A Bisimulation between DPLL(\(T\)) and a Proof-Search Strategy for the Focused Sequent Calculus

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Abstract

We describe how the Davis-Putnam-Logemann-Loveland procedure DPLL is bisimilar to the goal-directed proof-search mechanism described by a standard but carefully chosen sequent calculus. We thus relate a procedure described as a transition system on states to the gradual completion of incomplete proof-trees.

For this we use a focused sequent calculus for polarised classical logic, for which we allow analytic cuts. The focusing mechanisms, together with an appropriate management of polarities, then allows the bisimulation to hold: The class of sequent calculus proofs that are the images of the DPLL runs finishing on UNSAT, is identified with a simple criterion involving polarities.

We actually provide those results for a version DPLL(\(T\)) of the procedure that is parameterised by a background theory \(T\) for which we can decide whether conjunctions of literals are consistent. This procedure is used for Satisfiability Modulo Theories (SMT) generalising propositional SAT. For this, we extend the standard focused sequent calculus for propositional logic in the same way DPLL(\(T\)) extends DPLL: with the ability to call the decision procedure for \(T\).

DPLL(\(T\)) is implemented as a plugin for PSYCHE, a proof-search engine for this sequent calculus, to provide a sequent-calculus based SMT-solver.

*Categories and Subject Descriptors*

F.4.1 [Mathematical Logic]: Mechanical theorem proving

*Keywords*

polarised logic; focused sequent calculus; DPLL(\(T\))

1. Introduction

The sequent calculus is a versatile formalism that can be used to describe goal-directed proof-search, the foundational paradigm of a broad range of tools, from higher-order proof-assistants to logic programming. Not only is the gradual bottom-up construction of proof-trees in sequent calculus the basis of analytic tableaux methods, but it has also been shown to describe mechanisms as diverse as, on the one hand, type inhabitation / proof-construction in (the basic theory of) Coq [24] and, on the other hand, computation in ProLog and \(\lambda\)-ProLog [28].

This paper sets the foundations for applying the same methodology to the automated techniques developed to solve the Satisfiability Modulo Theories (SMT) family of problems, making them available to systems based on goal-directed proof-search.

Such problems generalise propositional SAT-problems: instead of considering the satisfiability of conjunctive normal forms (CNF) over propositional variables, SMT problems are concerned with the satisfiability of CNF over atomic propositions from a theory \(T\) such as linear arithmetic or bit vectors. Given a procedure deciding the consistency -with respect to \(T\)- of a conjunction of atoms or negated atoms, SMT-solving organises a cooperation between this procedure and SAT-solving techniques, thus providing a decision procedure for SMT-problems. This smart extension of the successful SAT-solving techniques opened a prolific area of research and led to the implementation of ever-improving tools, namely SMT-solvers, now crucial to a number of applications in software verification. The architecture of SMT-solvers is based on the extension of the Davis, Putnam, Logemann and Loveland (DPLL) procedure [13, 14] for solving SAT-problems to a procedure called DPLL(\(T\)) [32] addressing SMT-problems.

This paper does not try to improve the DPLL(\(T\)) technique itself, or current SMT-solvers based on it, but makes a step towards the integration of the technique into a sequent calculus framework. A now wide literature achieves the integration of SMT-solvers in various tools, using the blackbox approach. For instance, several proof assistants propose an infrastructure allowing the user to call an external SMT-solver as a blackbox and re-interpret its output to reconstruct a proof within the system [2, 8, 9, 35]. This blackbox infrastructure is natural, given the available tools, their development history, and the communities that designed their techniques.

Here, we aim at a new and deeper integration where DPLL(\(T\)) is performed within the system. Recently, an internal implementation of some SMT-techniques was made available in the Coq proof assistant [25], but is very specific to Coq’s reflection feature [10] (and therefore can hardly be adapted to a framework without reflection).

We rather investigate a broader and more basic context where we can perform each of the steps of DPLL(\(T\)) as the standard steps of proof-search in sequent calculus: the gradual and goal-directed construction of a proof-tree. This allows the DPLL(\(T\)) algorithm to be applied up-to-a-point, where a switch to another technique can be made (depending on the newly generated goals), whereas the use of reflection or of a blackbox call only works when the entire goal can be treated by a (full) run of DPLL(\(T\)).

The results in this paper can be seen as an abstract description of DPLL(\(T\)) that aims at providing a better proof-theoretical understanding of how different theorem proving techniques (e.g. tableaux, resolution, DPLL(\(T\)),...), geared towards different logical fragments, could efficiently cooperate inside the same prover.
This was not explicitly the concern of previous abstract descriptions of DPLL$(T)$ that have been proposed in the literature for the purpose of studying the non-trivial properties of its implementations in proof-theoretical terms. These use resolution trees [7] or most often transition systems [31, 32] based on rewrite rules.

While offering a seemingly convenient way of representing branching (and hence backtracking), the sequent calculus turns out to be a somewhat more rigid setting (than the aforementioned transition systems), tied to the root-first decomposition of formulae trees. Other descriptions of DPLL$(T)$ based on trees (e.g. [34]) offer some work-arounds, but are specifically designed for the job (of describing DPLL$(T)$).

In this paper we show how proof-search in a rather standard sequent calculus, called LK$^p(T)$, can simulate DPLL$(T)$. More precisely, we identify an elementary version of DPLL$(T)$ that is the direct extension of the Classical DPLL procedure to a background theory $T$, as well as being a restriction of the Full DPLL Modulo Theories system, both of which can be found in [32].

For this we do not tailor the sequent calculus to DPLL$(T)$, but we do use well-known sequent calculus features: not only do we allow the use of analytic cuts, but we also use polarities and focusing. Arising from Linear Logic [1, 17], the last two features also make sense in classical logic [18, 22, 26]; while Gentzen’s original rules offer a lot of non-determinism in the proof-search, these features provide a tight control on the breadth of the search space.

Such control allows us to derive a stronger result than the mere simulation of DPLL$(T)$: the proofs in LK$^p(T)$ that are the images of DPLL$(T)$ runs finishing on UNSAT can be characterised by a simple criterion only involving the way polarities are assigned to literals and the way formulae are placed into the focus of sequents (the device implementing focusing). From this criterion we directly get a simple proof-search strategy that is bisimilar to DPLL$(T)$ runs: that which performs the depth-first completion of incomplete proof-trees (starting with the leftmost open leaf), using any inference steps satisfying the given criterion on polarities and focusing.

That way, we ensure that bottom-up proof-search in sequent calculus can be as efficient as the DPLL$(T)$ procedure.

In order to validate the theoretical simulation of DPLL$(T)$, and to evaluate its efficiency as a proof-search method in LK$^p(T)$, we have implemented the simulation as a plugin for PSYCHE [20, 33], a proof-search tool based on LK$^p(T)$. This tool has a modular architecture in a style similar to LCF [29]: a simple kernel offers proof-search primitives for LK$^p(T)$, and plugins are programmed with these primitives to drive the kernel through the construction of a proof-tree. Thanks to this modularity, the correctness of PSYCHE’s output only relies on that of the simple kernel, while efficiency of proof-search is left to the plugin that implements an identified proof-search strategy of interest. We discuss the performance of our DPLL$(T)$ plugin on a standard benchmark.

The paper is organised as follows: section 2 describes the sequent calculus LK$^p(T)$ and section 3 presents the elementary system for DPLL$(T)$; section 4 presents the simulation of that system in LK$^p(T)$; section 5 identifies the corresponding proof-search strategy in LK$^p(T)$ to complete the bisimulation. Finally, section 6 provides an overview of the PSYCHE proof-search tool and of its DPLL$(T)$ plugin, before we discuss related works and perspectives in sections 7 and 8.

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1. that allows more advanced features such as backjumping

2. In brief, the inference rules decomposing the connectives of the same polarity can be chained without losing completeness - see e.g. [1, 28].

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2. Focusing for proof-search

In this section we briefly review the proof-search motivation for focused proof systems, and present the focused sequent calculus LK$^p(T)$ for propositional classical logic modulo a theory.

2.1 Background

At the basis of logic programming, proof-search on Horn clauses can be understood as a meaningful computational paradigm because this class of formulae makes a simple goal-directed proof-search strategy logically complete (with well-identified backtrack points and a reasonably efficient covering of the proof-search space). This still holds when the class is extended to hereditary Harrop formulae [28], and can hold on a wider class of formulae if logical connectives and atoms are tagged with polarities: positive or negative. Polarities emerged with the help of linear logic [17] and Andreoli’s focusing results [1], informally described below:

- Negative connectives can be decomposed, in sequent calculus style, with invertible inference rules that are called asynchronous: a proof-search strategy can perform the bottom-up application of those rules as basic proof-search steps without loss of generality (if the goal was provable, it remains provable after applying the step); in other words, no backtracking is necessary on the application of such steps, even though other steps were possible.

- Positive connectives are the (De Morgan’s) duals of negative connectives, and their decomposition rules, which are called synchronous, are not necessarily invertible.

Clearly, asynchronous rules can be applied eagerly, i.e. can be chained, without creating backtrack points and losing completeness; it turns out that synchronous rules (although possibly creating backtrack points) can also be chained without losing completeness. This result can be expressed as the completeness of a sequent calculus with a focus device, which syntactically highlights a formula in the sequent and forces the next proof-search step to decompose it with a synchronous rule, keeping the focus on its newly-revealed sub-formulae. Focusing considerably reduces the proof-search space, otherwise heavily redundant when Gentzen-style inference rules are used.

A sequent with a positive atom in focus must be proved immediately by an axiom on that atom; hence, the polarity of atoms greatly affects the shape of proofs. As illustrated in e.g. [26], the following sequent expresses the Fibonacci logic program (in some language where addition is primitive) and a goal $\text{fib}(n, p)$ (where $n$ and $p$ are closed terms):

$$\text{fib}(0, 0),$$
$$\text{fib}(1, 1),$$
$$\forall p. \text{fib}(i, p_1) \Rightarrow \text{fib}(i + 1, p_2) \Rightarrow \text{fib}(i + 2, p_1 + p_2)$$
$$\vdash \text{fib}(n, p)$$

The goal will be proved with backward-reasoning if the fib atoms are negative (yielding a proof of exponential size in $n$), and forward-reasoning if they are positive (yielding many proofs, one of which being linear).

In classical logic, polarities of connectives and atoms do not affect the provability of formulae, but still greatly affect the shape of proofs, and hence the basic proof-construction steps. This paper shows how the DPLL$(T)$ steps correspond to proof-construction steps for an appropriate management of polarities. Our focused sequent calculus LK$^p(T)$ for classical logic builds on previous systems based on LC [12, 18, 22], and is syntactically closest to LKF [26], now quite standard.

In order to make logical sense of e.g. the primitive addition in the Fibonacci example above, we only enrich LKF with the ability to call a decision procedure to decide the consistency of conjunctions of literals w.r.t. a theory (i.e. the same as for DPLL$(T)$): for
a theory that equates $1 + 1$ and 2, a call to the procedure proves $p(2), p^+(1 + 1) +$ in one step (unlike LKF’s syntactic checks).

System LKF also assumes that all atoms come with a pre-determined polarity, whereas LK^ evidence allows on-the-fly polarisation of atoms: the root of a proof-tree might have none of its atoms polarised, but atoms may become positive or negative as progress is made in the proof-search.

### 2.2 The LK^ evidence sequent calculus with analytic cuts

This sequent calculus (and this logic) involves a notion of literal and a notion of theory. The reader can safely see behind this terminology the standard notions from proof theory and automated reasoning. However at this point, very little is required from or assumed about those two notions.

**Definition 1 (Literals)**

Let $L$ be a set of elements called literals, equipped with an involutive function called negation from $L$ to $L$. In the rest of this paper, a possibly primed or indexed lowercase letter always denotes a literal, and $l^+$ its negation.

Another ingredient of LK^ evidence is a theory $T$, given in the form of an inconsistency predicate, a notion that we now introduce:

**Definition 2 (Inconsistency predicates)**

An inconsistency predicate is a predicate over sets of literals

- satisfied by the set $\{l, l^+\}$ for every literal $l$;
- that is upward closed (if a subset of a set satisfies the predicate, so does the set);
- such that if the sets $P, l$ and $P, l^+$ satisfy it then so does $P$.

The smallest inconsistency predicate is called the syntactical inconsistency predicate. If a set $P$ of literals satisfies the syntactically inconsistent predicate, we say that $P$ is syntactically inconsistent, denoted $P \vdash$. Otherwise $P$ is syntactically consistent.

The theory $T$ in the notion LK^ evidence is described by means of an (other) inconsistency predicate, called the semantical inconsistency predicate, which will be a formal parameter of the inference system defining LK^ evidence.

If a set $P$ of literals satisfies the semantical inconsistency predicate, we say that $P$ is semantically inconsistent or inconsistent modulo theory, denoted by $P \vdash \bot$. Otherwise $P$ is semantically consistent or consistent modulo theory.

**Definition 3 (Formulæ, negation)**

Let $L$ be a set of literals. The formulæ of propositional polarised classical logic are given by the following grammar:

Formulæ $A, B, \ldots ::= l$ where $l$ ranges over $L$

| $A^\wedge B | A^\vee B | T^+ | \bot^+$
| $A^\wedge B | A^\vee B | T^- | \bot^-$

The size of a formula $A$, denoted $|A|$, is its size as a tree (number of nodes).

Let $P \subseteq L$ be syntactically consistent. Intuitively, it represents the set of literals declared to be positive.

We define $P$-positive formulæ and $P$-negative formulæ as the formulæ generated by the following grammars:

$P$-positive formulæ $P, \ldots ::= p | A^\wedge B | A^\vee B | T^+ | \bot^+$

$P$-negative formulæ $N, \ldots ::= p^+ | A^\wedge B | A^\vee B | T^- | \bot^-$

where $p$ ranges over $P$.

Formulæ that are neither $P$-positive nor $P$-negative are said to be $P$-unpolarised.

Negation is recursively extended into an involutive map on formulæ as follows:

\[
\begin{align*}
(A^\wedge B)^\bot & ::= A^\wedge B^\bot \quad (A^\vee B)^\bot ::= A^\vee B^\bot \\
(T^+) & ::= \bot^- \\
(L^+) & ::= T^+
\end{align*}
\]

**Remark 1** Note that, given a syntactically consistent set $P$ of literals, negations of $P$-positive formulæ are $P$-negative and vice versa.

**Notation 4** A possibly primed or indexed $\Gamma$ always denotes a set of formulæ. By $\Gamma^ $, we denote the subset of elements of $\Gamma$ that are literals, and we write $l \in \Gamma$ if $l$ or $l^+$ appears in $\Gamma$.

The system LK^ evidence is the sequent calculus defined by the rules of Figure 1, which fall into three categories: synchronous, asynchronous, and structural rules, and manipulate two kinds of sequents:

\[ \Gamma \vdash^ P A \]

where the formula $A$ is in the focus of the sequent $\Gamma \vdash^ P \Gamma'$ where $P$ is a syntactically consistent set of literals declared to be positive.

A sequent of the second kind where $\Gamma'$ is empty is called developed.

The gradual proof-tree construction defined by the bottom-up application of the inference rules of LK^ evidence, is a goal-directed mechanism whose intuition can be given as follows:

Asynchronous rules are invertible: $(\wedge^-)$ and $(\vee^-)$ are applied eagerly when trying to construct the proof-tree of a given sequent; (Store) is applied when hitting a positive formula or a negative literal on the right-hand side of a sequent, storing its negation on the left; (Pol) is the on-the-fly polarisation rule, applied on demand, for instance when a right-hand side literal is of undetermined polarity and therefore cannot yet be stored.

When the right-hand side of a sequent becomes empty (i.e. the sequent is developed), a sanity check can be made with (Init2) to check the semantical consistency of the stored literals (w.r.t. the theory), otherwise a choice must be made to place a positive formulæ in focus, using rule (Select), before applying synchronous rules like $(\wedge^+)$ and $(\vee^+)$. Each such rule decomposes the formulæ in focus, keeping the revealed sub-formulæ in the focus of the corresponding premises, until a positive literal or a non-positive formulæ is obtained: the former case must be closed immediately with (Init1) calling the decision procedure, and the latter case uses the (Release) rule to drop the focus and start applying asynchronous rules again. The synchronous and the structural rules are in general not invertible, so each application of those yields in general a backtrack point in the proof-search.

Notice that an invariant of such a proof-tree construction process is that the left-hand side of a sequent only contains negative formulæ and positive literals.

**Notation 6** When $F$ is a formulæ of unpolarised propositional logic and $\Psi$ is a set of such formulæ, $\Psi \vdash F$ means that $\Psi$ entails $F$ in propositional classical logic. Given a theory $T$ (given by a semantical inconsistency predicate), we define the set of all theory lemmas as $\Psi_T := \{l_1 \vee \cdots \vee l_n \mid l_1^+, \ldots, l_n^+ \vdash T\}$ and generalise the notation $\vdash_T$ to write $\Psi \vdash_T F$ when $\Psi_T, \Psi \vdash F$. In that case we say that $F$ is a semantical consequence of $\Psi$. For any polarised formulæ $A, B, \ldots$:

\[\begin{align*}
(A^\wedge B)^\bot & ::= A^\wedge B^\bot \quad (A^\vee B)^\bot ::= A^\vee B^\bot \\
(T^+) & ::= \bot^- \\
(L^+) & ::= T^+
\end{align*}\]

\[\begin{align*}
(A^\wedge B)^\bot & ::= A^\wedge B^\bot \quad (A^\vee B)^\bot ::= A^\vee B^\bot \\
(T^+) & ::= \bot^- \\
(L^+) & ::= T^+
\end{align*}\]
formula $A$, let $\overline{A}$ be the unpolared formula obtained by removing all polarities on connectives.

**Theorem 2 (Cut-elimination and Completeness of $\mathcal{L}K^\mathcal{P}(T)$ [16])**

- The following analytic cut-rule is admissible in $\mathcal{L}K^\mathcal{P}(T)$:

  $\frac{\Gamma \vdash^p \top \quad \Gamma \vdash^p l \quad l \in \Gamma}{\Gamma \vdash^p \top}$ (cut)

- If $\models_{\mathcal{T}} F$, then for all $A$ such that $\overline{A} = F$ and all $\mathcal{P}$, we can prove $\vdash^p A$ in $\mathcal{L}K^\mathcal{P}(T)$.

The meta-theory of $\mathcal{L}K^\mathcal{P}(T)$, in particular the proofs of the above, can be found in [16].

### 3. The elementary DPLL($T$) procedure

Intuitively, DPLL($T$) aims at proving the inconsistency of a set of clauses with respect to a theory. We therefore retain from the previous section the notion of literal and inconsistencies, and introduce clauses:

**Definition 7 (Clause)**

A clause is a finite set of literals, which can be seen as their disjunction.

In the rest of the paper, a possibly indexed uppercased $C$ always denotes a clause. The empty clause is denoted by $\bot$. The number of literals in a clause $C$ is denoted $\sharp(C)$. The possibly indexed symbol $\phi$ always denotes finite sets of clauses $\{C_1, \ldots, C_n\}$, which can also be seen as a Conjunctive Normal Form (CNF). We use $\sharp(\phi)$ to denote the sum of the sizes of the clauses in $\phi$. Finally, $\text{lit}(\phi)$ denotes the set of literals that appear in $\phi$ or whose negations appear in $\phi$.

Viewing clauses as disjunctions of literals and sets of clauses as CNF, we will generalise Notation 6, writing for instance $\phi \models \neg C$ or $\phi \models_{\mathcal{T}} \neg C$, as well as $\phi \models_{\mathcal{T}} C$ or $\phi \models_{\mathcal{T}} C$.

**Definition 8 (Decision literals and sequences)**

We consider a (single) copy of the set $\mathcal{L}$ of literals, denoted $\mathcal{L}^d$, whose elements are called decision literals, which are just tagged clones of the literals in $\mathcal{L}$. Decision literals are denoted by $\top$.

We use the possibly indexed symbol $\Delta$ to denote a finite sequence of possibly tagged literals, with $\bot$ denoting the empty sequence. We also use $\Delta_1$, $\Delta_2$, and $\Delta_1, \Delta_2$ to denote the suggested concatenation of sequences.

For such a sequence $\Delta$, we write $\overline{\Delta}$ for the subset of $\mathcal{L}$ containing all the literals in $\Delta$ with their potential tags removed. The sequences that DPLL($T$) will construct will always be duplicate-free, so the difference between $\Delta$ and $\overline{\Delta}$ is just a matter of tags and ordering. When the context is unambiguous, we will sometimes use $\Delta$ when we mean $\overline{\Delta}$.

We define $\text{Clos}(\Delta) := \{l \mid \Delta, l \vdash_{\mathcal{T}} \top\}$, the closure of a sequence $\Delta$ by semantical entailment. For any set of clauses $\phi$, the set of literals occurring in $\phi$ that are semantically entailed by $\Delta$ is denoted by $\text{Clos}(\phi)(\Delta) := \text{Clos}(\Delta) \cap \text{lit}(\phi)$.

**Remark 3** Semantical consequences are the analogues of the consequences of a partial boolean assignment in the context of a DPLL procedure for propositional logic without theory. Obviously, if $l \in \Delta$, then $l \in \text{Clos}(\Delta)$. If $\phi_1 \subseteq \phi_2$, then for any $\Delta$, $\text{Clos}_{\phi_1}(\Delta) \subseteq \text{Clos}_{\phi_2}(\Delta)$.

We can now describe the elementary DPLL($T$) procedure as a transition system between states.

**Definition 9 (Elementary DPLL($T$))**

A state of the DPLL($T$) procedure is either the state UNSAT, or a pair denoted $\Delta\parallel\phi$, where $\phi$ is a set of clauses and $\Delta$ is a sequence of possibly tagged literals. The transition rules of the elementary DPLL($T$) procedure are given in Fig. 2.

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This exponent tag is a standard notation, standing for "decision".
Decide $\Delta \models \phi \Rightarrow \Delta, l^d \models \phi$ where $l \in \text{lit}(\phi)$ and $l \notin \Delta$ and $l^+ \notin \Delta$.

Propagate $\Delta \models \phi, C \lor l \Rightarrow \Delta, l \models \phi, C \lor l$ where $\Delta \models \neg C$ and $l \notin \Delta$ and $l^+ \notin \Delta$.

Propagate$_T$ $\Delta \models \phi \Rightarrow \Delta, l \models \phi$ where $\Delta \models \neg C$ and there is no decision literal in $\Delta$.

Fail $\Delta \models \phi, C \Rightarrow \text{UNSAT}$.

Fail$_T$ $\Delta \models \phi \Rightarrow \text{UNSAT}$.

Backtrack $\Delta, l^d, \Delta_2 \models \phi, C \Rightarrow \Delta, \Delta_1 \models \phi, C$ where $\Delta, l, \Delta_2 \models \neg C$ and there is no decision literal in $\Delta_2$.

Backtrack$_T$ $\Delta, l^d, \Delta_2 \models \phi \Rightarrow \Delta, l \models \phi$ where $\Delta, l, \Delta_2 \models \neg C$ and there is no decision literal in $\Delta_2$.

Figure 2. Elementary DPLL($T$)

This transition system is an extension of the Classical DPLL procedure, as presented in [32], to the background theory $T$. The first four rules are explicitly taken from the Abstract DPLL Modulo Theories system of [32].$^7$ The other rules of that system (namely $T$-Backjump, $T$-Learn, $T$-Forget, etc.) are not considered here in their full generality, but specific cases and combinations are covered by the rest of our elementary DPLL($T$) system, so that it is logically complete.$^8$ Note that this transition system is not deterministic: for instance the Decide rule can be applied from any state and it furthermore does not enforce a strategy for picking the literal to be tagged among the eligible elements of $\text{lit}(\phi)$. At the level of implementation, this (non deterministic) transition system is turned into a deterministic algorithm, whose efficiency can be transposed into a proof-search process for subsets of $\Delta$.\$^9$

Now, coming back to the intuition of "filling its holes" of the incomplete tree. We should read in complete sub-trees to be constructed. We removed the Pure Literal rule, in general unsound in presence of a theory $T$.\$^9$

Definition 10 (Representation of clauses as formulae)

An LK($T$) formula $C'$ represents a DPLL($T$) clause $\langle l_j \rangle_{j=1 \ldots p}$ if $C' = l_1 \lor \ldots \lor l_p \lor \bot$. A set of formulae $\phi'$ represents a set of clauses $\phi$ if there is a bijection $f$ from $\phi$ to $\phi'$ such that for all clauses $C$ in $\phi$, $f(C)$ represents $C$.$^5$

Remark 4 If $C'$ represents $C$, then $\sharp(C') \leq 2\sharp(C)$ (there are fewer symbols $\lor$ than there are literals in $C$).

Remark 5 The length of $[\Delta]$ is the number of decision literals in $\Delta$.

Now, coming back to the DPLL($T$) transition sequence $\Delta \models \phi \Rightarrow \Delta | \phi$ and its intuitive counterpart in sequent calculus, we have to formalise the notion of incomplete proof-tree together with the notion of "filling its holes":

Definition 11 (Backtrack points)

The backtrack points $[\Delta]$ of a sequence $\Delta$ of possibly tagged literals is the list of sets of untagged literals recursively defined by the following rules, where $[]$ and $::$ are the standard list constructors.

$$
[\langle \rangle] :: := \emptyset
$$

$$
[\Delta, l] :: := [\Delta]
$$

$$
[\Delta, l^d] :: := \Delta, l^+ :: [\Delta]
$$

Remark 5 The length of $[\Delta]$ is the number of decision literals in $\Delta$.
Definition 12 (Incomplete proof-tree, extension)

An incomplete proof-tree in LK^e(T) is a tree labelled with sequents,
- whose leaves are tagged as either open or closed;
- whose open leaves are labelled with developed sequents;
- and such that every node that is not an open leaf, together with its children, forms an instance of the LK^e(T) rules.

The size of an incomplete proof-tree is its number of nodes.

An incomplete proof-tree π is an extension of π, if there is a tree (edge and nodes preserving) homomorphism from π to π'. It is an n-extension of π, if moreover the difference of size between π' and π is less than or equal to n.

Remark 6
An incomplete proof-tree that has no open leaf is (isomorphic to) a well-formed complete LK^e(T) proof of the sequent labelling its root. In that case, we say the proof-tree is complete.

The intuition that an intermediate DPLL(T) state describes an "interface" between an incomplete proof-tree and the complete proof-trees that should be plugged into its holes, is formalised as follows:

Definition 13 (Correspondance)

An incomplete proof-tree π corresponds to a DPLL(T) state ∆ || φ if:
- the length of ∆ :: []∆ is the number of open leaves of π;
- if ∆_i is the i-th element of ∆ :: []∆, then the i-th open leaf of π (taken left-to-right) is labelled by a developed sequent of the form ∆'_i, φ'_i ⊢ ∆_i;
  - φ'_i represents φ (in the sense of Definition 10);
  - Clo_{φ'_i}(∆'_i) = Clo_{φ}(∆_i).

An incomplete proof-tree π corresponds to the state UNSAT if it has no open leaf.

Remark 7
In the general case, different incomplete proof-trees might correspond to the same DPLL(T) state (just like different DPLL(T) runs may reach that state from the initial one).

Note that we do not require anything from the conclusion of an incomplete proof-tree corresponding to ∆ || φ; just as our correspondence says nothing about the DPLL(T) transitions taking place after ∆ || φ (nor about the trees to be plugged into the open leaves), it says nothing about the transitions taking place before ∆ || φ (nor about the incomplete proof-tree, except for its open leaves).

If an incomplete proof-tree π corresponds to a DPLL(T) state ∆ || φ where there are no decision literals in Δ, then there is exactly one open leaf in π, and it is labelled by a sequent of the form ∆', φ' ⊢ ⊥, where φ' represents φ and Clo_{φ'}(Δ') = Clo_{φ}(Δ).

To the initial state ∅ || φ of a run of the DPLL(T) procedure corresponds the incomplete proof-tree consisting of one node (both root and open leaf) labelled with the sequent φ' ⊢ ⊥, where φ' represents φ.

The simulation theorem below provides a systematic way of interpreting any DPLL(T) transition as a completion of incomplete proof-trees that preserves the correspondence given in Definition 13 and controls the growth of the proof trees.

Theorem 8 (Simulation of DPLL(T) in LK^e(T))

If ∆ || φ ⇒ S₂ is a valid DPLL(T) transition, and π₁ is an incomplete proof tree in LK^e(T) corresponding to ∆ || φ, then there exists a (2(∥φ∥ + 3))-extension π₂ of π₁ that corresponds to S₂.

Proof: By case analysis on the nature of the transition, completing the leftmost open leaf of π₁:
- Decide:
  ∆ || φ ⇒ ∆, l⁺ || φ where l /∈ ∆, l⁺ /∈ ∆, l ∈ lit(φ).
  Let π₁ be an incomplete proof-tree corresponding to ∆ || φ. The leftmost leaf (corresponding to ∆) is of the form ∆', φ' ⊢ ⊥ where φ' represents φ and Clo_{φ'}(Δ') = Clo_{φ}(Δ).
  We extend π₁ into π₂ by replacing the leftmost leaf by the following (incomplete) proof-tree:
  \[
  \begin{array}{l}
  \Delta', l, \phi' \vdash \Delta^l \\
  \Delta', l, \phi' \vdash \Delta^l \\
  \Delta', l, \phi' \vdash \Delta^l \\
  \Delta', l, \phi' \vdash \Delta^l \\
  \end{array}
  \]

Note that we use here the analytic cut rule of LK^e(T). π₂ is a 3-extension of π₁ that corresponds to ∆, l⁺ || φ. Indeed, we have ∆, l⁺ || φ = (π₁), l⁺ || φ and Clo_{φ'}(Δ', l⁺) = Clo_{φ}(Δ', l). The two new leaves are tagged as open.

- Propagate:
  ∆ || φ, C ∨ l ⇒ ∆ || φ, C ∨ l where ∆ || φ, C ∨ l is the node with the formula ∆ || φ, C ∨ l
  Let π₁ be an incomplete proof-tree corresponding to ∆ || φ, C ∨ l. The open leaf corresponding to ∆ is of the form ∆', φ' ⊢ ⊥, where φ' represents φ and Clo_{φ'}(Δ', l⁺) =
\[
\begin{align*}
\text{Clo}_{\phi, C \land \phi} (\Delta'). & \quad \text{Let } C = l_1 \lor \ldots \lor l_n. \text{ From } \Delta \models -C \text{ we get } \\
\forall i, l_i^+ \in \Delta \subseteq \text{Clo}_{\phi, C \land \phi} (\Delta) = \text{Clo}_{\phi, C \land \phi} (\Delta'). & \quad \text{We extend } \pi_1 \text{ into } \pi_2 \text{ by replacing this leaf by the following (incomplete) proof-tree:} \\
\Delta', l', C' \models \neg \Delta. & \quad \Delta', l', C' \models \neg \Delta \quad \text{The top-right rules can be applied since } l_i^+ \in \Delta \text{ and } \Delta', l_i \models \neg \Delta. \quad \text{The open leaves are closed.} \\
\Delta', l', C' \models \neg \Delta[l_i^+] & \quad \Delta', l', C' \models \neg \Delta[l_i^+] \quad \text{where } l_i^+ \text{ is the open leaf corresponding to } \Delta_i. \quad \text{We extend } \pi_1 \text{ into } \pi_2 \text{ by replacing the open leaf by the following (incomplete) proof-tree:}
\end{align*}
\]
Corollary 9

If $\exists \phi \models \varphi$, UNSAT and $\varphi'$ represents $\varphi$ then there is a complete proof in LK$^p(T)$ of $\varphi'$, of size smaller than $(2g(\varphi) + 3)n$.

5. Completing the bismulation

Now the point of having mentioned quantitative information in Theorem 8, via the notion of $n$-extension, is to motivate the idea that performing proof-search directly in LK$^p(T)$ is in essence not less efficient than running DPLL(T), which we have a linear bound in the length of the DPLL(T) run and (the proportionality ratio is itself an affine function of the size of the original problem).

We also need to make sure that this final proof-tree is indeed found as efficiently as running DPLL(T), which can be done by identifying, in LK$^p(T)$, a (complete) search space that is isomorphic to (and hence no wider than) that of DPLL(T). We analyse this for a proof-search strategy, in LK$^p(T)$, that exactly captures the proof-extensions that we have used in the simulation of DPLL(T), i.e. the proof of Theorem 8:

Definition 14 (DPLL(T)-extensions)

An incomplete proof tree $\pi_2$ is a DPLL(T)-extension of an incomplete proof tree $\pi_1$ if:

1. it extends $\pi_1$ by replacing its leftmost open leaf with an incomplete proof-tree of one of the forms:
   \[
   \Gamma, A \vdash_p [A] \quad \Gamma, A \vdash_p \Gamma \vdash_p t \quad \Gamma \vdash_p t \quad \Gamma \vdash_p t \quad \Gamma \vdash_p \Gamma_{\text{lit}} \vdash \tau
   \]
   where
   (a) $A$ is a (positive) conjunction of literals that are all in $P$ except maybe one that is $P$-unpolared
   (b) the only instances of (Pol) in the above proof are of the form
   \[
   \Gamma \vdash_p t \quad \Gamma \vdash_p \Gamma_{\text{lit}} \vdash \tau
   \]
   (c) $\Gamma \vdash_p t$ with $\Gamma_{\text{lit}}, t \vdash \tau$
2. any incomplete proof-tree satisfying point 1. and extended by $\pi_2$ is $\pi_2$ itself.

Given a DPLL(T)-extension, we can now identify a DPLL(T) transition that the extension simulates, in the sense of Theorem 8:

Theorem 10 (Simulation of the strategy back into DPLL(T))

If $\pi_2$ is a DPLL(T)-extension of $\pi_1$, and $\pi_1$ corresponds to $\Delta \varphi$, then there is a (unique) DPLL(T) transition $\Delta \varphi \Rightarrow S_2$ such that $\pi_2$ corresponds to $S_2$.

Proof: By case analysis on the shape of the incomplete proof-tree replacing the leftmost open leaf of $\pi_1$. Out of the four shapes of definition 14:

- for the first one: if there is no $P$-unpolared literal in $A$, it is simulated by a Fail or Backtrack (depending on whether $\pi_2$ is complete) on the clause represented by $A$; if not, it is simulated by Propagate on the clause represented by $A$;
- the second one is simulated by Decide on $t$;
- the third one is simulated by Propagate$_\tau$ on $l$;
- the fourth one is simulated by Fail$_\tau$ or Backtrack$_\tau$ (depending on whether $\pi_2$ is complete).

The details are the same as in the proof of Theorem 8.  \[\square\]

If a complete proof-tree of LK$^p(T)$, whose conclusion is an SMT-problem, systematically uses the rules in the way described by the above shapes, then it is the image of a DPLL(T) run.

While it could be envisaged to simulate DPLL(T) in a Gentzen-style sequent calculus (with a variant of Theorem 8), the above definition and theorem reveal the advantage of using a focused sequent calculus for polarised logic: Definition 14 presents, different ways of starting the extension of an open branch (whose leaf sequent is developed), each one of Them corresponding to a specific DPLL(T) transition; then focusing takes care of the following steps of the extension so that, when hitting developed sequents again, the exact simulation of the DPLL(T) transition has been performed.

In order for proof-search mechanisms to exactly match DPLL(T) transitions, focusing therefore provides the right level of granularity and (together with an appropriate management of polarities) the right level of determinism.

Corollary 11 (Bismulation)

The correspondence relation (see definition 13) between incomplete proof trees and DPLL(T) states is a bismulation for the transition system defined on incomplete proof-trees of LK$^p(T)$ by the strategy of DPLL(T)-extensions and on states by DPLL(T).

Finally, obtaining this tight result is the reason why we identified the elementary DPLL(T) system, a restriction of the Abstract DPLL Modal Theory system of [32]:

Modern SMT-solvers feature some mechanisms that are not part of our (logically complete) elementary DPLL(T) system but increase efficiency, such as backjumping and lemma learning (cf. rules $T$-Backjump, $T$-Learn in [32]).

It is possible to simulate those rules in LK$^p(T)$ by using general cuts, by extending with identical steps several open branches of incomplete proof-trees, and possibly by using explicit weakenings (depending on whether we adapt the correspondence between DPLL(T) states and incomplete proof-trees).

But the price of this is high at the theoretical level: With such "parallel extensions" of incomplete proof-trees, it is not clear how to count the sizes of proofs and extensions in a meaningful way, so the quantitative aspects of Theorem 8 and Corollary 9 are compromised; neither is it clear which criterion on proof-trees (and on how to extend them) identifies the proof-construction strategy that is the exact image of a DPLL(T) procedure featuring those advanced mechanisms. In other words, it is not clear how to obtain such a tight correspondence.

6. DPLL(T) in PSIYCHE

The above description of DPLL(T) as a proof-search mechanism of LK$^p(T)$ has been implemented as a plugin for PSIYCHE [20, 33], a proof-search engine for LK$^p(T)$.

6.1 PSIYCHE’s overview

PSYCHE is a proof-search tool for the sequent calculus presented in Section 2, implemented in OCaml. The construction of a proof-tree for a given formula results from the interaction between a small kernel and plugins, which only agree on the data structures used in their interaction. The former is parametrised by a decision procedure for $T$ and implements the rules in LK$^p(T)$ (bottom-up), offering an API to apply synchronous rules (mainly, the choice of focus) while automatically applying asynchronous rules which are invertible, and therefore represent no backtrack point. A plugin implements a strategy that drives the kernel, by calling its API

\[i.e. \text{it corresponds to an initial state of DPLL(T)}\]

\[\text{mostly by specifying the management of polarities}\]
functions, towards either a proof or the guarantee that no proof exists. Plugins can be used to import efficient automated reasoning techniques in this goal-directed framework by specifying in which order (and to which depth) the branches of the search-space should be explored. The implementation of plugins is however not trusted for soundness: the worst that a plugin can do is crash the program, not affect its output. A expanded description of PSYCHE can be found in [20].

6.2 A plugin for DPLL(T)

Version 1.5 of PSYCHE comes with a plugin implementing the simulation of DPLL(T) as described in Section 4. Every time a rule is applied, the plugin calls the API function that takes as input a formula to focus on, feeding it with the appropriate clause. That function also accepts the alternative instruction of making a cut, which is what the plugin uses to simulate Decide. Backtrack and Propagate are done eagerly by the technique of watched literals [30].

As mentioned in the previous section, the theoretical simulation of the full DPLL(T) system, with backjumping and lemma learning requires extending several branches of open proof-trees with parallel steps. To avoid compromising on efficiency, our plugin for PSYCHE implements these advanced features, without departing from the theory described in this paper but using an alternative technique: it simply uses memoisation for the proof-search function. This is used to close, in one single step, any branch that would otherwise be closed by repeating the same steps as in another sub-proof. In particular, doing this avoids repeating, several times, the proof-construction steps of a “parallel extension” corresponding to a single backjump.

Memoisation is also a way of performing clause-learning: Even though our DPLL(T) plugin never actually adds a learnt clause to the original set of clauses (which it could actually do with a general cut), it rather relies on the memoisation table: a learnt clause \( C \) is a clause for which we know that \( \phi \models T C \), and that is made available for Fail, Backtrack or Propagate. Such a clause corresponds to a key \( \phi', C^+ \vdash \) of the memoisation table, with its proof as value. A state where \( C \) can be used for Fail or Backtrack is necessarily a sequent weakening \( C' \vdash \) with extra formulae or literals, so the proof recorded in the memoisation table can be plugged there to close the current branch. When \( C \) can be used for Propagate, it suffices to make a cut on the missing literal: one branch will be closed by plugging-in the proof recorded in the memoisation table, while the other branch will continue the simulation.

The memoisation table is filled-in by clause-learning: our plugin adds an entry whenever it builds a complete proof of some sequent \( \Delta \vdash \) and no previous entry \( \Delta' \vdash \) exists with \( \Delta' \subseteq \Delta \), or whenever it concludes that some sequent \( \Delta \vdash \) is not provable and no previous entry \( \Delta' \vdash \) exists with \( \Delta \subseteq \Delta' \). For the table to cut computation as often as possible, a pre-processing step is applied to a proof-tree before it enters the table: it is pruned from every formula that is not used in the proof, which is easy to do for complete proofs (eager weakening is applied a posteriori by inspection of the inductive structure). PSYCHE’s kernel instead performs pruning on-the-fly, whenever an inference is added to complete proofs. Since proof-completion can be seen as finding a conflict, pruning by eager weakening is a conflict analysis process naturally provided by structural proof theory. Of course, the efficiency of pruning relies on the efficiency of the decision procedure in providing a small inconsistent subset whenever it decides that a set of literals is inconsistent.

6.3 Examples

PSYCHE’s webpage (examples section) shows the output of PSYCHE when it is run, with its DPLL(T) plugin, on the instance that we used in Fig. 3 to illustrate the elementary DPLL(T) system. On such a small example, the \( \text{EPLX} \) source produced by PSYCHE can be compiled, produced in pdf format the proof-tree in LK\( ^n \) (that corresponds to the second run of Fig. 3). The treating bigger problems, we have evaluated the efficiency of PSYCHE on two standard benchmarks for automated decision procedures. The first set of problems is composed of purely propositional formulae (SAT problems) and is a subset of the Satisfiability Library (SATLIB) benchmark [21]. The second set of problems is composed of quantifier-free problems in the theory of linear rational arithmetic (QLRA) and is a fragment of the Satisfiability Modulo Theory Library (SMTLIB) benchmark [3]. All the results can be found on the PSYCHE website [33]. Comparing the current implementation of PSYCHE with state-of-the-art SAT and SMT solvers would make little sense: the current version of PSYCHE plugin does not incorporate techniques that are now folklore for SAT solvers, nor any kind of pre-processing. Moreover the decision procedure used by version 1.5 of PSYCHE for linear arithmetic is a very naive version of the simplex algorithm, and most importantly it does not implement the incremental version of this algorithm which is best suited to the needs of an SMT solver. We however believe that the performance results obtained by PSYCHE are promising and we plan to enrich its collection of plugins in order to benefit at least from pre-processing [15] and restart policies [19]. We also plan to interface PSYCHE with decision procedures for a broader range of theories, in order to be able to cover more benchmarks from the SMTLIB.

7. Related Work

In this paper we used the focused sequent calculus LK\( ^n \) (for polarised classical logic, which considerably narrows the search space provided by Gentzen’s sequent calculus. As already mentioned in Section 2.1, LK\( ^n \) is a variant of LKF [26], enriched with the ability to polarise atoms on-the-fly, and of course the ability to call a decision procedure as in DPLL(T).

We also allow an analytic cut-rule, notwithstanding that proof-search in sequent calculus is usually done in a cut-free system, as the cut-rule usually unreasonably widens the search-space. Analytic cuts only concern atomic cut-formulae among the finitely many atoms present in the rest of the sequent. Allowing them does widen the search space (which we narrowed in other ways) but this sometimes permits to draw quicker conclusions, in a way similar to DPLL(T)’s Decide rule. Miller and Nigam [27] have already shown how to use analytic cuts to incorporate tables into proofs, i.e. make sure that, once an atom is proved or known to be true (from a table of lemmas), the subsequent proof-search never tries to re-prove it. This is achieved by giving, to the atom that is cut, two opposite polarities in the two premises of the cut. The simulation of DPLL(T)’s Decide rule in LK\( ^n \) uses the same trick.

Moreover, their approach seems to have strong links with the memoisation table that our PSYCHE implementation uses to avoid re-proving sequents. So far, our memoisation table is not reflected in the sequent calculus itself, but we could envisage adapting their approach, possibly capturing the interaction with the memoisation table by the use of general cuts. Proof-search may then depart from the mere gradual construction of a proof-tree. But an appealing idea is that the cleverness that goes into finding a Theory lemma (i.e. a key of the memoisation table) translates as the cleverness that goes into picking a good cut-formula during proof-search. Indeed, for the mere simulation of our elementary DPLL(T) system, no memoisation table is required, just as no cut-formula ever needs to be picked.

Also note that the simulation of elementary DPLL(T) is tight and can be quantified: the bounds that we computed show that
proof-search in LK$^0(T)$ is no less efficient. Lifting this to the advanced features would require taking memoisation and/or general cuts into account (future work). This also hints at the field of proof complexity: the power of DPLL with or without backjumping, clause learning, etc, has been connected to proof complexity via Resolution systems [7]. But there is also a literature on proof complexity in sequent calculus (with/without cuts), which our approach should relate to.

While resolution systems have also been a way to relate DPLL and its variants to formal proof theory, the present paper was motivated by the import of such algorithms in a goal-directed framework, on which many systems (e.g. ProLog) are based. Resolution trees have also been used to perform conflict analysis; but since abstract presentations of DPLL(T) [32] abstract away the conflict analysis method (so as to accommodate any strategy), so did we (in the theory). On the other hand, our implementation in PSYCHE performs a conflict analysis according to a proof-theoretical method: eager weakenings a posteriori; and this we intend to compare to conflict analysis methods based on resolution trees or conflict graphs.

However, the handling of quantifiers is an important aspect of CLP, so relating our work to [23] would require extending LK$^0(T)$ with quantifiers, which is work-in-progress.

8. Conclusion and Further Work

In this paper we have identified an elementary DPLL(T) procedure and established a bismutation with the gradual construction of proof-trees in LK$^0(T)$ according to a simple strategy.

While LK$^0(T)$ differs from the inference system of e.g. [34] (e.g. we still take formulae to be trees and inference rules to organise the root-first decomposition of their connectives, rather than using DPLL(T)’s more flexible structures), it would be interesting to capture some of the related systems that extend DPLL(T) with e.g. full first-order logic and/or equality [4–6]. The full version of LK$^0(T)$ is indeed designed for handling quantifiers and equalities, so we hope to relate it to other techniques such as unification, parameterization, superposition, etc.

This is also our challenge for the PSYCHE implementation: ideally, try to show how the smart mechanisms that we know to be efficient at proving something, can be emulated or decomposed into an interaction between our kernel and a specific plug-in to be programmed. This would turn PSYCHE into a modular and collaborative platform, a standard format of problem-solving mechanisms. On which many systems (e.g. ProLog) are based. Resolution trees have also been used to perform conflict analysis; but since abstract presentations of DPLL(T) [32] abstract away the conflict analysis method (so as to accommodate any strategy), so did we (in the theory).

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