Integrating Decision Procedures with Boolean Solving

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*Procédures de décision et vérification de programmes:*
Lecture 5
GOAL: lift decision procedures for the satisfiability problem of conjunctions of (ground) atoms to handle arbitrary Boolean combinations of (ground) atoms

Key technique: **integrate**

1. decision procedure for conjunctions of ground atoms with
2. a Boolean solver to handle the Boolean structure of the formula via
3. an abstract-check-refine schema

Systems based on this idea are called **(Lazy) Satisfiability Modulo Theory (SMT) Solvers**
There is one key pre-requisite to this lecture

1. **DPLL** algorithm (or... how to build a Boolean model of a propositional formula)

For ease of reference, the following slide contains the essential concept which will be used in the rest of this lecture...
Let $S$ be a set of clauses

**Unit Resolution**

$$
\frac{S \cup \{L, C \lor \bar{L}\}}{S \cup \{L, C\}} \quad \text{if } \quad \overline{\overline{A}} := A
$$

$$
\frac{S \cup \{L, C \lor L\}}{S \cup \{L\}} \quad \text{if } \quad \overline{A} := \neg A
$$

**Unit Subsumption**

**Splitting**

$$
\frac{S}{S \cup \{A\} \mid S \cup \{
eg A\}} \quad \text{if } A \text{ is an atom occurring in } S
$$

**Exercise** (mandatory): explain how a Boolean model can be extracted from the application of these rules
(Hint: think of the derivation trees and collect the various unit clauses...)

Pre-requisite: the DPLL algorithm
Index

1. Satisfiability Modulo Theory: introduction
   - Conflict sets and conflict clauses

2. SMT Solvers: an algorithmic view
   - SMT solver: a refinement of the basic algorithm

3. Abstract DPLL
What about Arbitrary Boolean Combinations?

- Let $\phi$ be an arbitrary Boolean combinations of atoms in a given theory $T$ with decidable satisfiability problem.
- How can we check the $T$-satisfiability of $\phi$?
  - Convert $\phi$ into DNF $\bigvee_i \Phi_i$ and check each disjunct $\Phi_i$ for $T$-satisfiability.
- **Main problem**: conversion to DNF may result in a blow-up in space of the resulting formula.
- Unfortunately, the problem is very hard since it subsumes the propositional satisfiability problem which is NP-hard.
  - We are interested in developing efficient solutions (in practice).
  - Engineering rather than logic.
Satisfiability Modulo Theory (SMT) Systems
http://combination.cs.uiowa.edu/smtlib

- Class of systems based on a the following idea: integrate
  - a Boolean solver (to handle the propositional structure of the formula)
  - a satisfiability procedure for the theory $T$
- All lazy SMT solvers the systems are based on an abstract-check-refine process
- There also exist eager SMT solvers that reduce satisfiability modulo theory to a pure Boolean satisfiability problem upfront and then use a SAT solver
Lazy SMT: Generic System Architecture

- Based on an abstract-check-refine process:
  - the input formula $\phi$ is abstracted to a propositional formula $\phi^p$
  - the Boolean solver enumerates the propositional assignments $\beta_1^p, ..., \beta_n^p$ of $\phi^p$
  - each conjunction $\beta_i$ of $T$-literals is checked for $T$-satisfiability
  - in case of $T$-unsatisfiability of $\beta_i$, refine the abstract propositional formula
Observations on SMT (I)

- Use an **off-the-shelf** Boolean solver to leverage its advantages, e.g.
  - SAT solvers can scale up
  - they return just one propositional satisfying assignment
  - they can accept incrementally new clauses...

**Scenario:** let $\beta^p$ be a propositional assignment of $\phi^p$ such that $\beta (\beta^p = \text{abs}(\beta))$ is $T$-unsatisfiable

How can we modify $\phi^p$ so that it return a new satisfying assignment distinct from $\beta^p$?

$$\phi^p := \phi^p \land \neg \beta^p$$
Observations on SMT (II)

- Did we really get a practically efficient system?
  - Well... no! Since we enumerate all possible satisfying assignments of $\phi^p$ which may be exponentially many in the number of propositional literals in $\phi$ or, equivalently, we have still to check exponentially many the $T$-satisfiability of a conjunction of literals
  - So, even though the satisfiability procedure for $T$ is very fast, it is invoked an exponential number of times and this may result in unacceptable degradation of the performances of the system
- How can we improve this situation in practice? In theory, this exponential blow-up is unavoidable!
Observations on SMT (II)

- To overcome this problem in practice, all SMT systems are based on the same technique which is a refinement of the technique to modify $\phi^p$ so that the Boolean solver return a new satisfying assignment distinct from $\beta^p$, namely:

$$\phi^p := \phi^p \land \neg \beta^p$$

- We assume that the satisfiability procedure for $T$ has a more complex interface: besides returning sat/unsat, it also returns a subset $\pi$ of the input set $\beta$ of the literals which is still $T$-unsatisfiable so that we can perform the following

$$\phi^p := \phi^p \land \neg \pi^p$$

Why should this be beneficial in practice?
Let us consider the following (unsatisfiable) conjunction of literals in the theory of uninterpreted function symbols:

$$\beta := \ell_1 \land \cdots \land f(x) \neq f(y) \land x = y \land \cdots \land \ell_n$$

where each $\ell_i$ is an equality for $i = 1, \ldots, n$

- If we use the naïve approach of enumerating all possible assignment, we will consider $2^{n-2}$ conjunctions of literals (corresponding to $2^{n-2}$ Boolean assignments) whose $T$-unsatisfiability is the responsibility of the same set $\{x = y, f(x) \neq f(y)\}$ of only two literals

- Instead, if we add the (propositional abstraction of the) negation of $\{x = y, f(x) \neq f(y)\}$ to $\phi^p$, we will prune the search space of the Boolean solver of all the $2^{n-2}$ propositional assignments in one shot!
Observations on SMT (II)

- If $n = 100$ (which is a small/medium number for practical applications), then we avoid invoking $2^{98}$, i.e. around $3.2 \cdot 10^{29}$, prop. assignments!

- In practice, this situation is quite common: unsatisfiability of large conjunction of literals usually depends on a small sub-set of the input literals.
- So, this technique gives spectacular results in practice and makes SMT system capable of tackling very large Boolean combinations of atoms in a given theory.
  - the $T$-unsatisfiable subset $\pi$ of the input literals is called conflict set and $\neg\pi$ is called a conflict clause.
How can we obtain conflict sets from sat. proc?

- Naïve approach: by guessing
  - after detecting the unsatisfiability of a set $\beta$ of literals...
  - pick a literal $\ell$ in $\beta$ and check the unsatisfiability of $\beta \setminus \{\ell\}$: if it is still unsatisfiable, then $\beta := \beta \setminus \{\ell\}$; otherwise pick another literal
  - stop when all literals have been considered and return $\beta$ as the conflict set
- **Pros**: very little effort in implementation
- **Cons**: not very efficient since each Boolean assignment, there are $n$ calls to the satisfiability procedure (this is a linear blow-up but it may be significant for applications especially if the satisfiability procedure is quadratic or cubic)
How can we obtain conflict sets from sat. proc?

- Some properties of the theory $T$ under-consideration may improve a little bit the situation:
  - if $T$ is convex, we can partition the set $\beta$ into $\Gamma \cup \Delta$ where $\Gamma$ is the set of equalities and $\Delta$ is the set of disequalities
  - the conflict set $\pi$ consists of equalities from $\Gamma$ plus at most one disequality from $\Delta$
  - the previous algorithm to compute a conflict set can then be applied to $\Gamma \subset \beta$ (not on the whole $\beta$)
- More efficient alternative: modify the satisfiability procedure to return a conflict set while checking for satisfiability
About the “quality” of conflict set?

• The smaller (in cardinality) the conflict set, the higher the number of propositional assignments are eliminated
  ▶ This requires to consider all subsets of a given set: an exponential number ➤ not practical!
• So, we focus on minimal conflict set: a conflict set is minimal if all its (strict) subsets are satisfiable
  ▶ Unfortunately, even for simple theories such as the theory of uninterpreted function symbol, the problem is NP-hard!
• In practice, we are already happy with approximations of minimal conflict sets
SMT System: main algorithm

function Bool + T (φ: quantifier-free formula)
    \( \phi^p \leftarrow \text{nf}(\text{fol2prop}(\phi)) \)
    while Bool-satisfiable(\( \phi^p \)) do
        \( \beta^p \leftarrow \text{pick_assign}(\phi^p) \)
        (\( \rho, \pi \)) \leftarrow T-satisfiable(prop2fol(\( \beta^p \)))
        if \( \rho = \text{sat} \) then return sat
        \( \phi^p \leftarrow \phi^p \land \neg \text{fol2prop}(\pi) \)
    end while
    return unsat
end
SMT System: main algorithm (remarks)

- **Propositional abstraction**: $\text{fol2prop}(\phi)$ returns the propositional abstraction $\phi^p$ of $\phi$ and $\text{fol2prop}$ is its inverse
  - $\text{fol2prop}(\phi)$ maps each atom in $\phi$ to a propositional letter, then it is extended homomorphically over the Boolean structure of $\phi$
  - $\text{nf}$ computes a normal form (e.g. CNF) of the input formula so that the Boolean solver is capable of handling it (SAT solvers can handle only formulae in CNF)
Computation of CNF

Let $\Phi = (a_1 \land \cdots \land a_m) \lor (b_1 \land \cdots \land b_n)$

Equivalent CNF:

$$\land_{i=1}^{m} \land_{j=1}^{n} (a_i \lor b_j)$$

Equisatisfiable CNF:

$$(X \lor Y) \land (X \iff a_1 \land \cdots \land a_m) \land (Y \iff b_1 \land \cdots \land b_n)$$

where $(X \iff a_1 \land \cdots \land a_m) \land (Y \iff b_1 \land \cdots \land b_n)$ can be represented as a conjunction of clauses (Exercise).
SMT System: main algorithm (remarks)

- The **Boolean solver** is supposed to provide two interface functionalities:
  - **Bool-satisfiable** which checks whether the input formula is propositionally satisfiable or not
  - **pick_assign** returns a satisfying assignment of the input formula
- It is usually also assumed that it is possible to **incrementally add** a clause to those already in the solver
  - Notice that this requirement is not crucial: SAT solvers are so fast that even restarting them from scratch each time a conflict clause is added is not dramatic in term of performances!
The satisfiability procedure for a theory $T$ is supposed to have the following interface: $T$-satisfiable$(\beta)$ returns the pair $(\rho, \pi)$ where

- $\rho$ is sat when $\beta$ is $T$-satisfiable ($\pi$ is left unspecified in this case)
- $\rho$ is unsat when $\beta$ is $T$-unsatisfiable and $\pi \subseteq \beta$ is $T$-unsatisfiable
The main loop of the algorithm searches a propositional assignment whose refinement is $T$-satisfiable.

- If such an assignment is found, then the satisfiability of the input formula is returned.
- If all assignments have been found unsatisfiable, then the unsatisfiability of the input formula is reported.
• **Lemma** [Return on sat] $\phi$ is $T$-satisfiable iff there exists an assignment $\beta^p$ of $\phi^p$ such that
  - $\beta^p$ (propositionally) satisfies $\phi^p$
  - $\beta$ is $T$-satisfiable

• **Lemma** [Loop invariant] If $\beta^p$ is a set of literals satisfying $\phi^p$ and $\beta$ is $T$-unsatisfiable, then
  - $\phi$ is $T$-satisfiable iff $\phi \land \neg \beta$ is
SMT System: correctness of the main algorithm

- **Lemma** [Loop variant] If $\beta^p$ is a set of literals satisfying $\phi^p$ and $\beta$ is $T$-unsatisfiable, then
  - the number of propositional assignments of $\phi^p \land \neg \beta^p$ are strictly less than those of $\phi^p$

- **Theorem** [Correctness] The function $\text{Bool} + T$ is a decision procedure for checking the satisfiability of arbitrary Boolean combination of ground atoms
SMT System: refining the main algorithm

- Observation: SAT solvers are capable of incrementally building a satisfying assignment...
  - it would be interesting to incrementally check the $T$-satisfiability of the corresponding literals while the assignment is built
  - in this way, we will discover unsatisfiability early and prune the search space of the Boolean solver with a “better” conflict set
- The idea is that before a split on a certain propositional letter (a non-deterministic step), it is wise to check whether the satisfying assignment built so far is satisfiable: if it is not, there is no point to continue and create a branching point!
  - the conflict set returned will contain literals higher-up in the search tree and this prunes substantially the search space of the SAT solver
SMT System: refining the main algorithm

function Bool + T (φ: quantifier-free formula)
  \( \phi^p \leftarrow nf(fol2prop(\phi)) \)
  while Bool-satisfiable(\( \phi^p \)) do
    \( \beta^p \leftarrow \emptyset \)
    repeat
      \( \beta^p \leftarrow extend\_partial\_assign(\phi^p, \beta^p) \)
      \((\rho, \pi) \leftarrow T\text{-satisfiable}(prop2fol(\beta^p)) \)
      if \((\rho = \text{sat} \land |\beta^p| = |Atoms(\phi)|)\) then return sat
    until \( \rho = \text{unsat} \)
    \( \phi^p \leftarrow \phi^p \land \neg fol2prop(\pi) \)
  end while
  return unsat
end
The new function `extend_partial_assign(\phi^p, \beta^p)` monotonically extends the input (partial) assignment.

- A possible implementation of this is to eagerly perform all the computations of the DPLL algorithm which do not create branching point and then return the resulting assignment so that its refinement can be checked for $T$-satisfiability.

- Notice that we can only report the satisfiability of the input formula when a **total** assignment is found $T$-satisfiable.

- The inner loop of the new version of the algorithm must terminate since $|Atoms(\phi)| - |\beta^p|$ is monotonically decreasing and $|Atoms(\phi)| - |\beta^p| \geq 0$. 

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SMT System: refining the main algorithm
Further complications with conflict sets

- Consider the case when $T = T_1 \cup T_2$ and you have used the Nelson-Oppen combination schema to build a sat proc for $T$ by modularly combining those for $T_1$ and $T_2$
- Assume also that each satisfiability procedure is capable of building a conflict set: how can you build a conflict set for $T$?
  - notice that because of purification, the conflict sets computed by each sat proc are not on the original signature...
  - So, it cannot be a subset of the input (non-purified) set of literals!
  - we need to track the equalities exchanged between the two procedures in order to map the conflict set on the extended signature to a subset of the original set of literals
Abstract DPLL


- A rule-based framework to specify DPLL
- Rules handle a data-structure $M \parallel F$ where $M$ is a partial assignment of Boolean variables, and $F$ is a set of clauses
- Case-split on Boolean variables implemented via a backtrack/backjump mechanism
- Easy extension from DPLL to DPLL(T)
- Study of different efficient strategies
### Classical DPLL (à la Nieuwenhuis-Oliveras-Tinelli)

<table>
<thead>
<tr>
<th>Procedure</th>
<th>Rule</th>
<th>Condition</th>
</tr>
</thead>
<tbody>
<tr>
<td>Propagate</td>
<td>$M \parallel F, C \lor l$ $\vdash M \models F, C \lor l$</td>
<td>if $M \models \neg C$, $l$ undefined in $M$</td>
</tr>
<tr>
<td>Decide</td>
<td>$M \parallel F$ $\vdash M \models^d F$</td>
<td>if $l$ or $\neg l$ in $F$, $l$ undefined in $M$</td>
</tr>
<tr>
<td>Fail</td>
<td>$M \parallel F, C$ $\vdash \bot$</td>
<td>if $M \models \neg C$, no decision literals in $M$</td>
</tr>
<tr>
<td>Backtrack</td>
<td>$M \models^d N \parallel F, C$ $\vdash M \models \neg l \parallel F, C$</td>
<td>if $\begin{cases} M \models^d N \models \neg C \ no decision literals in $N$ \end{cases}$</td>
</tr>
</tbody>
</table>
Conflict Driven Clause Learning: SAT level

Learn \[ M \parallel F \vdash M \parallel F, C \]

if each atom of \( C \) in \( F \) or in \( M \)

Forget \[ M \parallel F, C \vdash M \parallel F \]

if \( F \models C \)

Backjump \[ M I^d N \parallel F, C \vdash M I' \parallel F, C \]

if \( M I^d N \models \neg C \)

\( \exists C', I' : F, C \models C' \lor I' \)

if \( M \models \neg C' \)

\( I' \) undefined in \( M \)

\( I' \) or \( \neg I' \) in \( F \) or in \( M I^d N \)
Conflict Driven Clause Learning: SMT level

\[ \begin{align*}
T - \text{Learn} & \quad M \parallel F & \vdash & & M \parallel F, C & \text{if} & \begin{cases} & \text{each atom of } C \text{ in } F \text{ or in } M \\ & F \models_T C \end{cases} \\
T - \text{Forget} & \quad M \parallel F, C & \vdash & & M \parallel F & \text{if} & F \models_T C \\
T - \text{Backjump} & \quad M \mid^d N \parallel F, C & \vdash & & M \mid^d C \parallel F, C & \text{if} & \begin{cases} & M \mid^d N \models \neg C \\ & \exists C', l' : F, C \models_T C' \lor l' \\ & M \models \neg C' \\ & l' \text{ undefined in } M \\ & l' \text{ or } \neg l' \text{ in } F \text{ or in } M \mid^d N \end{cases} 
\end{align*} \]
SMT Solving: suggested reading

- **Introduction to DPLL(T)**

- **Overview on SMT solvers (both eager and lazy approaches)**

- **In-depth discussion about lazy SMT solvers**