

# Systems and Solving Techniques for Knowledge Representation

## – (Normal) ASP solving [Part I - SAT] –

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## Goal of this part

In this part of the lectures the goal is to show how the main solvers for normal (non-disjunctive) ASP solvers, e.g.

- CMODELS,
- SMODELS,
- CLASP,
- ...

try to solved the program at hand.

We will NOT do these by presenting algorithm's behavior as usually done, i.e. by means of pseudo-code descriptions.

# Algorithms presentation: Motivation for alternative way

## Issue

- Usually solving algorithms are presented by means of pseudo-code descriptions, but
- some communities have experienced that analyzing such algorithms on this basis may not be fruitful.

## Instead ...

- more formal descriptions, based on mathematically precise but possibly simple objects, can be useful, and
- can allow for, e.g. a uniform representation.

# Algorithms presentation: Motivation for alternative way

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## Instead ...

- more formal descriptions, based on mathematically precise but possibly simple objects, can be useful, and
- can allow for, e.g. a uniform representation.

# Abstract solvers

**Abstract solvers** are a relatively new methodology for **describing**, **comparing** and **composing** solving procedures in an abstract way via graphs, where

- the states of computation are represented as nodes,
- the solving techniques as edges between such nodes,
- the solving process as a path in the graph, and
- formal properties of the procedures are reduced to related graph's properties.

# What are they good for?

- **Describing** abstract solving procedures in a clear mathematical and unified way via graphs.
- **Comparing** solving techniques employed in different procedures by means of comparison of related graphs.
- **Combining** abstract solving procedures, by means of modular addition/deletion of techniques/edges, to design novel abstract procedures.

# What are they not (that) good for?

- **Specifying** (low level) implementation details.
- **Arguing** about the efficiency of an implementation built on this basis.

# Outline

- 1 Abstract Solvers for SAT [Nieuwenhuis et al., 2006]
- 2 Abstract Solvers for non-disjunctive ASP [Lierler, 2011]



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## DPLL SAT solving: Some notation

Given a set  $X$  of atoms,

- a record relative to  $X$  is a string  $L$  composed of literals over  $X$  or the symbol  $\perp$ , with no repetitions.
- Some literal  $l$ , called *decision literal*, may be annotated as  $l^\Delta$ .

### (In)Consistent records

We say that a record  $L$  is *inconsistent* if it contains both a literal  $l$  and its complement  $\bar{l}$ , or if it contains  $\perp$ , and *consistent* otherwise.

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# DPLL algorithm for SAT solving

The Davis-Putnam-Logemann-Loveland algorithm [Davis and Putnam, 1960, Davis et al., 1962] is the most famous and used (backtracking-based) algorithm for solving the propositional satisfiability (SAT) problem.

Propositional formulas are in Conjunctive Normal Form (CNF), i.e. set of clauses.

A *model* of a CNF formula  $F$  is a (total) assignment to variables satisfying  $F$ .

# DPLL SAT solving: Nodes of the graph $DPLL_F$

A *state* relative to (a set of atoms)  $X$  is either

- 1 a record relative to  $X$ , or
- 2 the distinguished state *SAT* or *UNSAT*.

## Nodes of the graph $DPLL_F$

- The set of nodes of graph  $DPLL_F$  consists of the states relative to the set of atoms  $atoms(F)$  appearing in  $F$ .
- A node in the graph is **terminal** if no edge originates from it.
- The state  $\emptyset$  is called **initial**.

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## DPLL SAT solving: Edges of the graph $DPLL_F$

*Conclude* :  $L \implies UNSAT$  if  $\left\{ \begin{array}{l} L \text{ is inconsistent and} \\ L \text{ contains no decision literals} \end{array} \right.$

*Backtrack* :  $L \wedge L' \implies L \bar{l}$  if  $\left\{ \begin{array}{l} L \wedge L' \text{ is inconsistent and} \\ L' \text{ contains no decision literals} \end{array} \right.$

*Unit* :  $L \implies L \vee l$  if  $\left\{ \begin{array}{l} l \text{ does not occur in } L \text{ and} \\ F \text{ contains a clause } C \vee l \text{ and} \\ \text{all the literals of } \bar{C} \text{ occur in } L \end{array} \right.$

*Decide* :  $L \implies L \wedge l$  if  $\left\{ \begin{array}{l} L \text{ is consistent and} \\ \text{neither } l \text{ nor } \bar{l} \text{ occur in } L \end{array} \right.$

*Success* :  $L \implies SAT$  if no other rule applies

**Figure** : Transition rules that justify edges of the graph  $DPLL_F$ .

# DPLL SAT solving: Example (I)



Figure : Example of path in  $DPLL_{\{a \vee b, \bar{a} \vee c\}}$ .



# DPLL SAT solving: Example (I)

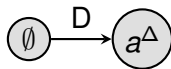


Figure : Example of path in  $DPLL_{\{avb, \bar{a}vc\}}$ .

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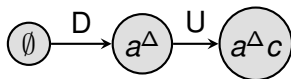


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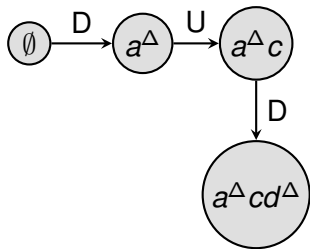


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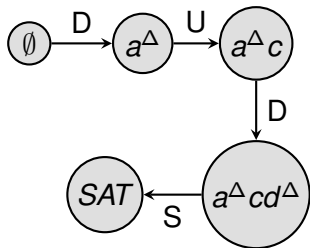


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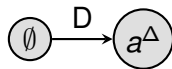


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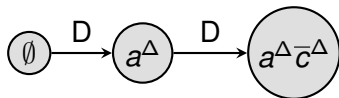


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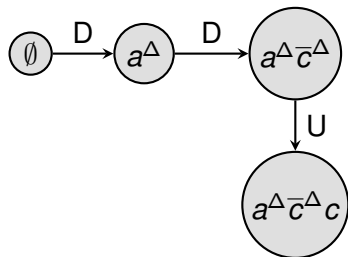


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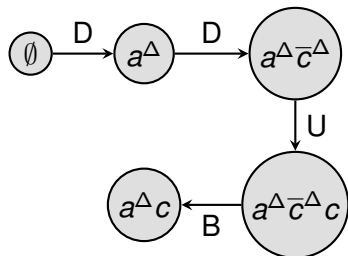


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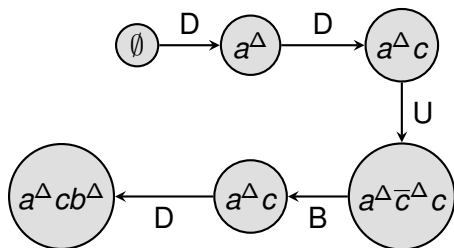


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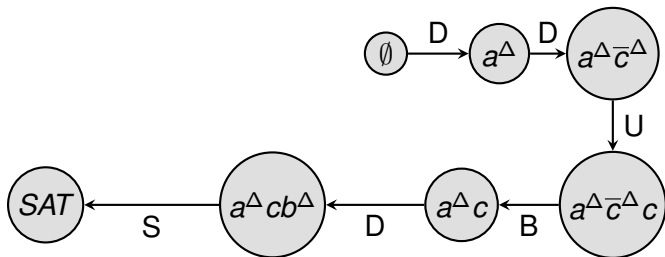


Figure : Example of path in  $DPLL_{\{a \vee b, \bar{a} \vee c\}}$ .

## (Abstract) Solver description

Graph + Formula/Program's instantiation +  
Rule's ordering

# DPLL SAT solving: Formal result

[Nieuwenhuis et al., 2006]; Proposition 1 in [Lierler, 2011]

## Theorem

*For any CNF formula  $F$ ,*

- 1 *graph  $DPLL_F$  is finite and acyclic,*
- 2 *any terminal state reachable from  $\emptyset$  in  $DPLL_F$  other than UNSAT is SAT, and*
- 3 *UNSAT is reachable from  $\emptyset$  in  $DPLL_F$  if and only if  $F$  is unsatisfiable.*

# DPLL SAT solving: Graphical alternative for Examples

|                 |                                |                  |                                      |
|-----------------|--------------------------------|------------------|--------------------------------------|
| Initial state : | $\emptyset$                    | Initial state :  | $\emptyset$                          |
| <i>Decide</i>   | $\implies a^\Delta$            | <i>Decide</i>    | $\implies a^\Delta$                  |
| <i>Unit</i>     | $\implies a^\Delta c$          | <i>Decide</i>    | $\implies a^\Delta \bar{c}^\Delta$   |
| <i>Decide</i>   | $\implies a^\Delta c b^\Delta$ | <i>Unit</i>      | $\implies a^\Delta \bar{c}^\Delta c$ |
| <i>Success</i>  | $\implies SAT$                 | <i>Backtrack</i> | $\implies a^\Delta c$                |
|                 |                                | <i>Decide</i>    | $\implies a^\Delta c b^\Delta$       |
|                 |                                | <i>Success</i>   | $\implies SAT$                       |

Figure : Examples of paths in  $DPLL_{\{a \vee b, \bar{a} \vee c\}}$ .

# CDCL algorithm for SAT

The Conflict-Driven Clause Learning algorithm for SAT “extends” the DPLL algorithm with optimized backtracking techniques, i.e. *backjumping* and *learning* borrowed from CSP [Prosser, 1993].

See, e.g. [Bayardo and Schrag, 1997, Marques-Silva and Sakallah, 1996, Zhang et al., 2001]

# CDCL algorithm for SAT: Intuition

- Backjumping is the ability of, instead of backtracking chronologically to the last decision as in Backtrack, back-jumping over decision literals that “were not directly responsible for the inconsistency”;
- Learning adds clauses, to be conjoined with the original formula, in order to prevent that a similar inconsistency is encountered again in another part of the search tree.



# CDCL SAT solving: Extended states

## $DPLL_{Learn_F}$ graph

- 1 Its nodes are **extended states** relative to  $F$ , and
- 2 its edges are justified by **extended, updated and additional** transition rules wrt  $DPLL_F$ .

For a CNF formula  $F$ , an *extended state* relative to  $F$  is either

- 1 a pair  $(L, \Gamma)$ , written  $L \parallel \Gamma$ , where
  - $L$  is a record relative to  $atoms(F)$ , and
  - $\Gamma$  is a set of clauses over  $atoms(F)$  that are entailed by  $F$ ;or
- 2 the distinguished state *SAT* or *UNSAT*.

## Initial state

The (extended) initial state is  $\emptyset \parallel \emptyset$ .

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## Initial state

The (extended) initial state is  $\emptyset \parallel \emptyset$ .

# CDCL SAT solving: Updated and extended rules (I)

*Conclude* :  $L \parallel \Gamma \implies UNSAT$  if  $\left\{ \begin{array}{l} L \text{ is inconsistent and} \\ L \text{ contains no decision literals} \end{array} \right.$

*Backjump* :  $L \wedge L' \parallel \Gamma \implies L' \parallel \Gamma$  if  $\left\{ \begin{array}{l} L \wedge L' \text{ is inconsistent and} \\ F \models l' \vee \bar{l} \end{array} \right.$

*UnitLearn* :  $L \parallel \Gamma \implies L \wedge l \parallel \Gamma$  if  $\left\{ \begin{array}{l} l \text{ does not occur in } L \text{ and} \\ F \cup \Gamma \text{ contains a clause } C \vee l \text{ and} \\ \text{all the literals of } \bar{C} \text{ occur in } L \end{array} \right.$

*Decide* :  $L \parallel \Gamma \implies L \wedge l \parallel \Gamma$  if  $\left\{ \begin{array}{l} L \text{ is consistent and} \\ \text{neither } l \text{ nor } \bar{l} \text{ occur in } L \end{array} \right.$

# CDCL SAT solving: Additional transition rules

*Learn*:  $L \parallel \Gamma \implies L \parallel C \cup \Gamma$  if  $\left\{ \begin{array}{l} \text{every atom in } C \text{ occurs in } F \text{ and} \\ F \models C \end{array} \right.$

# CDCL SAT solving: Updated and extended rules (II)

- Conclude* :  $L \parallel \Gamma \implies UNSAT$  if  $\begin{cases} L \text{ is inconsistent and} \\ L \text{ contains no decision literals} \end{cases}$
- Backjump* :  $L \wedge L' \parallel \Gamma \implies L' \parallel \Gamma$  if  $\begin{cases} L \wedge L' \text{ is inconsistent and} \\ F \models l' \vee \bar{l} \end{cases}$
- UnitLearn* :  $L \parallel \Gamma \implies L \wedge l \parallel \Gamma$  if  $\begin{cases} l \text{ does not occur in } L \text{ and} \\ F \cup \Gamma \text{ contains a clause } C \vee l \text{ and} \\ \text{all the literals of } \bar{C} \text{ occur in } L \end{cases}$
- Decide* :  $L \parallel \Gamma \implies L \wedge \bar{l} \parallel \Gamma$  if  $\begin{cases} L \text{ is consistent and} \\ \text{neither } l \text{ nor } \bar{l} \text{ occur in } L \end{cases}$
- Success* :  $L \parallel \Gamma \implies SAT$  if no other rule applies other than *Learn*

# CDCL SAT solving: Example

|                  |   |
|------------------|---|
| Initial state :  | $\emptyset \parallel \emptyset$                         |
| <i>Learn</i>     | $\Rightarrow \emptyset \parallel \{b \vee c\}$          |
| <i>Decide</i>    | $\Rightarrow \bar{b}^\Delta \parallel \{b \vee c\}$     |
| <i>UnitLearn</i> | $\Rightarrow \bar{b}^\Delta c \parallel \{b \vee c\}$   |
| <i>UnitLearn</i> | $\Rightarrow \bar{b}^\Delta c a \parallel \{b \vee c\}$ |
| <i>Success</i>   | $\Rightarrow \text{SAT}$                                |

Figure : Example of path in  $DPLL_{\{a \vee b, \bar{a} \vee c\}}$ .

# CDCL SAT solving: Formal result

## Theorem

*For any formula  $F$ ,*

- 1 every path in  $DPLL_{\text{Learn}_F}$  uses only finitely many times edges justified by transition rules other than Learn,*
- 2 any terminal state reachable from  $\emptyset \parallel \emptyset$  in  $DPLL_{\text{Learn}_F}$  other than UNSAT is SAT, and*
- 3 UNSAT is reachable from  $\emptyset \parallel \emptyset$  in  $DPLL_{\text{Learn}_F}$  if and only if  $F$  is unsatisfiable.*

## CDCL SAT solving: Additional transition rules (II)

*Learn* :  $L \parallel \Gamma \implies L \parallel C \cup \Gamma$  if  $\left\{ \begin{array}{l} \text{every atom in } C \text{ occurs in } F \text{ and} \\ F \models C \end{array} \right.$

When Learning comes into play, SAT solvers usually implement two more techniques

- Restart starts the search from scratch but maintaining the learned clauses;
- Forget deletes a previously added clause.



## CDCL SAT solving: Additional transition rules (II)

*Learn* :  $L \parallel \Gamma \implies L \parallel C \cup \Gamma$  if  $\left\{ \begin{array}{l} \text{every atom in } C \text{ occurs in } F \text{ and} \\ F \models C \end{array} \right.$

*Restart* :  $L \parallel \Gamma \implies \emptyset \parallel \Gamma$

*Forget* :  $L \parallel C \cup \Gamma \implies L \parallel \Gamma$

# DPLL SAT solving: States (slightly modified)

A *state* relative to (a set of atoms)  $X$  is either

- 1 A record relative to  $X$ ,
- 2  $Ok(L)$  where  $L$  is a record relative to  $X$ , or
- 3 The distinguished state *UNSAT*.

## States and graphs

- The set of nodes of  $DPLL_F$  consists of the states relative to the set of atoms appearing in  $F$   $atoms(F)$ .
- A node in the graph is *terminal* if no edge originates from it.
- The state  $\emptyset$  is called *initial*.
- Each formula  $F$  determines its DPLL graph  $DPLL_F$ .

## DPLL SAT solving: States (slightly modified)

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# DPLL SAT solving: Transition rules (slightly modified)

*Conclude* :  $L \implies UNSAT$  if  $\begin{cases} L \text{ is inconsistent and} \\ L \text{ contains no decision literals} \end{cases}$

*Backtrack* :  $L \wedge L' \implies L\bar{l}$  if  $\begin{cases} L \wedge L' \text{ is inconsistent and} \\ L' \text{ contains no decision literals} \end{cases}$

*Unit* :  $L \implies L \vee l$  if  $\begin{cases} l \text{ does not occur in } L \text{ and} \\ F \text{ contains a clause } C \vee l \text{ and} \\ \text{all the literals of } \bar{C} \text{ occur in } L \end{cases}$

*Decide* :  $L \implies L \wedge l$  if  $\begin{cases} L \text{ is consistent and} \\ \text{neither } l \text{ nor } \bar{l} \text{ occur in } L \end{cases}$

*Success* :  $L \implies Ok(L)$  if no other rule applies

# DPLL SAT solving: Formal result (slightly modified)

## Theorem

For any CNF formula  $F$ ,

- 1 graph  $DPLL_F$  is finite and acyclic,
- 2 any terminal state reachable from  $\emptyset \parallel \emptyset$  in  $DPLL_F$  other than UNSAT is  $Ok(L)$ , with (the assignment that can be built from)  $L$  being a model of  $F$ , and
- 3 UNSAT is reachable from  $\emptyset$  in  $DPLL_F$  if and only if  $F$  is unsatisfiable.

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# Outline

- 1 Abstract Solvers for SAT [Nieuwenhuis et al., 2006]
- 2 Abstract Solvers for non-disjunctive ASP [Lierler, 2011]

## Non-disjunctive programs

A *program*  $\Pi$  consists of finitely many rules of the form

$$a \leftarrow b_1, \dots, b_l, \text{not } b_{l+1}, \dots, \text{not } b_m$$

where

- the *head*  $a$  is an atom or  $\perp$ , and
- in the *body*  $b_1, \dots, b_l, \text{not } b_{l+1}, \dots, \text{not } b_m$ , each  $b_i (1 \leq i \leq m)$  is an atom.

We can identify a rule with the clause

$$a \vee \overline{b_1} \vee \dots \vee \overline{b_l} \vee b_{l+1} \vee \dots \vee b_m$$

and also with the set of its elements.

*Answer sets* are defined in terms of *reduct* and minimality [Gelfond and Lifschitz, 1988].



# SAT-based Generate&Test procedure [Lierler, 2008]

We first present a modification of the  $DPLL_F$  graph.

## Setting

- $F$  is a CNF formula,
- $G$  is formula formed from atoms in  $atoms(F)$ .

## Graph $GT_{F,G}$

- The nodes are the same as  $DPLL_F$ .
- The edges are justified by the transition rules of  $DPLL_F$  and

$$Test : \quad L \implies \bar{L} \quad \text{if} \quad \left\{ \begin{array}{l} L \text{ is consistent and} \\ G \models \bar{L} \text{ and} \\ I \in L \end{array} \right.$$

# SAT-based Generate&Test procedure: Formal result

## Theorem

*For any CNF formula  $F$  and a formula  $G$  formed from  $\text{atoms}(F)$*

- 1 graph  $GT_{F,G}$  is finite and acyclic,*
- 2 any terminal state reachable from  $\emptyset$  in  $GT_{F,G}$  other than UNSAT is  $Ok(L)$ , with  $L$  being a model of  $F \wedge G$ , and*
- 3 UNSAT is reachable from  $\emptyset$  in  $GT_{F,G}$  if and only if  $F \wedge G$  is unsatisfiable.*

# Comparing solving procedures through graphs

At the beginning of this lecture, we have mentioned that solving procedures can be conveniently compared through the study of their related graphs.

As an example, it is easy to see that the graph  $DPLL_F$  is a subgraph of  $GT_{F,G}$ .

# Comparing solving procedures through graphs

At the beginning of this lecture, we have mentioned that solving procedures can be conveniently compared through the study of their related graphs.

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