### Proofs in Conflict-Driven Theory Combination

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- procedure for deciding the satisfiability of Boolean formulae
- ▶ uses assignments of Boolean values to variables, e.g., *I*←true

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  - can also use first-order assignments
  - models theory reasoning with modules made of inference rules

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**This paper** concerns the dual situation of **unsatisfiable** formulae: there exists a proof (of the formula's negation)

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Here, we start by adding learning mechanisms to CDSAT.

Conflict-driven theory combination

The CDSAT system - with learning

**Proof production** 

1. Conflict-driven theory combination













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It involves a trail where a putative model is being specified.

It relies on a notion of conflict between the putative model and the formula it should satisfy.

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unsatisfiable in Linear Rational Arithmetic (LRA).

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▶ No! Clash of bounds suggests a better conflict explanation,

by inferring 
$$l_0 + 2l_2$$
, i.e.,  $(-y < -2)$   
It rules out  $y \leftarrow 0$ ,

but also many values that would fail for the same reasons.

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Now undo the guess but keep I<sub>3</sub>.
Conflict-driven reasoning can be used for (other) theories

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- Now undo the guess but keep  $I_3$ .
- and so on...

(when there is no guess to undo, problem is UNSAT)

# Traditional architecture of SMT-solving



\* e.g. equality sharing / Nelson-Oppen [NO79]

In CDSAT

... the theory combination is organised directly in the main conflict-driven loop:

As in MCSAT, trail contains

- ► Boolean assignments a ← true
- ► First-order assignments y ← 3/4



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  - $a \leftarrow \mathsf{true}$
- ► First-order assignments y ← 3/4

Features of conflict-driven satisfiability:

- Boolean theory can have the same status as other theories.
- ► Theory-specific reasoning often consists of fine-grained reasoning inferences, e.g., Fourier-Motzkin resolution for LRA: (t<sub>1</sub> < x), (x < t<sub>2</sub>) ⊢ t<sub>1</sub> < t<sub>2</sub>



## 2. The CDSAT system - with learning

A set of inferences of the form

$$(t_1 \leftarrow \mathfrak{c}_1), \ldots, (t_k \leftarrow \mathfrak{c}_k) \vdash_{\mathcal{T}} (l \leftarrow \mathfrak{b})$$

where

- ► each  $t_i \leftarrow c_i$  is a single  $\mathcal{T}$ -assignment (a term  $t_i$  and a  $\mathcal{T}$ -value  $c_i$  of matching sorts)
- I←b is a single Boolean assignment
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  Abbreviations: (*I*←true) as *I* and (*I*←false) as *Ī*
- Soundness requirement: Every model of the premisses is a model of the conclusion: (t<sub>1</sub>← c<sub>1</sub>),...,(t<sub>k</sub>← c<sub>k</sub>) ⊨ (l← b)

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Examples:

$$\begin{array}{l} (x \leftarrow \sqrt{2}), (y \leftarrow \sqrt{2}) \vdash_{\mathsf{NLRA}} (x \cdot y \simeq 2) \\ (I_1 \lor \cdots \lor I_n), \overline{I_1} \ldots, \overline{I_{n-1}} \vdash_{\mathsf{Bool}} I_n \end{array}$$
(unit propagation)

What is a theory module? (Equality inferences)

All theory modules have the equality inferences:

 $t_1 \leftarrow \mathfrak{c}_1, t_2 \leftarrow \mathfrak{c}_2 \vdash_{\mathcal{T}} t_1 \simeq t_2$  if  $\mathfrak{c}_1$  and  $\mathfrak{c}_2$  are the same value  $t_1 \leftarrow \mathfrak{c}_1, t_2 \leftarrow \mathfrak{c}_2 \vdash_{\mathcal{T}} t_1 \not\simeq t_2$  if  $\mathfrak{c}_1$  and  $\mathfrak{c}_2$  are distinct values

$$\begin{array}{ccc} \vdash_{\mathcal{T}} & t_1 \simeq t_1 \\ t_1 \simeq t_2 \vdash_{\mathcal{T}} & t_2 \simeq t_1 \\ t_1 \simeq t_2, t_2 \simeq t_3 \vdash_{\mathcal{T}} & t_1 \simeq t_3 \end{array}$$

reflexivity symmetry transitivity

Search states: simply trails.

A trail is a stack of justified assignments  $_{H\vdash}(t\leftarrow \mathfrak{c})$  and decisions  $_?(t\leftarrow \mathfrak{c})$  coming from different theories

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Example (trail grows from left to right):

$$_{\emptyset \vdash}(x \simeq z), _{\emptyset \vdash}(y \simeq z), _?(x \leftarrow \sqrt{2}), _?(y \leftarrow \mathsf{blue}), _?(x \leftarrow \mathsf{red}), _{H \vdash}(x \neq y)$$

where *H* is {( $y \leftarrow blue$ ), ( $x \leftarrow red$ )}

Everything is on the trail, including assertions from the input problem, with empty justifications

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In this paper, new rule for solving/exiting conflicts: Learn

Input problem  $H_0$  including:  $(\neg l_2 \lor \neg l_4 \lor \neg l_5)$ with  $l_4 = (x \le y)$  and  $l_5 = (f(x) \le f(y))$  in a theory where f is monotonic

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First conflict: $\langle \Gamma; (\neg l_2 \lor \neg l_4 \lor \neg l_5), l_2, l_4, l_5 \rangle$ Resolving  $l_5$ : $\langle \Gamma; (\neg l_2 \lor \neg l_4 \lor \neg l_5), l_2, l_4 \rangle$ In first conflict, both  $l_4$  and  $l_5$  depend on the latest decision  $_2l_4$ .

After applying Resolve, only  $I_4$  does. Time to stop conflict analysis.

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Then Deduce can derive  $\overline{I_4}$  as before:

 $\Gamma_0, \ _?A_1, \ _?I_2, \ _{H'\vdash}(\neg I_2 \lor \neg I_4), \ _{\{(\neg I_2 \lor \neg I_4), \ I_2\}\vdash}\overline{I_4}$ 

## Example: exiting a conflict learning a clause & restarting

Input problem  $H_0$  including:  $(\neg l_2 \lor \neg l_4 \lor \neg l_5)$ with  $l_4 = (x \le y)$  and  $l_5 = (f(x) \le f(y))$  in a theory where f is monotonic Initial trail  $\Gamma_0$  including:  $\emptyset \vdash (\neg l_2 \lor \neg l_4 \lor \neg l_5)$ 

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 $\langle \Gamma; \mathbf{E} \uplus \mathbf{H} \rangle \longrightarrow \Gamma', \mathbf{E} \vdash L$  if L is a "clausal form of  $\mathbf{H}$ ",  $L \notin \Gamma, \overline{L} \notin \Gamma$ 

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"Clausal forms of H" reify H in Boolean logic:

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- generalises the CADE'2017 one (sufficient for completeness)
- models clause learning by reifying (Boolean parts of) conflicts
- models clause learning + restarts,

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Which version to apply depends on your search strategy (particularly for restarts) All version are OK with respect to termination of CDSAT

# 3. Proof production

• For every assignment  $_{H\vdash}A$  on the trail,  $H\models A$ ;

• For every conflict state  $\langle \Gamma; E \rangle$ ,  $E \models \bot$ .

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Deduce

$$\label{eq:Gamma-formula} \begin{split} \Gamma & \longrightarrow & \Gamma, \ _{J\vdash}(t {\leftarrow} \mathfrak{b}) & \text{ if } J \vdash_{\mathcal{T}} (t {\leftarrow} \mathfrak{b}) \text{ and } J \subseteq \Gamma, \\ & \text{ and } t {\leftarrow} \overline{\mathfrak{b}} \text{ is not in } \Gamma \end{split}$$

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Conflict

$$\Gamma \qquad \longrightarrow \quad \langle \Gamma; J, (t \leftarrow \overline{\mathfrak{b}}) \rangle \quad \text{if } J \vdash_{\mathcal{T}} (t \leftarrow \mathfrak{b}) \text{ and } J \subseteq \Gamma, \\ \text{and } t \leftarrow \overline{\mathfrak{b}} \text{ is in } \Gamma$$

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Resolve

 $\langle \Gamma; \mathbf{E} \uplus \{\mathbf{A}\} \rangle \longrightarrow \langle \Gamma; \mathbf{E} \cup \mathbf{H} \rangle \quad \text{if }_{\mathbf{H} \vdash} \mathbf{A} \text{ is in } \Gamma$ 

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#### Learn

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if *L* is a "clausal form" of *H*  $L \notin \Gamma$ ,  $\overline{L} \notin \Gamma$ , and  $E \subseteq \Gamma'$ 

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To keep track of the soundness invariants, we need to refer to theory inferences

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is annotated as

$$a_1(t_1 \leftarrow \mathfrak{c}_1), \ldots, a_k(t_k \leftarrow \mathfrak{c}_k) \vdash_{\mathcal{T}} j_{\mathcal{T}}: (l \leftarrow \mathfrak{b})$$

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Examples:  $a_1(x \leftarrow \sqrt{2}), a_2(y \leftarrow \sqrt{2}) \vdash_{\mathsf{NLRA}} \mathsf{eval}(\{a_1, a_2\}): (x \cdot y \simeq 2)$ (evaluation inference)  $a_0(h \lor (x \lor \lor \lor h)) = a_1(\overline{h})$ 

 $\stackrel{a_0}{(I_1 \vee \cdots \vee I_n)}, \stackrel{a_1}{(\overline{I_1})}, \dots, \stackrel{a_{k-1}}{(\overline{I_{n-1}})} \vdash_{\mathsf{Bool}} \frac{\mathsf{UP}(a_0, \{a_1, \dots, a_n\}): I_n}{(\mathsf{unit propagation})}$ 

## Proof-terms and proof-carrying CDSAT

- A proof-carrying trail is a stack
  - of justified assignments  $H \vdash_{j:} (t \leftarrow \mathfrak{c})$
  - ▶ and decisions ?(t←c)
- A proof-carrying conflict state is of the form  $\langle \Gamma; H; c \rangle$

 $\ldots$  where j and c respectively range over

Deduction proof terms j ::= in  $| j_T |$  lem(*H.c*) Conflict proof term c ::= cfl( $j_T$ , a) | res(j, <sup>a</sup>A.c)

A is an input	J ⊢ <sub>T</sub> j <sub>T</sub> : L	$E \uplus H$	$\vdash c: \perp$	,
$\emptyset \vdash in : A$	$J \vdash j_T : L$	E⊢ len	(H.c): L	1
J ⊢ <sub>T</sub> j <sub>T</sub>	: L	H ⊢ <u>j</u> :A	$E, {}^{a}A \vdash c: \bot$	
$\overline{J \cup \{{}^{a}\overline{L}\}} \vdash cfl($	$j_{\mathcal{T}}, a$ ): $\perp$	$E \cup H \vdash$	$res(j, {}^{a}A.c): \bot$	

 $\frac{A \text{ is an input}}{\emptyset \vdash \text{ in}: A} \quad \frac{J \vdash_{\mathcal{T}} j_{\mathcal{T}}: L}{J \vdash j_{\mathcal{T}}: L} \quad \frac{E \uplus H \vdash c: \bot}{E \vdash \text{ lem}(H.c): L} L \text{ clausal form of } H$  $\frac{J \vdash_{\mathcal{T}} j_{\mathcal{T}}: L}{J \cup \{{}^{a}\overline{L}\} \vdash \text{ cfl}(j_{\mathcal{T}}, a): \bot} \quad \frac{H \vdash j: A \quad E, {}^{a}A \vdash c: \bot}{E \cup H \vdash \text{ res}(j, {}^{a}A.c): \bot}$ 

Rules of CDSAT are adapted so as to use those proof-terms, and the soundness invariants are materialised as:

Theorem

- ► For every assignment  $_{H\vdash i}$  A on the trail,  $H\vdash j$ : A
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CDSAT is a search procedure for the resulting system

An SMT-problem with input clauses  $C_1, \ldots, C_n$  is treated by running CDSAT on the initial trail  $\bigcup_{n \in \mathbb{N}} C_1, \ldots, \bigcup_{n \in \mathbb{N}} C_n$ 

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But the CDSAT system can accept inputs with first-order assignments, e.g:  $(x \leftarrow 3/4)$ ,  $y \leftarrow in: (x \le y)$ ,  $y \leftarrow in: (y \le 0)$  Such problems are called SMA problems.

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If there are no first-order inputs and the problem is unsat, then the final proof-term will not mention any deduction proof-term  $H \vdash j: L$  nor any conflict proof  $H \vdash c: \bot$ such that H contains a first-order assignment

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Easy optimisation in that case:

the construction of any such proof-term during the run can be omitted Theory modules do not have to provide theory proofs  $H \vdash_{\mathcal{T}} j_{\mathcal{T}}: L$ if H contains a first-order assign. (typically: evaluation inferences)

### Different views about proof objects

Proof-carrying CDSAT can be considered exactly as defined above, where in,  $j_{\mathcal{T}}$ , lem(*H*.*c*), cfl( $j_{\mathcal{T}}$ , *a*), res(j, <sup>*a*</sup>*A*.*c*) are terms.

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Another proof format is desired for output? Just interpret the terms in that format after the run (proof reconstruction) Different views about proof objects

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Another proof format is desired for output? Just interpret the terms in that format after the run (proof reconstruction)

Alternatively, proof-carrying CDSAT can directly manipulate proofs in the format, if equipped with the operations corresponding to the term constructs. The proof-terms *denote* the manipulated proofs,

but are never constructed.

## Example: resolution proofs

If input contains no first-order assignments, resolution trees (or DAGs) form a proof format equipped with the right operations

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Leaves of resolution proofs are labeled by

- either literals corresponding to input assignments  $\emptyset \vdash in : A$
- or theory lemmas corresponding to theory proofs  $J \vdash_{\mathcal{T}} j_{\mathcal{T}} : L$ Internal nodes are obtained by applying resolution rule,

corresponding to  $H \vdash \operatorname{res}(j, {}^{a}A.c) : \bot$  constructs.

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Leaves of resolution proofs are labeled by

• either literals corresponding to input assignments  $\emptyset \vdash in : A$ 

► or theory lemmas corresponding to theory proofs  $J \vdash_{\mathcal{T}} j_{\mathcal{T}} : L$ Internal nodes are obtained by applying resolution rule, corresponding to  $H \vdash \operatorname{res}(j, {}^{a}A.c) : \bot$  constructs.

If input does contains first-order assignments (SMA problems) the resolution format has to be slightly extended, so that it manipulates guarded clauses of the form

$$\{(t_1 \leftarrow \mathfrak{c}_1), \ldots, (t_n \leftarrow \mathfrak{c}_n)\} \Rightarrow C$$

where  $(t_1 \leftarrow c_1), \ldots, (t_n \leftarrow c_n)$  are first-order assign. guarding clause C Details in the paper.

Other "proof format":

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#### LCF: answers that are correct-by-construction

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No proof object needs to be built in memory

#### CDSAT is well-suited to the LCF approach 1/2

Given a type **assign** for multiple assignments and **single\_assign** for singleton assignments, a trusted kernel defines

```
type deduction = assign*single_assign
type conflict = assign
```

and exports

### CDSAT is well-suited to the LCF approach 2/2

If the empty assignment is constructed in type conflict, input problem is guaranteed to be unsat, provided the kernel primitives and the implementation of theory proofs are trusted (code for the search plan does not have to be certified)

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If the empty assignment is constructed in type conflict, input problem is guaranteed to be unsat, provided the kernel primitives and the implementation of theory proofs are trusted (code for the search plan does not have to be certified)

Answer is correct-by-construction, no proof object in memory.

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### Ongoing and future work

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   Currently working on more performance-driven implementation.
- Issue of cost: Penalty of building proof-terms? Penalty of having a code developed in the correct-by-construction approach?
- Use proof-terms for interpolation? (See Tanja Schindler's talk on interpolation in a related context! 16:30 at VMCAI)

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# Adapting the rules

Deduce	
$\Gamma \longrightarrow \ \Gamma, \ _{J \vdash j_{\mathcal{T}}:}(t \leftarrow \mathfrak{b})$	if $J \vdash_{\mathcal{T}} \underline{j_{\mathcal{T}}}: (t \leftarrow \mathfrak{b}), \ J \subseteq \Gamma$ , and $t \leftarrow \overline{\mathfrak{b}}$ is not in $\Gamma$
Conflict	
$\Gamma \longrightarrow \langle \Gamma; J, (t \leftarrow \overline{\mathfrak{b}}); cfl(j_k, a) \rangle$	if $J \vdash_{\mathcal{T}} \underline{j_{\mathcal{T}}}: (t \leftarrow \mathfrak{b}), \ J \subseteq \Gamma$ , and $t \leftarrow \overline{\mathfrak{b}}$ is in $\Gamma$ with id a
Resolve	
$\langle \Gamma; E \uplus \{A\}; c \rangle \longrightarrow \langle \Gamma; E \cup H; \operatorname{res}(j,$	$^{a}A.c)\rangle$
	if $H \vdash_{j:} A$ is in $\Gamma$ with id $a$
Learn	
$\langle \Gamma; E \uplus H; c \rangle \longrightarrow \Gamma', E \vdash \operatorname{lem}(H.c): L$	if <i>L</i> is a "clausal form" of <i>H</i> $L \notin \Gamma$ , $\overline{L} \notin \Gamma$ , and $E \subseteq \Gamma'$