Formal Software Verification Project Report SAT Solver

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Abstract. The abstract should briefly summarize the contents of the paper in 150-250 words.

1 Introduction

The Boolean Satisfiability Problem (SAT) consists in determining whether, given a formula of propositional logic, there exists a valuation that satisfies it, i.e., a mapping from the unknowns of the formula to boolean values such that the formula holds. Great numbers of problems in computer science can easily be reduced to SAT, and even though SAT is known to be \mathcal{NP} -complete by the Cook-Levin theorem of 1971 [1] and its many proofs since then, there exist many efficient SAT-Solvers for certain classes of formulas.

In this project, we implement a small formally verified SAT-Solver in the CoQ proof assistant for formulas containing conjunctions, disjunctions, implications, and negations. To that end, we start by formalizing the syntax of such formulas. Then, we introduce their semantic interpretation given a specified valuation. Following the implementation of a syntactic optimizer that we show to simplify a formula to its minimal form, we conclude this work by implementing the actual solver based on a brute-force search algorithm and prove it to be both correct and complete.

2 Implementation and Encountered Problems

In this section, we present our implementation choices and compare them to discarded alternatives to justify them. We also describe what difficulties we encountered during development and how we addressed them.

2.1 Syntax

First, we formalize the abstract syntax of boolean formulas containing conjunctions, disjunctions, implications, and negations.

Abstract Syntax. We state an inductive definition of the type form with a set of constructors describing how to build up a formula:

$$p, q ::= x \mid \texttt{true} \mid \texttt{false} \mid p \land q \mid p \lor q \mid p \to q \mid \neg p \tag{1}$$

For identifiers, we introduce the type id with a single constructor Id that wraps a string. We also define an equality function comparing two identifiers and returning true if and only if their contained strings are equal. To ease proof development in the rest of the project, we prove some basic lemmas about the function by case distinction, namely reflexivity, equivalence with propositional equality, and decidability of equality of identifiers.

Concrete Syntax. To make reading and writing formulas easier, we use CoQ's Notations system, as well as a Coercion from an identifier as a formula. The main difficulty lies in assigning precedence levels to the individual constructors that pay respect to the commonly assumed binding of operators. Given the binds stronger relation (>), that is:

$$\neg > \land > \lor > \to \tag{2}$$

For example, $x \vee \neg y \wedge z \to \mathsf{false}$ is interpreted as $(x \vee ((\neg y) \wedge z) \to \mathsf{false}$.

2.2 Semantics

Now that we can write a formula in CoQ, we want to define its semantics, i.e., if we interpret it as true or false.

Valuations. To do so, we need to know which boolean values to replace all identifiers occurring in a formula with. Therefore, we define the type **valuation**. A *valuation* is a function that, being passed an identifier, returns **true** or **false**, or, in other words, a valuation is a total map from identifiers to booleans. Our implementation is analogous to the total map defined by Pierce et al. in [2].

Again, to simplify writing and reading valuations, we introduce a Notation x !-> b;; v to override a valuation v with a new value b for x. At first, we used just a single semicolon, but this caused conflicts with list notations. Additionally, we define an empty valuation to map all identifiers to false.

We also specify some lemmas of [2] for later reasonings. Their proofs rely on the functional extensionality axiom, stating that two functions are equal if and only if their applications to all their possible arguments are equal. It is known to be compatible with CoQ's logical core.

Interpreter. The valuation type allows us to define a recursive interpretation function interp. Applied to a valuation and a formula, it returns true if and only if the formula holds by traversing a formula bottom-up and pattern matching on it. All the necessary functions, such as andb, orb, and negb are already implemented in Coq. They even suffice to compute the result of an implication since $p \to q$ is known to be equivalent to $\neg p \lor q$.

2.3 Optimizer

Sometimes, a formula's interpretation can be derived or, at least, simplified on a purely syntactical level, leaving it in a form that is easier to read and reason upon and marginally reducing the computation effort needed by the interpretation function while preserving the semantics. In this part of the project, we therefore introduce an optimization function optim.

Minimality. A key challenge was formally defining what a simplification is and what property the result of optim should have.

A simplification reduces the applications of the binary operators \land , \lor and \rightarrow to one of their two arguments or an atom, i.e., the boolean values true and false. Additionally, when applied to an atom, the unary operation \neg can be simplified to an atom.

The aim is to leave a formula p in a form that cannot be simplified further on a syntactical level. We assume this to be the case in exactly two mutually exclusive situations:

- -p is an atom, or
- p does not contain any atoms.

When one of the situations applies, we say p is in *minimal form*. Formalizing this definition in CoQ, we require either a fixpoint returning true if and only if a formula does not contain atoms or an inductive proposition. While we initially opted for a fixpoint, proving the correctness of our optimizer later turned out easier with an inductive proposition. A formula does not contain an atom if it is an identifier or if all its subformulas do not contain atoms.

We settle on a set of laws that our optimizer uses and allow it to fulfill the above-mentioned correctness and minimality requirements:

$$\begin{array}{lll} \operatorname{true} \wedge p \equiv p & p \wedge \operatorname{true} \equiv p & \operatorname{false} \wedge p \equiv \operatorname{false} & p \wedge \operatorname{false} \equiv \operatorname{false} & (3) \\ \operatorname{true} \vee p \equiv \operatorname{true} & p \vee \operatorname{true} \equiv \operatorname{true} & \operatorname{false} \vee p \equiv p & p \vee \operatorname{false} \equiv p & (4) \\ \operatorname{true} \rightarrow p \equiv p & p \rightarrow \operatorname{true} \equiv \operatorname{true} & \operatorname{false} \rightarrow p \equiv \operatorname{true} & p \rightarrow \operatorname{false} \equiv \neg p & (5) \\ \neg \operatorname{true} \equiv \operatorname{false} & \neg \operatorname{false} \equiv \operatorname{true} & (6) \end{array}$$

Note that we leave out some further simplification potential. For improved provability, we only concentrate on laws involving at least one atom. One could also go beyond and, e.g., add laws such as $p \land \neg p \equiv \mathtt{false}$ for an arbitrary formula p to this set.

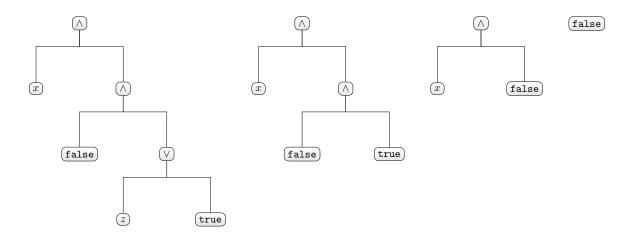


Fig. 1. Successive syntactical optimization steps on $x \land (false \land (z \lor true))$.

Implementation. The actual implementation of the optimizer in CoQ requires a bit of thought. Indeed, our first implementation attempt was a top-down traversal of the abstract syntax tree of the given formula and successive application of the listed simplifications. However, this misses some optimization potential as it may only become available after optimizing subformulas. To try and remedy this, we wrote a recursive function repeatedly applying optim on a formula until reaching a fixpoint. While a fixpoint is eventually reached in practice, CoQ cannot know and rejects the function as new arguments to optim are not obviously smaller.

Instead, our revised approach only traverses the passed formula once and hence only requires linear runtime. We achieve this by performing a post-order depth-first-search (DFS) on the abstract syntax tree of the passed formula. This way, optimizations of subformulas are directly taken into account. Even though sometimes it is sufficient to simplify only one or even no subformula, we always traverse the whole tree for easier proving of the required properties. Figure 1 illustrates such a case, where for the formula $z \wedge (\mathtt{false} \wedge (z \vee \mathtt{true}))$, the subformula $z \vee \mathtt{true}$ is simplified to \mathtt{true} first, even though $\mathtt{false} \wedge (z \vee \mathtt{true})$ could directly be simplified to \mathtt{false} from the get-go. Note, however, that the formula is translated to \mathtt{false} without even running the interpreter just by its syntax.

Properties. To close off this section, we must formally verify that our optimizer truly meets its requirements. First and foremost, the following theorem holds:

Theorem 1. For all valuations v and formulas p, interp v p = interp v (optim p), i.e., the optimizer is correct since it preserves the semantics of formulas.

Proof. We proceed by induction on the structure of p. The non-trivial cases $p = q_1 \land q_2$, $p = q_1 \lor q_2$ and $p = q_1 \to q_2$ are all shown by case distinction on optim q_1 and optim q_2 and application of the induction hypotheses claiming the optimizer is correct for q_1 and q_2 . $p = \neg q$ follows the same pattern for a single subformula q.

The main challenges for formally proving this theorem in CoQ do not lie in the reasoning but rather in removing redundancies. Consequently, as for many other proofs of the project, we decided to first write a mostly manually written version before looking out for automation potential. In a final step, we filtered out common patterns through custom tactics through the Ltac language.

We also prove this second theorem to show our optimizer is actually exhaustive:

Theorem 2. For all formulas p, the result of the optimizer optim p is in minimal form.

Proof. We proceed by induction on the structure of p. For the cases $p = q_1 \land q_2$, $p = q_2 \lor q_2$, $p = q_1 \to q_2$ and $p = \neg q$, we perform case distinctions on the induction hypotheses (minimality of the subformulas), as well as the results of the optimizer applied to them. If necessary, we invert the induction hypotheses involving the deconstructed optimizer results. Then, it follows directly from the hypotheses set that p either does not contain atoms or, on the contrary, is an atom.

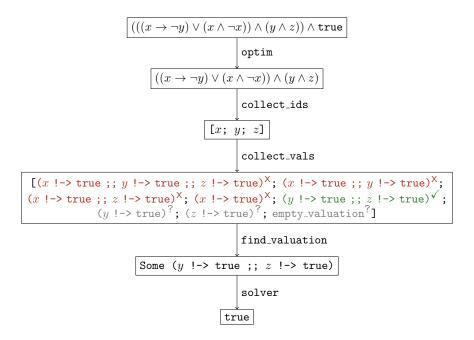


Fig. 2. Exemplary solving process for $(((x \to \neg y) \lor (x \land \neg x)) \land (y \land z)) \land \text{true}.$

For this CoQ proof, many cases are very similar, but, e.g., differ in the actual atom p is equal to, reducing the potential for automation, as one must carefully introduce a fitting witness for the left branch of minimality, i.e., exists b, $p = form_bool$ b. Nevertheless, we factor out the recurrent patterns of deconstructing and rewriting with the induction hypotheses and destructing optim q for a subformula q and then inverting one of the induction hypotheses. In some rare cases, some forms are shelved in the process, requiring the use of Unshelve, which we were unaware of beforehand.

2.4 Solver

In this final section, we can finally deal with our SAT solver. Our idea is to collect representatives for all classes of valuations that are susceptible of satisfying a given formula and test them on the formula until one positive interpretation is found or no valuation matches. Indeed, only identifiers that are in a formula influence its interpretation, and we can safely map all identifiers not contained in a formula to false in our collected valuations:

Lemma 1. For any formula, two otherwise identical valuations differing in the value of an identifier that is not contained in the formula lead to an identical interpretation.

Initially, we wanted to write a function that directly collects all relevant valuations for a given formula. While this is possible, it would have inevitably collected duplicate valuations as comparing valuations is not computable, as valuations are, in turn, functions themselves. Duplicate valuations can, however, lead to a significantly increased runtime of our solver in some cases. For instance, consider the unsatisfiable formula $x \land \neg x$. The initial function would collect $[x !-> true ; x !-> true ; empty_valuation]$. We would then check the valuation x !-> true twice, even though it does not satisfy the formula. Generalized to examples with many duplicate identifiers in respective subformulas, this does not scale.

Instead, we decide to implement our solver with three individual functions. We name them explicitly rather than defining them as anonymous fixpoints and can hence show some detached properties in CoQ. Throughout this section, we will use the formula $(((x \to \neg y) \lor (x \land \neg x)) \land (y \land z)) \land \text{true}$ as running example to illustrate the solving steps as depicted in figure 2.

Extracting Identifiers. To tackle the above-mentioned problem, we provide a dedicated recursive function collect_ids that collects all identifiers of a formula without duplicates.

We wanted the result of the function to be a finite set but did not find a standard library implementation that fully corresponded to our expectations. Thus, we first decided to return a list and take

```
Fixpoint ids_union (11 12 : list id) :=

match 11 with

| | \Rightarrow 12

| x:: xs \Rightarrow if existsb (eqb_id x) 12 then ids_union xs 12 else ids_union xs (x :: 12)
```

List. 1. Custom duplicate free list merging function.

care ourselves to not include any duplicates. The abandoned implementation of a function merging the lists of collect identifiers of two subformulas without duplicates can be found in listing 1. We formally proved that elements included in either 11 or 12 are all kept in the result, that any element of the result is originally contained in 11 or 12, and that if NoDup 11 and NoDup 12, then NoDup (ids_union 11 12) holds.

After some deeper search through the standard library, we, in the end, opted for the finite set implementation provided in $Coq.Lists.ListSet^1$. Even though it does not hide many implementation details, it comes with some predefined set manipulation operations and some lemmas about its properties, which would not have justified a similar custom re-implementation. To confirm it matches our expectations, we show that set_add behaves as the identity function if we try adding an identifier x to a set that already contains it and that it behaves as an append of the one-element list [x] to the end of the set, which in fact is just a list, else.

The most relevant property of collect_ids is however the following:

Lemma 2. An identifier is contained in a formula p if and only if it is contained in collect_ids p.

Collecting Valuations.

2.5 Miscellaneous

Difficulty with Import/Export

2.6 A Subsection Sample

Please note that the first paragraph of a section or subsection is not indented. The first paragraph that follows a table, figure, equation etc. does not need an indent, either.

Subsequent paragraphs, however, are indented.

Sample Heading (Third Level) Only two levels of headings should be numbered. Lower level headings remain unnumbered; they are formatted as run-in headings.

Sample Heading (Fourth Level) The contribution should contain no more than four levels of headings. Table 1 gives a summary of all heading levels.

Table 1. Table caption	ns should be placed above the tables.
nple	Font size and style

0	1	Font size and style
		14 point, bold
1st-level heading	1 Introduction	12 point, bold
2nd-level heading	2.1 Printing Area	10 point, bold
3rd-level heading	Run-in Heading in Bold. Text follows	10 point, bold
4th-level heading	Lowest Level Heading. Text follows	10 point, italic

Displayed equations are centered and set on a separate line.

$$x + y = z \tag{7}$$

Please try to avoid rasterized images for line-art diagrams and schemas. Whenever possible, use vector graphics instead (see Fig. 3).

¹ https://coq.inria.fr/doc/v8.9/stdlib/Coq.Lists.ListSet.html

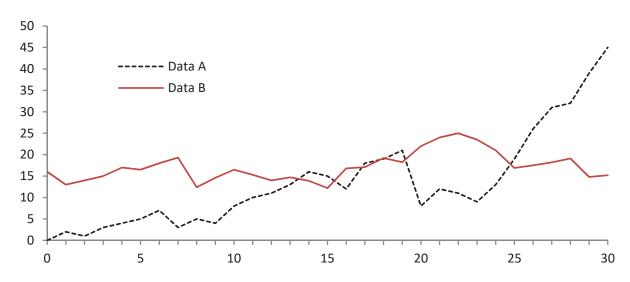


Fig. 3. A figure caption is always placed below the illustration. Please note that short captions are centered, while long ones are justified by the macro package automatically.

Theorem 3. This is a sample theorem. The run-in heading is set in bold, while the following text appears in italics. Definitions, lemmas, propositions, and corollaries are styled the same way.

Proof. Proofs, examples, and remarks have the initial word in italics, while the following text appears in normal font.

3 Conclusion

References

- [1] S. A. Cook. "The complexity of theorem-proving procedures". In: *Proceedings of the Third Annual ACM Symposium on Theory of Computing*. STOC '71. Shaker Heights, Ohio, USA: Association for Computing Machinery, 1971, pp. 151–158. ISBN: 9781450374644. DOI: 10.1145/800157.805047.
- [2] B. C. Pierce, A. A. de Amorim, C. Casinghino, M. Gaboardi, M. Greenberg, C. Hriţcu, V. Sjöberg, and B. Yorgey. *Logical Foundations*. Ed. by B. C. Pierce. Vol. 1. Software Foundations. Version 6.7, http://softwarefoundations.cis.upenn.edu. Electronic textbook, 2024.