new T -atoms and T -lemmas “on-demand”
 - ...

T-backjumping and *T*-learning

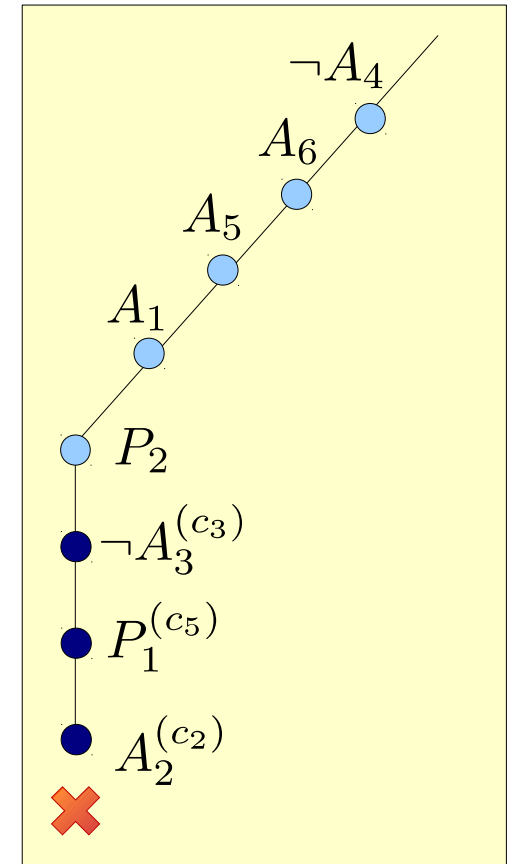
- When unsat, *T*-solver can produce **reason for inconsistency**
 - ***T*-conflict set**: inconsistent subset of the input constraints
- *T*-conflict clause given as input to the CDCL **conflict analysis**
 - Drives non-chronological backtracking (backjumping)
 - Can be learned by the SAT solver
- The 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

Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	:	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$

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

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



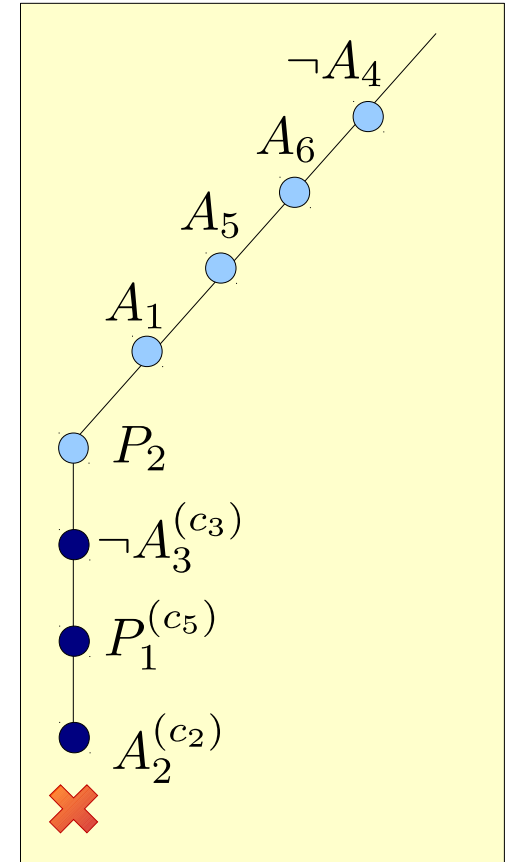
Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	:	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$

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

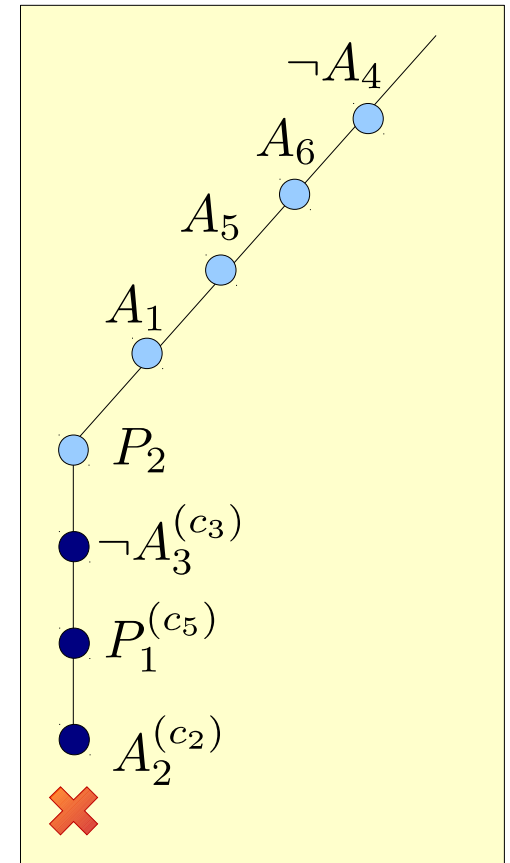
$$M' = \{ \neg(3x_1 - x_3 \leq 6), (x_3 = 3x_5 + 4), (x_2 - x_4 \leq 6), \neg(2x_2 - x_3 > 2), \neg(3x_1 - 2x_2 \leq 3), (x_1 - x_5 \leq 1) \}$$

T-conflict set



Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	:	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$

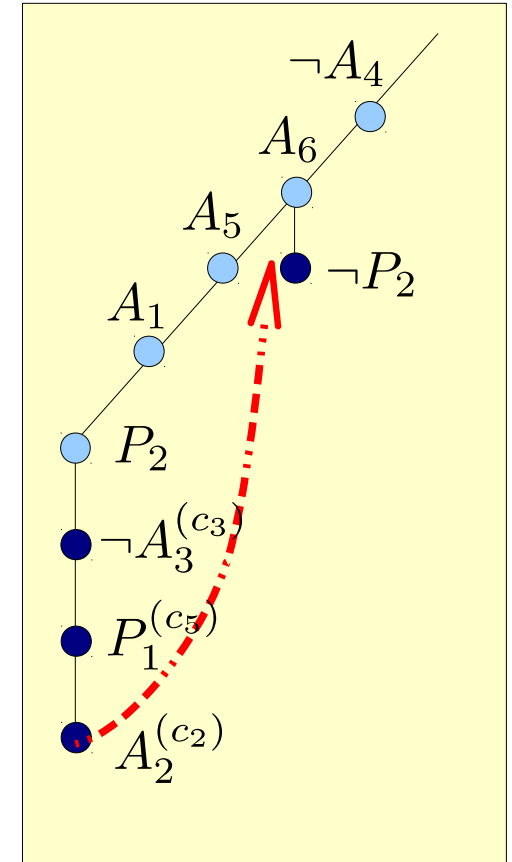


Conflict analysis:

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

Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	:	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$
c_8	:	$A_4 \vee \neg A_6 \vee \neg P_2$	

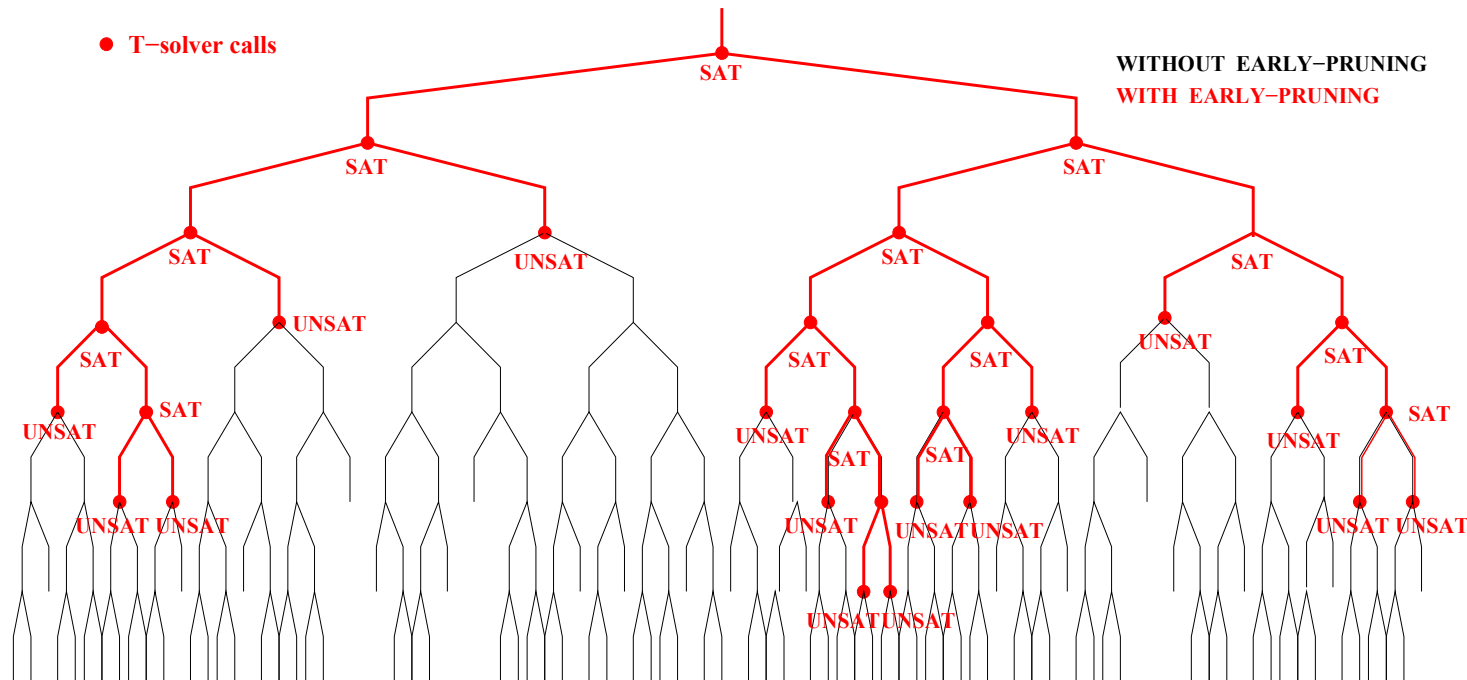


Conflict analysis:

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

Early pruning

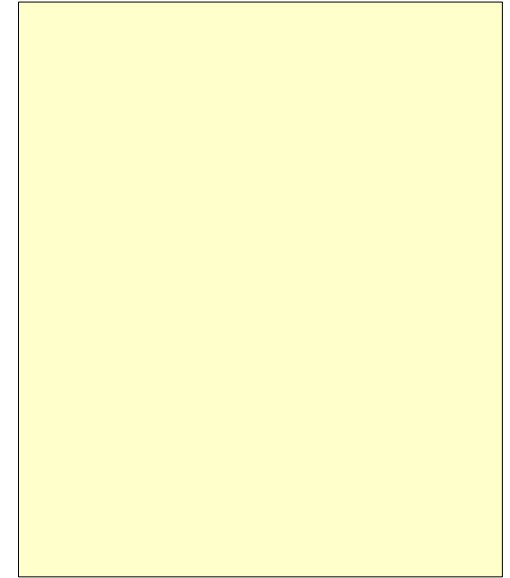
- Invoke T -solver on **intermediate assignments**, during the CDCL search
 - If **unsat** is returned, can backtrack immediately
- **Advantage:** can drastically prune the search tree
- **Drawback:** possibly many **useless (expensive) T -solver calls**



- Different strategies to call T -solver
 - Eagerly, every time a new atom is assigned
 - After every round of BCP
 - Heuristically, based on some statistics (e.g. effectiveness, ...)
 - **No need of 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

φ	$\stackrel{\text{def}}{=}$	φ^{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	:	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$

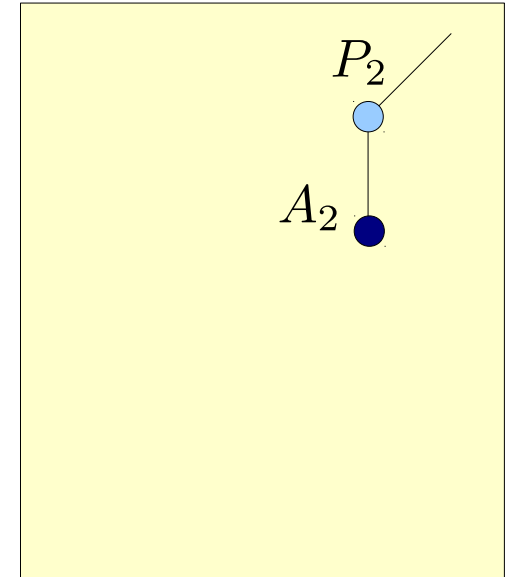


Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$

$$M = [P_2, A_2]$$

SAT

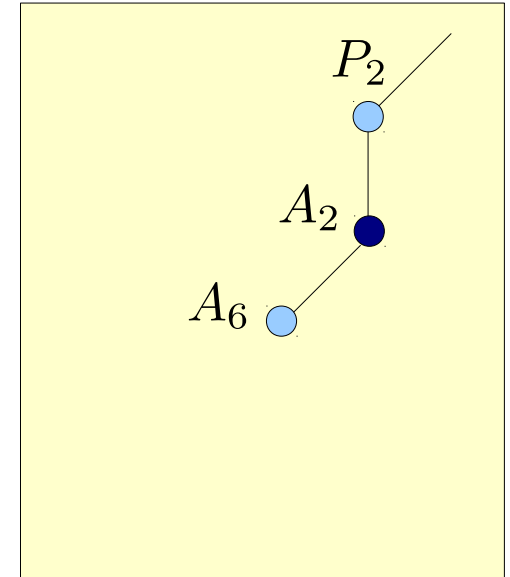


Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$

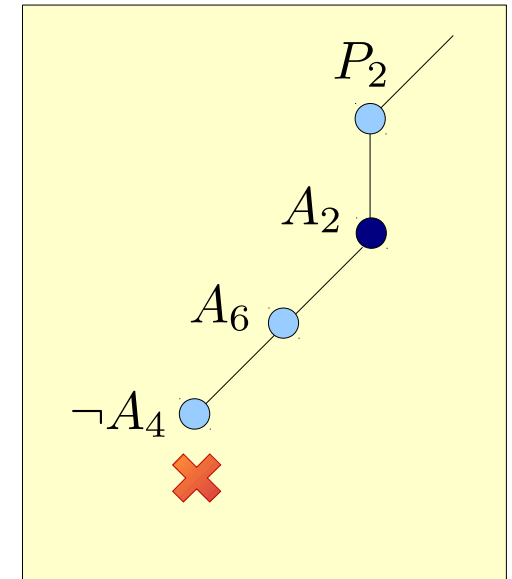
$$M = [P_2, A_2, A_6]$$

SAT



Example

φ	$\stackrel{\text{def}}{=}$	φ^{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

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

T-solver incrementality

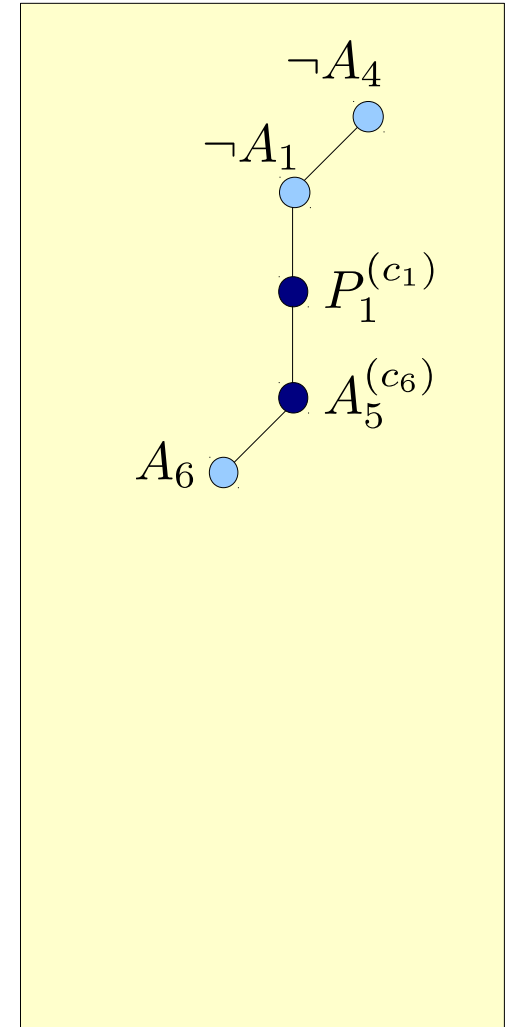
- With early pruning, *T*-solvers invoked **very frequently** on **similar** problems
 - **Stack** of constraints (the assignment stack of CDCL) **incrementally updated**
- **Incrementality**: when a new constraint is added, no need to redo all the computation “from scratch”
- **Backtrackability**: support cheap (stack-based) removal of constraints without “resetting” the internal state
- **Crucial for efficiency**
 - Distinguishing feature for effective integration in DPLL(*T*)

- T-solvers might support **deduction** of unassigned constraints
 - If early pruning check on M returns **sat**, T -solver might also return a set D of **unassigned atoms** 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

φ	$\stackrel{\text{def}}{=}$	φ^{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	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$
c_8	$P_2 \vee (2x_2 - 3x_1 \geq 5) \vee$ $(x_3 + x_5 - 4x_1 \geq 0)$		$P_2 \vee A_7 \vee A_8$

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



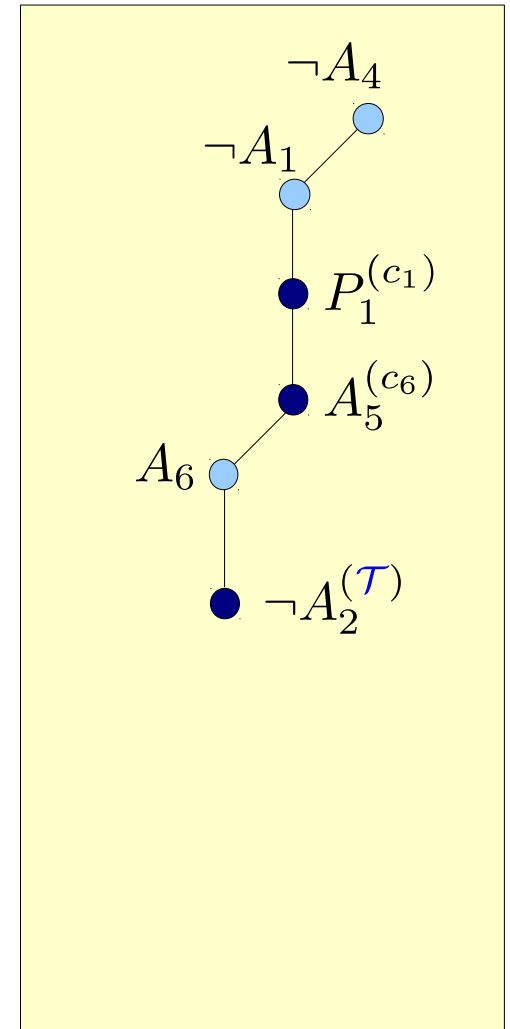
Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	:	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$	$P_1 \vee A_6 \vee \neg P_2$
c_8	:	$P_2 \vee (2x_2 - 3x_1 \geq 5) \vee$ $(x_3 + x_5 - 4x_1 \geq 0)$	$P_2 \vee A_7 \vee A_8$

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

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

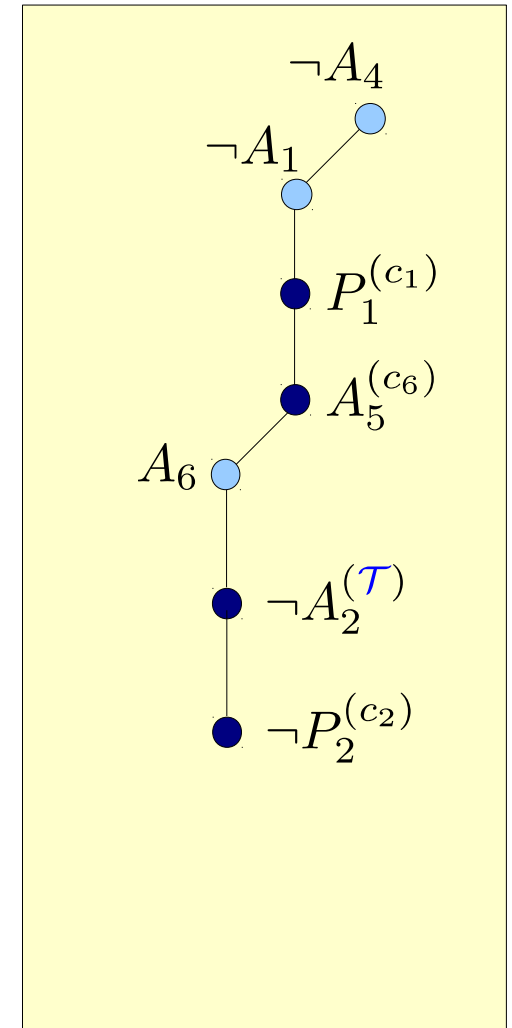
$$\neg(x_1 - x_5 \leq 1) \equiv \neg A_2$$



Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$
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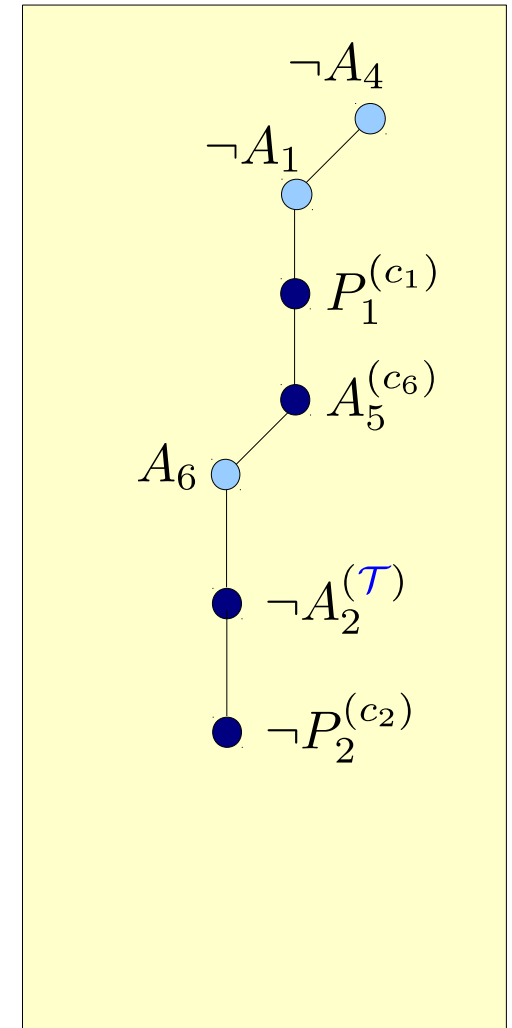
$$M = [\neg A_4, \neg A_1, P_1, A_5, A_6, \neg A_2, \neg P_2]$$



Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$
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$$M = [\neg A_4, \neg A_1, P_1, A_5, A_6, \neg A_2, \neg P_2]$$



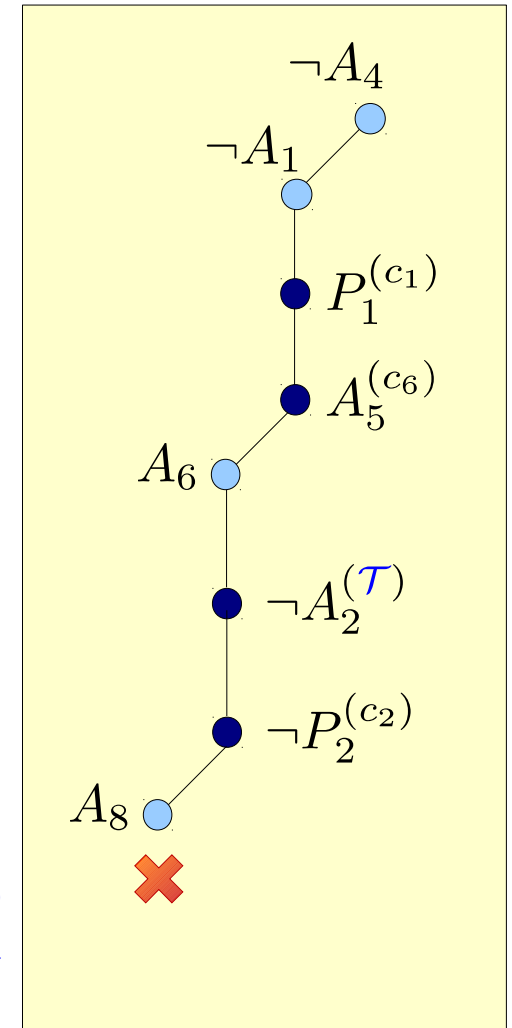
Example

φ	$\stackrel{\text{def}}{=}$	φ^{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	$\neg(3x_1 - x_3 \leq 6) \vee \neg P_1$		$\neg A_4 \vee \neg P_1$
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c_7	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$
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$$M = [\neg A_4, \neg A_1, P_1, A_5, A_6, \neg A_2, \neg P_2, A_8]$$

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

\perp

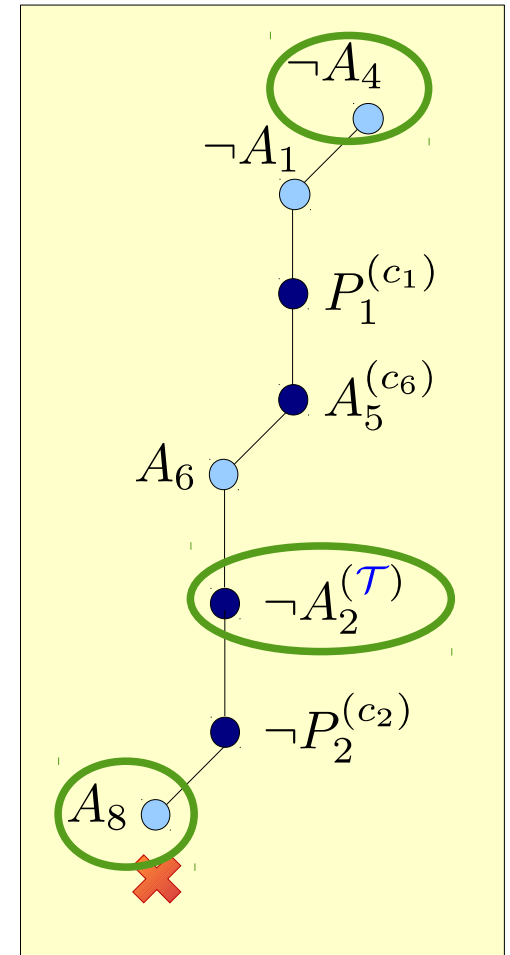


Example

φ	$\stackrel{\text{def}}{=}$	φ^{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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$$M = [\neg A_4, \neg A_1, P_1, A_5, A_6, \neg A_2, \neg P_2, A_8]$$

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



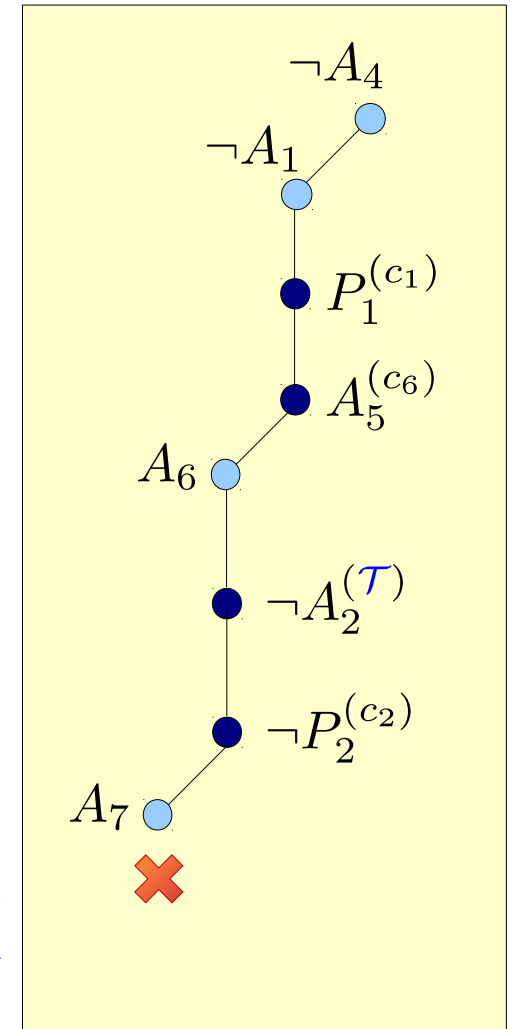
Conflict analysis →
compute
T-reason for $\neg A_2$

Example

φ	$\stackrel{\text{def}}{=}$	φ^{Bool}	$\stackrel{\text{def}}{=}$
c_1	$(2x_2 - x_3 > 2) \vee P_1$		$A_1 \vee P_1$
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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	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$
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$$M = [\neg A_4, \neg A_1, P_1, A_5, A_6, \neg A_2, \neg P_2, A_7]$$

$$\frac{\frac{\neg(3x_1 - x_3 \leq 6) \quad \neg(2x_2 - x_3 > 2)}{\neg(3x_1 - 2x_2 \leq 4)} \quad (2x_2 - 3x_1 \geq 5)}{\perp}$$

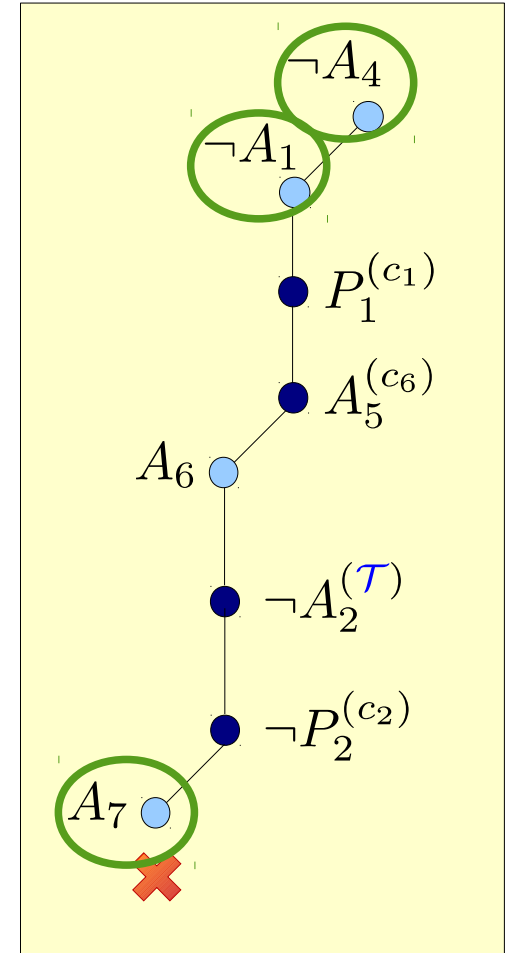


Example

φ	$\stackrel{\text{def}}{=}$	φ^{Bool}	$\stackrel{\text{def}}{=}$
c_1	$(2x_2 - x_3 > 2) \vee P_1$		$A_1 \vee P_1$
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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	$(x_2 - x_4 \leq 6) \vee \neg P_1$		$A_5 \vee \neg P_1$
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$$M = [\neg A_4, \neg A_1, P_1, A_5, A_6, \neg A_2, \neg P_2, A_7]$$

$$\frac{\neg(3x_1 - x_3 \leq 6) \quad \neg(2x_2 - x_3 > 2)}{\neg(3x_1 - 2x_2 \leq 4) \quad (2x_2 - 3x_1 \geq 5)} \perp$$



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

Filtering of assignments

- Remove **unnecessary literals** from current assignment M
 - **Irrelevant** literals: l s.t. $M \setminus \{l\} \models \varphi$ (φ arbitrary, not CNF)
 - **Ghost** literals: l occurs only in clauses satisfied by $M \setminus \{l\}$
 - **Pure** literals: $\neg l \in M$ and l occurs only positively in φ
 - **Note**: this is **not** the pure-literal rule of SAT!
- **Pros:**
 - reduce effort for T -solver
 - increases the chances of finding a solution
- **Cons:**
 - may weaken the effect of early pruning (esp. with T -propagation)
 - may introduce overhead in SAT search
- Typically used for **expensive theories**

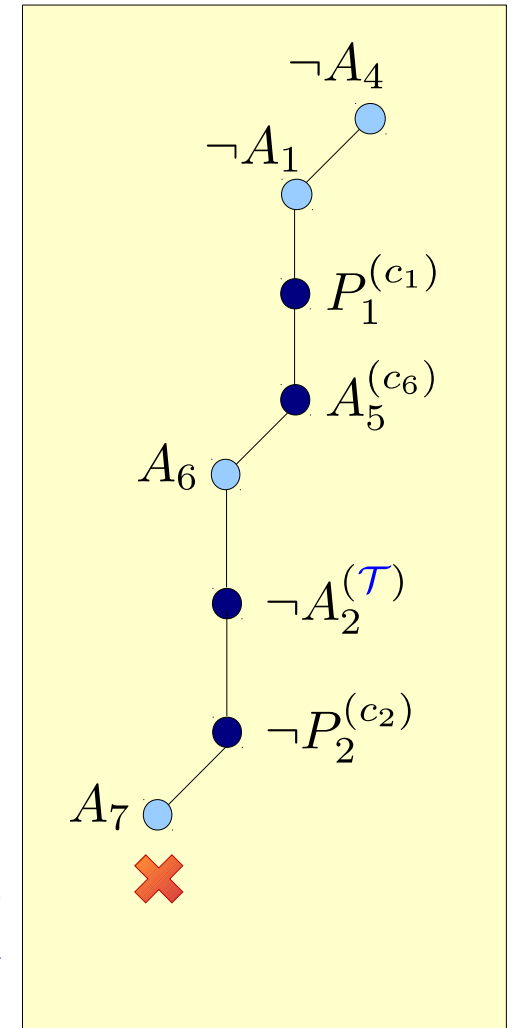
Example

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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	$P_1 \vee (x_3 = 3x_5 + 4) \vee \neg P_2$		$P_1 \vee A_6 \vee \neg P_2$
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$$M = [\neg A_4, \neg A_1, P_1, A_5, A_6, \neg A_2, \neg P_2, A_7]$$

$$\frac{\neg(3x_1 - x_3 \leq 6) \quad \neg(2x_2 - x_3 > 2)}{\neg(3x_1 - 2x_2 \leq 4) \quad (2x_2 - 3x_1 \geq 5)}$$

\perp



Example

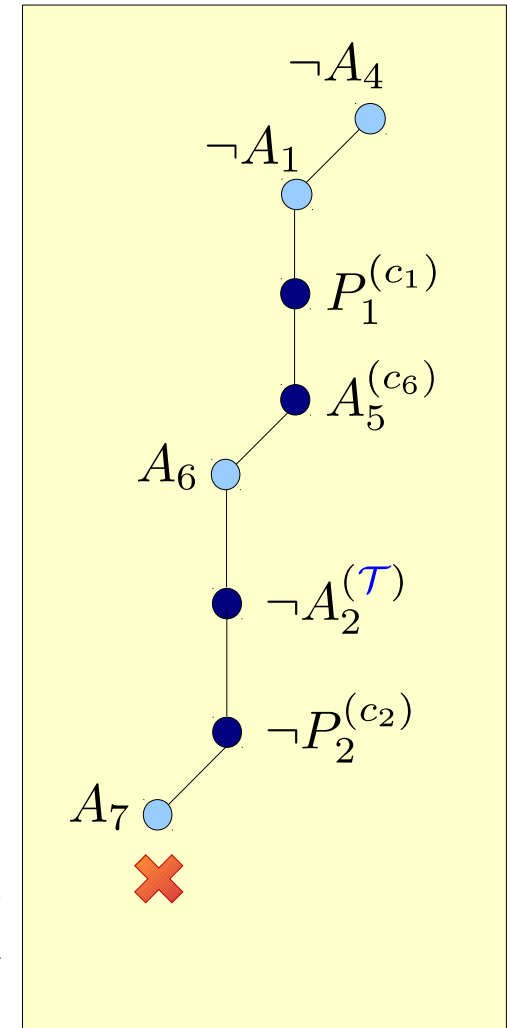
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c_2	$\neg P_2 \vee (x_1 - x_3 < 1)$		$\neg P_2 \vee A_2$
c_3	$\neg(3x_1 - 2x_2 \leq 3)$		$\neg A_3 \vee \neg P_2$
c_4	$\neg(3x_1 - x_3 \leq 6)$		$\neg A_4 \vee \neg P_1$
c_5	$P_1 \vee (3x_1 - 2x_2 \leq 3)$		$P_1 \vee A_3$
c_6	$(x_2 - x_4 \leq 6) \vee \neg P_1$		$A_5 \vee \neg P_1$
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c_8	$P_2 \vee (2x_2 - 3x_1 \geq 5) \vee$ $(x_3 + x_5 - 4x_1 \geq 0)$		$P_2 \vee A_7 \vee A_8$

Ghost!

$$M = [\neg A_4, \neg A_1, P_1, A_5, A_6, \neg A_2, \neg P_2, A_7]$$

$$\frac{\neg(3x_1 - x_3 \leq 6) \quad \neg(2x_2 - x_3 > 2)}{\neg(3x_1 - 2x_2 \leq 4) \quad (2x_2 - 3x_1 \geq 5)}$$

⊥

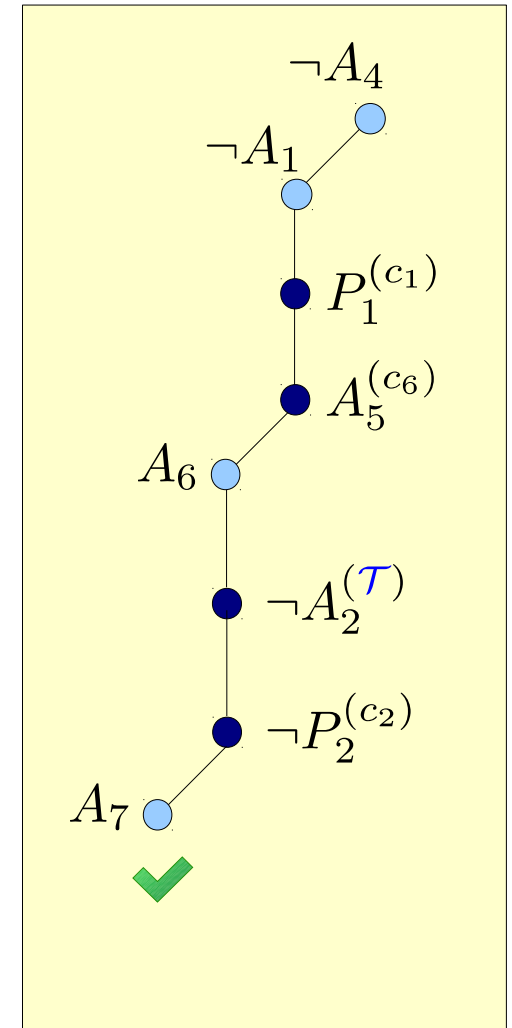


Example

φ	$\stackrel{\text{def}}{=}$	φ^{Bool}	$\stackrel{\text{def}}{=}$
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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	$P_1 \vee (3x_1 - 2x_2 \leq 3)$		$P_1 \vee A_3$
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c_8	$P_2 \vee (2x_2 - 3x_1 \geq 5) \vee$ $(x_3 + x_5 - 4x_1 \geq 0)$		$P_2 \vee A_7 \vee A_8$

$$M = [\neg A_4, \quad , P_1, A_5, A_6, \neg A_2, \neg P_2, A_7]$$

SAT



T-atoms and T-lemmas on demand

- Some T-solvers might need to perform **internal case splits** to decide satisfiability

- **Example:** linear integer arithmetic

$$(x - 3y \leq 0), (y - 2x \leq 0), (x + 3y \leq 3) \mapsto \begin{cases} \text{case } (y \leq 0) \mapsto \perp \\ \text{case } (y \geq 1) \mapsto \perp \end{cases}$$

- **Splitting on-demand:** use the SAT solver for case splits
 - Encode splits as T-valid clauses (**T-lemmas**) with **fresh T-atoms**
 - Generated **on-the-fly during search**, when needed
 - **Benefits:** reuse the efficient SAT search
 - simplify the implementation
 - exploit advanced search-space exploration techniques (backjumping, learning, restarts, ...)
 - **Potential drawback:** “pollute” the SAT search

T-atoms and *T*-lemmas on demand

- *T*-solver can now return **unknown** also for complete checks
 - In this case, it must also produce one or more *T*-lemmas
 - SAT solver learns the lemmas and continues searching
 - eventually, *T*-solver **can decide** sat/unsat
- **Termination** issues
 - If SAT solver drops lemmas, might get into an infinite loop
 - similar to the Boolean case (and the “basic” SMT case), similar solution (e.g. monotonically increase # of kept lemmas)
 - *T*-solver can generate an **infinite number of new *T*-atoms!**
 - For several theories (e.g. linear integer arithmetic, arrays) enough to draw new *T*-atoms **from a finite set** (dependent on the input problem)

T-solver interface example

```
class TheorySolver {
    bool tell_atom(Var boolatom, Expr tatom);

    void new_decision_level();
    void backtrack(int level);

    void assume(Lit l);
    lbool check(bool approx);

    void get_conflict(LitList &out);

    Lit get_next_implied();
    bool get_explanation(Lit implied, LitList &out);

    bool get_lemma(LitList &out);

    Expr get_value(Expr term);
};
```

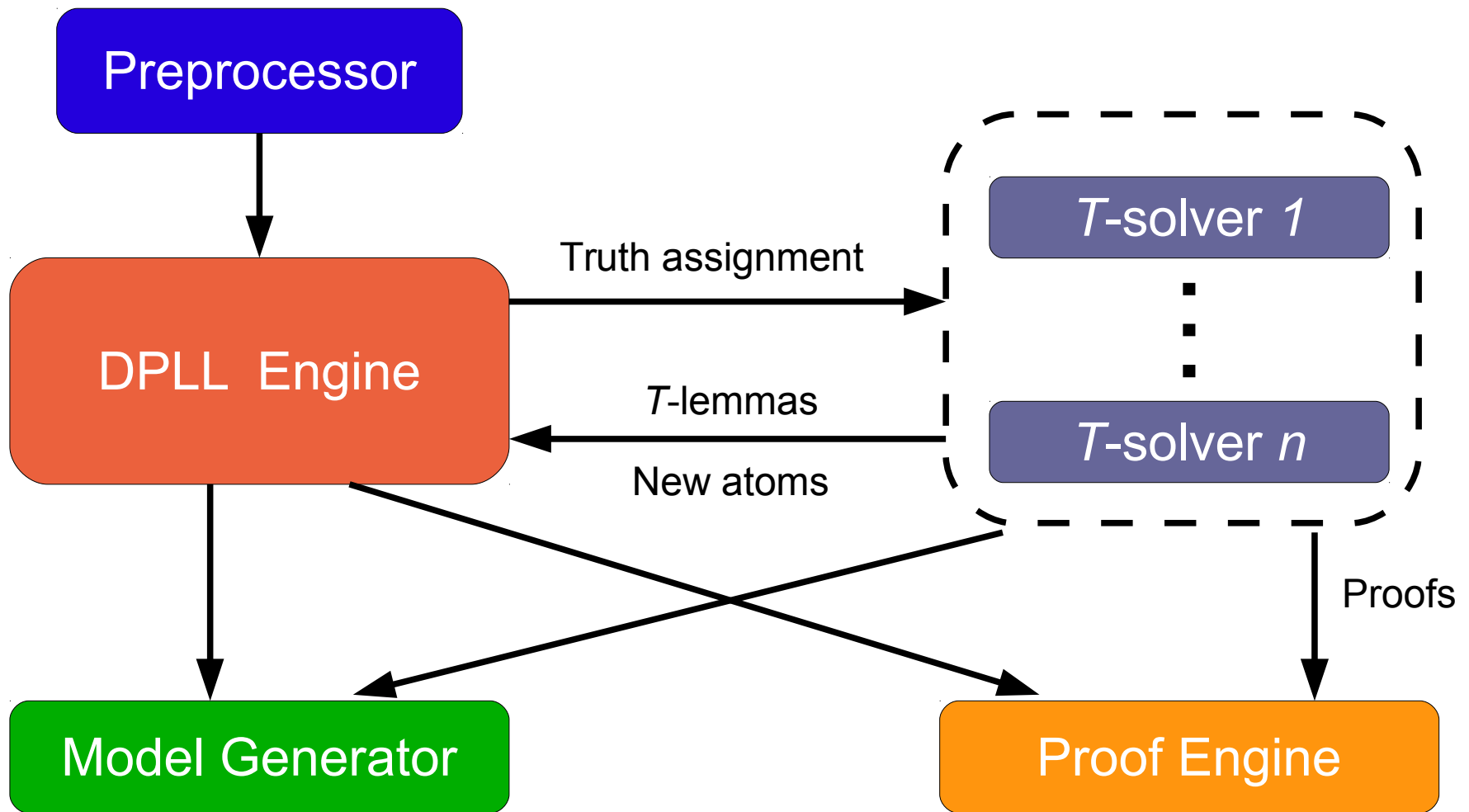
DPLL(T) example

```
def DPLL-T():
    while True:
        conflict = False
        if unit_propagation():
            res = T.check(!all_assigned())
            if res == False: conflict = True
            elif res == True: conflict = theory_propagation()
            elif learn_T_lemmas(): continue
            elif !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)
```

DPLL(T) example

```
def DPLL-T():  
    while True:  
        conflict = False  
        if unit_propagation():  
            call T.assume(lit)  
        res = T.check(!all_assigned())  
        call T.get_next_implied()  
        if res == False: conflict = True  
        elif res == True: conflict = theory_propagation()  
        elif learn_T_lemmas(): continue  
        elif !all_assigned(): decide()  
        call T.get_lemma()  
        else:  
            build_model()  
            call T.new_decision_level()  
            T.assume(lit)  
            return SAT  
        else: conflict = True  
        call T.get_value(e)  
        if conflict:  
            lvl, cls = conflict_analysis()  
            call T.get_conflict(c)  
            T.get_explanation(l, e)  
            if lvl < 0: return UNSAT  
            else:  
                backtrack(lvl)  
                call T.backtrack(lvl)  
                learn(cls)
```

An example lazy SMT architecture (MathSAT)



Outline



Introduction

CDCL-based SAT solvers

The DPLL(T) architecture

Some relevant T-solvers

Combination of theories

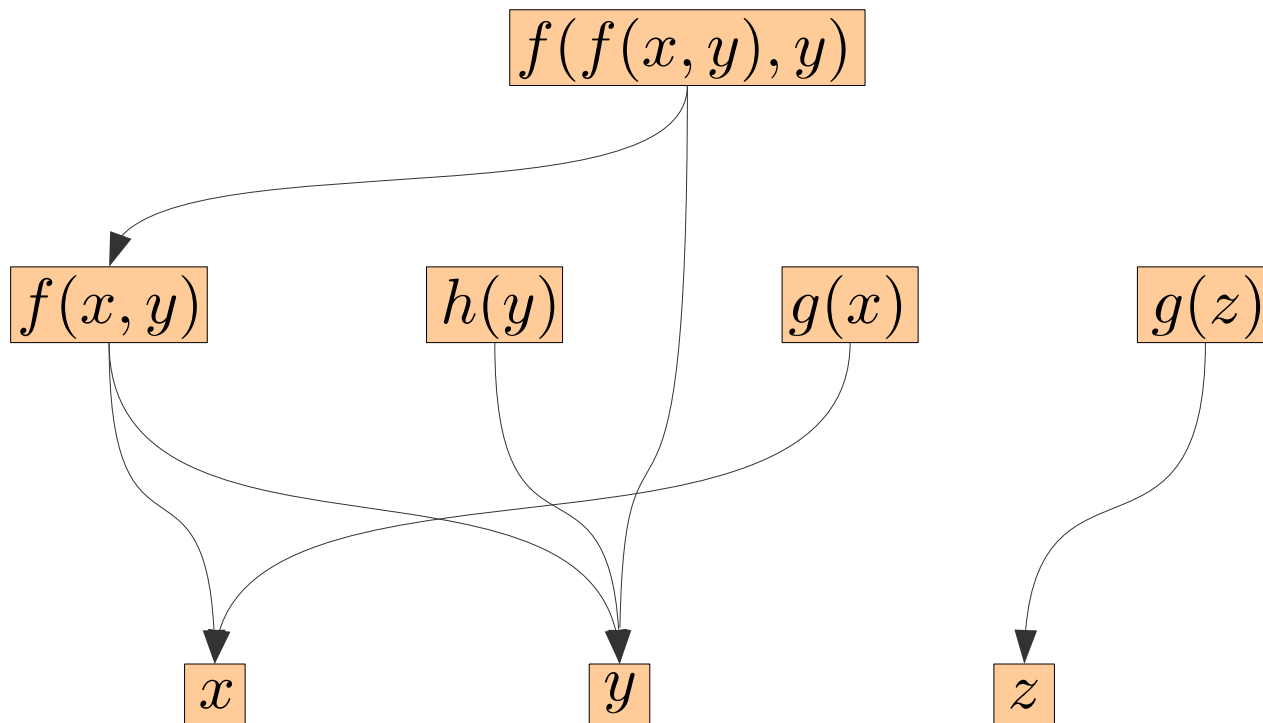
Equality (EUF)

- Polynomial time $O(n \log n)$ **congruence closure** procedure
- Fully incremental and backtrackable (stack-based)
- Supports **efficient T-propagation**
 - Exhaustive for positive equalities
 - Incomplete for disequalities
- Lazy explanations and conflict generation

- Typically used as a “**core**” T-solver
- Supports efficient extensions, e.g.
 - Integer offsets
 - Bit-vector slicing and concatenation

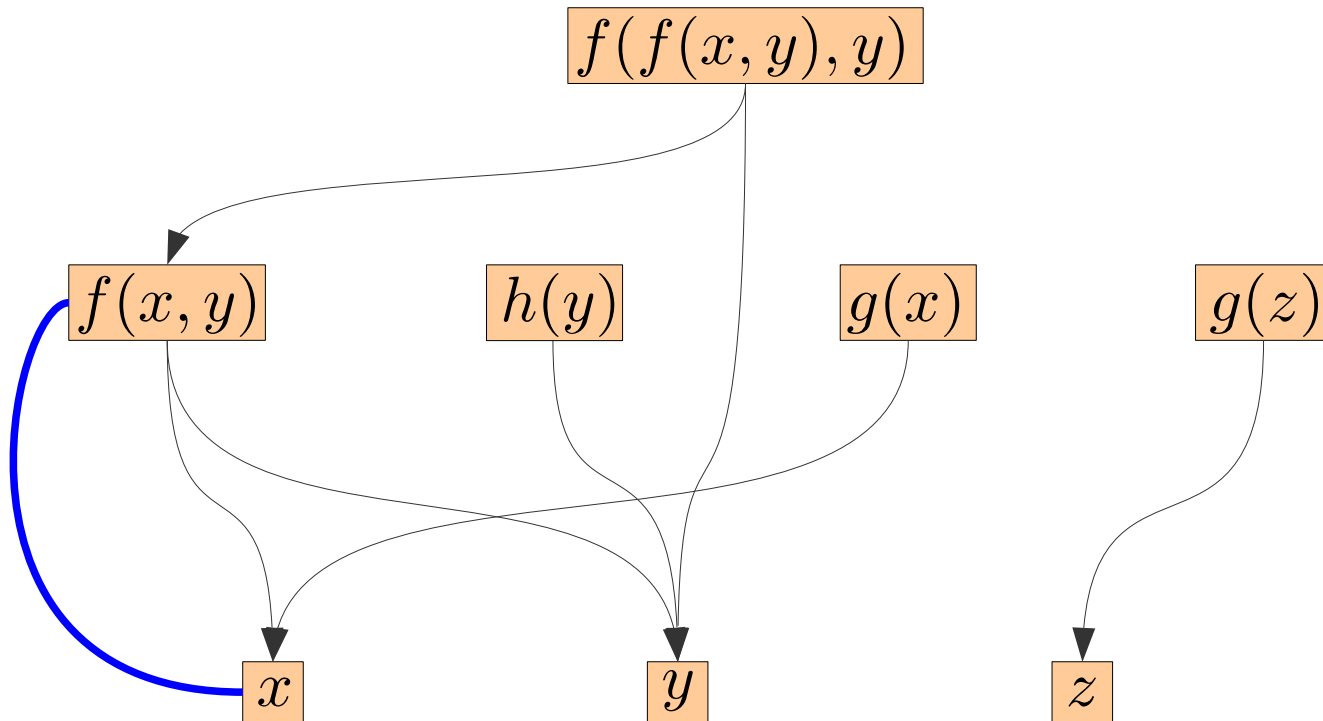
Example

$[(f(x, y) = x), (h(y) = g(x)), (f(f(x, y), y) = z), \neg(g(x) = g(z))]$



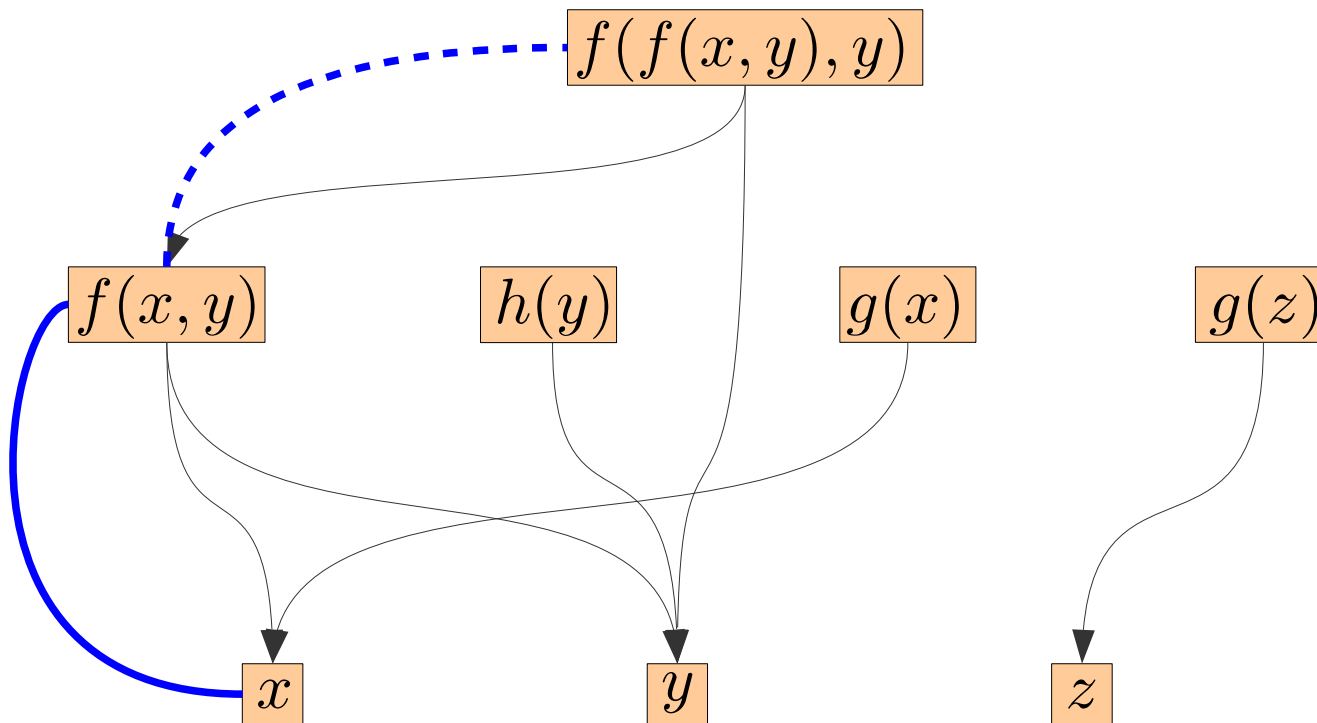
Example

$(f(x, y) = x), (h(y) = g(x)), (f(f(x, y), y) = z), \neg(g(x) = g(z))$



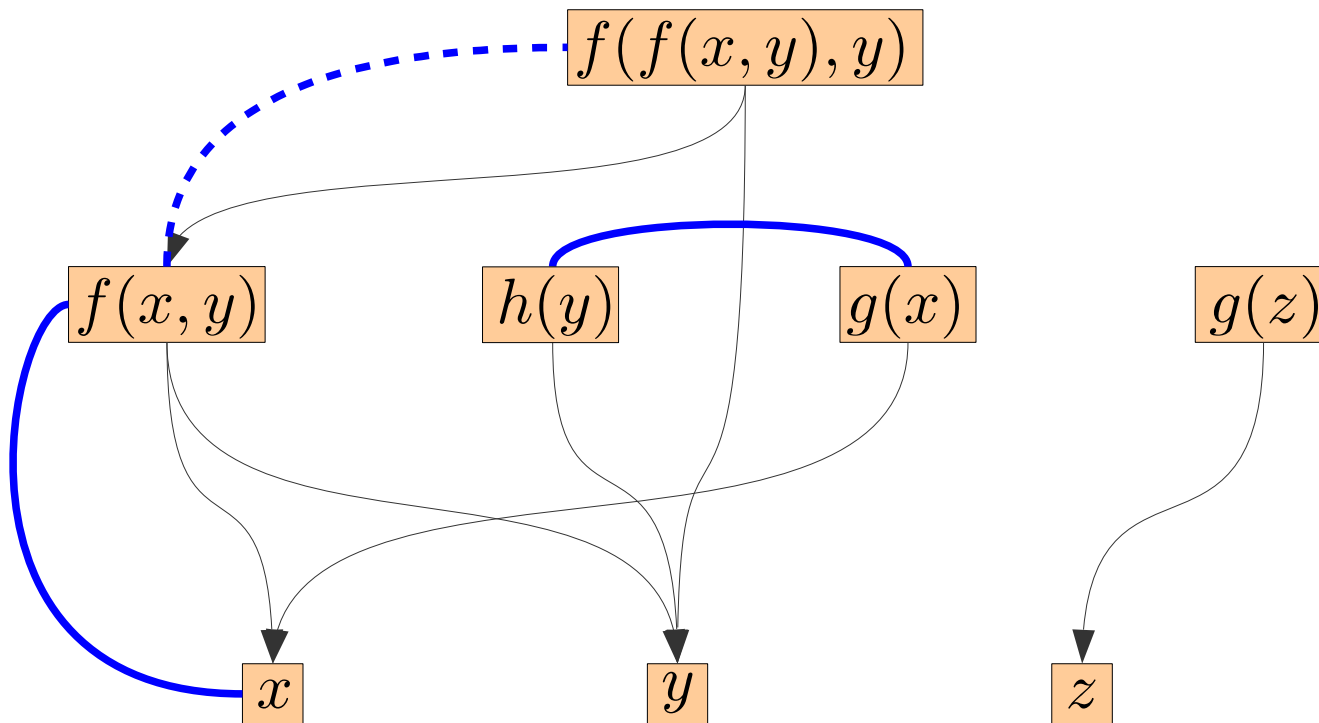
Example

$(f(x, y) = x), (h(y) = g(x)), (f(f(x, y), y) = z), \neg(g(x) = g(z))$



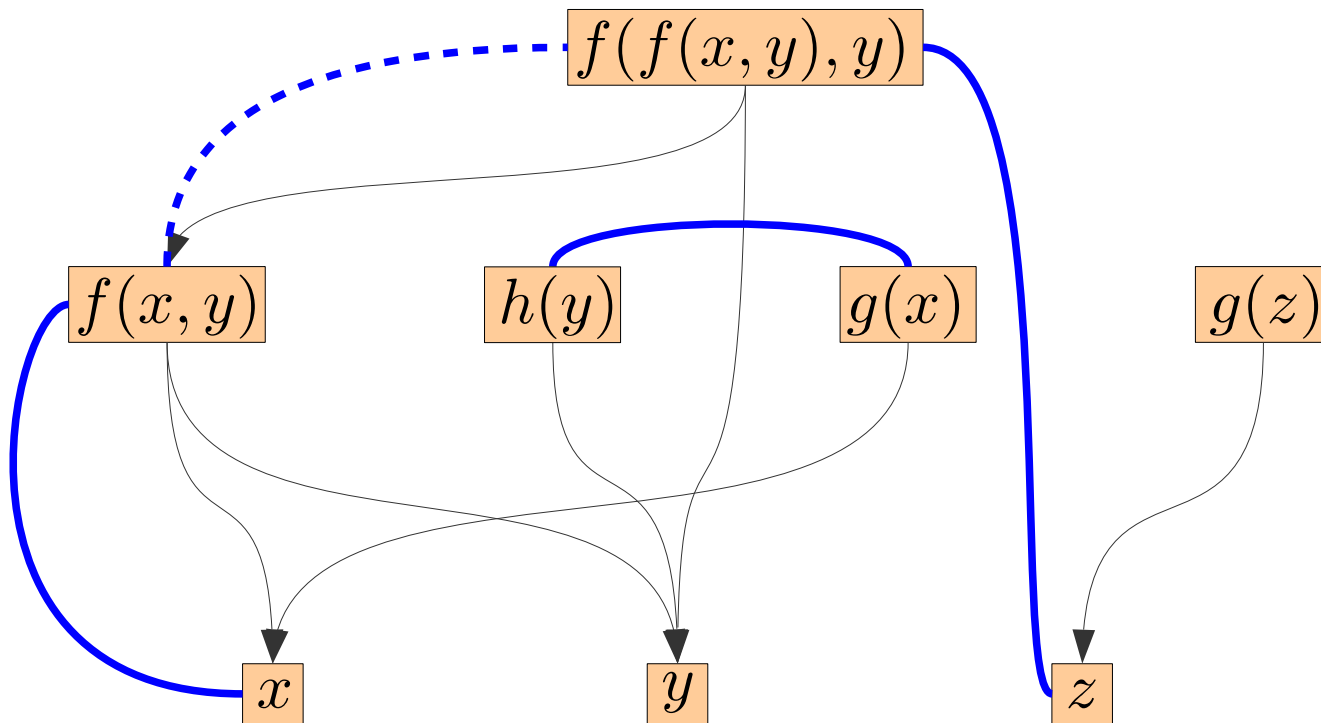
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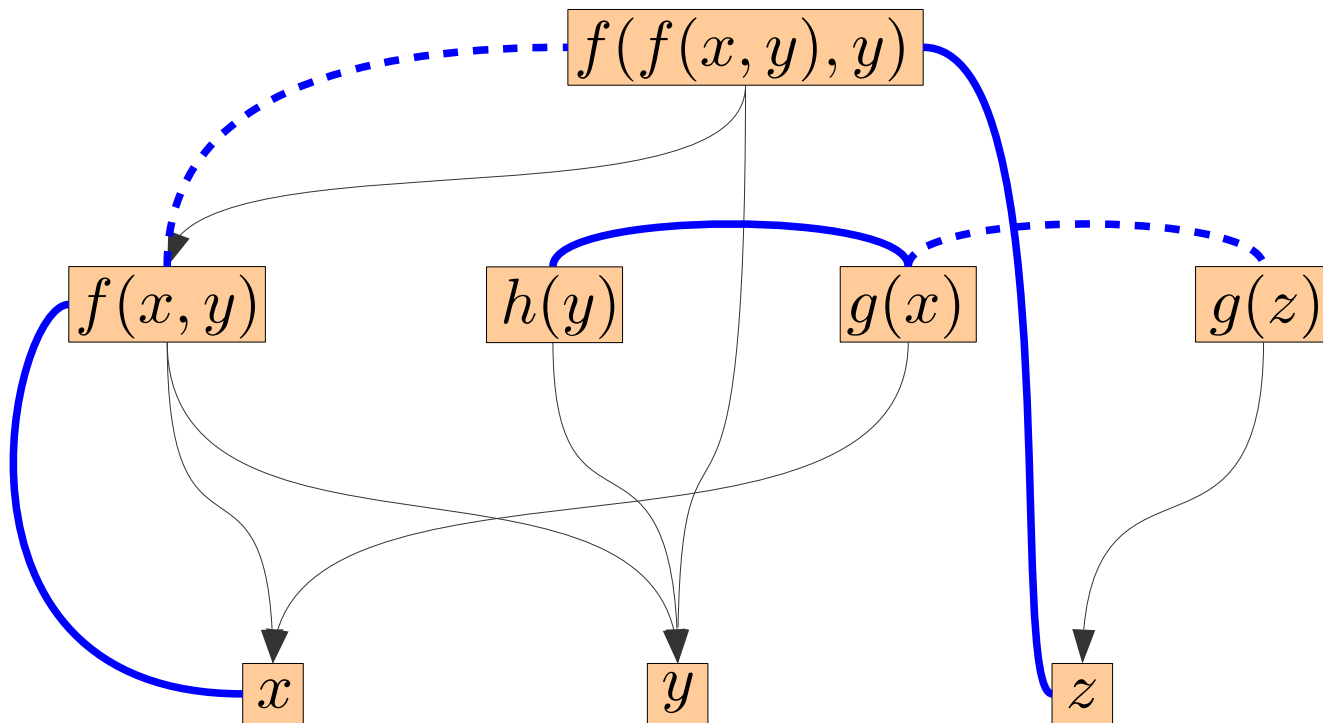
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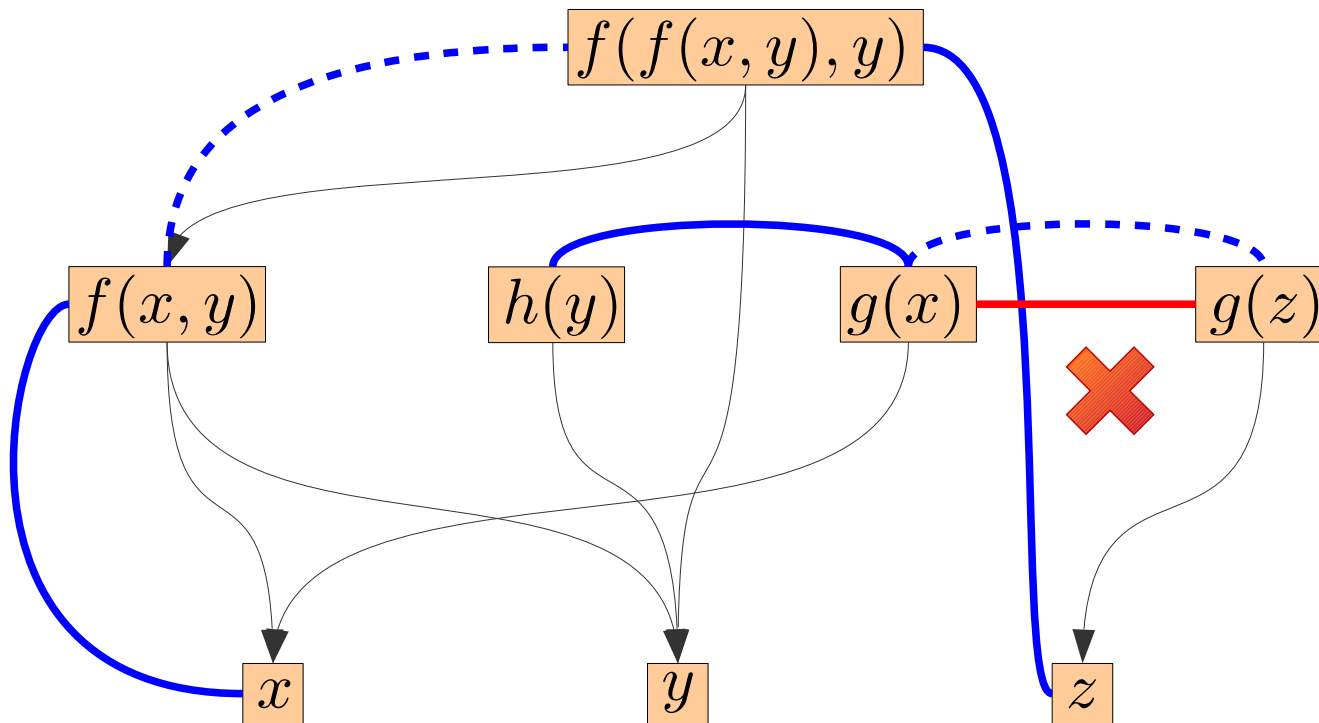
Example

$[(f(x, y) = x), (h(y) = g(x)), (f(f(x, y), y) = z), \neg(g(x) = g(z))]$



Example

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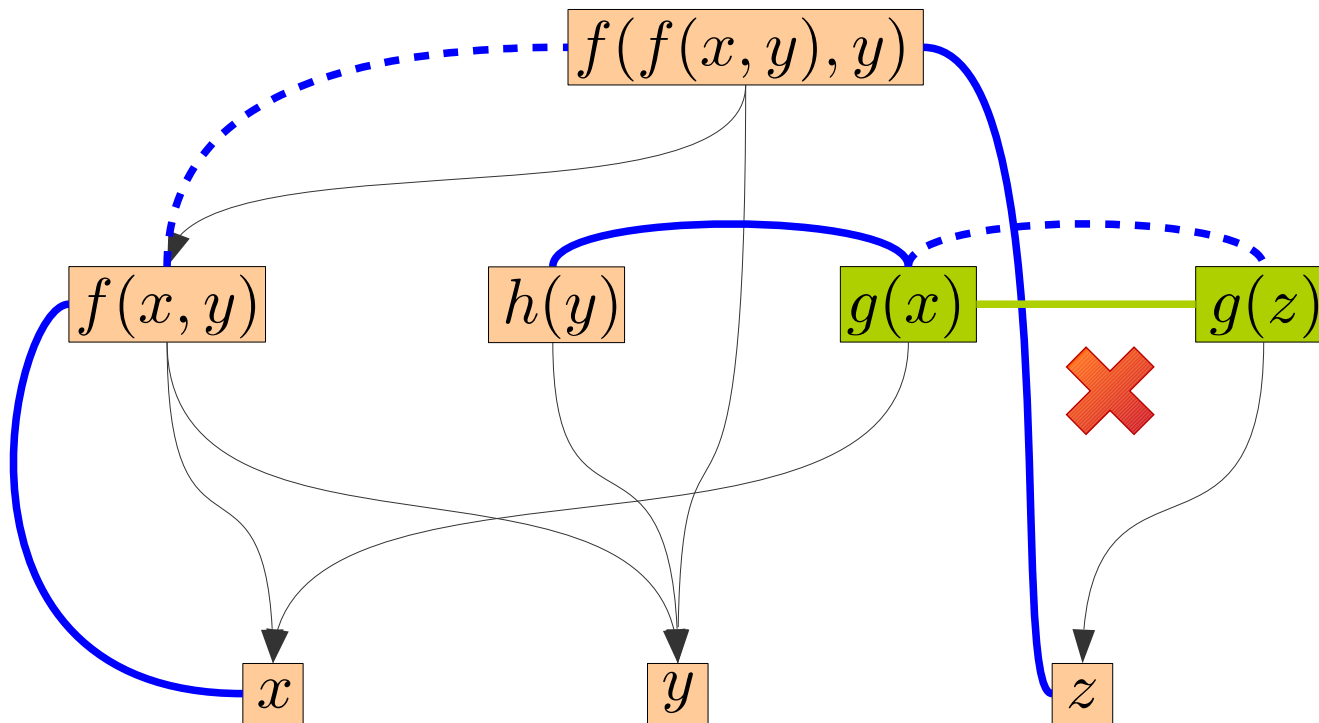


Example

$[(f(x, y) = x), (h(y) = g(x)), (f(f(x, y), y) = z), \boxed{\neg(g(x) = g(z))}]$

get_conflict():

$\neg(g(x) = g(z))$

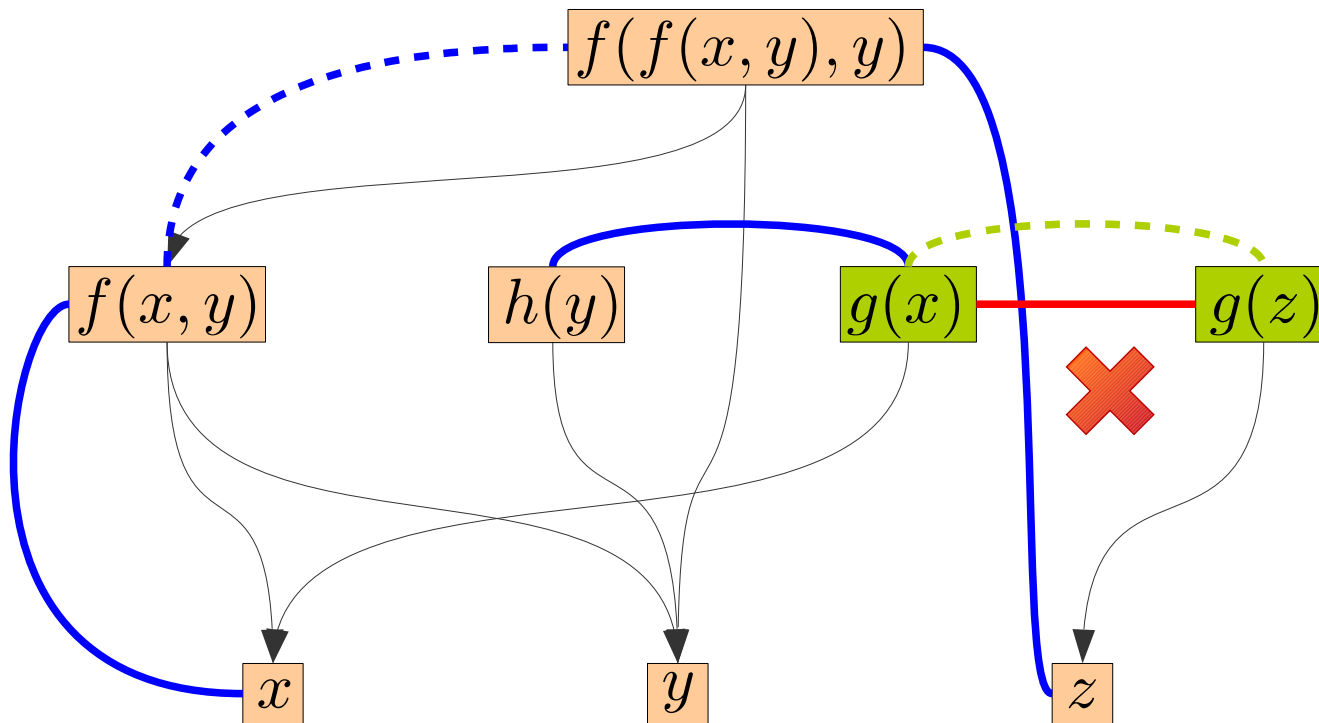


Example

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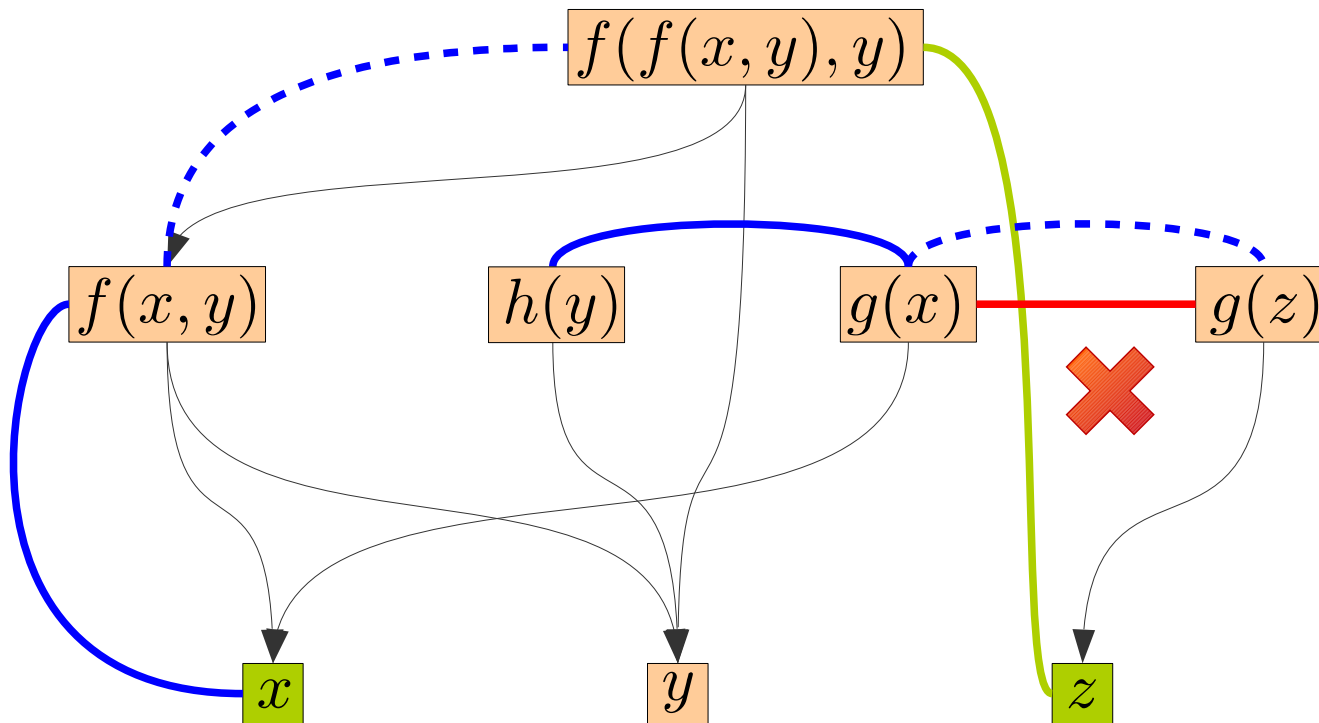


Example

$[(f(x, y) = x), (h(y) = g(x)), (f(f(x, y), y) = z), \boxed{\neg(g(x) = g(z))}]$

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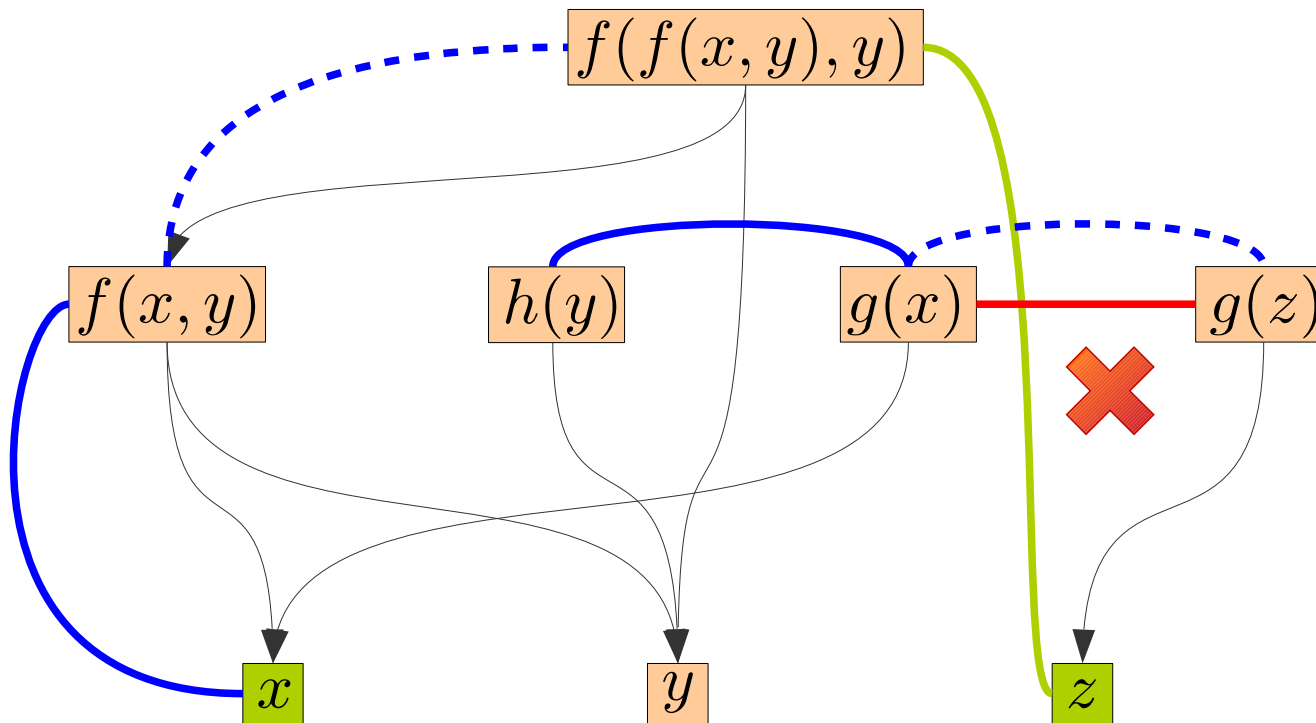
Example

$[(f(x, y) = x), (h(y) = g(x)), (f(f(x, y), y) = z), \boxed{\neg(g(x) = g(z))}]$

get_conflict():

$\neg(g(x) = g(z))$

$(f(f(x, y), y) = z)$



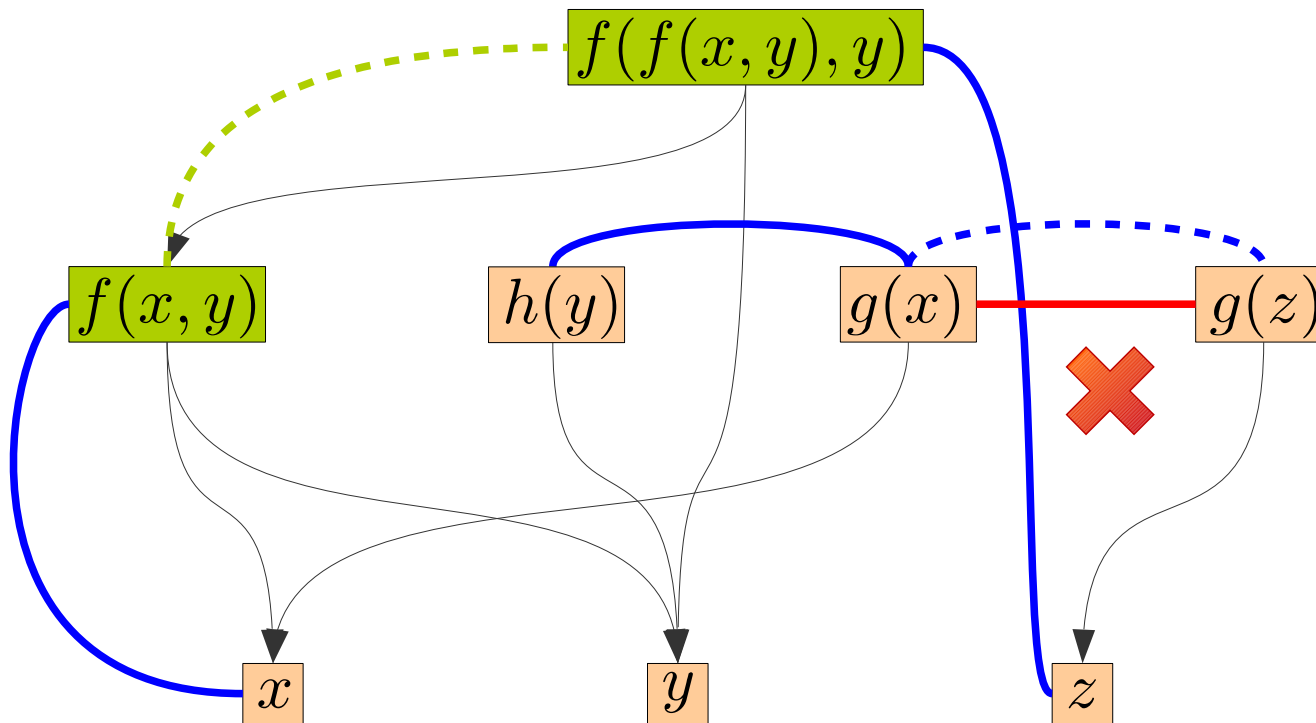
Example

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`get_conflict():`

$\neg(g(x) = g(z))$

$(f(f(x, y), y) = z)$



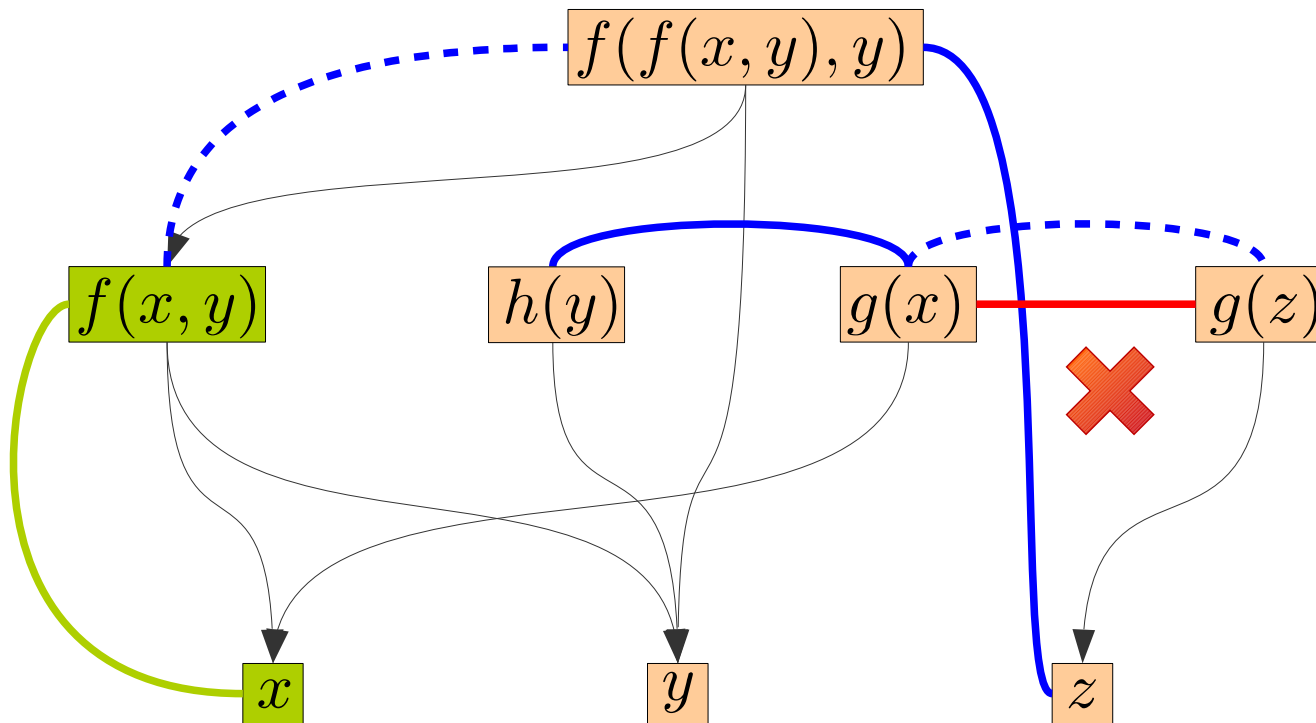
Example

$[(f(x, y) = x), (h(y) = g(x)), (f(f(x, y), y) = z), \boxed{\neg(g(x) = g(z))}]$

get_conflict():

$\neg(g(x) = g(z))$

$(f(f(x, y), y) = z)$



Example

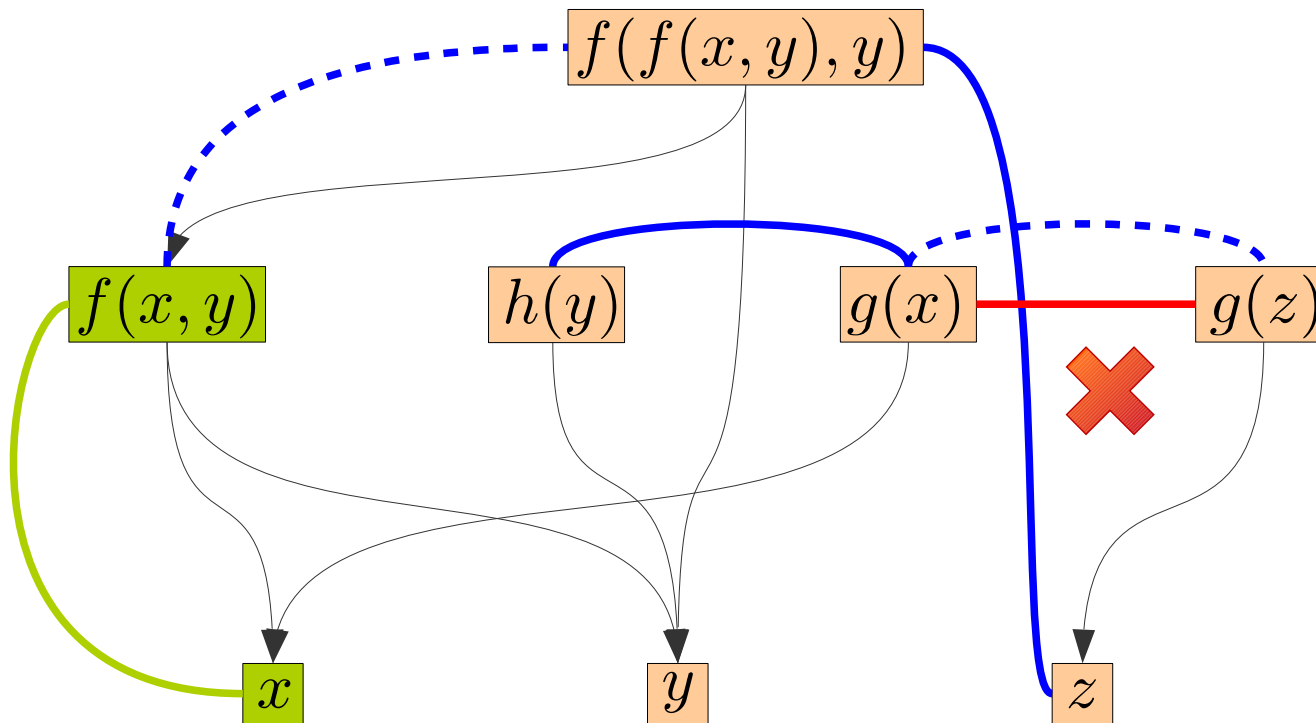
$[(f(x, y) = x), (h(y) = g(x)), (f(f(x, y), y) = z), \boxed{\neg(g(x) = g(z))}]$

`get_conflict():`

$\neg(g(x) = g(z))$

$(f(f(x, y), y) = z)$

$(f(x, y) = x)$



Example

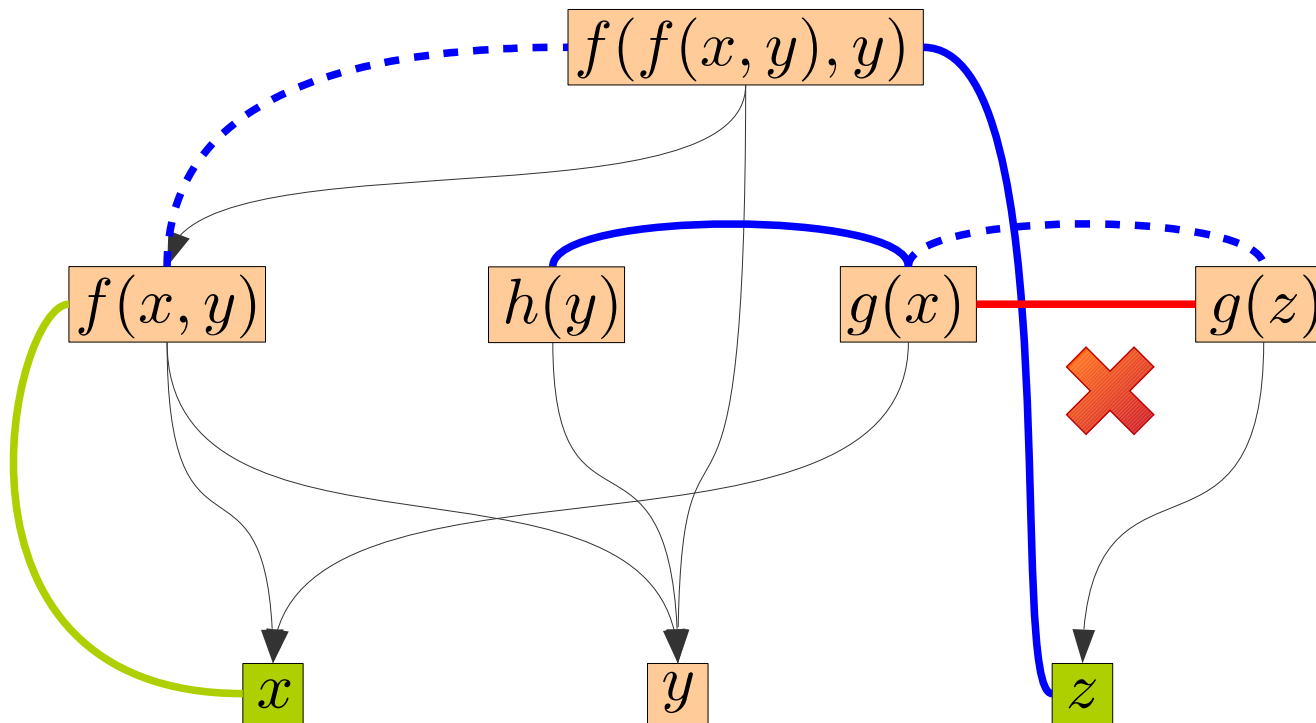
$[(f(x, y) = x), (h(y) = g(x)), (f(f(x, y), y) = z), \boxed{\neg(g(x) = g(z))}]$

get_conflict():

$\neg(g(x) = g(z))$

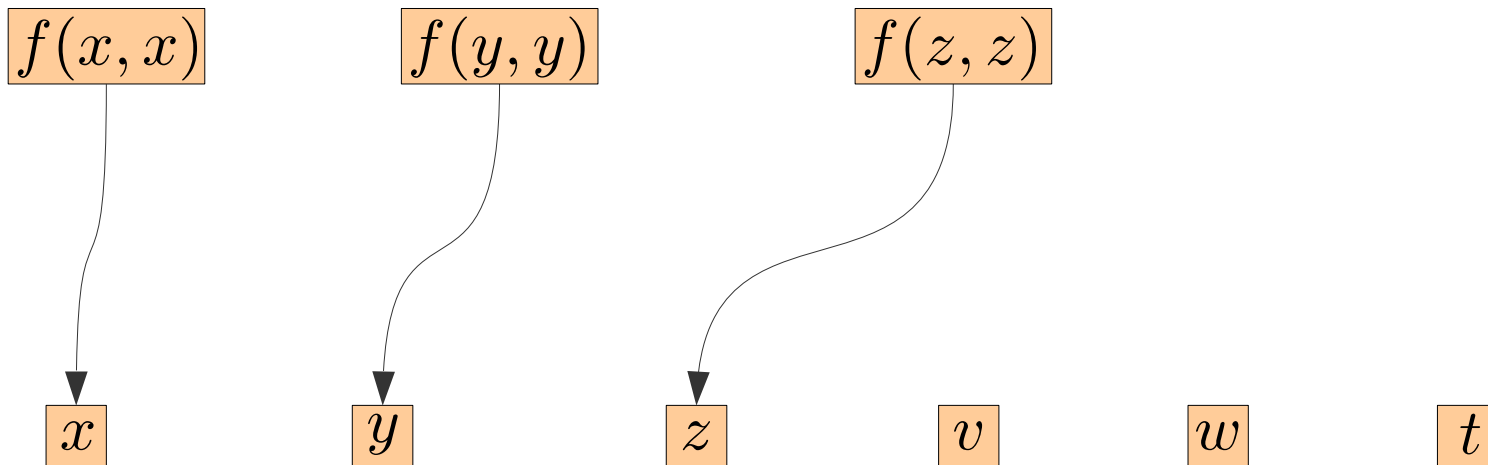
$(f(f(x, y), y) = z)$

$(f(x, y) = x)$



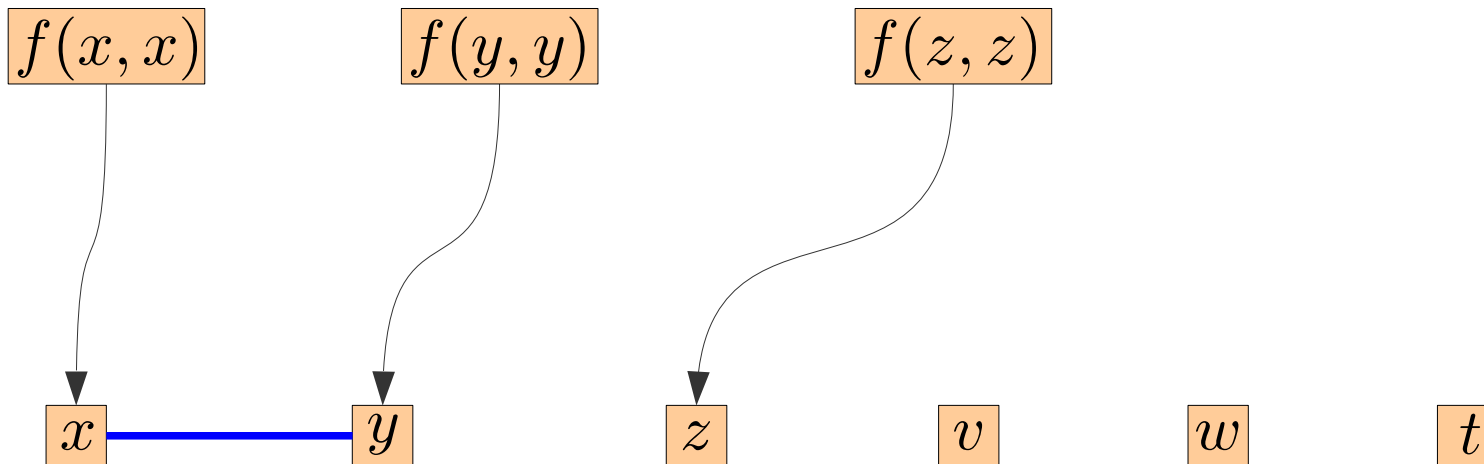
Example: redundant explanations

$[(x = y), (x = z), (f(x, x) = v), (f(y, y) = w), (f(z, z) = t), \neg(v = t)]$



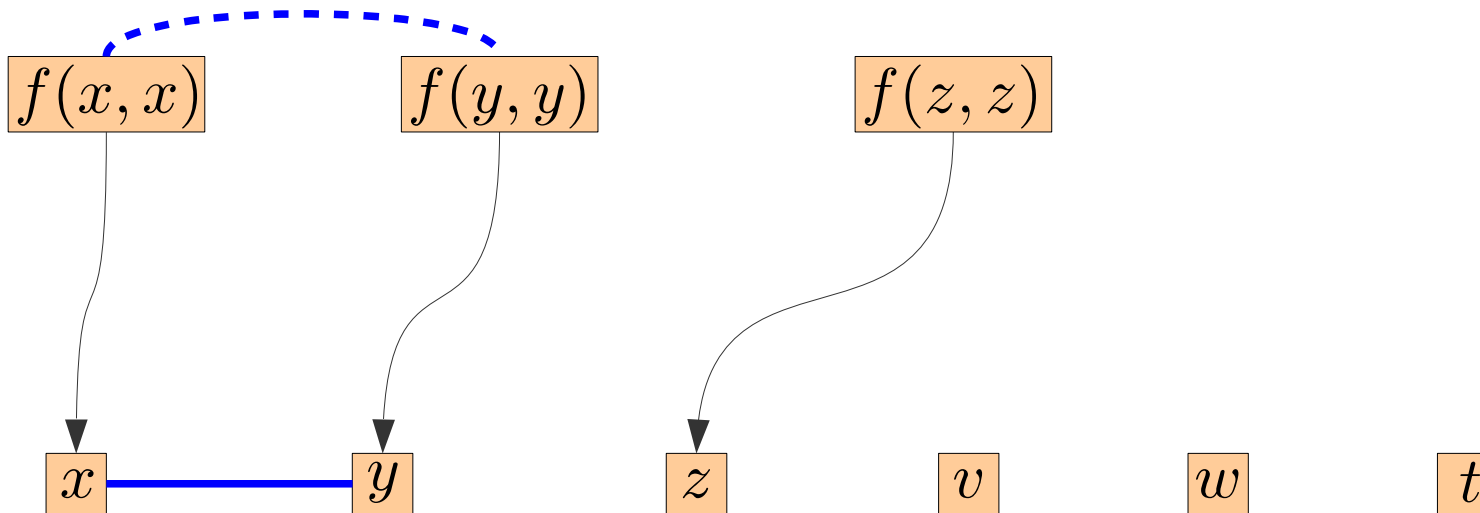
Example: redundant explanations

$(x = y), (x = z), (f(x, x) = v), (f(y, y) = w), (f(z, z) = t), \neg(v = t)$



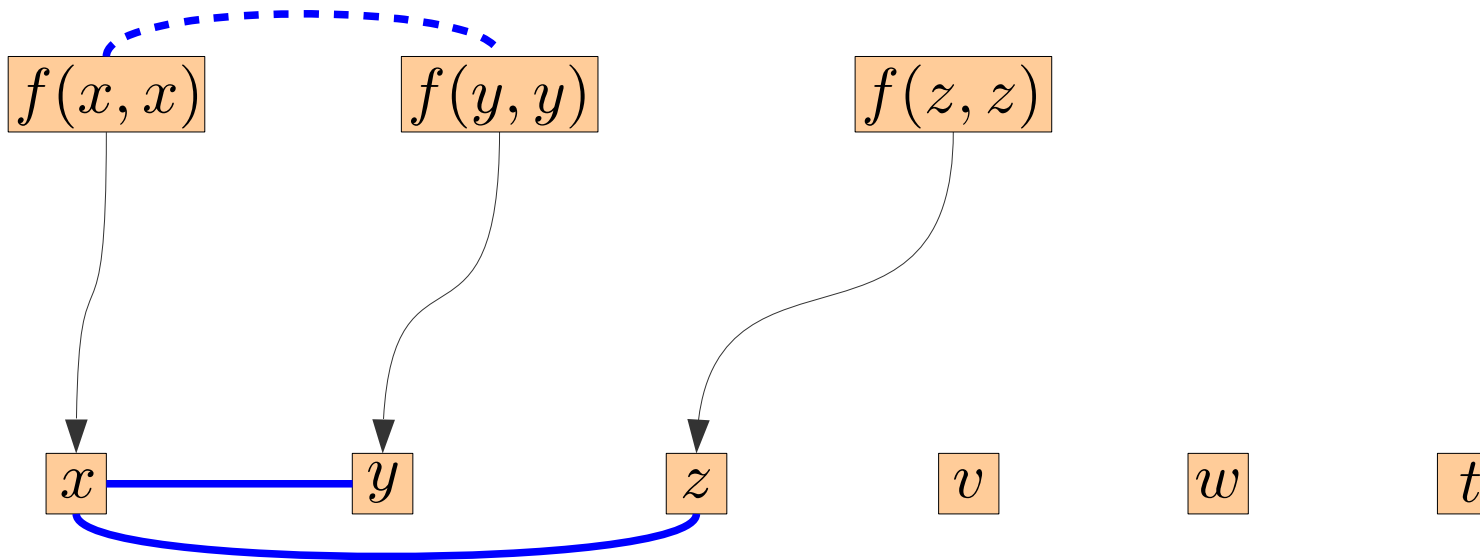
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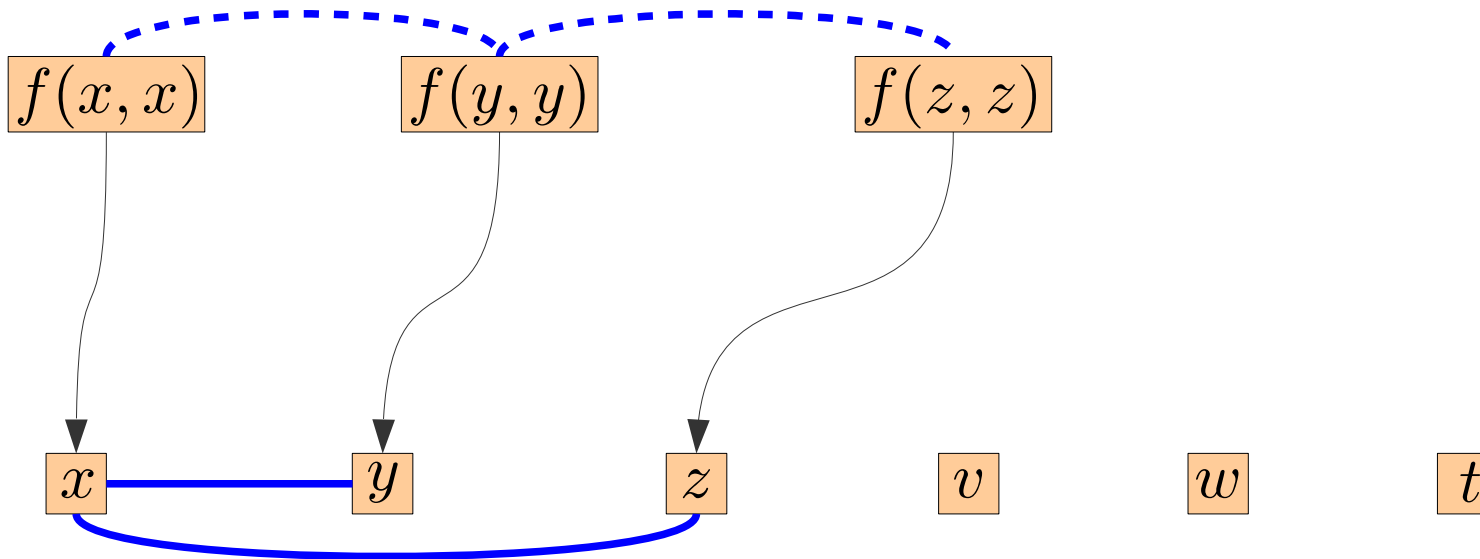
Example: redundant explanations

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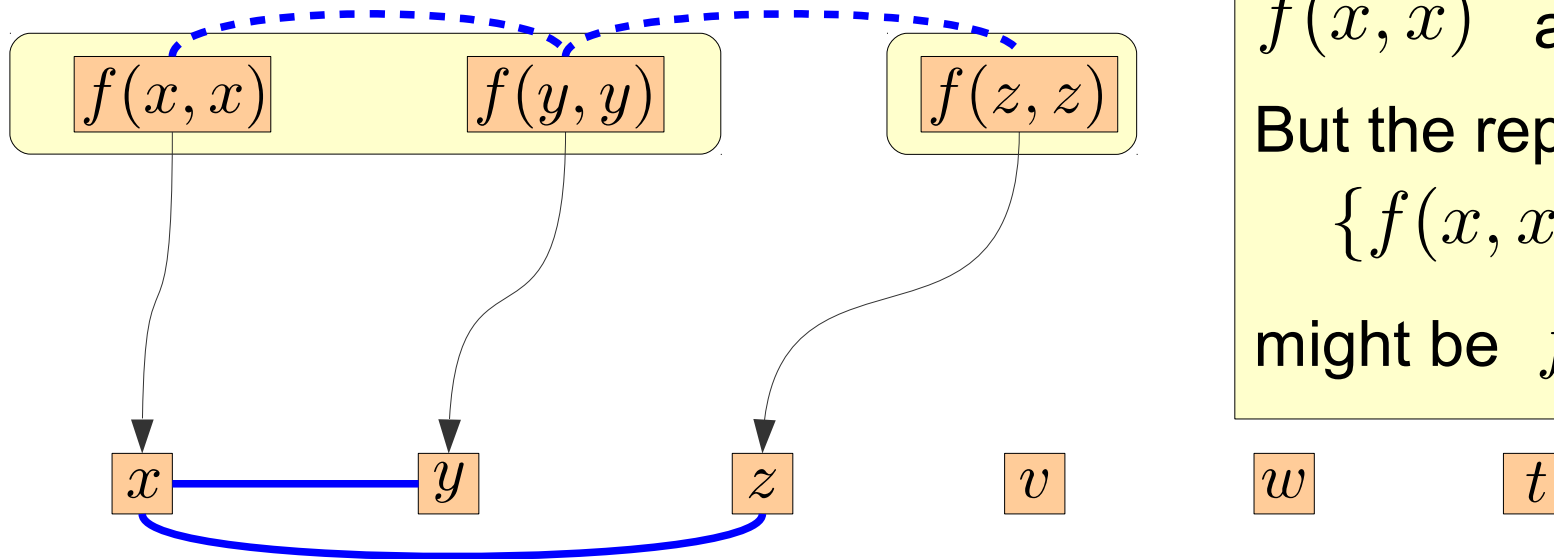
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Example: redundant explanations

$[(x = y), (x = z), (f(x, x) = v), (f(y, y) = w), (f(z, z) = t), \neg(v = t)]$

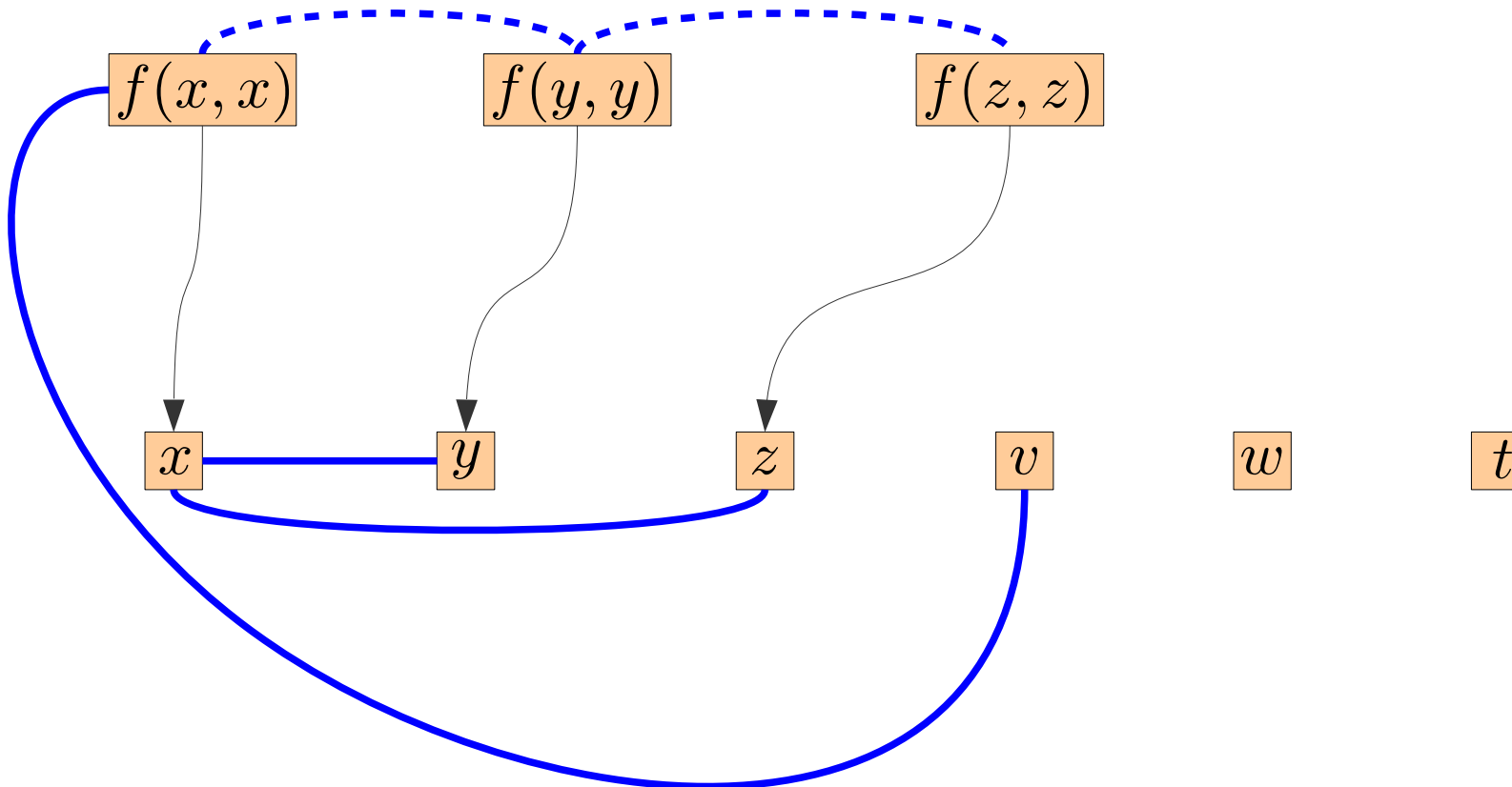


Here we connect the **equivalence classes** of $f(x, x)$ and $f(z, z)$

But the representative for $\{f(x, x), f(y, y)\}$ might be $f(y, y)$

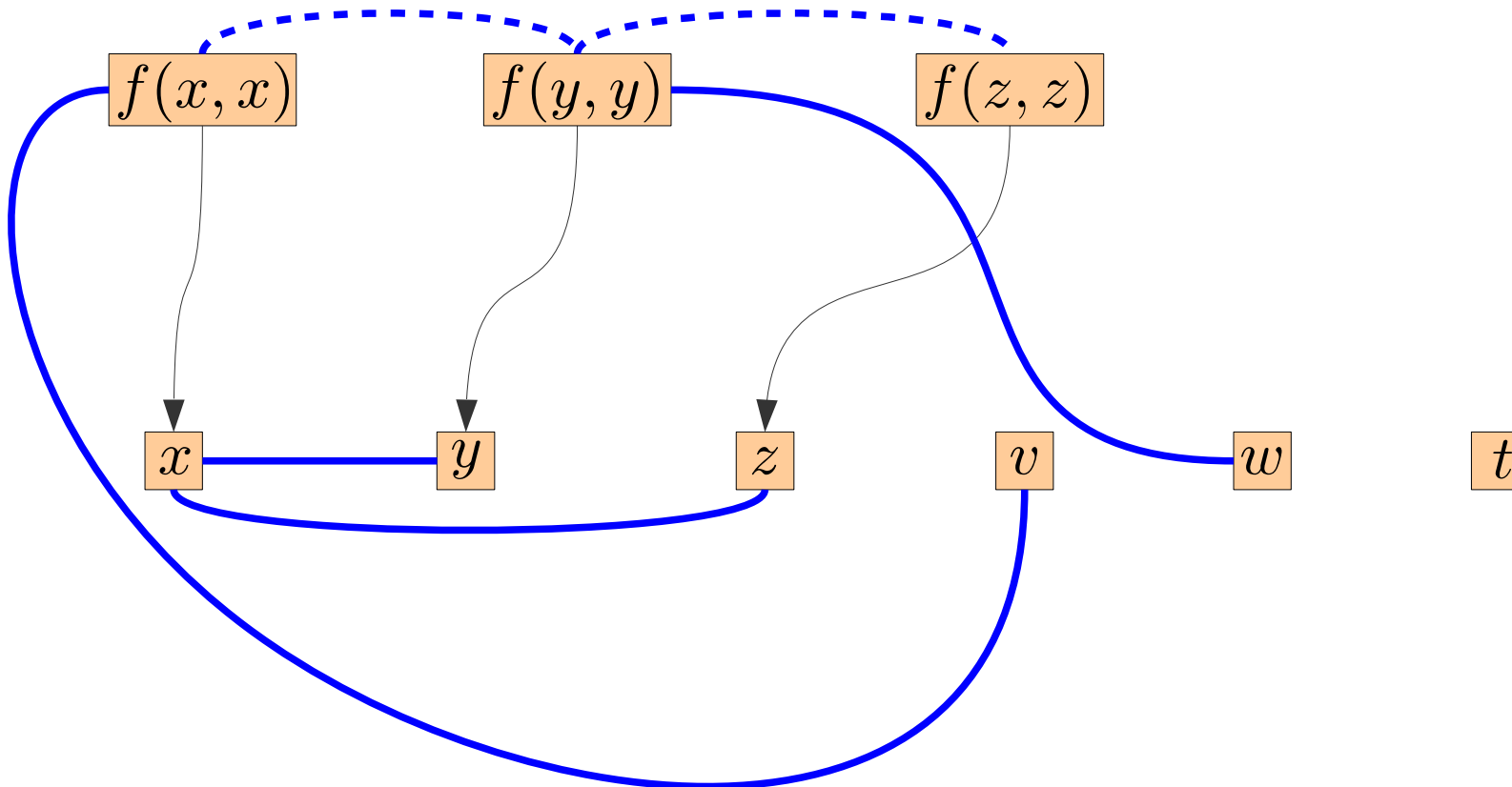
Example: redundant explanations

$[(x = y), (x = z), (f(x, x) = v), (f(y, y) = w), (f(z, z) = t), \neg(v = t)]$



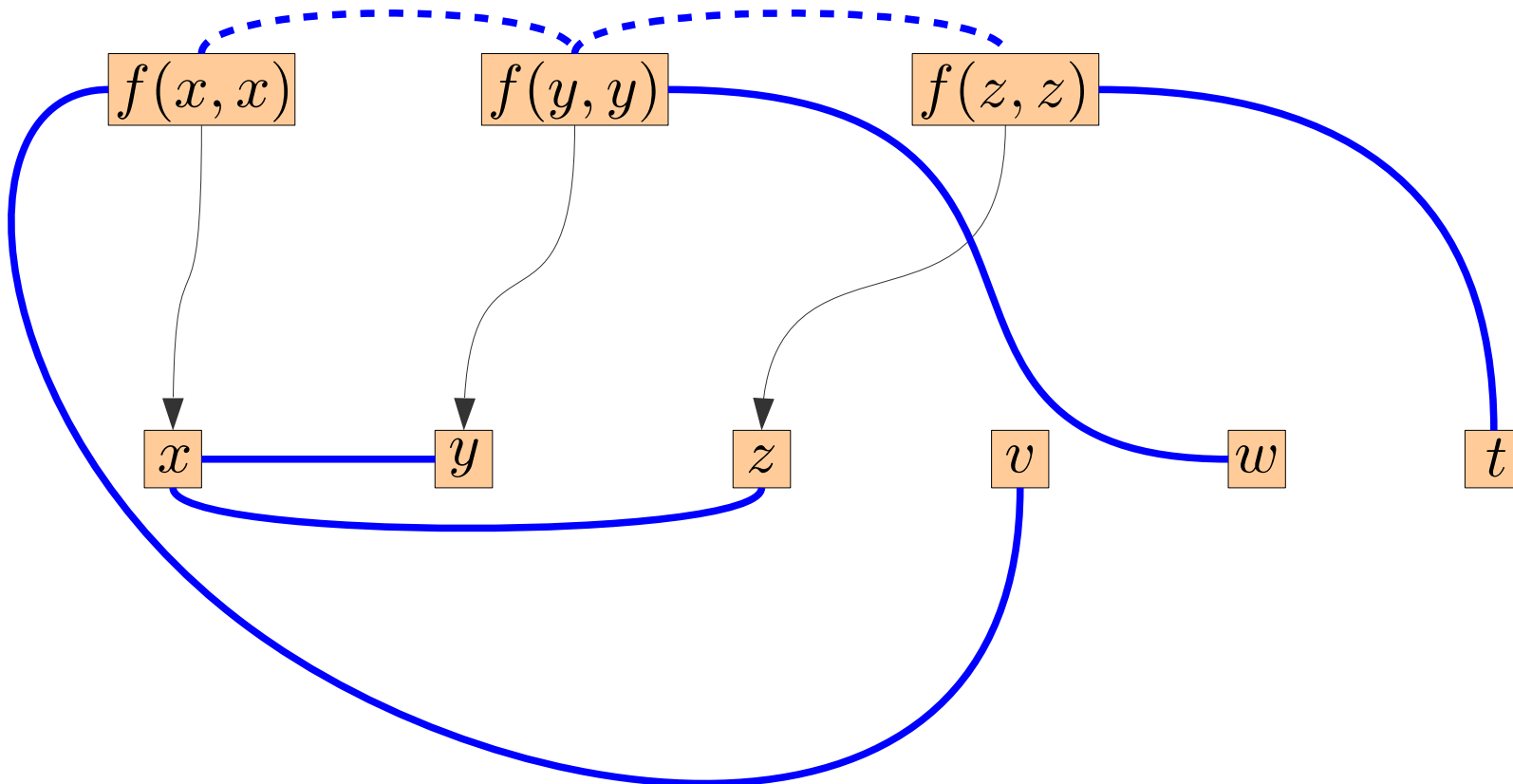
Example: redundant explanations

$[(x = y), (x = z), (f(x, x) = v), (f(y, y) = w), (f(z, z) = t), \neg(v = t)]$



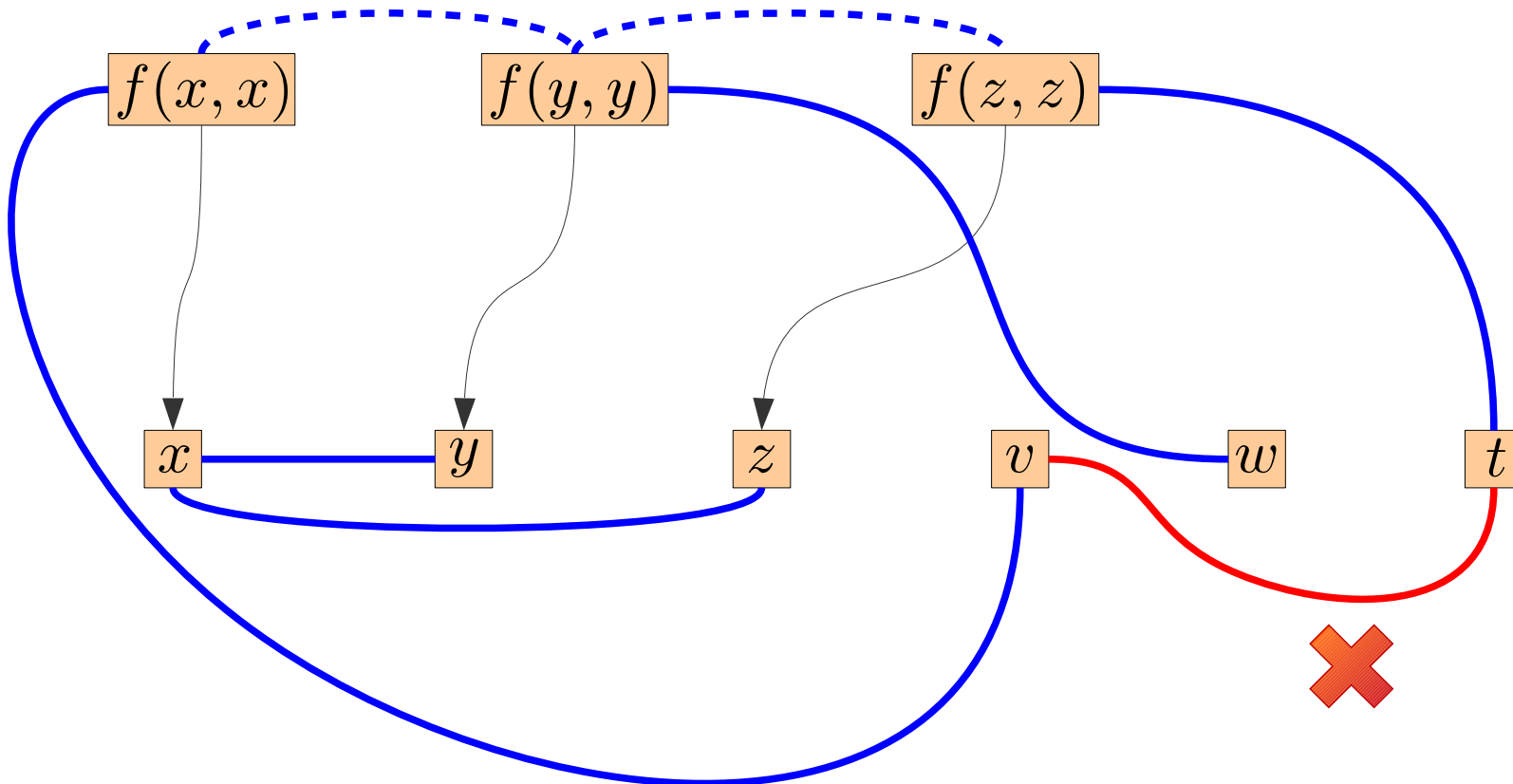
Example: redundant explanations

$$[(x = y), (x = z), (f(x, x) = v), (f(y, y) = w), (f(z, z) = t)] \neg(v = t)$$



Example: redundant explanations

$[(x = y), (x = z), (f(x, x) = v), (f(y, y) = w), (f(z, z) = t), \neg(v = t)]$



Linear Rational Arithmetic (LRA)

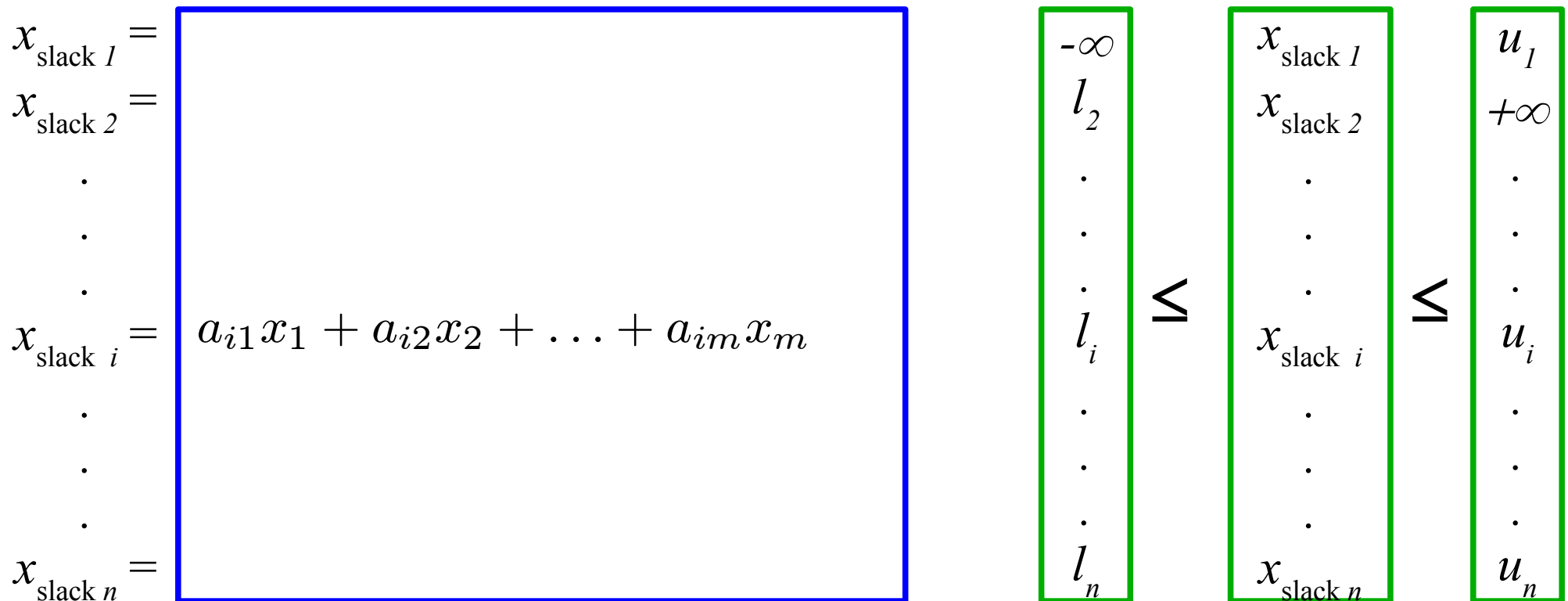
- Constraints of the form $\sum_i a_i x_i \leq c$
- Variant of **simplex** specifically designed for DPLL(T)
 - **Very efficient** backtracking
 - Incremental checks
 - **Cheap deduction** of unassigned literals
 - **Minimal explanations** generation
 - Can handle efficiently also strict inequalities
 - Rewrite $(t < 0)$ to $(t + \varepsilon \leq 0)$, treat ε symbolically
 - Worst-case exponential (although LRA is polynomial), but **fast in practice**

Simplex for DPLL(T)

Preprocessing: $\sum a_h x_h \leq u \mapsto x_{\text{slack}} = \sum a_h x_h \wedge x_{\text{slack}} \leq u$

Tableau of equations (fixed) + bounds (added/removed)

Candidate solution β always consistent with the tableau

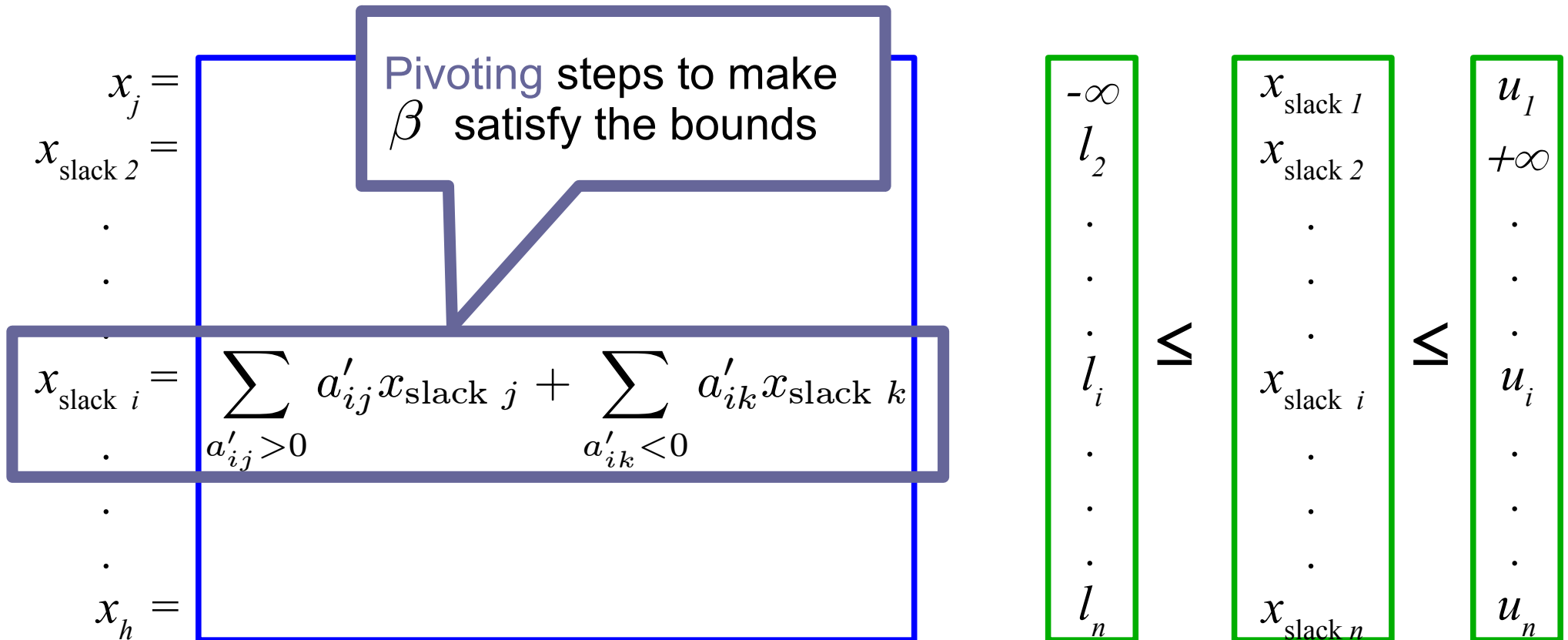


Simplex for DPLL(T)

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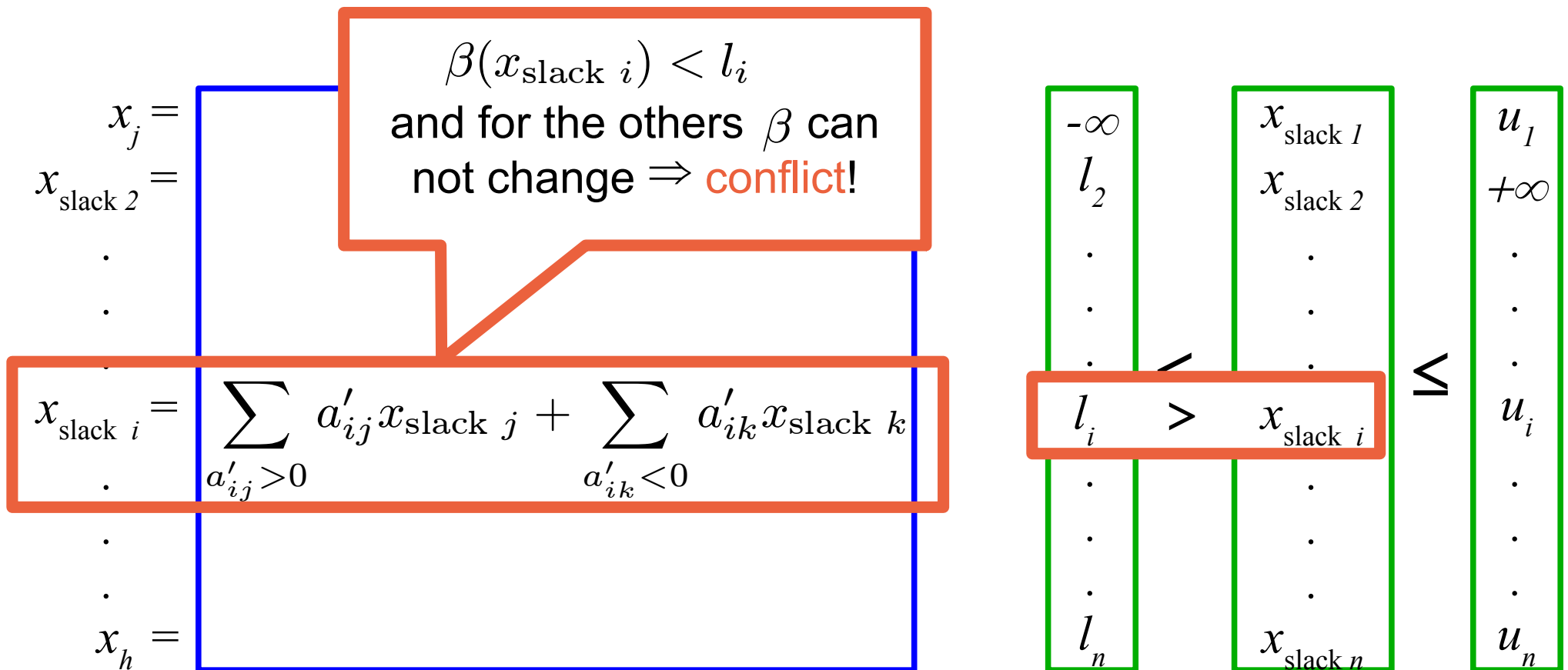


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Simplex for DPLL(T)

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Tableau of equations (fixed) + bounds (added/removed)

Candidate solution β always consistent with the tableau

$$x_j =$$

$$x_{\text{slack } 2} =$$

⋮

⋮

$$x_{\text{slack } i} = \sum_{a'_{ij} > 0} a'_{ij} x_{\text{slack } j} + \sum_{a'_{ik} < 0} a'_{ik} x_{\text{slack } k}$$

⋮

⋮

$$x_h =$$

$\beta(x_{\text{slack } i}) < l_i$
and for the others β can
not change \Rightarrow **conflict!**

`get_conflict():`

$\{(\sum a_h x_h \leq u)\}_j$
for $x_{\text{slack } j} \cup$

$\{(\sum a_h x_h \geq l)\}_k$
for $x_{\text{slack } k} \cup$

$\{(\sum a_h x_h \geq l)\}_i$
for $x_{\text{slack } i}$

Example

$$[(3x_2 - x_1 \leq 1), (x_1 + x_2 \geq 0), (x_3 - 2x_1 \geq 3), (2x_3 \leq 1)]$$

Example

$$\underbrace{[(3x_2 - x_1 \leq 1)]}_{s_1}, \underbrace{[(x_1 + x_2 \geq 0)]}_{s_2}, \underbrace{[(x_3 - 2x_1 \geq 3)]}_{s_3}, \underbrace{[(2x_3 \leq 1)]}_{s_4}$$

tableau

$$s_1 = 3x_2 - x_1$$

$$s_2 = x_1 + x_2$$

$$s_3 = x_3 - 2x_1$$

$$s_4 = 2x_3$$

bounds

$$-\infty \leq s_1 \leq 1$$

$$0 \leq s_2 \leq \infty$$

$$3 \leq s_3 \leq \infty$$

$$-\infty \leq s_4 \leq 1$$

candidate solution β

$$x_1 \mapsto 0$$

$$x_2 \mapsto 0$$

$$x_3 \mapsto 0$$

$$s_1 \mapsto 0$$

$$s_2 \mapsto 0$$

$$s_3 \mapsto 0$$

$$s_4 \mapsto 0$$

Example

$$\underbrace{[(3x_2 - x_1 \leq 1)]}_{s_1}, \underbrace{[(x_1 + x_2 \geq 0)]}_{s_2}, \underbrace{[(x_3 - 2x_1 \geq 3)]}_{s_3}, \underbrace{[(2x_3 \leq 1)]}_{s_4}$$

tableau

$$s_1 = 3x_2 - x_1$$

$$s_2 = x_1 + x_2$$

$$s_3 = x_3 - 2x_1$$

$$s_4 = 2x_3$$

bounds

$$-\infty \leq s_1 \leq 1$$

$$0 \leq s_2 \leq \infty$$

$$3 \leq s_3 \leq \infty$$

$$-\infty \leq s_4 \leq 1$$

candidate solution β

$$x_1 \mapsto 0$$

$$x_2 \mapsto 0$$

$$x_3 \mapsto 0$$

$$s_1 \mapsto 0$$

$$s_2 \mapsto 0$$

$$s_3 \mapsto 0$$

$$s_4 \mapsto 0$$

Find a bound violation

Example

$$\underbrace{[(3x_2 - x_1 \leq 1)]}_{s_1}, \underbrace{[(x_1 + x_2 \geq 0)]}_{s_2}, \underbrace{[(x_3 - 2x_1 \geq 3)]}_{s_3}, \underbrace{[(2x_3 \leq 1)]}_{s_4}$$

tableau

$$s_1 = 3x_2 - x_1$$

$$s_2 = x_1 + x_2$$

$$s_3 = x_3 - 2x_1$$

$$s_4 = 2x_3$$

bounds

$$-\infty \leq s_1 \leq 1$$

$$0 \leq s_2 \leq \infty$$

$$3 \leq s_3 \leq \infty$$

$$-\infty \leq s_4 \leq 1$$

candidate solution β

$$x_1 \mapsto 0$$

$$x_2 \mapsto 0$$

$$x_3 \mapsto 0$$

$$s_1 \mapsto 0$$

$$s_2 \mapsto 0$$

$$s_3 \mapsto 0$$

$$s_4 \mapsto 0$$

Pick a variable for pivoting

Example

$$\underbrace{[(3x_2 - x_1 \leq 1)]}_{s_1}, \underbrace{[(x_1 + x_2 \geq 0)]}_{s_2}, \underbrace{[(x_3 - 2x_1 \geq 3)]}_{s_3}, \underbrace{[(2x_3 \leq 1)]}_{s_4}$$

tableau

$$\begin{aligned} s_1 &= 3x_2 - \frac{1}{2}x_3 + \frac{1}{2}s_3 \\ s_2 &= \frac{1}{2}x_3 - \frac{1}{2}s_3 + x_2 \\ x_1 &= \frac{1}{2}x_3 - \frac{1}{2}s_3 \\ s_4 &= 2x_3 \end{aligned}$$

bounds

$$\begin{aligned} -\infty &\leq s_1 \leq 1 \\ 0 &\leq s_2 \leq \infty \\ 3 &\leq s_3 \leq \infty \\ -\infty &\leq s_4 \leq 1 \end{aligned}$$

candidate solution β

$$\begin{array}{l} x_1 \mapsto -\frac{3}{2} \\ x_2 \mapsto 0 \\ x_3 \mapsto 0 \\ s_1 \mapsto \frac{3}{2} \\ s_2 \mapsto -\frac{3}{2} \\ s_3 \mapsto 3 \\ s_4 \mapsto 0 \end{array}$$

Pivot and update β

Example

$$\underbrace{[(3x_2 - x_1 \leq 1)]}_{s_1}, \underbrace{[(x_1 + x_2 \geq 0)]}_{s_2}, \underbrace{[(x_3 - 2x_1 \geq 3)]}_{s_3}, \underbrace{[(2x_3 \leq 1)]}_{s_4}$$

tableau

$$\begin{aligned} s_1 &= 3x_2 - \frac{1}{2}x_3 + \frac{1}{2}s_3 \\ s_2 &= \frac{1}{2}x_3 - \frac{1}{2}s_3 + x_2 \\ x_1 &= \frac{1}{2}x_3 - \frac{1}{2}s_3 \\ s_4 &= 2x_3 \end{aligned}$$

bounds

$$\begin{aligned} -\infty &\leq s_1 \leq 1 \\ 0 &\leq s_2 \leq \infty \\ 3 &\leq s_3 \leq \infty \\ -\infty &\leq s_4 \leq 1 \end{aligned}$$

candidate solution β

$$\begin{aligned} x_1 &\mapsto -\frac{3}{2} \\ x_2 &\mapsto 0 \\ x_3 &\mapsto 0 \\ s_1 &\mapsto \frac{3}{2} \\ s_2 &\mapsto -\frac{3}{2} \\ s_3 &\mapsto 3 \\ s_4 &\mapsto 0 \end{aligned}$$

Find a bound violation

Example

$$[(3x_2 - x_1 \leq 1), (x_1 + x_2 \geq 0), (x_3 - 2x_1 \geq 3), (2x_3 \leq 1)]$$

s_1 s_2 s_3 s_4

tableau

$$\begin{aligned}
 s_1 &= 3x_2 - \frac{1}{2}x_3 + \frac{1}{2}s_3 \\
 s_2 &= \frac{1}{2}x_3 - \frac{1}{2}s_3 + x_2 \\
 x_1 &= \frac{1}{2}x_3 - \frac{1}{2}s_3 \\
 s_4 &= 2x_3
 \end{aligned}$$

bounds

$$\begin{aligned}
 -\infty &\leq s_1 \leq 1 \\
 0 &\leq s_2 \leq \infty \\
 3 &\leq s_3 \leq \infty \\
 -\infty &\leq s_4 \leq 1
 \end{aligned}$$

candidate solution β

$$\begin{array}{lcl}
 x_1 & \mapsto & -\frac{3}{2} \\
 x_2 & \mapsto & 0 \\
 x_3 & \mapsto & 0 \\
 s_1 & \mapsto & \frac{3}{2} \\
 s_2 & \mapsto & -\frac{3}{2} \\
 s_3 & \mapsto & 3 \\
 s_4 & \mapsto & 0
 \end{array}$$

Pick a variable for pivoting

Example

$$\underbrace{[(3x_2 - x_1 \leq 1)]}_{s_1}, \underbrace{[(x_1 + x_2 \geq 0)]}_{s_2}, \underbrace{[(x_3 - 2x_1 \geq 3)]}_{s_3}, \underbrace{[(2x_3 \leq 1)]}_{s_4}$$

tableau

$$\begin{aligned} s_1 &= 3s_2 - 2x_3 + 2s_3 \\ x_2 &= s_2 - \frac{1}{2}x_3 + \frac{1}{2}s_3 \\ x_1 &= \frac{1}{2}x_3 - \frac{1}{2}s_3 \\ s_4 &= 2x_3 \end{aligned}$$

bounds

$$\begin{aligned} -\infty &\leq s_1 \leq 1 \\ 0 &\leq s_2 \leq \infty \\ 3 &\leq s_3 \leq \infty \\ -\infty &\leq s_4 \leq 1 \end{aligned}$$

candidate solution β

$$\begin{aligned} x_1 &\mapsto -\frac{3}{2} \\ x_2 &\mapsto \frac{3}{2} \\ x_3 &\mapsto 0 \\ s_1 &\mapsto 6 \\ s_2 &\mapsto 0 \\ s_3 &\mapsto 3 \\ s_4 &\mapsto 0 \end{aligned}$$

Pivot and update β

Example

$$\underbrace{[(3x_2 - x_1 \leq 1)]}_{s_1}, \underbrace{[(x_1 + x_2 \geq 0)]}_{s_2}, \underbrace{[(x_3 - 2x_1 \geq 3)]}_{s_3}, \underbrace{[(2x_3 \leq 1)]}_{s_4}$$

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Find a bound violation

Example

$$\underbrace{[(3x_2 - x_1 \leq 1)]}_{s_1}, \underbrace{[(x_1 + x_2 \geq 0)]}_{s_2}, \underbrace{[(x_3 - 2x_1 \geq 3)]}_{s_3}, \underbrace{[(2x_3 \leq 1)]}_{s_4}$$

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Pick a variable for pivoting

Example

$$\underbrace{[(3x_2 - x_1 \leq 1)]}_{s_1}, \underbrace{[(x_1 + x_2 \geq 0)]}_{s_2}, \underbrace{[(x_3 - 2x_1 \geq 3)]}_{s_3}, \underbrace{[(2x_3 \leq 1)]}_{s_4}$$

tableau

$$\begin{aligned}x_3 &= -\frac{1}{2}s_1 + \frac{3}{2}s_2 + s_3 \\x_2 &= \frac{1}{4}s_1 + \frac{1}{4}s_2 \\x_1 &= -\frac{1}{4}s_1 + \frac{3}{4}s_2 \\s_4 &= -s_1 + 3s_2 + 2s_3\end{aligned}$$

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candidate solution β

$$\begin{aligned}x_1 &\mapsto -\frac{1}{4} \\x_2 &\mapsto \frac{1}{4} \\x_3 &\mapsto \frac{5}{2} \\s_1 &\mapsto 1 \\s_2 &\mapsto 0 \\s_3 &\mapsto 3 \\s_4 &\mapsto 5\end{aligned}$$

Pivot and update β

Example

$$\underbrace{[(3x_2 - x_1 \leq 1)]}_{s_1}, \underbrace{[(x_1 + x_2 \geq 0)]}_{s_2}, \underbrace{[(x_3 - 2x_1 \geq 3)]}_{s_3}, \underbrace{[(2x_3 \leq 1)]}_{s_4}$$

tableau

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candidate solution β

$$\begin{array}{l}x_1 \mapsto -\frac{1}{4} \\x_2 \mapsto \frac{1}{4} \\x_3 \mapsto \frac{5}{2} \\s_1 \mapsto 1 \\s_2 \mapsto 0 \\s_3 \mapsto 3 \\s_4 \mapsto 5\end{array}$$

No suitable variable for pivoting!
Conflict

Example

$$[(3x_2 - x_1 \leq 1), (x_1 + x_2 \geq 0), (x_3 - 2x_1 \geq 3), (2x_3 \leq 1)]$$

s_1 s_2 s_3 s_4

tableau

$$\begin{aligned}
 x_3 &= -\frac{1}{2}s_1 + \frac{3}{2}s_2 + s_3 \\
 x_2 &= \frac{1}{4}s_1 + \frac{1}{4}s_2 \\
 x_1 &= -\frac{1}{4}s_1 + \frac{3}{4}s_2 \\
 s_4 &= -s_1 + 3s_2 + 2s_3
 \end{aligned}$$

bounds

$$\begin{aligned}
 -\infty &\leq s_1 \leq 1 \\
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candidate solution β

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 \end{aligned}$$

Explanation:

$$(3x_2 - x_1 \leq 1)$$

Example

$$[(3x_2 - x_1 \leq 1), (x_1 + x_2 \geq 0), (x_3 - 2x_1 \geq 3), (2x_3 \leq 1)]$$

s_1 s_2 s_3 s_4

tableau

$$\begin{aligned}
 x_3 &= -\frac{1}{2}s_1 + \frac{3}{2}s_2 + s_3 \\
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Explanation:

$$(3x_2 - x_1 \leq 1), (x_1 + x_2 \geq 0)$$

Example

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Example

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Explanation:

$$(3x_2 - x_1 \leq 1), (x_1 + x_2 \geq 0), (x_3 - 2x_1 \geq 3), (2x_3 \leq 1)$$

Linear Integer Arithmetic (LIA)

- **NP-complete** problem
- **Popular approach**: simplex + **branch and bound**
 - Approximate checks solve only over the rationals
 - In complete checks, force integrality of variables by adding either:
 - **Branch and bound** lemmas $(x \leq \lfloor c \rfloor) \vee (x \geq \lceil c \rceil)$
 - **Cutting plane** lemmas
 - Inequalities entailed by the current constraints, excluding only non-integer solutions
 - Gomory cuts commonly used
 - Using splitting on-demand
 - Might also include other **specialized sub-solvers** for tractable fragments
 - E.g. specialized equational reasoning

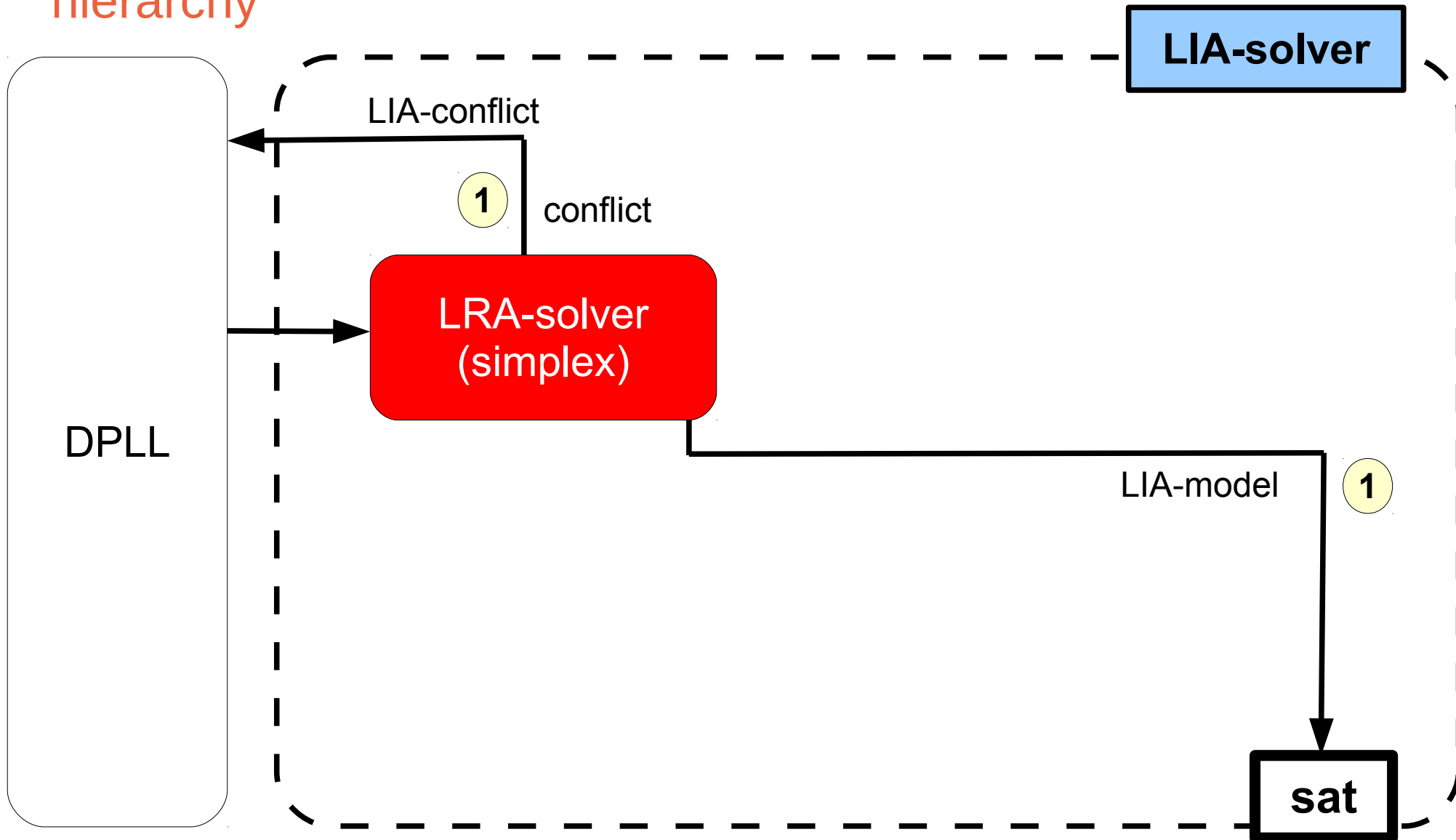
Layered architecture for LIA (MathSAT)

- Cooperation of several sub-modules organized as a **layered hierarchy**



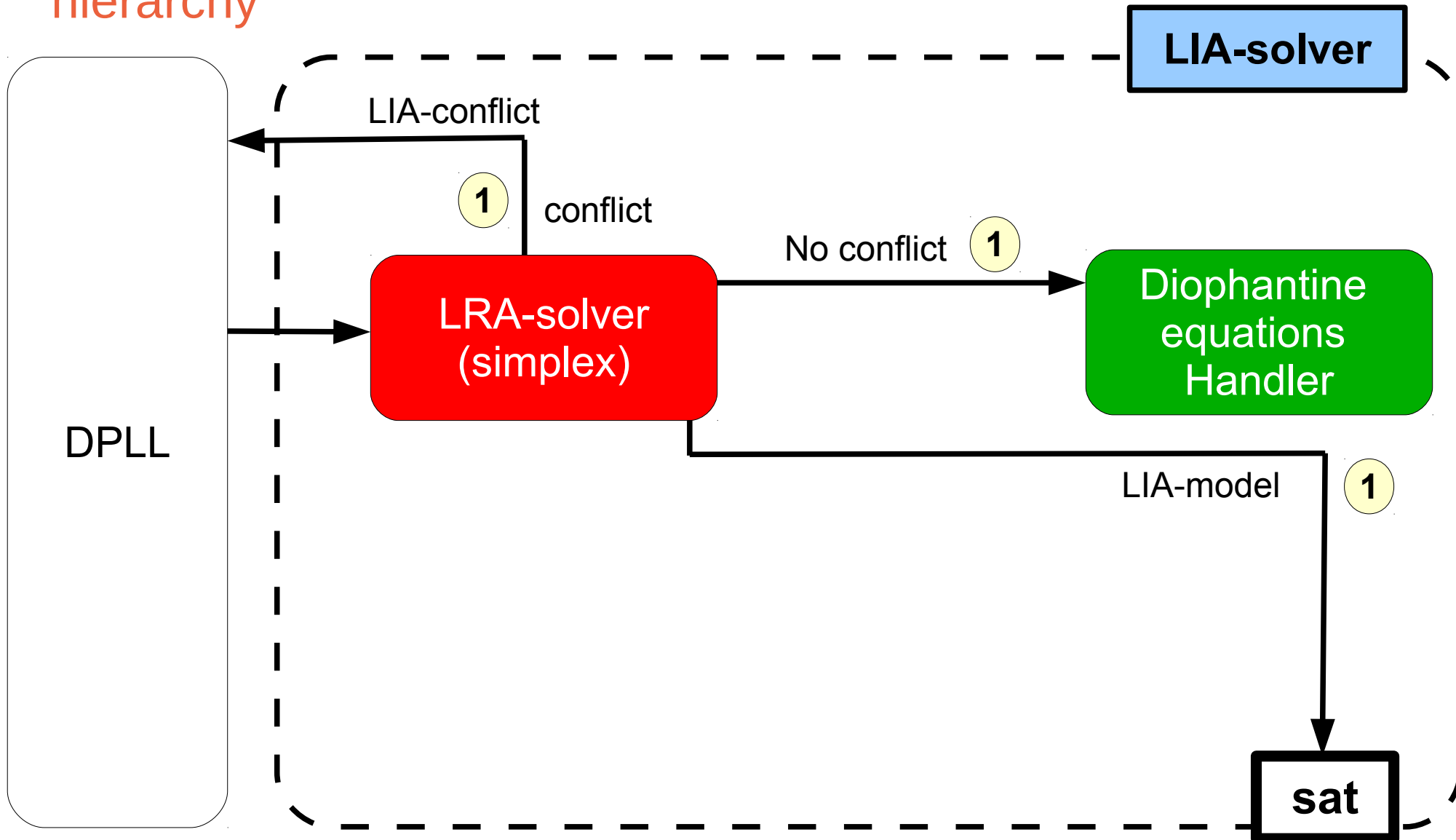
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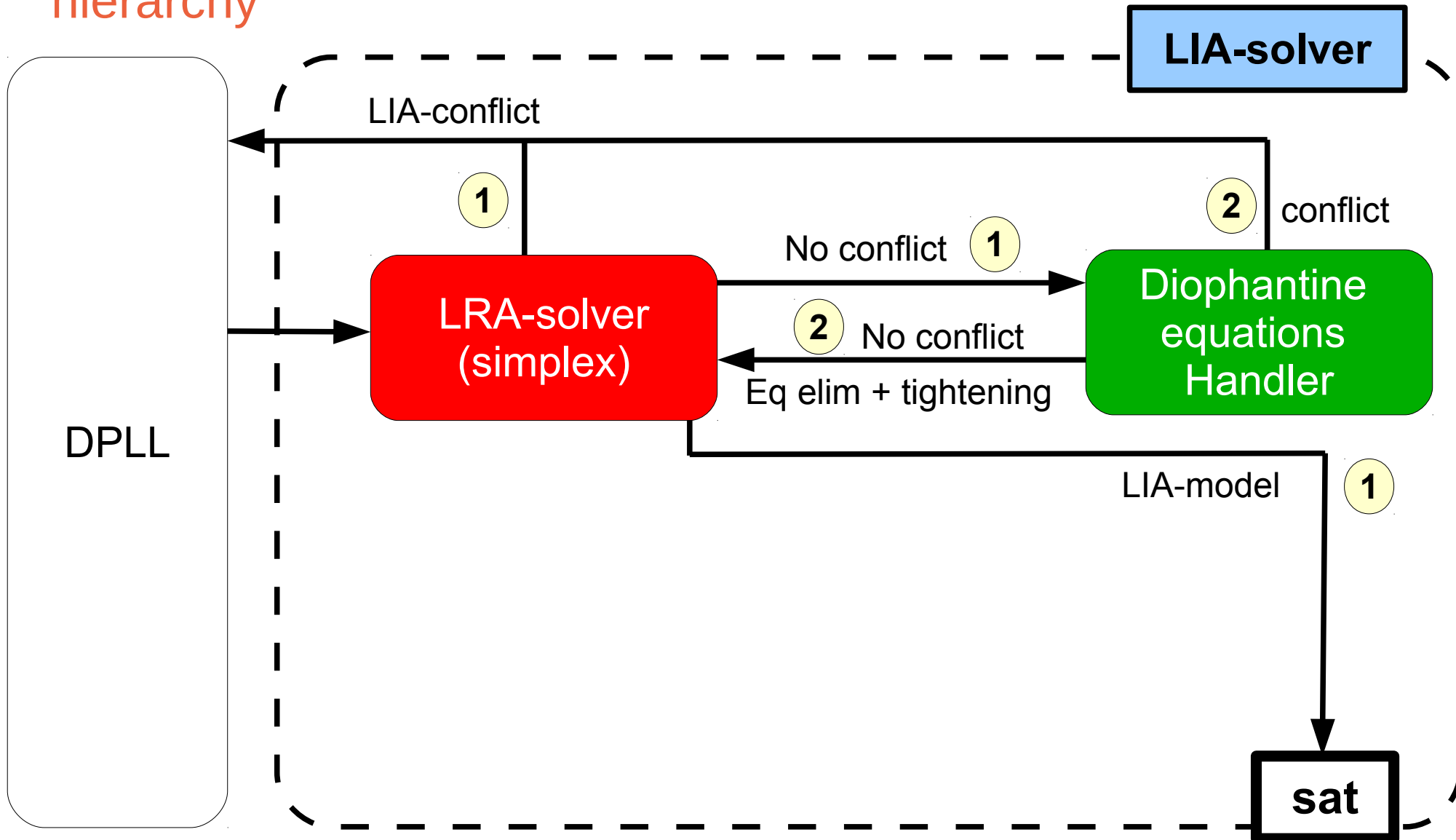
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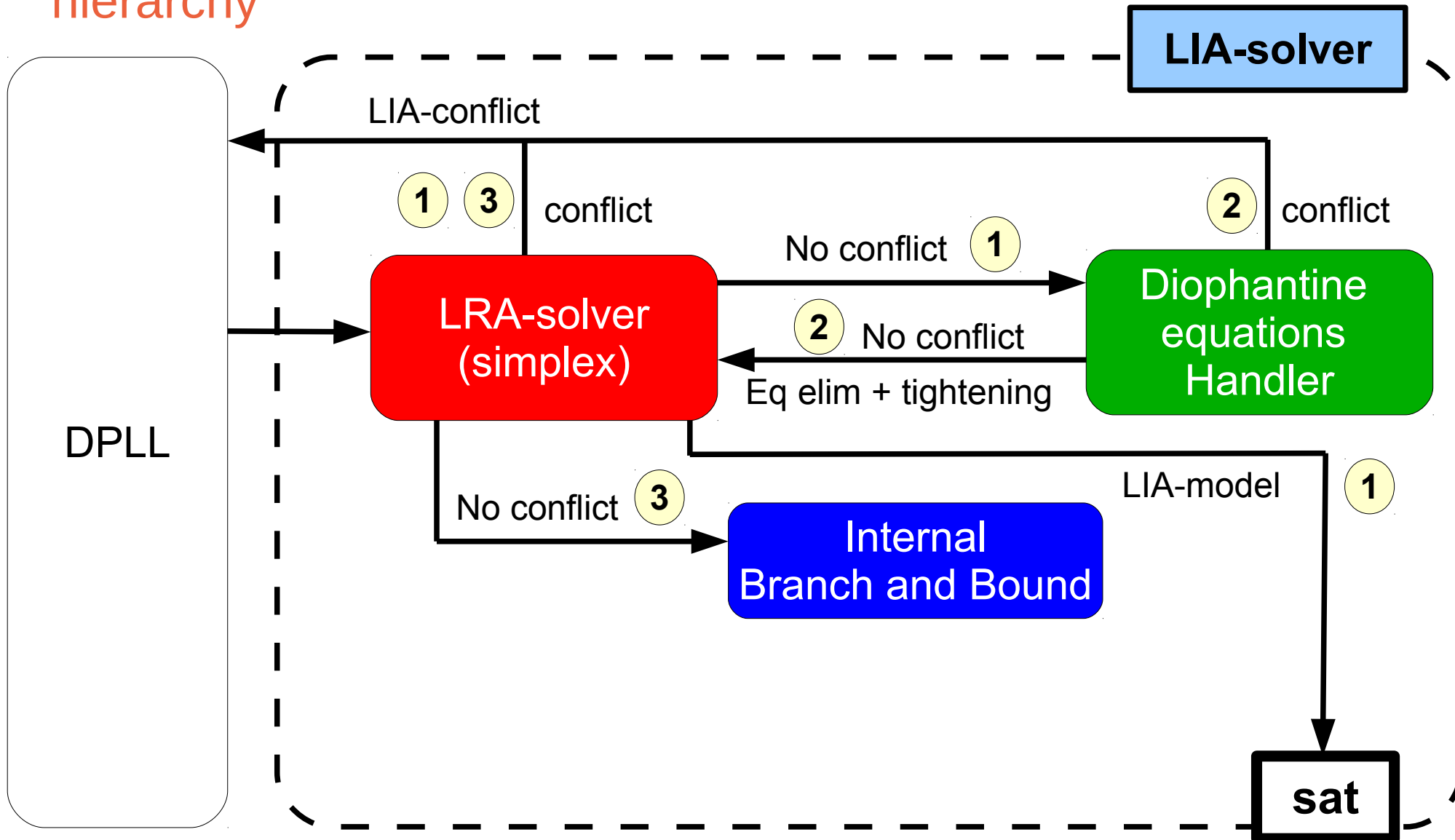
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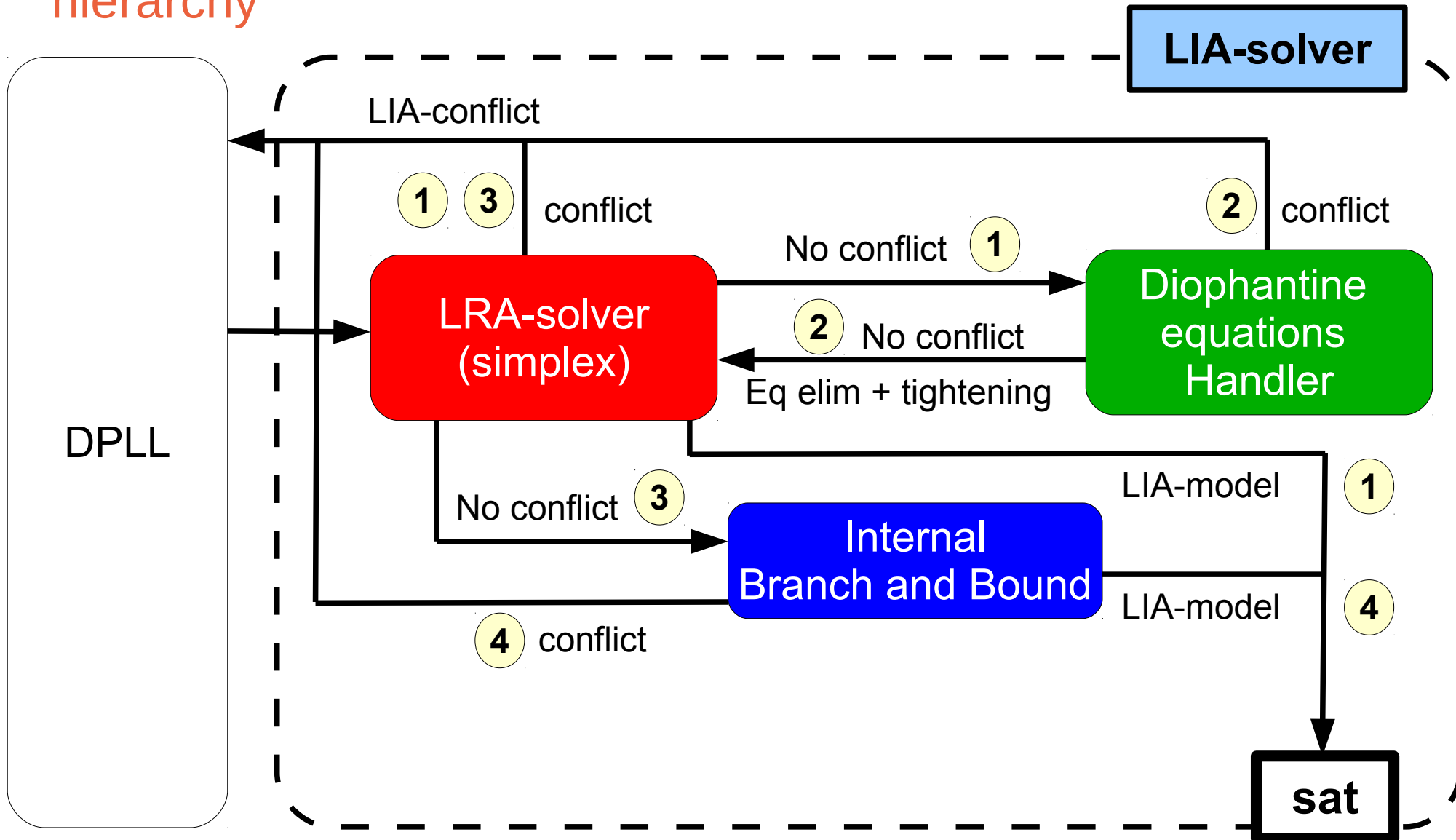
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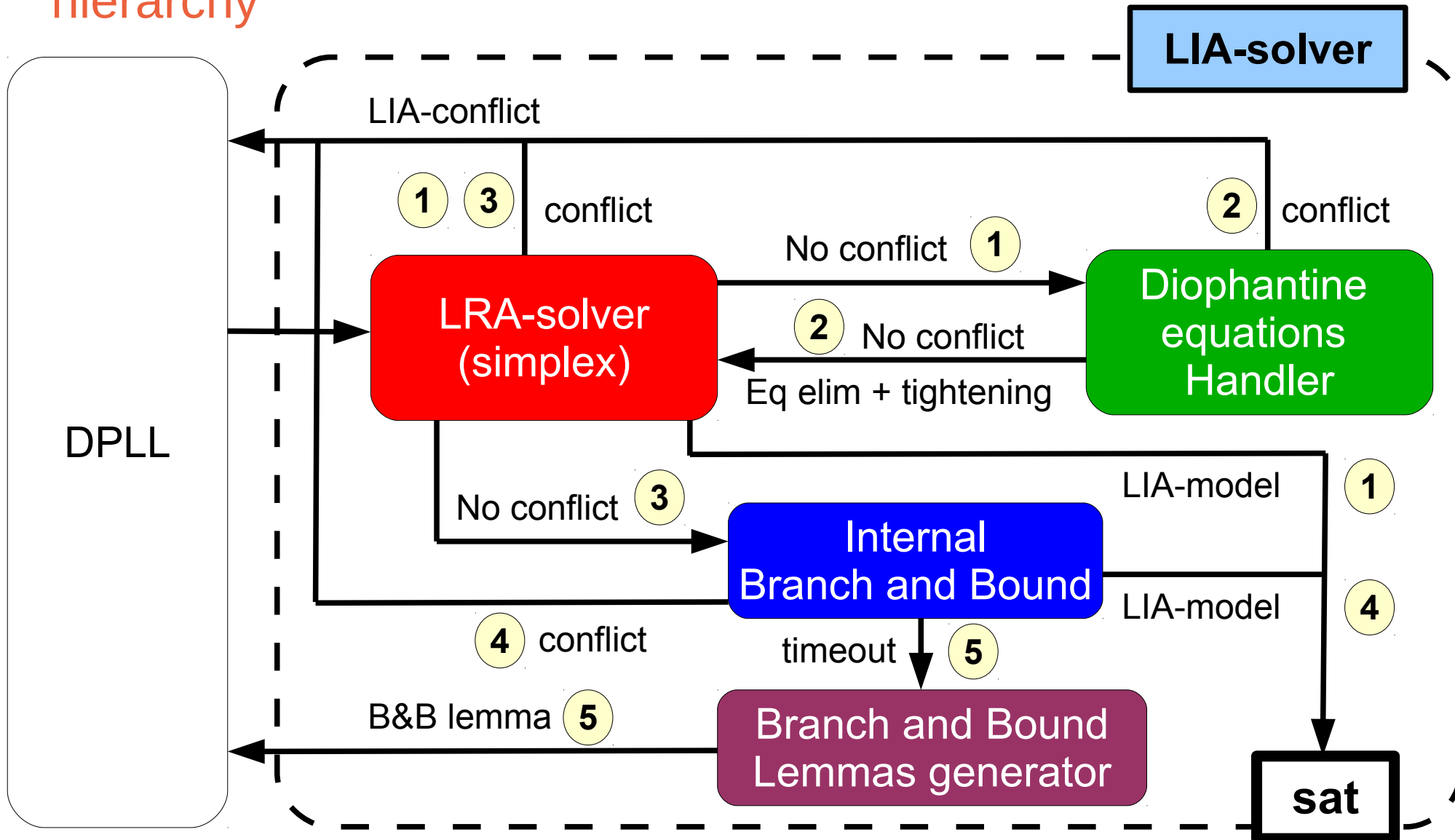
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- Cooperation of several sub-modules organized as a **layered hierarchy**



The Diophantine equation handler

- **Polynomial-time** procedure for solving systems of equations in LA(Z) (Diophantine)
- Similar to the first part of the Omega test [Pug91]
- **Extension of Gaussian elimination to integer constraints**
 - Given $\sum_j a_{ij}x_j + a_{ik}x_k + c_i$ where $|a_{ik}|$ is the smallest:
 - Rewrite into $a_{ik} \cdot (x_k + \sum_{j \neq k} a_{ij}^q x_j + c_i^q) + (\sum_{j \neq k} a_{ij}^r x_j + c_i^r)$
where $a_{ij} = a_{ik} \cdot a_{ij}^q + a_{ij}^r$
 - Introduce a **fresh var** x_t and add the equation
$$x_t = x_k + \sum_{j \neq k} a_{ij}^q x_j + c_i^q$$
 to the system
 - **Substitute** x_k with $a_{ik}x_t + (\sum_{j \neq k} a_{ij}^r x_j + c_i^r)$ in the other equations
 - Return **unsat** when there is an equation $\sum_j a_{hj}x_j + c_h$ such that $\text{GCD}(a_{h1}, \dots, a_{hn})$ does not divide c_h
 - Return **sat** when the system is in triangular form

The Diophantine equation handler

■ Features:

- Can generate *T*-lemmas when unsat is detected
 - Can actually generate detailed proofs of unsatisfiability as linear combinations of the input constraints
- When sat is detected, the solution can be used to eliminate equalities
 - Allows for performing tightening of inequalities, with which the simplex solver can discover more conflicts
- Can be performed incrementally with efficient backtracking

Tightening - Example

$$E := \begin{cases} 2x_1 - 5x_3 = 0 \\ x_2 - 3x_4 = 0 \end{cases} \quad I := \begin{cases} -2x_1 - x_2 - x_3 \leq -7 \\ 2x_1 + x_2 + x_3 \leq 8 \end{cases}$$

- Give E to the Diophantine equations handler

- Obtain (parametric) solution $S := \begin{cases} x_1 = 2x_3 + t \\ x_2 = 3x_4 \\ x_3 = 2t \end{cases}, t \text{ fresh}$

- Substitute x_1, x_2, x_3 in I , obtaining $I' := \begin{cases} -3x_4 - 12t \leq -7 \\ 3x_4 + 12t \leq 8 \end{cases}$

- Tighten, obtaining $I'' := \begin{cases} -\frac{3}{3}x_4 - \frac{12}{3}t \leq \lfloor -\frac{7}{3} \rfloor \\ \frac{3}{3}x_4 + \frac{12}{3}t \leq \lfloor \frac{8}{3} \rfloor \end{cases}$

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- Substitute x_1, x_2, x_3 in I , obtaining $I' := \begin{cases} -3x_4 - 12t \leq -7 \\ 3x_4 + 12t \leq 8 \end{cases}$

- Tighten, obtaining $I'' := \begin{cases} -x_4 - 4t \leq -3 \\ x_4 + 4t \leq 2 \end{cases}$

- Give I'' to the LRA-solver \Rightarrow **conflict**

Branch and Bound

- Given an LRA-model μ for the current set of constraints S

- If there is an integer variable z such that $\mu(z) \notin \mathbb{Z}$

- S is LIA-consistent iff either $S \cup \{(z \leq \lfloor \mu(z) \rfloor)\}$
or $S \cup \{(z \geq \lceil \mu(z) \rceil)\}$ is

- **Branch and Bound** idea: recursively solve subproblems

$S^i \cup \{(z_j \leq \lfloor \mu^i(z_j) \rfloor)\}$ $S^i \cup \{(z_j \geq \lceil \mu^i(z_j) \rceil)\}$

until either a LIA-model is found or all of them are
LA(Q)-inconsistent

- **Implementation**: a popular approach is to use “splitting on-demand”

- Create new clause (lemma) $(z \leq \lfloor \mu(z) \rfloor) \vee (z \geq \lceil \mu(z) \rceil)$
and send it to DPLL, and continue searching

Branch and Bound with splitting on-demand



- Advantages:
 - Ease of implementation
 - No need to support case splits within the theory solver, can reuse DPLL
 - Exploit “for free” all the search space pruning techniques of modern DPLL solvers
 - Backjumping
 - Learning
 - ...

- However, in our setting splitting on-demand has also **some drawbacks**

Splitting on-demand: drawbacks - 1

- Can not fully exploit equality elimination and tightening
 - New variables introduced by the Diophantine equations handler and tightened inequalities are **invisible** to DPLL
 - New variables can't be used in branch-and-bound lemmas
 - Tightened inequalities are thrown away when returning to DPLL

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$$\text{Example } S := \begin{cases} z = 2y - 3w \\ z + 3w \leq -1 \\ 2x + z \geq 1 \\ x + w \leq 0 \end{cases}$$

$$\mu_{LRA} := \begin{cases} x = \frac{2}{5} \\ y = -\frac{1}{2} \\ z = \frac{1}{5} \\ w = -\frac{2}{5} \end{cases}$$

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After substitution of z and tightening:

$$S' := \begin{cases} y \leq -1 \\ 2x + 2y - 3w \geq 1 \\ x + w \leq 0 \end{cases}$$

$$\mu'_{LRA} := \begin{cases} x = \frac{3}{5} \\ y = -1 \\ w = -\frac{3}{5} \end{cases}$$

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Example $S := \begin{cases} z = 2y - 3w \\ z + 3w \leq -1 \\ 2x + z \geq 1 \\ x + w \leq 0 \end{cases}$ After substitution of z and tightening:
 $S' := \begin{cases} y \leq -1 \\ 2x + 2y - 3w \geq 1 \\ x + w \leq 0 \end{cases}$

If we branch on $(x \leq \lfloor \frac{3}{5} \rfloor)$ (i.e. $(x \leq 0)$)

Then the **simplex** finds a **LIA-model** for $S' \cup \{(x \leq 0)\}$

However, the model found for $S \cup \{(x \leq 0)\}$ is **not good** in LA(Z)

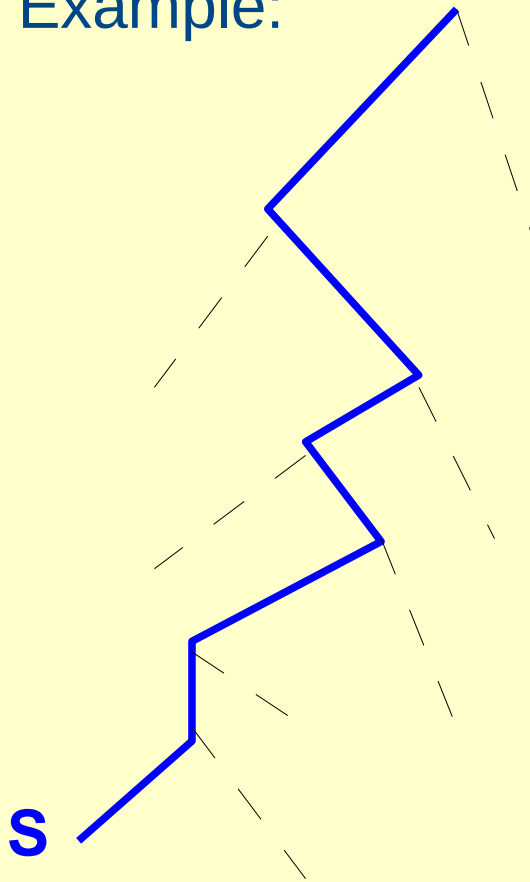
Splitting on-demand: drawbacks - 2

- Branch and bound lemmas might cease to be useful upon backtracking
 - Branch and bound aimed at finding a LIA-model for the constraints in the current branch
 - Splitting on-demand adds “global” lemmas
 - They might “pollute” the search space
 - Overhead in DPLL

Splitting on-demand: drawbacks - 3

- Non-chronological backtracking (backjumping) might hurt sometimes

- Example:



- Set of constraints S in the current DPLL branch

LA(Q)-model for S s.t. $\mu(x_k) = q_k \notin \mathbb{Z}$

Branch and bound lemma:

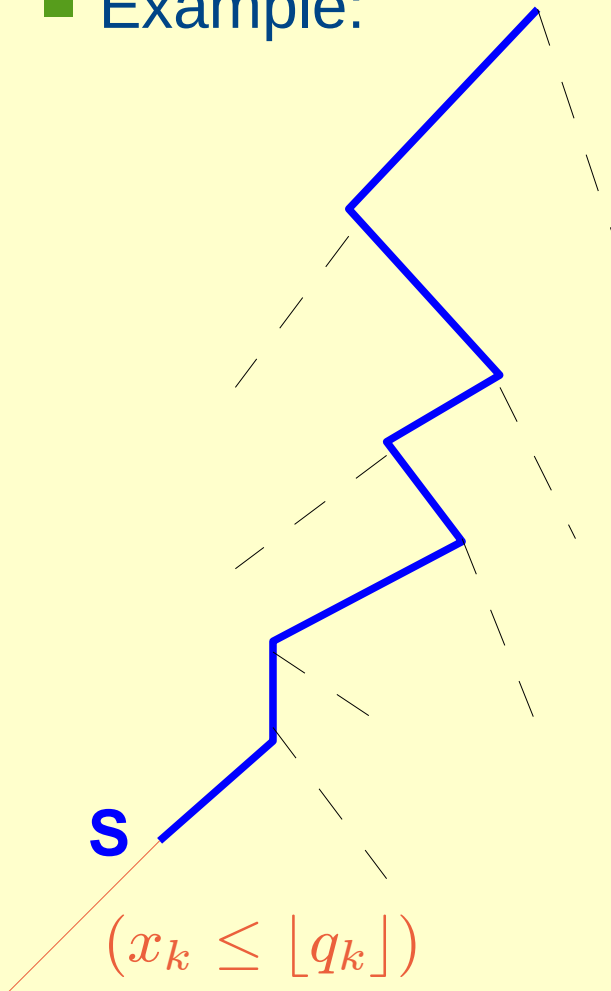
$$(x_k \leq \lfloor q_k \rfloor) \vee (x_k \geq \lceil q_k \rceil)$$

Suppose the LA(Q)-model for $S \cup \{(x_k \geq \lceil q_k \rceil)\}$ is also a LA(Z)-model, but we branch on $(x_k \leq \lfloor q_k \rfloor)$ first

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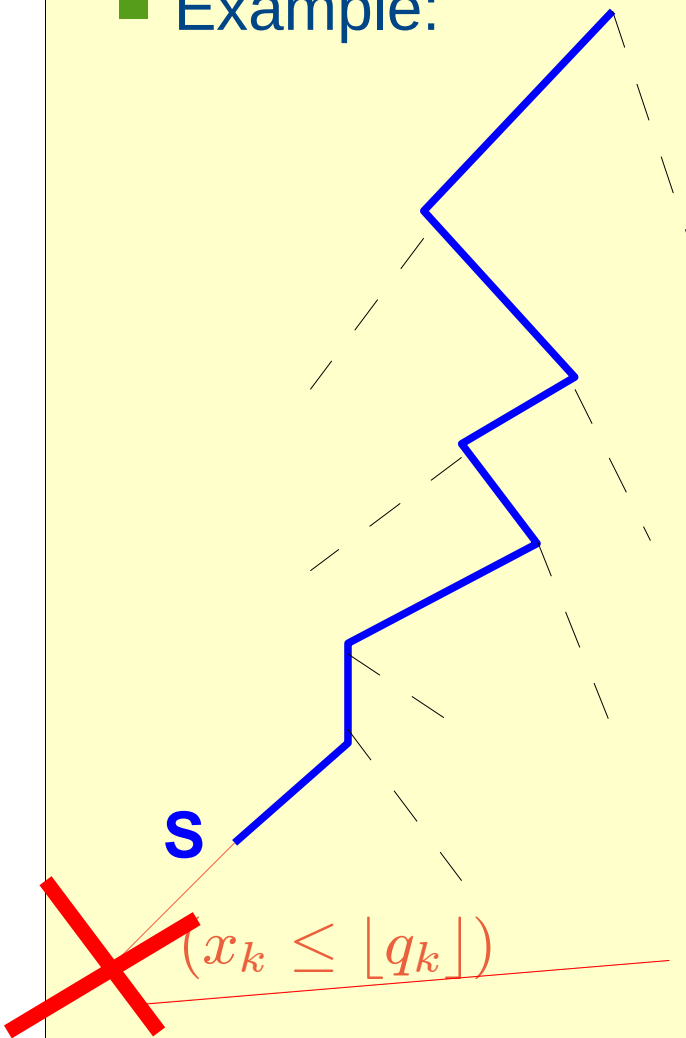
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$$(x_k \leq \lfloor q_k \rfloor) \vee (x_k \geq \lceil q_k \rceil)$$

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Conflict!

Splitting on-demand: drawbacks - 3

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- Example:

- Set of constraints S in the current DPLL branch

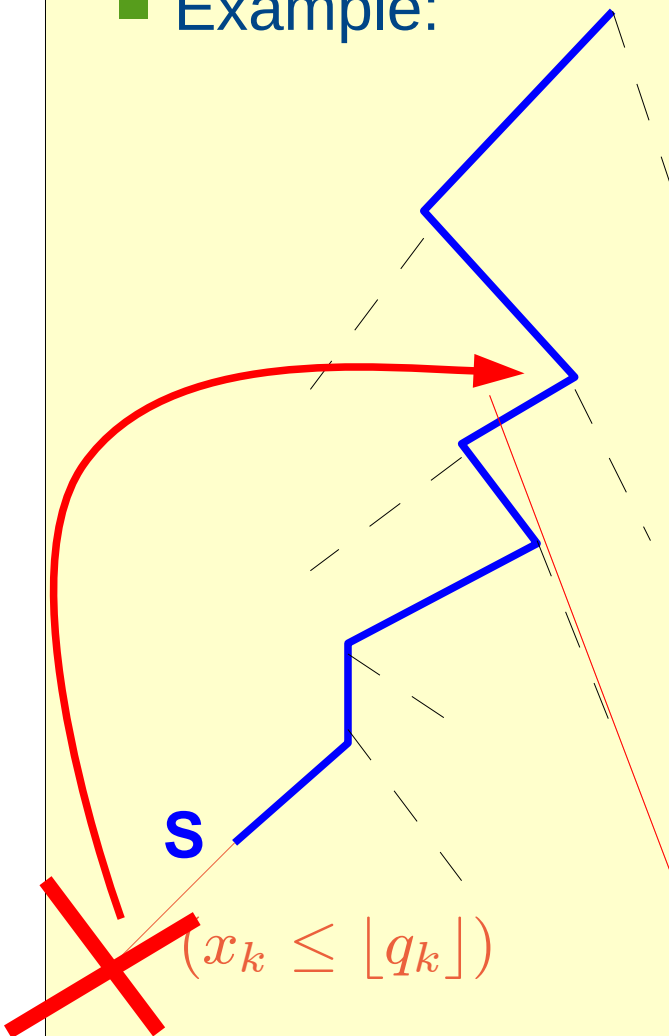
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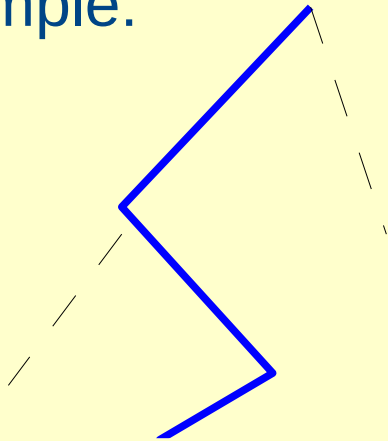
**Backjumping after
conflict analysis**



Splitting on-demand: drawbacks - 3

- Non-chronological backtracking (backjumping) might hurt sometimes

- Example:



- Set of constraints S in the current DPLL branch

LA(Q)-model for S s.t. $\mu(x_k) = q_k \notin \mathbb{Z}$

Branch and bound lemma:

$$(x_k \leq \lfloor q_k \rfloor) \vee (x_k \geq \lceil q_k \rceil)$$

Suppose the LA(Q)-model for $S \cup \{(x_k \geq \lceil q_k \rceil)\}$ is also a LA(Z)-model, but we branch on $(x_k \leq \lfloor q_k \rfloor)$ first

After backjumping, might need to redo a lot of expensive computations (equality elimination, tightening) before finding S again

Splitting on-demand: drawbacks - 4

- Difficult to use dedicated heuristics for exploring the branch-and-bound search tree
 - Several sophisticated heuristics developed in the ILP community, crucial for performance
 - Might not be straightforward to integrate with those commonly used in DPLL

Internal branch and bound

- **Possible solution** (MathSAT): perform branch and bound search within the LIA-solver
 - Start from the result of **equality elimination + tightening**
 - Use **dedicated heuristics** for selecting the variables on which to branch
 - Do **not backjump** past the starting point
 - **Remove redundant constraints** before starting branch and bound
 - E.g. by exploiting polarity of variables
- **Only perform a bounded (small) number of case splits, and then revert to splitting on-demand**
 - Keep the benefits of splitting on-demand for hard problems

- Most solvers use an **eager approach** for BV, not DPLL(T)
 - **Heavy preprocessing** based on **rewriting rules** + SAT encoding (“bit-blasting”)

$$\begin{aligned} \text{■ Example: } & (x_{[1]} \neq 0_{[1]}) \wedge (y_{[31]} :: x_{[1]} \% 2_{[32]} = 0_{[32]}) \mapsto \\ & (x_{[1]} = 1_{[1]}) \wedge (y_{[31]} :: x_{[1]} \% 2_{[32]} = 0_{[32]}) \mapsto \\ & (y_{[31]} :: 1_{[1]} \% 2_{[32]} = 0_{[32]}) \mapsto \perp \end{aligned}$$

- Alternative: **lazy bit-blasting**, compatible with DPLL(T)
 - Use a second SAT solver as T-solver for BV
 - bit-blast only BV-atoms, not the whole formula
 - Boolean skeleton of the formula handled by the main SAT solver
 - **Easier integration** with other T-solvers and DPLL(T)
 - Can integrate additional **specialized sub-solvers**
- Eager still better performance-wise

Lazy bit-blasting: implementation

- **For each BV-atom** α occurring in the input formula, create a fresh Boolean “**label**” variable l_α , and bit-blast to SAT-BV the formula $(l_\alpha \leftrightarrow \alpha)$
- **Exploit SAT solving under assumptions**
 - When the main solver generates the BV-assignment $\alpha_1 \dots \alpha_n$
 - Invoke SAT-BV **with the assumptions** $l_{\alpha_1} \dots l_{\alpha_n}$
 - If unsat, generate an **unsat core of the assumptions** $l_{\alpha_i} \dots l_{\alpha_j}$
 - From its negation, generate a BV-lemma $\neg\alpha_i \vee \dots \vee \neg\alpha_j$ and send it to the main solver as a blocking clause, like in standard DPLL(T)

SAT solving under assumptions

Modern CDCL-based SAT solvers allow for solving a CNF formula φ **under assumptions on the values of some literals** $\{l_1, l_2, \dots, l_n\}$

- Logically equivalent to checking $\varphi \wedge \bigwedge_i l_i$
- But l_1, \dots, l_n are assumed only temporarily
 - A limited but very useful form of **incremental solving**
- If φ is **unsat** under the assumptions $\{l_1, l_2, \dots, l_n\}$ we can ask the SAT solver to compute an **unsatisfiable core of the assumptions**
 - A subset $\{l_j, \dots, l_k\} \subseteq \{l_1, \dots, l_n\}$ that is sufficient for proving unsatisfiability

SAT under assumptions - implementation



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- **Modify the branching heuristics** of the CDCL solver to always pick the next unassigned literal from $\{l_1, l_2, \dots, l_n\}$ before other literals
 - The **first n decision levels** of the trail always correspond to the assumptions
- If an **assumption** literal is **assigned to false**, return **unsat**
 - Can only happen by unit propagation at a level $< n$
- **Unsat core**: start **conflict analysis** from the falsified assumption literal $\neg l_j$, and use the “**decision**” strategy to collect all the involved assumptions

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit | Reason

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
$\neg A_3$	—

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

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$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
$\neg A_3$	—
A_4	—

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

Assumptions


$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
$\neg A_3$	—
A_4	—
A_5	c_4



Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

Assumptions


$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
$\neg A_3$	—
A_4	—
A_5	c_4



$$c_5 : \neg A_4 \vee \neg A_5 \quad c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_6^l : A_3 \vee \neg A_4$$

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
$\neg A_3$	—

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
$\neg A_3$	—
$\neg A_4$	c_6^l

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
$\neg A_3$	—
$\neg A_4$	c_6^l
A_5	c_2



Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
$\neg A_3$	—
$\neg A_4$	c_6^l
A_5	c_2



$$c_3 : A_4 \vee \neg A_5 \quad c_2 : A_1 \vee A_5 \vee A_4$$

$$c_7^l : A_1 \vee A_4$$

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

$$c_7^l : A_1 \vee A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

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$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

$$c_7^l : A_1 \vee A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
A_4	c_7^l

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

$$c_7^l : A_1 \vee A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$



Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
A_4	c_7^l
A_3	c_6^l

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

$$c_7^l : A_1 \vee A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
A_4	c_7^l
A_3	c_6^l

$$c_6^l : A_3 \vee \neg A_4$$

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

$$c_7^l : A_1 \vee A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
A_4	c_7^l
A_3	c_6^l

$$c_6^l : A_3 \vee \neg A_4$$

$$c_7^l : A_1 \vee A_4$$

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

$$c_7^l : A_1 \vee A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
A_4	c_7^l
A_3	c_6^l

$$c_6^l : A_3 \vee \neg A_4 \quad c_7^l : A_1 \vee A_4$$

$$A_1 \vee A_3$$

Example

Input clauses

$$c_1 : \neg A_2 \vee A_1 \vee A_5$$

$$c_2 : A_1 \vee A_5 \vee A_4$$

$$c_3 : A_4 \vee \neg A_5$$

$$c_4 : A_3 \vee A_5 \vee \neg A_4$$

$$c_5 : \neg A_4 \vee \neg A_5$$

$$c_6^l : A_3 \vee \neg A_4$$

$$c_7^l : A_1 \vee A_4$$

Assumptions

$$\neg A_1$$

$$\neg A_2$$

$$\neg A_3$$

Trail

Lit	Reason
$\neg A_1$	—
$\neg A_2$	—
A_4	c_7^l
A_3	c_6^l

$$c_6^l : A_3 \vee \neg A_4 \quad c_7^l : A_1 \vee A_4$$

$$A_1 \vee A_3$$

Arrays (A)

- **Read (rd)** and **write (wr)** operations over arrays
- **Equality** over array variables (extensionality)
- **Example:** $wr(a, i, x) = wr(b, i, rd(a, j, y)) \wedge \neg(a = b)$
- **Common approach:** reduction to EUF via **lazy axiom instantiation**

- **read-over-write:**
$$\forall a. \forall i. \forall x. (rd(wr(a, i, x), i) = x)$$
$$\forall a. \forall i. \forall j. \forall x. ((i \neq j) \rightarrow rd(wr(a, i, x), j) = rd(a, j))$$

- **extensionality:**
$$\forall a. \forall b. ((a \neq b) \rightarrow \exists i. (rd(a, i) \neq rd(b, i)))$$

- Add **lemmas on-demand** by instantiating the quantified variables with terms occurring in the input formula
 - Using **smart “frugal” strategies:** **check** candidate solution, instantiate only **(potentially) violated axioms**

Example

$$\neg(j = k), \neg(\text{rd}(\text{wr}(a, i, x), j) = \text{rd}(a, j)), \neg(\text{rd}(\text{wr}(a, i, x), k) = \text{rd}(a, k))$$

EUF solution (equivalence classes):

$$\{a, \text{wr}(a, i, x)\}, \{\text{rd}(\text{wr}(a, i, x), j)\}, \{\text{rd}(\text{wr}(a, i, x), k)\}, \\ \{x, i, j\}, \{k\}, \{\text{rd}(a, j)\}, \{\text{rd}(a, k)\}$$

Example

$$\neg(j = k), \neg(\text{rd}(\text{wr}(a, i, x), j) = \text{rd}(a, j)), \neg(\text{rd}(\text{wr}(a, i, x), k) = \text{rd}(a, k))$$

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Add violated lemma: $(i \neq k) \rightarrow (\text{rd}(\text{wr}(a, i, x), k) = \text{rd}(a, k))$

Example

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Example

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EUF solution (equivalence classes):

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EUF solution (equivalence classes):

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Add violated lemma: $(i \neq j) \rightarrow (\text{rd}(\text{wr}(a, i, x), j) = \text{rd}(a, j))$

EUF solver returns **UNSAT**

Outline



Introduction

CDCL-based SAT solvers

The DPLL(T) architecture

Some relevant T-solvers

Combination of theories

- Very often in practice more than one theory is needed

- Example (from intro):

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

- How to build solvers for $\text{SMT}(T_1 \dots T_n)$ that are both **efficient** and **modular**?
 - Can we **reuse** T_i -solvers and **combine** them?
 - Under what conditions?
 - How do we go from $\text{DPLL}(T)$ to $\text{DPLL}(T_1 \dots T_n)$?

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 (other than =)
 - 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
 - **Examples:** EUF, LIA, LRA, A
 - **Counterexample:** BV
 - *(Extensions exist to deal with some non-stably-infinite theories)*

The Nelson-Oppen method

How it works (for $T_1 \cup T_2$)

- Preprocessing **purification** step on the input formula φ
 - **Pure formula**: no atom containing symbols of different T_i 's (except =)
 - By **labeling** subterms

■ **Example:**

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

The Nelson-Oppen method

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- T_i -solvers cooperate by **exchanging (disjunctions of) entailed interface equalities**
 - I.e., equalities between shared variables

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

- T_i -solvers cooperate by **exchanging (disjunctions of) entailed interface equalities**

- I.e., equalities between shared variables

Interface variables

Example

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

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

Example

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

$$\models$$

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

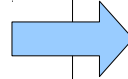
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Example

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$$(x_1 = x_3)$$

\models

$$(x_5 = x_6)$$

Example

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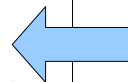
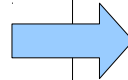
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$$(x_5 = x_6)$$

$$\models$$

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

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Example

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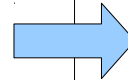
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$$(x_5 = x_6)$$

\models

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

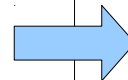
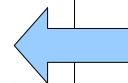
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$$(x_1 = x_3)$$

\models

$$(x_5 = x_6)$$



$$(x_2 = x_3)$$



Example

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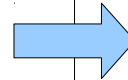
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$$(x_5 = x_6)$$

\models

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

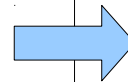
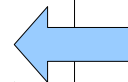
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$$(x_1 = x_3)$$

\models

$$(x_5 = x_6)$$



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



Example

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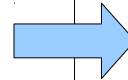
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$$(x_5 = x_6)$$

\models

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

$$\neg(f(x_1) = f(x_2)), \text{ EUF} \\ \neg(f(x_2) = f(x_4)), \\ (f(x_3) = x_5), (f(x_1) = x_6)$$



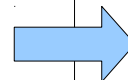
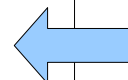
$$(x_1 = x_3)$$

$$(x_1 = x_4)$$

\models

$$(x_5 = x_6)$$

No more  deductions possible



$$(x_2 = x_3)$$

$$(x_2 = x_4)$$



DPLL(T) for combined theories

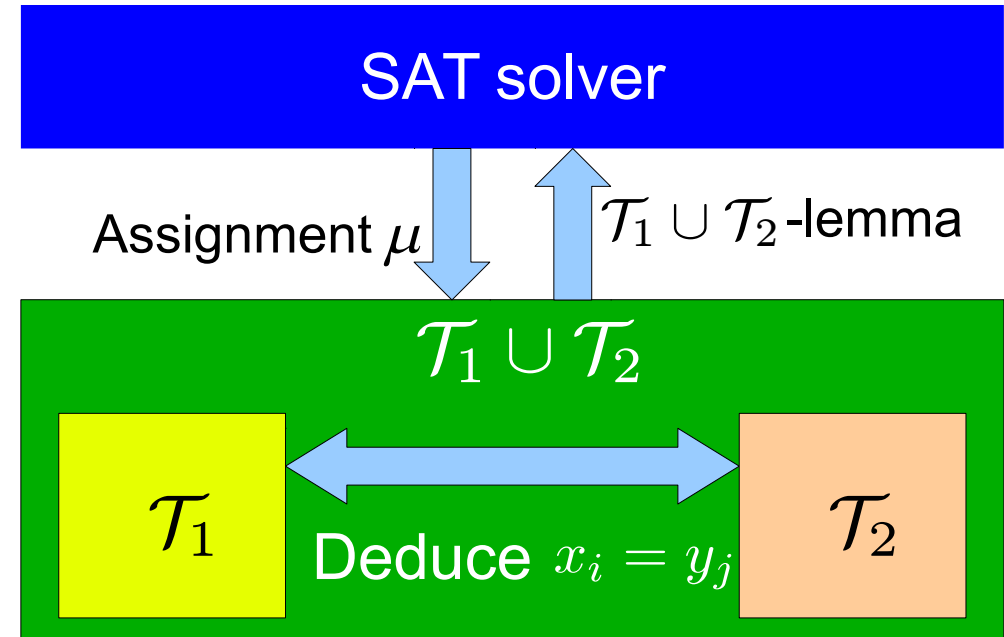
- **Traditional approach:**
a **single** combined
Nelson-Oppen T -solver

- T_i -solvers exchange
(disjunctions of) implied
interface equalities
internally

- Interface equalities
invisible to the SAT solver

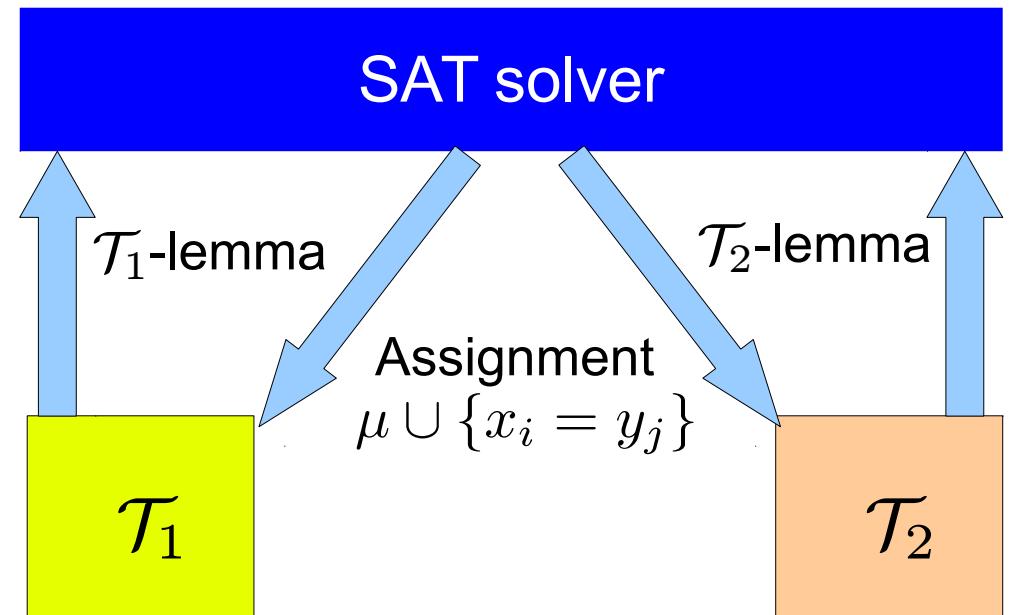
- **Drawbacks:** T_i -solvers need to:

- be **deduction complete** for interface equalities
- be able to **perform case splits** internally



Delayed Theory Combination

- **Alternative** to traditional approach
 - Each T_i -solver interacts directly and only with the SAT solver
 - SAT solver takes care of (all or part of) the combination
 - **Augment the Boolean search space** with the possible interface equalities ($x_i = y_j$)
- **Advantages:**
 - No need of complete deduction of interface equalities
 - Case analysis via **splitting on-demand**



- **Model-based** heuristic:
 - Initially, no interface equalities generated
 - When a solution is found, **check against all the possible interface equalities**
 - If T_1 and T_2 **agree** on the implied equalities, return **SAT**
 - Otherwise, branch on equalities **implied by T_1 -model but not by T_2 -model**
 - **Optimistic approach**, similar to axiom instantiation
- Still allow T_i -solvers to **exchange equalities internally**
 - But **no requirement of completeness**
 - **Avoids “polluting” the SAT space** with equality deductions leading to conflicts

Example

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

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

Example

LIA $(x_1 \geq 0), (x_1 \leq 1), (x_2 \geq x_6)$
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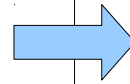
$x_1 \mapsto 1 \quad x_2 \mapsto 2$

LIA-model: $x_3 \mapsto 0 \quad x_4 \mapsto 1$

$x_5 \mapsto 0 \quad x_6 \mapsto 1$

\models

$(x_1 = x_4)$



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

EUF-model: $\{x_1\} \{x_2\} \{x_3\} \{x_4\}$
 $\{x_5, f(x_3)\} \{x_6, f(x_1)\}$
 $\{f(x_2)\} \{f(x_4)\}$

$\not\models$

$(x_1 = x_4)$

Branch on $(x_1 = x_4)$

$(x_1 = x_4)$

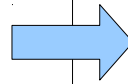
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LIA-model: $x_1 \mapsto 1 \quad x_2 \mapsto 2$
 $x_3 \mapsto 0 \quad x_4 \mapsto 1$
 $x_5 \mapsto 0 \quad x_6 \mapsto 1$

\models

$(x_3 = x_5)$
 $(x_4 = x_6)$



$(x_1 = x_4)$

$(x_3 = x_5)$

$(x_4 = x_6)$

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

EUUF-model: $\{x_1, x_4\} \{x_2\} \{x_3\}$
 $\{x_5, f(x_3)\}$
 $\{x_6, f(x_1), f(x_4)\} \{f(x_2)\}$

$\not\models$

$(x_3 = x_5)$
 $(x_4 = x_6)$

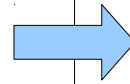
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$\neg(f(x_1) = f(x_2)),$ EUF
 $\neg(f(x_2) = f(x_4)),$
 $(f(x_3) = x_5), (f(x_1) = x_6)$

$\{x_1, x_4, x_6, f(x_1), f(x_4)\}$

EUF-model: $\{x_3, x_5, f(x_3)\}$

$\{f(x_2)\}\{x_2\}$



$(x_1 = x_4)$

$(x_3 = x_5)$

$(x_4 = x_6)$

Selected bibliography

DISCLAIMER: this is not meant to be complete, just a starting point. Apologies to missing authors/works

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■ Theory solvers

- Detlefs, Nelson, Saxe. **Simplify: a theorem prover for program checking**. J. ACM 2005
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■ Theory solvers (cont'd)

- Dutertre, de Moura. **A Fast Linear-Arithmetic Solver for DPLL(T)**. CAV 2006
- de Moura, Bjørner. **Model-based Theory Combination**. Electr. Notes Theor. Comput. Sci. 2008
- Brummayer, Biere. **Lemmas on Demand for the Extensional Theory of Arrays**. JSAT 2009
- de Moura, Bjørner. **Generalized, efficient array decision procedures**. FMCAD 2009
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Thank You

VTSA summer school 2015

Exploiting SMT for Verification of Infinite-State Systems

2. Interpolation in SMT and in Verification

Alberto Griggio

Fondazione Bruno Kessler – Trento, Italy

Outline



Introduction

Interpolants in Formal Verification

Computing interpolants in SMT

- (Craig) Interpolant for an ordered pair (A, B) of formulae s.t.

$A \wedge B \models_T \perp$ (or: $A \models_T \neg B$) is a formula I s.t.

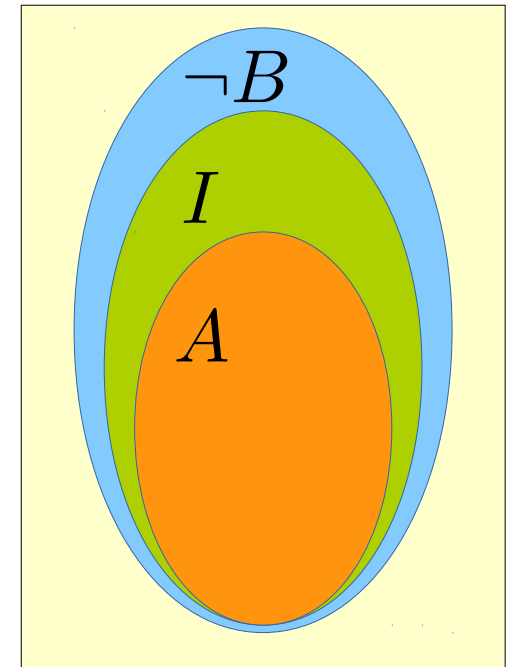
- $A \models_T I$
- $I \wedge B \models_T \perp$ ($I \models_T \neg B$)
- All the uninterpreted (in T) symbols of I are shared between A and B

- Why are interpolants useful?

- Overapproximation of A relative to B

- Overapprox. of $\exists_{\{x \notin B\}} \vec{x}. A$

- “Local” explanation of why A is inconsistent with B



Importance of interpolation

Several important applications in formal verification:

- **Approximate image computation** for model checking of infinite-state systems
- **Predicate discovery** for Counterexample-Guided Abstraction Refinement
- Approximation of transition relation for infinite-state systems
- An alternative to (lazy) predicate abstraction for program verification
- Automatic generation of loop invariants
- ...

Outline



Introduction

Interpolants in Formal Verification

Computing interpolants in SMT

Symbolic transition systems

- State variables X
- Initial states formula $I(X)$
- Transition relation formula $T(X, X')$
- A state σ is an assignment to the state vars $\bigwedge_{x_i \in X} x_i = v_i$
- A path of the system S is a sequence of states $\sigma_0, \dots, \sigma_k$ such that $\sigma_0 \models I$ and $\sigma_i, \sigma'_{i+1} \models T$
- A k -step (symbolic) unrolling of S is a formula
$$I(X^0) \wedge \bigwedge_{i=0}^{k-1} T(X^i, X^{i+1})$$
 - Encodes all possible paths of length up to k
- A state property is a formula P over X
 - Encodes all the states σ such that $\sigma \models P$

Forward reachability checking

■ Forward image computation

- Compute all states reachable from σ in one transition:

$$\text{Img}(\sigma(X)) := \exists X'. \sigma(X) \wedge T(X, X')[X/X']$$

- Prove that a set of states $\text{Bad}(X)$ is **not reachable**:

$$R(X) := \text{I}(X)$$



$$\text{Img}(R(X))$$

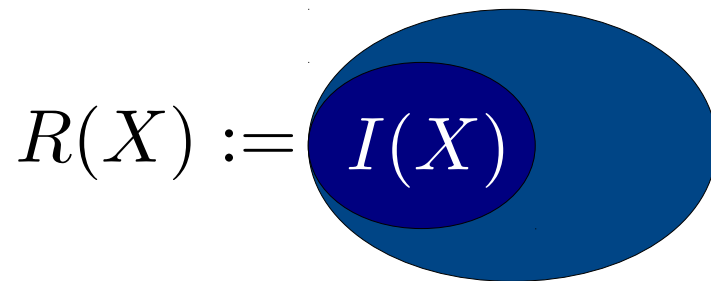
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$$\text{Img}(R(X))$$

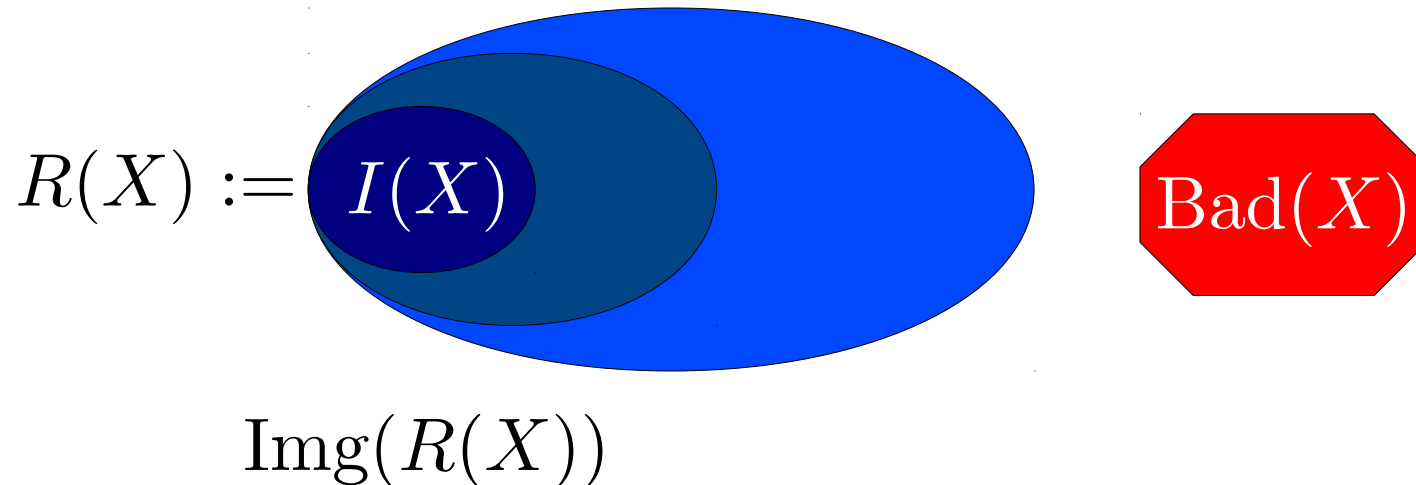
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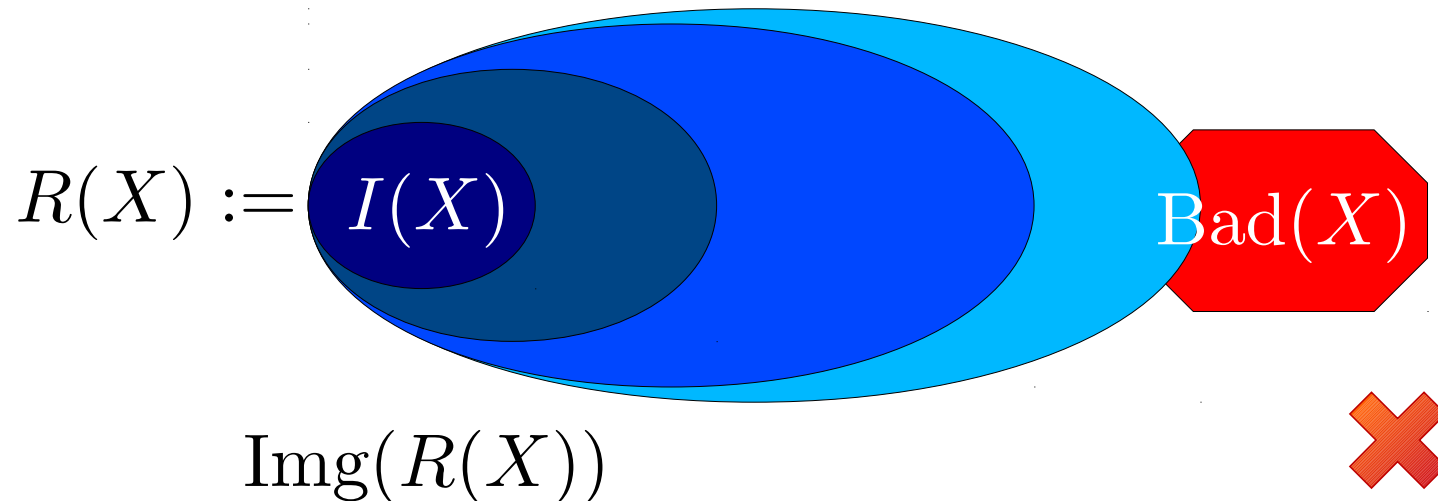
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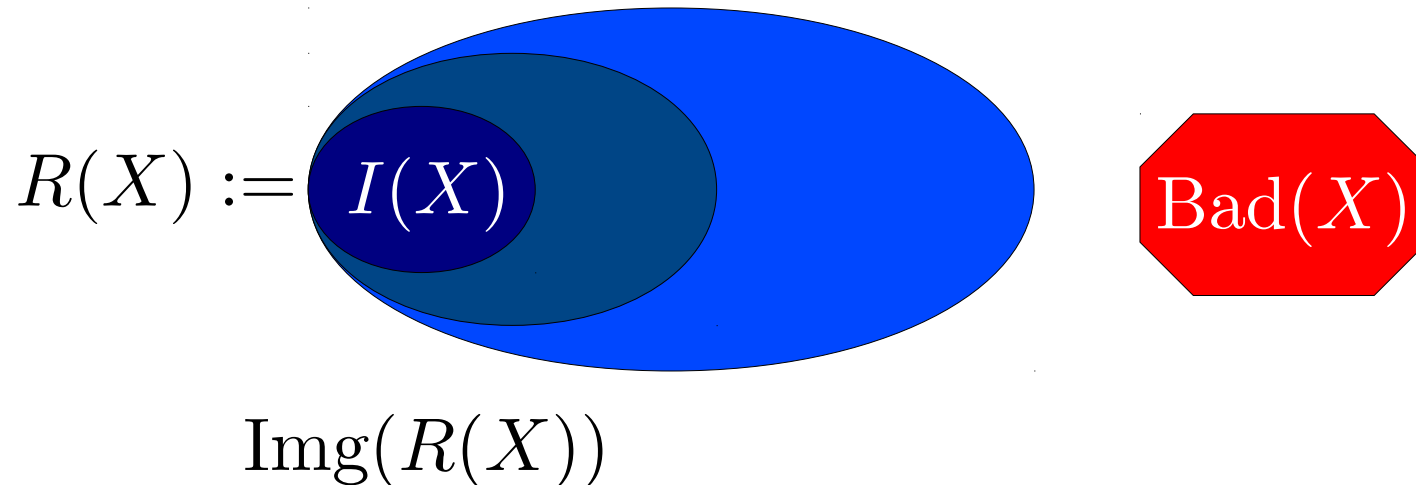
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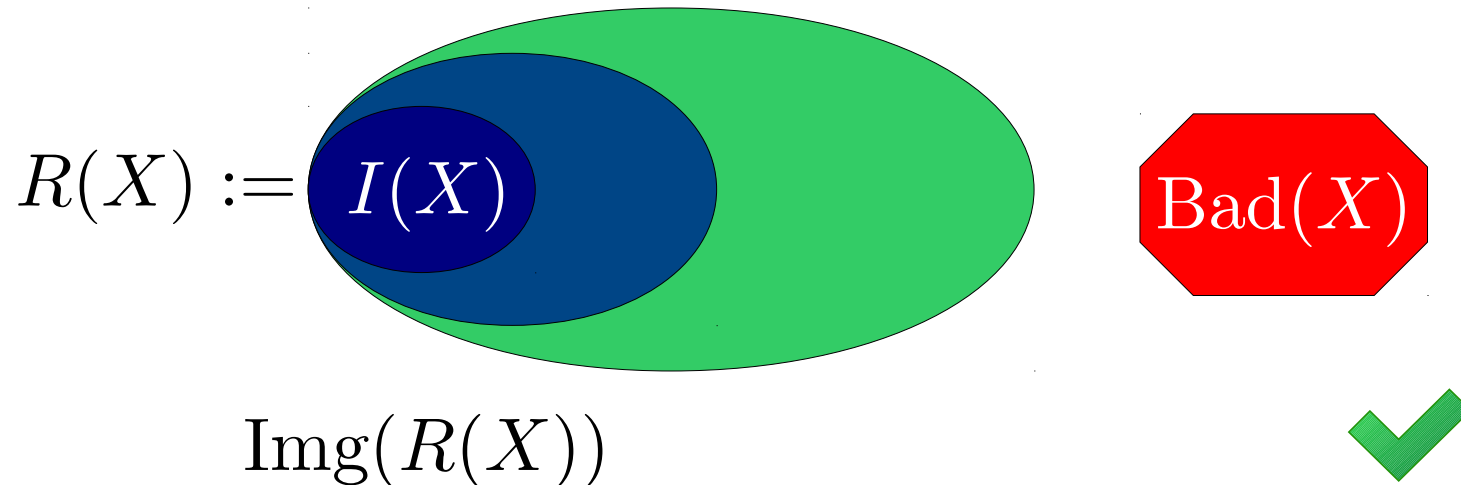
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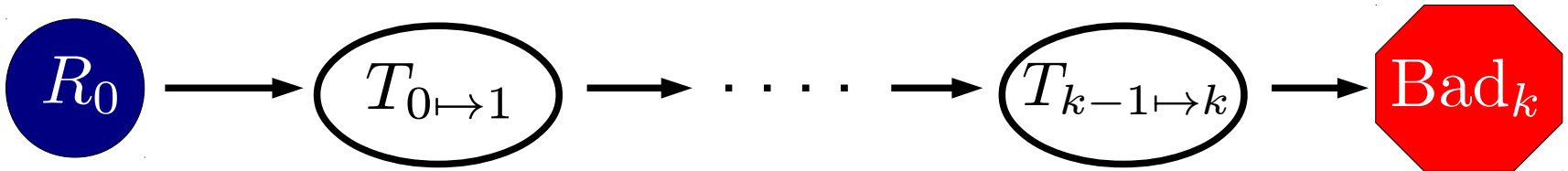


Interpolation-based reachability

- Image computation requires **quantifier elimination**, which is typically **very expensive** (both in theory and in practice)
- Interpolation-based algorithm (McMillan CAV'03): use interpolants to **overapproximate image computation**
 - **much more efficient** than the previous algorithm
 - interpolation is often much cheaper than quantifier elimination
 - abstraction (overapproximation) accelerates convergence
 - termination is still guaranteed for finite-state systems

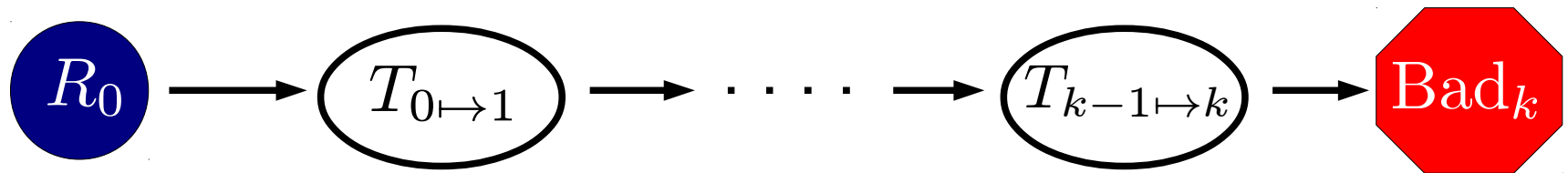
Interpolation-based reachability

- Set $R(X) := I(X)$
- Check satisfiability of $R_0 \wedge \bigwedge_{i=0}^{k-1} T_i \wedge \text{Bad}_k$



Interpolation-based reachability

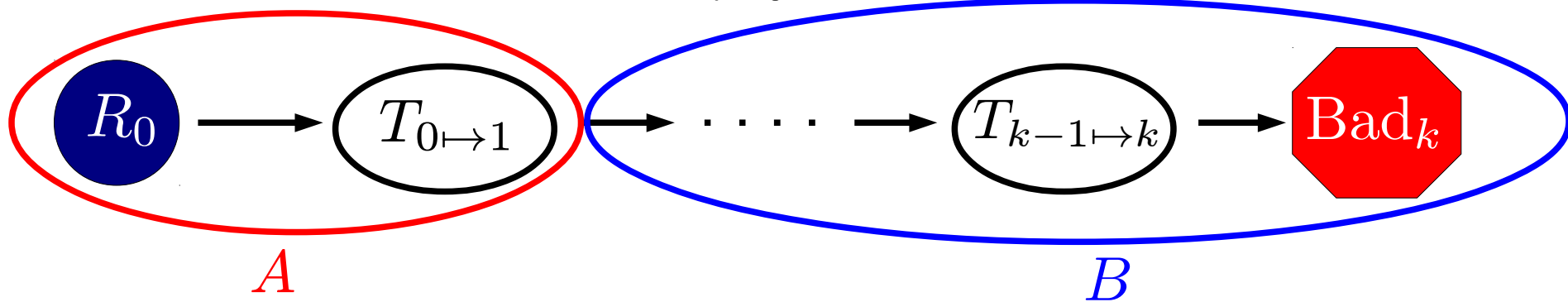
- Set $R(X) := I(X)$
- Check satisfiability of $R_0 \wedge \bigwedge_{i=0}^{k-1} T_i \wedge \text{Bad}_k$



- If **SAT**:
 - If $R \equiv I$, return **REACHABLE** the unrolling hits Bad
 - else, increase k and repeat

Interpolation-based reachability

- Set $R(X) := I(X)$
- Check satisfiability of $R_0 \wedge \bigwedge_{i=0}^{k-1} T_i \wedge \text{Bad}_k$



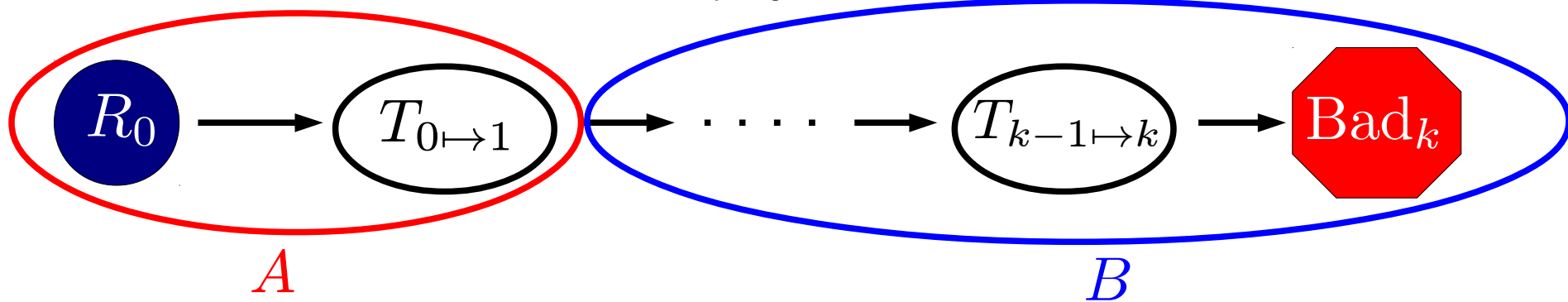
- If UNSAT:

- Set $\varphi(X) := \text{Interpolant}(A, B)[X'/X]$

φ is an abstraction of the forward image guided by the property

Interpolation-based reachability

- Set $R(X) := I(X)$
- Check satisfiability of $R_0 \wedge \bigwedge_{i=0}^{k-1} T_i \wedge \text{Bad}_k$



- If **UNSAT**:

- Set $\varphi(X) := \text{Interpolant}(A, B)[X'/X]$

φ is an **abstraction** of the forward image
guided by the property

- If $\varphi \models R$, return **UNREACHABLE** fixpoint found
- else, set $R(X) := R(X) \vee \varphi(X)$ and continue