A Verified Automatic Prover Based on Ordered Resolution

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First-order theorem provers based on superposition, such as E, SPASS, and Vampire, play an important role in formal software verification. They are based on sophisticated logical calculi that combine ordered resolution and equality reasoning. They also employ advanced algorithms, data structures, and heuristics. As a step towards verifying the correctness of state-of-the-art provers, we specify, using the Isabelle/HOL proof assistant, a purely functional ordered resolution prover and formally establish its soundness and refutational completeness. Methodologically, we apply stepwise refinement to obtain, from an abstract specification of a nondeterministic prover, a verified deterministic program, written in a subset of Isabelle/HOL from which we extract purely functional Standard ML code that constitutes a semidecision procedure for first-order logic.

CCS Concepts: • Theory of computation → Logic and verification; Automated reasoning; • Software and its engineering → Completeness;

Additional Key Words and Phrases: automatic theorem provers, proof assistants, first-order logic, refinement

ACM Reference Format:

1 INTRODUCTION

Formal verification of programs aims at ensuring correctness by mechanically checking their behavior with respect to a logical specification. Automatic theorem provers based on superposition, such as E [Schulz 2013b], SPASS [Weidenbach et al. 2009], and Vampire [Kovács and Voronkov 2013], are often employed as backends in verification tools. They are used to discharge verification conditions [Bobot et al. 2011; Paulson and Blanchette 2012], but also to generate loop invariants [Kovács and Voronkov 2009].

Superposition is a highly successful logical calculus for first-order logic with equality, which generalizes both ordered resolution [Bachmair and Ganzinger 2001] and ordered (unfailing) completion [Bachmair and Ganzinger 2001] and ordered (unfailing) completion [Bachmair et al. 1989].

Resolution does not operate on first-order formulas but instead on sets of clauses. A clause is an n-ary disjunction of literals $L_1 \lor \cdots \lor L_n$ whose free variables are interpreted universally. Each literal is either an atom $A$ or its negation $\neg A$. An atom is a symbol applied to a tuple of terms—e.g., $\text{divides}(2, n)$. The empty clause, which is false, is denoted by $\bot$. Resolution works by refutation: Conceptually, the calculus proves a conjecture $\forall \bar{x}. C$ from axioms $A$ by deriving $\bot$ from $A \cup \{\exists\bar{x}. \neg C\}$, indicating its unsatisfiability.

Compared with plain resolution, ordered resolution relies on an order on the atoms to restrict the search space. Another important difference is that it uses a redundancy criterion to discard subsumed clauses at any point; for example, $p(x) \lor q(x)$ and $p(5)$ are subsumed by $p(x)$.

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2018. 2475-1421/2018/1-ART1 $15.00
https://doi.org/
Using formal verification, we aim to develop trustworthy programs. But why should anyone trust verification tools? In particular, modern superposition provers are highly optimized programs that rely on sophisticated calculi, with a rich metatheory, and specialized data structures. In this paper, we propose an answer by verifying, in Isabelle/HOL [Nipkow et al. 2002], a purely functional prover based on ordered resolution. The verification relies on stepwise refinement [Wirth 1971]. Four layers are connected by three refinement steps:

- Our starting point, layer 1 (Section 4), is an abstract Prolog-style nondeterministic resolution prover in a highly general form, as presented by Bachmair and Ganzinger [2001] and as formalized by Schlichtkrull et al. [2018a,b]. It operates on possibly infinite sets of clauses. Its soundness and refutational completeness are inherited by the other layers.
- Layer 2 (Section 5) operates on finite multisets of clauses and introduces a priority queue to ensure that logical inferences are performed in a fair manner, guaranteeing completeness: Given a valid conjecture, the prover will eventually find a proof.
- Layer 3 (Section 6) is a deterministic program that works on finite lists, committing to a concrete strategy for assigning priorities to clauses. However, it is not fully executable: It abstracts over operations on atoms and employs logical specifications instead of executable functions for some auxiliary notions.
- Finally, layer 4 (Section 7) is a fully executable program. It provides a concrete datatype for atoms and executable definitions for all auxiliary notions, including unifiers, clause subsumption, and the order on atoms.

From layer 4, we can extract Standard ML code by invoking Isabelle’s code generator [Haftmann and Nipkow 2010]. The resulting prover serves first and foremost as a proof of concept: It uses an efficient calculus (layer 1) and a reasonable strategy to ensure fairness (layers 2 and 3), but it depends on naive list-based data structures. Further refinement steps will be required to obtain a prover that is competitive with the state of the art.

The refinement steps connect vastly different levels of abstraction, spanning much of computer science. The most abstract level is occupied by an infinitary logical calculus and the semantics of first-order logic. Soundness and completeness relate these two notions. At the functional programming level, soundness amounts to a safety property: Whenever the program terminates normally, its outcome is correct, whether it is a proof or a finite saturation witnessing unprovability. Correspondingly, refutational completeness is a liveness property: The program will always terminate normally with a proof if the conjecture is valid. Our executable functional prover demonstrates that, far from being academic exercises, Bachmair and Ganzinger’s [2001] framework and its formalization by Schlichtkrull et al. [2018a,b] accurately capture the metatheory of actual provers.

To our knowledge, our program is the first verified prover for first-order logic implementing an optimized calculus. It is also the first example of the application of refinement in this context. This methodology has been used to verify SAT solvers [Blanchette et al. 2018; Marić 2010], which decide the satisfiability of propositional formulas, but first-order logic is semidecidable—sound and complete provers are guaranteed to terminate only for unsatisfiable (i.e., provable) clause sets. This complicates the transfer of completeness results across refinement layers.

Our contributions are as follows:

- We unveil a verified sound and complete first-order prover based on ordered resolution.
- We propose a general methodology, using modern tools, for refining an abstract Prolog-style definition of a refutational prover to an ML-style functional program, applicable to provers and other nondeterministic semidecision procedures that can be stated abstractly.
- We present a reusable library of Isabelle/HOL definitions, lemmas, and proofs that supports the methodology. These concern atoms, terms, substitutions, and derivation chains.
In addition, we offer a few “proof pearls”—smaller proving puzzles that illustrate specific techniques and that we find instructive or elegant. Our work is connected to the IsaFoL (Isabelle Formalization of Logic) project, which aims at developing a library of results about logic and automated reasoning. The Isabelle source files are available in the IsaFoL repository and in the Archive of Formal Proofs. The parts specific to the functional prover refinement amount to about 4000 lines of source text. A convenient way to study the files is to open them in the Isabelle/jEdit development environment, as explained in the repository’s readme file. This will ensure that logical symbols are rendered properly and will let you inspect proof states. The files were created using Isabelle version 2017, but the repositories will be updated to track Isabelle’s evolution.

2 ISABELLE/HOL

Isabelle is a generic proof assistant that supports multiple object logics. Its most developed instantiation, Isabelle/HOL, provides a version of classical higher-order logic (HOL) that supports rank-1 (top-level) polymorphism, Haskell-style type classes, and Hilbert’s choice operator. Unlike the type theories that underlie Agda and Coq, HOL has no built-in notion of computation or executability. Nonetheless, a substantial fragment of HOL corresponds closely to Standard ML or Haskell and can be exported to these languages using a code generator.

Isabelle’s syntax is inspired by both ML and traditional mathematical conventions. The types are built from type variables and n-ary type constructors, normally written in postfix notation (e.g., ‘a list). The infix type constructor ‘a ⇒ ‘b is interpreted as the (total) function space from ‘a to ‘b. Propositions are terms of type bool, a datatype equipped with the constructors False and True. The familiar logical symbols ∀, ∃, ¬, ∧, ∨, =⇒, ⇔, and = are normal functions, although the quantifiers and equality on functions fall outside the executable fragment.

Isabelle adheres to a tradition initiated by the LCF system: All logical inferences are derived by a small trusted kernel, and types and functions are defined rather than axiomatized to guard against inconsistencies. Isabelle/HOL provides high-level specification mechanisms inspired by typed functional programming (e.g., ML) and logic programming (e.g., Prolog). These let us define large classes of types and operations, such as inductive datatypes, recursive functions, inductive predicates, and their coinductive counterparts. For example, the codatatype and corec commands can be used to define codatatype and productive corecursive functions in the style of Haskell, and the coinductive command can be used to introduce coinductive predicates. Internally, Isabelle synthesizes suitable low-level nonrecursive definitions and derives the user specifications via primitive inferences. This foundational approach allows the system to provide a highly expressive, trustworthy specification language.

Isabelle proofs are expressed in a language called Isar. It encourages a declarative, hierarchical style reminiscent of the format suggested by Lamport, but with alphanumeric labels to identify intermediate proof steps. Isar also supports low-level tactics that manipulate the proof state directly, similar to those offered by Coq and other systems.

Most Isabelle formalizations are structured using locales. A locale is a parameterized module, similar to an ML functor. The parameters may be types or terms satisfying some assumptions. For example, Isabelle/HOL provides the following basic specifications:
locale semigroup =  
  fixes * :: 'a ⇒ 'a ⇒ 'a  
  assumes (a * b) * c = a * (b * c)  
locale monoid = semigroup +  
  fixes 1 :: 'a  
  assumes 1 * a = a and a * 1 = a

The semigroup locale is parameterized by a type 'a and a binary operation * on 'a, which must be associative. The monoid locale inherits these parameters and assumptions and enriches them with a constant 1 assumed to be left- and right-neutral. Once a locale is declared, we can enter its scope at any point in a formal development. Within a locale’s scope, we can use its parameters and assumptions in definitions, lemma statements, and proofs.

To actually use a locale, we must instantiate the parameters with concrete types and terms. For example, we can instantiate monoid by taking ('a, *, 1) to be (nat, +, 0) or (nat, ×, 1), where nat is the type of natural numbers and 0, 1, +, × have their usual semantics. Before we can retrieve the definitions and lemmas from a locale, we must discharge the assumptions (e.g., 0 + a = a for all a :: nat). If a locale is parameterized by exactly one type variable, it can be introduced as a type class instead. This can be useful to offload some bureaucracy onto the type system, but it has its limitations: As in Haskell, a type class can be instantiated with a given type at most once.

3 ATOMS AND SUBSTITUTIONS

The first three refinement layers are based on an abstract library of first-order atoms and substitutions. In the fourth and final layer, the abstract framework is instantiated with concrete datatypes and functions. We start from the library of clausal logic developed by Blanchette et al. [2018], which is parameterized by a type 'a of logical atoms. Literals are defined as an inductive datatype with constructors for positive and negative literals:

datatype 'a literal =  
  Pos 'a  
  | Neg 'a

The type of clauses is then defined as the abbreviation 'a clause = 'a literal multiset, where multiset is the type constructor of finite multisets. Thus, the clause ¬A ∨ B, where A and B are arbitrary atoms, is represented by the multiset {Neg A, Pos B}, and the empty clause ⊥ is represented by {}. The complement operation is defined as ¬ Neg A = Pos A and ¬ Pos A = Neg A for any atom A.

In automated reasoning, it is customary to view clauses as multisets of literals rather than as sets. One reason is that multisets behave more naturally under substitution. For example, applying the substitution {y ↔ x} to the two-literal clause p(x) ∨ p(y) results in a two-literal clause p(x) ∨ p(x), preserving the structure of the clause.

The truth value of ground (i.e., variable-free) atoms is given by a Herbrand interpretation: a set, of type 'a set, of all true ground atoms. The “models” predicate |= is defined as I |= A ⇔ A ∈ I. This definition is lifted to literals, clauses, and sets of clauses in the usual way:

I |= Pos A ⇔ A ∈ I  
I |= Neg A ⇔ A /∈ I  
I |= C ⇔ ∃L ∈ C. I |= L  
I |= D ⇔ ∀C ∈ D. I |= C

A set of clauses D is satisfiable if there exists an interpretation I such that I |= D.

Ordered resolution crucially depends on a notion of substitution and of most general unifier (MGU). These auxiliary concepts are provided by a third-party library, IsaFoR (Isabelle Formalization of Rewriting) [Thiemann and Sternagel 2009]. To reduce our dependency on external libraries, we hide them behind abstract locales parameterized by a type of atoms 'a and a type of substitutions 's. We will usually think of the atoms as being first-order terms, which can be either a variable or a symbol applied to a list of first-order terms. Another possibility would be to use applicative first-order terms, also called λ-free higher-order terms. A substitution is modeled as a function...
from variables to terms. Substitutions can be applied to first-order terms by mapping them onto the terms’ variables.

We start by defining a locale substitution_ops that declares the basic operations on substitutions: application (·), identity (id), and composition (o):

```
locale substitution_ops =
  fixes
    · :: 'a ⇒ 's ⇒ 'a and
    id :: 's and
    o :: 's ⇒ 's ⇒ 's
```

Within the locale’s scope, we introduce a number of derived concepts. Ground atoms are defined as atoms that are left unchanged by substitutions:

```
is_ground A ⇐⇒ ∀σ. A = A · σ
```

A ground substitution is a substitution whose application always results in ground atoms:

```
is_ground σ ⇐⇒ ∀A. is_ground(A · σ)
```

Nonstrict and strict generalization are defined as

```
generalizes A B ⇐⇒ ∃σ. A · σ = B
strictly generalizes A B ⇐⇒ generalizes A B ∧ ¬ generalizes B A
```

The operators on atoms are lifted to literals, clauses, and sets of clauses. The grounding of a clause is defined as

```
grounding_of C = {C · σ | is_ground σ}
```

The operator is lifted to sets of clauses in the obvious way. Clause subsumption is defined as

```
subsumes C D ⇐⇒ ∃σ. C · σ ⊆ D
strictly subsumes C D ⇐⇒ subsumes C D ∧ ¬ subsumes D C
```

Unifiers and MGUs are characterized as follows, where \( A :: 'a set \) represents a unification constraint \( A_1 = \cdots = A_k \) and \( S :: 'a set set \) represents a set of unification constraints:

```
is_unifier σ A ⇐⇒ |A · σ| ≤ 1
is_unifier σ S ⇐⇒ ∀A ∈ S. is_unifier σ A
is_mgu σ S ⇐⇒ is_unifier σ S ∧ (∀τ. is_unifier τ S ⇒ ∃γ. τ = σ o γ)
```

The next locale, substitution, characterizes the substitution_ops operations using assumptions. A separate locale is necessary because we cannot interleave assumptions and definitions in a single locale. In addition, substitution fixes a function for renaming clauses apart (so that they do not share any variables) and a function that, given a list of atoms, constructs an atom with these as subterms:

```
locale substitution = substitution_ops +
  fixes
    renamings_apart :: 'a clause list ⇒ 's list and
    atm_of_atms :: 'a list ⇒ 'a
  assumes
    A · id = A and
    A · (σ o τ) = (A · σ) · τ and
    (∀A. A · σ = A · τ) ⇒ σ = τ and
    is_ground_cls (C · σ) ⇒ ∃τ. is_ground τ ∧ C · τ = C · σ and
    wf strictly generalizes and
    |renamings_apart Cs| = |Cs|
```
\[ \rho \in \text{renamings\_apart} \ C_s \implies \text{is\_renaming} \ \rho \ \text{and} \]
\[ \text{var\_disjoint} \ (Cs \cdot \text{renamings\_apart} \ Cs) \ \text{and} \]
\[ \text{atm\_of\_atms} \ As \cdot \sigma = \text{atm\_of\_atms} \ Bs \iff \text{map} \ ((\lambda A. A \cdot \sigma) As) Bs = Bs \]

The above definition is presented to give a flavor of our development. We refer to the Isabelle theory files for the precise definitions. Inside the locale, we prove further properties of the substitution\_ops operations. Notably, we prove well-foundedness of the strictly\_subsumes predicate based on the well-foundedness of strictly\_generalizes, which is stated as an assumption. The atm\_of\_atms operation is needed for encoding a clause into a single atom in this well-foundedness proof.

Finally, a third locale, mgu, extends substitution by fixing a function mgu :: ‘a set set ⇒ ‘s option that computes an MGU \( \sigma \) given a set of unification constraints. If a unifier exists, it returns Some \( \sigma \); otherwise, it returns None.

4 BACHMAIR AND GANZINGER’S PROVER

The formalization by Schlichtkrull et al. [2018a,b] of a nondeterministic ordered resolution prover presented by Bachmair and Ganzinger [2001] forms layer 1 of our refinement. Resolution is first defined on ground terms and proved sound and complete with respect to a propositional semantics. First-order ordered resolution is then defined and proved sound, and the ground completeness result is lifted to obtain completeness of the first-order resolution prover. The resolution inference rule is \( n \)-ary, with an optional “selection” mechanism to guide the proof search. In this paper, we disable selection and hence only need to consider the binary case, which can be implemented efficiently and forms the basis of modern provers such as E, SPASS, and Vampire.

The ordered resolution calculus is parameterized by a total order \( > \) ("larger than") on atoms. The ground version of the calculus consists of the single inference rule

\[
\frac{C \lor A \lor \cdots \lor A \neg A \lor D}{C \lor D}
\]

where atom \( A \) must be larger than all the atoms in clause \( C \) and larger than or equal to all the atoms in clause \( D \). The side condition is not necessary for soundness, but it rules out many unnecessary inferences, thereby pruning the search space. Because clauses are defined as multisets, the order of the literals in a clause is immaterial; \( \neg A \lor B \) and \( B \lor \neg A \) are the same clause.

For first-order logic, the order on atoms \( > \) is extended to an order \( > \) on nonground atoms so that \( B > A \) if and only if for all ground substitutions \( \sigma \), we have \( B \cdot \sigma > A \cdot \sigma \). The nonground version of the calculus consists of the single inference rule

\[
\frac{C \lor A_1 \lor \cdots \lor A_k \neg A \lor D}{(C \lor D) \cdot \sigma}
\]

where \( \sigma \) is the (canonical) MGU that solves the unification problem \( A_1 \overset{\gamma}{\equiv} \cdots \overset{\gamma}{\equiv} A_k \overset{\gamma}{\equiv} A \), each \( A_i \cdot \sigma \) is strictly \( > \)-maximal with respect to the atoms in \( C \cdot \sigma \), and \( A \cdot \sigma \) is \( > \)-maximal with respect to the atoms in \( D \cdot \sigma \). An important detail is that to achieve completeness, the rule must be adapted slightly to rename apart the variables occurring in different premises.

Resolution works by saturation. A set of clauses \( D \) is saturated if any conclusion from premises in \( D \) is already in \( D \). The ordered resolution calculus is refutationally complete, meaning that any unsatisfiable saturated set of clauses necessarily contains \( \bot \).

Resolution provers exploit the calculus’s completeness in the following way. They start with a finite set of initial clauses—the input problem—and successively add conclusions from premises in the set. If the inference rule is applied in a fair fashion on the available clauses, the set reaches saturation at the limit; if the set is unsatisfiable, this means \( \bot \) is eventually derived, after finitely many steps. Crucially, not only do efficient provers add clauses to their working set, they also
remove clauses that are deemed redundant. This requires a refined notion of saturation. We call a set of clauses \( D \) saturated up to redundancy, formally saturated_upto \( D \), if any inference from nonredundant clauses in \( D \) yields a redundant conclusion.

Bachmair and Ganzinger’s nondeterministic first-order prover, called RP, captures the “dynamic” aspects of saturation. It builds on the first-order ordered resolution rule and a redundancy criterion. The redundant clauses are those that are tautological (i.e., clauses of the form \( C \lor A \lor \neg A \)) and those that are subsumed by another clause in the working set—for example, the clauses \( B \) and \( A \lor B \) are both subsumed if the working set already contains \( B \). Furthermore, RP takes advantage of subsumption resolution, which can be expressed as an inference rule:

\[
\frac{D \lor L \quad C \lor (D \lor \neg L) \cdot \sigma}{C \lor D \cdot \sigma}
\]

The conclusion subsumes the second premise, which may therefore be deleted.

The RP prover is defined as an inductive predicate \( \leadsto \) on states, where a state is a triple \( S = (N, P, O) \) of new clauses \( N \), processed clauses \( P \), and old clauses \( O \). Initially, \( N \) is the input problem (including the negated conjecture), and \( P \cup O \) is empty. Clauses can be removed if they are tautological or subsumed or after subsumption resolution has been applied. When all clauses in \( N \) have been processed (either removed entirely or moved to \( P \)), a clause from \( P \) can be chosen for inference computation: This clause is moved to \( O \), and all its conclusions with premises from the other old clauses are introduced to form the new \( N \).

In Isabelle, an inductive predicate is specified as a set of Horn-style introduction rules, as in Prolog, but with the conclusion on the right. RP is defined as follows:

\[
\text{inductive } \leadsto :: \text{'a state } \Rightarrow \text{'a state } \Rightarrow \text{bool where}
\begin{align*}
\text{Neg } A \in C \land \text{Pos } A \in C & \Rightarrow (N \cup \{C\}, P, O) \leadsto_1 (N', P, O) \\
| D \in P \cup O \land \text{subsumes } D \in C & \Rightarrow (N \cup \{C\}, P, O) \leadsto_2 (N', P, O) \\
| D \in N' \land \text{strictly_subsumes } D \in C & \Rightarrow (N', P \cup \{C\}, O) \leadsto_3 (N', P, O) \\
| D \in N' \land \text{strictly_subsumes } D \in C & \Rightarrow (N', P, O \cup \{C\}) \leadsto_4 (N', P, O) \\
| D \in P \cup O \land \text{reduces } D \in C \land L & \Rightarrow (N' \cup \{C \cup \{L\}\}, P, O) \leadsto_5 (N' \cup \{C\}, P, O) \\
| D \in N' \land \text{reduces } D \in C \land L & \Rightarrow (N', P \cup \{C \cup \{L\}\}, O) \leadsto_6 (N', P \cup \{C\}, O) \\
| D \in N' \land \text{reduces } D \in C \land L & \Rightarrow (N', P \cup \{C \cup \{L\}\}) \leadsto_7 (N', P \cup \{C\}, O) \\
| (N \cup \{C\}, P, O) & \leadsto_8 (N', P \cup \{C\}, O) \\
| (\{\}, P \cup \{C\}, O) & \leadsto_9 (\text{concl_of } \text{`infers_between } O \land C, P, O \lor \{C\} \text{})
\end{align*}
\]

Subscripts on \( \leadsto \) identify the rules. The notation \( f ' X \) stands for the image of the set (or multiset) \( X \) under function \( f \), \text{infers_between } O \land C \) calculates all the inferences whose premises are a subset of \( O \cup \{C\} \) that contains \( C \), and reduces \( D \lor C \land L \Leftrightarrow \exists D' \lor \sigma \). \( D = D' \lor \{L\} \land \neg L = L' \lor \sigma \land D' \lor \sigma \subseteq C \).

\textbf{Example 4.1.} There are many ways to derive \( \bot \) from the unsatisfiable clause set \( \{p(x), \neg p(a) \lor \neg p(b)\} \). The derivation on the left-hand side below relies on the two mandatory rules (rules 8 and 9). On the right-hand side, we show a shorter derivation that exploits reduction and subsumption to
avoid performing resolution inferences. In both cases, we assume a reasonable term order.

\[
\begin{align*}
\sim_8 ((\neg p(a) \land p(x), \neg p(f(x), y) \lor p(x, y)), \{\}, \{\})
\end{align*}
\]

\[
\sim_9 ((\neg p(a), p(x, f(x), y)) \lor p(x, y), \{\}, \{\})
\]

\[
\sim_8 ((\neg p(a, a), p(x, f(x)), y)) \lor p(x, y), \{\}, \{\})
\]

\[
\sim_9 ((\neg p(a), p(x, f(x))) \lor p(x, y), p(x, x))
\]

\[
\sim_8 ((\neg p(a, a), p(x, f(x))))
\]

\[
\sim_8 \ldots
\]

We can leave \(\neg p(a, a)\) in \(\mathcal{P}\) forever and always generate more clauses of the form \(p(x, i^i(x))\), for increasing values of \(i\). This emphasizes the importance of employing a fair strategy for moving clauses from \(\mathcal{P}\) to \(\mathcal{O}\).

Formally, a derivation is a possibly infinite sequence of states \(S_0 \leadsto S_1 \leadsto S_2 \leadsto \cdots\). In Isabelle, this is expressed by the codatatype of lazy lists:

\[
\begin{align*}
\text{codatatype} \quad 'a \text{ list} = \\
| \text{LNil} \quad | \text{LCons} \quad 'a \quad (\text{a list})
\end{align*}
\]

Lazy list operation names are prefixed by an \(L\) or \(l\) to distinguish them from the corresponding operations on finite lists. For example, \(\text{lhd} \; xs\) yields \(xs\)'s head (if \(xs \neq \text{LNil}\)), and \(\text{lnth} \; xs \; i\) yields the \((i + 1)\)st element of \(xs\) (if \(i < |xs|\)).

We capture the mathematical notation \(S_0 \leadsto S_1 \leadsto S_2 \leadsto \cdots\) formally as chain \((\leadsto)\) \(Ss\), where \(Ss\) is a lazy list of states and chain is a coinductive predicate:

\[
\begin{align*}
\text{coinductive} \quad \text{chain} :: (\text{a} \Rightarrow 'a \Rightarrow \text{bool}) \Rightarrow 'a \text{ list} \Rightarrow \text{bool} \quad \text{where}
\end{align*}
\]

\[
\begin{align*}
\text{chain} \; R \; (\text{LCons} \; x \; \text{LNil})
\end{align*}
\]

\[
\begin{align*}
\text{chain} \; R \; x \; \land \; R \; x \; (\text{lhds} \; xs) \implies \text{chain} \; R \; (\text{LCons} \; x \; xs)
\end{align*}
\]

Coinduction is used to allow infinite chains. The base case is needed to allow finite chains. Chains cannot be empty.

Another important notion is that of the limit of a sequence \(xs\) of sets. It is defined as the set of elements that are members of all positions of \(xs\) except for an at most finite prefix:

\[
\begin{align*}
\text{definition} \quad \text{Liminf} :: 'a \text{ set} \; \text{list} \Rightarrow 'a \text{ set} \quad \text{where}
\end{align*}
\]

\[
\begin{align*}
\text{Liminf} \; xs = \bigcup_{i < |xs|} \bigcap_{j \leq i < |xs|} \text{lnth} \; xs \; j
\end{align*}
\]

Liminf and other operators working on clause sets are lifted pointwise to states. For example, the limit of a sequence of states is defined as \(\text{Liminf} \; Ss = (\text{Liminf} \; Ns, \text{Liminf} \; Ps, \text{Liminf} \; Os)\), where
\[^{Ns, Ps, and Os are the projections of the \(N\), \(P\), and \(O\) components of \(S\). For the rest of this section, we assume that \(S\) is a derivation.

The soundness theorem states that if \(RP\) derives \(\bot\) (represented by the multiset \(\{\}\)) from a set of clauses, that set must be unsatisfiable:

\[
\text{theorem } \text{RP}_\text{sound}:
\{\} \in \text{Liminf } S \Rightarrow \neg \text{satisfiable } (\text{grounding}_\text{of } (\text{lhds } S))
\]

A stronger, finer-grained notion of soundness relates models before and after a transition:

\[
\text{theorem } \text{RP}_\text{model}:
\quad S \leadsto S' \Rightarrow (I \models \text{grounding}_\text{of } S' \iff I \models \text{grounding}_\text{of } S)
\]

The canonical way of expressing the unsatisfiability of a set or multiset of first-order clauses with respect to Herbrand interpretations is as the unsatisfiability of its grounding.

Completeness of the prover can only be guaranteed when its rules are executed in a fair order, such that clauses do not get stuck forever in \(N\) or \(P\). Accordingly, fairness is defined as \(\text{Liminf } N_s = \text{Liminf } P_s = \{\}\). The completeness theorem states that the limit of a fair derivation is saturated:

\[
\text{theorem } \text{RP}_\text{saturated}_\text{if}_\text{fair}:
\quad \text{fair } S \Rightarrow \text{saturated}_\text{upto } (\text{Liminf } (\text{grounding}_\text{of } S))
\]

In particular, if the initial problem is unsatisfiable, \(\bot\) must appear in the \(O\) component of the limit of any fair derivation:

\[
\text{corollary } \text{RP}_\text{complete}_\text{if}_\text{fair}:
\quad \text{fair } S \land \neg \text{satisfiable } (\text{grounding}_\text{of } (\text{lhds } S)) \Rightarrow \{\} \in O\_\text{of } (\text{Liminf } S)
\]

5 ENSURING FAIRNESS

The second refinement layer is the prover \(RP_w\), which ensures fairness by assigning a weight to every clause and by organizing the set of processed clauses—the \(P\) component of a state—as a priority queue, where lighter clauses are chosen before heavier clauses. By assigning heavier weights to newer clauses, we can guarantee that all derivations are fair.

Another necessary ingredient for completeness is that derivations must be complete; for example, the incomplete derivation consisting of the single state \(\{(C), \{\}, \{\}\}\) is not fair because \(C\) is never processed. This requirement is expressed formally as \(\text{full}_\text{chain } (\leadsto_w) S\), where the \(\text{full}_\text{chain}\) predicate is defined coinductively as

\[
\text{coinductive } \text{full}_\text{chain} :: ('a \Rightarrow 'a \Rightarrow \text{bool}) \Rightarrow 'a \text{ llist} \Rightarrow \text{bool where}
\]
\[
\forall y. \neg R x y \Rightarrow \text{full}_\text{chain} R (\text{LCons } x \text{ LNil})
\]
\[
\land \text{full}_\text{chain} R xs \land R x (\text{lhds } xs) \Rightarrow \text{full}_\text{chain} R (\text{LCons } x \text{ xs})
\]

and characterized by the equivalence

\[
\text{lemma } \text{full}_\text{chain}_\text{iff}_\text{chain}:
\quad \text{full}_\text{chain} R xs \iff \text{chain } R xs \land (\text{finite } xs \Rightarrow \forall y. \neg R (\text{lhd } xs) y)
\]

For the rest of this section, we fix a full chain \(S\) such that \(P\_\text{of } (\text{lhds } S) = O\_\text{of } (\text{lhds } S) = \{\}\).

Because each \(RP_w\) rule corresponds to an \(RP\) rule, it is straightforward to lift the soundness and completeness results from \(RP\) to \(RP_w\). The main difficulty is to show that the priority queue ensures fairness of full derivations, which is needed to obtain an unconditional completeness theorem for \(RP_w\), without the assumption fair \(S\).

5.1 Definition

The weight of a clause \(C\), which defines its priority in the queue, may depend both on the clause itself and on when it was generated. As a result, the \(RP_w\) prover represents clauses by a pair \((C, i)\), where \(i\) is the timestamp—the larger the timestamp, the newer the clause. A state is now a quadruple...
\[ S = (N, P, O, t), \] where the first three components are finite multisets and \( t \) is the timestamp to assign to the next generation of clauses. Formally, we have the following type abbreviations:

- **type_synonym** \( \text{wclause} = \text{a clause} \times \text{nat} \)
- **type_synonym** \( \text{wstate} = \text{wclause multiset} \times \text{wclause multiset} \times \text{wclause multiset} \times \text{nat} \)

We extend the `FO_resolution_prover` locale, in which \( \text{RP} \) is defined, with a weight function that, for any given clause, is strictly monotone with respect to the timestamp, so that older copies of a clause are preferred to newer ones:

- **locale** `weighted_FO_resolution_prover = FO_resolution_prover`
  - **fixes** `weight : 'a wclause ⇒ nat`
  - **assumes** `i < j ⇒ weight (C, i) < weight (C, j)`

The \( \text{RP}_w \) prover uses \( 'a wclause \) for clauses. It is defined inductively as follows:

**inductive** \( \sim_w :: 'a wstate ⇒ 'a wstate ⇒ \text{bool} \) where

\[
\begin{align*}
\text{Neg} A \in C \land \text{Pos} A \in C & \implies (N \uplus \{(C, i)\}, P, O, t) \sim_{w_1} (N, P, O, t) \\
D \in \text{fst} \ ('(P \uplus O) \land \text{subsumes} D C) & \implies (N + \{(C, i)\}, P, O, t) \sim_{w_2} (N, P, O, t) \\
D \in \text{fst} \ ('N \land C) \in \text{fst} \ 'P \land \text{strictly_subsumes} D C & \implies (N, P, O, t) \sim_{w_3} (N, \{\{E, k\} \in P. E \neq C\}, O, t) \\
D \in \text{fst} \ 'N \land \text{strictly_subsumes} D C & \implies (N, P, O \uplus \{(C, i)\}, t) \sim_{w_4} (N, P, O, t) \\
D \in \text{fst} \ '((P \uplus O) \land \text{reduces} D C) L & \implies (N \uplus \{(C \uplus \{L, i\}\}, P, O, t) \sim_{w_5} (N \uplus \{(C, i)\}, P, O, t) \\
D \in \text{fst} \ 'N \land \text{reduces} D C L & \land \bigvee (j. (C \uplus \{L\}, j) \in P \implies j \leq i) & \implies (N, P \uplus \{(C \uplus \{L\}, i)\}, O, t) \sim_{w_6} (N, P \uplus \{(C, i)\}, O, t) \\
D \in \text{fst} \ 'N \land \text{reduces} D C L & \implies (N, P, O \uplus \{(C \uplus \{L\}, i)\}, t) \sim_{w_7} (N, P \uplus \{(C, i)\}, O, t) \\
\langle \forall (D, j) \in P. \text{weight} (C, i) \leq \text{weight} (D, j) \rangle \land \\
N = \text{mset_set} ((\lambda D. (D, t)) \text{ concl_of } \text{ infers_between (set_mset (fst O)) C}) & \implies (\{\}, P \uplus \{(C, i)\}, O, t) \sim_{w_9} (N, \{\{D, j\} \in P. D \neq C\}, O \uplus \{C, i\}, t + 1)
\end{align*}
\]

where \( \text{fst} \) is the function that returns the first component of a pair, \( \text{mset_set} \) converts a set to the multiset with exactly one copy of each element in the set, and \( \text{set_mset} \) converts a multiset to the set of elements in the multiset. Each \( \text{RP}_w \) rule \( i \) corresponds to \( \text{RP} \) rule \( i \).

\( \text{RP}_w \) uses finite multisets for representing \( N, P, \) and \( O \). They offer a compromise between the layer 1 representation as sets and the layer 3 implementation as lists. Finite multisets help eliminate some unfair derivations:

- The finiteness condition guarantees that each clause in \( N \) gets the opportunity to move to \( P \) (and further to \( O \)).
- The set-based \( \text{RP} \) allows idle transitions, such as \((N \cup \{C\}, P, O) \sim (N, P \cup \{C\}, O)\) where \( C \in N \cap P \). The use of multisets and \( \uplus \) precludes such steps in \( \text{RP}_w \).

In the inductive definition of \( \text{RP}_w \), the last rule, which computes inferences, assigns timestamp \( t \) to each newly computed clause \( D \) and increments \( t \). Since we want \( P \) to work as a priority queue, we let the prover choose a clause \( C \) with the smallest weight.

Timestamps are preserved when clauses are moved between \( N, P, \) and \( O \). They are also preserved by reduction steps (rules 5 to 7), even though reduction alters the clauses by removing unnecessary literals. This works because reduction can only happen finitely many times—a \( k \)-literal clause can be reduced at most \( k \) times. Therefore, there is no danger of divergence due to an infinite chain of reductions. Incidentally, it would also be possible to assign the current \( t \) as the reduced clause’s
This demonstrates how working out the details and making all assumptions explicit using a proof assistant can help us clarify fine theoretical points.

To prevent this behavior, the RP_w rules are formulated so that whenever they remove the earliest copy of any clause \( C \in P \), they also remove all its copies from \( P \). This property is captured by the following lemma, which is proved by case distinction on the rules:

\[
\begin{align*}
\text{lemma preserve_min}_P: \\
S \sim_w S' \land (C, i) \in P_\text{of } S \land (\forall k. (C, k) \in P_\text{of } S \Rightarrow k \geq i) \land C \in \text{fst } P_\text{of } S' \Rightarrow \\
(C, i) \in P_\text{of } S'
\end{align*}
\]

This completes our review of RP_w. As an intermediate step towards a more concrete prover, we restrict the weight function to be a linear equation that considers both timestamps and clause sizes:

\[
\text{locale weighted_FO_resolution_prover_with_size_timestamp_factors = FO_resolution_prover +}
\]

\[
\begin{align*}
| | :: 'a \Rightarrow \text{nat and} \\
\text{size_factor :: nat and} \\
\text{timestamp_factor :: nat} \\
\text{assumes} \\
\text{timestamp_factor > 0}
\end{align*}
\]

\[
\begin{align*}
\text{fun weight :: 'a wclause \Rightarrow nat where} \\
\text{weight } (C, i) = \text{size_factor } \times |C| + \text{timestamp_factor } \times i
\end{align*}
\]

where \( |\text{C}| = \sum_{A: A \in C} \forall A \in C |A| \). It is easy to prove that this definition of weight is strictly monotone and hence that this locale is a sublocale of \text{weighted_FO_resolution_prover}. This gives us a correspondingly specialized version of RP_w that will form the basis of further refinement steps.

The idea of organizing \( P \) as a priority queue is well known in the automated reasoning community. It is mentioned in a footnote in Bachmair and Ganzinger [2001, p. 44], but they require their weight function to be monotone not only in the timestamp but also in the clause size, claiming that this is necessary to ensure fairness. Although it often makes sense to prefer small clauses to large ones, our proof reveals that clause size is irrelevant for fairness, even in the presence of reductions. This demonstrates how working out the details and making all assumptions explicit using a proof assistant can help us clarify fine theoretical points.
Example 5.1. The following derivation, based on the function weight \((C, i) = |C| + i\), follows the second derivation of Example 4.1:

\[
\begin{align*}
\sim_{w_8} & \{\{\neg p(a) \lor \neg p(b), 0\}, \{p(x), 0\}, \{2\} \}\, (\forall x) \\
\sim_{w_8} & \{\{\neg p(a) \lor \neg p(b), 0\}, \{(p(x), 0), \neg p(a) \lor \neg p(b), 0\}, \{2\} \}
\end{align*}
\]

Due to the weight function, the clause \(p(x)\) must be moved from \(P\) to \(O\) before \(\neg p(a) \lor \neg p(b)\).

5.2 Refinement Proofs

To lift the soundness and completeness results about RP to RP\(_w\), we must first show that any possible behavior of RP\(_w\) on states of type \(wstate\) is a possible behavior of RP on the corresponding values of type \(state\), without timestamps. Formally:

**Lemma weighted\(_{RP}\)_imp\(_{RP}\):**

\(S \sim_w S' \implies state\_of\ S \sim state\_of\ S'\)

The proof is by straightforward induction on the introduction rules of RP\(_w\), with one difficult case. Inference computation (rule 9) converts a set to a finite multiset using \(mset\_set\). This operation is undefined for infinite sets. Thus, we must show that from a finite set of clauses, only a finite set of inferences may be performed by \(\text{inference}\_between\):

**Lemma finite\(_{ord}\)_FO\(_\)_resolution\(_\)inferences\(_\)between:**

\(\text{finite } D \implies \text{finite } (\text{inference\_between } D \ C)\)

Our formal proof caters for \(n\)-ary resolution, but in our application we only need the binary case. A binary resolution inference takes two premises, of the form \(CAA = C \lor A_1 \lor \cdots \lor A_k\) and \(DA = \neg A \lor D\), and produces a conclusion \(E = (C \lor D) \cdot \sigma\). It can be represented compactly by a tuple of the form \((CAA, DA, AA, A, E)\), where \(AA = A_1 \lor \cdots \lor A_k\). We must show that the set of such tuples returned by \(\text{inference\_between}\) is finite, assuming \(D\) is finite.

First, observe that the \(E\) component of a tuple is fully determined by the other four components. Hence it suffices to consider tuples of the form \((CAA, DA, AA, A, A)\). Let \(DC = D \cup \{C\}\), and let \(n\) be the length of the longest clause in \(DC\). Moreover, let \(A = \bigcup_{D \in DC} \text{atms\_of } D \) and \(AA = \{B \mid \text{set\_mset } |B| \leq n\} \). Then all inferences between \(D\) and \(C\) belong to \(DC \times DC \times AA \times A\), which is a cartesian product of finite sets.

5.3 Soundness and Completeness Proofs

Using the refinement theorem, it is easy to lift the \(RP\_model\) theorem (Section 4) to RP\(_w\):

**Theorem weighted\(_{RP}\)_model:**

\(S \sim_w S' \implies (I \models \text{grounding\_of } S' \iff I \models \text{grounding\_of } S)\)

Completeness is considerably more difficult. We first show that the use of timestamps ensures that all full RP\(_w\) derivations are fair. From this fact follows unconditional completeness.

In principle, a full derivation could be unfair by virtue of being finite and ending in a state such as \(\mathcal{N}\) or \(P\) is nonempty. However, this is impossible because a transition of rule 8 or 9 could then

be taken from the last state, contradicting the hypothesis that the derivation is full. Hence, finite full derivations are necessarily fair:

**Lemma** fair_if_finite:

\[ \text{finite } S_S \Rightarrow \text{fair (lmap state_of } S_S) \]

There are two ways in which an infinite derivation \( S_S \) in \( \text{RP}_w \) could be unfair: A clause could get stuck forever in \( N \), or in \( P \). We show that the \( N \) case is impossible by defining a measure on states that decreases with respect to the lexicographic extension of \( > \) on natural numbers to pairs, which is a well-founded relation. The measure is

**Abbreviation** \( \text{RP}_\text{basic\_measure} :: \{ 'a \text{ wstate} \Rightarrow \text{nat}^2 \} \) where

\[
\text{RP}_\text{basic\_measure} (N, P, O, t) \equiv (\text{sum } ((\lambda(C, _). \text{ |C| } + 1) \cdot (N \cup P \cup O)), \text{ |N|})
\]

The first component of the pair is the total size of all the clauses in the state, plus 1 for each clause that ensures that empty clauses are counted. The second component is the number of clauses in \( N \).

It is easy to see why the measure is decreasing. Rule 9, inference computation, is not applicable due to our assumption that a clause remains stuck in \( N \). Rule 8, which moves a clause from \( N \) to \( P \), decreases the measure’s second component while leaving the first component unchanged.

The other rules decrease the first component since they remove clauses or literals. Formally:

**Lemma** weighted_RP_basic_measure_decreasing_N:

\[
S \sim_w S' \land (C, _) \in N_\text{of } S \Rightarrow
(\text{RP}_\text{basic\_measure } S', \text{RP}_\text{basic\_measure } S) \in \text{RP}_\text{basic\_rel}
\]

where \( \text{RP}_\text{basic\_rel} = \text{natLess } <\text{lex}> \text{natLess} \).

What about the case where a clause \( C \) is stuck in \( P \)? Lemma preserve_min_P (Section 5.1) states that in any step, either all copies of a clause \( C \in P \) are removed or the one with minimum timestamp is preserved. It follows that \( C \)'s timestamp will either remain stable or decrease over time. Since \( > \) is well founded on natural numbers, eventually a fixed \( i \) will be reached and will belong to the limit:

**Lemma** persistent_wclause_in_P_if_persistent_clause_in_P:

\[ \exists i. (C, i) \in \text{Liminf } (\text{lmap set_mset } \circ \text{P_of } S_S) \]

Again, we define a measure, but it must also decrease when inferences are computed and new clauses appear in \( N \). (In this case, \( \text{RP}_\text{basic\_measure} \) may increase.) Our new measure is parameterized by a predicate \( p \) that can be used to filter out undesirable clauses:

**Abbreviation** \( \text{RP}_\text{filtered\_measure} :: \{ ('a \text{ wclause } \Rightarrow \text{bool} ) \Rightarrow ('a \text{ wstate } \Rightarrow \text{nat}^2 \} \) where

\[
\text{RP}_\text{filtered\_measure} p (N, P, O, t) \equiv (\text{sum } ((\lambda(C, _). \text{ |C| } + 1) \cdot \{(D \in N \cup P \cup O | p D)\}, \{(D \in N | p D)\}, \{(D \in P | p D)\}))
\]

Notice that \( \text{RP}_\text{filtered\_measure} (\lambda_. \text{ True}) \) essentially amounts to \( \text{RP}_\text{basic\_measure} \). In the formalization, we use \( \text{RP}_\text{filtered\_measure} (\lambda_. \text{ True}) \) to avoid code duplication.

Suppose the clause \( C \) that is stuck in \( P \) has weight \( w \) in the limit, and suppose that a clause \( D \) is moved from \( P \) to \( O \) by the inference computation rule. That clause’s weight must be at most \( w \); otherwise, it would not have been preferred to \( C \).

Infinite derivations necessarily consist of segments each consisting of finitely many applications of rules other than rule 9 followed by an application of rule 9: \( (\sim_{w^1-8} \circ \sim_{w^9})^o \). Since each application of rule 9 increases the \( t \) component of the state, eventually we reach a state in which \( t > w \). As a consequence of strict monotonicity of weight, any clauses generated by inference computation from that point on will have weights above \( C \)'s, and if \( C \) remains stuck, then so must these clauses. Thus, we can ignore these clauses altogether, by using \( \lambda(C, i). i \leq w \) as the filter \( p \).

We adapt the corresponding relation to consider the extra argument:
We add a component $\omega$.

The third refinement layer defines a functional program $\text{RP}_3$ that embodies a specific rule application strategy, thereby eliminating the nondeterminism present in $\text{RP}_w$. Clauses are now represented as lists, and multisets of clauses as lists of lists. Although the program is deterministic, some auxiliary functions are specified mathematically and are not directly executable; making these executable is the objective of the fourth refinement layer (Section 7).

6 ELIMINATING NONDETERMINISM

The third refinement layer defines a functional program $\text{RP}_3$ that embodies a specific rule application strategy, thereby eliminating the nondeterminism present in $\text{RP}_w$. Clauses are now represented as lists, and multisets of clauses as lists of lists. Although the program is deterministic, some auxiliary functions are specified mathematically and are not directly executable; making these executable is the objective of the fourth refinement layer (Section 7).

6.1 Definition

Our prover corresponds roughly to the following pseudocode:

$$
\text{function } \text{RP}_3(N, P, O, t) \text{ is}
\begin{align*}
\text{repeat forever} \quad & \\
\text{if } \bot \in P \cup O \text{ then} & \quad \text{return } P \cup O \\
\text{else if } N = P = \{ \} & \quad \text{return } O \\
\text{else if } N = \{ \} & \quad \\
& \quad \text{let } C \text{ be a minimal-weight clause in } P; \\
& \quad N := \text{conclusions of all inferences from } O \cup \{C\} \text{ involving } C, \text{ with timestamp } t;
\end{align*}
$$
A Verified Automatic Prover Based on Ordered Resolution

move $C$ from $\mathcal{P}$ to $\mathcal{O}$;
$t := t + 1$

else
remove an arbitrary clause $C$ from $\mathcal{N}$;
reduce $C$ using $\mathcal{P} \uplus \mathcal{O}$;
if $C = \bot$ then
  return $\{ \bot \}$
else if $C$ is neither a tautology not subsumed by a clause in $\mathcal{P} \uplus \mathcal{O}$ then
  reduce $\mathcal{P}$ using $C$;
  reduce $\mathcal{O}$ using $C$, moving any reduced clauses from $\mathcal{O}$ to $\mathcal{P}$;
  remove all clauses from $\mathcal{P}$ and $\mathcal{O}$ that are strictly subsumed by $C$;
  add $C$ to $\mathcal{P}$

The function should be invoked with $\mathcal{N}$ as the input problem, $\mathcal{P} = \mathcal{O} = \{ \}$, and an arbitrary timestamp $t$. The loop is loosely modeled after the proof procedure implemented in Vampire [Voronkov 2014, Section 3]. Instead of finite multisets, the actual $RP_d$ definition in Isabelle uses finite lists, bringing us closer to executable code. The # operator abbreviates the Cons constructor, and @ is the append operator.

The list-based representations compel us to introduce the following type abbreviations:

- `type_synonym 'a lclause = 'a literal list`
- `type_synonym 'a dclause = 'a lclause × nat`
- `type_synonym 'a dstate = 'a dclause list × 'a dclause list × 'a dclause list × nat`

A state is a tuple $(\mathcal{N}, \mathcal{P}, \mathcal{O}, t)$ as before, but with different types.

The prover is defined inside a locale that inherits `weighted_FO_resolution_prover_with_size_timestamp_factors`. The core function, $RP_d$ _step_, performs a single iteration of the main loop. Here is the complete definition, excluding auxiliary functions:

```plaintext
fun $RP_d$ _step :: 'a dstate ⇒ 'a dstate where
$RP_d$ _step $(\mathcal{N}, \mathcal{P}, \mathcal{O}, t)$ =
if $\exists C_i \in \mathcal{P} @ \mathcal{O}$. fst $C_i = []$ then
  ([], [], remdups $\mathcal{P} @ \mathcal{O}$, $t + |\text{remdups } \mathcal{P}|)$
else
  (case $\mathcal{N}$ of
    [] ⇒
    (case $\mathcal{P}$ of
      [] ⇒ $(\mathcal{N}, \mathcal{P}, \mathcal{O}, t)$
    | $P_0 \# \mathcal{P}'$ ⇒
      let
        $(C, i) = \text{select_min_weight_clause } P_0 \mathcal{P}'$;
        $\mathcal{N}' = \text{map } (\lambda D. (D, t)) (\text{remdups } (\text{resolve_rename } C C$
          @ concat (map (\text{resolve_rename_either_way } C \circ \text{fst }) \mathcal{O})))$;
        $\mathcal{P} = \text{filter } (\lambda (D, j). \text{mset } D \neq \text{mset } C) \mathcal{P}$;
        $\mathcal{O} = (C, i) \# \mathcal{O}$;
        $t = t + 1$
      in
      $(\mathcal{N}', \mathcal{P}, \mathcal{O}, t)$
    | $(C, i) \# \mathcal{N}$ ⇒
      let
        $C = \text{reduce } (\text{map } \text{fst } (\mathcal{P} @ \mathcal{O})) [\] C$
```
in
if C = [] then
    ([], [], [], (t + 1)
else if is_tautology C ∨ subsume (map fst (P @ O)) C then
    (N, P, O, t)
else
    let
        P = reduce_all C P;
        (back_to_P, O) = reduce_all2 C O;
        P = back_to_P P O;
        O = filter ((¬) o strictly_subsume [C] o fst) O;
        P = filter ((¬) o strictly_subsume [C] o fst) P;
        P = (C, t) # P
    in
    (N, P, O, t)

The code above relies on some nonexecutable constructs, such as the existential quantifier. The quantifier is unproblematic because it ranges over a finite set, but some of the auxiliary functions rely on infinite quantification. Notably, subsumption of D by C is defined as ∃σ. C · σ ⊆ D (Section 3), where σ ranges over all substitutions. Nonexecutable constructs are acceptable if we know that we can replace them by equivalent executable constructs further down the refinement chain; for example, an implementation of subsumption can compute a finite set of candidates for σ using matching, instead of blindly enumerating all possibilities.

The prover’s main program is a tail-recursive function that repeatedly calls RP_d_step until a final state, of the form ([], [], O, t), is reached, at which point it returns O stripped of its timestamps:

**partial_function (option) RP_d :: 'a dstate ⇒ 'a clause list option where**

RP_d S = if is_final S then Some (map fst (O of S)) else RP_d (RP_d_step S)

Since there are no guarantees that the recursion will terminate, we cannot introduce the function using the fun command [Krauss 2006], which is restricted to well-founded recursion. Instead, we use partial_function (option) [Krauss 2010], which puts the computation in an option monad. The function’s result is of the form Some R if the recursion terminates and None if the computation diverges. Executing the function would never actually return None, but it is convenient to define it mathematically in this way. For example, it allows us to state and prove a characterization such as the following, which can be used to replace a terminating call RP_d S by a finite iteration RP_d_step^k S:

**lemma deterministic_RP_SomeD:**

RP_d S = Some R ⇒ ∃S’ k. RP_d_step^k S = S’ ∧ is_final S’ ∧ R = map fst (O of S’)

### 6.2 Refinement Proofs

Using refinement, we connect the RP_d_step function to the RP_w predicate. RP_d_step has a coarser granularity than RP_w: A single invocation on a nonfinal state S can amount to a chain of RP_w transitions. This is captured by the following weak-refinement property:

**lemma nonfinal_deterministic_RP_step:**

¬ is_final S ⇒ wstate_of S wstate_of (RP_d_step S)

where wstate_of converts RP_d states to RP_w states. The entire proof, including key lemmas, is about 1300 lines long. It follows the case distinctions present in the definition of RP_d_step:

**case ∃Ci ∈ P @ O. fst Ci = []**
By induction on |remdups $P$|, there must exist a derivation of the form

$$\text{wstate}_0 (\mathcal{N}, P, O, t) \xrightarrow{w2} \text{wstate}_0 ([], P, O, t)$$

$$\xrightarrow{w9} \text{wstate}_0 (\mathcal{N'}, P', (C, i) \# O, t + 1)$$

$$\xrightarrow{w} \text{wstate}_0 ([], [], \text{remdups } P' \@ O, t + |\text{remdups } P'|)$$

for $P' = \text{filter } (\lambda(D, j). \ \text{mset } D \neq \text{mset } C) \ P$ and suitable $\mathcal{N'}$ and $(C, i) \in \mathcal{P}$. The last step is justified by the induction hypothesis.

**case $\mathcal{N} = \mathcal{P} = []$:**

Contradiction with the assumption that $(\mathcal{N}, P, O, t)$ is a nonfinal state.

**case $\mathcal{N} = []$:**

It suffices to show that the transition

$$\text{wstate}_0 ([], P, O, t) \xrightarrow{w9} \text{wstate}_0 (\mathcal{N'}, P', (C, i) \# O, t + 1)$$

is possible, where $(C, i) \in \mathcal{P}$ is a minimal-weight clause and

$$\mathcal{N'} = \text{map } (\lambda.D. (D, t)) (\text{remdups } (\text{resolve_rename } C C @ \text{concat } (\text{map } (\text{resolve_rename_each_way } C \circ \text{fst } O))))$$

$$\mathcal{P'} = \text{filter } (\lambda(D, j). \ \text{mset } D \neq \text{mset } C) \ P$$

The main proof obligation is that $\mathcal{N'}$, converted to multisets, equals the multiset $\text{mset}_\text{set} (((\lambda D. (D, t)) \bowtie \text{concl}_\text{of} \bowtie \text{infers}_\text{between} (\text{set}_\text{mset } (\text{fst } \setminus O)) C)$ specified in rule $\xrightarrow{w9}$. The distance between the functional program and its mathematical specification is at its greatest here. The proof is tedious but straightforward.

**otherwise:**

Let $C' = \text{reduce } (\text{map fst } P @ \text{map fst } O) [] C$. If $C' = [],$ then

$$\text{wstate}_0 ((C, i) \# \mathcal{N'}, P, O, t)$$

$$\xrightarrow{w5} \text{wstate}_0 ([], [], \text{remdups } P' \@ O, t)$$

$$\xrightarrow{w3} \text{wstate}_0 ([], [], \text{remdups } P' \@ O, t)$$

$$\xrightarrow{w4} \text{wstate}_0 ([], [], \text{remdups } P' \@ O, t)$$

$$\xrightarrow{w5} \text{wstate}_0 (\mathcal{N'}, [], [], [], t)$$

$$\xrightarrow{w2} \text{wstate}_0 ([], [], [], [], t)$$

$$\xrightarrow{w9} \text{wstate}_0 ([], [], [], [], t)$$

Otherwise, if $\text{is_tautology } C' \vee \text{subsume } (\text{map fst } (P @ O)) C'$, then

$$\text{wstate}_0 ((C, i) \# \mathcal{N}, P, O, t)$$

$$\xrightarrow{w5} \text{wstate}_0 ((C', i) \# \mathcal{N}, P, O, t)$$

$$\xrightarrow{w1,2} \text{wstate}_0 (\mathcal{N}, P, O, t)$$
Otherwise:

\[
\begin{align*}
\text{wstate_of } ((C, i) \# \mathcal{N}', \mathcal{P}, \mathcal{O}, t) \\
\sim_{w5} \text{wstate_of } ((C', i) \# \mathcal{N}', \mathcal{P}, \mathcal{O}, t) \\
\sim_{w6} \text{wstate_of } ((C', i) \# \mathcal{N}', \mathcal{P}', \mathcal{O}, t) \\
\sim_{w7} \text{wstate_of } ((C', i) \# \mathcal{N}', \text{back_to}_\mathcal{P} @ \mathcal{P}', \mathcal{O}', t) \\
\sim_{w4} \text{wstate_of } ((C', i) \# \mathcal{N}', \text{back_to}_\mathcal{P} @ \mathcal{P}', \mathcal{O}'', t) \\
\sim_{w3} \text{wstate_of } ((C', i) \# \mathcal{N}', \mathcal{P}'', \mathcal{O}'', t) \\
\sim_{w8} \text{wstate_of } ((\mathcal{N}', (C', i) \# \mathcal{P}'', \mathcal{O}'', t) \\
\end{align*}
\]

for suitable clause lists \(\mathcal{P}', \text{back_to}_\mathcal{P}, \mathcal{O}', \mathcal{O}'', \) and \(\mathcal{P}''\).

The above refinement theorem, about computations from nonfinal states, is complemented by the following trivial result concerning final states:

**Lemma**  
final_deterministic_RP_step:  
is_final \(\mathcal{S}\) \(\Rightarrow\) \(\text{RP}_d\_step \mathcal{S} = \mathcal{S}\)

### 6.3 Soundness and Completeness Proofs

Let \(\mathcal{S}_0 = (\mathcal{N}_0, [], [], \ell_0)\) be an arbitrary initial state. For \(\text{RP}_d\), soundness means that whenever \(\text{RP}_d\) \(\mathcal{S}_0\) terminates with some clause set \(R\), then \(R\) is a saturation that satisfies the same models as \(\mathcal{N}_0\). In addition, if \(\mathcal{N}_0\) is unsatisfiable, then \(R\) contains \(\perp\), which provides a simple syntactic check for unsatisfiability. Completeness means that divergence is possible only if \(\mathcal{N}_0\) is satisfiable. Note that for satisfiable clause sets \(\mathcal{N}_0\), both termination and divergence are possible.

To lift soundness and completeness results from \(\text{RP}_w\) to \(\text{RP}_d\), we first define \(\mathcal{S}s\) as a full chain of nontrivial \(\text{RP}_d\) steps starting from \(\mathcal{S}_0\). Formally, we let \(\mathcal{S}s = \text{derivation}_from \mathcal{S}_0\), with

**primecorec** derivation_from :: ‘a dstate \(\Rightarrow\) ‘a dstate list where
derivation_from \(\mathcal{S}\) = LCons \(\mathcal{S}\) (if is_final \(\mathcal{S}\) then LNil else derivation_from (\(\text{RP}_d\_step} \mathcal{S}\))

Based on \(\mathcal{S}s\), we let \(w\mathcal{S}s = \text{lmap wstate_of } \mathcal{S}s\) and note that \(w\mathcal{S}s\) is a full chain of “big” \(\sim_w\) steps.

Using a lemma that will be proved in Section 6.4, we obtain a full chain \(s w\mathcal{S}s\) of “small” \(\sim_{w}\) steps. This chain satisfies the conditions postulated on \(\mathcal{S}s\) in Section 5.3, allowing us to lift the results presented there.

The soundness results are proved in a nameless locale, or context, that assumes termination:

**context**  
fixes \(R\) :: ‘a lclause list  
assumes \(\text{RP}_d \mathcal{S}_0 = \text{Some } R\)

The definition of \(\text{RP}_d\), using partial_function, gives us an induction rule restricted to the case where \(\text{RP}_d\) terminates (i.e., returns a Some value). This rule can be used to prove that \(\mathcal{S}s\) and hence \(w\mathcal{S}s\) and \(s w\mathcal{S}s\) are finite sequences.

Soundness takes the form of a pair of theorems that lift weighted_RP_model and weighted_RP_saturated:

**Theorem** deterministic_RP_model:  
\(\mathcal{I} \models \text{grounding_of } \mathcal{N}_0 \iff \mathcal{I} \models \text{grounding_of } R\)

**Theorem** deterministic_RP_saturated:  
\(\text{saturated_upto (grounding_of } R)\)

Admittedly, the terminology is somewhat confusing. For \(\text{RP}\) and \(\text{RP}_w\), it is natural—indeed, it is

In most applications, all that matters is the satisfiability status of the set \( \mathcal{N}_0 \). It can be retrieved syntactically:

**corollary** deterministic\_RP\_refutation:
\[ \neg \text{ satisfiable } (\text{grounding}_\text{of } \mathcal{N}_0) \iff \{ \} \in R \]

Completeness is proved in a separate nameless locale that assumes nontermination: \( \text{RP}_d \mathcal{S}_0 = \text{None} \). The strongest result we prove is that this assumption implies the satisfiability of \( \mathcal{N}_0 \):

**theorem** deterministic\_RP\_complete:
\[ \text{ satisfiable } (\text{grounding}_\text{of } \mathcal{N}_0) \]

The proof is by contradiction:

Assume that \( \neg \text{ satisfiable } (\text{grounding}_\text{of } \mathcal{N}_0) \). Hence, by weighted\_RP\_complete we have \( \{ \} \in O_\text{of } ss \text{w}s \) s. It is easy to show that \( ss \text{w}s \) s’s limit is a subset of \( w \text{S}s \)'s limit; hence \( \{ \} \in O_\text{of } w \text{S}s \). This implies the existence of a natural number \( k \) such that \( \{ \} \in O_\text{of } (\text{lnth } w \text{s}s \ k) \). Hence \( \{ \} \in O_\text{of } (\text{RP}_d \text{step}^k \mathcal{S}_0) \). However, by an induction on \( k \), we can show that \( \text{RP}_d \) must terminate after at most \( k \) iterations, contradicting the assumption that \( \text{RP}_d \) diverges.

### 6.4 A Coinductive Puzzle

A single “big” step of the deterministic prover \( \text{RP}_d \) may consist of multiple “small” steps of the weighted prover \( \text{RP}_w \). To transfer the results from \( \text{RP}_w \) to \( \text{RP}_d \), we must expand \( \text{RP}_d \)'s big steps. The core of the expansion is an abstract property of chains and a relation’s transitive closure:

Let \( R \) be a relation and \( \mathbf{x} \) a chain of \( R^+ \) transitions. There exists a chain of \( R \) transitions that embeds \( \mathbf{x} \) — i.e., that contains all elements of \( \mathbf{x} \) in the same order and with only finitely many elements inserted between each pair of consecutive elements of \( \mathbf{x} \).

On finite chains, this property would follow by straightforward induction. But the completeness proof must also consider infinite chains. To prove the property on infinite chains requires us to use coinduction and corecursion up-to techniques.

The desired property is formalized as follows:

**lemma** chain\_tranclp\_imp\_exists\_chain:
\[ \text{chain } R^+ \mathbf{x} \implies \exists \mathbf{y}. \text{chain } R \mathbf{y} \land \mathbf{x} \subseteq \mathbf{y} \land \text{lhd } \mathbf{x} = \text{lhd } \mathbf{y} \land \text{llast } \mathbf{x} = \text{llast } \mathbf{y} \]

where the embedding \( \subseteq \) of lazy lists is defined coinductively using the function \( + \), which prepends a finite list to a lazy list:

**coinductive** \( \subseteq \) ‘a list ⇒ ’a list ⇒ bool where
\[ \begin{align*}
\text{finite } \mathbf{x}s \Rightarrow \text{LNil} \subseteq \mathbf{x}s \\
| \mathbf{x}s \subseteq \mathbf{y}s \Rightarrow \text{LCons } \mathbf{x}s \mathbf{x} \subseteq \mathbf{z}s \Leftrightarrow \text{LCons } \mathbf{x} \mathbf{y}s \\
\end{align*} \]

**fun** \( + \) ‘a list ⇒ ’a list ⇒ ’a list where
\[ \begin{align*}
[] + \mathbf{x}s &= \mathbf{x}s \\
(z \# \mathbf{z}s) + \mathbf{x}s &= \text{LCons } z (\mathbf{z}s + \mathbf{x}s) \\
\end{align*} \]

The definition of \( \subseteq \) ensures that infinite lazy lists only embed other infinite lazy lists, but not the finite ones. Formally: \( \mathbf{x}s \subseteq \mathbf{y}s \Rightarrow (\text{finite } \mathbf{x}s \Leftrightarrow \text{finite } \mathbf{y}s) \). The unguarded calls to \( \text{llast} \) may seem worrying, but the function is conveniently defined to always return the same unspecified element for infinite lists (i.e., \( \neg \text{finite } \mathbf{x}s \land \neg \text{finite } \mathbf{y}s \Rightarrow \text{llast } \mathbf{x}s = \text{llast } \mathbf{y}s \)).

To prove chain\_tranclp\_imp\_exists\_chain, we instantiate the existential quantifier by the following corecursively defined witness:

**corec** \( \text{wit} \) ‘a ⇒ ’a ⇒ bool ⇒ ’a list ⇒ ’a list where
\[ \text{wit } R \mathbf{x}s = (\text{case } \mathbf{x}s \text{ of} \]

\begin{verbatim}
LCons x (LCons y ys) \Rightarrow LCons x (\text{pick } R x y +\!+ \text{wit } R xs) | _ \Rightarrow xs
\end{verbatim}

Here pick \( R x y \) returns an arbitrary finite list of \( R \)-related intermediate states connecting the \( R^+ \)-related states \( x \) and \( y \). Formally,

\[
\text{pick } R x y = \text{SOME } zs. \text{ chain } R (\text{list_of } (x \# zs @ [y]))
\]

where \text{list_of} converts finite lists into lazy list and \text{SOME} is Hilbert’s choice operator. Thus, pick satisfies the characteristic property \( R^+ \) \( x y \Rightarrow \text{chain } R (\text{list_of } (x \# \text{pick } R x y @ [y])) \). The use of Hilbert choice makes pick, and \text{wit}, nonexecutable. This is acceptable because these constants are used only in the proofs and not in the actual prover’s code.

The definition of \text{wit} is not primitivesly coreductive. Although there is a guarding \text{LCons}-constructor, the corecursive call occurs under ++, which makes the productivity of this function subtle. This syntactic structure of the definition is called \text{corecursive up to ++}. Ultimately, \text{wit} is productive because ++ does not remove any \text{LCons} constructors from its second arguments. A slightly weaker requirement, called \text{friendliness}, is supported by Isabelle’s \texttt{corec} command [Blanchette et al. 2017]. Hence, ++ must be registered as a “friend,” which involves a one-line proof, for the above definition to be accepted by Isabelle.

The four conjuncts in \text{chain_tranclp_imp_exists_chain} are then discharged separately under the common assumption \( R^+ \) \( xs \). In increasing difficulty: \( \text{lhd } (\text{wit } R xs) = \text{lhd } xs \) follows by simple rewriting. Next, \text{llast } (\text{wit } R xs) = \text{llast } xs \) requires an induction in the case of finite chains \( xs \). For any infinite chain \( zs \) of type ‘a \text{llist}, \text{llast } zs \) is defined as a fixed but not further specified value of type ‘a. The properties \( xs \subseteq \text{wit } R xs \) and \text{chain } R (\text{wit } R xs) \) require a coinduction on \( \subseteq \) and chain, respectively. In keeping with the definitional principle of corecursion up to ++, plain coinduction on \( \subseteq \) and chain does not suffice and we must use coinduction up to ++ on \( \subseteq \) and chain. We contrast the coinduction (left) and coinduction up to ++ (right) rules for chain:

\[
\forall xs. P \Rightarrow (\exists z. xs = \text{LCons } z \text{ Nil}) \lor \\
(\exists z zs. xs = \text{LCons } z zs \land \\
P zs \land R z (\text{lhd } zs)) \\
\forall xs. P \Rightarrow (\exists z zs ys. xs = \text{LCons } z (ys ++ zs) \land \\
P zs \land \text{chain } R (z \# ys @ [\text{lhd } zs]) \\
\forall xs. P \Rightarrow \text{chain } R xs \\
\forall xs. P \Rightarrow \text{chain } R xs
\]

The property \text{chain_tranclp_imp_exists_chain} easily extends to full chains (because the last element in the case of finite chains remains unchanged), as required in Section 6.3.

7 OBTAINING EXECUTABLE CODE

Our deterministic prover RP\(_d\) is already quite close to being an executable program. There are two main ingredients missing: a concrete representation of terms, over which we have abstracted so far, and an executable algorithm for clause subsumption.

7.1 First-Order Terms

First-order terms are a core data structure in various fields of computer science, be it logic, rewriting, (tree) automata theory, or programming languages (as types of the simply typed λ-calculus). It should be no surprise that various formalizations of terms exist. We instantiate our abstract notion of atoms using a particularly comprehensive formalization of terms developed as part of the IsaFoR library [Thiemann and Sternagel 2009]. This rewriting-independent part of IsaFoR has recently migrated to the Archive of Formal Proofs [Sternagel and Thiemann 2018].

IsaFoR terms are defined as the following datatype:
To simplify notation, in this paper we fix \( f = \nu = \text{nat} \) and abbreviate \((f, \nu) \text{ term}\) by \text{term}.

In the formalization, polymorphic types are used whenever possible. IsaFoR also defines the standard monadic term substitution \( \cdot : \text{term} \Rightarrow (\nu \Rightarrow \text{term}) \Rightarrow \text{term} \) and a function \text{unify} : \((\text{term} \times \text{term}) \Rightarrow \text{lsubst} \Rightarrow \text{lsubst} \), where \text{lsubst} = \((\nu \times \text{term}) \text{ list}\) is the list-based representation of a finite substitution. The function \text{unify} computes the MGU for a list of unification constraints that is compatible with a given substitution. IsaFoR includes a wealth of theorems about the defined functions, including the correctness of unify and the well-foundedness of strict term generalization > : \text{term} \Rightarrow \text{term} \Rightarrow \text{bool} defined by \( t > s \iff (\exists \sigma. \ s \cdot \sigma = t) \land (\exists \sigma. \ t \cdot \sigma = s) \).  

This infrastructure allows us to conveniently instantiate our locales \text{substitution_ops}, \text{substitution}, and \text{mgu}. We instantiate the type \( 'a \) of atoms with \text{term} and the type \( '\nu \) of substitutions with \( '\nu \Rightarrow \text{term} \) and the constants \( \cdot, \text{id}, \circ, \) and \text{atm_of_atms} with \( \cdot, \text{Var}, \lambda \sigma \tau. \sigma \tau, \) and \text{Fun} 0, respectively. (The function symbol name 0 is arbitrary.) For the computation of the MGU, there is a slight type mismatch: IsaFoR offers a list-based unifier, whereas our locale requires the type \text{term set set} \Rightarrow ('\nu \Rightarrow \text{term}) \text{ option}. It is easy to translate a finite set of finite sets of terms (where the inner sets of terms are the ones to be unified) into a finite list of pairs of constraints. To be executable, the translation requires us to sort the terms contained in a set with respect to some arbitrary (but executable) linear order on terms.

Only the function \text{renamings_apart} was not present in IsaFoR. We supply this definition:

\begin{verbatim}
  datatype ('f, 'nu) term =
      Var 'nu
    | Fun 'f ((('f, 'nu) term) list)

  fun renamings_apart :: term clause list \Rightarrow ('nu \Rightarrow \text{term}) list where
    renamings_apart [] = []
  | renamings_apart (C # Cs) =
      let
        \sigma s = renamings_apart Cs;
      \sigma = \lambda \nu. \nu + max \{0\} \cup \text{vars_clause_list} (Cs \cdot \sigma) + 1
      in \sigma \# \sigma s

  where \text{vars_clause_list} :: term clause list \Rightarrow 'nu set returns the variables contained in a list of clauses. The creation of fresh variable names relies on \( '\nu = \text{nat} \).

Finally, the \text{FO_resolution_prover} locale further requires that the type of atoms supports two comparison operators: a well-order \( > \) and a comparison \( > \) that is stable under substitution (i.e., \( B > A \Rightarrow B \cdot \sigma > A \cdot \sigma \)). Moreover, \( > \) and \( \geq \) must coincide on ground atoms. Our approach is to instantiate \( > \) with the Knuth–Bendix order (KBO) [Knuth and Bendix 1970] on terms, which is formalized in IsaFoR [Sternagel and Thiemann 2013]. KBO is executable, stable under substitution, well-founded, and total on ground terms. The well-order \( > \), which must be total on \text{all} terms, is then defined as an arbitrary extension of a partial well-founded order \( > \) to a well-order, using Hilbert choice. This makes \( > \) nonexecutable, which is unproblematic given that \( > \) is used only in proofs and not in the actual prover’s code (which relies on \( > \)).

Working with different orders poses a slight technical challenge in Isabelle. Orders are organized as type classes, which are comfortable to work with as they hide the order assumption. However, a type class can be instantiated with a concrete order at most once—in our case by \( > \). This instantiation propagates to subsequent definitions, such as sorting or computing the minimum. To use a different order for sorting, we must resort to lower-level definitions that are explicitly parameterized by the comparison operation. This is inconvenient when we are defining programs and even more so when we are reasoning about them.
7.2 Clause Subsumption

The second hurdle concerns clause subsumption. Its mathematical definition, subsumes \( C D \iff \exists \sigma. \ C \cdot \sigma \subseteq D \), involves an infinite quantification ranging over substitutions.

The problem of deciding whether such a substitution exists is NP-complete [Kapur and Narendran 1986]. We start with the following naive code. In contrast to the mathematical definition, which operates on multisets of literals, our function operates on lists:

\[
\text{fun subsumes_list :: term literal list } \Rightarrow \text{term literal list } \Rightarrow \text{osubst } \Rightarrow \text{bool }
\]

\[\text{where subsumes_list } [ ] \text{ Ks } \sigma = \text{True} \]

\[\text{subsumes_list } (L \# Ls) \text{ Ks } \sigma = \]

\[(\exists K \in \text{set Ks}. \text{is_pos } K = \text{is_pos } L \land \]

\[\text{case match_term_list } [(\text{atm_of } L, \text{atm_of } K)] \text{ } \sigma \text{ of} \]

\[\text{None } \Rightarrow \text{False} \]

\[\mid \text{Some } \rho \Rightarrow \text{subsumes_list } Ls \text{ (remove1 } K \text{ Ks } \rho) \]

In the type declaration, osubst abbreviates \( \forall \Rightarrow \text{term option} \). The function recurses on its first argument. In the recursive case, we must consider all possible matching literals for \( L \) from \( Ks \) compatible with the substitution \( \sigma \). The bounded existential quantification that expresses this nondeterminism can be executed by iterating over the finite list \( Ks \). The functions is_pos and \( \text{atm_of} \) are the discriminator and selector for literals. The function match_term_list is provided by IsaFoR. It attempts to extend a given substitution into Some matcher for a list of matching constraints, given as term pairs. If the extension is impossible, match_term_list returns None.

This substitution-passing style is typical of purely functional implementations of unification and matching procedures and is inherited by our subsumes_list.

It is easy to prove that the above executable function correctly implements clause subsumption:

\[\text{subsumes } (\text{mset } Ls) \text{ (mset } Ks \text{)} = \text{subsumes_list } Ls \text{ Ks } \lambda x. \text{None}, \]

where mset converts lists to multisets by forgetting the order of the elements. After the registration of this equation, Isabelle’s code generator will rewrite any code that contains the nonexecutable left-hand side to use the executable right-hand side instead.

Clause subsumption is a hot spot in a resolution prover. This has led to the empirical studies of various heuristics to improve on the naive exhaustive search [Schulz 2013a; Tammet 1998]. Following Tammet [1998], we implement a heuristic that often reduces the number of calls to match_term_list, which is linear in the size of the input terms, by first performing a simpler, imprecise comparison. For example, terms with different root symbols will never match, and these can be compared in constant time. Similarly, literals with opposite polarities cannot match.

Accordingly, we sort our (list-represented) clauses with respect to a literal quasi-order (i.e., a transitive and reflexive relation) leq_lit such that

\[\text{is_pos } L = \text{is_pos } K \land \text{match_term_list } [(\text{atm_of } L, \text{atm_of } K)] \text{ } \sigma = \text{Some } \rho \Rightarrow \text{leq_lit } L K \]

Any quasi-order satisfying this property can be used in a refinement of subsumes_list to remove too small literals (with respect to leq_lit), as highlighted in gray below:

\[
\text{fun subsumes_list’ :: term literal list } \Rightarrow \text{term literal list } \Rightarrow \text{osubst } \Rightarrow \text{bool }
\]

\[\text{where subsumes_list’ } [ ] \text{ Ks } \sigma = \text{True} \]

\[\mid \text{subsumes_list’ } (L \# Ls) \text{ Ks } \sigma = \]

\[\text{let Ks } = \text{filter } (\text{leq_lit } L) \text{ Ks in} \]

\[(\exists K \in \text{set Ks}. \text{is_pos } K = \text{is_pos } L \land \]

\[\text{case match_term_list } [(\text{atm_of } L, \text{atm_of } K)] \text{ } \sigma \text{ of} \]

\[\text{None } \Rightarrow \text{False} \]

\[\mid \text{Some } \rho \Rightarrow \text{subsumes_list’ } Ls \text{ (remove1 } K \text{ Ks } \rho) \]

The theorem subsumes_list $Ls Ks \rho = \text{subsumes_list}'$ (sort leq_lit $Ls$) $Ks \rho$ allows the code generator to refine the unoptimized version. In our prover, we let leq_lit be a quasi-order that considers negative literals smaller than positive ones, that considers variables smaller than nonvariable terms, and that sorts terms according to a total order on their root symbols.

This refinement is a local optimization: It requires us to explicitly sort one of the input clauses. A more efficient but also more intrusive refinement would be to maintain the invariant that all clauses in the prover’s state are sorted with respect to leq_lit. Sorting $Ls$ for each invocation of clause subsumption could then be avoided, and filtering $Ks$ could be performed more efficiently. However, maintaining the invariant would require changes throughout the prover’s code.

7.3 The End Result

Finally, Isabelle can export our prover to Standard ML, Haskell, OCaml, or Scala. The command

```ml
export_code prover in SML module_name RP
```

generates a Standard ML module containing the implementation of our prover in slightly more than 1000 lines of code, including dependencies. The generated module exports the ML function

```ml
val prover : ((nat, nat) term literal list * nat) list -> bool
```

Even though in Isabelle we have proved that for any unsatisfiable input prover will terminate and return False, the code generator guarantees only partial correctness of its output: If the generated program terminates on the ML input generated from the Isabelle term $t$ and evaluates to the Boolean result $b$, the proposition $t = b$ is provable in Isabelle. (There is recent work towards providing stronger guarantees [Hupel and Nipkow 2018].) By soundness, we also know that the Boolean $b$ indicates the satisfiability of the input clause set.

After we have worked hard to obtain an executable prover, it would be a shame not to run it on some example. We selected benchmark MSC015 from the TPTP library [Sutcliffe 2017], a particularly challenging family $\Phi_n$ of first-order problems. Each problem consists of the following $n + 2$ clauses (2 unit clauses and $n$ two-literal clauses):

\[
\neg p(b, \ldots, b) \quad p(a, \ldots, a) \\
\neg p(a, b, \ldots, b) \lor p(b, a, \ldots, a) \\
\neg p(x_1, a, b, \ldots, b) \lor p(x_1, b, a, \ldots, a) \\
\vdots \\
\neg p(x_1, \ldots, x_{n-2}, a, b) \lor p(x_1, \ldots, x_{n-2}, b, a) \\
\neg p(x_1, \ldots, x_{n-2}, x_{n-1}, a) \lor p(x_1, \ldots, x_{n-2}, x_{n-1}, b)
\]

A comment in the benchmark warns us that back in 2007, no prover could solve the $\Phi_{23}$ within an hour. Even in 2018, only one prover solves $\Phi_{22}$ within 300 s, and four provers solve $\Phi_{20}$ within 300 s. Our verified prover solves $\Phi_{20}$ in 100 s and $\Phi_{22}$ in 200 s. Although our prover cannot yet challenge state-of-the-art provers in general, its performance is respectable and could be improved further using refinement.

8 DISCUSSION AND RELATED WORK

We found Bachmair and Ganzinger’s [2001] chapter and its formalization by Schlichtkrull et al. [2018a,b] suitable as a starting point for a verified prover. Nonetheless, we faced some difficulties, notably concerning the identification of suitable refinement layers. We developed layers 2, 3, and 4 largely in parallel, with each of the authors working on a separate layer. Bringing layer 2 into a state such that it both ensures fairness and could be refined further by layer 3 required several iterations.
Stepwise refinement helped us achieve separation of concerns: fairness, determinism, and executability were achieved successively. Another strength of refinement is that it allows us to prove results at a high level of abstraction; for example, the fairness of layer 2 is inherited by layers 3 and 4 and could be inherited by further layers. The main weakness of refinement is that some nontrivial machinery is necessary to lift results from one layer to the next. We believe the gain in modularity makes this worthwhile.

It took us quite some time to design a suitable measure to prove the fairness of the layer 2 prover $RP_w$. Our solution amounts to advancing to a state carrying a suitably high timestamp and filtering out all overly heavy clauses. Initially, our proof consisted of two steps—advancing and filtering—each with its own measure. This proof gave us the insurance that $RP_w$ was fair, but we found that combining the measures is both more succinct and more intelligible.

The main goal of our formalization effort was not to obtain a “QED” as quickly as possible but to investigate how to harness a modern proof assistant to formalize the metatheory of automatic theorem provers. We found Isabelle suitable for this verification task. The Isar proof language allows us to state key intermediate steps, as in a paper proof. Standard tactics, including Isabelle’s simplifier, can be used to discharge proof obligations. The Sledgehammer tool [Paulson and Blanchette 2012] uses superposition provers and SMT (satisfiability-modulo-theories) solvers to swiftly identify which lemmas are necessary to prove a goal; standard Isabelle tactics are then used to certify the proof. Isabelle’s support for coinductive methods, including the \texttt{coinductive}, \texttt{codatatype}, and \texttt{corec} commands, helps reason about infinite processes. Locales are a useful abstraction for defining the refinement layers. And the libraries included in the Isabelle distribution, the Archive of Formal Proofs, and the third-party IsaFoR certainly saved us months of work.

The Archive also includes a refinement framework [Lammich 2013], which has been used in a separate effort to connect the imperative code of an efficient SAT solver to an abstract calculus [Blanchette et al. 2018]. The framework is helpful in a variety of situations, including when the refinement relation between a concrete and an abstract data representation is not a function. But since converting a list to a multiset (between our levels 3 and 2) or a multiset to a set (between levels 2 and 1) is a function, we did not see a need to employ it. Moreover, the framework is currently not designed for refining semidecision procedures, as acknowledged privately by its author, Peter Lammich. We conjecture that its support for separation logic could be useful if we were to refine the prover further to obtain imperative code.

Thanks to the verification, we can trust to a very high extent that our ordered resolution prover is sound and complete. To make the prover’s performance competitive with E, SPASS, and Vampire, we would need to extend the current work along two axes. First, we should use superposition, together with its extensive simplification machinery, as the base calculus. A good starting point would be to apply our methodology to Peltier’s [2016] formalization of (a generalization of) superposition. Given that most of a modern superposition prover’s code consists of heuristics, which are easy to verify, the full verification of a competitive superposition prover appears to be a realistic objective for a forthcoming Ph.D. thesis. Second, the chain of refinement should be continued to cover optimized algorithms and data structures. These could be specified by refining layer 4 further, along the lines of Fleury et al.’s [2018] refinement of an imperative SAT solver.

In computer science, metatheories and implementations are often left unconnected. A metatheory may inspire an implementation, or vice versa, but the connection is rarely made explicit. By formalizing the metatheory, the implementation, and their connection, we can demonstrate not only the implementation’s correctness but also the metatheory’s adequacy for describing potential implementations. In particular, we have now confirmed that Bachmair and Ganzinger [2001] (with the exceptions noted by Schlichtkrull et al. [2018b]) accurately describe the abstract principles of an executable functional prover, even though they provide few details beyond layer 1.
We built our verified prover on Schlichtkrull et al.'s [2018a,b] formalization of ordered resolution. Related efforts, developed using Isabelle/HOL, include Peltier’s [2016] formalization of superposition and Schlichtkrull’s [2018] formalization of unordered resolution. However, these developments only cover logical calculi; had we started with any of them, the first step would have been to define an abstract prover in the style of layer 1 and prove basic properties about it. Another related effort is Hirokawa et al.’s [2017] formalization of ordered completion, which (like ordered resolution) can be regarded as a special case of superposition.

Formalizing a theorem proving tool using a theorem proving tool is a thrilling (if self-referential) prospect for many researchers. An early result is Ridge and Margetson’s [2005] verified first-order prover, based on a sequent calculus for first-order logic without full first-order terms but only variables. Kumar et al. [2016] formalized the soundness of a proof assistant for higher-order logic. Jensen et al. [2018] verified the soundness of a kernel for a proof assistant for first-order logic that includes a tableau prover. There are several verified SAT solvers [Blanchette et al. 2018; Lescuyer 2011; Marić 2008, 2010; Oe et al. 2012; Shankar and Vaucher 2011]. Among these, two implement efficient data structures such as the two watched literals [Blanchette et al. 2018; Oe et al. 2012]. SAT being a decidable problem, termination has been proved for most solvers. First-order logic, on the other hand, is semidecidable, which is partly what makes our present work original. Lifting, via refinement, an abstract completeness result expressed in terms of the limit of a possibly infinite derivation to a possibly nonterminating functional program is something we have not found anywhere in the literature.

A pragmatic approach to combine the efficiency of unverified code with the trustworthiness of verified code consists of checking certificates produced by reasoning tools—e.g., proofs produced by SAT solvers [Cruz-Filipe et al. 2017; Lammich 2017]. Researchers from the first-order theorem proving community are now advocating this approach for their systems as well [Reger and Suda 2017]. An ad hoc version of this approach is used in Sledgehammer and similar tools to reconstruct proofs found by first-order provers [Blanchette et al. 2016; Kaliszyk and Urban 2013].

9 CONCLUSION

Starting from Schlichtkrull et al.’s [2018a,b] abstract specification of an ordered resolution prover, we verified, through a refinement chain, a purely functional prover that uses lists as its main data structure. The resulting program is interesting in its own right and could be refined further to obtain an implementation that is competitive with the state of the art.

The stepwise refinement methodology is a keystone of our approach, and we found it entirely adequate for this kind of work. Each refinement step cleanly isolates concerns, yielding intelligible proof obligations. Refinement also helped us identify an unnecessary assumption in Bachmair and Ganzinger [2001] and generally clarify the argument. Lifting results from one layer to another required some thought, especially the completeness results, which correspond to liveness properties. Having now established a methodology and built basic formal libraries, we expect that verifying other saturation-based provers, using Isabelle/HOL or other systems, will be substantially easier.

ACKNOWLEDGMENTS

Alexander Bentkamp, Mathias Fleury, Andreas Halkjær From, Carsten Fuhs, Peter Lammich, Mark Summerfield, and Jørgen Villadsen suggested many textual improvements. Blanchette has received funding from the European Research Council (ERC) under the European Union’s Horizon 2020 research and innovation program (grant agreement No. 713999, Matryoshka).
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