

# Abstract DPLL and Abstract DPLL Modulo Theories

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# Overview of the talk

- Motivation: SAT and SMT
- Propositional case
  - ◆ The Basic DPLL System
  - ◆ The DPLL System
- SMT case
  - ◆ Very Lazy Theory Learning
  - ◆ Lazy Theory Learning
  - ◆ Theory propagation

# Propositional satisfiability: SAT

- Deciding the satisfiability of a propositional formula is a very important problem
- **Theoretical interest:** first established NP-Complete problem, phase transition, ...
- **Practical interest:** applications to scheduling, planning, logic synthesis, verification, ...
  - ◆ Successful procedure: DPLL + **backjumping** + **learning**

# Satisfiability Modulo Theories

- Some problems are more naturally expressed in other logics
  - ◆ Pipelined microprocessors: logic **EUUF**, atoms are  $f(g(a, b), c) = g(c, a)$
  - ◆ Timed automata: **separation logic**, atoms are  $a < b + 2$
  - ◆ Software verification: **combination** of theories, e.g.  $5 + car(a + 2) = cdr(a + 1)$
- Deciding the satisfiability of a (**ground**) formula with respect to a background theory has lots of applications (SMT problem)

# Lifting SAT to SMT

- **Eager approach:** obtain an equisatisfiable propositional formula and use a SAT solver (UCLID)
- **Lazy approach:** abstract the formula into a propositional one and use a theory decision procedure to refine it (CVC, ICS, MathSAT, TSAT++, ...)
- **DPLL(T):** smarter way to use the theory information

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# The Basic DPLL Procedure

- Tries to incrementally **build a model**  $M$  for the CNF formula  $F$ .
- $M$  is augmented by **deciding** a literal or **deducing** one from  $M$  and  $F$ .
- When a wrong decision is detected, the procedure **backtracks**.

We will model it with a transition system between states:

$$M \parallel F \Longrightarrow M' \parallel F'$$

# The Basic DPLL System

Extending the model:

UnitProp

$$M \parallel F, C \vee l \implies M l \parallel F, C \vee l \quad \mathbf{if} \quad \begin{cases} M \models \neg C \\ l \text{ is undefined in } M \end{cases}$$

Decide

$$M \parallel F \implies M l^d \parallel F \quad \mathbf{if} \quad \begin{cases} l \text{ or } \neg l \text{ occurs in } F \\ l \text{ is undefined in } M \end{cases}$$



# The Basic DPLL System

Repairing the model:

Fail

$$M \parallel F, C \implies \text{fail} \text{ if } \left\{ \begin{array}{l} M \models \neg C \\ M \text{ contains no decision literals} \end{array} \right.$$

Backjump

$$M l^d N \parallel F \implies M l' \parallel F \text{ if } \left\{ \begin{array}{l} \text{for some clause } C \vee l' : \\ F \models C \vee l' \text{ and } M \models \neg C \\ l' \text{ is undefined in } M \\ l' \text{ or } \neg l' \text{ occurs in } F \end{array} \right.$$

# Basic DPLL System - Example

$\emptyset$	$\parallel$	$\bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, 6 \vee \bar{5} \vee \bar{2}$	$\implies$	(Decide)
1	$\parallel$	$\bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, 6 \vee \bar{5} \vee \bar{2}$	$\implies$	(UnitProp)
1 2	$\parallel$	$\bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, 6 \vee \bar{5} \vee \bar{2}$	$\implies$	(Decide)
1 2 3	$\parallel$	$\bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, 6 \vee \bar{5} \vee \bar{2}$	$\implies$	(UnitProp)
1 2 3 4	$\parallel$	$\bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, 6 \vee \bar{5} \vee \bar{2}$	$\implies$	(Decide)
1 2 3 4 5	$\parallel$	$\bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, 6 \vee \bar{5} \vee \bar{2}$	$\implies$	(UnitProp)
1 2 3 4 5 $\bar{6}$	$\parallel$	$\bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, 6 \vee \bar{5} \vee \bar{2}$	$\implies$	(Backjump)
1 2 5	$\parallel$	$\bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, 6 \vee \bar{5} \vee \bar{2}$	$\implies$	...

# Basic DPLL System - Example

$$\begin{array}{l}
 \dots \\
 1\ 2\ 3\ 4\ 5\ \bar{6} \parallel \bar{1}\vee 2, \bar{3}\vee 4, \bar{5}\vee \bar{6}, 6\vee \bar{5}\vee \bar{2} \implies (\text{Backjump}) \\
 1\ 2\ \bar{5} \parallel \bar{1}\vee 2, \bar{3}\vee 4, \bar{5}\vee \bar{6}, 6\vee \bar{5}\vee \bar{2}
 \end{array}$$

In this case  $F \models \bar{1}\vee \bar{5}$  we have by resolution

$$\frac{\frac{\bar{1}\vee 2 \quad 6\vee \bar{5}\vee \bar{2}}{\bar{1}\vee 6\vee \bar{5}} \quad \bar{5}\vee \bar{6}}{\bar{1}\vee \bar{5}}$$

and before deciding 3, we could have deduced  $\bar{5}$ .

# Basic DPLL System-Correctness

- $\emptyset \parallel F \Longrightarrow! fail$  iff  $F$  is unsatisfiable
- $\emptyset \parallel F \Longrightarrow! M \parallel F$  iff  $F$  is satisfiable

Key ingredients:

- All rules decrease with respect to a **well-founded ordering** between states
- When  $M$  **falsifies** a clause in  $F$ , either Fail or Backjump apply.

# The DPLL System

Learning and forgetting clauses:

Learn

$$M \parallel F \implies M \parallel F, C \text{ if } \begin{cases} \text{all atoms of } C \text{ occur in } F \\ F \models C \end{cases}$$

Forget

$$M \parallel F, C \implies M \parallel F \text{ if } F \models C$$

The DPLL system **terminates** if no clause is **learned/forgotten** infinitely often

# The DPLL system - Strategies

- Applying one rule of the Basic DPLL system between each two Learn ensures termination
- In practice, Learn is usually (but not only) applied right after Backjump.
- A common strategy is to apply the rules using the following priorities:
  1. If there is a clause in  $F$  which is false in  $M$  apply Fail or Backjump + Learn
  2. Apply UnitProp
  3. Apply Decide

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# Very Lazy Approach for SMT

$$\underbrace{g(a)=c}_1 \wedge \left( \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a)=d}_3 \right) \wedge \underbrace{c \neq d}_{\bar{4}}$$

- SAT solver returns model  $[1, \bar{2}, \bar{4}]$



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- Send  $\{1, \bar{2} \vee 3, \bar{4}, \bar{1} \vee 2\}$  to SAT solver

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

# Very Lazy Approach - Modelling

- The process within the SAT solver is modelled using the DPLL system
- The interaction between the theory solver and the SAT solver is modelled with the rule

Very Lazy Theory Learning

$$M \perp M_1 \parallel F \implies \emptyset \parallel F, \bar{l}_1 \vee \dots \vee \bar{l}_n \vee \bar{l} \text{ if } \begin{cases} M \perp M_1 \models F \\ \{l_1, \dots, l_n\} \subseteq M \\ l_1 \wedge \dots \wedge l_n \models_T \bar{l} \end{cases}$$

# Lazy approach

- Detects  $T$ -inconsistent **partial** models using  
Lazy Theory Learning

$$M \models M_1 \parallel F \implies M \models M_1 \parallel F, \bar{l}_1 \vee \dots \vee \bar{l}_n \vee \bar{l} \text{ if } \begin{cases} \{l_1, \dots, l_n\} \subseteq M \\ l_1 \wedge \dots \wedge l_n \models_T \bar{l} \\ \bar{l}_1 \vee \dots \vee \bar{l}_n \vee l \notin F \end{cases}$$

- The learnt clause is **false** in  $M \models M_1$  and hence either Backjump or Fail apply

# Lazy approach - Strategies

- A **common strategy** is to apply the rules using the following priorities:
  1. If there is a clause in  $F$  which is false in  $M$  apply Fail or Backjump + Learn
  2. If the model is  $T$ -inconsistent apply Lazy Theory Learning + ( Backjump or Fail)
  3. Apply UnitProp
  4. Apply Decide



# DPLL(T) - Eager T-Propagation

- Use the theory information as soon as possible by eagerly applying

Theory Propagate

$$M \parallel F \implies M l \parallel F \text{ if } \begin{cases} M \models_T l \\ l \text{ or } \bar{l} \text{ occurs in } F \\ l \text{ is undefined in } M \end{cases}$$

# Eager T-Propagation - Example

$$\underbrace{g(a)=c}_1 \wedge \left( \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a)=d}_3 \right) \wedge \underbrace{c \neq d}_{\bar{4}}$$

$$\emptyset \parallel 1, \bar{2} \vee 3, \bar{4} \implies (\text{UnitProp})$$

$$1 \parallel 1, \bar{2} \vee 3, \bar{4} \implies (\text{Theory Propagate})$$

$$1\ 2 \parallel 1, \bar{2} \vee 3, \bar{4} \implies (\text{UnitProp})$$

$$1\ 2\ 3 \parallel 1, \bar{2} \vee 3, \bar{4} \implies (\text{Theory Propagate})$$

$$1\ 2\ 3\ 4 \parallel 1, \bar{2} \vee 3, \bar{4} \implies (\text{Fail})$$

*fail*

# Eager Theory Propagation

- By eagerly applying Theory Propagate any  $M$  will be  $T$ -consistent, since  $M_1 \wedge l$  is  $T$ -inconsistent iff  $M_1 \models_T \bar{l}$
- Therefore, Lazy Theory Learning will never apply
- For some logics, e.g. separation logic, this approach is extremely effective
- For some other, e.g. EUF, it is too expensive to detect all  $T$ -consequences

# Non-Exhaustive T-Propagation

- If Theory Propagate is not eagerly applied, Lazy Theory Learning is **needed** to repair  $T$ -inconsistent models
- The six rules of the DPLL system plus Theory Propagate and Lazy Theory Learning provide a **decision procedure** for SMT
- **Termination** is usually ensured this way:
  - ◆ Between two Learn applications some rule of the Basic DPLL is applied
  - ◆ Apply Backjump or Fail immediately after Lazy Theory Learning

# Conclusions

- The DPLL procedure can be modelled in an **abstract** way
- Modern features such as **backjumping**, **learning** (also **restarts**) can be captured with our transition systems
- Extensions to **SMT** are possible
- It allows one to describe the **strategies** of concrete systems in a clean way