



# Conflict-Driven Satisfiability for Theory Combination: Lemmas, Modules, and Proofs

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## Abstract

Search-based satisfiability procedures try to build a model of the input formula by simultaneously proposing candidate models and deriving new formulae implied by the input. *Conflict-driven* procedures perform non-trivial inferences only when resolving conflicts between formulae and assignments representing the candidate model. CDSAT (*Conflict-Driven SATisfiability*) is a method for conflict-driven reasoning in *unions of theories*. It combines inference systems for individual theories as *theory modules* within a solver for the union of the theories. This article augments CDSAT with a more general *lemma learning* capability and with *proof generation*. Furthermore, theory modules for several theories of practical interest are shown to fulfill the requirements for *completeness* and *termination* of CDSAT. Proof generation is accomplished by a *proof-carrying* version of the CDSAT transition system that produces *proof objects* in memory accommodating multiple proof formats. Alternatively, one can apply to CDSAT the *LCF approach to proofs* from interactive theorem proving, by defining a kernel of reasoning primitives that guarantees the correctness by construction of CDSAT proofs.

**Keywords** Lemma learning · Proof generation · Satisfiability modulo theories · Satisfiability modulo assignment

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## 1 Introduction

The satisfiability problem is one of checking if a given formula has a model. In the propositional case (SAT) the input is usually a formula in conjunctive normal form (a set of clauses), and a model is an assignment of truth values to propositional variables that satisfies all the clauses. Many SAT solvers employ a *conflict-driven search strategy*, known as Conflict-Driven Clause Learning (CDCL), in which the solver extends a partial assignment until it satisfies all clauses, or a conflict arises as the assignment falsifies a clause. Non-trivial inference steps are performed in response to a conflict to roll back the partial assignment and direct the search elsewhere [44,45]. This conflict-driven style inspired the design of several solvers for quantifier-free fragments of arithmetic (e.g., [8,16,17,21,36,37,40,46] for more references). These *conflict-driven theory solvers* decide the satisfiability of sets of literals in the theory.

The problem of deciding the satisfiability of a quantifier-free formula in a theory is known as Satisfiability Modulo a Theory (SMT). MCSAT, for *Model Constructing SATisfiability*, integrates a CDCL-based SAT solver and a conflict-driven model-constructing theory solver [6,24,31,34,35,56]. CDSAT, for *Conflict-Driven SATisfiability*, generalizes MCSAT to a *generic* union of *disjoint* theories whose solvers may or may not be model-constructing [11].

In CDSAT, both Boolean and *first-order* terms are given assignments in a trail representing the candidate model as a partial assignment. First-order terms are assigned constant symbols representing individuals of the corresponding sort in a model's domain (e.g., integer terms are assigned integer constants). Since CDSAT accepts such first-order assignments also as part of the input, CDSAT is an engine to determine the satisfiability of a quantifier-free formula modulo a union of theories (SMT) and possibly *modulo an initial assignment* of values to terms appearing in the input formula. We call this generalization of SMT *satisfiability modulo assignment* (SMA).

CDSAT is presented as a *transition system* that combines *multiple* theory solvers, all or some of which are conflict-driven, into a conflict-driven solver for the union of the theories. More precisely, CDSAT combines *theory inference systems*, called *theory modules*, and performs the conflict-driven search for all theories. A theory module is an abstraction of a theory solver. Propositional logic is regarded as one of the theories. Every theory module can *expand* the trail in two ways: either by deciding the value of a term or by performing an inference. Theory inferences are applied to *propagate* consequences of the assignments on the trail, to *detect* conflicts, and to *explain* such conflicts, all in the respective theory. A conflict in CDSAT is a set of assignments from the trail that is unsatisfiable. The inferences can be used to transform the conflict into a Boolean one, susceptible of conflict analysis. The analysis *solves* the conflict, producing a lemma and undoing some assignments on the trail.

In conflict-driven reasoning, it is essential that the system *learns a lemma* from a solved conflict, because the lemma immediately thwarts any attempt to repeat a failed search path. In CDSAT [11], lemma learning is limited to the case of backjumping that simply flips the truth assignment of a Boolean term that was involved in the conflict. The first contribution of the present article is a CDSAT transition system with a *more general and more flexible lemma learning capability*. The new lemma learning mechanism subsumes the old one, allows both learn-and-backjump and learn-and-restart, and it enables CDSAT to form and learn new clauses. With this addition, CDSAT reduces to CDCL, if propositional logic is the only theory, and to MCSAT, if propositional logic and another theory with a conflict-driven solver are the only theories.

The theory modules need to satisfy a completeness property that is strong enough to ensure that CDSAT determines whether the input problem has a model in the union of the theories. For such a model to exist, the theories need to agree on what they have in common. As disjoint theories only share equality and sorts, the theories need to agree on which shared terms are equal and on the cardinalities of shared sorts. The standard approach to this problem in the literature is the *equality-sharing (Nelson-Oppen) scheme* (e.g., [9,23,42,48,49]) for a survey covering also several extensions). In order to reach an agreement on which shared terms are equal, each theory solver propagates the (disjunctions of) equalities between shared variables that are entailed by its part of the problem. For cardinalities, the equality-sharing method requires the theories to be *stably infinite* (every satisfiable formula has a model with countably infinite domain), so that the shared sorts can be interpreted as countably infinite domains.

For model-constructing theory solvers the equality-sharing scheme can be implemented by *model-based theory combination* (MBTC) [23]. In MBTC a theory solver may decide an equality, between terms occurring in the problem, that is true in its candidate model, even if it is not entailed by its part of the problem. If it turns out later that such an equality is not entailed, it will cause a conflict, so that the responsible solver will retract it and amend its model. MBTC was born out of the observation that, especially when the input is found satisfiable, it is generally less expensive for a theory solver to enumerate the equalities satisfied in a candidate theory model than those entailed.

In the equality-sharing method, including MBTC, solvers are combined as *black-boxes*. If a conflict-driven model-constructing theory solver is included, as in MBTC, its model-constructing and conflict-driven operations remain hidden inside the black-box. When combination of theories by equality sharing or MBTC is integrated with the CDCL procedure in the DPLL( $\mathcal{T}$ ) or CDCL( $\mathcal{T}$ ) paradigm (e.g., [13,23,42,50]), the candidate model on the trail and the public conflict-driven reasoning is propositional. The CDCL procedure plays a central role, while the theory solvers are satellites that signal theory conflicts or submits theory lemmas to CDCL.

In CDSAT all theory modules, including a CDCL module for propositional logic, cooperate as peers to build a model for the union of the theories on the shared trail, and the conflict-driven reasoning happens in the union of the theories. Each theory module has a *view* of the shared trail, comprising its theory assignments as well as equalities or disequalities implied by assignments of other theories. The idea of MBTC is subsumed, since in CDSAT any theory module can decide an equality that is true in its view of the shared trail. Furthermore, CDSAT does *not* require stable infiniteness, provided there is a *leading theory*, its module is *complete*, and the other modules are *leading-theory-complete*. The leading theory knows all the sorts in the union of theories, and it aggregates any constraints that the theories may impose on the cardinality of shared sorts. The aggregated constraints are enforced as axioms or theorems of the leading theory and as inference rules of its module. A theory module is *complete* if it can expand any assignment that is not satisfied by a model of its theory. *Leading-theory-completeness* also requires that the module can expand an assignment, if its theory model does not agree on cardinalities of shared sorts and equality of shared terms with a model of the leading theory.

In previous work [11] we showed that if the theory modules are *sound* and *leading-theory-complete*, CDSAT is *sound* and *complete*. Furthermore, we exemplified the notion of theory module by listing theory modules for *propositional logic*, also known as the Boolean theory (Bool), the *theory of equality with uninterpreted function symbols* (EUF), the *theory of arrays* (Arr), and *linear rational arithmetic* (LRA). A non-conflict-driven solver can be abstracted into a *black-box theory module*, whose only inference rule invokes the solver

to detect a theory conflict on the trail. However, it was not shown that these modules are leading-theory-complete. The second contribution of the present article is a *collection of completeness theorems* showing that the above theory modules are leading-theory-complete for all suitable leading theories. Moreover, we prove that if all modules are black-boxes, CDSAT emulates the equality-sharing method (covering also MBTC), and we demonstrate the role of the leading theory by considering the case where at-most cardinality constraints need to be enforced.

A key difference between conflict-driven theory reasoning and conflict-driven propositional reasoning is that theory inference rules may explain conflicts by inferences that generate *new* (i.e., non-input) terms. If the transition system allows this kind of expansion, *termination* requires that all new terms come from a *finite basis* (e.g., [24]). For conflict-driven reasoning in a union of theories, this issue must be approached *locally*, because no inference system should be authorized to generate infinitely many terms, and *globally*, because the interaction of multiple inference systems should preserve finiteness. In previous work [11] we showed that if every theory module is equipped with a *finite local basis* for its theory, and a *finite global basis* for the union of the theories does exist, CDSAT is guaranteed to halt. However, finite local bases for the above modules were not exhibited. The third contribution of the present article is a *collection of finite local bases* for those modules, with a technique for the *generic construction of a finite global basis* from given finite local bases.

Proofs are important in SMT/SMA, because many applications require the solver to generate either a satisfying assignment or a proof of unsatisfiability. CDCL-based SAT solvers generate proofs by resolution [57]. Since such proofs are huge, more sophisticated and compact proof formats have been investigated (e.g., [22,29,32,33]). The DPLL( $\mathcal{T}$ ) or CDCL( $\mathcal{T}$ ) paradigm naturally supports the generation of proofs by resolution, where the theory lemmas are plugged in as leaves with black-box subproofs [3,12,27,38]. This style has been implemented in solvers such as Z3 [3], VERiT [2,27], and CVC4 [38] and extended in several ways (e.g., [2,38]). In CDSAT, the CDCL-based SAT solver loses its centrality as the only conflict-driven component, and all theory modules contribute directly to the proof, including new terms. Even if propositional resolution with theory subproofs is chosen as the final proof format, CDSAT proofs cannot be reconstructed in the same way as CDCL( $\mathcal{T}$ ) proofs, because the structure and the operations of CDSAT differ from those of CDCL( $\mathcal{T}$ ). The fourth contribution of the present article comprises *two approaches to generate CDSAT proofs*.

The first approach is a *proof-carrying CDSAT transition system*, where *proof terms* record the information needed to generate proofs. We describe different ways to turn proof terms into proofs, including producing resolution proofs with theory lemmas. These proof objects can then be checked directly by a verified checker [53] or exported to a proof format verifiable by proof checkers. Thus, proof-carrying CDSAT can slot into pipelines from proof-search to proof-checking [1,5,7], where a minimal amount of proof information (e.g., an unsatisfiable core) may be sufficient for a theorem prover to regenerate a proof in its own format. The second approach works by specifying a small *kernel* of primitives in LCF style [30,47], so that building proof objects in memory can be avoided. If CDSAT is implemented on top of this kernel, the LCF-type abstraction ensures that an *unsat* answer is correct by construction, and CDSAT can be used as a trusted external oracle for interactive proof tools.

In summary, the original contributions of the present article include:

1. An extension of the CDSAT transition system with a *more general and more flexible lemma learning capability*;

2. Definitions of *finite local bases* and proofs of *leading-theory-completeness* of the modules for Bool, EUF, Arr, and LRA, as well as for a generic black-box module, and a generic module for at-most cardinality constraints;
3. A general technique to *construct a finite global basis* for a union of theories from finite local bases of the theories; and
4. Approaches to endow CDSAT with *proof generation* either by producing proof objects in memory or in LCF style.

This article is organized as follows. Section 2 contains basic definitions for CDSAT. Section 3 is subdivided in three parts: Sect. 3.1 describes the CDSAT transition system with *enhanced lemma learning*; Sect. 3.2 illustrates via examples the novel lemma learning capabilities; and Sect. 3.3 presents other definitions, including that of *leading-theory-completeness*, discussing how the *soundness*, *completeness*, and *termination* results for CDSAT [11] extend to the transition system in Sect. 3.1. Section 4 presents theory modules, *local finite bases*, and *leading-theory-completeness theorems* for Bool, LRA, EUF, Arr, a generic stably infinite theory, and a generic theory with at-most cardinality constraints. Section 5 portrays the technique to get a global basis from local ones. Sections 6 and 7 cover the two approaches to proof generation in CDSAT.

Lemma learning and proof generation for CDSAT appeared in a conference version [10] of the present article.

## 2 Basic Definitions

Let  $\mathcal{T}_1, \dots, \mathcal{T}_n$  be *disjoint theories*, each defined by its signature  $\Sigma_k = (S_k, F_k)$  and axiomatization  $\mathcal{A}_k$ , where  $S_k$  is the set of sorts and  $F_k$  is the set of symbols, for all  $k, 1 \leq k \leq n$ . Every theory has the sort *prop* of the Boolean values and sorted equality symbols:  $\simeq_S = \{\simeq_s : s \times s \rightarrow \text{prop} \mid s \in S_k\} \subseteq F_k$ . The sorts of equalities may be omitted. Disjointness means that the theories do not share symbols except equality on shared sorts. Often one of the theories is the Boolean theory Bool, with the logical connectives  $\neg, \wedge,$  and  $\vee$  as symbols. Formulae are terms of sort *prop*. The union of  $\mathcal{T}_1, \dots, \mathcal{T}_n$  is denoted  $\mathcal{T}_\infty$ , with signature  $\Sigma_\infty = (S_\infty, F_\infty)$ , where  $S_\infty = \bigcup_{k=1}^n S_k$  and  $F_\infty = \bigcup_{k=1}^n F_k$ , and axiomatization  $\bigcup_{k=1}^n \mathcal{A}_k$ .

Let  $\mathcal{T}, \Sigma,$  and  $S$  stand for  $\mathcal{T}_k, \Sigma_k,$  and  $S_k$  ( $1 \leq k \leq n$ ), or for  $\mathcal{T}_\infty, \Sigma_\infty,$  and  $S_\infty$ . We assume a collection  $\mathcal{V} = (\mathcal{V}^s)_{s \in S}$  of disjoint sets of *variables*, where  $\mathcal{V}^s$  is the set of variables of sort  $s$ . We use  $x, y,$  and  $z$  for variables,  $t$  and  $u$  for terms of any sort,  $l$  and  $p$  for formulae, and  $\sqsubseteq$  for the *subterm ordering*. If  $\Sigma = (S, F)$  is a signature with  $F \subseteq F_\infty$ , the  $\Sigma$ -foreign subterms of a term  $t$  are those subterms whose root symbol is not in  $F$ , including variables. Non-variable  $\Sigma$ -foreign subterms can be regarded as variables, without replacing them explicitly with new variables. This is accomplished by defining the *free  $\Sigma$ -variables* of  $t$  as its  $\Sigma$ -foreign subterms with a  $\sqsubseteq$ -maximal occurrence. For a term  $t$ , the set of its free  $\Sigma$ -variables is denoted  $\text{fv}_\Sigma(t)$ , and the set of its free  $\Sigma$ -variables of sort  $s$  is denoted  $\text{fv}_\Sigma^s(t)$ . For a set  $X$  of terms,  $\text{fv}_\Sigma(X) = \{u \mid u \in \text{fv}_\Sigma(t), t \in X\}$  and  $\text{fv}_\Sigma^s(X) = \{u \mid u \in \text{fv}_\Sigma^s(t), t \in X\}$ .

A  $\mathcal{T}[\mathcal{V}]$ -model  $\mathcal{M}$  interprets each  $s \in S$  as a non-empty domain  $s^\mathcal{M}$  with  $\text{prop}^\mathcal{M} = \{\text{true}, \text{false}\}$ , each  $v \in \mathcal{V}^s$  as an element  $v^\mathcal{M}$  in  $s^\mathcal{M}$ , each  $f \in F$  with  $f : (s_1 \times \dots \times s_m) \rightarrow s$  as a function  $f^\mathcal{M}$  from  $s_1^\mathcal{M} \times \dots \times s_m^\mathcal{M}$  to  $s^\mathcal{M}$ , and each  $\simeq_s$  as the function  $\simeq_s^\mathcal{M}$  from  $s^\mathcal{M} \times s^\mathcal{M}$  to  $\{\text{true}, \text{false}\}$  that returns true if and only if its arguments are the same element. The interpretation of terms and formulae is defined as usual, with the interpretation of term  $t$  denoted  $\mathcal{M}(t)$ . We write  $\mathcal{T}$ -model when the variables do not matter.

CDSAT works with *assignments* that assign to terms *values* of the appropriate sort. For example, assuming theories Bool, Arr, and a fragment of arithmetic,  $((x > 1) \vee (y < 0)) \leftarrow \text{true}$ ,  $y \leftarrow -1$ ,  $z \leftarrow \sqrt{2}$ ,  $(\text{store}(a, i, v) \simeq b) \leftarrow \text{true}$ ,  $\text{select}(a, j) \leftarrow 3$ , and  $(\text{select}(a, j) \simeq v) \leftarrow \text{true}$  are assignments. The standard approach to define what the *values* are is to extend the signature with sorted constant symbols to name all individuals in the domains used to interpret the sorts (e.g., the appropriate set of numerals for a fragment of arithmetic).

For each  $\mathcal{T}_k$ ,  $1 \leq k \leq n$ , a *conservative theory extension*  $\mathcal{T}_k^+$  is a theory with signature  $\Sigma_k^+ = (S_k, F_k^+)$ , where  $F_k^+$  adds to  $F_k$  a possibly empty set of *new* constant symbols, called  $\mathcal{T}_k$ -*values*, accompanied by new axioms as needed (e.g.,  $\sqrt{2}$  with  $\sqrt{2} \cdot \sqrt{2} \simeq 2$ ). For numerals, as for true and false, a  $\mathcal{T}_k$ -value is both the domain element and the constant symbol that names it.  $F_k^+$  may be infinite, but it is countable (e.g., using the algebraic reals as real numbers). The *trivial extension* only adds {true, false} as  $\mathcal{T}_k$ -values. We assume that the extended theories are still *disjoint* except for true and false.

The union of  $\mathcal{T}_1^+, \dots, \mathcal{T}_n^+$  is a conservative extension  $\mathcal{T}_\infty^+$  of  $\mathcal{T}_\infty$ , with signature  $\Sigma_\infty^+ = (S_\infty, F_\infty^+)$  for  $F_\infty^+ = \bigcup_{k=1}^n F_k^+$ . *Conservativity* means that  $\mathcal{T}^+$ -unsatisfiability implies  $\mathcal{T}$ -unsatisfiability for  $\Sigma$ -formulae: if CDSAT detects  $\mathcal{T}_\infty^+$ -unsatisfiability, the problem is  $\mathcal{T}_\infty$ -unsatisfiable; if the problem is  $\mathcal{T}_\infty$ -satisfiable, there is a  $\mathcal{T}_\infty^+$ -model that CDSAT can discover. The symbols c and q, possibly with subscripts, are used for values, reserving b for true or false.

Recalling that  $\mathcal{T}$  stands for either a  $\mathcal{T}_k$  ( $1 \leq k \leq n$ ) or  $\mathcal{T}_\infty$ , a  $\mathcal{T}$ -assignment is an assignment of  $\mathcal{T}$ -values to  $\mathcal{T}_\infty$ -terms. Formally, a  $\mathcal{T}$ -assignment is a set  $J = \{u_1 \leftarrow c_1, \dots, u_m \leftarrow c_m\}$ , where, for all  $i$ ,  $1 \leq i \leq m$ ,  $u_i$  is a  $\mathcal{T}_\infty$ -term and  $c_i$  a  $\mathcal{T}$ -value of the same sort. The set of terms that occur in  $J$  is  $G(J) = \{t \mid t \trianglelefteq u_i, 1 \leq i \leq m\}$ , and  $G_s(J)$  is the subset of the terms of sort  $s$  in  $G(J)$ . The set of free variables of  $J$  is  $\text{fv}_\Sigma(J) = \{u \mid u \in \text{fv}_\Sigma(t), (t \leftarrow c) \in J\}$ . We use  $J$  for generic  $\mathcal{T}$ -assignments,  $A$  for generic singleton assignments,  $L$  or  $K$  for Boolean singleton assignments,  $H$  and  $E$  for  $\mathcal{T}_\infty$ -assignments.

The *flip*  $\bar{L}$  of  $L$  assigns to the same formula the opposite Boolean value. Since  $\neg$  is a function symbol in the signature of Bool, one can write  $l \leftarrow \text{true}$  and  $l \leftarrow \text{false}$ , that are one the flip of the other, and also  $\neg l \leftarrow \text{true}$  and  $\neg l \leftarrow \text{false}$ , that are also one the flip of the other. Clearly,  $l \leftarrow \text{true}$  and  $\neg l \leftarrow \text{false}$  are equivalent and so are  $l \leftarrow \text{false}$  and  $\neg l \leftarrow \text{true}$ . The simplest form is preferred when writing assignments: for example, if  $l$  is  $\neg a$ , where  $a$  is a propositional variable, it is preferable to use  $a \leftarrow \text{true}$  and  $a \leftarrow \text{false}$ . Furthermore,  $l \leftarrow \text{true}$  is abbreviated as  $l$ ,  $l \leftarrow \text{false}$  as  $\bar{l}$ , and  $(t \simeq u) \leftarrow \text{false}$  as  $t \not\approx u$ .

An assignment is *plausible* if for no  $L$  it contains both  $L$  and  $\bar{L}$ . A *Boolean* assignment only assigns Boolean values, while a *first-order* assignment only assigns non-Boolean values. An *SMT problem* is presented as a plausible Boolean assignment  $\{l_1 \leftarrow \text{true}, \dots, l_m \leftarrow \text{true}\}$ , abbreviated  $\{l_1, \dots, l_m\}$ , while an *SMA problem* also includes first-order assignments.

The *theory view*, or  $\mathcal{T}$ -view,  $H_{\mathcal{T}}$  of a  $\mathcal{T}_\infty$ -assignment  $H$  comprises the  $\mathcal{T}$ -assignments in  $H$  and all equalities or disequalities between terms of a sort in  $S$  that are entailed by first-order assignments in  $H$ . If  $\{x \leftarrow 3, y \leftarrow 3, z \leftarrow 4\} \subseteq H$ , the  $\mathcal{T}$ -view  $H_{\mathcal{T}}$  also includes  $x \simeq y$ ,  $x \not\approx z$ , and  $y \not\approx z$ , for every  $\mathcal{T}$  having the sort of  $x$ ,  $y$ , and  $z$ . If  $H$  is Boolean,  $H_{\mathcal{T}} = H$ . As a  $\mathcal{T}_i$ -assignment ( $1 \leq i \leq n$ ) is a special case of  $\mathcal{T}_\infty$ -assignment, the  $\mathcal{T}$ -view of a  $\mathcal{T}_i$ -assignment is also defined.

A  $\mathcal{T}^+$ -model  $\mathcal{M}$  endorses a  $\mathcal{T}$ -assignment  $J$ , written  $\mathcal{M} \models J$ , if  $\mathcal{M}$  satisfies  $u \simeq c$  for all pairs  $(u \leftarrow c) \in J$ . It follows that if  $\{u \leftarrow c, t \leftarrow c\} \subseteq J$ , then  $\mathcal{M}$  also satisfies  $u \simeq t$ . If  $\mathcal{M} \models J_{\mathcal{T}}$ , that is,  $\mathcal{M}$  endorses the  $\mathcal{T}$ -view of  $J$ , then  $\mathcal{M}$  also satisfies  $u \not\approx t$ , for all pairs  $u \leftarrow c_1$  and  $t \leftarrow c_2$  in  $J$  with  $c_1 \neq c_2$ . Thus,  $\mathcal{M} \models J_{\mathcal{T}}$  is generally stronger than  $\mathcal{M} \models J$ .

A  $\mathcal{T}$ -assignment  $J$  is *satisfiable*, if there is a  $\mathcal{T}^+$ -model  $\mathcal{M}$  such that  $\mathcal{M} \models J_{\mathcal{T}}$ , and it is *unsatisfiable* otherwise, written  $J \models \perp$ . We write  $J \models L$  if  $\mathcal{M} \models L$  for all  $\mathcal{T}^+$ -models  $\mathcal{M}$  such that  $\mathcal{M} \models J_{\mathcal{T}}$ . All this applies to a  $\mathcal{T}_{\infty}$ -assignment  $H$ , for which we say that  $\mathcal{M}$  *globally endorses*  $H$  if  $\mathcal{M} \models H_{\mathcal{T}_{\infty}}$ , also written  $\mathcal{M} \models^G H$  to emphasize “globally.”

A *theory module*  $\mathcal{I}_k$  for theory  $\mathcal{T}_k$  ( $1 \leq k \leq n$ ) is an inference system with inferences of the form  $J \vdash_{\mathcal{I}_k} L$ , or  $J \vdash_k L$  for short, where  $J$  is a  $\mathcal{T}_k$ -assignment and  $L$  is a Boolean assignment. Theory modules are required to be *sound*: if  $J \vdash_k L$  then  $J \models L$ . In the sequel, *assignment* stands for  $\mathcal{T}_{\infty}$ -assignment.

### 3 CDSAT with Lemma Learning

In this section we present the *CDSAT transition system with lemma learning*. Section 3.1 presents the CDSAT transition system as in [11], except that the Backjump rule of [11] is replaced by a new LearnBackjump rule that introduces a more general and more flexible lemma learning mechanism. Section 3.2 analyzes in detail the working of LearnBackjump: it is more general because it enables CDSAT to learn new clauses, whereas Backjump only flips a Boolean term; it is more flexible because it enables a CDSAT search plan to choose the destination level upon backjumping; and it can simulate the Backjump rule. Section 3.3 includes the definitions of *basis* and *leading-theory-completeness* (from [11]), and discusses how the arguments of the proofs of *soundness*, *termination*, and *completeness* of CDSAT in [11] are modified to have LearnBackjump in place of Backjump.

#### 3.1 The CDSAT Transition System with Lemma Learning

CDSAT works with a *trail*  $\Gamma$ , defined as a sequence of distinct singleton assignments that are either *decisions* or *justified assignments*. A *decision* is written  $\gamma A$  to convey guessing, and it can be either a Boolean or a first-order assignment. A *justified assignment* is written  ${}_{H\vdash}A$ , where  $H$ , the *justification* of  $A$ , is a set of singleton assignments that appear before  $A$  in  $\Gamma$ . The elements of the *input assignment*  $H_0$  are listed in  $\Gamma$  as justified assignments with empty justification. The only justified assignments that are first-order assignments are the input first-order assignments of an SMA problem; all non-input justified assignments are Boolean. A non-input justified assignment  ${}_{J\vdash}L$  is due to either an inference  $J \vdash_k L$  for some theory  $\mathcal{T}_k$ ,  $1 \leq k \leq n$ , or a conflict-solving transition. A justified assignment  ${}_{H\vdash}A$  is *sound* if for all  $\mathcal{T}_{\infty}^+$ -models  $\mathcal{M}$ , if  $\mathcal{M} \models^G H_0 \cup H$  then  $\mathcal{M} \models A$ . A trail can be seen as an *assignment* by ignoring order and justifications.

Given a trail  $\Gamma$  with assignments  $A_0, \dots, A_m$ , the *level* of a singleton assignment  $A_i$ ,  $0 \leq i \leq m$ , is given by  $\text{level}_{\Gamma}(A_i) = 1 + \max\{\text{level}_{\Gamma}(A_j) \mid j < i\}$ , if  $A_i$  is a decision, and  $\text{level}_{\Gamma}(A_i) = \text{level}_{\Gamma}(H)$ , if  $A_i$  is a justified assignment  ${}_{H\vdash}A_i$ . The *level* of a set of singleton assignments  $H \subseteq \Gamma$  is given by  $\text{level}_{\Gamma}(H) = 0$ , if  $H = \emptyset$ , and  $\text{level}_{\Gamma}(H) = \max\{\text{level}_{\Gamma}(A) \mid A \in H\}$ , otherwise. As the level of  ${}_{H\vdash}A$  depends on its justification, not on its position on the trail, the trail is *not* organized as a stack, and  ${}_{H\vdash}A$  can be added to the trail *after* assignments of greater level. This behavior and assignment  $A$  are called *late propagation*.  $\Gamma^{\leq m}$  denotes the restriction of  $\Gamma$  to its elements of level at most  $m$ .

The *state* of a CDSAT-derivation is either a *trail*  $\Gamma$  or a *conflict state*  $\langle \Gamma; E \rangle$ , where  $\Gamma$  is a trail, and  $E$  is a *conflict*, that is, an assignment such that  $E \subseteq \Gamma$  and  $H_0 \cup E \models \perp$ . The CDSAT transition system features *trail rules*, denoted  $\longrightarrow$ , and *conflict-state rules*, denoted  $\Longrightarrow$ , with transitive closure  $\Longrightarrow^*$  and  $\uplus$  for disjoint union (see Fig. 1). As CDSAT may place

TRAIL RULES		
In the trail rules, let $1 \leq k \leq n$ .		
Decide	$\Gamma \longrightarrow \Gamma, ?A$	if $A$ is an acceptable $\mathcal{T}_k$ -assignment for $\mathcal{I}_k$ in $\Gamma_{\mathcal{T}_k}$
The next three rules share the conditions: $J \subseteq \Gamma$ , $(J \vdash_k L)$ , and $L \notin \Gamma$ .		
Deduce	$\Gamma \longrightarrow \Gamma, J \vdash L$	if $\bar{L} \notin \Gamma$ and $L$ is in $\mathcal{B}$
Fail	$\Gamma \longrightarrow \text{unsat}$	if $\bar{L} \in \Gamma$ and $\text{level}_\Gamma(J \cup \{\bar{L}\}) = 0$
ConflictSolve	$\Gamma \longrightarrow \Gamma'$	if $\bar{L} \in \Gamma$ , $\text{level}_\Gamma(J \cup \{\bar{L}\}) > 0$ , and $\langle \Gamma; J \cup \{\bar{L}\} \rangle \Longrightarrow^* \Gamma'$
CONFLICT STATE RULES		
UndoClear	$\langle \Gamma; E \uplus \{A\} \rangle \Longrightarrow \Gamma^{\leq m-1}$	if $A$ is a first-order decision of level $m > \text{level}_\Gamma(E)$
Resolve	$\langle \Gamma; E \uplus \{A\} \rangle \Longrightarrow \langle \Gamma; E \cup H \rangle$	if $(H \vdash A) \in \Gamma$ and for no first-order decision $A' \in H$ $\text{level}_\Gamma(A') = \text{level}_\Gamma(E \uplus \{A\})$
UndoDecide	$\langle \Gamma; E \uplus \{L\} \rangle \Longrightarrow \Gamma^{\leq m-1}, ?\bar{L}$	if $(H \vdash L) \in \Gamma$ and for a first-order decision $A' \in H$ $m = \text{level}_\Gamma(E) = \text{level}_\Gamma(L) = \text{level}_\Gamma(A')$
LearnBackjump	$\langle \Gamma; E \uplus H \rangle \Longrightarrow \Gamma^{\leq m}, E \vdash L$	if $L$ is a clausal form of $H$ , $L$ is in $\mathcal{B}$ , $L \notin \Gamma$ , $\bar{L} \notin \Gamma$ , and $\text{level}_\Gamma(E) \leq m < \text{level}_\Gamma(H)$

**Fig. 1** The CDSAT transition system with lemma learning

on the trail assignments for *new* (i.e., non-input) terms, for termination all terms must come from a *finite* set  $\mathcal{B}$ , called *global basis*, which is determined based on the input and does not change during the derivation. While terms come from  $\mathcal{B}$ , values come from  $F_\infty^+$ , which may be infinite: a derivation will use a finite subset of  $F_\infty^+$  that is not fixed beforehand. An assignment  $H$  is in  $\mathcal{B}$  if  $t \in \mathcal{B}$  for all  $(t \leftarrow c) \in H$ .

Rule **Decide** adds a decision  $u \leftarrow c$  if it is *acceptable* for a theory module  $\mathcal{I}_k$  in its view  $\Gamma_{\mathcal{T}_k}$  of trail  $\Gamma$ . Acceptability comprises three requirements: (1)  $\Gamma$  does not assign a  $\mathcal{T}_k$ -value to  $u$ ; (2) if  $u \leftarrow c$  is first-order, there is no inference  $J \cup \{u \leftarrow c\} \vdash_{\mathcal{I}_k} L$  such that  $\bar{L} \in \Gamma_{\mathcal{T}_k}$  for  $J \subseteq \Gamma_{\mathcal{T}_k}$ ; and (3)  $u$  is *relevant* to  $\mathcal{T}_k$  in  $\Gamma_{\mathcal{T}_k}$ . The latter means that either (i)  $u \in G(\Gamma_{\mathcal{T}_k})$ ,  $\mathcal{T}_k$  has its sort and values for it, so that  $\mathcal{I}_k$  can decide an assignment to  $u$ ; or (ii)  $u$  is an equality  $u_1 \simeq u_2$  such that  $u_1, u_2 \in G(\Gamma_{\mathcal{T}_k})$ ,  $\mathcal{T}_k$  has their sort, but does not have values for their sort, so that  $\mathcal{I}_k$  can decide the truth value of  $u_1 \simeq u_2$ .

By Condition (1), if  $L \in \Gamma$ , both  $L$  and  $\bar{L}$  are unacceptable for all theories. By Condition (2), if  $\{x \leftarrow 1, \bar{x} < \bar{y}\} \subseteq \Gamma$ , then  $y \leftarrow 2$  is unacceptable for LRA, as  $\{x \leftarrow 1, y \leftarrow 2\} \vdash_{\mathcal{I}_{\text{LRA}}} x < y$  by an LRA-evaluation inference (cf. Sect. 4.4). By Condition (3), if  $\{f(u_1) \leftarrow \text{red}, u_2 \leftarrow \text{yellow}\} \subseteq \Gamma$ , where  $f$  is a function from colors to colors,  $u_1 \leftarrow \text{yellow}$  is relevant to a theory of colors by (i), while  $u_1 \simeq u_2$  is relevant to EUF by (ii), provided EUF has the sort of colors. A decision  $u \leftarrow c$  is *forced* when  $c$  is the only acceptable value for  $u$ , such as if  $\{u \simeq t, t \leftarrow c\} \subseteq \Gamma$  for EUF, or  $\{u \leq t, t \leq u, t \leftarrow c\} \subseteq \Gamma$  for LRA.

Rule **Deduce** expands  $\Gamma$  with a Boolean singleton assignment justified by a theory inference  $J \vdash_k L$  from assignments  $J$  already in  $\Gamma$ . Sound theory inferences yield sound justified assignments. The system proceeds with decisions and deductions until a conflict arises: if  $J \vdash_k L$  and  $\bar{L} \in \Gamma$ , the assignment  $J \cup \{\bar{L}\}$  is a conflict. **Deduce** encompasses *propagation*, by deducing an assignment entailed in the theory by assignments in  $\Gamma$ , and *conflict explanation*, by performing theory inferences that allow a theory conflict to surface in  $\Gamma$  as a Boolean conflict. For example, given a series of decisions  $u_2 \leftarrow \text{yellow}$ ,  $f(u_1) \leftarrow \text{red}$ ,  $u_1 \leftarrow \text{yellow}$ ,  $f(u_2) \leftarrow \text{blue}$ , a module for the theory of colors can

deduce (propagate)  $f(u_1) \not\approx f(u_2)$  and  $u_1 \simeq u_2$  by equality inferences (cf. Fig. 3 in Sect. 3.3), and a module for EUF can detect the conflict by deducing  $\{u_1 \simeq u_2, f(u_1) \not\approx f(u_2)\} \vdash_{\text{EUF}} \perp$  (cf. Sect. 4.2). If the conflict is at level 0, rule Fail returns unsat. Otherwise, rule ConflictSolve returns the trail  $\Gamma'$  produced by the conflict-state rules, so that the search can resume.

The conflict-state rules handle both first-order and Boolean assignments and their interplay. Conflict-solving as in CDCL involves flipping a Boolean assignment  $L$  into  $\bar{L}$ , recording that  $L$  was tried and failed. A first-order assignment  $u \leftarrow c$  cannot be flipped: its complement would be the set of all other values for  $u$ , which is not a singleton and not even a finite set in general. Thus,  $u \leftarrow c$  is *undone*, not flipped. Since  $u \leftarrow c$  may appear in justifications, undoing it requires the removal of all its consequences (i.e., justified assignments with  $u \leftarrow c$  in the justification). A Boolean decision  $\bar{L}$  is then forced (for a consequence  $L$  of  $u \leftarrow c$  in the conflict) to prevent repeating the same first-order decision  $u \leftarrow c$  that caused a conflict. Another remark is that since each assignment has a level, conflict solving can proceed by considering an assignment that stands out, because its level is the greatest level in the conflict. Clearly, such an assignment is not unique in general.

The main workhorse of conflict solving is the Resolve rule. Resolve *explains* a conflict  $E \uplus \{A\}$  by replacing a *justified assignment*  $A$  with its justification  $H$  (see Fig. 1). Since  $H_0 \cup E \uplus \{A\} \models \perp$ , and  $H \vdash A$  is sound,  $H_0 \cup E \cup H \models \perp$  follows, and  $E \cup H$  is still a conflict. If  $A$  is first-order, it is an input assignment ( $H = \emptyset$ ), and Resolve removes  $A$  from the conflict, not from the trail. For example, Resolve turns a conflict  $\{u_1 \simeq u_2, f(u_1) \not\approx f(u_2)\}$  into  $\{u_1 \simeq u_2, f(u_1) \leftarrow \text{red}, f(u_2) \leftarrow \text{blue}\}$ , if  $\{f(u_1) \leftarrow \text{red}, f(u_2) \leftarrow \text{blue}\} \vdash f(u_1) \not\approx f(u_2)$  is on the trail. Intuitively, one can think of Resolve continuing to unfold the conflict, until it contains an assignment  $A$  that stands out in the above sense. This outstanding assignment  $A$  is either a first-order or a Boolean assignment.

If  $A$  is a first-order assignment with  $\text{level}_\Gamma(A) = m$ , rule UndoClear applies by going back to  $\Gamma^{\leq m-1}$  (see Fig. 1), which means that it undoes  $A$  and clears the trail of all its consequences. Note that  $m-1 \geq 0$  implies  $m > 0$ , that is,  $A$  is a decision. UndoClear solves the above conflict produced by Resolve by removing  $f(u_2) \leftarrow \text{blue}$ , whose level is  $m = 4$ , and  $\{f(u_1) \leftarrow \text{red}, f(u_2) \leftarrow \text{blue}\} \vdash f(u_1) \not\approx f(u_2)$ . Going back to  $\Gamma^{\leq m-1}$  does not represent a loop, because  $\Gamma^{\leq m-1}$  is new, as it must contain some late propagation. Indeed,  $A$  was acceptable when it was decided, which means it did not cause a conflict. If  $A$  became later part of a conflict, it must be that some late propagation  $L$  with  $\text{level}_\Gamma(L) < m$  was added to the trail *after*  $A$ , so that  $L$  is in  $\Gamma^{\leq m-1}$ . In the example, the late propagation is  $u_1 \simeq u_2$  on level 3. A Deduce step based on  $u_1 \simeq u_2 \vdash_{\text{EUF}} f(u_1) \simeq f(u_2)$  adds  $f(u_1) \simeq f(u_2)$  to the trail, making  $f(u_2) \leftarrow \text{red}$  a forced decision.

If  $A$  is a Boolean assignment  $L$  in a conflict  $E \uplus \{L\}$  such that  $\text{level}_\Gamma(E) = m$  and  $\text{level}_\Gamma(L) > m$ , CDSAT [11] applies the Backjump rule, which solves the conflict by producing the trail  $\Gamma^{\leq m}, E \vdash \bar{L}$ . In other words, it jumps back to level  $m$  and adds to the trail the justified assignment  $E \vdash \bar{L}$ , which is sound because  $H_0 \cup (E \uplus \{L\}) \models \perp$  yields  $H_0 \cup E \models \bar{L}$ . Here the Backjump rule is replaced with the more general LearnBackjump, which behaves like Backjump when  $H$  and  $L$  in its definition in Fig. 1 are  $\{L\}$  and  $\bar{L}$ , respectively. LearnBackjump and the notion of *clausal form* mentioned in Fig. 1 will be illustrated in Sect. 3.2.

Last, UndoDecide covers a situation where Resolve cannot apply. Reconsider Resolve explaining a conflict  $E \uplus \{A\}$  by replacing a Boolean justified assignment  $A$  with its justification  $H$ . The condition of Resolve in Fig. 1 requires that  $H$  does *not* contain a first-order decision  $A'$  such that  $\text{level}_\Gamma(A') = m = \text{level}_\Gamma(E \uplus \{A\})$ . Indeed, suppose that  $A$  is  $\{A'\} \vdash L$  and  $\text{level}_\Gamma(E) < m$ : if Resolve unfolds  $E \uplus \{A\}$  into  $E \uplus \{A'\}$ , UndoClear becomes applica-

ble. If UndoClear undoes  $A'$ , and then Decide retries  $A'$ , and Deduce reiterates  $\langle A' \rangle \vdash L$ , the system loops. Thus, Resolve is forbidden, and either UndoDecide or LearnBackjump applies. If an assignment other than  $L$  in the conflict has level  $m$  (see the condition of UndoDecide in Fig. 1), UndoDecide undoes  $A'$  and its consequences by going back to  $\Gamma^{\leq m-1}$  and decides  $\bar{L}$ . If  $L$  is the *only* assignment of level  $m$  in the conflict, LearnBackjump applies like Backjump.

The CDSAT transition system is non-deterministic, as it leaves room for heuristic choices. Thus, multiple CDSAT-derivations from a given input exist. The addition of a *search plan* that controls the application of the transition rules yields a *CDSAT procedure*, whose derivation from a given input is unique.

### 3.2 Lemma Learning in CDSAT

The CDCL procedure can *learn* a propositional resolvent that was generated to explain a conflict. For example, consider a CDSAT trail containing

$$\emptyset \vdash a \vee b, \emptyset \vdash \neg a \vee d, \dots, \gamma \bar{d}, \gamma \bar{b}, \langle \bar{d}, \neg a \vee d \rangle \vdash \bar{a},$$

where  $\emptyset \vdash a \vee b$  and  $\emptyset \vdash \neg a \vee d$  belong to the input assignment  $H_0$  and have level 0;  $\gamma \bar{d}$  has level 1,  $\gamma \bar{b}$  has level 2, and  $\langle \bar{d}, \neg a \vee d \rangle \vdash \bar{a}$  is a late propagation and has level 1. Clause  $a \vee b$  is a conflict clause for CDCL, and  $\{a \vee b, \bar{a}, \bar{b}\}$  is a conflict for CDSAT. The CDCL procedure can learn  $b \vee d$ , resolvent of  $a \vee b$  and the CDCL justification  $\neg a \vee d$  of  $\bar{a}$ .

CDSAT [11] can apply the Backjump rule, which fires when CDSAT reaches a conflict state  $\langle \Gamma; E \uplus \{L\} \rangle$ , where  $\text{level}_\Gamma(E) = m$  and  $\text{level}_\Gamma(L) > m$ , producing the trail  $\Gamma^{\leq m}, \langle E \rangle \vdash \bar{L}$ . In the example,  $E = \{a \vee b, \bar{a}\}$ ,  $L = \bar{b}$ ,  $\text{level}_\Gamma(E) = 1$ ,  $\text{level}_\Gamma(L) = 2 > 1$ , and Backjump produces the trail

$$\emptyset \vdash a \vee b, \emptyset \vdash \neg a \vee d, \dots, \gamma \bar{d}, \langle \bar{d}, \neg a \vee d \rangle \vdash \bar{a}, \langle a \vee b, \bar{a} \rangle \vdash b.$$

Alternatively, CDSAT [11] can Resolve the conflict into  $\{a \vee b, \bar{d}, \neg a \vee d, \bar{b}\}$  and then apply Backjump to yield the trail

$$\emptyset \vdash a \vee b, \emptyset \vdash \neg a \vee d, \dots, \gamma \bar{d}, \langle \bar{d}, \neg a \vee d \rangle \vdash \bar{a}, \langle a \vee b, \bar{d}, \neg a \vee d \rangle \vdash b.$$

Either way, CDSAT [11] learns  $b$ , not  $b \vee d$ . The new rule LearnBackjump of Fig. 1 enables CDSAT to learn the justified assignment  $\langle a \vee b, \neg a \vee d \rangle \vdash b \vee d$  from the conflict  $\{a \vee b, \bar{d}, \neg a \vee d, \bar{b}\}$ , forming clause  $b \vee d$  from the subset  $\{\bar{d}, \bar{b}\}$  of the conflict.

In general, LearnBackjump empowers CDSAT to turn *any Boolean subset* of a conflict into a *disjunction* of Boolean terms, that the system can learn, and that we call *clause*, slightly abusing the terminology because Boolean terms are formulæ. This requires  $\vee \in F_\infty$ , which is the case whenever  $\mathcal{T}_\infty$  includes propositional logic. If  $\vee \notin F_\infty$ , only unit clauses will be learned. Suppose that  $E \uplus H$  is a conflict, where  $H$  contains only Boolean assignments. This means that  $H_0 \cup (E \uplus H) \models \perp$ , where  $H_0$  is the input assignment. If  $H$  is a singleton  $L$ , we have  $H_0 \cup (E \uplus \{L\}) \models \perp$ , hence  $H_0 \cup E \models \bar{L}$ , and  $\langle E \rangle \vdash \bar{L}$  can be learned. If  $H$  is not a singleton, it can be rewritten as the singleton

$$\left( \left( \bigwedge_{(l \leftarrow \text{true}) \in H} l \right) \wedge \left( \bigwedge_{(l \leftarrow \text{false}) \in H} \neg l \right) \right) \leftarrow \text{true}$$

**Fig. 2** Propositional extract from a CDSAT derivation

Input problem  $H_0$  including:  $(\neg l_4 \vee l_5), (\neg l_2 \vee \neg l_4 \vee \neg l_5)$   
 Initial trail  $\Gamma_0$  including:  $\emptyset \vdash (\neg l_4 \vee l_5), \emptyset \vdash (\neg l_2 \vee \neg l_4 \vee \neg l_5)$   
 Extending  $\Gamma_0$  into  $\Gamma = \Gamma_0, ?A_1, ?l_2, ?A_3, ?l_4, (\neg l_4 \vee l_5), l_4 \vdash l_5$   
 (involving unrelated decisions  $A_1$  and  $A_3$ )  
 First conflict:  $\langle \Gamma; (\neg l_2 \vee \neg l_4 \vee \neg l_5), l_2, l_4, l_5 \rangle$   
 Applying Resolve to  $l_5$ :  $\langle \Gamma; (\neg l_2 \vee \neg l_4 \vee \neg l_5), l_2, l_4, (\neg l_4 \vee l_5) \rangle$

whose flip is  $((\bigwedge_{(l \leftarrow \text{true}) \in H} \neg l) \wedge (\bigwedge_{(l \leftarrow \text{false}) \in H} l)) \leftarrow \text{false}$ . In order to get a clause, the latter assignment can be rewritten in the equivalent form

$$\left( \left( \bigvee_{(l \leftarrow \text{true}) \in H} \neg l \right) \vee \left( \bigvee_{(l \leftarrow \text{false}) \in H} l \right) \right) \leftarrow \text{true}$$

leading to the next definition.

**Definition 1** (*Clausal form of an assignment in a conflict*) Given a conflict  $E \uplus H$ , where  $H$  is a Boolean assignment, the *clausal form* of  $H$  is the singleton Boolean assignment  $((\bigvee_{(l \leftarrow \text{true}) \in H} \neg l) \vee (\bigvee_{(l \leftarrow \text{false}) \in H} l)) \leftarrow \text{true}$ , or, equivalently,  $((\bigwedge_{(l \leftarrow \text{true}) \in H} l) \wedge (\bigwedge_{(l \leftarrow \text{false}) \in H} \neg l)) \leftarrow \text{false}$ .

The new rule LearnBackjump allows CDSAT to perform *learning and backjumping*, or *learning and restart*, and it subsumes the Backjump rule [11], adding the capability of *learning clauses*. We examine these features in this order. *Learning and backjumping* is the generic behavior of LearnBackjump. This rule singles out a Boolean subset  $H$  of the conflict  $E \uplus H$ , such that  $\text{level}_\Gamma(H) > \text{level}_\Gamma(E)$ . Then, it solves the conflict by jumping back to a level  $m$ , such that  $\text{level}_\Gamma(E) \leq m < \text{level}_\Gamma(H)$ , and learning a clausal form  $L$  of  $H$ . The system learns  $L$  by adding to the trail the justified assignment  $E \vdash L$ , which is *sound*, because  $H_0 \cup (E \uplus H) \vdash \perp$  implies  $H_0 \cup E \vdash L$ , as  $L$  is a clausal form of  $H$ . As  $L$  may be a new Boolean term, it must belong to  $\mathcal{B}$ . Note that  $H$  does not necessarily contain all Boolean assignments in the conflict: it is the search plan that chooses a Boolean subset  $H$  and a destination level  $m$ .

**Example 1** Consider the conflict on the last line of Fig. 2. If LearnBackjump is applied with  $H = \{l_2, l_4\}$ , and  $E = \{(\neg l_2 \vee \neg l_4 \vee \neg l_5), (\neg l_4 \vee l_5)\}$ ,  $((\neg l_2 \vee \neg l_4) \leftarrow \text{true})$  is a clausal form of  $H$ ,  $\text{level}_\Gamma(H) = 4$ , and  $\text{level}_\Gamma(E) = 0$ , so that any destination level  $m$  such that  $0 \leq m < 4$  can be picked. A standard choice for  $m$  would be the second highest level in the conflict, namely  $m = 2$ , in which case the LearnBackjump step jumps over decision  $A_3$  and yields

$$\Gamma_0, ?A_1, ?l_2, (\neg l_2 \vee \neg l_4 \vee \neg l_5), (\neg l_4 \vee l_5) \vdash (\neg l_2 \vee \neg l_4).$$

The derivation continues from level 2 with  $\neg l_2 \vee \neg l_4$  added to level 0.

We consider next *learning and restart*. It is common to restart after learning a clause, and search plans with aggressive restart proved successful in SAT solving. LearnBackjump makes this kind of search plan possible in CDSAT. Assume that the destination level  $m$  is chosen to be the *smallest*, that is,  $m = \text{level}_\Gamma(E)$ . If  $\text{level}_\Gamma(E)$  is 0, LearnBackjump produces a trail of the form  $\Gamma \stackrel{\leq 0}{\vdash} E \vdash L$ , performing a *restart* and adding  $E \vdash L$  to level 0.

**Example 2** The LearnBackjump step of Example 1 with destination level  $m = 0$  generates

$$\Gamma_0, (\neg l_2 \vee \neg l_4 \vee \neg l_5), (\neg l_4 \vee l_5) \vdash (\neg l_2 \vee \neg l_4).$$

We analyze next how LearnBackjump subsumes the Backjump rule [11] recalled in Sect. 3.1 and at the beginning of this section. LearnBackjump behaves in the same way as Backjump, if it picks as  $H$  a singleton  $L$ , as  $\bar{L}$  is a clausal form of a singleton Boolean assignment  $L$  in a conflict. However, while Backjump goes back to level  $m = \text{level}_\Gamma(E)$ , LearnBackjump allows to choose any destination level  $m$  such that  $\text{level}_\Gamma(E) \leq m < \text{level}_\Gamma(L)$ .

**Example 3** In the conflict on the last line of Fig. 2, the level of  $l_4$  is greater than that of the rest of the conflict  $E = \{(\neg l_2 \vee \neg l_4 \vee \neg l_5), l_2, (\neg l_4 \vee l_5)\}$ , as  $\text{level}_\Gamma(l_4) = 4 > \text{level}_\Gamma(E) = 2$ . Thus, Backjump could apply; LearnBackjump mimics it with  $H = \{l_4\}$  and  $m = 2$  to yield

$$\Gamma_0, ?A_1, ?l_2, (\neg l_2 \vee \neg l_4 \vee \neg l_5), l_2, (\neg l_4 \vee l_5) \vdash \bar{l}_4.$$

Alternatively, if  $m = 3$ , LearnBackjump produces

$$\Gamma_0, ?A_1, ?l_2, ?A_3, (\neg l_2 \vee \neg l_4 \vee \neg l_5), l_2, (\neg l_4 \vee l_5) \vdash \bar{l}_4.$$

The side-conditions  $\bar{L} \notin \Gamma$  and  $L \notin \Gamma$  prevent LearnBackjump from breaking plausibility or adding to the trail a clause that is already there.

**Example 4** Consider the first conflict in Fig. 2:  $\langle \Gamma; (\neg l_2 \vee \neg l_4 \vee \neg l_5), l_2, l_4, l_5 \rangle$ . For a LearnBackjump step with  $E = \{(\neg l_2 \vee \neg l_4 \vee \neg l_5)\}$  and  $H = \{l_2, l_4, l_5\}$ , we have  $\text{level}_\Gamma(H) = 4$  and  $\text{level}_\Gamma(E) = 0$ . Regardless of the choice of destination level  $m$ ,  $0 \leq m < 4$ , a clausal form of  $H$  is redundant since clause  $\neg l_2 \vee \neg l_4 \vee \neg l_5$  is already on the trail and LearnBackjump does not add it.

Unlike Backjump, LearnBackjump does *not* require that the conflict contains a singleton assignment  $L$  of level greater than the rest of the conflict.

**Example 5** In the first conflict in Fig. 2 both  $l_4$  and  $l_5$  have level 4. If we apply LearnBackjump with  $H = \{l_4, l_5\}$ ,  $E = \{(\neg l_2 \vee \neg l_4 \vee \neg l_5), l_2\}$ ,  $\text{level}_\Gamma(H) = 4$ ,  $\text{level}_\Gamma(E) = 2$ , and destination level  $m = 2$ , the resulting trail is

$$\Gamma_0, ?A_1, ?l_2, E \vdash \neg l_4 \vee \neg l_5$$

where  $(\neg l_4 \vee \neg l_5) \leftarrow \text{true}$  on level 2 is a clausal form of  $H$ .

We inspect last the learning of clauses. In CDCL, the last conflict clause generated prior to backjumping is called *backjump clause*: the procedure learns this clause and jumps back to a prior level, undoing at least one decision and satisfying the learned clause by placing one of its literals on the trail. An *assertion clause* is a conflict clause such that only one of its literals, termed *assertion literal*, is falsified on the current, or greatest, level of the trail. The *First Unique Implication Point* (UIP) heuristic [45] picks as backjump clause the *first* generated assertion clause, as destination level the smallest where the assertion literal is undefined and all other literals of the assertion clause are false, and places the assertion literal on the trail.

The Backjump rule of CDSAT [11] generalizes this behavior, taking into account that, unlike in CDCL, a CDSAT trail is *not* a stack. Backjump applies to a conflict  $E \uplus \{L\}$  such that  $\text{level}_\Gamma(L) > \text{level}_\Gamma(E)$ , but  $\text{level}_\Gamma(L)$  is *not* necessarily the current one, and Backjump puts  $E \vdash \bar{L}$  on the trail *without learning an assertion clause*. However, if a CDSAT conflict has the form  $E \uplus \{L\}$  with  $\text{level}_\Gamma(L) > \text{level}_\Gamma(E)$ , it is possible to extract from the conflict an assertion clause, and LearnBackjump does it.

Let  $\kappa = l_1 \vee \dots \vee l_q$  be an assertion clause and  $l_q$  its literal such that  $L = (l_q \leftarrow \text{false})$  is on the current level. Assume that  $\kappa \in \mathcal{B}$ . In order to learn  $\kappa$ , it suffices to take the Boolean

subset  $H = H' \uplus \{L\}$  of the conflict that makes  $\kappa$  false: for all  $i, 1 \leq i \leq q, (l_i \leftarrow \text{false}) \in H$  if and only if  $l_i \in \kappa$  and  $(l_i \leftarrow \text{true}) \in H$  if and only if  $\neg l_i \in \kappa$ . By Definition 1, the assignment  $K = (\kappa \leftarrow \text{true})$  is a clausal form of  $H$ . Let  $E$  be the rest of the conflict. Then the system applies LearnBackjump with destination level  $m = \text{level}_\Gamma(E \uplus H')$ , which means that  $m \geq \text{level}_\Gamma(H')$ . This choice satisfies the condition  $\text{level}_\Gamma(E) \leq m < \text{level}_\Gamma(H)$ , because  $\text{level}_\Gamma(E) \leq \text{level}_\Gamma(E \uplus H') < \text{level}_\Gamma(H) = \text{level}_\Gamma(L)$ . This LearnBackjump step yields the trail

$$\Gamma^{\leq m}, \underline{E} \vdash K,$$

and  $\kappa$  is learned. The theory module for Bool features inference rules for *unit propagation* (see Sect. 4.1) that allow the inference:

$$\{K\} \uplus H' \vdash_{\text{Bool}} \bar{L}. \tag{1}$$

Indeed,  $K$  is  $(l_1 \vee \dots \vee l_{q-1} \vee l_q) \leftarrow \text{true}$ , and  $H'$  makes  $l_1, \dots, l_{q-1}$  false, so that unit propagation infers  $l_q$ . Since  $L$  makes  $l_q$  false,  $\bar{L}$  makes  $l_q$  true. Because the destination level  $m$  of the LearnBackjump step was chosen in such a way that  $m \geq \text{level}_\Gamma(H')$ , the premises  $K, H'$  of inference (1) are all on the trail  $\Gamma^{\leq m}, \underline{E} \vdash K$ . Furthermore, literal  $l_q$  is in  $\mathcal{B}$ , since  $L$  was on the trail. Thus, all conditions for a Deduce step with inference (1) are met. The resulting trail is

$$\Gamma^{\leq m}, \underline{E} \vdash K, \{K\} \uplus H' \vdash \bar{L}$$

which is similar to the  $\Gamma^{\leq m}, \underline{E} \uplus H' \vdash \bar{L}$  produced by Backjump, except for the learned clause  $K$ . The advantage is that  $K$  can be reused in future branches of the search. The smaller the level of  $\underline{E} \vdash K$ , which is  $\text{level}_\Gamma(E)$ , the longer  $K$  may remain on the trail and be used for inferences.

**Example 6** Continuing Example 1 from

$$\Gamma_0, ?A_1, ?l_2, (\neg l_2 \vee \neg l_4 \vee \neg l_5), (\neg l_4 \vee l_5) \vdash (\neg l_2 \vee \neg l_4),$$

rule Deduce with inference (1) generates

$$\Gamma_0, ?A_1, ?l_2, (\neg l_2 \vee \neg l_4 \vee \neg l_5), (\neg l_4 \vee l_5) \vdash (\neg l_2 \vee \neg l_4), (\neg l_2 \vee \neg l_4), l_2 \vdash \bar{l}_4.$$

A comparison between Examples 3 and 6 shows the difference between LearnBackjump imitating Backjump, and a LearnBackjump Deduce sequence that backjumps, learns the assertion clause, and asserts the assertion literal by Deduce. A CDSAT search plan may restrict the application of LearnBackjump to IUIP assertion clauses and couple it with Deduce systematically.

### 3.3 Soundness, Completeness, and Termination with Lemma Learning

In this section we present definitions about theory modules that appeared in [11], but must be reproduced because they are indispensable for what follows in Sects. 4 and 5, and we discuss how replacing Backjump with LearnBackjump is harmless for the *soundness, termination, and completeness* of CDSAT.

Every theory module includes the *equality inference rules* of Fig. 3. The first two rules allow any module to infer an assignment to an equality from assignments to its sides. Indeed, in the presence of first-order assignments, there are two ways to make an equality  $t_1 \simeq_s t_2$

$$\begin{aligned}
& t_1 \leftarrow c, t_2 \leftarrow c \vdash t_1 \simeq_s t_2 \text{ if } c \text{ is a } \mathcal{T}\text{-value of sort } s \\
& t_1 \leftarrow c_1, t_2 \leftarrow c_2 \vdash t_1 \not\simeq_s t_2 \text{ if } c_1 \text{ and } c_2 \text{ are distinct } \mathcal{T}\text{-values of sort } s \\
& \quad \vdash t_1 \simeq_s t_1 \text{ (reflexivity)} \\
& \quad t_1 \simeq_s t_2 \vdash t_2 \simeq_s t_1 \text{ (symmetry)} \\
& t_1 \simeq_s t_2, t_2 \simeq_s t_3 \vdash t_1 \simeq_s t_3 \text{ (transitivity)}
\end{aligned}$$

**Fig. 3** Equality inference rules, where  $t_1$ ,  $t_2$ , and  $t_3$  are terms of sort  $s$

true: either assign the same value to  $t_1$  and  $t_2$  or assign true to  $t_1 \simeq_s t_2$ . Dually, there are two ways to make  $t_1 \simeq_s t_2$  false: either assign distinct values to  $t_1$  and  $t_2$  or assign false to  $t_1 \simeq_s t_2$ . The first two equality rules provide a bridge between these two ways (cf. the two ways for a term to be *relevant* to a theory in Sect. 3.1). The remaining equality rules are standard rules for *reflexivity*, *symmetry*, and *transitivity*.

In order to explain theory conflicts, theory inferences may introduce new (i.e., non-input) terms. For termination, all new terms must come from a finite *local basis* associated with the module and dependent on the input problem. We say that a set  $X$  of terms is *closed* if (i) it is closed with respect to the subterm ordering, or  $\leq$ -closed for short: for all  $u \in X$ ,  $t \leq u$  implies  $t \in X$ , and (ii) it is closed with respect to equality: for all  $t, u \in X$  of sort  $s$ ,  $s \neq \text{prop}$ ,  $(t \simeq_s u) \in X$ . The second condition excludes *prop*, because otherwise a non-empty closed set is necessarily infinite, as it would contain, for all terms  $t$  of sort  $s$ , the infinite series  $l_1 = (t \simeq_s t)$ ,  $l_2 = (l_1 \simeq_{\text{prop}} l_1)$ ,  $l_3 = (l_2 \simeq_{\text{prop}} l_2)$ , etc. The *closure*  $\Downarrow X$  of a set  $X$  of terms is the smallest closed set containing  $X$ . The closure operation is *idempotent*, as  $\Downarrow(\Downarrow X) = \Downarrow X$ , and *monotone*: if  $X \subseteq Y$  then  $\Downarrow X \subseteq \Downarrow Y$ .

**Definition 2** (*Basis*) A *basis* for theory  $\mathcal{T}$  with signature  $\Sigma$  is a function *basis* from sets of terms to sets of terms, such that for all sets  $X$  of terms:

- $X \subseteq \text{basis}(X)$  (*extensiveness*),
- If  $X$  is finite then  $\text{basis}(X)$  is finite (*finiteness*),
- $\text{basis}(X) = \text{basis}(\Downarrow X) = \Downarrow \text{basis}(X)$  (*closedness*),
- For all sets  $Y$  of terms, if  $X \subseteq Y$  then  $\text{basis}(X) \subseteq \text{basis}(Y)$  (*monotonicity*),
- $\text{basis}(\text{basis}(X)) = \text{basis}(X)$  (*idempotence*), and
- $\text{fv}_\Sigma(\text{basis}(X)) \subseteq \text{fv}_\Sigma(X) \cup \mathcal{V}_\infty$  (*no introduction of foreign terms*).

For each theory  $\mathcal{T}_k$  in the union,  $1 \leq k \leq n$ , the theory module  $\mathcal{I}_k$  has a basis, called *local basis* and denoted  $\text{basis}_{\mathcal{I}_k}$  or  $\text{basis}_k$ , such that for all sets  $X$  of terms (e.g.,  $X = G(H_0)$  for input assignment  $H_0$  in a CDSAT-derivation),  $\text{basis}_k(X)$  contains all terms that  $\mathcal{I}_k$  can generate starting from those in  $X$ . Given a  $\mathcal{T}$ -assignment  $J$ , we abbreviate  $\text{basis}(G(J))$  as  $\text{basis}(J)$ . The global basis  $\mathcal{B}$  is *stable* if for all  $k$ ,  $1 \leq k \leq n$ ,  $\text{basis}_k(\mathcal{B}) \subseteq \mathcal{B}$ .

**Definition 3** (*Assignment expansion*) A  $\mathcal{T}$ -module  $\mathcal{I}$  with local basis *basis* *expands* a  $\mathcal{T}$ -assignment  $J$  by adding either (1) a  $\mathcal{T}$ -assignment  $A$  that is acceptable for  $\mathcal{I}$  in  $J$ , or (2) a Boolean assignment  $l \leftarrow b$  derived by an  $\mathcal{I}$ -inference  $J' \vdash_{\mathcal{I}} (l \leftarrow b)$  such that  $J' \subseteq J$ ,  $(l \leftarrow b) \notin J$ , and  $l \in \text{basis}(J)$ .

Case (1) covers *Decide* and Case (2) covers *Deduce*, *Fail*, and *ConflictSolve*.

**Definition 4** (*One-theory-completeness*) Given theory  $\mathcal{T}$ , a  $\mathcal{T}$ -module  $\mathcal{I}$  is *complete* for  $\mathcal{T}$ , if, for all plausible  $\mathcal{T}$ -assignments  $J$ , either  $\mathcal{I}$  can expand  $J$  or there exists a  $\mathcal{T}^+[\text{fv}_\Sigma(J)]$ -model  $\mathcal{M}$  such that  $\mathcal{M} \models J$ .

For completeness in a union  $\mathcal{T}_\infty$  of theories, the theories need to agree on cardinalities of shared sorts and equalities between shared terms. CDSAT achieves this by requiring that

every theory agrees on both counts with a *leading theory*, say  $\mathcal{T}_1$ , which has all the sorts, that is, such that  $S_1 = S_\infty$ .

**Definition 5** (*Leading-theory-compatibility*) Let  $\mathcal{T}_1$  be the leading theory,  $\mathcal{T}$ ,  $\Sigma$ , and  $S$  stand for  $\mathcal{T}_k$ ,  $\Sigma_k$ , and  $S_k$ ,  $2 \leq k \leq n$ , and  $N$  be a set of terms. A  $\mathcal{T}$ -assignment  $J$  is *leading-theory-compatible with  $\mathcal{T}$  sharing  $N$* , if for all  $\mathcal{T}_1^+[\mathcal{V}_1]$ -model  $\mathcal{M}_1$  such that  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$  with  $\text{fv}_{\Sigma_1}(J \cup N) \subseteq \mathcal{V}_1$ , there exists a  $\mathcal{T}^+[\mathcal{V}]$ -model  $\mathcal{M}$  with  $\text{fv}_{\Sigma}(J \cup N) \subseteq \mathcal{V}$ , such that (i)  $\mathcal{M} \models J$ , (ii) for all sorts  $s \in S$ ,  $|s^{\mathcal{M}}| = |s^{\mathcal{M}_1}|$ , and (iii) for all  $s \in S$  and terms  $u, u' \in N$  of sort  $s$ ,  $\mathcal{M}(u) = \mathcal{M}(u')$  if and only if  $\mathcal{M}_1(u) = \mathcal{M}_1(u')$ .

Since in a worst-case scenario all terms are shared, the next definition picks as set of shared terms the set of all terms occurring in the assignment.

**Definition 6** (*Leading-theory-completeness*) For a non-leading theory  $\mathcal{T}$ , a  $\mathcal{T}$ -module  $\mathcal{I}$  is *leading-theory-complete*, if for all plausible  $\mathcal{T}$ -assignments  $J$ , either  $\mathcal{I}$  can expand  $J$  or  $J$  is leading-theory-compatible with  $\mathcal{T}$  sharing  $G(J)$ .

Note that if  $\mathcal{I}$  cannot expand  $J$ , all applicable equality inference steps (see Fig. 3) have been applied, and therefore  $J = J_{\mathcal{T}}$ .

The next theorem summarizes the requirements for soundness, termination, and completeness of CDSAT: Sects. 4 and 5 will show how to fulfill those for completeness and termination, respectively.

**Theorem 1** *CDSAT with lemma learning and global basis  $\mathcal{B}$  is*

- **Sound:** *if the theory modules are sound, whenever a CDSAT-derivation reaches state unsat, the input problem is unsatisfiable;*
- **Terminating:** *if  $\mathcal{B}$  is finite and closed, every CDSAT-derivation from an input problem in  $\mathcal{B}$  is guaranteed to terminate; and*
- **Complete:** *if there is a leading theory  $\mathcal{T}_1$ , module  $\mathcal{I}_1$  is complete for  $\mathcal{T}_1$ , modules  $\mathcal{I}_k$ 's,  $2 \leq k \leq n$ , are leading-theory-complete, and  $\mathcal{B}$  is stable, whenever a CDSAT-derivation from an input problem in  $\mathcal{B}$  reaches a state other than unsat such that no transition rule applies, there exists a  $\mathcal{T}_\infty^+$ -model that globally endorses the assignment on the trail, hence the input problem.*

We conclude this section with a discussion of how the soundness, termination, and completeness arguments for CDSAT [11] carry over to CDSAT with LearnBackjump. The proof of soundness rests on soundness of the theory modules and on showing that CDSAT transitions transform sound states into sound states, meaning that justified assignments are sound and conflicts are indeed conflicts: LearnBackjump does not change this, because it adds sound justified assignments.

The proof of termination begins by using acceptability of decisions to show that a CDSAT trail does not contain distinct assignments to the same term, unless they are input assignments. For Boolean assignments, this means that CDSAT rules preserve plausibility, and so does LearnBackjump, since in essence it flips a Boolean assignment. Next, the closedness of  $\mathcal{B}$  and the relevance of decided terms<sup>1</sup> are employed to show that if the input assignment  $H_0$  is in  $\mathcal{B}$ , so are all derived trails: this holds also with LearnBackjump, because the learned clause is required to be in  $\mathcal{B}$ . Then, one uses the finiteness of  $\mathcal{B}$  to get an upper bound on trail

<sup>1</sup> This proof works also if relevance (see Sect. 3.1) of term  $u$  is weakened to  $u \in \mathcal{B}$ , which allows CDSAT to decide the value of  $u$  even if  $u \notin G(\Gamma)$ .

length, hence a trail measure, and shows that CDSAT transitions reduce the trail measure with respect to a well-founded ordering: LearnBackjump does it like Backjump.

For completeness, one preliminarily observes that if  $\mathcal{B}$  is stable, then it is closed (by extensiveness and closedness of all  $\text{basis}_k$ 's, see Definition 2). Then, one uses the closedness of  $\mathcal{B}$  and the completeness of the theory modules to show that, whenever a CDSAT-derivation reaches a state other than unsat such that no transition rule applies, its trail  $\Gamma$  is model-describing. Replacing Backjump with LearnBackjump preserves this result, because if LearnBackjump does not apply, Backjump does not apply either, as LearnBackjump subsumes Backjump.  $\Gamma$  is *model-describing* if  $\Gamma_{\mathcal{T}_1}$  is endorsed by a  $\mathcal{T}_1^+$ -model, and for all  $k$ ,  $2 \leq k \leq n$ ,  $\Gamma_{\mathcal{T}_k}$  is leading-theory-compatible with  $\mathcal{T}_k$  sharing the set of shared terms of the problem. The generic assignment  $J$  of the definitions of leading-theory-compatibility and leading-theory-completeness (see Definitions 5 and 6) is instantiated to  $\Gamma_{\mathcal{T}_k}$ , and a theory module  $\mathcal{I}_k$ ,  $2 \leq k \leq n$ , is leading-theory-complete sharing  $G(\Gamma_{\mathcal{T}_k})$ , hence sharing the set of shared terms of the problem, since the latter is a subset of  $G(\Gamma_{\mathcal{T}_k})$  for all problems. The proof of completeness is achieved by showing that a model-describing trail is globally endorsed by a  $\mathcal{T}_\infty^+$ -model, which is independent of transition rules.

## 4 Completeness of Theory Modules

In previous work we defined theory modules for Bool, EUF, Arr, LRA, and generic Nelson-Oppen theories [11]. In this section we add a theory module for a generic *non-stably infinite theory*, we specify *local bases* for all these theory modules, and we prove that all these theory modules are *leading-theory-complete for all suitable leading theories*, fulfilling a key requirement for the completeness of CDSAT (see Theorem 1).

A theory module is an *inference system*, that is, a *set of inference rules*, and it represents an *abstraction* with respect to a theory satisfiability procedure. A theory satisfiability procedure implements the inference rules of the module, a *search plan*, and other algorithmic components, such as those of a full-fledged CDCL procedure for Bool, a congruence-closure algorithm for EUF, or an LRA-procedure that keeps polynomials in normal form as sums of monomials and maintains lower and upper bounds for each rational variable.

We begin with a lemma that will be used several times in the sequel. Given a  $\mathcal{T}$ -assignment  $J$ , let  $\simeq_s^J$  be the binary relation over  $G_s(J)$  defined by  $t_1 \simeq_s^J t_2$  if and only if  $(t_1 \simeq_s t_2) \in J$ . The lemma shows that if module  $\mathcal{I}$  for theory  $\mathcal{T}$  *cannot expand*  $J$ , the relation  $\simeq_s^J$  is an equivalence, and  $J$  provides  $\mathcal{T}$ -values for all terms that are relevant to  $\mathcal{T}$ . For terms of sort  $s$  other than `prop`, this result relies on two hypotheses: first,  $J$  does not exhaust the supply of  $s$ -sorted  $\mathcal{T}$ -values, so that a decision is doable; second, the only  $\mathcal{I}$ -rules with first-order assignments as premises are equality inferences (see Fig. 3), so that the analysis of acceptability of decisions is module-independent. If  $\mathcal{T}^+$  offers infinitely many  $s$ -sorted  $\mathcal{T}$ -values, the first hypothesis is satisfied a priori.

**Lemma 1** *If  $\mathcal{T}$ -module  $\mathcal{I}$  cannot expand a plausible  $\mathcal{T}$ -assignment  $J$ , then:*

1. *For all sorts  $s \in S \setminus \{\text{prop}\}$ , the relation  $\simeq_s^J$  is an equivalence, and if  $\{t_1 \leftarrow c_1, t_2 \leftarrow c_2\} \subseteq J$ , then  $c_1$  and  $c_2$  are identical if and only if  $t_1 \simeq_s^J t_2$ ;*
2. *Assignment  $J$  gives a value to every formula that is relevant to  $\mathcal{T}$  in  $J$ ;*
3. *Assignment  $J$  gives a value to every term  $t$  of sort  $s \in S \setminus \{\text{prop}\}$  that is relevant to  $\mathcal{T}$  in  $J$ , provided that (i) there exists a  $\mathcal{T}$ -value of sort  $s$  that  $J$  does not use, and (ii) the only  $\mathcal{I}$ -inferences involving first-order assignments of sort  $s$  are equality inferences.*

**Proof** All claims are proved by way of contradiction.

1. Assume that  $\simeq_s^J$  is not reflexive. This means there exists a term  $t \in G_s(J)$  such that  $(t \simeq_s t) \notin J$ . The Boolean assignment  $t \simeq_s t$  can be derived by reflexivity (see Fig. 3), and  $(t \simeq_s t) \in \text{basis}_{\mathcal{I}}(J)$  since  $\text{basis}_{\mathcal{I}}(J)$  is closed and therefore contains all equalities between terms in  $G_s(J)$  for  $s \neq \text{prop}$ . Thus,  $\mathcal{I}$  can expand  $J$ , which is a contradiction. The cases for symmetry and transitivity are analogous. Similarly, assume that  $\{t_1 \leftarrow c_1, t_2 \leftarrow c_2\} \subseteq J$ ,  $c_1$  and  $c_2$  are identical, but  $(t_1 \simeq_s t_2) \notin J$ : then  $\mathcal{I}$  can expand  $J$  by an equality inference deriving  $t_1 \simeq_s t_2$ . Conversely, assume  $(t_1 \simeq_s t_2) \in J$ , and  $c_1$  and  $c_2$  are distinct: by plausibility  $(t_1 \not\simeq_s t_2) \notin J$ , and  $\mathcal{I}$  can expand  $J$  by an equality inference deriving  $t_1 \not\simeq_s t_2$ .
2. Assume  $l$  is a relevant formula without assigned value. Then  $l \leftarrow b$  (for either truth value) is acceptable for  $\mathcal{I}$  in  $J$ , and therefore  $\mathcal{I}$  can expand  $J$ .
3. Assume that  $J$  does not assign a value to such a relevant term  $t$ . We find an acceptable assignment for  $t$ , so that  $\mathcal{I}$  can expand  $J$ . It suffices to find a value that does not cause a conflict (see Sect. 3.1 for acceptability). Consider the  $\simeq_s^J$ -equivalence class  $e$  of  $t$  (the relation  $\simeq_s^J$  is an equivalence by Part (1)). If none of the terms in  $e$  are assigned a value in  $J$ , then  $t \leftarrow c$ , where  $c$  is the  $\mathcal{T}$ -value of sort  $s$  that  $J$  does not use, is acceptable, because otherwise there would be an assignment  $(t_2 \leftarrow c_2) \in J$  and an equality inference  $t \leftarrow c, t_2 \leftarrow c_2 \vdash t \not\simeq_s t_2$  such that  $(t \simeq_s t_2) \in J$ , meaning  $t_2 \in e$  is assigned a value. If for a term  $t_1 \in e$ ,  $J$  contains  $t_1 \leftarrow c_1$ , then  $t \leftarrow c_1$  is acceptable, because otherwise there would be an assignment  $(t_2 \leftarrow c_2) \in J$  and an equality inference  $t \leftarrow c_1, t_2 \leftarrow c_2 \vdash (t \simeq_s t_2) \leftarrow b$  such that  $(t \simeq_s t_2) \leftarrow b \in J$ : if  $b$  is true, then  $c_1$  is  $c_2$ ,  $t_1 \simeq_s^J t_2$  (by Part (1)), hence  $t \simeq_s^J t_2$  by transitivity (since  $t_1 \in e$ ), so that  $\{t \simeq_s t_2, t \not\simeq_s t_2\} \subseteq J$ , violating plausibility; if  $b$  is false, then  $c_1$  and  $c_2$  are distinct,  $(t \simeq_s t_2) \in J$ , hence  $t \simeq_s^J t_2$ , and, by transitivity (since  $t_1 \in e$ ),  $t_1 \simeq_s^J t_2$ , so that  $c_1$  and  $c_2$  should be identical by Part (1). □

The definition of leading-theory-compatibility with theory  $\mathcal{T}$  (see Definition 5) refers to a generic set  $N$  of shared terms and considers models whose sets of variables include the set of free variables of  $J \cup N$ , for  $J$  a  $\mathcal{T}$ -assignment. The definition of leading-theory-completeness (see Definition 6) instantiates  $N$  to be  $G(J)$  in order to cover all possible sets of shared terms. Thus, when proving leading-theory-completeness we are interested in showing the existence of a  $\mathcal{T}$ -model whose set of variables includes  $\text{fv}_{\Sigma}(J \cup G(J))$ , with  $\Sigma$  the signature of theory  $\mathcal{T}$ . Clearly,  $\text{fv}_{\Sigma}(J \cup G(J)) = \text{fv}_{\Sigma}(G(J))$ . On the other hand, in general,  $\text{fv}_{\Sigma}(G(J)) \neq \text{fv}_{\Sigma}(J)$ , because there can be two  $\Sigma$ -foreign terms  $u, t \in G(J)$  such that  $u \triangleleft t$ , so that  $u \in \text{fv}_{\Sigma}(G(J))$ , but  $u \notin \text{fv}_{\Sigma}(J)$ . The following remark is stated as a corollary of Lemma 1, because Lemma 1 will be applied to show that a  $\mathcal{T}$ -assignment  $J$  assigns values to all terms in  $G(J)$ , or to all equalities between terms in  $G(J)$ , and then the following corollary will be applied to conclude that in such cases  $\text{fv}_{\Sigma}(G(J)) = \text{fv}_{\Sigma}(J)$ , so that it suffices to build a  $\mathcal{T}$ -model whose set of variables includes  $\text{fv}_{\Sigma}(J)$ .

**Corollary 1** *For all signatures  $\Sigma = (S, F)$  and assignments  $J$ , if either (1) for all terms  $t \in G(J)$  there is an assignment  $(t \leftarrow c) \in J$ , or (2) for all distinct terms  $t, u \in G_s(J)$  of sort  $s \in S \setminus \{\text{prop}\}$  there is an assignment  $((t \simeq_s u) \leftarrow b) \in J$ , then  $\text{fv}_{\Sigma}(G(J)) = \text{fv}_{\Sigma}(J)$ .*

**Proof** The direction  $\text{fv}_{\Sigma}(J) \subseteq \text{fv}_{\Sigma}(G(J))$  is trivially true, as  $(t \leftarrow c) \in J$  implies  $t \in G(J)$ . The direction  $\text{fv}_{\Sigma}(G(J)) \subseteq \text{fv}_{\Sigma}(J)$  follows from either hypothesis. □

The corollary is true regardless of signature  $\Sigma$ ; however, it will be applied to a  $\mathcal{T}$ -assignment  $J$  and the signature  $\Sigma$  of theory  $\mathcal{T}$ .

The following lemma is useful to prove leading-theory-completeness for a theory module  $\mathcal{I}_1$  and then extend the result to a module  $\mathcal{I}_2$  with additional inference rules, that is, such that  $\mathcal{I}_1 \subseteq \mathcal{I}_2$  (modules are sets of inference rules).

**Lemma 2** *Let  $\mathcal{I}_1$  and  $\mathcal{I}_2$  such that  $\mathcal{I}_1 \subseteq \mathcal{I}_2$  be modules for a theory  $\mathcal{T}$ . If all inference rules in  $\mathcal{I}_2 \setminus \mathcal{I}_1$  take only Boolean assignments as premises, then if  $\mathcal{I}_1$  is leading-theory-complete,  $\mathcal{I}_2$  also is leading-theory-complete.*

**Proof** We need to show that  $\mathcal{I}_2$  can expand a plausible  $\mathcal{T}$ -assignment  $J$  whenever  $\mathcal{I}_1$  can (see Definition 6). If  $\mathcal{I}_1$  expands  $J$  by an inference (Case (2) of Definition 3), then  $\mathcal{I}_2$  can do it too, since  $\mathcal{I}_1 \subseteq \mathcal{I}_2$ . If  $\mathcal{I}_1$  expands  $J$  by a decision (Case (1) of Definition 3), we need to show that the decision is acceptable also for  $\mathcal{I}_2$ . By way of contradiction, suppose that the decision is acceptable for  $\mathcal{I}_1$  but not for  $\mathcal{I}_2$ . By definition of acceptability (see Sect. 3), this means that the decision is a first-order assignment  $u \leftarrow c$  and there is an inference  $J' \cup \{u \leftarrow c\} \vdash_{\mathcal{I}_2} L$  with  $\bar{L} \in J$  and  $J' \subseteq J$ . Furthermore, this  $\mathcal{I}_2$ -inference applies a rule in  $\mathcal{I}_2 \setminus \mathcal{I}_1$ , because  $u \leftarrow c$  is acceptable for  $\mathcal{I}_1$ . It follows that this rule in  $\mathcal{I}_2 \setminus \mathcal{I}_1$  takes a first-order assignment as premise, contradicting the hypothesis that all rules in  $\mathcal{I}_2 \setminus \mathcal{I}_1$  take only Boolean assignments as premises.  $\square$

Let  $x$  be an arbitrary variable of sort  $\text{prop}$ ,  $\top$  stand for  $(x \simeq_{\text{prop}} x) \leftarrow \text{true}$ , and  $\perp$  for  $x \not\approx_{\text{prop}} x$ : no model endorses  $\perp$  and  $\vdash \top$  is an equality inference.

### 4.1 Propositional Logic

For propositional logic the signature  $\Sigma_{\text{Bool}}$  has only the sort  $\text{prop}$  and symbols  $\simeq_{\text{prop}}$  for equality,  $\neg: \text{prop} \rightarrow \text{prop}$  for negation,  $\vee: (\text{prop} \times \text{prop}) \rightarrow \text{prop}$  for disjunction, and  $\wedge: (\text{prop} \times \text{prop}) \rightarrow \text{prop}$  for conjunction. Let  $\text{Bool}^+$  be the trivial extension, and  $\mathcal{I}_{\text{Bool}}^{\text{eval}}$  the module that only adds to the equality inference rules of Fig. 3 an inference rule for *evaluation* of formulæ:

$$l_1 \leftarrow b_1, \dots, l_m \leftarrow b_m \vdash_{\text{Bool}} l \leftarrow b$$

where  $l$  is in the closure of formulæ  $l_1, \dots, l_m$  under the  $\Sigma_{\text{Bool}}$ -connectives, and  $b$  is its truth value determined by  $b_1, \dots, b_m$  and the truth tables. Given a set  $X$  of terms,  $\text{basis}_{\text{Bool}}(X)$  contains all subformulæ of formulæ in  $X$  by closedness (see Definition 2), and all disjunctions of subformulæ in  $X$  for lemma learning.

**Theorem 2** *Module  $\mathcal{I}_{\text{Bool}}^{\text{eval}}$  is leading-theory-complete for all leading theories.*

**Proof** Let  $J$  be a plausible Boolean assignment that  $\mathcal{I}_{\text{Bool}}^{\text{eval}}$  cannot expand. Since all formulæ in  $G_{\text{prop}}(J)$  are relevant to  $\text{Bool}$ ,  $J$  assigns them values by Part (2) of Lemma 1. This has two consequences: first,  $\text{fv}_{\Sigma_{\text{Bool}}}(G(J)) = \text{fv}_{\Sigma_{\text{Bool}}}(J)$  by Corollary 1; second,  $J$  determines a unique  $\text{Bool}^+[\text{fv}_{\Sigma_{\text{Bool}}}(J)]$ -model  $\mathcal{M}$  such that  $\mathcal{M} \models J$ . We show that  $J$  is leading-theory-compatible with  $\text{Bool}$  sharing  $G(J)$ . Let  $\mathcal{T}_1$  be a leading theory. Since  $J$  is a Boolean assignment,  $J_{\mathcal{T}_1} = J$ . For all  $\mathcal{T}_1^+[\mathcal{V}_1]$ -model  $\mathcal{M}_1$  such that  $\text{fv}_{\Sigma_1}(G(J)) \subseteq \mathcal{V}_1$  and  $\mathcal{M}_1 \models J$ , we have that  $|\text{prop}^{\mathcal{M}}| = |\text{prop}^{\mathcal{M}_1}| = 2$ , and for all terms  $l$  and  $p$  in  $G_{\text{prop}}(J)$ ,  $\mathcal{M}(l) = \mathcal{M}(p)$  if and only if  $\mathcal{M}_1(l) = \mathcal{M}_1(p)$ , since this happens if and only if  $l$  and  $p$  are assigned the same value in  $J$ .  $\square$

Let  $\mathcal{I}_{\text{Bool}}$  be the module that adds to  $\mathcal{I}_{\text{Bool}}^{\text{eval}}$  rules for negation elimination, conjunction elimination, and *unit propagation* as in CDCL:

$$\begin{array}{l} \neg l \vdash_{\text{Bool}} \bar{l} \quad \overline{l_1 \vee \dots \vee l_m} \vdash_{\text{Bool}} \bar{l}_i \quad l_1 \vee \dots \vee l_m, \{\bar{l}_j \mid j \neq i\} \vdash_{\text{Bool}} l_i \\ \overline{\neg l} \vdash_{\text{Bool}} l \quad l_1 \wedge \dots \wedge l_m \vdash_{\text{Bool}} l_i \quad \overline{l_1 \wedge \dots \wedge l_m}, \{l_j \mid j \neq i\} \vdash_{\text{Bool}} \bar{l}_i \end{array}$$

where  $1 \leq j, i \leq m$ . Then by Lemma 2 we have

**Corollary 2** *Module  $\mathcal{I}_{\text{Bool}}$  is leading-theory-complete for all leading theories.*

### 4.2 The Theory of Equality

For the theory of equality EUF, with signature  $\Sigma_{\text{EUF}} = (S, \simeq_S \cup F)$ , the extension  $\text{EUF}^+$  may either be trivial or add a countably infinite set of values for each sort in  $S \setminus \{\text{prop}\}$  and no axioms. A minimal module  $\mathcal{I}_{\text{EUF}}^m$  complements the equality inference rules (see Fig. 3) with an inference rule

$$(t_i \simeq u_i)_{i=1\dots m}, f(t_1, \dots, t_m) \not\simeq f(u_1, \dots, u_m) \vdash_{\text{EUF}} \perp \tag{2}$$

for all  $f \in F$ , that fires when the trail violates a congruence axiom of equality. In case of non-trivial extension, the equality inference rules are the only rules that make use of first-order assignments, and values are employed as labels of congruence classes of terms. For example, the first-order assignment

$$t_1 \leftarrow c, t_2 \leftarrow c, t_3 \leftarrow c_3, t_4 \leftarrow c_4, t_5 \leftarrow c_5$$

and the Boolean assignment

$$t_1 \simeq t_2, t_1 \not\simeq t_3, t_1 \not\simeq t_4, t_1 \not\simeq t_5, t_3 \not\simeq t_4, t_3 \not\simeq t_5, t_4 \not\simeq t_5$$

represent the same four congruence classes. The first-order assignment is an optimization, because it encodes equalities and disequalities between terms without listing them explicitly, whereas a Boolean assignment requires  $\binom{m}{2}$  hence  $O(m^2)$  literals for  $m$  terms in the worst case.

The local basis  $\text{basis}_{\text{EUF}}$  has to ensure that all formulae that may be needed to reason about equality are available. Given a set  $X$  of terms, by closedness  $\text{basis}_{\text{EUF}}(X)$  contains all equalities between subterms of terms in  $X$  of a sort  $s$  other than  $\text{prop}$ . Then  $\text{basis}_{\text{EUF}}$  adds the following equalities between formulae: the formula  $\top$ , and all equalities  $l \simeq_{\text{prop}} l'$ , such that either (i)  $l$  and  $l'$  are formulae in  $X$  with the same root symbol  $f \in F$ , or (ii)  $X$  contains terms  $f(t_1, \dots, t_m, l, u_1, \dots, u_m)$  and  $f(t'_1, \dots, t'_m, l', u'_1, \dots, u'_m)$  with  $f \in F$ . We prove completeness assuming the non-trivial  $\text{EUF}^+$ : the proof rests on showing that if  $\mathcal{I}_{\text{EUF}}^m$  cannot expand an assignment, all equalities are determined.

**Theorem 3** *Module  $\mathcal{I}_{\text{EUF}}^m$  is leading-theory-complete for all leading theories.*

**Proof** Let  $\mathcal{T}_1$  be a leading theory, with signature  $\Sigma_1$  and extension  $\mathcal{T}_1^+$ , and  $J$  a plausible EUF-assignment that  $\mathcal{I}_{\text{EUF}}^m$  cannot expand. We show that  $J$  is leading-theory-compatible with EUF sharing  $G(J)$ . We begin by observing that every formula  $l \in G_{\text{prop}}(J)$  is relevant to EUF, and therefore  $J$  assigns a value to  $l$  by Part (2) of Lemma 1 (†). For  $s$  other than  $\text{prop}$ , every term  $u \in G_s(J)$  is relevant to EUF, as  $\text{EUF}^+$  has (infinitely many) values for such sorts. Moreover, the only EUF-inferences using first-order assignments are equality inferences, and therefore  $J$  assigns a value to every such term  $u$  by Part (3) of Lemma 1 (‡). It follows that  $\text{fv}_{\Sigma_{\text{EUF}}}(G(J)) = \text{fv}_{\Sigma_{\text{EUF}}}(J)$  by Corollary 1. Let  $\mathcal{M}_1$  be a  $\mathcal{T}_1^+[\mathcal{V}_1]$ -model such that

$\text{fv}_{\Sigma_1}(G(J)) \subseteq \mathcal{V}_1$  and  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$ . We build an  $\text{EUF}^+[\mathcal{V}]$ -model  $\mathcal{M}$  with  $\mathcal{V} = \text{fv}_{\Sigma_{\text{EUF}}}(J)$  that fulfills the requirements for leading-theory-compatibility (see Definition 5). First,  $\mathcal{M}$  interprets the sorts in  $S$  as  $\mathcal{M}_1$  does. This suffices for Part (ii) of Definition 5. Second,  $\mathcal{M}$  interprets every variable  $t \in \text{fv}_{\Sigma_{\text{EUF}}}(J)$  as  $\mathcal{M}_1(t)$ , every EUF-value  $c$  such that  $(t \leftarrow c) \in J$  as  $\mathcal{M}_1(t)$ , and every other EUF-value arbitrarily. The interpretation of EUF-values is well-defined, because if  $\{t \leftarrow c, u \leftarrow c\} \subseteq J$  then  $(t \simeq u) \in J_{\mathcal{T}_1}$ , by definition of  $\mathcal{T}_1$ -view and because  $\mathcal{T}_1$  has all the sorts, so that  $\mathcal{M}_1(t) = \mathcal{M}_1(u)$  since  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$ . Third and last,  $\mathcal{M}$  interprets every symbol  $f: (s_1 \times \dots \times s_m) \rightarrow s$  in  $F$  as follows: for all elements  $e_1 \in s_1^{\mathcal{M}_1} \dots e_m \in s_m^{\mathcal{M}_1}$ , if  $G(J)$  contains no term  $f(t_1, \dots, t_m)$  such that  $\mathcal{M}_1(t_1) = e_1, \dots, \mathcal{M}_1(t_m) = e_m$ , then  $f^{\mathcal{M}}(e_1, \dots, e_m)$  is an arbitrary element in  $s^{\mathcal{M}_1}$ ; otherwise,  $f^{\mathcal{M}}(e_1, \dots, e_m)$  is  $\mathcal{M}_1(f(t_1, \dots, t_m))$ . Note that  $f^{\mathcal{M}}$  is well-defined: indeed, if there is in  $G(J)$  another term  $f(u_1, \dots, u_m)$  such that  $\mathcal{M}_1(u_1) = e_1, \dots, \mathcal{M}_1(u_m) = e_m$ , then by  $(\dagger)$  and  $(\ddagger)$ ,  $J$  assigns values to  $t_1, \dots, t_m, u_1, \dots, u_m, f(t_1, \dots, t_m)$ , and  $f(u_1, \dots, u_m)$ . Also,  $J$  contains assignments  $(t_i \simeq u_i) \leftarrow b_i$ , for  $1 \leq i \leq m$ , and  $(f(t_1, \dots, t_m) \simeq f(u_1, \dots, u_m)) \leftarrow b$ , because otherwise an equality inference could expand it. The truth values  $b_1, \dots, b_m$  are all true, because  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$ . The truth value  $b$  is true, as otherwise inference rule (2) could expand  $J$ . Since  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$ ,  $\mathcal{M}_1(f(t_1, \dots, t_m)) = \mathcal{M}_1(f(u_1, \dots, u_m))$ , and  $f^{\mathcal{M}}$  is well-defined. This completes the construction of  $\mathcal{M}$ . For Part (i) of Definition 5, we need to show that for all  $(t \leftarrow c) \in J$ , we have  $\mathcal{M}(t) = \mathcal{M}_1(t) = c^{\mathcal{M}}$ . For Part (iii) of Definition 5, we need to show that for all  $t \in G(J)$ , we have  $\mathcal{M}(t) = \mathcal{M}_1(t)$ . Both claims are proved by a straightforward induction on the structure of terms.  $\square$

If  $\text{EUF}^+$  is trivial,  $\mathcal{I}_{\text{EUF}}^m$  is still leading-theory complete. The proof follows the same pattern: it is simpler as there are no EUF-values and no first-order assignments, and the key point is that the assignment gives a value to  $t \simeq_s u$  for all terms  $t$  and  $u$  of sort  $s \in S \setminus \{\text{prop}\}$ . Let  $\mathcal{I}_{\text{EUF}}$  be the module obtained by adding to  $\mathcal{I}_{\text{EUF}}^m$  inference rules that propagate consequences of assignments on the trail, according to the congruence axioms of equality for all  $f \in F$ :

$$\begin{aligned} (t_i \simeq u_i)_{i=1\dots m} \vdash_{\text{EUF}} f(t_1, \dots, t_m) \simeq f(u_1, \dots, u_m) \\ (t_i \simeq u_i)_{i=1\dots m, i \neq j}, f(t_1, \dots, t_m) \not\simeq f(u_1, \dots, u_m) \vdash_{\text{EUF}} t_j \not\simeq u_j. \end{aligned}$$

**Corollary 3** *Module  $\mathcal{I}_{\text{EUF}}$  is leading-theory-complete for all leading theories.*

### 4.3 The Theory of Arrays

The theory of arrays  $\text{Arr}$  features sorts for *arrays*, *indices*, and *values*, and function symbols to *select* and *store* array elements. Given a set of *basic sorts* that includes  $\text{prop}$ , let  $\Rightarrow$  be the *array sort constructor*, so that  $I \Rightarrow V$  is the sort of arrays with indices of sort  $I$  and values of sort  $V$ . We use  $a, b, c$ , and  $d$  for variables of an  $I \Rightarrow V$  sort,  $u$  and  $v$  for variables of sort  $V$ , and  $i, j$ , and  $k$  for variables of sort  $I$ . In signature  $\Sigma_{\text{Arr}} = (S, F)$ , the set of sorts  $S$  is the free closure of the set of basic sorts with respect to  $\Rightarrow$ , and the set of symbols  $F$  is

$$\begin{aligned} \simeq_S \cup \{ & \{\text{select}_{I \Rightarrow V} : (I \Rightarrow V) \times I \rightarrow V \} & | (I \Rightarrow V) \in S\} \\ & \cup \{ \{\text{store}_{I \Rightarrow V} : (I \Rightarrow V) \times I \times V \rightarrow (I \Rightarrow V) \} \} & | (I \Rightarrow V) \in S\} \\ & \cup \{ \{\text{diff}_{I \Rightarrow V} : (I \Rightarrow V) \times (I \Rightarrow V) \rightarrow I \} & | (I \Rightarrow V) \in S\}. \end{aligned}$$

Sort subscripts can be omitted when clear, and  $\text{store}(a, i, v)$  and  $\text{select}(a, i)$  may be abbreviated as  $a[i]:=v$  and  $a[i]$ . The symbol  $\text{diff}$  is the Skolem function symbol in the clausal form of the  $\rightarrow$  direction of the *extensionality* axiom  $\forall a \forall b ((\forall i a[i] \simeq b[i]) \leftrightarrow a \simeq b)$ : the

function *diff* maps two arrays to an index, called *witness*, where they differ. Similar to EUF, the extension  $\text{Arr}^+$  may either be trivial, or add a countably infinite set of values for each sort in  $S \setminus \{\text{prop}\}$  and no axioms. Module  $\mathcal{I}_{\text{Arr}}$  augments the equality inference rules of Fig. 3 with inference rules that apply when the trail violates an array axiom. Rules (3)–(5) detect violations of the *congruence* axioms for the  $\Sigma_{\text{Arr}}$ -symbols:

$$a \simeq b, i \simeq j, a[i] \not\simeq b[j] \vdash_{\text{Arr}} \perp \quad (3)$$

$$a \simeq b, i \simeq j, u \simeq v, (a[i]:=u) \not\simeq (b[j]:=v) \vdash_{\text{Arr}} \perp \quad (4)$$

$$a \simeq c, b \simeq d, \text{diff}(a, b) \not\simeq \text{diff}(c, d) \vdash_{\text{Arr}} \perp. \quad (5)$$

Violations of the *select-over-store* axioms

$$\forall a \forall b \forall i \forall j \forall v (i \simeq j \rightarrow \text{select}(\text{store}(a, i, v), j) \simeq v)$$

$$\forall a \forall b \forall i \forall j \forall v (i \not\simeq j \rightarrow \text{select}(\text{store}(a, i, v), j) \simeq \text{select}(a, j))$$

are detected by rules (6) and (7) of  $\mathcal{I}_{\text{Arr}}$ :

$$b \simeq (a[i]:=v), i \simeq j, b[j] \not\simeq v \vdash_{\text{Arr}} \perp \quad (6)$$

$$b \simeq (a[i]:=v), i \not\simeq j, j \simeq k, a[j] \not\simeq b[k] \vdash_{\text{Arr}} \perp. \quad (7)$$

The last inference rule builds into  $\mathcal{I}_{\text{Arr}}$  the extensionality axiom:

$$a \not\simeq b \vdash_{\text{Arr}} a[\text{diff}(a, b)] \not\simeq b[\text{diff}(a, b)]. \quad (8)$$

Most Arr-satisfiability procedures replace every disequality between arrays with a disequality between their elements at the witness index in a preprocessing phase (see [9, Sect. 6 and 7] for more background and references). Rule (8) is the only rule of  $\mathcal{I}_{\text{Arr}}$  that produces new terms. Similar to  $\mathcal{I}_{\text{EUF}}$ ,  $\mathcal{I}_{\text{Arr}}$  reasons about Arr-values, if present, by the equality inference rules, treating Arr-values as labels of equivalence classes.

For the local basis, for all sets  $X$  of terms,  $\text{basis}_{\text{Arr}}(X)$  is the smallest closed set  $Y$  such that  $X \subseteq Y$ ,  $\top \in Y$ , and:

1. For all terms  $l_1$  and  $l_2$  of sort *prop* that occur as subterms of terms in  $Y$  with *select*, *store*, or *diff* as root symbol,  $(l_1 \simeq_{\text{prop}} l_2) \in Y$ ;
2. For all terms  $t, u \in Y$  of an array sort,  $t[\text{diff}(t, u)] \in Y$  and  $u[\text{diff}(t, u)] \in Y$ .

Clause (1) adds equalities between formulae that may be needed and whose presence is not already guaranteed by closedness of local bases. Clause (2) adds the terms that may be generated by rule (8); it preserves finiteness, because *diff* produces terms of an index sort which is structurally smaller, in terms of the array sort constructor  $\Rightarrow$ , than the array sort of its arguments.

As arrays represent functions that can be updated, a model can interpret an array as an *updatable function* and an array sort as a set of updatable functions. Given generic sets  $\mathcal{U}$  and  $\mathcal{V}$ , let  $\mathcal{V}^{\mathcal{U}}$  denote the set of functions from  $\mathcal{U}$  to  $\mathcal{V}$ . A set  $\mathcal{W} \subseteq \mathcal{V}^{\mathcal{U}}$  is an *updatable function set from  $\mathcal{U}$  to  $\mathcal{V}$* , if every function obtained by a finite number of updates to a function in  $\mathcal{W}$  is in  $\mathcal{W}$ . As done for EUF, we prove completeness assuming a non-trivial extension.

**Theorem 4** *Module  $\mathcal{I}_{\text{Arr}}$  is leading-theory-complete for all leading theories  $\mathcal{T}_1$  such that for all  $\mathcal{T}_1$ -models  $\mathcal{M}_1$  and array sorts  $I \Rightarrow V$  of  $\Sigma_{\text{Arr}}$ , there is an updatable function set  $X$  from  $I^{\mathcal{M}_1}$  to  $V^{\mathcal{M}_1}$  such that  $|(I \Rightarrow V)^{\mathcal{M}_1}| = |X|$ .*

**Proof** Let  $J$  be a plausible Arr-assignment that  $\mathcal{I}_{\text{Arr}}$  cannot expand. We show that  $J$  is leading-theory-compatible with Arr sharing  $G(J)$ . By the same reasoning at the beginning of the proof

of Theorem 3,  $J$  assigns values to all terms in  $G(J)$  ( $\dagger$ ), and  $\text{fv}_{\Sigma_{\text{Arr}}}(G(J)) = \text{fv}_{\Sigma_{\text{Arr}}}(J)$  by Corollary 1. Let  $\mathcal{T}_1$  be a leading theory that satisfies the hypothesis,  $\Sigma_1$  its signature,  $\mathcal{T}_1^+$  its extension, and  $\mathcal{M}_1$  a  $\mathcal{T}_1^+[\mathcal{V}_1]$ -model such that  $\text{fv}_{\Sigma_1}(G(J)) \subseteq \mathcal{V}_1$  and  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$ . For all array sorts  $I \Rightarrow V$  of  $\Sigma_{\text{Arr}}$ , let  $X$  be the updatable function set from  $I^{\mathcal{M}_1}$  to  $V^{\mathcal{M}_1}$  such that  $|(I \Rightarrow V)^{\mathcal{M}_1}| = |X|$ . We organize the proof in two parts.

1. *Definition of a bijective function  $\phi: (I \Rightarrow V)^{\mathcal{M}_1} \rightarrow X$ :*

We pick an updatable function  $f_0 \in X$  that will be used as default in the sequel. Then, we begin by defining the restriction  $\phi_Y$  of  $\phi$  to the finite subset  $Y \subseteq (I \Rightarrow V)^{\mathcal{M}_1}$  consisting of those elements  $a$  such that  $\mathcal{M}_1(t) = a$  for some term  $t \in G(J)$ . For all  $a \in Y$ , let  $\mathcal{R}_a \subseteq I^{\mathcal{M}_1} \times V^{\mathcal{M}_1}$  be the relation defined by the following set of pairs:

$$\begin{aligned} & \{(\mathcal{M}_1(i), \mathcal{M}_1(t[i])) \mid t[i] \in G(J), \mathcal{M}_1(t) = a\} \cup \\ & \{(\mathcal{M}_1(i), \mathcal{M}_1(u)) \mid t[i] := u \in G(J), \mathcal{M}_1(t[i] := u) = a\} \cup \\ & \{(\mathcal{M}_1(i), \mathcal{M}_1(t[i])) \mid t[k] := u \in G(J), \mathcal{M}_1(t[k] := u) = a, \mathcal{M}_1(i) \neq \mathcal{M}_1(k)\}. \end{aligned}$$

In other words,  $\mathcal{R}_a$  is the set of index-value pairs dictated by those terms in  $G(J)$  where either `select` is applied to an array term that  $\mathcal{M}_1$  interprets as  $a$  or the application of `store` forms an array term that  $\mathcal{M}_1$  interprets as  $a$ . Since  $G(J)$  is finite,  $\mathcal{R}_a$  is finite. Also,  $\mathcal{R}_a$  is a partial function  $\mathcal{R}_a: I^{\mathcal{M}_1} \rightarrow V^{\mathcal{M}_1}$ , because otherwise  $\mathcal{I}_{\text{Arr}}$  could expand  $J$  by rules (6)–(7). Let  $\phi_Y(a)$  be the total function that is identical to  $\mathcal{R}_a$  where  $\mathcal{R}_a$  is defined, and maps every  $e \in I^{\mathcal{M}_1}$  where  $\mathcal{R}_a$  is undefined to  $f_0(e) \in V^{\mathcal{M}_1}$ . Since  $\mathcal{R}_a$  is finite,  $\phi_Y(a)$  differs from  $f_0$  by finitely many updates, and therefore  $\phi_Y(a) \in X$ .

Next, we show that  $\phi_Y$  is injective. By way of contradiction, suppose that there are two elements  $a, a' \in Y$  such that  $a \neq a'$  and  $\phi_Y(a) = \phi_Y(a')$ . Since  $a, a' \in Y$ , it is  $a = \mathcal{M}_1(t)$  and  $a' = \mathcal{M}_1(t')$  for some terms  $t, t' \in G(J)$ . This means that  $\mathcal{M}_1 \models t \neq t'$ . By ( $\dagger$ ),  $J$  assigns values to  $t$  and  $t'$ , and therefore it also assigns a truth value  $\mathfrak{b}$  to  $t \simeq t'$ , because otherwise an equality inference could expand it. Also,  $((t \simeq t') \leftarrow \mathfrak{b}) \in J_{\mathcal{T}_1}$  by definition of theory view. Since  $\mathcal{M}_1 \models t \neq t'$  and  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$ ,  $\mathfrak{b}$  must be false, or, equivalently,  $(t \neq t') \in J$ . Therefore, also  $t[\text{diff}(t, t')] \neq t'[\text{diff}(t, t')] \in J$ , because otherwise  $\mathcal{I}_{\text{Arr}}$  could expand  $J$  by rule (8). As before,  $(t[\text{diff}(t, t')] \neq t'[\text{diff}(t, t')]) \in J_{\mathcal{T}_1}$ . Since  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$ , it follows that  $\mathcal{M}_1(t[\text{diff}(t, t')]) \neq \mathcal{M}_1(t'[\text{diff}(t, t')])$ . By definition of  $\phi_Y(a)$  for a generic  $a$ , we have:

$$\begin{aligned} \phi_Y(a)(\mathcal{M}_1(\text{diff}(t, t'))) &= \mathcal{M}_1(t[\text{diff}(t, t')]) \\ \phi_Y(a')(\mathcal{M}_1(\text{diff}(t, t'))) &= \mathcal{M}_1(t'[\text{diff}(t, t')]). \end{aligned}$$

Since the two right-hand sides are different, the two left-hand sides are also different, so that  $\phi_Y(a) \neq \phi_Y(a')$ , a contradiction.

Given that  $\phi_Y$  is injective, we can extend  $\phi_Y$  to the sought-after bijective function  $\phi$ , by taking as pre-images of the elements of  $X$  that are not images of elements of  $Y$  other elements of  $(I \Rightarrow V)^{\mathcal{M}_1}$  and there are enough distinct such elements as  $|(I \Rightarrow V)^{\mathcal{M}_1}| = |X|$ .

2. *Construction of an  $\text{Arr}^+[\mathcal{V}]$ -model  $\mathcal{M}$  with  $\mathcal{V} = \text{fv}_{\Sigma_{\text{Arr}}}(J)$ :*

The first part of the definition of  $\mathcal{M}$  follows the same pattern as in the proof of Theorem 3:  $\mathcal{M}$  interprets all sorts in  $S$ , all variables  $t \in \text{fv}_{\Sigma_{\text{Arr}}}(J)$ , and all  $\text{Arr}$ -values  $c$  such that  $(t \leftarrow c) \in J$ , as  $\mathcal{M}_1$  does, and all other  $\text{Arr}$ -values arbitrary. The point on sorts suffices for Part (ii) of Definition 5. Then, for all array sorts  $I \Rightarrow V$ ,  $\mathcal{M}$  interprets the `select`, `store` and `diff` symbols as follows:

- For all array-index pairs  $(a, e) \in (I \Rightarrow V)^{\mathcal{M}} \times I^{\mathcal{M}}$ , let  $\text{select}_{I \Rightarrow V}^{\mathcal{M}}(a, e) = \phi(a)(e) \in V^{\mathcal{M}}$ ;

- For all array-index-value triples  $(a, e, v) \in (I \Rightarrow V)^{\mathcal{M}} \times I^{\mathcal{M}} \times V^{\mathcal{M}}$ , let  $f \in X$  be the function mapping  $e$  to  $v$  and every other  $e' \in I^{\mathcal{M}}$  to  $\phi(a)(e') \in V^{\mathcal{M}}$ ; then  $\text{store}_{I \Rightarrow V}^{\mathcal{M}}(a, e, v) = \phi^{-1}(f) \in (I \Rightarrow V)^{\mathcal{M}}$ ;
- For all pairs  $(a, a') \in (I \Rightarrow V)^{\mathcal{M}} \times (I \Rightarrow V)^{\mathcal{M}}$  with  $a \neq a'$ ,  $\text{diff}_{I \Rightarrow V}^{\mathcal{M}}(a, a') = e \in I^{\mathcal{M}}$  such that  $\phi(a)(e) \neq \phi(a')(e)$ , and  $\text{diff}_{I \Rightarrow V}^{\mathcal{M}}(a, a)$  is an arbitrary element of  $I^{\mathcal{M}}$ .

By construction,  $\mathcal{M}$  satisfies the Arr-axioms and it is an  $\text{Arr}^+[\text{fv}_{\Sigma_{\text{Arr}}}(J)]$ -model. Parts (i) and (iii) of Definition 5 follow by induction on the term structure.  $\square$

The same property holds for the trivial  $\text{Arr}^+$  with an almost identical proof, except that non-Boolean terms in  $G(J)$  are not assigned values. Rules obtained from rules (3)–(7) by removing the last premise and adding its flip as conclusion can be added to  $\mathcal{I}_{\text{Arr}}$ , preserving leading-theory-completeness by Lemma 2.

### 4.4 Linear Rational Arithmetic

Theory LRA has signature  $\Sigma_{\text{LRA}}$  with sorts  $S_{\text{LRA}} = \{\text{prop}, \mathbb{Q}\}$  and set of symbols  $F_{\text{LRA}}$  with equality symbols  $\simeq_{\{\text{prop}, \mathbb{Q}\}}$ , the constant  $1 : \mathbb{Q}$ , the symbol  $+: (\mathbb{Q} \times \mathbb{Q}) \rightarrow \mathbb{Q}$  for addition, the predicates  $<, \leq : (\mathbb{Q} \times \mathbb{Q}) \rightarrow \text{prop}$  for the orderings, and the collection of unary function symbols  $\{c : \mathbb{Q} \rightarrow \mathbb{Q} \mid c \in \mathbb{Q}\}$ , indexed by the set  $\mathbb{Q}$  of the rational numbers, for multiplication by constants. The extension  $\text{LRA}^+$  adds constants for all rational numbers, namely  $\Sigma_{\text{LRA}}^+ = (\{\text{prop}, \mathbb{Q}\}, F_{\text{LRA}} \cup \{\tilde{q} : \mathbb{Q} \mid q \in \mathbb{Q}\})$  with axioms  $\tilde{q} \simeq_{\mathbb{Q}} q \cdot 1$  for all  $q \in \mathbb{Q}$ .

The *Fourier-Motzkin (FM) algorithm* [41,43,52] determines the satisfiability of a set of linear inequalities over the reals, by applying *variable elimination* until either it generates a contradiction, in the form of a constraint  $0 \leq \tilde{q}$  with negative  $\tilde{q}$ , in which case the algorithm returns unsatisfiable, or it eliminates all variables, in which case the algorithm returns satisfiable.

Variable elimination works as follows: select a variable  $x$ ; if  $x$  appears only with positive, or negative, coefficients, remove all constraints where  $x$  appears; otherwise, compute *all linear combinations* of constraints  $t_1 + c_1 \cdot x \leq u_1$  and  $t_2 - c_2 \cdot x \leq u_2$  where  $x$  appears with positive and negative coefficient, respectively, generating the constraint  $c_2 \cdot t_1 + c_1 \cdot t_2 \leq c_2 \cdot u_1 + c_1 \cdot u_2$  (if a premise is a strict inequality, the result is also strict). Alternatively, the constraints where  $x$  appears with positive coefficient are rearranged into *upper bounds*  $x \leq t$ , those where  $x$  appears with negative coefficient are rearranged into *lower bounds*  $u \leq x$ , and the computation of all linear combinations is replaced by that of *all transitive closures*, concatenating  $u \leq x$  and  $x \leq t$  to generate  $u \leq t$  (if a premise is strict, the result is also).

Since a linear combination eliminating  $x$  recalls a propositional resolution inference eliminating the propositional variable of the literals resolved upon, a *single* linear combination, or transitive closure step, is known as *Fourier-Motzkin (FM) resolution*. A variable  $x$  appearing only with one sign parallels a *pure literal* in a set of clauses, and the elimination of all constraints where  $x$  occurs reminds one of the *pure literal rule* that eliminates, or deems satisfied, all clauses where a pure literal occurs [18]. The FM-algorithm resembles the *level-saturation strategy* for resolution: select a propositional variable  $l$ , add all resolvents generated by resolving upon  $l$ , remove all clauses where  $l$  appears, and repeat, until either the empty clause arises or the set is emptied.

Given  $m$  constraints in  $n$  variables the FM-algorithm generates  $\frac{m \cdot 2^n}{4^n}$  constraints in the worst case (e.g., [41]), whereas the *simplex algorithm* (e.g., [41,52]) is exponential in the worst case [39], but polynomial in practice [54]. Thus, most LRA-satisfiability procedures adopt a modern version of the *simplex algorithm* that deals also with strict inequalities [26].

**Fig. 4** An infinite series of FM-resolution inferences from input  $R = \{l_0, l_1, l_2\}$

$$\begin{aligned}
 l_0 &: -2 \cdot x - y < 0 \\
 l_1 &: x + y < 0 \\
 l_2 &: x < -1 \\
 l_3 &: -y < -2 \quad (l_0 + 2l_2) \\
 l_4 &: x < -2 \quad (l_1 + l_3) \\
 l_5 &: -y < -4 \quad (l_0 + 2l_4) \\
 l_6 &: x < -4 \quad (l_1 + l_5) \\
 l_7 &: -y < -8 \quad (l_0 + 2l_6) \\
 &\dots
 \end{aligned}$$

However, conflict-driven LRA-satisfiability procedures [21,40,46] apply FM-resolution only to explain LRA-conflicts, just like the CDCL procedure applies resolution only to explain Boolean conflicts [44,45]. These procedures stand to the FM-algorithm like CDCL stands to level-saturation by resolution. Similar to MCSAT [35], CDSAT embeds a conflict-driven LRA-satisfiability procedure and applies FM-resolution only to solve conflicts. A difference between CDSAT and MCSAT in this regard is that the Deduce rule of CDSAT covers both propagation and conflict explanation, allowing CDSAT to apply FM-resolution more liberally. Therefore, we consider an  $\mathcal{I}_{\text{LRA}}$  module that features *FM-resolution*:

$$t_1 <_1 x, x <_2 t_2 \vdash_{\text{LRA}} t_1 <_3 t_2,$$

where  $t_1, t_2$ , and  $t_3$  are terms of sort  $\mathbf{Q}$ ,  $<_1, <_2, <_3 \in \{<, \leq\}$ , and  $<_3$  is  $<$  if and only if either  $<_1$  or  $<_2$  is  $<$ . Since FM-resolution concatenates inequalities,  $\mathcal{I}_{\text{LRA}}$  has *equality elimination* rules to replace an equality by inequalities:

$$t_1 \simeq_{\mathbf{Q}} t_2 \vdash_{\text{LRA}} t_1 \leq t_2 \quad t_1 \simeq_{\mathbf{Q}} t_2 \vdash_{\text{LRA}} t_2 \leq t_1,$$

and *positivization* rules to handle flipped inequalities based on the totality of the ordering on  $\mathbf{Q}$ :

$$\overline{t_1 < t_2} \vdash_{\text{LRA}} t_2 \leq t_1 \quad \overline{t_1 \leq t_2} \vdash_{\text{LRA}} t_2 < t_1.$$

Let  $t_0, \dots, t_m$  be terms of sort  $\mathbf{Q}$ , and  $l$  a formula whose normal form is in the closure of  $t_1, \dots, t_m$  with respect to the symbols of  $F_{\text{LRA}}$ . Module  $\mathcal{I}_{\text{LRA}}$  also has an *evaluation* rule to evaluate the truth value of  $l$  when values for  $t_1, \dots, t_m$  are available on the trail:

$$t_1 \leftarrow \tilde{q}_1, \dots, t_m \leftarrow \tilde{q}_m \vdash_{\text{LRA}} l \leftarrow b.$$

For example,  $(z \leftarrow 1) \vdash_{\text{LRA}} (w + 2 \simeq_{\mathbf{Q}} w + z) \leftarrow \text{false}$  is an evaluation inference, which does not need a value for  $w$ , because the normal form of  $w + 2 \simeq_{\mathbf{Q}} w + z$  is  $-z + 2 \simeq_{\mathbf{Q}} 0$ . Let  $x$  be a free  $\Sigma_{\text{LRA}}$ -variable of sort  $\mathbf{Q}$  that does not occur free in  $t_0, t_1$ , and  $t_2$ . The last rule of  $\mathcal{I}_{\text{LRA}}$ , beside the equality rules of Fig. 3, is *disequality elimination*, which detects a situation where there is no value for  $x$ :

$$t_1 \leq x, x \leq t_2, t_1 \simeq_{\mathbf{Q}} t_0, t_2 \simeq_{\mathbf{Q}} t_0, x \not\leq_{\mathbf{Q}} t_0 \vdash_{\text{LRA}} \perp.$$

FM-resolution and disequality elimination apply also to formulae reducible to the form of their premises, as in  $y - x < 0, 3 \cdot x < 5 \vdash_{\text{LRA}} y < \frac{5}{3}$  for FM-resolution.

The FM-algorithm bundles in one step all FM-resolutions on one variable, eliminating it altogether; as there are finitely many variables, the algorithm terminates. However, other strategies for applying FM-resolution may generate infinitely many terms as shown in Fig. 4: the never-halting series alternates FM-resolutions on a variable  $x$  with FM-resolutions on another variable  $y$ .

In CDSAT, the FM-algorithm can be emulated with  $\mathcal{I}_{\text{LRA}}$ -inferences, as a series of Deduce transitions applied with a *level-saturation search plan*. The local basis can be set to those

terms that are newly generated by the algorithm, in finite numbers, so that termination follows. This search plan thus provides a decision procedure for satisfiability, but it is not conflict-driven and is as inefficient as the FM-algorithm. Other search plans may be more interesting, but may raise termination issues: were it not for the finite global basis of CDSAT that forces termination, the infinite series of Fig. 4 could also be emulated in CDSAT, for instance as a never-ending search phase that never generates any conflict. While a conflict-driven search plan would not apply Deduce in this manner, this infinite series may ensue also if Deduce only explains conflicts, as detailed in the following example.

**Example 7** Consider the set  $R = \{l_0: -2 \cdot x - y < 0, l_1: x + y < 0, l_2: x < -1\}$  of Fig. 4. Suppose that Decide tries  $y \leftarrow 0$ . The LRA-procedure sees LRA-conflict  $\{-2 \cdot x - y < 0, x < -1, y \leftarrow 0\}$  and explains it by the FM-resolution inference  $\{-y < 2 \cdot x, 2 \cdot x < -2\} \vdash_{\text{LRA}} -y < -2$ , so that Deduce places  $l_3: -y < -2$  on the trail. Literal  $l_3$  is a late propagation, as it has level 0, but it comes after the first decision. The evaluation inference  $y \leftarrow 0 \vdash_{\text{LRA}} \overline{-y < -2}$  reveals conflict  $\{y \leftarrow 0, -y < -2\}$  on the trail. Since its level is 1, ConflictSolve fires, and UndoClear solves the conflict by undoing  $y \leftarrow 0$ . If Decide tries next  $x \leftarrow -2$ , the LRA-procedure detects LRA-conflict  $\{x + y < 0, -y < -2, x \leftarrow -2\}$  and explains it by the FM-resolution inference  $\{x < -y, -y < -2\} \vdash_{\text{LRA}} x < -2$ , so that Deduce puts  $l_4: x < -2$  on the trail. The evaluation inference  $x \leftarrow -2 \vdash_{\text{LRA}} \overline{x < -2}$  exposes conflict  $\{x \leftarrow -2, x < -2\}$  on the trail. Since its level is 1, ConflictSolve applies, and UndoClear solves the conflict by retracting  $x \leftarrow -2$ . A subsequent Decide with  $y \leftarrow 3$  causes LRA-conflict  $\{-2 \cdot x - y < 0, x < -2, y \leftarrow 3\}$ , that Deduce explains by the FM-resolution  $\{-y < 2 \cdot x, 2 \cdot x < -4\} \vdash_{\text{LRA}} -y < -4$ , generating  $l_5: -y < -4$ . The evaluation inference  $y \leftarrow 3 \vdash_{\text{LRA}} \overline{-y < -4}$  gets conflict  $\{y \leftarrow 3, -y < -4\}$  on the trail, and the same ConflictSolve UndoClear pair of transitions undoes  $y \leftarrow 3$ . Again, a Decide with  $x \leftarrow -3$  leads to LRA-conflict  $\{x + y < 0, -y < -4, x \leftarrow -3\}$ , explained by Deduce with the FM-resolution  $\{x < -y, -y < -4\} \vdash_{\text{LRA}} x < -4$ , so that  $l_6: x < -4$  is added to the trail. Evaluation inference  $x \leftarrow -3 \vdash_{\text{LRA}} \overline{x < -4}$  unveils conflict  $\{x \leftarrow -3, x < -4\}$  on the trail, so that ConflictSolve and UndoClear apply, repealing  $x \leftarrow -3$ . The last FM-resolution in Fig. 4 may respond to a Decide with  $y \leftarrow 5$ , so that LRA-conflict  $\{-2 \cdot x - y < 0, x < -4, y \leftarrow 5\}$  is explained by Deduce with FM-resolution  $\{-y < 2 \cdot x, 2 \cdot x < -8\} \vdash_{\text{LRA}} -y < -8$ , adding  $l_7: -y < -8$  to the trail. The evaluation inference  $y \leftarrow 5 \vdash_{\text{LRA}} \overline{-y < -8}$  shows conflict  $\{y \leftarrow 5, -y < -8\}$  on the trail, so that ConflictSolve applies and UndoClear removes  $y \leftarrow 5$ .

A well-known solution to this problem assumes a *total ordering*  $<_{\text{LRA}}$  on  $\Sigma_{\text{LRA}}$ -variables of sort Q and restricts FM-resolution by requiring that the resolved variable  $x$  is  $<_{\text{LRA}}$ -maximum in both premises [11,21,35,40,46].

**Example 8** Assuming  $y <_{\text{LRA}} x$ , the first FM-resolution step in Fig. 4, namely  $\{-y < 2 \cdot x, 2 \cdot x < -2\} \vdash_{\text{LRA}} -y < -2$ , still applies, as it eliminates the  $<_{\text{LRA}}$ -maximum variable  $x$ , and generates  $l_3: -y < -2$ . The second FM-resolution step of the diverging series, namely  $\{x < -y, -y < -2\} \vdash_{\text{LRA}} x < -2$ , is barred, because  $y$  is not the  $<_{\text{LRA}}$ -maximum variable in the premises. Thus, all CDSAT-derivations embedding that diverging series of FM-resolution inferences are excluded. Multiple CDSAT-derivations discover that  $R$  is LRA-unsatisfiable. One that does it by mere theory propagations at level 0 begins with Deduce placing  $l_3$  on the trail. Another Deduce applies FM-resolution to compute linear combination  $l_0 + 2l_1$  as  $\{-y < 2 \cdot x, 2 \cdot x < -2 \cdot y\} \vdash_{\text{LRA}} -y < -2 \cdot y$ , adding the normal form  $l_4: y < 0$  of  $-y < -2 \cdot y$  to the trail. A third Deduce with FM-resolution inference  $\{2 < y, y < 0\} \vdash_{\text{LRA}} 2 < 0$ , computing linear combination  $-l_3 + l_4$ , expands the trail with  $l_5: 2 < 0$ . The evaluation step  $\emptyset \vdash_{\text{LRA}} \overline{2 < 0}$  leads to a Fail transition as  $2 < 0$  has level 0.

In CDSAT, termination is ensured by the finiteness of the global basis  $\mathcal{B}$  which restricts Deduce. For completeness,  $\mathcal{B}$  must be stable, requiring in particular that  $\text{basis}_{\text{LRA}}(\mathcal{B}) \subseteq \mathcal{B}$  for the local basis  $\text{basis}_{\text{LRA}}$ . Thus,  $\text{basis}_{\text{LRA}}$  must be limited so as to never introduce infinitely many terms, which is obtained by incorporating the restriction to FM-resolution as follows. For all sets  $X$  of terms,  $\text{basis}_{\text{LRA}}(X)$  is the smallest closed set  $Y$  such that  $X \subseteq Y$ ,  $\top \in Y$ , and, for all terms  $t_1$  and  $t_2$  of sort  $Q$ :

1. If  $t_1 \simeq_Q t_2 \in Y$  then  $t_1 \leq t_2 \in Y$  and  $t_2 \leq t_1 \in Y$ ;
2. If  $(t_1 < t_2) \in Y$  then  $(t_2 \leq t_1) \in Y$ , and if  $(t_1 \leq t_2) \in Y$  then  $(t_2 < t_1) \in Y$ ;
3. If  $(t_1 \prec_1 x) \in Y$  and  $(x \prec_2 t_2) \in Y$ , then  $(t_1 \prec_3 t_2) \in Y$ , where  $\prec_1, \prec_2, \prec_3 \in \{<, \leq\}$ ,  $\prec_3$  is  $<$  if and only if either  $\prec_1$  or  $\prec_2$  is  $<$ , and  $x$  is the  $\prec_{\text{LRA}}$ -maximum variable in both  $\text{fv}_{\Sigma_{\text{LRA}}}^Q(t_1 \prec_1 x)$  and  $\text{fv}_{\Sigma_{\text{LRA}}}^Q(x \prec_2 t_2)$ .

Clauses (1) and (2) add the terms that may be generated by the equality elimination and positivization rules, respectively, which do not challenge finiteness. Clause (3) adds the terms that may be inferred by FM-resolution, and it preserves finiteness thanks to the  $\prec_{\text{LRA}}$ -based restriction. The side-condition of Deduce (see Fig. 1) ensures that the evaluation rule evaluates a formula in  $\mathcal{B}$ .

For  $\mathcal{I}_{\text{LRA}}$  to be complete with FM-resolution thus restricted, it suffices to add the following inference rule, named *detection of an empty solution space*:

$$\{y_1 \leftarrow \tilde{q}_1, \dots, y_m \leftarrow \tilde{q}_m\} \uplus E \vdash_{\text{LRA}} \perp$$

where  $y_1, \dots, y_m$  are  $\Sigma_{\text{LRA}}$ -variables of sort  $Q$ ,  $E$  is an LRA-assignment such that for all  $x$  in  $\text{fv}_{\Sigma_{\text{LRA}}}^Q(E)$ ,  $x \prec_{\text{LRA}} y_i$  or  $x = y_i$  for some  $i$ ,  $1 \leq i \leq m$ , and  $\{y_1 \leftarrow \tilde{q}_1, \dots, y_m \leftarrow \tilde{q}_m\} \uplus E$  is unsatisfiable. Alternatively, and in practice, since Deduce applies FM-resolution to explain LRA-conflicts typically due to decisions on rational variables, one may adopt a search plan that selects rational variables for decisions in  $\prec_{\text{LRA}}$ -increasing order. We call such a search plan *sensible*. An LRA-assignment  $J$  generated by a sensible search plan is also termed *sensible* and has the following property: for all variables  $x, y \in \text{fv}_{\Sigma_{\text{LRA}}}^Q(J)$ , if  $x \prec_{\text{LRA}} y$  and  $J$  assigns a value to  $y$ , then  $J$  assigns a value to  $x$ .

Toward completeness, since  $\mathcal{I}_{\text{LRA}}$  does not fulfill Condition (ii) of Part (3) of Lemma 1, we prove another lemma. Preliminarily we observe that the evaluation rule of  $\mathcal{I}_{\text{LRA}}$  subsumes the equality inference rules that take as premises first-order assignments (the first two in Fig. 3), and therefore we can assume that evaluation and detection of an empty solution space are the only rules of  $\mathcal{I}_{\text{LRA}}$  that deal with first-order assignments. Also, the only sort of LRA other than  $\text{prop}$  is  $Q$ , and all terms in  $G_Q(J)$  are relevant to LRA in an LRA-assignment  $J$ . Given LRA-assignment  $J$  and variable  $x \in \text{fv}_{\Sigma_{\text{LRA}}}^Q(J)$ , a *unit constraint* [35] about  $x$  in  $J$  is a singleton Boolean assignment  $L \in J$  where only  $x$  is unassigned:  $J$  assigns a value to all  $y, y \in \text{fv}_{\Sigma_{\text{LRA}}}^Q(L)$  and  $y \neq x$ .

**Lemma 3** *If module  $\mathcal{I}_{\text{LRA}}$  cannot expand a plausible LRA-assignment  $J$ , then  $J$  assigns values to all terms in  $G_Q(J)$ .*

**Proof** As a preliminary remark, all Boolean assignments in  $J$  concern terms of the form  $t_1 \leq t_2$ ,  $t_1 < t_2$ , or  $t_1 \not\prec_Q t_2$ , because otherwise an equality elimination or positivization inference rule could expand  $J$ . We begin by showing that  $J$  assigns values to all variables in  $\text{fv}_{\Sigma_{\text{LRA}}}^Q(J)$ . By way of contradiction, assume this is not the case, and let  $x$  be the  $\prec_{\text{LRA}}$ -smallest variable to which  $J$  does not assign a value. If  $J$  is sensible, for all variables  $y \in \text{fv}_{\Sigma_{\text{LRA}}}^Q(J)$  such that  $y \neq x$ , if  $J$  assigns a value to  $y$  then  $y \prec_{\text{LRA}} x$  ( $\dagger$ ). No LRA-assignment  $x \leftarrow \tilde{q}$  is acceptable for  $\mathcal{I}_{\text{LRA}}$  in  $J$  ( $\ddagger$ ), because otherwise  $\mathcal{I}_{\text{LRA}}$  could expand  $J$  by a decision. Property

( $\ddagger$ ) implies that for all values  $\tilde{q}$  there exist  $L \in \tilde{J}$  and  $J' \subseteq J$  such that  $J' \cup \{x \leftarrow \tilde{q}\} \vdash_{\text{LRA}} \bar{L}$ . This means that the space of possible solutions for  $x$  is empty: we distinguish three cases.

1. For variable  $x$  the lower bound is greater than the upper bound:  
 $E = \{t_1 \leq x, x \leq t_2, t_1 \leftarrow \tilde{q}_1, t_2 \leftarrow \tilde{q}_2\} \subseteq J$  and  $\tilde{q}_2 < \tilde{q}_1$ ; every assignment  $x \leftarrow \tilde{q}$  triggers an evaluation inference contradicting either  $t_1 \leq x$  or  $x \leq t_2$  or both. It follows that  $t_1 \leq x$  and  $x \leq t_2$  are unit constraints about  $x$  in  $J$ , because the evaluation rule determines the value of a Boolean term when all its rational subterms are assigned. If  $x$  is the  $\prec_{\text{LRA}}$ -maximum variable in  $\text{fv}_{\Sigma_{\text{LRA}}}^{\text{Q}}(t_1 \leq x) \cup \text{fv}_{\Sigma_{\text{LRA}}}^{\text{Q}}(x \leq t_2)$ , the FM-resolution inference  $\{t_1 \leq x, x \leq t_2\} \vdash_{\text{LRA}} t_1 \leq t_2$  is enabled. If  $J$  is sensible, this is guaranteed by ( $\ddagger$ ). Otherwise, if  $y_1, \dots, y_m$  are the variables other than  $x$  in  $\text{fv}_{\Sigma_{\text{LRA}}}^{\text{Q}}(t_1 \leq x) \cup \text{fv}_{\Sigma_{\text{LRA}}}^{\text{Q}}(x \leq t_2)$ ,  $x \prec_{\text{LRA}} y_i$  for some  $i, 1 \leq i \leq m$ . Since  $t_1 \leq x$  and  $x \leq t_2$  are unit constraints about  $x$  in  $J$ ,  $\{y_1 \leftarrow \tilde{q}_3, \dots, y_m \leftarrow \tilde{q}_{3+k}\} \subseteq J$  ( $3+k = m$ ). As  $\{y_1 \leftarrow \tilde{q}_3, \dots, y_m \leftarrow \tilde{q}_{3+k}\} \uplus E$  is unsatisfiable, an inference by the detection of an empty solution space rule is enabled.
2. For variable  $x$  the lower bound and the upper bound are equal, but one of them is strict: either  $E = \{t_1 < x, x \leq t_2, t_1 \leftarrow \tilde{q}, t_2 \leftarrow \tilde{q}\} \subseteq J$  or  $E = \{t_1 \leq x, x < t_2, t_1 \leftarrow \tilde{q}, t_2 \leftarrow \tilde{q}\} \subseteq J$ . The reasoning is the same as in Case (1) except that the enabled instance of FM-resolution is either  $\{t_1 < x, x \leq t_2\} \vdash_{\text{LRA}} t_1 < t_2$  or  $\{t_1 \leq x, x < t_2\} \vdash_{\text{LRA}} t_1 < t_2$ .
3. The lower bound and the upper bound for  $x$  are equal, and neither is strict, but a disequality excludes the only possible value:  $\{t_1 \leq x, x \leq t_2, t_1 \simeq t_0, t_2 \simeq t_0, x \not\approx t_0\} \subseteq J$ , so that a disequality elimination is enabled.

In all three cases an inference is enabled, contradicting the hypothesis. Thus,  $J$  assigns values to all variables in  $\text{fv}_{\Sigma_{\text{LRA}}}^{\text{Q}}(J)$ . We complete the proof by showing that  $J$  assigns values to all non-variable terms  $t \in G_{\text{Q}}(J)$ . Since  $J$  assigns values to all variables  $x_1, \dots, x_r$  in  $t$  (i.e.,  $\{x_1 \leftarrow \tilde{q}_1, \dots, x_r \leftarrow \tilde{q}_r\} \subseteq J$ ), these assignments dictates a value  $\tilde{q}$  for  $t$ . If  $t \leftarrow \tilde{q}$  is acceptable for  $\mathcal{I}_{\text{LRA}}$  in  $J$ ,  $\mathcal{I}_{\text{LRA}}$  can expand  $J$  deciding  $t \leftarrow \tilde{q}$ , a contradiction. If  $t \leftarrow \tilde{q}$  is not acceptable for  $\mathcal{I}_{\text{LRA}}$  in  $J$ , it means that  $t \leftarrow \tilde{q}$  enables an evaluation step generating  $\bar{L}$  for some  $L \in J$ ; then also  $\{x_1 \leftarrow \tilde{q}_1, \dots, x_r \leftarrow \tilde{q}_r\}$  enables an evaluation inference generating  $\bar{L}$ , and  $\mathcal{I}_{\text{LRA}}$  can expand  $J$ , again a contradiction.  $\square$

The above lemma applies to any plausible LRA-assignment  $J$ , without requiring that  $J$  is sensible: if  $J$  is sensible, the rule for detection of an empty solution space plays no role in the proof of the lemma, whereas it does if  $J$  is not sensible (cf. Case (1) in the proof). It follows that if the CDSAT search plan is sensible, and therefore all generated assignments are sensible, the rule for detection of an empty solution space is unnecessary.

**Theorem 5** *Module  $\mathcal{I}_{\text{LRA}}$  is leading-theory-complete for all leading theories whose models interpret Q as an infinite set.*

**Proof** Let  $J$  be a plausible LRA-assignment that  $\mathcal{I}_{\text{LRA}}$  cannot expand. We show that  $J$  is leading-theory-compatible with LRA sharing  $G(J)$ . Assignment  $J$  gives values to all terms in  $G_{\text{prop}}(J)$  by Part (2) of Lemma 1 and to all terms in  $G_{\text{Q}}(J)$  by Lemma 3 ( $\ddagger$ ). Since  $S_{\text{LRA}} = \{\text{prop}, \text{Q}\}$ ,  $G(J) = G_{\text{prop}}(J) \uplus G_{\text{Q}}(J)$ , and  $\text{fv}_{\Sigma_{\text{LRA}}}(G(J)) = \text{fv}_{\Sigma_{\text{LRA}}}(J)$  by Corollary 1. Let  $\mathcal{T}_1$  be a leading theory, and  $\mathcal{M}_1$  a  $\mathcal{T}_1^+[\mathcal{V}_1]$ -model such that  $\text{fv}_{\Sigma_1}(G(J)) \subseteq \mathcal{V}_1$ ,  $\mathcal{M}_1 \models \mathcal{T}_1$ , and  $|\mathcal{Q}^{\mathcal{M}_1}|$  is infinite. We define an LRA $^+$  $[\mathcal{V}]$ -model  $\mathcal{M}$  with  $\mathcal{V} = \text{fv}_{\Sigma_{\text{LRA}}}(J)$ , and we show that it satisfies Definition 5.  $\mathcal{M}$  interprets Q as  $\mathbb{Q}$ , every symbol in  $\Sigma_{\text{LRA}}^+$  in the standard way (e.g.,  $\tilde{q}$  as  $q$ ), and every Q-sorted  $\Sigma_{\text{LRA}}$ -variable  $x \in \text{fv}_{\Sigma_{\text{LRA}}}(J)$  as  $\tilde{q}$  for  $(x \leftarrow \tilde{q}) \in J$ . For Part (i) of Definition 5, we show that  $\mathcal{M} \models J$ . For all  $(t \leftarrow c) \in J$ , there are three cases:  $t$  is either

a  $\Sigma_{\text{LRA}}$ -variable, or a formula, or a non-variable term of sort  $Q$ . If  $t$  is a  $\Sigma_{\text{LRA}}$ -variable, then  $\mathcal{M}(t) = c^{\mathcal{M}}$  by construction of  $\mathcal{M}$ . Otherwise,  $J$  assigns values to all  $Q$ -sorted subterms of  $t$  by  $(\dagger)$ . If  $t$  is a formula and  $\mathcal{M}(t) \neq c^{\mathcal{M}}$ , then  $\mathcal{I}_{\text{LRA}}$  can expand  $J$  with an evaluation inference deriving  $\bar{t} \leftarrow \bar{c}$ . If  $t$  is a non-variable term of sort  $Q$  and  $\mathcal{M}(t) \neq c^{\mathcal{M}}$ , then  $\mathcal{I}_{\text{LRA}}$  can expand  $J$  with an evaluation inference deriving  $t \not\approx_Q t$ . Both conclusions contradict the hypothesis that  $\mathcal{I}_{\text{LRA}}$  cannot expand  $J$ , and therefore  $\mathcal{M}(t) = c^{\mathcal{M}}$  holds. For Part (ii) of Definition 5,  $Q^{\mathcal{M}}$  is countably infinite. If  $Q^{\mathcal{M}_1}$  is countably infinite, we are done. Otherwise,  $|Q^{\mathcal{M}_1}|$  is some larger infinite cardinality:  $|Q^{\mathcal{M}_1}| > |Q^{\mathcal{M}}|$ . Since  $\Sigma_{\text{LRA}}^+$  is countable, by the Löwenheim-Skolem theorem, there exists another  $\text{LRA}^+[\mathcal{V}]$ -model  $\mathcal{M}'$  such that  $|Q^{\mathcal{M}'}| = |Q^{\mathcal{M}_1}|$  and  $\mathcal{M}'$  agrees with  $\mathcal{M}$  on everything else. For Part (iii) of Definition 5, we observe that  $J \subseteq J_{\mathcal{T}_1}$  by the definition of theory view, since  $J$  is an LRA-assignment and  $\mathcal{T}_1$  has the sorts of LRA, so that LRA-values are also  $\mathcal{T}_1$ -values. Thus,  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$  implies  $\mathcal{M}_1 \models J$ . For all  $t, t' \in G_{\text{prop}}(J)$ ,  $\mathcal{M}_1 \models J$  and  $\mathcal{M} \models J$  suffice for  $\mathcal{M}(t) = \mathcal{M}(t')$  if and only if  $\mathcal{M}_1(t) = \mathcal{M}_1(t')$ . For all  $t, t' \in G_Q(J)$ ,  $J$  assigns a truth value to  $t \simeq_Q t'$ , because if this were not the case,  $J$  could be expanded by an equality inference, since  $J$  gives values to  $t$  and  $t'$  by  $(\dagger)$ . Thus,  $\mathcal{M}_1 \models J$  and  $\mathcal{M} \models J$  imply  $\mathcal{M}(t) = \mathcal{M}(t')$  if and only if  $(t \simeq_Q t') \in J$  if and only if  $\mathcal{M}_1(t) = \mathcal{M}_1(t')$ . □

A module  $\mathcal{I}$  is *unit-constraint complete* [35] for sort  $s$  of its theory  $\mathcal{T}$ , if for all trails  $\Gamma$  and unassigned variables  $x$  of sort  $s$  for which  $\Gamma$  contains a unit constraint, module  $\mathcal{I}$  offers either an acceptable assignment  $x \leftarrow c$  or an inference revealing a conflict. The above results show that  $\mathcal{I}_{\text{LRA}}$  is unit-constraint complete for  $Q$ . In general, unit-constraint completeness is subsumed by the CDSAT completeness requirements on theory modules.

### 4.5 Generic Theories: Stable Infiniteness and Beyond

We consider first a generic theory  $\mathcal{T}$  with signature  $\Sigma = (S, F)$  that can be part of a combination by equality sharing (e.g., [42,48,49]): (i) there exists a decision procedure for the  $\mathcal{T}$ -satisfiability of conjunctions, or, equivalently, sets of  $\mathcal{T}$ -literals; and (ii)  $\mathcal{T}$  is *stably infinite* (every  $\mathcal{T}$ -satisfiable  $\Sigma$ -formula has a  $\mathcal{T}$ -model with countably infinite domains for all sorts in  $S \setminus \{\text{prop}\}$ ). In equality sharing the decision procedures cooperate by exchanging equalities between shared variables toward building an *arrangement*, namely a satisfiable set of sorted equalities and disequalities telling whether any two variables of the same sort are equal (e.g., [9, Sect. 3] for more background). CDSAT handles  $\mathcal{T}$  with a *black-box theory module*  $\mathcal{I}_{\mathcal{T}}^{\text{bb}}$ . The extension  $\mathcal{T}^+$  either is trivial or adds a countably infinite set of values for each sort  $s \in S \setminus \{\text{prop}\}$  and no axioms. Module  $\mathcal{I}_{\mathcal{T}}^{\text{bb}}$  includes the equality inference rules and a *black-box inference rule*

$$l_1 \leftarrow b_1, \dots, l_m \leftarrow b_m \vdash_{\mathcal{T}} \perp,$$

where  $l_1, \dots, l_m$  are  $\Sigma$ -formulae ( $\Sigma$ -atoms as  $\Sigma$  has no connectives). A black-box inference  $J \vdash_{\mathcal{T}} \perp$  applies if the set of literals defined by the Boolean assignment  $J$ , namely  $C_J = \{l \mid (l \leftarrow \text{true}) \in J\} \cup \{\neg l \mid (l \leftarrow \text{false}) \in J\}$ , is found  $\mathcal{T}$ -unsatisfiable by the  $\mathcal{T}$ -satisfiability procedure. If  $\mathcal{T}^+$  is non-trivial, the only rules of  $\mathcal{I}_{\mathcal{T}}^{\text{bb}}$  that may use first-order  $\mathcal{T}$ -assignments are the equality inference rules, and  $\mathcal{T}$ -values act as labels of congruence classes of terms. The local basis only adds  $\top$ : for all sets  $X$  of terms,  $\text{basis}_{\mathcal{T}}(X) = X \cup \{\top\}$ . Indeed, in equality sharing, no new terms introduced by non-trivial inferences are shared.

**Theorem 6** *Module  $\mathcal{I}_{\mathcal{T}}^{\text{bb}}$  is leading-theory-complete for all leading theories whose models interpret all sorts other than  $\text{prop}$  as countably infinite sets.*

**Proof** Let  $J$  be a plausible  $\mathcal{T}$ -assignment that  $\mathcal{T}_T^{\text{bb}}$  cannot expand. We show that  $J$  is leading-theory-compatible with  $\mathcal{T}$  sharing  $G(J)$  (see Definition 5). Let  $\mathcal{T}_1$  be a leading theory satisfying the hypothesis,  $\Sigma_1$  its signature,  $\mathcal{T}_1^+$  its extension, and  $\mathcal{M}_1$  any  $\mathcal{T}_1^+[\mathcal{V}_1]$ -model such that  $\text{fv}_{\Sigma_1}(G(J)) \subseteq \mathcal{V}_1$  and  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$ . We distinguish two cases depending on the choice of  $\mathcal{T}^+$ .

1. Trivial  $\mathcal{T}^+$ : all terms  $l \in G_{\text{prop}}(J)$  including all equalities  $t \simeq_s u$  for  $t, u \in G_s(J)$  of sort  $s \in S \setminus \{\text{prop}\}$  are relevant to  $\mathcal{T}$ , so that  $J$  assigns them values by Part (2) of Lemma 1 ( $\dagger$ ), and  $\text{fv}_{\Sigma}(G(J)) = \text{fv}_{\Sigma}(J)$  by Corollary 1.  $C_J$  is  $\mathcal{T}$ -satisfiable, because otherwise  $\mathcal{T}_T^{\text{bb}}$  could expand  $J$  with a black-box inference. Thus, there exists a  $\mathcal{T}$ -model  $\mathcal{M}'$  of  $C_J$ . Since  $\mathcal{T}^+$  is trivial, it suffices to interpret the Boolean values as themselves to get from  $\mathcal{M}'$  a  $\mathcal{T}^+[\mathcal{V}]$ -model  $\mathcal{M}$ , with  $\mathcal{V} = \text{fv}_{\Sigma}(J)$ , such that  $\mathcal{M} \models J$ , fulfilling Part (i) of Definition 5. For Part (ii), since  $\mathcal{T}$  is stably infinite, we let  $\mathcal{M}$  interpret every sort in  $S \setminus \{\text{prop}\}$  as a countably infinite set, thus agreeing with  $\mathcal{M}_1$ . For Part (iii),  $J \subseteq J_{\mathcal{T}_1}$  by definition of theory view, so that  $\mathcal{M}_1 \models J_{\mathcal{T}_1}$  implies  $\mathcal{M}_1 \models J$ . For all terms  $t, u \in G_{\text{prop}}(J)$ ,  $J$  gives them values by ( $\dagger$ ),  $\mathcal{M}(t) = \mathcal{M}(u)$  if and only if  $\{t \leftarrow b, u \leftarrow b\} \subseteq J$  since  $\mathcal{M} \models J$ , and  $\mathcal{M}_1(t) = \mathcal{M}_1(u)$  if and only if  $\{t \leftarrow b, u \leftarrow b\} \subseteq J$  since  $\mathcal{M}_1 \models J$ , so that  $\mathcal{M}$  and  $\mathcal{M}_1$  agree. For all terms  $t, u \in G_s(J)$  with  $s \in S \setminus \{\text{prop}\}$ ,  $J$  gives a value to  $t \simeq_s u$  by ( $\dagger$ ),  $\mathcal{M}(t) = \mathcal{M}(u)$  if and only if  $(t \simeq_s u) \in J$  since  $\mathcal{M} \models J$ ,  $\mathcal{M}_1(t) = \mathcal{M}_1(u)$  if and only if  $(t \simeq_s u) \in J$  since  $\mathcal{M}_1 \models J$ , so that  $\mathcal{M}$  and  $\mathcal{M}_1$  agree.
2. Non-trivial  $\mathcal{T}^+$ : by the same reasoning in the proof of Theorem 3, assignment  $J$  gives values to all terms in  $G(J)$  ( $\dagger$ ), and  $\text{fv}_{\Sigma}(G(J)) = \text{fv}_{\Sigma}(J)$  by Corollary 1. Also,  $J$  assigns a truth value to  $t \simeq_s u$  for all  $t, u \in G_s(J)$  of sort  $s \in S \setminus \{\text{prop}\}$  ( $\ddagger$ ), because otherwise  $\mathcal{T}$  could expand  $J$  with an equality inference, since  $J$  assigns values to  $t$  and  $u$  by ( $\dagger$ ) and  $(t \simeq_s u) \in \text{basis}_{\mathcal{T}}(J)$  by closedness of the local basis. As in Case (1), the set  $C_J$  has a  $\mathcal{T}$ -model  $\mathcal{M}'$ . We show how to get from  $\mathcal{M}'$  a  $\mathcal{T}^+[\mathcal{V}]$ -model  $\mathcal{M}$ , with  $\mathcal{V} = \text{fv}_{\Sigma}(J)$ , such that  $\mathcal{M} \models J$ . Let  $\mathcal{M}$  interpret the sorts in  $S$ , the symbols in  $F$ , and the  $\Sigma$ -variables as  $\mathcal{M}'$  does. The Boolean values are interpreted as themselves. For a non-Boolean  $\mathcal{T}^+$ -value  $c$  of sort  $s$ , either it is never used in  $J$  or there is some assignment  $(t \leftarrow c) \in J$ . In the first case we let  $\mathcal{M}$  interpret  $c$  arbitrarily in  $s^{\mathcal{M}}$ . In the second case  $t$  appears in equalities in  $J$  by ( $\ddagger$ ), hence  $t$  appears in  $C_J$ , and we define  $c^{\mathcal{M}}$  as  $\mathcal{M}'(t)$ . The interpretation of  $\mathcal{T}^+$ -values is well-defined, because if  $\{t_1 \leftarrow c, t_2 \leftarrow c\} \subseteq J$ , then  $(t_1 \simeq_s t_2) \in J$  by ( $\ddagger$ ) and  $(t_1 \simeq_s t_2) \in C_J$ , so that  $\mathcal{M}'(t_1) = \mathcal{M}'(t_2)$ . By construction,  $\mathcal{M} \models J$ , and the rest of the proof is the same as in Case (1).  $\square$

This theorem shows that the *equality-sharing method is a special case of the CDSAT framework*. Indeed, when the  $\mathcal{T}$ -module cannot expand a  $\mathcal{T}$ -assignment  $J$  (the  $\mathcal{T}$ -view of the trail) it follows that: (1) there exists a  $\mathcal{T}$ -model endorsing  $J$ , and (2)  $J$  determines the truth value of all equalities, hence it defines an arrangement of shared variables. If this is the case for all theories in  $\mathcal{T}_{\infty}$ , an endorsing  $\mathcal{T}_{\infty}$ -model also exists, by the CDSAT completeness theorem [11, Theorem 4]. This remark applies also to *model-based theory combination* [23], which is a way to implement equality sharing and relies on equality sharing for completeness. Theorem 6 still holds if the black-box rule is restricted to apply only to  $\mathcal{T}$ -unsatisfiable cores or *minimal  $\mathcal{T}$ -unsatisfiable assignments*, where it suffices to remove an element to make the assignment  $\mathcal{T}$ -satisfiable.

We describe next how CDSAT also handles a generic *non-stably infinite* theory  $\mathcal{T}$  with signature  $\Sigma = (S, F)$ . Suppose  $\mathcal{T}$  is stably infinite for the sorts in  $S \setminus \{\text{prop}, s_1, \dots, s_k\}$ , whereas all  $\mathcal{T}$ -models interpret sorts  $s_1, \dots, s_k$  as sets of fixed finite cardinalities  $m_1, \dots, m_k$ , respectively. The proof of Theorem 6 can be adapted to prove the following.

**Theorem 7** *Module  $\mathcal{I}_{\mathcal{T}}^{\text{bb}}$  is leading-theory-complete for all leading theories whose models interpret all sorts in  $S \setminus \{\text{prop}, s_1, \dots, s_k\}$  as countably infinite sets and  $s_1, \dots, s_k$  as sets of cardinality  $m_1, \dots, m_k$ , respectively.*

For example,  $\mathcal{T}$  could be a theory of bitvectors of different lengths, where for all  $l$ ,  $1 \leq l \leq k$ ,  $s_l$  is the sort  $bv[l]$  of bitvectors of length  $l$  and  $m_l = 2^l$ . Theorem 7 does not need  $k$  to be finite: for bitvectors,  $l$  could range over all nonzero natural numbers. Thus, the cardinality constraints in  $\mathcal{T}$  affect the choice of the leading theory  $\mathcal{T}_1$ , for which  $S_1 = S_\infty$ . If the leading theory can be picked so that all theory modules involved in the combination are leading-theory complete, the cardinality constraints in  $\mathcal{T}$  are imposed to the other theories sharing  $\{s_1, \dots, s_k\}$  or a subset thereof. More generally, different theories in the union  $\mathcal{T}_\infty$  may pose cardinality requirements on a shared sort  $s$ , and the leading theory  $\mathcal{T}_1$  acts as an aggregator of such requirements (see [11, Examples 9 and 10]). Once chosen, the leading theory  $\mathcal{T}_1$  needs a theory module  $\mathcal{I}_1$  that can be used in CDSAT and that enforces the cardinality constraints.

We illustrate this point for an *at-most- $m$  cardinality constraint* on sort  $s$ , given an integer  $m > 0$ . The constraint can be expressed by the sentence  $\forall x_0, \dots, \forall x_m. \bigvee_{0 \leq i \neq k \leq m} x_i \simeq_s x_k$ , for  $x_0, \dots, x_m$  distinct variables of sort  $s$ , which could be an axiom or a theorem of one of the non-leading theories in  $\mathcal{T}_\infty$ , or an axiom of the leading theory  $\mathcal{T}_1$ , resulting from aggregating cardinality constraints from non-leading theories in  $\mathcal{T}_\infty$ . For instance, if  $\mathcal{T}_2$  entails the at-most- $m_1$  cardinality constraint on  $s$  and  $\mathcal{T}_3$  entails the at-most- $m_2$  cardinality constraint on  $s$ , the leading theory  $\mathcal{T}_1$  is picked so that its models satisfy the at-most-min( $m_1, m_2$ ) cardinality constraint. Then, theory module  $\mathcal{I}_1$  for  $\mathcal{T}_1$  includes the *at-most- $m$  inference rule*:

$$\bigwedge_{0 \leq i \neq k \leq m} u_i \not\simeq_s u_k \vdash_{\mathcal{T}_1} \perp$$

where  $u_0, \dots, u_m$  are any  $m + 1$  distinct terms of sort  $s$ . If  $\mathcal{T}_1^+$  is non-trivial with values for sort  $s$ , the at-most- $m$  inference rule can be abbreviated as  $u_0 \leftarrow c_0, \dots, u_m \leftarrow c_m \vdash_{\mathcal{T}_1} \perp$ , for  $c_0, \dots, c_m$  any distinct  $m+1$   $\mathcal{T}_1$ -values of sort  $s$ . If both  $\mathcal{T}$  and the leading theory  $\mathcal{T}_1$  have non-trivial extensions,  $\mathcal{T}^+$  and  $\mathcal{T}_1^+$  use different sets of constant symbols to name  $s$ -sorted elements, and the construction of a  $\mathcal{T}_\infty$ -model for an assignment  $J$  that cannot be expanded establishes a bijection between the  $s$ -sorted  $\mathcal{T}$ -values that appear in  $J$  and the  $s$ -sorted  $\mathcal{T}_1$ -values that appear in  $J$  (see [11, Sect. 9.3, Theorem 4]). A module with an at-most- $m$  inference rule satisfies a lemma that complements Lemma 1.

**Lemma 4** *If a  $\mathcal{T}$ -module  $\mathcal{I}$  with the at-most- $m$  inference rule for sort  $s$  cannot expand a plausible  $\mathcal{T}$ -assignment  $J$ , the relation  $\simeq_s^J$  is an equivalence with at most  $m$  equivalence classes.*

**Proof** By definition of  $\simeq_s^J$  (see the text in Sect. 4 preceding Lemma 1), for all  $t_1, t_2 \in G_s(J)$ ,  $t_1 \simeq_s^J t_2$  if and only if  $(t_1 \simeq_s t_2) \in J$ . By Part (1) of Lemma 1 the relation  $\simeq_s^J$  is an equivalence. If  $\simeq_s^J$  had  $m+1$  equivalence classes,  $J$  would contain an instance of the premises of the at-most- $m$  inference rule for sort  $s$ , and  $\mathcal{I}$  could expand  $J$ , a contradiction.  $\square$

This lemma suffices to obtain the following theorem that says how to build a leading theory and its module to enforce the at-most cardinality constraint coming from the theories in  $\mathcal{T}_\infty$ . Given a theory  $\mathcal{T}_1$ , let  $\mathcal{T}_1^{s \leq m}$  be  $\mathcal{T}_1$  plus the at-most- $m$  cardinality constraint on sort  $s$  as additional axiom. Given a theory module  $\mathcal{I}$ , let  $\mathcal{I}^{s \leq m}$  be  $\mathcal{I}$  plus the at-most- $m$  inference rule on sort  $s$ .

**Theorem 8** *If  $\mathcal{I}_1$  is sound and complete for theory  $\mathcal{T}_1$ , then  $\mathcal{I}_1^{s \leq m}$  is sound and complete for theory  $\mathcal{T}_1^{s \leq m}$ .*

Once the enforcement of cardinality constraints is handled by the leading theory module, it is not necessary to handle them in other modules.

**Theorem 9** *Given a theory  $\mathcal{T}$  with signature  $\Sigma = (S, F)$  and a leading theory  $\mathcal{T}_1$  that entails the at-most- $m$  constraint on sort  $s \in S$ , a  $\mathcal{T}$ -module  $\mathcal{I}$  is leading-theory complete if and only if  $\mathcal{I}^{s \leq m}$  is leading-theory complete.*

**Proof** The  $(\Rightarrow)$  direction holds by Lemma 2. The  $(\Leftarrow)$  direction holds because whenever the at-most- $m$  inference rule of  $\mathcal{I}^{s \leq m}$  can be applied to expand an assignment  $J$ , there can be no  $\mathcal{T}_1$ -model endorsing  $J_{\mathcal{T}_1}$  so that leading-theory compatibility is vacuously true.  $\square$

In summary, the completeness of a leading theory module with the appropriate at-most rules ensures that cardinality constraints on shared sorts are satisfied; and all theories sharing those sorts concur on their cardinalities by leading-theory-completeness of their modules.

### 5 Global Basis Construction

Termination of CDSAT requires the global basis  $\mathcal{B}$  to be *finite* and *closed*, and completeness requires it to be *stable* (see Theorem 1). The meaning of stability of  $\mathcal{B}$  (for all  $k, 1 \leq k \leq n$ ,  $\text{basis}_k(\mathcal{B}) \subseteq \mathcal{B}$ ) is that the global basis “contains” the local bases  $\text{basis}_1, \dots, \text{basis}_n$  associated to the theory modules  $\mathcal{I}_1, \dots, \mathcal{I}_n$  for theories  $\mathcal{T}_1, \dots, \mathcal{T}_n$ : for all sets  $X$  of terms, if  $X \subseteq \mathcal{B}$  then for all  $k, 1 \leq k \leq n$ ,  $\text{basis}_k(X) \subseteq \text{basis}_k(\mathcal{B})$  by monotonicity (see Definition 2) and  $\text{basis}_k(X) \subseteq \mathcal{B}$  by stability. Thus, for all input assignments  $H$ , if  $H$  is in  $\mathcal{B}$ , or, equivalently,  $G(H) \subseteq \mathcal{B}$ , no  $\mathcal{I}_k$ -inference can take us outside of  $\mathcal{B}$ . In this section we show how to build a stable global basis from local ones.

The existence of a finite stable global basis does not necessarily follow from that of local bases. Given input assignment  $H$  and  $X_0 = G(H)$ , a module  $\mathcal{I}_k$  may introduce a term  $u_0$  in  $Y_0 = \text{basis}_k(X_0)$ , which prompts  $\mathcal{I}_j$  to introduce a term  $t_1$  in  $X_1 = \text{basis}_j(\text{basis}_k(X_0))$ , which in turns prompts  $\mathcal{I}_k$  to introduce a term  $u_1$  in  $Y_1 = \text{basis}_k(\text{basis}_j(\text{basis}_k(X_0)))$ , and so on. In other words, even if all these sets are finite,  $\bigcup_{m \geq 0} X_m$  may be infinite, where  $X_{m+1} = \text{basis}_j(\text{basis}_k(X_m))$ . The aim is to find sufficient conditions on local bases to avoid such cyclic behavior. Since the problem arises from a cyclic alternation, the point is whether it is possible to *permute* local bases, relating  $\text{basis}_j(\text{basis}_k(X))$  and  $\text{basis}_k(\text{basis}_j(X))$ . To this end, we introduce the following notions.

**Definition 7 (Production and consumption of a sort)** Let  $\text{basis}$  be a basis for theory  $\mathcal{T}$  with signature  $\Sigma = (S, F)$ . For all sorts  $s \in S$ ,  $\text{basis}$  *produces sort  $s$*  if for some closed set of terms  $X$  and term  $t$  of sort  $s$ ,  $t \in \text{basis}(X) \setminus X$ ;  $\text{basis}$  *consumes sort  $s$*  if for some closed set of terms  $X$  and term  $t$  of sort  $s$ ,  $\text{basis}(X \uplus \{t\}) \not\subseteq \downarrow(\text{basis}(X) \uplus \{t\})$ , where  $t$  is either a  $\Sigma$ -variable or an equality whose strict subterms are in  $X$ .

In plain words,  $\text{basis}$  produces sort  $s$  if its application to a closed set  $X$  yields some term  $t$  of sort  $s$  which is not in  $X$  and does not arise from the closure of  $X$ , since  $X$  is already closed;  $\text{basis}$  consumes sort  $s$  if its application to  $X \uplus \{t\}$ , where  $t$  is a term of sort  $s$ , yields some term  $u$  which is not in  $\downarrow(\text{basis}(X) \uplus \{t\})$ , where  $\text{basis}$  is applied to  $X$  only. The restrictions on what term  $t$  can be depend on what suffices for forthcoming Lemma 5.

**Example 9** Most local bases produce  $\text{prop}$ , as they add  $\top$ ;  $\text{basis}_{\text{Bool}}$  produces and consumes  $\text{prop}$ , as it forms clauses for lemma learning;  $\text{basis}_{\text{EUF}}$  only produces  $\text{prop}$  by adding equalities and does not consume any sort. For  $\text{Arr}$ ,  $\text{basis}_{\text{Arr}}$  produces all sorts in  $\Sigma_{\text{Arr}}$  and consumes

all array sorts: given array terms  $t$  and  $u$  of sort  $I \Rightarrow V$ ,  $\text{basis}_{\text{Arr}}$  consumes  $I \Rightarrow V$  and produces sorts  $I$  and  $V$ , by introducing terms  $\text{diff}(t, u)$ ,  $t[\text{diff}(t, u)]$ , and  $u[\text{diff}(t, u)]$ ; it produces also array sorts, because arrays can be values or indices, as there can be arrays of arrays and array-indexed arrays. For LRA,  $\text{basis}_{\text{LRA}}$  produces sorts  $\text{prop}$  and  $Q$  and only consumes  $Q$ , where  $Q$  is produced when polynomials are reduced to normal form. For example, the FM-resolution  $y-x < 0, 3 \cdot x < 5 \vdash_{\text{LRA}} y < \frac{5}{3}$  produces  $\text{prop}$  by introducing  $y < \frac{5}{3}$  and  $Q$  by introducing  $\frac{5}{3}$ . The bases of  $\mathcal{T}_{\text{bb}}$  and  $\mathcal{T}_{\text{bb}}^m$  only produce  $\text{prop}$  and do not consume any sort.

The next move is to define a *theory ordering* on the theories that captures producer-consumer dependencies between their local bases: for all  $k, j$  such that  $1 \leq k \neq j \leq n$ , let  $\mathcal{T}_k < \mathcal{T}_j$  if there exists a sort  $s$  that  $\text{basis}_k$  produces and  $\text{basis}_j$  consumes. By the contrapositive, if  $\mathcal{T}_k \not< \mathcal{T}_j$ , then  $\text{basis}_j$  is independent of  $\text{basis}_k$ , hence  $\text{basis}_j(\text{basis}_k(X)) \subseteq \text{basis}_k(\text{basis}_j(X))$  for all  $X$ . Intuitively, if  $<$  is *acyclic*, the cyclic behavior described above cannot happen. Formally, if  $<$  is acyclic, the listing of the theories and a basis for  $\mathcal{T}_\infty$  can be defined accordingly: for all  $k$  and  $j, 1 \leq k < j \leq n$ , if  $\mathcal{T}_k < \mathcal{T}_j$  then  $k < j$ , and for all sets  $X$  of terms,  $\text{basis}_\infty(X) = \text{basis}_n(\dots \text{basis}_1(X))$ . The next lemma shows that under these hypotheses local bases can be permuted.

**Lemma 5** *If  $\mathcal{T}_1, \dots, \mathcal{T}_n$  are disjoint theories with an acyclic ordering  $<$ , then for all  $k$  and  $j, 1 \leq k < j \leq n$ , and for all finite closed sets  $X$  of terms: (1) For all  $\trianglelefteq$ -closed sets  $Y$  of terms such that  $X \subseteq Y \subseteq \text{basis}_j(X)$ , it holds that  $\text{basis}_k(Y) \subseteq \Downarrow(\text{basis}_k(X) \cup Y)$ ; and (2)  $\text{basis}_k(\text{basis}_j(X)) \subseteq \text{basis}_j(\text{basis}_k(X))$ .*

**Proof** First,  $k < j$  implies  $j \not< k$ , hence  $\mathcal{T}_j \not< \mathcal{T}_k$ , so that no sort produced by  $\text{basis}_j$  is consumed by  $\text{basis}_k$ .

1. The proof is by induction on the (finite) cardinality of  $Y \setminus X$ , denoted  $|Y \setminus X|$ . For the base case, if  $|Y \setminus X| = 0$  (i.e.,  $Y = X$ ), the claim is trivially true. For the induction hypothesis, let the claim be true for any such  $Y$  with  $|Y \setminus X| = q \geq 0$ . For the induction step, let  $|Y \setminus X| = q + 1$  and  $t$  be a term of largest size (symbol count) in  $Y \setminus X$ . By hypothesis,  $t$  is in  $\text{basis}_j(X)$ . Since the theories are disjoint,  $t$  is either  $\Sigma_j$ -foreign, or  $\Sigma_k$ -foreign, or an equality. By the last property of a basis in Definition 2,  $t$  being in  $\mathcal{V}_\infty$  is the only way that it can be  $\Sigma_j$ -foreign, in which case it is also  $\Sigma_k$ -foreign. Therefore, it is either  $\Sigma_k$ -foreign or an equality. By  $\trianglelefteq$ -closedness of  $Y$ , all strict subterms of  $t$  are in  $Y \setminus \{t\}$ , hence in  $\Downarrow(Y \setminus \{t\})$ . Since  $t \in (\text{basis}_j(X) \setminus X)$ , the sort  $s$  of  $t$  is produced by  $\text{basis}_j$ . Last,  $\text{basis}_k$  does not consume  $s$ , because  $\text{basis}_k$  does not depend on  $\text{basis}_j$ , as  $k < j$  by hypothesis. By Definition 7 applied to  $\text{basis}_k$  and the closed set  $\Downarrow(Y \setminus \{t\})$ , we get

$$\text{basis}_k(\Downarrow(Y \setminus \{t\}) \cup \{t\}) \subseteq \Downarrow(\text{basis}_k(\Downarrow(Y \setminus \{t\})) \cup \{t\}) \quad (\dagger).$$

Next, we observe that  $X \subseteq (Y \setminus \{t\})$ , since  $X \subseteq Y$  and  $t \in Y \setminus X$ , and  $Y \setminus \{t\}$  is  $\trianglelefteq$ -closed: indeed, if it were  $t \triangleleft u$  for some term  $u \in Y$ , then either  $u \in X$ , in which case  $t \in X$ , because  $X$  is closed, or  $u \in Y \setminus X$ , in which case  $t$  would not be a term of greatest size in  $Y \setminus X$ . Therefore, we can apply the induction hypothesis to  $Y \setminus \{t\}$  and get

$$\text{basis}_k(Y \setminus \{t\}) \subseteq \Downarrow(\text{basis}_k(X) \cup (Y \setminus \{t\})) \quad (\ddagger).$$

Then the claim is established as follows:

$$\begin{aligned}
 \text{basis}_k(Y) &= \text{basis}_k((Y \setminus \{t\}) \cup \{t\}) \\
 &\subseteq \text{basis}_k(\Downarrow(Y \setminus \{t\}) \cup \{t\}) && \text{by monotonicity of } \text{basis}_k \\
 &\subseteq \Downarrow(\text{basis}_k(\Downarrow(Y \setminus \{t\})) \cup \{t\}) && \text{by } (\dagger) \\
 &= \Downarrow(\text{basis}_k(Y \setminus \{t\}) \cup \{t\}) && \text{by closedness of } \text{basis}_k \\
 &\subseteq \Downarrow(\Downarrow(\text{basis}_k(X) \cup (Y \setminus \{t\})) \cup \{t\}) && \text{by } (\ddagger) \\
 &\subseteq \Downarrow\Downarrow(\text{basis}_k(X) \cup (Y \setminus \{t\}) \cup \{t\}) \\
 &= \Downarrow(\text{basis}_k(X) \cup Y) && \text{by idempotence of } \Downarrow.
 \end{aligned}$$

2. The second claim is derived as follows:

$$\begin{aligned}
 \text{basis}_k(\text{basis}_j(X)) &\subseteq \Downarrow(\text{basis}_k(X) \cup \text{basis}_j(X)) && \text{by Claim 1} \\
 &\subseteq \Downarrow(\text{basis}_j(\text{basis}_k(X)) \cup \text{basis}_j(X)) && \text{by extensiveness} \\
 &= \Downarrow(\text{basis}_j(\text{basis}_k(X))) && \text{as } X \subseteq \text{basis}_k(X) \\
 &\quad \text{and by monotonicity of } \text{basis}_j \\
 &= \text{basis}_j(\text{basis}_k(X)) && \text{by closedness.}
 \end{aligned}$$

□

Next, we use Lemma 5 to show that  $\text{basis}_\infty(X)$  is stable.

**Lemma 6** *If  $T_1, \dots, T_n$  are disjoint theories with an acyclic ordering that defines  $\text{basis}_\infty$ , then for all  $k, 1 \leq k \leq n$ ,  $\text{basis}_k(\text{basis}_\infty(X)) = \text{basis}_\infty(X)$  for all finite sets  $X$  of terms.*

**Proof** We prove a more general property, namely that  $\forall j, 1 \leq k \leq j \leq n$ , we have  $\text{basis}_k(\text{basis}_j(\dots \text{basis}_1(X))) = \text{basis}_j(\dots \text{basis}_1(X))$ . The  $\supseteq$ -direction holds by extensiveness of  $\text{basis}_k$ . For the  $\subseteq$ -direction, the proof is by induction on  $j$ . For the base case, if  $j=k$ , the claim holds by idempotence. For the induction hypothesis, let the claim be true for  $j$ . For the induction step, we prove the claim for  $j+1$ . Let  $Z$  stand for  $\text{basis}_j(\dots \text{basis}_1(X))$ . Since  $k < j+1$  and  $Z$  is closed by closedness of the bases,  $\text{basis}_k(\text{basis}_{j+1}(Z)) \subseteq \text{basis}_{j+1}(\text{basis}_k(Z))$  holds by Lemma 5. Then,  $\text{basis}_k(Z) \subseteq Z$  holds by induction hypothesis, and  $\text{basis}_{j+1}(\text{basis}_k(Z)) \subseteq \text{basis}_{j+1}(Z)$  follows by monotonicity of  $\text{basis}_{j+1}$ . □

**Theorem 10** *Given disjoint theories  $T_1, \dots, T_n$  with modules  $\mathcal{I}_1, \dots, \mathcal{I}_n$  and local bases  $\text{basis}_1, \dots, \text{basis}_n$  such that the theory ordering is acyclic, if  $\text{basis}_\infty$  is defined based on the theory ordering, then for all input assignments  $H$ , the set  $\mathcal{B} = \text{basis}_\infty(G(H))$  is a finite stable global basis.*

**Proof** The function  $\text{basis}_\infty$  is a basis for the union theory  $\mathcal{T}_\infty$  according to Definition 2, as it inherits the properties of local bases. Thus,  $\mathcal{B}$  is finite, as  $G(H)$  is finite, and it is stable by Lemma 6. □

For example,  $\text{basis}_{\text{Bool}}(\text{basis}_{\mathcal{I}}(\text{basis}_{\text{EUF}}(\text{basis}_{\text{LRA}}(\text{basis}_{\text{Arr}}(G)(H))))$ , where  $\mathcal{I}$  is a black-box module for a generic theory  $\mathcal{T}$ , is a global basis for the union of theories Bool,  $\mathcal{T}$ , LRA, EUF, and Arr, given an input assignment  $H$ . The next section opens the part of the article devoted to *proof generation* in CDSAT.

## 6 Proof Reconstruction: Proof-Carrying CDSAT

When a derivation terminates detecting unsatisfiability, it is desirable to return a proof. *Proof reconstruction* is the activity of extracting a proof from the final state of the derivation, provided that the final state contains enough information. In this section we instrument CDSAT for proof reconstruction.

$$\begin{array}{l}
a^1(t_1 \leftarrow c), a^2(t_2 \leftarrow c) \vdash \text{eq}(a_1, a_2): t_1 \simeq t_2 \quad \text{if } c \text{ is a } \mathcal{T}\text{-value of sort } s \\
a^1(t_1 \leftarrow c_1), a^2(t_2 \leftarrow c_2) \vdash \text{neq}(a_1, a_2): t_1 \not\simeq t_2 \quad \text{if } c_1 \text{ and } c_2 \text{ are distinct } \mathcal{T}\text{-values of sort } s \\
\vdash \text{refl}: t \simeq t \quad (\text{reflexivity}) \\
a(t_1 \simeq t_2) \vdash \text{sym}(a): t_2 \simeq t_1 \quad (\text{symmetry}) \\
a^1(t_1 \simeq t_2), a^2(t_2 \simeq t_3) \vdash \text{trans}(a_1, a_2): t_1 \simeq t_3 \quad (\text{transitivity})
\end{array}$$

**Fig. 5** Annotated equality inference rules, where  $t_1$ ,  $t_2$ , and  $t_3$  are terms of sort  $s$

## 6.1 Theory Proofs

Because CDSAT combines theory modules, proof reconstruction in CDSAT requires that all theory modules produce proofs. Therefore, we assume that each theory module is equipped with a *proof annotation system* that *annotates* theory inferences with *theory proofs*:

$$J \vdash_k j_k : L$$

states that module  $\mathcal{I}_k$  infers  $L$  from  $J$  with theory proof  $j_k$ . A theory proof from  $\mathcal{I}_k$  is called  $\mathcal{I}_k$ -*proof*. Theory proofs, hence CDSAT proofs, may refer to singleton assignments by means of *identifiers*. A  $\mathcal{T}$ -*assignment with identifiers* is a set of triples  $a(t \leftarrow c)$ , where  $a$  is the *identifier* of  $t \leftarrow c$ . From now on *all assignments are assignments with identifiers*, the trail contains a  $\mathcal{T}_\infty$ -assignment with identifiers, and the subset relation between assignments take identifiers into account. For example,  $\mathcal{I}_{\text{Bool}}$ -inference (1) from Sect. 3.2 can be annotated with a theory proof denoted  $\text{UP}(a, \{a_1, \dots, a_n\})$ , as follows:

$$\{a K\} \uplus H' \vdash_{\text{Bool}} \text{UP}(a, \{a_1, \dots, a_n\}) : \bar{L}$$

where UP stands for unit propagation, and  $a_1, \dots, a_n$  are the identifiers of the assignments in  $H'$ . Annotated  $\mathcal{I}_{\text{LRA}}$ -inferences include instances of Fourier-Motzkin resolution and of the evaluation rule:

$$\begin{array}{l}
a^1(e_1 \leq t), a^2(t \leq e_2) \vdash_{\text{LRA}} \text{FM}(a_1, a_2): (e_1 \leq e_2), \\
a^1(x_1 \leftarrow q_1), \dots, a^m(x_m \leftarrow q_m) \vdash_{\text{LRA}} \text{eval}(\{a_1, \dots, a_m\}): l \leftarrow b.
\end{array}$$

Equality inferences are annotated with theory proofs as shown in Fig. 5. Assuming  $J$  is  $a^1(t_1 \simeq u_1), \dots, a^m(t_m \simeq u_m)$ , an instance of annotated  $\mathcal{I}_{\text{EUF}}$ -inference is:

$$J \vdash_{\text{EUF}} \text{Cong}(a_1, \dots, a_m): f(t_1, \dots, t_m) \simeq f(u_1, \dots, u_m)$$

where Cong stands for congruence.

If an assignment appears on the trail, its identifier in any theory proof is the same as its identifier on the trail: for a Deduce transition supported by a theory inference  $J \vdash_k j_k : L$ , the assignments in  $J$  appear on the trail  $\Gamma$ , and their identifiers in  $j_k$  are the same as in  $\Gamma$ . Since identifiers are mere names, theory proof annotations are stable under their permutations: any permutation  $\pi$  of identifiers transforms a theory proof  $j_k$  into a theory proof  $\pi(j_k)$ , such that, if  $a^1(t_1 \leftarrow c_1), \dots, a^m(t_m \leftarrow c_m) \vdash_k j_k : L$  then

$$\pi^{(a_1)}(t_1 \leftarrow c_1), \dots, \pi^{(a_m)}(t_m \leftarrow c_m) \vdash_k \pi(j_k) : L.$$

For example, if  ${}^1(x \leftarrow c_1), {}^4(y \leftarrow c_1) \vdash_k \text{Cong}(1, 4): f(x) \simeq f(y)$ , with  $\pi(1) = 4$  and  $\pi(4) = 1$ , it is also  ${}^4(x \leftarrow c_1), {}^1(y \leftarrow c_1) \vdash_k \text{Cong}(4, 1): f(x) \simeq f(y)$ . The assumption that theory modules have proof annotation systems is not a restriction, as the proof annotation system can be a trivial one that uses a dummy theory proof for all theory inferences. The resulting theory proofs convey no information, which is acceptable if they are not required to offer more.

$$\begin{array}{c}
 \frac{A \text{ is initial}}{\emptyset \vdash \text{in}(A): A} \quad \frac{J \vdash_k j_k: L}{J \vdash j_k: L} \quad \frac{E \uplus H \vdash c: \perp}{E \vdash \text{lem}(H.c): L} \quad L \text{ is a clausal form of } H \\
 \\
 \frac{J \vdash j_k: L}{J \cup \{\alpha \bar{L}\} \vdash \text{cfl}(j_k, a): \perp} \quad \frac{H \vdash j: A \quad E \uplus \{\alpha A\} \vdash c: \perp}{E \cup H \vdash \text{res}(j, \alpha A.c): \perp}
 \end{array}$$

Fig. 6 Proof system for the CDSAT proof terms

### 6.2 Proof Terms, Proof System, and Invariants for CDSAT

In order to enable CDSAT to compose theory proofs into CDSAT proofs, we will equip the CDSAT transition system with the capability of building *proof terms*. These proof terms keep track of *soundness invariants* that ensure that transitions do not change the problem, so that invariant-preserving transition rules are sound. The *CDSAT soundness invariants* are:

1. For all justified assignments  $H \vdash A$  on the trail,  $H_0 \cup H \models A$ , and
2. For all conflict states  $\langle \Gamma; E \rangle$ ,  $H_0 \cup E \models \perp$ ,

where  $H_0$  is the input, or initial, assignment. A proof term is either a *deduction proof term* recording why a justified assignment is on the trail to enforce (1), or a *conflict proof term* recording why a conflict is a conflict to enforce (2). The two kinds of proof terms are defined mutually recursively as follows.

**Definition 8** (*CDSAT proof terms*) A *CDSAT proof term* is

- Either a *deduction proof term*  $j ::= \text{in}(A) \mid j_k \mid \text{lem}(H.c)$ ,
- Or a *conflict proof term*  $c ::= \text{cfl}(j_k, a) \mid \text{res}(j, \alpha A.c)$ , where:

(i) *in*, *lem*, *cfl*, and *res*, abbreviating *initial*, *lemma*, *conflict*, and *resolve*, respectively, are the CDSAT *proof constructors*; (ii)  $A$  is a singleton assignment,  $j_k$  is a  $\mathcal{T}_k$ -proof for some  $k$ ,  $1 \leq k \leq n$ ,  $H$  is a Boolean assignment with identifiers,  $a$  is the identifier of a singleton Boolean conflicting assignment in  $\text{cfl}(j_k, a)$ , and the identifier of  $A$  in  $\text{res}(j, \alpha A.c)$ ; and (iii) the dot notation means that  $\text{res}(j, \alpha A.c)$  binds  $a$  in  $c$  and  $\text{lem}(H.c)$  binds the identifiers of  $H$  in  $c$ .

The CDSAT proof terms come with the *CDSAT proof system* in Fig. 6. Its first three rules establish the derivability of judgments of the form  $H \vdash j: A$ . Proof term  $\text{in}(A)$  witnesses the fact that an initial assignment  $A$  holds. The second rule *coerces* a theory proof  $j_k$  into a CDSAT deduction proof term. The third rule says that, whenever there is a conflict including a Boolean assignment  $H$ , a clausal form of  $H$  is a lemma entailed by the rest of the conflict. The proof term  $\text{lem}(H.c)$  carries  $H$  to record which part of the conflict became a lemma. If identifiers of assignments in  $H$  occur in  $c$ , such occurrences are bound in  $\text{lem}(H.c)$ . The last two rules establish the derivability of judgments of the form  $E \vdash c: \perp$ . Proof term  $\text{cfl}(j_k, a)$  witnesses the conflict between the conclusion  $L$  of a theory inference, whose theory proof is coerced into a CDSAT deduction proof term, and its flip  $\bar{L}$  (with identifier  $a$ ). Proof term  $\text{res}(j, \alpha A.c)$  is the only construct that combines two subproofs, connecting the conclusion  $A$  of the left premise with the hypothesis  $\alpha A$  of the right premise: the proof of  $A$  from  $H$  is plugged as a subproof in the proof of  $\perp$  from  $E \uplus \{\alpha A\}$  to get a proof of  $\perp$  from  $E \cup H$ . Any occurrence of  $a$  in  $c$  is bound in  $\text{res}(j, \alpha A.c)$ . The following theorem connects provability in the CDSAT proof system with endorsement, showing the *soundness* of the CDSAT proof system.

TRAIL RULES		
In the trail rules, let $1 \leq k \leq n$ .		
Decide	$\Gamma \longrightarrow \Gamma, ?A$	if $A$ is an acceptable $\mathcal{T}_k$ -assignment for $\mathcal{I}_k$ in $\Gamma_{\mathcal{T}_k}$
The next three rules share the conditions: $J \subseteq \Gamma$ , $(J \vdash_k j_k : L)$ , and $L \notin \Gamma$ .		
Deduce	$\Gamma \longrightarrow \Gamma, \overline{J} \vdash_{j_k} : L$	if $\overline{L} \notin \Gamma$ and $L$ is in $\mathcal{B}$
Fail	$\Gamma \longrightarrow \text{unsat}(c)$	if ${}^a\overline{L} \in \Gamma$ and $\langle \Gamma; J \cup \{{}^a\overline{L}\}; \text{cfl}(j_k, a) \rangle \Longrightarrow^* \langle \Gamma; \emptyset; c \rangle$
ConflictSolve	$\Gamma \longrightarrow \Gamma'$	if ${}^a\overline{L} \in \Gamma$ and $\langle \Gamma; J \cup \{{}^a\overline{L}\}; \text{cfl}(j_k, a) \rangle \Longrightarrow^* \Gamma'$
CONFLICT STATE RULES		
UndoClear	$\langle \Gamma; E \uplus \{{}^aA\}; c \rangle \Longrightarrow \Gamma^{\leq m-1}$	if $A$ is a first-order decision such that $\text{level}_\Gamma(A) = m > \text{level}_\Gamma(E)$
In the next two rules $A'$ is a first-order decision.		
Resolve	$\langle \Gamma; E \uplus \{{}^aA\}; c \rangle \Longrightarrow \langle \Gamma; E \cup H; \text{res}(j, {}^aA.c) \rangle$	if $(\overline{H} \vdash_{j_k} : A) \in \Gamma$ and for no $A' \in H$ $\text{level}_\Gamma(A') = \text{level}_\Gamma(E \uplus \{A\})$
UndoDecide	$\langle \Gamma; E \uplus \{{}^aL\}; c \rangle \Longrightarrow \Gamma^{\leq m-1}, ?\overline{L}$	if $(\overline{H} \vdash_{j_k} : L) \in \Gamma$ and for an $A' \in H$ $m = \text{level}_\Gamma(E) = \text{level}_\Gamma(L) = \text{level}_\Gamma(A')$
LearnBackjump	$\langle \Gamma; E \uplus H; c \rangle \Longrightarrow \Gamma^{\leq m}, \overline{E} \vdash_{\text{lem}(H.c)} : L$	if $L$ is a clausal form of $H$ , $L$ is in $\mathcal{B}$ , $L \notin \Gamma$ , $\overline{L} \notin \Gamma$ , and $\text{level}_\Gamma(E) \leq m < \text{level}_\Gamma(H)$

Fig. 7 The proof-carrying CDSAT transition system

**Theorem 11** *If  $H \vdash j : A$ , then  $H_0 \cup H \models A$ ; if  $E \vdash c : \perp$ , then  $H_0 \cup E \models \perp$ .*

**Proof** The proof is by structural induction. The base case covers in and coercion: if  $H \vdash j : A$  has the form  $\emptyset \vdash \text{in}(A) : A$ , then  $A$  is initial, which means that  $A \in H_0$  and  $H_0 \models A$ ; if  $H \vdash j : A$  has the form  $J \vdash j_k : L$ , then  $J \vdash_k j_k : L$  and by soundness of theory inferences  $J \models L$ , hence  $H_0 \cup J \models L$ . The induction step covers lem, cfl, and res. If  $H \vdash j : A$  has the form  $E \vdash \text{lem}(H.c) : L$ , then  $E \uplus H \vdash c : \perp$ , by induction hypothesis  $H_0 \cup E \uplus H \models \perp$ , hence  $H_0 \cup E \models L$  as  $L$  is a clausal form of  $H$ . If  $E \vdash c : \perp$  has the form  $J \cup \{{}^a\overline{L}\} \vdash \text{cfl}(j_k, a) : \perp$ , then  $J \vdash j_k : L$ , by induction hypothesis  $H_0 \cup J \models L$ , hence  $H_0 \cup J \cup \{{}^a\overline{L}\} \models \perp$ . If  $E \vdash c : \perp$  has the form  $E \cup H \vdash \text{res}(j, {}^aA.c) : \perp$ , then  $H \vdash j : A$  and  $E \uplus \{{}^aA\} \vdash c : \perp$ ; by induction hypothesis  $H_0 \cup H \models A$  and  $H_0 \cup E \uplus \{A\} \models \perp$ , hence  $H_0 \cup E \cup H \models \perp$ .  $\square$

### 6.3 The Proof-Carrying CDSAT Transition System

In the *proof-carrying CDSAT transition system* (see Fig. 7), justified assignments are decorated with deduction proof terms, and conflict states are triples of the form  $\langle \Gamma; E; c \rangle$ , where  $c$  is a conflict proof term. A justified assignment  $\overline{H} \vdash_{j_k} : A$  carries a deduction proof term  $j$  such that  $H \vdash j : A$ . Initial assignments have the form  $\emptyset \vdash \text{in}(A) : A$  where the deduction proof term presents the initial assignment as a premise of the proof.

Comparing Figs. 1 and 7, Decide is unchanged as a decision does not carry a proof term; Deduce is modified as the supporting theory inference  $J \vdash_k j_k : L$  is annotated with a theory proof that the added justified assignment  $\overline{J} \vdash_{j_k} : L$  carries with itself. In Fig. 1 the choice between Fail and ConflictSolve depends on the level of the conflict, whereas in Fig. 7 it

depends on the outcome of the conflict resolution phase, because proof-carrying CDSAT fires Fail and returns *unsat*, if it can return a proof of unsatisfiability. If the outcome of the conflict resolution phase is a trail, the conflict was solved and ConflictSolve applies; if it is a state  $\langle \Gamma; \emptyset; c \rangle$ , the system is still in conflict state, but there is no conflict to solve. The system concludes that the input assignment is  $\mathcal{T}_\infty^+$ -unsatisfiable and that proof term  $c$  encodes the discovered proof of unsatisfiability: Fail applies and the derivation terminates in state *unsat*( $c$ ) returning proof term  $c$ . It is simple to show that the conflict state rules of proof-carrying CDSAT reduce a conflict state  $\langle \Gamma; E; c_1 \rangle$ , where  $E \neq \emptyset$ , to one of the form  $\langle \Gamma; \emptyset; c \rangle$  if and only if  $\text{level}_\Gamma(E) = 0$ ; and that they solve the conflict producing some trail  $\Gamma'$  different from  $\Gamma$  if and only if  $\text{level}_\Gamma(E) > 0$ .

Continuing with the conflict state rules, UndoClear and UndoDecide are unchanged. Proof-carrying LearnBackjump generates the proof term  $\text{lem}(H.c)$ , recording that the learned lemma  $L$  is a clausal form of  $H$ , and turning the conflict proof term  $c$  that represents a proof of unsatisfiability of  $E \uplus H$  into a deduction proof term that represents a proof of  $L$  from  $E$ . The main rule for proof reconstruction is proof-carrying Resolve, which combines proof term  $c$ , witnessing the unsatisfiability of the conflict, with proof term  $j$  witnessing that one of the assignments in the conflict, named  $A$ , follows from prior assignments:  $A$  is retained in the proof term  $\text{res}(j, {}^a A.c)$ . By applying this mechanism, proof-carrying CDSAT connects a proof of why a conflict  $E$  follows from  $H_0$  with a proof of why  $E$  is unsatisfiable, and generates a proof of unsatisfiability of  $H_0$ . The following theorem shows that proof-carrying CDSAT maintains *provability invariants* connecting the operations of the transition system with provability in the CDSAT proof system.

**Theorem 12** *For all proof-carrying CDSAT-derivations*

- If a trail containing  $\overline{H} \vdash_j A$  is generated, then  $H \vdash j : A$ ;
- If a conflict state  $\langle \Gamma; E; c \rangle$  is reached, then  $E \vdash c : \perp$ .

**Proof** The two claims are proved simultaneously by induction on the number of transition steps yielding the justified assignment or the conflict state, respectively. The base case covers input justified assignments, justified assignments placed on the trail by Deduce, and conflict states of the form  $\langle \Gamma; J \cup \{{}^a \overline{L}\}; \text{cfl}(j_k, a) \rangle$ . A justified assignment  $\emptyset \vdash_{\text{in}(A)} A$  is on the trail because  $A$  is initial:  $\emptyset \vdash_{\text{in}(A)} A$  follows by the *in* rule of the CDSAT proof system. For a justified assignment  $J \vdash_{j_k} L$  placed on the trail by Deduce,  $J \vdash_k j_k : L$ , hence  $J \vdash j_k : L$  by coercion. For a conflict state of the form  $\langle \Gamma; J \cup \{{}^a \overline{L}\}; \text{cfl}(j_k, a) \rangle$ , we have  $J \vdash_k j_k : L$ , hence  $J \cup \{{}^a \overline{L}\} \vdash \text{cfl}(j_k, a) : \perp$  by the *cfl* rule. The inductive step covers a justified assignment placed on the trail by LearnBackjump and conflict states generated by Resolve. For a justified assignment  $\overline{E} \vdash_{\text{lem}(H.c)} L$ , by induction hypothesis  $E \uplus H \vdash c : \perp$ , and  $L$  is a clausal form of  $H$ , so that  $E \vdash \text{lem}(H.c) : L$  by the *lem* rule. For a conflict state of the form  $\langle \Gamma; E \cup H; \text{res}(j, {}^a A.c) \rangle$ , by induction hypothesis  $E \uplus \{{}^a A\} \vdash c : \perp$ , and  $(\overline{H} \vdash_j A) \in \Gamma$ , so that  $E \cup H \vdash \text{res}(j, {}^a A.c) : \perp$  by the *res* rule. □

Theorems 11 and 12 together show that proof-carrying CDSAT builds a trace of proof terms in such a way to keep track of the CDSAT soundness invariants through the provability invariants. The next example expands Fig. 2 into a full derivation by proof-carrying CDSAT. For the sake of readability, we omit identifiers and we abuse the formalism by building proof terms made of assignments rather than their identifiers.

**Example 10** Assume that the input assignment for the example in Fig. 2 is  $\{\neg l_4 \vee l_5, \neg l_2 \vee \neg l_4 \vee \neg l_5, l_2 \vee (z \simeq y), x \not\leq 0 \vee l_2, x \not\leq 0 \vee l_4, f(z) \leftarrow \text{blue}, f(y) \leftarrow \text{red}\}$ , where  $l_2$  is  $y \geq 0$ ,

$l_4$  is  $x+y>0$ , and  $x \not\leq 0$  abbreviates  $\neg(x \leq 0)$ . The initial trail  $\Gamma_0$  contains these assignments. The derivation proceeds as in Fig. 2 with decisions  $?A_1, ?l_2, ?A_3, ?l_4$ , and propagation  $(\neg l_4 \vee l_5), l_4 \vdash l_5$ , but here we assume that  $A_1$  is  $x \leftarrow^{3/4}$ . Let  $\Gamma_1$  be the trail up to this point. The first conflict state is  $\langle \Gamma_1; \{\neg l_2 \vee \neg l_4 \vee \neg l_5, l_2, l_4, l_5\}; c_1 \rangle$  with conflict proof term

$$c_1 = \text{cfl}(\text{UP}(\neg l_2 \vee \neg l_4 \vee \neg l_5, \{l_2, l_4\}), l_5)$$

that registers the conflict between  $l_5$  and  $\neg l_5$ , the latter derived by Unit Propagation from  $\neg l_2 \vee \neg l_4 \vee \neg l_5, l_2$ , and  $l_4$ . The Resolve step from Fig. 2 replaces  $l_5$  in the conflict by its justification  $\{\neg l_4 \vee l_5, l_4\}$ , yielding conflict state  $\langle \Gamma_1; \{\neg l_2 \vee \neg l_4 \vee \neg l_5, l_2, l_4, \neg l_4 \vee l_5\}; c_2 \rangle$ . The associated conflict proof term is

$$c_2 = \text{res}(\text{UP}(\neg l_4 \vee l_5, \{l_4\}), l_5.c_1),$$

which plugs on top of the leaf  $l_5$  of  $c_1$  its theory proof  $\text{UP}(\neg l_4 \vee l_5, \{l_4\})$ . Let LearnBackjump solve the conflict as in Example 1 by placing on the trail

$$\overline{(\neg l_2 \vee \neg l_4 \vee \neg l_5), (\neg l_4 \vee l_5)} \vdash_{j_1} : \neg l_2 \vee \neg l_4,$$

with deduction proof term  $j_1 = \text{lem}(\{l_2, l_4\}.c_2)$  recording that conflict proof  $c_2$  showed that the learned lemma  $\neg l_2 \vee \neg l_4$  is inferred from  $\neg l_2 \vee \neg l_4 \vee \neg l_5$  and  $\neg l_4 \vee l_5$ . Proceeding as in Example 6, a Deduce step adds  $\overline{(\neg l_2 \vee \neg l_4), l_2} \vdash_{j_2} : \overline{l_4}$ , with deduction proof term  $j_2 = \text{UP}(\neg l_2 \vee \neg l_4, \{l_2\})$ . Suppose that the derivation continues with a Deduce step encapsulating the LRA evaluation inference

$$(x \leftarrow^{3/4}) \vdash_{j_3} : \overline{x \leq 0},$$

with deduction proof term  $j_3 = \text{eval}(\{x \leftarrow^{3/4}\})$ . The resulting trail  $\Gamma_2$  contains the initial assignments followed by

$$?A_1, ?l_2, \overline{(\neg l_2 \vee \neg l_4 \vee \neg l_5), (\neg l_4 \vee l_5)} \vdash_{j_1} : (\neg l_2 \vee \neg l_4), \overline{(\neg l_2 \vee \neg l_4), l_2} \vdash_{j_2} : \overline{l_4}, A_1 \vdash_{j_3} : \overline{(x \leq 0)}.$$

At this stage, the LRA-procedure detects the LRA-conflict  $\{l_2, \overline{l_4}, \overline{x \leq 0}\}$  and explains it by the FM-resolution inference

$$0 \leq y, y \leq -x \vdash_{\text{LRA}} \text{FM}(0 \leq y, y \leq -x) : 0 \leq -x,$$

mirrored in CDSAT proof system as the inference

$$y \geq 0, \overline{x+y>0}, \overline{x \leq 0} \vdash_{c_3} : \perp,$$

(see the cfl rule in Fig. 6) with conflict proof term

$$c_3 = \text{cfl}(\text{FM}(0 \leq y, y \leq -x), 0 \leq -x).$$

The conflict state is  $\langle \Gamma_2; l_2, \overline{l_4}, \overline{x \leq 0}; c_3 \rangle$ . A Resolve step replaces  $\overline{l_4}$  by its justification  $\{\neg l_2 \vee \neg l_4, l_2\}$ , producing conflict state  $\langle \Gamma_2; \{l_2, \neg l_2 \vee \neg l_4, \overline{x \leq 0}\}; c_4 \rangle$  with conflict proof term  $c_4 = \text{res}(j_2, \overline{l_4}.c_3)$  that expands upward leaf  $\overline{l_4}$  of  $c_3$  with its proof  $j_2$ . The system exits from this conflict by a LearnBackjump transition that jumps back to level 1 and learns lemma  $(x \leq 0) \vee \neg l_2$  with deduction proof term  $j_4 = \text{lem}(\{x \leq 0, l_2\}.c_4)$ . The resulting trail  $\Gamma_3$  contains the initial assignments followed by

$$?A_1, \overline{(\neg l_2 \vee \neg l_4 \vee \neg l_5), (\neg l_4 \vee l_5)} \vdash_{j_1} : (\neg l_2 \vee \neg l_4), A_1 \vdash_{j_3} : \overline{(x \leq 0)}, \overline{(\neg l_2 \vee \neg l_4)} \vdash_{j_4} : (x \leq 0) \vee \neg l_2.$$

With the last lemma the system has learned that if  $y \geq 0$  ( $l_2$ ) implies  $\overline{x+y>0}$  ( $\neg l_4$ ), then  $y \geq 0$  implies  $x \leq 0$ . Next, Deduce expands  $\Gamma_3$  with the assignments

$$(x \leq 0) \vee \neg l_2, \overline{x \leq 0} \vdash_{j_5} : \overline{l_2}, \quad l_2 \vee (z \simeq y), \overline{l_2} \vdash_{j_6} : z \simeq y,$$

carrying proof terms  $j_5 = \text{UP}(x \leq 0 \vee \neg l_2, \{\overline{x \leq 0}\})$  and  $j_6 = \text{UP}(l_2 \vee (z \simeq y), \{\overline{l_2}\})$ . A Deduce step for EUF inference  $\{f(z) \leftarrow \text{blue}, f(y) \leftarrow \text{red}\} \vdash_{j_7} : f(z) \not\approx f(y)$  further adds

$$f(z) \leftarrow \text{blue}, f(y) \leftarrow \text{red} \vdash_{j_7} : f(z) \not\approx f(y)$$

where  $j_7 = \text{neq}(f(z) \leftarrow \text{blue}, f(y) \leftarrow \text{red})$ . Let  $\Gamma_4$  be the resulting trail. At this point ConflictSolve fires, as the EUF inference  $\{z \simeq y, f(z) \not\approx f(y)\} \vdash_{c_5} : \perp$  leads to conflict state  $\langle \Gamma_4; \{f(z) \not\approx f(y), z \simeq y\}; c_5 \rangle$ , with conflict proof term

$$c_5 = \text{cfl}(\text{Cong}(z \simeq y), f(z) \not\approx f(y)).$$

A Resolve step yields conflict state  $\langle \Gamma_4; \{f(z) \not\approx f(y), l_2 \vee (z \simeq y), \overline{l_2}\}; c_6 \rangle$ , with conflict proof term  $c_6 = \text{res}(j_6, (z \simeq y).c_5)$  that expands upward leaf  $z \simeq y$  of  $c_5$  with its proof  $j_6$ . Similarly, another Resolve step produces conflict state

$$\langle \Gamma_4; \{f(z) \not\approx f(y), l_2 \vee (z \simeq y), (x \leq 0) \vee \neg l_2, \overline{x \leq 0}\}; c_7 \rangle,$$

with conflict proof term  $c_7 = \text{res}(j_5, \overline{l_2}.c_6)$  that expands upward leaf  $\overline{l_2}$  of  $c_6$  with its proof  $j_5$ . The conflict is solved by a LearnBackjump transition that jumps back to level 0 and learns  $x \leq 0$  as

$$f(z) \not\approx f(y), l_2 \vee (z \simeq y), (x \leq 0) \vee \neg l_2 \vdash_{j_8} : (x \leq 0),$$

with deduction proof term  $j_8 = \text{lem}(\{\overline{x \leq 0}\}.c_7)$ . As a byproduct, decision  $A_1$  is gone, and the resulting trail  $\Gamma_5$  contains the initial assignments, the learned lemmas, namely  $\neg l_2 \vee \neg l_4$ ,  $x \leq 0 \vee l_2$ , and  $x \leq 0$ , and the level 0 propagation  $f(z) \not\approx f(y)$ . Lemma  $x \leq 0$  enables Deduce to perform two unit propagations involving input clauses:

$$x \not\leq 0 \vee l_2, x \leq 0 \vdash_{j_9} : l_2, \quad x \not\leq 0 \vee l_4, x \leq 0 \vdash_{j_{10}} : l_4,$$

with  $j_9 = \text{UP}(x \not\leq 0 \vee l_2, \{x \leq 0\})$  and  $j_{10} = \text{UP}(x \not\leq 0 \vee l_4, \{x \leq 0\})$ . Let  $\Gamma_6$  be  $\Gamma_5$  thus expanded. At this point the conflict in  $\langle \Gamma_6; \{\neg l_2 \vee \neg l_4, l_2, l_4\}; c_8 \rangle$  is at level 0. Conflict proof term  $c_8 = \text{cfl}(\text{UP}(\neg l_2 \vee \neg l_4, \{l_2\}), l_4)$  records the conflict between  $l_4$  and  $\neg l_4$ , the latter derived by Unit Propagation from  $\neg l_2 \vee \neg l_4$  and  $l_2$ . While CDSAT would halt, proof-carrying CDSAT performs the series of Resolve steps in Fig. 8. When Resolve replaces a justified assignment by its justification, the proof term evolves as seen before. When Resolve removes an initial assignment from the conflict, the corresponding leaf in the associated proof term gets surrounded by the in constructor to mark it as a leaf of the final proof. When the conflict is empty, Fail fires returning  $\text{unsat}(c_{21})$ .

This example shows how lemma learning avoids repeating work: if conflict  $\{\neg l_2 \vee \neg l_4 \vee \neg l_5, l_2, l_4, \neg l_4 \vee l_5\}$  is solved by Backjump (see Example 3), rather than LearnBackjump (see Examples 1 and 10), trail  $\Gamma_1$  of Example 10 would be expanded with

$$(\neg l_2 \vee \neg l_4 \vee \neg l_5), (\neg l_4 \vee l_5), l_2 \vdash_{j_{11}} : \overline{l_4}$$

where  $j_{11} = \text{lem}(\{l_4\}.c_2)$ . Conflict  $\{\neg l_2 \vee \neg l_4, l_2, l_4\}$  would not be detected as in Example 10, and more steps would be necessary to discover it. Such steps would build another proof of  $\neg l_2 \vee \neg l_4$ , possibly identical to the first forgotten one. Furthermore, lemmas can be reused

$\{\neg l_2 \vee \neg l_4, l_2, x \leq 0 \vee l_4, x \leq 0\}$	$c_9 = \text{res}(j_{10}, l_4.c_8)$
$\{\neg l_2 \vee \neg l_4, x \leq 0 \vee l_2, x \leq 0 \vee l_4, x \leq 0\}$	$c_{10} = \text{res}(j_9, l_2.c_9)$
$\{\neg l_2 \vee \neg l_4, x \leq 0 \vee l_2, x \leq 0 \vee l_4, f(z) \not\approx f(y), l_2 \vee z \approx y, x \leq 0 \vee \neg l_2\}$	$c_{11} = \text{res}(j_8, x \leq 0.c_{10})$
$\{\neg l_2 \vee \neg l_4, x \leq 0 \vee l_2, x \leq 0 \vee l_4, f(z) \not\approx f(y), l_2 \vee z \approx y\}$	$c_{12} = \text{res}(j_4, x \leq 0 \vee \neg l_2.c_{11})$
$\{\neg l_2 \vee \neg l_4, x \leq 0 \vee l_2, x \leq 0 \vee l_4, f(z) \not\approx f(y)\}$	$c_{13} = \text{res}(\text{in}(l_2 \vee z \approx y), l_2 \vee z \approx y.c_{12})$
$\{\neg l_2 \vee \neg l_4, x \leq 0 \vee l_2, x \leq 0 \vee l_4, f(z) \leftarrow \text{blue}, f(y) \leftarrow \text{red}\}$	$c_{14} = \text{res}(j_7, f(z) \not\approx f(y).c_{13})$
$\{\neg l_2 \vee \neg l_4, x \leq 0 \vee l_2, x \leq 0 \vee l_4, f(z) \leftarrow \text{blue}\}$	$c_{15} = \text{res}(\text{in}(f(y) \leftarrow \text{red}), f(y) \leftarrow \text{red}.c_{14})$
$\{\neg l_2 \vee \neg l_4, x \leq 0 \vee l_2, x \leq 0 \vee l_4\}$	$c_{16} = \text{res}(\text{in}(f(z) \leftarrow \text{blue}), f(z) \leftarrow \text{blue}.c_{15})$
$\{\neg l_2 \vee \neg l_4, x \leq 0 \vee l_2\}$	$c_{17} = \text{res}(\text{in}(x \leq 0 \vee l_4), x \leq 0 \vee l_4.c_{16})$
$\{\neg l_2 \vee \neg l_4\}$	$c_{18} = \text{res}(\text{in}(x \leq 0 \vee l_2), x \leq 0 \vee l_2.c_{17})$
$\{\neg l_2 \vee \neg l_4 \vee \neg l_5, \neg l_4 \vee l_5\}$	$c_{19} = \text{res}(j_1, \neg l_2 \vee \neg l_4.c_{18})$
$\{\neg l_2 \vee \neg l_4 \vee \neg l_5\}$	$c_{20} = \text{res}(\text{in}(\neg l_4 \vee l_5), \neg l_4 \vee l_5.c_{19})$
$\emptyset$	$c_{21} = \text{res}(\text{in}(\neg l_2 \vee \neg l_4 \vee \neg l_5), \neg l_2 \vee \neg l_4 \vee \neg l_5.c_{20})$

**Fig. 8** Conflicts and conflicts proof terms produced by the final series of steps in Example 10

and may appear multiple times in the final proof term. Lemma  $\neg l_2 \vee \neg l_4$  is used twice in Example 10, and it occurs twice in  $c_{13} - c_{21}$ : once in  $c_8$ , and once in  $c_{12}$ , because  $c_{12}$  contains  $j_4$ , which contains  $c_4$ , which contains  $j_2$ , where  $\neg l_2 \vee \neg l_4$  appears. Also, deduction proof terms referring to first-order decisions do not appear in the final proof term: an inspection of  $c_{21}$  shows that  $j_3$  is absent. The reason is that first-order decisions play a role in finding models, not proofs. An easy optimization avoids constructing deduction proof terms involving first-order decisions.

## 7 Proof Reconstruction: From Proof Terms to Proofs

The motivation for using the proof-carrying CDSAT system is the ability to justify the unsatisfiability of an input with a proof. When CDSAT concludes  $\text{unsat}(c)$ , the proof term  $c$  and its associated derivation of  $\vdash c : \perp$  can be considered as a proof of unsatisfiability of the input, following Theorem 11. If need be, the rules of Fig. 6 can be used for proof checking. If another proof format is preferred,  $c$  indicates how a proof in that format can be *reconstructed*, having CDSAT traced in  $c$  how a contradiction was reached from a logical point of view. Indeed, a deduction proof term  $j$  with  $H \vdash j : A$  (resp. a conflict proof term  $c$  with  $E \vdash c : \perp$ ) can be decoded into, or can be seen as denoting, a proof of  $H_0 \cup H \models A$  (resp.  $H_0 \cup E \models \perp$ ) in the format of choice.

A first option is to decode proof terms into proofs after CDSAT halts, in a post-processing phase. A second option consists of identifying first the proof operations corresponding to the rules of Fig. 6 in the target proof format and then reading the proof-carrying CDSAT system as manipulating directly the proofs *denoted* by proof terms such as  $\text{in}(A)$ ,  $j_k$ ,  $\text{lem}(H.c)$ ,  $\text{cfl}(j_k, a)$ , and  $\text{res}(j, {}^a A.c)$ . In other words, a CDSAT-based solver would build in memory not the proof terms, but the proofs themselves. Of course, the execution of the above-mentioned proof operations during a CDSAT derivation may increase its runtime. In any case, proof-carrying CDSAT is modular in the way theory proofs are handled, reconstructed, and checked. In Sect. 7.1 we exemplify proof reconstruction by showing how CDSAT proof terms can be turned into *resolution-style proof trees*. In Sect. 7.2 we discuss yet another alternative consisting of applying to CDSAT the *LCF approach* to proofs.

### 7.1 Proof Format Based on Resolution

A resolution proof is usually represented as a resolution proof tree with nodes labeled by clauses. However, CDSAT views logical connectives as interpreted symbols of theory Bool, treats formulæ as terms of sort prop, allows assignments such as  $(l_1 \vee l_2) \leftarrow \text{false}$ , and does not assume that the input is a set of clauses. Therefore, we distinguish between *object-level clauses* of the form  $l_1 \vee \dots \vee l_m$ , where the  $l_i$ 's are terms of sort prop and  $\vee$  is the symbol for disjunction in  $\Sigma_{\text{Bool}}$ , and *CDSAT clauses* written  $L_1 || \dots || L_m$ , where the  $L_i$ 's are singleton Boolean assignments and  $||$  is a meta-level symbol for disjunction. Reconstructed proofs will use object-level clauses for input clauses and CDSAT clauses for generated clauses. Since in CDSAT there are first-order assignments, we introduce *guarded CDSAT clauses* of the form  $H \rightarrow C$ , where  $H$  is a set of first-order assignments, which can be empty, and  $C$  is a CDSAT clause. When there is no ambiguity, we use *clause* for CDSAT clause and *guarded clause* for guarded CDSAT clause. The *reconstruction of a resolution proof from a CDSAT proof term* yields a *CDSAT resolution proof*.

**Definition 9 (CDSAT Resolution Proof)** A *CDSAT resolution proof* is represented as a binary tree such that:

1. A leaf is labeled with either an input singleton assignment, or a guarded clause that is a *theory lemma*;
2. An internal node  $n$  is labeled with a guarded clause and has children  $n_1$  and  $n_2$  if its label can be inferred from those of  $n_1$  and  $n_2$  by one of the following inference rules:

$$\begin{array}{l}
 \text{Binary Resolution:} \\
 \text{Unit Resolution:} \\
 \text{First-order Assignment Elimination:}
 \end{array}
 \quad
 \begin{array}{c}
 \frac{H_1 \rightarrow C_1 || L \quad H_2 \rightarrow C_1 || \bar{L}}{H_1 \cup H_2 \rightarrow C_1 || C_2} \text{BR} \\
 \frac{L \quad H \rightarrow C || \bar{L}}{H \rightarrow C} \text{UR} \\
 \frac{A \quad \frac{H \rightarrow C}{H, A \rightarrow C}}{H \rightarrow C} \text{FOAE}
 \end{array}$$

where  $L$  and  $A$  are input singleton assignments labeling leaves.

Theory lemmas are treated as leaves, because theory proofs involve inference rules other than resolution. Since CDSAT treats propositional logic as a theory, there are also theory lemmas for Bool or Bool-lemmas. If  $L_0$  is a clausal form of  $\{L_1, \dots, L_m\}$  (see Definition 1), the following clauses are Bool-lemmas:

$$\emptyset \rightarrow \bar{L}_0 || \bar{L}_1 || \dots || \bar{L}_m \quad \emptyset \rightarrow L_0 || L_i \quad (1 \leq i \leq m). \tag{9}$$

Indeed,  $\bar{L}_0 || \bar{L}_1 || \dots || \bar{L}_m$  and  $L_0 || L_i, 1 \leq i \leq m$ , are tautologies, since  $L_0$  is a clausal form of  $\{L_1, \dots, L_m\}$ , hence they can label leaves. The first Bool-lemma allows one to transform the object-level clause to which  $L_0$  assigns true into a CDSAT clause:

$$\frac{\dots}{\frac{H \rightarrow C || L_0 \quad \emptyset \rightarrow \bar{L}_0 || \bar{L}_1 || \dots || \bar{L}_m}{H \rightarrow C || \bar{L}_1 || \dots || \bar{L}_m} \text{BR}$$

Conversely, the other lemmas allow one to turn a CDSAT clause  $\overline{L_1} \parallel \dots \parallel \overline{L_m}$  into an object-level clause:

$$\frac{\frac{\dots}{H \rightarrow C \parallel \overline{L_1} \parallel \dots \parallel \overline{L_m}}{\emptyset \rightarrow L_0 \parallel L_1} \text{BR}}{\dots} \frac{\emptyset \rightarrow L_0 \parallel L_m}{H \rightarrow C \parallel L_0} \text{BR}$$

These transformations can be synthesized into derivable inference rules:

$$\frac{H \rightarrow C \parallel L_0}{H \rightarrow C \parallel \overline{L_1} \parallel \dots \parallel \overline{L_m}} \text{Unfold} \quad \frac{H \rightarrow C \parallel \overline{L_1} \parallel \dots \parallel \overline{L_m}}{H \rightarrow C \parallel L_0} \text{Fold}$$

that involve Bool-lemmas only if  $m \geq 2$ , because if  $m = 1$ ,  $L_0$  is simply  $\overline{L_1}$ . An assignment  $H$  can be partitioned as  $H_{FO} \uplus \{L_1, \dots, L_n\}$ , where  $H_{FO}$  contains all the singleton first-order assignments in  $H$ , and  $L_1, \dots, L_n$  are all the singleton Boolean assignments in  $H$ . Let  $H_{\text{clause}}$  be the clause  $\overline{L_1} \parallel \dots \parallel \overline{L_n}$ , that is,  $H_{\text{clause}}$  is a clausal form of  $\{L_1, \dots, L_n\}$  (see Definition 1) written as a CDSAT clause. Then CDSAT proof terms are transformed into CDSAT resolution proofs by turning:

- a deduction proof term  $\emptyset \vdash \text{in}(A) : A$  into a leaf labeled  $A$ ;
- a deduction proof term  $H \vdash j : L$  of any other form into a proof of  $H_{FO} \rightarrow H_{\text{clause}} \parallel L$ ; and
- a conflict proof term  $H \vdash c : \perp$  into a proof of  $H_{FO} \rightarrow H_{\text{clause}}$ .

The application of this transformation to a proof term is denoted by surrounding the proof term with  $\llbracket$  and  $\rrbracket$ . The definition of this transformation is inductive and is given in Fig. 9, which should be read together with Fig. 6. The first, second, and third cases in Fig. 9 are base cases that yield leaves according to Case (1) in Definition 9. The remaining cases are inductive cases, where the recursive application of the transformation is represented as a subproof with the transformed proof term as premise and the result of the transformation as consequence. The rule for *res* in Fig. 6 is articulated into three rules in Fig. 9 distinguishing among *Unit Resolution*, *First-order Assignment Elimination*, and *Binary Resolution* according to Case (2) in Definition 9.

For instance, consider the transformation of a deduction proof term and a conflict proof term that encapsulate a unit propagation, where  $K$  is a clausal form of  $H = H' \uplus \{L\}$  as in inference (1) from Sect. 3.2:

$$\frac{\llbracket \{^a K\} \uplus H' \vdash \text{UP}(a, \{a_1, \dots, a_n\}) : \overline{L} \rrbracket}{\llbracket \{^a K\} \uplus H' \uplus \{^a L\} \vdash \text{cfl}(\text{UP}(a, \{a_1, \dots, a_n\}), a_0) : \perp \rrbracket}$$

where  $a_1, \dots, a_n$  are the identifiers of the elements of  $H'$ . Both transformations yield the Bool-lemma

$$\emptyset \rightarrow \overline{K} \parallel H'_{\text{clause}} \parallel \overline{L}$$

by applying the second and the fourth rules in Fig. 9, respectively.

Since a CDSAT answer of the form  $\text{unsat}(c)$  means  $\emptyset \vdash c : \perp$ , the resolution proof reconstructed from  $c$  is a *refutation*, as its conclusion is the empty clause  $\emptyset \rightarrow \emptyset$ . Since first-order decisions do not appear in proofs, first-order assignments may appear in proofs only if they belong to the initial assignment of an SMA problem. If the input problem is an SMT problem, and therefore contains no first-order assignments, the reconstructed proof involves only

$$\begin{array}{lcl}
 \llbracket \emptyset \vdash \text{in}(A) : A \rrbracket & := & A \\
 \llbracket J \vdash j_k : L \rrbracket & := & J_{\text{FO}} \rightarrow J_{\text{clause}} \parallel L \\
 \begin{array}{l} \llbracket E \vdash \text{lem}(H.c) : L \rrbracket \\ \text{where } L \text{ is a clausal form of } H \end{array} & := & \frac{\llbracket E \uplus H \vdash c : \perp \rrbracket}{E_{\text{FO}} \rightarrow E_{\text{clause}} \parallel H_{\text{clause}}} \text{Fold} \\
 \llbracket J \uplus \{^a \bar{L}\} \vdash \text{cfl}(j_k, a) : \perp \rrbracket & := & J_{\text{FO}} \rightarrow J_{\text{clause}} \parallel L \\
 \llbracket E \vdash \text{res}(\text{in}(L), ^a L.c) : \perp \rrbracket & := & \frac{L \quad E_{\text{FO}} \rightarrow E_{\text{clause}} \parallel \bar{L}}{E_{\text{FO}} \rightarrow E_{\text{clause}}} \text{UR} \\
 \begin{array}{l} \llbracket E \vdash \text{res}(\text{in}(A), ^a A.c) : \perp \rrbracket \\ \text{if } A \text{ is first-order} \end{array} & := & \frac{A \quad E_{\text{FO}} \uplus \{A\} \rightarrow E_{\text{clause}}}{E_{\text{FO}} \rightarrow E_{\text{clause}}} \text{FOAE} \\
 \begin{array}{l} \llbracket E \cup H \vdash \text{res}(j, ^a L.c) : \perp \rrbracket \\ \text{if } j \text{ is not of the form } \text{in}(A) \end{array} & := & \frac{\frac{\llbracket H \vdash j : L \rrbracket}{H_{\text{FO}} \rightarrow H_{\text{clause}} \parallel L} \quad \frac{\llbracket E \uplus \{^a L\} \vdash c : \perp \rrbracket}{E_{\text{FO}} \rightarrow E_{\text{clause}} \parallel \bar{L}}}{H_{\text{FO}} \cup E_{\text{FO}} \rightarrow H_{\text{clause}} \parallel E_{\text{clause}}} \text{BR}
 \end{array}$$

Fig. 9 Transformation of CDSAT proof terms into CDSAT resolution proofs

singleton Boolean assignments labeling leaves and guarded clauses of the form  $\emptyset \rightarrow C$  labeling leaves or internal nodes. In other words, the reconstructed proof is a resolution refutation in the standard sense, with leaves labeled by input assignments or theory lemmas. On the other hand, Bool-lemmas are a non-standard feature that also enables the *sharing* of resolution proofs. For instance, in the refutation of Example 10, the conflict proof term  $c_{19} = \text{res}(j_1, \neg l_2 \vee \neg l_4, c_{18})$  (see Fig. 8), with  $j_1 = \text{lem}(\{l_2, l_4\}, c_2)$ , yields  $L_1, L_2 \vdash c_{19} : \perp$ , where  $L_1$  is  $\neg l_4 \vee l_5$  and  $L_2$  is  $\neg l_2 \vee \neg l_4 \vee \neg l_5$ . The resolution proof reconstructed from  $c_{19}$  is:

$$\frac{\frac{\llbracket L_1, L_2, l_2, l_4 \vdash c_2 : \perp \rrbracket}{\emptyset \rightarrow \bar{L}_1 \parallel \bar{L}_2 \parallel \bar{l}_2 \parallel \bar{l}_4} \text{Fold} \quad \frac{\llbracket \neg l_2 \vee \neg l_4 \vdash c_{18} : \perp \rrbracket}{\emptyset \rightarrow (\neg l_2 \vee \neg l_4)} \text{BR}}{\emptyset \rightarrow \bar{L}_1 \parallel \bar{L}_2} \text{BR}$$

In the proof reconstructed from  $c_{21}$ , CDSAT clause  $\emptyset \rightarrow \bar{L}_1 \parallel \bar{L}_2$  resolves with initial assignments  $L_1$  and  $L_2$  to yield  $\emptyset \rightarrow \emptyset$ . The double occurrence of  $\neg l_2 \vee \neg l_4$  in  $c_{18}$  (see Example 10), means that the resolution proof  $\llbracket \neg l_2 \vee \neg l_4 \vdash c_{18} : \perp \rrbracket$  has two leaves labeled by the same Bool-lemma

$$\emptyset \rightarrow (\neg l_2 \vee \neg l_4) \parallel \bar{l}_2 \parallel \bar{l}_4.$$

An alternative refutation can be obtained by replacing those two leaves with the subproof translating  $c_2$ , and replacing  $(\neg l_2 \vee \neg l_4)$  by  $\bar{L}_1 \parallel \bar{L}_2$  in all nodes underneath. This avoids the explicit conversions between the object-level clause  $\neg l_2 \vee \neg l_4$  and the CDSAT clause  $\bar{l}_2 \parallel \bar{l}_4$ . However, in this alternative proof, the subtree for  $c_2$  is duplicated. Such duplications are

```

type deduction
type conflict
in : single_assign -> deduction
coerc : 'k theory_handler -> 'k theory_proof -> deduction
lem : conflict -> assign -> deduction
cfl : 'k theory_handler -> 'k theory_proof -> conflict
res : deduction -> conflict -> conflict
reveal : conflict -> assign

```

**Fig. 10** API (Application Programming Interface) exported by a CDSAT kernel

customary in resolution proof trees where there is only one kind of clauses. By distinguishing between object-level clauses and CDSAT clauses, and using the former for input clauses and the latter for generated clauses, CDSAT natively supports the sharing of subproofs that one obtains by replacing trees with directed acyclic graphs.

## 7.2 An LCF Architecture for CDSAT

Another example of proof format is the “dummy” one, where proofs do not contain any information other than *what they are supposed to be the proofs of*:

1. A deduction proof for proof term  $j$  with  $H \vdash j : L$  is the pair  $\langle H, L \rangle$ , and
2. A conflict proof for proof term  $c$  with  $H \vdash c : \perp$  is  $H$ .

Although this proof format does not allow any proof checking, the trustworthiness of a reasoner producing such proofs can still be established by the LCF programming abstraction [30,47]. This approach uses a type `theorem`, whose constructed inhabitants are provable formulæ. Actually, this type is defined as an alias for the type `formula` of formulæ, but this is known only to a fixed and well-identified piece of code, called the *LCF kernel*. This kernel hides the definition of `theorem` to the outside world and exports a range of kernel primitives to manipulate inhabitants of type `theorem` in a safe and provably correct way. For instance, assuming that  $\Rightarrow$  denotes implication, a primitive

```
modus_ponens : theorem -> theorem -> theorem
```

takes as arguments two inhabitants  $F$  and  $G$  of type `theorem`, checks that  $F$  is of the form  $G \Rightarrow R$ , and returns  $R$  as an inhabitant of `theorem`. The kernel can export a primitive that reveals that an inhabitant of `theorem` is a formula, but not one that casts any inhabitant of `formula` into type `theorem`. Thus, the existence of an inhabitant  $F$  of `theorem` witnesses the fact that  $F$  is provable, as an inhabitant only results from a series of correct manipulations by the kernel primitives: if the kernel code is trusted, then  $F$  can be trusted to be a theorem, while no proof has ever been constructed in memory. CDSAT is well-suited for the LCF approach: given a type `assign` for assignments and `single_assign` for singleton assignments, a trusted kernel defines types

```

type deduction = assign * single_assign
type conflict = assign

```

hides their definitions to the outside world and exports a range of primitives corresponding to the proof term constructs.

The signature in Fig. 10 lists hidden-type definitions and exported primitives. The primitives check that the conditions of the rules in Fig. 6 are met: `in` checks that its argument is one of the initial assignments; `lem` takes as arguments a conflict  $E$  and an assignment

$H$ , checks that  $H$  is Boolean and included in  $E$ , computes a clausal form  $L$  of  $H$ , and produces the deduction  $\langle E \setminus H, L \rangle$ , where  $\setminus$  is set subtraction. Primitive `res` takes a deduction  $\langle H, A \rangle$  and a conflict, checks that  $A$  occurs in the conflict, and returns the conflict where  $A$  is replaced by  $H$ . Primitives `coerc` and `cfl` take as arguments a  $\mathcal{T}_k$ -proof,  $1 \leq k \leq n$ , given as an inhabitant of `'k theory_proof`, a type parameterized by  $k$ . Their first argument is a handler for theory  $\mathcal{T}_k$ , whose type `'k theory_handler` is parameterized by a matching  $k$ , as implemented for example by a generalized algebraic datatype [19,55]. The handler allows the primitives to check that  $\mathcal{T}_k$  is one of the combined theories before coercing the  $\mathcal{T}_k$ -proof, trusted to be correct, into a deduction or a conflict. Proof-carrying CDSAT can be programmed on top of this kernel, so that, when it halts with answer `unsat(c)`, the proof term  $c$  is an inhabitant of type `conflict`. The `reveal` primitive applied to  $c$  will return the empty assignment. Although no proof has been constructed in memory, the answer is *correct by construction*.

## 8 Discussion

Conflict-driven satisfiability procedures work by building partial assignments, detecting conflicts when the assignment falsifies the input formula, and performing conflict-driven inferences to explain conflicts and reorient the search. In prior work [11], we presented CDSAT as a conflict-driven combination framework for disjoint theories and proved its soundness, termination, and completeness. The present article has extended the theoretical foundations of CDSAT in three main directions, namely *lemma learning*, properties of *theory modules* (i.e., the theory inference systems that CDSAT orchestrates), and *proof generation*.

We generalized the *lemma learning* capability of CDSAT, in such a way that new clauses can be *formed and learned* during backjumping from a conflict, and the destination level of backjumping can be chosen according to different heuristics, including *learn and restart*. Soundness, termination, and completeness of CDSAT are preserved.

We proved that the theory modules listed in [11] for the *Boolean theory*, *equality*, *arrays with extensionality*, and *linear rational arithmetic*, as well as generic black-box modules for stably infinite and non-stably infinite theories, are *complete with respect to all suitable leading theories* in a union of theories, and admit *finite local bases*. We also showed how to get a *finite global basis* for the union of the theories from finite local bases for the individual theories. These results mean that the assumptions for the termination and completeness of CDSAT can indeed be satisfied, complementing our previous general results [11]. By including black-box modules for stably infinite theories CDSAT subsumes the equality-sharing method for theory combination [42,48,49], also known as Nelson-Oppen scheme, including model-based theory combination [23]. By handling also non-stably infinite theories, CDSAT goes beyond equality sharing (see [9] for a survey of other approaches to this issue).

We presented a *proof-carrying CDSAT transition system* that constructs and carries *proof terms*, so that a proof can be reconstructed when CDSAT discovers that the input problem is unsatisfiable. CDSAT proof terms can be rendered in a number of proof formats, and the resulting proofs checked by a trusted checker, or shown to be correct by construction in LCF style. A translation to resolution-style proofs is illustrated as example.

Research on CDSAT as a new paradigm for theory combination has just begun, and there are many directions for future work. A main objective is an implementation of CDSAT, either in the form of a CDSAT-based prototype (e.g., [6,14]) or by extending the implementation of MCSAT in the Yices SMT solver [25]. An implementation can be used for exploring and

evaluating different *search plans* and *proof formats*, the latter in connection with the objective of efficient proof checking. The design of a CDSAT search plan involves both *global* issues about reasoning in the theory union and *local* issues about reasoning in each theory. At the local level, each theory search plan is in charge of detecting the *applicability* of inferences and the *acceptability* of decisions. At the global level, the search plan decides which CDSAT transition rule to apply next, coordinates the theory modules, prioritizing them with respect to both decisions and deductions, and controls lemma learning.

CDSAT proofs may be extended to account for *preprocessing and inprocessing techniques* (e.g., [2,38]), evaluating the *cost* of proof generation and proof checking (e.g., [2,3]), and studying proof formats to reduce it (e.g., [22]). At the foundational level, we may investigate empowering CDSAT to handle unions of non-disjoint theories (e.g., [20,28]), or formulæ with quantifiers, considering model-based conflict-driven instantiation (e.g., [4,51]), or integrations with first-order logic modules (e.g., [13,15]).

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