Lookahead Techniques

Marijn J.H. Heule

Carnegie Mellon University

http://www.cs.cmu.edu/~mheule/15816-f24/ Automated Reasoning and Satisfiability September 23, 2024 DPLL Procedure

Look-ahead Architecture

Look-ahead Learning

Autarky Reasoning

marijn@cmu.edu

DPLL Procedure

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marijn@cmu.edu

Davis Putnam Logemann Loveland [DP60,DLL62]

Recursive procedure that in each recursive call:

- Simplifies the formula (using unit propagation)
- Splits the formula into two subformulas
 - Variable selection heuristics (which variable to split on)
 - Direction heuristics (which subformula to explore first)

DPLL: Example

$$\begin{split} \Gamma_{\mathrm{DPLL}} &:= (x_1 \lor x_2 \lor \overline{x}_3) \land (\overline{x}_1 \lor x_2 \lor x_3) \land \\ (\overline{x}_1 \lor \overline{x}_2 \lor x_3) \land (x_1 \lor x_3) \land (\overline{x}_1 \lor \overline{x}_3) \end{split}$$

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DPLL: Example

 $\Gamma_{\text{DPLL}} := (\mathbf{x}_1 \lor \mathbf{x}_2 \lor \overline{\mathbf{x}}_3) \land (\overline{\mathbf{x}}_1 \lor \mathbf{x}_2 \lor \mathbf{x}_3) \land$ $(\overline{\mathbf{x}}_1 \lor \overline{\mathbf{x}}_2 \lor \mathbf{x}_3) \land (\mathbf{x}_1 \lor \mathbf{x}_3) \land (\overline{\mathbf{x}}_1 \lor \overline{\mathbf{x}}_3)$



DPLL with selection of (effective) decision variables by look-aheads on variables

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Look-ahead:

Assign a variable to a truth value

DPLL with selection of (effective) decision variables by look-aheads on variables

- Assign a variable to a truth value
- Simplify the formula

DPLL with selection of (effective) decision variables by look-aheads on variables

- Assign a variable to a truth value
- Simplify the formula
- Measure the reduction

DPLL with selection of (effective) decision variables by look-aheads on variables

- Assign a variable to a truth value
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- Measure the reduction
- Learn if possible

DPLL with selection of (effective) decision variables by look-aheads on variables

- Assign a variable to a truth value
- Simplify the formula
- Measure the reduction
- Learn if possible
- Backtrack

DPLL Procedure

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$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \overline{x}_6) \end{split}$$

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Look-ahead: Properties

- Very expensive
- Effective compared to cheap heuristics
- Detection of failed literals (and more)
- Strong on random k-SAT formulae
- Examples: march, OKsolver, kcnfs

DEMO

marijn@cmu.edu

Look-ahead: Reduction heuristics

Number of satisfied clauses

Look-ahead: Reduction heuristics

Number of satisfied clauses

Number of implied variables

Look-ahead: Reduction heuristics

- Number of satisfied clauses
- Number of implied variables
- New (reduced, not satisfied) clauses
 - Smaller clauses more important
 - Weights based on occurrences

Look-ahead: Architecture



Look-ahead: Pseudo-code of DPLL with lookahead

- 1: $\Gamma :=$ Simplify (Γ)
- 2: if Γ is empty then return satisfiable
- 3: if $\bot \in \Gamma$ then return unsatisfiable
- 4: $\langle \Gamma; l_{decision} \rangle := LookAhead (\Gamma)$
- 5: if $(\mathsf{DPLL}(\Gamma|\iota_{\mathrm{decision}})=\texttt{satisfiable})$ then
- 6: return satisfiable
- 7: return DPLL ($\Gamma | \overline{l_{decision}}$)

DPLL Procedure

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Local Learning

Look-ahead solvers do not perform global learning, in contrast to contrast to conflict-driven clause learning (CDCL) solvers

Instead, look-ahead solvers learn locally:

- Learn small (typically unit or binary) clauses that are valid for the current node and lower in the DPLL tree
- Locally learnt clauses have to be removed during backtracking

A literal l is called a failed literal if the look-ahead on l = 1 results in a conflict:

- failed literal l is forced to false followed by unit propagation
- **i** if both x and \overline{x} are failed literals, then backtrack

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$$\alpha = \{x_4 = 0, x_6 = 1, x_1 = 1\}$$

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$$\alpha = \{x_4 = 0, x_6 = 1, x_1 = 1, x_2 = 1\}$$

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$$\alpha = \{x_4 = 0, x_6 = 1, x_1 = 1, x_2 = 1, x_3 = 1\}$$

Hyper Binary Resolution [Bacchus 2002]

Definition (Hyper Binary Resolution Rule)

$$\frac{(x \lor x_1 \lor x_2 \lor \cdots \lor x_n) \ (\overline{x}_1 \lor x') \ (\overline{x}_2 \lor x') \ \dots \ (\overline{x}_n \lor x')}{(x \lor x')}$$



Hyper Binary Resolution Rule:

- combines multiple resolution steps into one
- uses one n-ary clauses and multiple binary clauses
- special case hyper unary resolution where x = x'

marijn@cmu.edu

Look-ahead: Hyper Binary Resolvents

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hyper binary resolvents: $(x_2 \lor \overline{x}_6)$ and $(x_2 \lor x_3)$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_3 \vee \mathbf{x}_4) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_2 \vee \mathbf{x}_3) \wedge \\ &(\overline{\mathbf{x}}_1 \vee \mathbf{x}_2) \wedge (\mathbf{x}_1 \vee \mathbf{x}_3 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_1 \vee \mathbf{x}_4 \vee \overline{\mathbf{x}}_5) \wedge \\ &(\mathbf{x}_1 \vee \overline{\mathbf{x}}_6) \wedge (\mathbf{x}_4 \vee \mathbf{x}_5 \vee \mathbf{x}_6) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_6) \end{split}$$

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hyper binary resolvents: $(x_2 \lor \overline{x}_6)$ and $(x_2 \lor x_3)$ Which one is more useful?

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$$\begin{split} &\Gamma_{\text{learning}} := (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_3 \vee \mathbf{x}_4) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_2 \vee \mathbf{x}_3) \wedge \\ &(\overline{\mathbf{x}}_1 \vee \mathbf{x}_2) \wedge (\mathbf{x}_1 \vee \mathbf{x}_3 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_1 \vee \mathbf{x}_4 \vee \overline{\mathbf{x}}_5) \wedge \\ &(\mathbf{x}_1 \vee \overline{\mathbf{x}}_6) \wedge (\mathbf{x}_4 \vee \mathbf{x}_5 \vee \mathbf{x}_6) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_6) \\ &\alpha = \{\mathbf{x}_1 = \mathbf{1}, \mathbf{x}_2 = \mathbf{1}, \mathbf{x}_3 = \mathbf{1}, \mathbf{x}_4 = \mathbf{1}\} \\ &\Gamma_{\text{learning}} := (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_3 \vee \mathbf{x}_4) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_2 \vee \mathbf{x}_3) \wedge \\ &(\overline{\mathbf{x}}_1 \vee \mathbf{x}_2) \wedge (\mathbf{x}_1 \vee \mathbf{x}_3 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_1 \vee \mathbf{x}_4 \vee \overline{\mathbf{x}}_5) \wedge \\ &(\mathbf{x}_1 \vee \overline{\mathbf{x}}_6) \wedge (\mathbf{x}_4 \vee \mathbf{x}_5 \vee \mathbf{x}_6) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_6) \\ &\alpha = \{\mathbf{x}_1 = \mathbf{0}\} \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} \coloneqq (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_3 \vee \mathbf{x}_4) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_2 \vee \mathbf{x}_3) \wedge \\ &(\overline{\mathbf{x}}_1 \vee \mathbf{x}_2) \wedge (\mathbf{x}_1 \vee \mathbf{x}_3 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_1 \vee \mathbf{x}_4 \vee \overline{\mathbf{x}}_5) \wedge \\ &(\mathbf{x}_1 \vee \overline{\mathbf{x}}_6) \wedge (\mathbf{x}_4 \vee \mathbf{x}_5 \vee \mathbf{x}_6) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_6) \\ &\alpha = \{\mathbf{x}_1 = \mathbf{1}, \mathbf{x}_2 = \mathbf{1}, \mathbf{x}_3 = \mathbf{1}, \mathbf{x}_4 = \mathbf{1}\} \\ &\Gamma_{\text{learning}} \coloneqq (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_3 \vee \mathbf{x}_4) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_2 \vee \mathbf{x}_3) \wedge \\ &(\overline{\mathbf{x}}_1 \vee \mathbf{x}_2) \wedge (\mathbf{x}_1 \vee \mathbf{x}_3 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_1 \vee \mathbf{x}_4 \vee \overline{\mathbf{x}}_5) \wedge \\ &(\mathbf{x}_1 \vee \overline{\mathbf{x}}_6) \wedge (\mathbf{x}_4 \vee \mathbf{x}_5 \vee \mathbf{x}_6) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_6) \\ &\alpha = \{\mathbf{x}_1 = \mathbf{0}, \mathbf{x}_6 = \mathbf{0}\} \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} \coloneqq (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_3 \vee \mathbf{x}_4) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_2 \vee \mathbf{x}_3) \wedge \\ &(\overline{\mathbf{x}}_1 \vee \mathbf{x}_2) \wedge (\mathbf{x}_1 \vee \mathbf{x}_3 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_1 \vee \mathbf{x}_4 \vee \overline{\mathbf{x}}_5) \wedge \\ &(\mathbf{x}_1 \vee \overline{\mathbf{x}}_6) \wedge (\mathbf{x}_4 \vee \mathbf{x}_5 \vee \mathbf{x}_6) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_6) \\ &\alpha = \{\mathbf{x}_1 = \mathbf{1}, \mathbf{x}_2 = \mathbf{1}, \mathbf{x}_3 = \mathbf{1}, \mathbf{x}_4 = \mathbf{1}\} \\ &\Gamma_{\text{learning}} \coloneqq (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_3 \vee \mathbf{x}_4) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_2 \vee \mathbf{x}_3) \wedge \\ &(\overline{\mathbf{x}}_1 \vee \mathbf{x}_2) \wedge (\mathbf{x}_1 \vee \mathbf{x}_3 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_1 \vee \mathbf{x}_4 \vee \overline{\mathbf{x}}_5) \wedge \\ &(\mathbf{x}_1 \vee \overline{\mathbf{x}}_6) \wedge (\mathbf{x}_4 \vee \mathbf{x}_5 \vee \mathbf{x}_6) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_6) \end{split}$$

 $\alpha = \{x_1 = 0, x_6 = 0, x_3 = 1\}$

Stålmarck's Method

In short, Stålmarck's Method is a procedure that generalizes the concept of necessary assignments.

For each variable x, $(\operatorname{Simplify}(\Gamma|_{\mathfrak{X}}) \cap \operatorname{Simplify}(\Gamma|_{\overline{\mathfrak{X}}})) \setminus \Gamma$ is added to Γ .

The above is repeated until fixpoint, i.e., until $\forall x : (Simplify(\Gamma|\chi) \cap Simplify(\Gamma|\overline{\chi})) \setminus F = \emptyset$

Afterwards the procedure is repeated using all pairs for variables x and y: Add (Simplify($\Gamma|_{x}y) \cap$ Simplify($\Gamma|_{\overline{x}\overline{y}}) \cap$ Simplify($\Gamma|_{\overline{x}\overline{y}} \cap$ Simplify($\Gamma|_{\overline{x}\overline{y}})$) \ Γ to Γ .

The second round is very expensive and can typically not be finished in reasonable time.

DPLL Procedure

Look-ahead Architecture

Look-ahead Learning

Autarky Reasoning

An autarky is a partial assignment that satisfies all clauses that are "touched" by the assignment

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An 1-autarky is a partial assignment that satisfies all touched clauses except one

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \overline{x}_6) \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \lor \overline{x}_3 \lor x_4) \land (\overline{x}_1 \lor \overline{x}_2 \lor x_3) \land \\ &(\overline{x}_1 \lor x_2) \land (x_1 \lor x_3 \lor x_6) \land (\overline{x}_1 \lor x_4 \lor \overline{x}_5) \land \\ &(x_1 \lor \overline{x}_6) \land (x_4 \lor x_5 \lor x_6) \land (x_5 \lor \overline{x}_6) \\ &\alpha = \{x_1 = 1\} \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_3 \vee \mathbf{x}_4) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_2 \vee \mathbf{x}_3) \wedge \\ &(\overline{\mathbf{x}}_1 \vee \mathbf{x}_2) \wedge (\mathbf{x}_1 \vee \mathbf{x}_3 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_1 \vee \mathbf{x}_4 \vee \overline{\mathbf{x}}_5) \wedge \\ &(\mathbf{x}_1 \vee \overline{\mathbf{x}}_6) \wedge (\mathbf{x}_4 \vee \mathbf{x}_5 \vee \mathbf{x}_6) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_6) \\ &\alpha = \{\mathbf{x}_1 = \mathbf{1}, \mathbf{x}_2 = \mathbf{1}\} \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \overline{x}_6) \\ &\alpha = \{x_1 = 1, x_2 = 1, x_3 = 1\} \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \overline{x}_6) \\ &\alpha = \{x_1 = 1, x_2 = 1, x_3 = 1, x_4 = 1\} \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_3 \vee \mathbf{x}_4) \wedge (\overline{\mathbf{x}}_1 \vee \overline{\mathbf{x}}_2 \vee \mathbf{x}_3) \wedge \\ &(\overline{\mathbf{x}}_1 \vee \mathbf{x}_2) \wedge (\mathbf{x}_1 \vee \mathbf{x}_3 \vee \mathbf{x}_6) \wedge (\overline{\mathbf{x}}_1 \vee \mathbf{x}_4 \vee \overline{\mathbf{x}}_5) \wedge \\ &(\mathbf{x}_1 \vee \overline{\mathbf{x}}_6) \wedge (\mathbf{x}_4 \vee \mathbf{x}_5 \vee \mathbf{x}_6) \wedge (\mathbf{x}_5 \vee \overline{\mathbf{x}}_6) \end{split}$$

$$\alpha = \{x_1 = 1, x_2 = 1, x_3 = 1, x_4 = 1\}$$

 $\Gamma_{\rm learning}$ satisfiability equivalent to $(x_5 \vee \overline{x}_6)$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{\mathbf{x}}_1 \lor \overline{\mathbf{x}}_3 \lor \mathbf{x}_4) \land (\overline{\mathbf{x}}_1 \lor \overline{\mathbf{x}}_2 \lor \mathbf{x}_3) \land \\ &(\overline{\mathbf{x}}_1 \lor \mathbf{x}_2) \land (\mathbf{x}_1 \lor \mathbf{x}_3 \lor \mathbf{x}_6) \land (\overline{\mathbf{x}}_1 \lor \mathbf{x}_4 \lor \overline{\mathbf{x}}_5) \land \\ &(\mathbf{x}_1 \lor \overline{\mathbf{x}}_6) \land (\mathbf{x}_4 \lor \mathbf{x}_5 \lor \mathbf{x}_6) \land (\mathbf{x}_5 \lor \overline{\mathbf{x}}_6) \end{split}$$

$$\alpha = \{x_1 = 1, x_2 = 1, x_3 = 1, x_4 = 1\}$$

 $\Gamma_{\rm learning}$ satisfiability equivalent to $(x_5 \vee \overline{x}_6)$

Could reduce computational cost on UNSAT

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \overline{x}_6) \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \overline{x}_6) \\ &\alpha = \{x_2 = 0\} \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \overline{x}_6) \\ &\alpha = \{x_2 = 0, x_1 = 0\} \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \overline{x}_6) \\ &\alpha = \{x_2 = 0, x_1 = 0, x_6 = 0\} \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \overline{x}_6) \\ &\alpha = \{x_2 = 0, x_1 = 0, x_6 = 0, x_3 = 1\} \end{split}$$

$$\begin{split} &\Gamma_{\text{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \overline{x}_6) \end{split}$$

$$\alpha = \{x_2 = 0, x_1 = 0, x_6 = 0, x_3 = 1\}$$

(local) 1-autarky resolvents: $(\overline{x}_2 \lor \overline{x}_4)$ and $(\overline{x}_2 \lor \overline{x}_5)$

Look-ahead: Autarky or Conflict on 2-SAT Formulae

Lookahead techniques can solve 2-SAT formulae in polynomial time. Each lookahead on l results: 1. in an autarky: forcing l to be true 2. in a conflict: forcing l to be false
Look-ahead: Autarky or Conflict on 2-SAT Formulae

Lookahead techniques can solve 2-SAT formulae in polynomial time. Each lookahead on l results: 1. in an autarky: forcing l to be true 2. in a conflict: forcing l to be false

SAT Game

by Olivier Roussel

http://www.cs.utexas.edu/~marijn/game/2SAT