# An Introduction to Satisfiability Modulo Theories

Albert Oliveras and Enric Rodríguez-Carbonell

Deduction and Verification Techniques
Session 1
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#### Overview of the session

- Motivation
- SMT
- Theories of Interest
- Eager approach
- Lazy approach
  - Optimizations
  - Theory propagation
  - DPLL(T) in depth



#### Introduction

- Historically, automated reasoning = uniform proof-search procedures for FO logic
- Little success: is FO logic the best compromise between expressivity and efficiency?
- Current trend is to gain efficiency by:
  - addressing only (expressive enough) decidable fragments of a certain logic
  - incorporate domain-specific reasoning, e.g.:
    - arithmetic reasoning
    - equality
    - data structures (arrays, lists, stacks, ...)



# Introduction (2)

#### Examples of this recent trend:

- SAT: use propositional logic as the formalization language
  - + high degree of efficiency
  - expressive (all NP-complete) but not natural encodings
- SMT: propositional logic + domain-specific reasoning
  - + improves the expressivity
  - certain (but acceptable) loss of efficiency

#### **GOAL OF THIS COURSE:**

study techniques, tools and applications of SAT/SMT



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## **Need and Applications of SMT**

- Some problems are more naturally expressed in other logics than propositional logic, e.g:
  - Software verification needs reasoning about equality, arithmetic, data structures, ...
- SMT consists of deciding the satisfiability of a (ground) FO formula with respect to a background theory
- Example (Equality with Uninterpreted Functions EUF):

$$g(a) = c \land (f(g(a)) \neq f(c) \lor g(a) = d) \land c \neq d$$

- Wide range of applications:
  - Predicate abstraction
  - Model checking
  - Equivalence checking

- Static analysis
- Scheduling
- Test-case generation
- **...**



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#### **Theories of Interest - EUF**

- $\blacksquare$  Equality with Uninterpreted Functions, i.e. "=" is equality
- If background logic is FO with equality, EUF is empty theory
- Consider formula

$$a*(f(b)+f(c)) = d \land b*(f(a)+f(c)) \neq d \land a = b$$

- Formula is UNSAT, but no arithmetic resoning is needed
- If we abstract the formula into  $h(a, g(f(b), f(c))) = d \land h(b, g(f(a), f(c))) \neq d \land a = b$  it is still UNSAT
- EUF is used to to abstract non-supported constructions
  - Non-linear multiplication
  - ALUs in circuits

#### **Theories of Interest - Arithmetic**

- Very useful for obvious reasons
- Restricted fragments support more efficient methods:
  - Bounds:  $x \bowtie k$  with  $\bowtie \in \{<,>,\leq,\geq\}$
  - Difference logic:  $x y \bowtie k$ , with  $\bowtie \in \{<, >, \le, \ge\}$
  - UTVPI:  $x \pm y \bowtie k$ , with  $\bowtie \in \{<,>,\leq,\geq\}$
  - Linear arithmetic, e.g.  $2x 3y + 4z \le 5$
  - Non-linear arithmetic, e.g.  $2xy + 4xz^2 5y \le 10$
  - Variables are either reals or integers

## **Theories of Interest - Arrays**

- Two interpreted function symbols read and write
- Theory is axiomatized by:
  - $\forall a \forall i \forall v \ (read(write(a, i, v), i) = v)$
  - $\forall a \forall i \forall j \forall v \ (i \neq j \rightarrow read(write(a, i, v), i) = read(a, j))$
- Sometimes extensionality is added:
  - $\forall a \forall b \ ( (\forall i (read(a,i) = read(b,i))) \rightarrow a = b$
- Is the following set of literals satisfiable?

$$write(a,i,x) \neq b$$
  $read(b,i) = y$   $read(write(b,i,x),j) = y$   $a = b$   $i = j$ 

- Used for:
  - Software verification
  - Hardware verification (memories)



#### **Theories of Interest - Fixed-width bit vectors**

- Constants represent vectors of bits
- Useful both for hardware and software verification
- Different type of operations:
  - String-like operations: concat, extract, ...
  - Logical operations: bit-wise not, or, and, ...
  - Arithmetic operations: add, substract, multiply, ...
- Assume bit-vectors have size 3. Is the formula SAT?

$$a[0:1] \neq b[0:1] \land (a|b) = c \land c[0] = 0 \land a[1] + b[1] = 0$$



#### **Theories of Interest - Combinations**

- In practice, theories are not isolated
- Software verifications needs arithmetic, arrays, bitvectors, ...
- Formulas of the following form usually arise:

$$a = b + 2 \land A = write(B, a + 1, 4) \land (read(A, b + 3) = 2 \lor f(a - 1) \neq f(b + 1))$$

The goal is to combine decision procedures for each theory

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#### Eager approach

- Methodology: translate problem into equisatisfiable propositional formula and use off-the-shelf SAT solver [Bryant, Velev, Pnueli, Lahiri, Seshia, Strichman, ...]
- Why "eager"?
  Search uses all theory information from the beginning
- Characteristics:
  - + Can use best available SAT solver
  - Sophisticated encodings are needed for each theory
- Tools: UCLID [Lahiri, Seshia and Bryant]



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Methodology:

Example: consider **EUF** and

$$\underbrace{g(a) = c}_{1} \land (\underbrace{f(g(a)) \neq f(c)}_{\overline{2}} \lor \underbrace{g(a) = d}_{3}) \land \underbrace{c \neq d}_{\overline{4}}$$

• SAT solver returns model  $[1, \overline{2}, \overline{4}]$ 

Methodology:

$$\underbrace{g(a) = c}_{1} \wedge (\underbrace{f(g(a)) \neq f(c)}_{\overline{2}} \vee \underbrace{g(a) = d}_{3}) \wedge \underbrace{c \neq d}_{\overline{4}}$$

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- Send  $\{1, \overline{2} \lor 3, \overline{4}, \overline{1} \lor 2 \lor 4\}$  to SAT solver

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- Send  $\{1, \overline{2} \lor 3, \overline{4}, \overline{1} \lor 2 \lor 4\}$  to SAT solver
- SAT solver returns model  $[1, 2, 3, \overline{4}]$
- Theory solver says *T*-inconsistent
- SAT solver detects  $\{1, \overline{2} \lor 3, \overline{4}, \overline{1} \lor 2 \lor 4, \overline{1} \lor \overline{2} \lor \overline{3} \lor 4\}$  UNSATISFIABLE



## Lazy approach (2)

• Why "lazy"? Theory information used lazily when checking *T*-consistency of propositional models

- Characteristics:
  - + Modular and flexible
  - Theory information does not guide the search
- Tools:
  - Barcelogic (UPC)
  - CVC3 (Univ. New York + Iowa)
  - DPT (Intel)

- MathSAT (Univ. Trento)
- Yices (SRI)
- Z3 (Microsoft)
- **...**

Several optimizations for enhancing efficiency:

Check *T*-consistency only of full propositional models



- Check T-consistency only of full propositional models
- Check T-consistency of partial assignment while being built



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- Check T-consistency of partial assignment while being built
- Given a T-inconsistent assignment M, add  $\neg M$  as a clause



- Check T-consistency only of full propositional models
- Check T-consistency of partial assignment while being built
- Given a T-inconsistent assignment M, add  $\neg M$  as a clause
- Given a *T*-inconsistent assignment *M*, identify a *T*-inconsistent subset  $M_0 \subseteq M$  and add  $\neg M_0$  as a clause



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- Upon a T-inconsistency, add clause and restart



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- Given a T-inconsistent assignment M, identify a T-inconsistent subset  $M_0 \subseteq M$  and add  $\neg M_0$  as a clause
- Upon a T-inconsistency, add clause and restart
- Upon a *T*-inconsistency, bactrack to some point where the assignment was still *T*-consistent



#### Lazy approach - Important points

Important and benefitial aspects of the lazy approach: (even with the optimizations)

- Everyone does what he/she is good at:
  - SAT solver takes care of Boolean information
  - Theory solver takes care of theory information
- Theory solver only receives conjunctions of literals
- Modular approach:
  - SAT solver and *T*-solver communicate via a simple API
  - SMT for a new theory only requires new *T*-solver
  - SAT solver can be embedded in a lazy SMT system with very few new lines of code (40?)



# Lazy approach - T-propagation

- As pointed out the lazy approach has one drawback:
  - Theory information does not guide the search
- How can we improve that?

T-Propagate:

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$$M \parallel F$$
  $\Rightarrow M l \parallel F \quad \text{if} \begin{cases} M \models_T l \\ l \text{ or } \neg l \text{ occurs in } F \text{ and not in } M \end{cases}$ 

- Search guided by *T*-Solver by finding T-consequences, instead of only validating it as in basic lazy approach.
- ▶ Naive implementation:: Add  $\neg l$ . If T-inconsistent then infer l. But for efficient Theory Propagation we need:
  - -T-Solvers specialized and fast in it.
  - -fully exploited in conflict analysis
- ullet This approach has been named  $\operatorname{DPLL}(T)$

$$\underbrace{g(a) = c}_{1} \land \underbrace{\left(\underbrace{f(g(a)) \neq f(c)}_{\overline{2}} \lor \underbrace{g(a) = d}_{3}\right)}_{3} \land \underbrace{c \neq d}_{\overline{4}}$$

$$\emptyset \parallel 1, \overline{2} \lor 3, \overline{4} \Rightarrow \text{(UnitPropagate)}$$

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$$12 \parallel 1, \overline{2} \lor 3, \overline{4} \Rightarrow \text{(UnitPropagate)}$$

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$$12 \parallel 1, \overline{2} \lor 3, \overline{4} \Rightarrow \text{(UnitPropagate)}$$

$$123 \parallel 1, \overline{2} \lor 3, \overline{4} \Rightarrow \text{(T-Propagate)}$$

$$\underbrace{g(a) = c}_{1} \wedge \underbrace{\left(\underbrace{f(g(a)) \neq f(c)}_{\overline{2}} \vee \underbrace{g(a) = d}_{3}\right)}_{3} \wedge \underbrace{c \neq d}_{\overline{4}}$$

$$0 \parallel 1, \overline{2} \vee 3, \overline{4} \Rightarrow \text{(UnitPropagate)}$$

$$1 \parallel 1, \overline{2} \vee 3, \overline{4} \Rightarrow \text{(T-Propagate)}$$

$$12 \parallel 1, \overline{2} \vee 3, \overline{4} \Rightarrow \text{(UnitPropagate)}$$

$$123 \parallel 1, \overline{2} \vee 3, \overline{4} \Rightarrow \text{(T-Propagate)}$$

$$1234 \parallel 1, \overline{2} \vee 3, \overline{4} \Rightarrow \text{(Fail)}$$

$$\underbrace{g(a) = c}_{1} \wedge \underbrace{\left(\underbrace{f(g(a)) \neq f(c)}_{\overline{2}} \vee \underbrace{g(a) = d}_{3}\right)}_{3} \wedge \underbrace{c \neq d}_{\overline{4}}$$

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$$123 \parallel 1, \overline{2} \vee 3, \overline{4} \Rightarrow \text{(T-Propagate)}$$

$$1234 \parallel 1, \overline{2} \vee 3, \overline{4} \Rightarrow \text{(Fail)}$$

$$fail$$



## DPLL(T) - Overall algorithm

High-levew view gives the same algorithm as a CDCL SAT solver:

```
while(true){
    while (propagate_gives_conflict()){
        if (decision_level==0) return UNSAT;
        else analyze_conflict();
    }
    restart_if_applicable();
    remove_lemmas_if_applicable();
    if (!decide()) returns SAT; // All vars assigned
}
```

#### Differences are in:

- propagate\_gives\_conflict
- analyze\_conflict



# DPLL(T) - Propagation

```
propagate_gives_conflict( ) returns Bool
    do {
      // unit propagate
      if ( unit_prop_gives_conflict() ) then return false
      // check T-consistency of the model
      if ( solver.is_model_inconsistent() ) then return false
      // theory propagate
      solver.theory_propagate()
    } while (someTheoryPropagation)
```



# $\mathsf{DPLL}(T)$ - Propagation (2)

- Three operations:
  - Unit propagation (SAT solver)
  - Consistency checks (*T*-solver)
  - Theory propagation (*T*-solver)
- Cheap operations are computed first
- If theory is expensive, calls to T-solver are sometimes skipped
- For completeness, only necessary to call *T*-solver at the leaves (i.e. when we have a full propositional model)
- Theory propagation is not necessary for completeness



# DPLL(T) - Conflict Analysis

Remember conflict analysis in SAT solvers:

```
C:= conflicting clause
while C contains more than one lit of last DL
    l:=last literal assigned in C
    C:=Resolution(C,reason(l))
end while

// let C = C' v l where l is UIP
backjump(maxDL(C'))
add l to the model with reason C
learn(C)
```



# DPLL(T) - Conflict Analysis (2)

Conflict analysis in DPLL(*T*):

```
if boolean conflict then C:= conflicting clause
else C:=\neg(\text{solver.explain\_inconsistency}())
while C contains more than one lit of last DL
    l:=last literal assigned in C
    C:=Resolution(C, reason(l))
end while
// let C = C' v l where l is UIP
backjump(maxDL(C'))
add 1 to the model with reason C
learn(C)
```



# DPLL(T) - Conflict Analysis (3)

What does explain\_inconsistency return?

- ▶ A (small) conjuntion of literals  $l_1 \land ... \land l_n$  such that:
  - They were in the model when *T*-inconsistency was found
  - It is *T*-inconsistent

What is now reason(l)?

- If l was unit propagated  $\longrightarrow$  clause that propagated it
- If *l* was *T*-propagated?
  - T-solver has to provide an explanation for l, i.e. a (small) set of literals  $l_1, \ldots, l_n$  such that:
    - They were in the model when l was T-propagated
    - $\bullet$   $l_1 \wedge \ldots \wedge l_n \models_T l$
  - Then reason(l) is  $\neg l_1 \lor ... \lor \neg l_n \lor l$



# DPLL(T) - Conflict Analysis (4)

Let *M* be of the form  $N, c = b, f(a) \neq f(b)$  and let *F* contain

$$a=b \lor g(a) \neq g(b), \qquad h(a)=h(c) \lor p, \qquad g(a)=g(b) \lor \neg p$$

$$h(a) = h(c) \vee p$$

$$g(a) = g(b) \vee \neg p$$

Take the following sequence:

- 1. Decide  $h(a) \neq h(c)$
- 2. T-Propagate  $a \neq b$  (due to  $h(a) \neq h(c)$  and c = b)
- 3. UnitPropagate  $g(a) \neq g(b)$
- 4. UnitPropagate p
- 5. Conflicting clause  $g(a) = g(b) \vee \neg p$

Explain
$$(a \neq b)$$
 is  $\{h(a) \neq h(c), c = b\}$ 

$$a = b \vee g(a) \neq g(b)$$

$$a = b \lor g(a) \neq g(b)$$

$$h(a) = h(c) \lor p \quad g(a) = g(b) \lor \neg p$$

$$h(a) = h(c) \lor g(a) = g(b)$$

$$h(a) = h(c) \lor c \neq b \lor a \neq b$$

$$h(a) = h(c) \lor a = b$$



$$h(a) = h(c) \lor c \neq b$$

## DPLL(T) - Some final remarks

- Completing a partial model is no longer a trivial task (no proper solution found so far)
- What about *T*-based decision heuristics? (no successful alternative found so far)
- What about producing proofs? 3 things to check
  - Explanations of inconsistencies are *T*-tautologies
  - Explanations of *T*-propagations are *T*-tautologies
  - Resolution propositional proof is correct
- See http://www.smtlib.org for benchmarks, theories, ...
- See http://www.smtcomp.org for competition results

