

Formalization of SAT Solvers

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Second Workshop on Formal and Automated Theorem Proving
and Applications

Overview

1 Introduction

- SAT problem and its applications
- Classic DPLL algorithm
- Modern DPLL modifications
- Verification of SAT solvers

2 Formalization of CNF propositional logic

3 State Transition Systems

- Formal system of Krstić and Goel
- Example of a simple system
- Formalization of state transition systems

4 Shallow embedding into HOL

- Code samples
- Verification

SAT problem

Definition (SAT problem)

Propositional satisfiability (SAT) problem is the problem of deciding if there is a truth assignment under which a given propositional formula (in CNF) evaluates to true. Satisfying truth assignment is a **model** of the formula.

Example

The formula

$$(x_1 \vee x_2) \wedge (\neg x_1 \vee \neg x_3) \wedge (\neg x_2 \vee x_3)$$

is true in the model $\{x_1, \neg x_2, \neg x_3\}$.

Example

The formula

$$(\neg x_1 \vee x_2) \wedge (\neg x_2 \vee x_3) \wedge (\neg x_3 \vee \neg x_1) \wedge x_1$$

is not satisfiable.

Applications of SAT solving

Many practical problems can be encoded in SAT.

- Electronic Design Automation
- Software and Hardware Verification
- Artificial Intelligence
- Planing and Scheduling
- Operations Research

SAT Solving Algorithms

Complete algorithms - for every SAT instance can either find its model or show that no model exists.

Stochastic algorithms - cannot show that no model exists, but can find a model of some large SAT instances very quickly.

We are only interested in **complete algorithms**.

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Classic DPLL - recursive formulation

```

function dpll ( $F$  : Formula) : (SAT, UNSAT)
begin
  if  $F$  is empty then                                     BASE
    return SAT
  else if there is an empty clause in  $F$  then
    return UNSAT
  else there is a unit clause  $[l]$  in  $F$  then             INFERENCE
    return dpll( $F[l \rightarrow \top]$ )
  else if there is a pure literal  $l$  in  $F$  then
    return dpll( $F[l \rightarrow \top]$ )
  else begin                                             SEARCH
    select a literal  $l$  occurring in  $F$ 
    if dpll( $F[l \rightarrow \top]$ ) = SAT then
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    else
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Progress in SAT Solving

- Spectacular improvements in the last decade.
- Possible to solve formulae with $\approx 10\,000$ variables and $\approx 1\,000\,000$ clauses

Reasons for this success

Conceptual enhancements of the DPLL procedure

- *backjumping*
- *conflict-driven lemma learning*
- *restarts*

Better implementation

- non-recursive implementation
- smart data-structures
- *two-watched literals scheme* for unit propagation,

Heuristic components

- *literal selection strategies*

Motivation

Goal

Have **trusted** SAT solvers.

Approaches

- Make SAT solvers produce **proofs of their claims** and verify those proofs by independent trusted checkers.
- Apply formal methods and **verify SAT solvers** themselves.

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Descriptions of modern SAT solvers

Concrete descriptions - Usually given in a form of programming language (pseudo)code. Close to real implementations, but hard to understand and reason about.

Abstract descriptions - Usually given as state transition systems. Easy to understand, formalize and reason about, but hide many important implementation details.

Approaches for verification

- 1 Verify **only abstract** descriptions.
- 2 Use **Hoare-logic style** verification for imperative code.
- 3 Formalize and verify SAT solvers by **shallow embedding into HOL** and automatically extract executable code.

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Syntax

Example

$$(x_1 \vee x_2) \wedge (\neg x_1 \vee \neg x_3) \wedge (\neg x_2 \vee x_3)$$

Model: $\{x_1, \neg x_2, \neg x_3\}$

Isabelle types

```
types    Variable = nat
datatype Literal   = Pos Variable
              | Neg Variable
types    Clause    = "Literal list"
types    Formula   = "Clause list"

types    Valuation = "Literal list"
```

Semantics

Definition

$v \models l$ if and only if $l \in v$

literalTrue :: "Literal => Valuation => bool"

$v \models \neg l$ if and only if $\bar{l} \in v$

literalFalse :: "Literal => Valuation => bool"

$v \models c$ if and only if $\exists l. l \in c \wedge v \models l$

clauseTrue :: "Clause => Valuation => bool"

$v \models \neg c$ if and only if $\forall l. l \in c \rightarrow v \models \neg l$

clauseFalse :: "Clause => Valuation => bool"

$v \models F$ if and only if $\forall c. c \in F \rightarrow v \models c$

formulaTrue :: "Formula => Valuation => bool"

$v \models \neg F$ if and only if $\exists c. c \in F \wedge v \models \neg c$

formulaFalse :: "Formula => Valuation => bool"

Semantics (cont.)

Definition

(consistent v) if and only if $(\neg \exists I. v \models I \wedge v \models \bar{I})$
consistent :: "Valuation => bool"
(model v F) if and only if (consistent $v \wedge v \models F$)
model :: "Valuation => Formula => bool"
(sat F) if and only if $(\exists v. \text{model } v F)$
satisfiable :: "Formula => bool"

Assertion trails

When building a non-recursive implementation the notion of valuation is extended.

Definition (Assertion trail)

Assertion trail is a list of literals, some of which are marked as *decision literals*. Decision literals split the trail into *levels*.

Example

A trail M could be $[+1, |-2, +6, |+5, -3, +4, |-7]$.

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Formal system of Krstić and Goel [KG07]

Decide:

$$\frac{I \in L \quad I, \bar{I} \notin M}{M := M \mid I}$$

UnitPropagate:

$$\frac{I \vee I_1 \vee \dots \vee I_k \in F \quad \bar{I}_1, \dots, \bar{I}_k \in M \quad I, \bar{I} \notin M}{M := M \mid I}$$

Conflict:

$$\frac{C = \text{no_cflct} \quad \bar{I}_1 \vee \dots \vee \bar{I}_k \in F \quad I_1, \dots, I_k \in M}{C := \{I_1, \dots, I_k\}}$$

Explain:

$$\frac{I \in C \quad I \vee \bar{I}_1 \vee \dots \vee \bar{I}_k \in F \quad I_1, \dots, I_k \prec I}{C := C \cup \{I_1, \dots, I_k\} \setminus \{I\}}$$

Learn:

$$\frac{C = \{I_1, \dots, I_k\} \quad \bar{I}_1 \vee \dots \vee \bar{I}_k \notin F}{F := F \cup \{I_1 \vee \dots \vee I_k\}}$$

Backjump:

$$\frac{C = \{I, I_1, \dots, I_k\} \quad \bar{I} \vee \bar{I}_1 \vee \dots \vee \bar{I}_k \in F \quad \text{level } I > m \geq \text{level } I_i}{C := \text{no_cflct} \quad M := M^{[m]} \bar{I}}$$

Forget:

$$\frac{C = \text{no_cflct} \quad c \in F \quad F \setminus c \models c}{F := F \setminus c}$$

Restart:

$$\frac{C = \text{no_cflct}}{M := M^{[0]}}$$

Solver state - (F, M, C)

- F - formula
- M - valuation (trail)
- C - conflict analysis clause

DPLL search - a simplified system

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Backtrack:

$$\frac{M \models \neg F \quad M = M' \mid I \quad M'' \text{ decisions } M'' = []}{M := M' \mid \bar{I}}$$

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Example

$F = [[-1, +2], [-1, -3], [-2, +4, +5], [+3, -4, -5], [-4, +5]]$

Function applied	sat?	M
	UNDEF	$[]$
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UnitProp ($c = [-1, -3], l = -3$)	UNDEF	$[+1, +2, -3]$
Decide ($l = +4$)	UNDEF	$[+1, +2, -3, +4]$
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Transition relation - formal definition

Definition (State)

State (M, F) is an ordered pair of an assertion trail M and formula F .

Definition

decide $(M_1, F_1) (M_2, F_2) \iff$

$$\exists l. \quad \text{var } l \in \text{vars } F_1 \wedge l \notin M_1 \wedge \bar{l} \notin M_1 \wedge \\ M_2 = M_1 @ l^\top \wedge F_2 = F_1$$

backtrack $(M_1, F_1) (M_2, F_2) \iff$

$$M_1 \models \neg F_1 \wedge \text{decisions } M_1 \neq [] \wedge \\ M_2 = \text{prefixBeforeLastDecision } M_1 @ \overline{\text{lastDecision } M_1}^\perp \wedge \\ F_2 = F_1$$

Transition relation - formal definition

Definition

$$(M_1, F_1) \rightarrow (M_2, F_2) \iff \begin{aligned} &\text{decide } (M_1, F_1) (M_2, F_2) \vee \\ &\text{backtrack } (M_1, F_1) (M_2, F_2) \vee \\ &\text{unitPropagate } (M_1, F_1) (M_2, F_2) \end{aligned}$$

The relation \rightarrow^* is the *transitive and reflexive closure* of the \rightarrow relation.

The state (M, F) is a **final state** if it is minimal wrt. the relation \rightarrow , i.e., if there is no state (M', F') st. $(M, F) \rightarrow (M', F')$.

Theorem (Soundness)

Let $([], F_0) \rightarrow^* (M, F)$.

• If

- 1 no conflict (i.e., $M \not\models \neg F$),
- 2 the rule *Decide* is not applicable
(i.e., $\text{var } l \in \text{vars } F_0, l \notin M \text{ and } \bar{l} \notin M$)

then F_0 is satisfiable and M is its model
(i.e., $\text{sat } F_0$ and $\text{model } M F_0$).

• If

- 1 conflict (i.e., $M \models \neg F$),
- 2 the rule *Backtrack* is not applicable
(i.e., $(\text{decisions } M) = []$),

then F_0 is unsatisfiable (i.e., $\neg \text{sat } F_0$).

Theorem (Soundness)

Let $([], F_0) \rightarrow^* (M, F)$.

• If

- ① no conflict (i.e., $M \not\models \neg F$),
- ② the rule *Decide* is not applicable
(i.e., $\text{var } l \in \text{vars } F_0, l \notin M \text{ and } \bar{l} \notin M$)

then F_0 is satisfiable and M is its model
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• If

- ① conflict (i.e., $M \models \neg F$),
- ② the rule *Backtrack* is not applicable
(i.e., $(\text{decisions } M) = []$),

then F_0 is unsatisfiable (i.e., $\neg \text{sat } F_0$).

Theorem (Pre-Completeness)

In every finite state (M, F) one of the following holds:

- 1 the rule *Backtrack* is not applicable
(i.e., $M \models \neg F$ and decisions $M = []$)
- 2 the rule *Decide* is not applicable
(i.e., $M \not\models \neg F$ and $\text{var } l \in \text{vars } F_0, l \notin M$ and $\bar{l} \notin M$)

Theorem (Termination)

Relation \rightarrow is well-founded, i.e., there is no infinite descending chain

$$([], F_0) \rightarrow (M_1, F_1) \rightarrow (M_2, F_2) \rightarrow \dots$$

How are these theorems proved?

Invariants

- Invariant*_{consistent}: consistent M
*Invariant*_{distinct}: distinct M
*Invariant*_{varsM}: $\text{vars } M \subseteq \text{vars } F$
*Invariant*_{impliedLiterals}: $\forall l. l \in M \implies (F @ \text{decisionsTo } l \ M) \models l$

Theorem

If $([], F_0) \rightarrow^ (M, F)$, then all invariants hold in the state (M, F) .*

How are these theorems proved?

The termination is proved using **well-founded orderings**.

Definition

$$l_1 \prec^{lit} l_2 \iff (\text{isDecision } l_1) \wedge \neg(\text{isDecision } l_2)$$

Definition

$$M_1 \succ_M M_2 \iff M_1 \prec_{lex}^{lit} M_2,$$

where \prec_{lex}^{lit} is a lexicographic extension of relation \prec^{lit} .

Definition

$$\begin{aligned} M_1 \succ_M^r M_2 \iff & (\text{consistent } M_1) \wedge (\text{distinct } M_1) \wedge (\text{vars } M_1) \subseteq VbI \\ & (\text{consistent } M_2) \wedge (\text{distinct } M_2) \wedge (\text{vars } M_2) \subseteq VbI \\ & M_1 \succ_M M_2 \end{aligned}$$

Overview

- 1 Introduction
 - SAT problem and its applications
 - Classic DPLL algorithm
 - Modern DPLL modifications
 - Verification of SAT solvers
- 2 Formalization of CNF propositional logic
- 3 State Transition Systems
 - Formal system of Krstić and Goel
 - Example of a simple system
 - Formalization of state transition systems
- 4 Shallow embedding into HOL
 - Code samples
 - Verification

Shallow embedding into HOL - characteristics

- Program is expressed as a set of recursive HOL functions.
- Proof methods are just standard induction principles and equational reasoning.
- No specialized program logic (e.g., Hoare logic) is necessary.
- Executable code can be automatically generated.
- Side-effects are impossible.

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Embedding of a SAT solver

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Embedding of a modern MiniSat-like SAT solver

Positive:

- Implementation follows state transition systems.
- All algorithms described by state transition systems are implemented.
- Also, major low-level implementation techniques are implemented.

Negative:

- There are no in-place modifications of data.
- Global variables are grouped to form a solver state which is explicitly passed around function calls.

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Solver state

```
record State =
  "getM"          :: LiteralTrail
  "getF"          :: Formula
  "getSATFlag"    :: ExtendedBool
  "getConflictFlag" :: bool
  "getConflictClause" :: pClause
  "getQ"          :: "Literal list"
  "getReason"     :: "Literal  $\Rightarrow$  pClause option"
  "getWatch1"     :: "pClause  $\Rightarrow$  Literal option"
  "getWatch2"     :: "pClause  $\Rightarrow$  Literal option"
  "getWatchList"  :: "Literal  $\Rightarrow$  pClause list"
  "getC"          :: Clause
  "getC1"         :: Literal
  "getC11"        :: Literal
```

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  "getWatchList"  :: "Literal  $\Rightarrow$  pClause list"
  "getC"          :: Clause
  "getCl"         :: Literal
  "getCl1"        :: Literal
```

applyDecide – implements Decide rule

```
definition applyDecide :: "State  $\Rightarrow$  Variable set  $\Rightarrow$  State"  
where  
  "applyDecide state decisionVars =  
    assertLiteral (selectLiteral state decisionVars) True state  
  "
```

The main solver function

```

definition solve :: "Formula  $\Rightarrow$  ExtendedBool"
where
"solve F0 = getSATFlag (solve_loop (initialize F0 initialState) (vars F0))"
definition solve_loop_body :: "State  $\Rightarrow$  Variable set  $\Rightarrow$  State"
where
"solve_loop_body state decisionVars =
  (let state_up = exhaustiveUnitPropagate state in
   (if (getConflictFlag state_up) then
    (if (currentLevel (getM state_up)) = 0 then
     state_up(| getSATFlag := False |)
    else
     let state_c = applyConflict state_up in
     let state_e = applyExplainUIP state_c in
     let state_l = applyLearn state_e in
     let state_b = applyBackjump state_l in
     state_b
    )
   else
    (if (vars (elements (getM state_up))  $\supseteq$  decisionVars) then
     state_up(| getSATFlag := TRUE |)
    else
     applyDecide state_up decisionVars
    )
  ))
"

```

solve_loop – a total recursive function

```
function (domintros, tailrec) solve_loop ::
  "State  $\Rightarrow$  Variable set  $\Rightarrow$  State"
where
  "solve_loop state decisionVars =
    (if (getSATFlag state)  $\neq$  UNDEF then
      state
    else
      let state' = solve_loop_body state decisionVars in
      solve_loop state' decisionVars
    )
  "
by pat_completeness auto
```

Two-watch literal scheme – the most complex function

primrec

notifyWatches_loop :: "Literal \Rightarrow pClause list \Rightarrow pClause list \Rightarrow State \Rightarrow State"

where

```
"notifyWatches_loop literal [] newWl state =
  state(| getWatchList := (getWatchList state)(literal := newWl) |)" |
"notifyWatches_loop literal (clause # list') newWl state =
  (let state' = (if Some literal = (getWatch1 state clause) then (swapWatches clause state)
    else state) in
  case (getWatch1 state' clause) of Some w1  $\Rightarrow$  (
  case (getWatch2 state' clause) of Some w2  $\Rightarrow$  (
  (if (literalTrue w1 (elements (getM state')))) then
    notifyWatches_loop literal list' (newWl @ [clause]) state'
  else
    (case (getNonWatchedUnfalsifiedLiteral ((getF state') ! clause) w1 w2 (getM state')) of
    Some l'  $\Rightarrow$ 
      notifyWatches_loop literal list' newWl (setWatch2 clause l' state') |
    None  $\Rightarrow$ 
      (if (literalFalse w1 (elements (getM state')))) then
        let state'' = state'(| getConflictFlag := True, getConflictClause := clause |) in
        notifyWatches_loop literal list' (newWl @ [clause]) state''
      else
        let state''' =
          state'(| getQ := (if w1 el (getQ state') then (getQ state') else (getQ state') @ [w1])) in
        let state'''' = (setReason w1 clause state''') in
        notifyWatches_loop literal list' (newWl @ [clause]) state''''))))))"
```

Total correctness

Theorem

$$\text{solve } F_0 = SAT \iff \text{sat } F_0$$

How is it proved?

- Correctness proofs for state transition systems were reused.
- New invariants (24 totally) were introduced.

A complex invariant

$$\begin{aligned} \forall c. c < |F| \implies M \models \neg (\text{watch}_1 c) \implies \\ & (\exists l. l \in c \wedge M \models l \wedge \text{level } l \leq \text{level } (\overline{\text{watch}_1 c})) \vee \\ & (\forall l. l \in c \wedge l \neq (\text{watch}_1 c) \wedge l \neq (\text{watch}_2 c) \implies \\ & M \models \neg l \wedge \text{level } \bar{l} \leq \text{level } (\overline{\text{watch}_1 c})). \end{aligned}$$

- Partial termination of recursive functions was proved, using the same orderings as for state transition systems.

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Some numbers

- Around 1 man-year effort.
- $\approx 25\,000$ lines of Isar code.
- Generated PDF-s ≈ 700 pages.

Further work

- Extract executable code from specifications.
- Use monadic programming to get imperative features

Non-monadic programming

```
definition setWatch1 :: "pClause  $\Rightarrow$  Literal  $\Rightarrow$  State  $\Rightarrow$  State"
where
  "setWatch1 clause literal state =
    let state' = state( getWatch1 := (getWatch1 state)(clause := Some literal) ) in
    addToWatchList literal clause state' "
```

Monadic programming

```
definition setWatch1 :: "nat  $\Rightarrow$  Literal  $\Rightarrow$  unit StateTransformer"
where
  "setWatch1 clause literal =
    do
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