Lookahead Techniques

Ruben Martins

Carnegie Mellon University

http://www.cs.cmu.edu/~rubenm/15816-f25/ Automated Reasoning and Satisfiability September 22, 2024

DPLL Procedure

Look-ahead Architecture

Look-ahead Learning

Autarky Reasoning

DPLL Procedure

Look-ahead Architecture

Look-ahead Learning

Autarky Reasoning

SAT Solving: DPLL

Davis Putnam Logemann Loveland [DP60,DLL62]

Recursive procedure that in each recursive call:

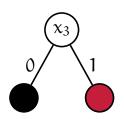
- Simplifies the formula (using unit propagation)
- Splits the formula into two subformulas
 - Variable selection heuristics (which variable to split on)
 - Direction heuristics (which subformula to explore first)

DPLL: Example

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

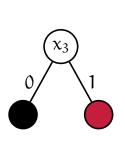
DPLL: Example

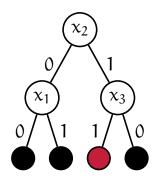
$$\Gamma_{\mathrm{DPLL}} := (x_1 \vee x_2 \vee \overline{x}_3) \wedge (\overline{x}_1 \vee x_2 \vee x_3) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge (x_1 \vee x_3) \wedge (\overline{x}_1 \vee \overline{x}_3)$$



DPLL: Example

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





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

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

Look-ahead:

Assign a variable to a truth value

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

Look-ahead:

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

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

Look-ahead:

- 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

Look-ahead:

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

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

Look-ahead:

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

DPLL Procedure

Look-ahead Architecture

Look-ahead Learning

Autarky Reasoning

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

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

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

$$\alpha = \{x_2 = 0, x_1 = 0\}$$

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

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

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

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

 ${\tt rubenm@cmu.edu} \hspace*{100mm} 10 \hspace*{0.5mm} / \hspace*{0.5mm} 25$

Look-ahead: Reduction heuristics

■ Number of satisfied clauses

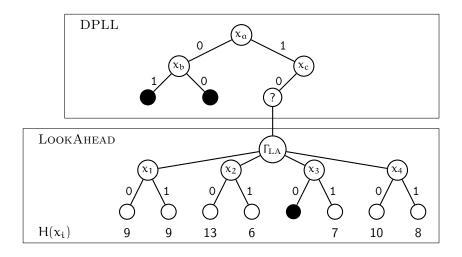
Look-ahead: Reduction heuristics

- 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

```
    Γ := Simplify (Γ)
    if Γ is empty then return satisfiable
    if ⊥ ∈ Γ then return unsatisfiable
    ⟨Γ; l<sub>decision</sub>⟩ := LookAhead (Γ)
    if (DPLL(Γ|l<sub>decision</sub>) = satisfiable) then
    return satisfiable
    return DPLL (Γ|l<sub>decision</sub>)
```

DPLL Procedure

Look-ahead Architecture

Look-ahead Learning

Autarky Reasoning

Local Learning

Look-ahead solvers do not perform global learning, in 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
- if both x and \overline{x} are failed literals, then backtrack

Failed literals can be generalized by double lookahead: assign two literals and learn a binary clause in case of a conflict.

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
- if both x and \overline{x} are failed literals, then backtrack

Failed literals can be generalized by double lookahead: assign two literals and learn a binary clause in case of a conflict.

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

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
- if both x and \overline{x} are failed literals, then backtrack

Failed literals can be generalized by double lookahead: assign two literals and learn a binary clause in case of a conflict.

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

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

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
- if both x and \overline{x} are failed literals, then backtrack

Failed literals can be generalized by double lookahead: assign two literals and learn a binary clause in case of a conflict.

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

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

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
- if both x and \overline{x} are failed literals, then backtrack

Failed literals can be generalized by double lookahead: assign two literals and learn a binary clause in case of a conflict.

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

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

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
- if both x and \overline{x} are failed literals, then backtrack

Failed literals can be generalized by double lookahead: assign two literals and learn a binary clause in case of a conflict.

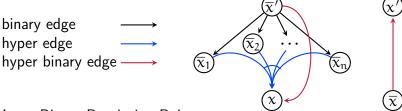
$$\begin{split} &\Gamma_{\mathrm{learning}} := (\overline{x}_1 \vee \overline{x}_3 \vee x_4) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge \\ &(\overline{x}_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\overline{x}_1 \vee x_4 \vee \overline{x}_5) \wedge \\ &(x_1 \vee \overline{x}_6)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

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

Hyper Binary Resolution [Bacchus 2002]

Definition (Hyper Binary Resolution Rule)

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



Hyper Binary Resolution Rule:

- combines multiple resolution steps into one
- uses one n-ary clauses and multiple binary clauses

 \blacksquare special case hyper unary resolution where x = x'

Look-ahead: Hyper Binary Resolvents

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

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

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

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

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

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

hyper binary resolvents:

$$(x_2 \lor \overline{x}_6)$$
 and $(x_2 \lor x_3)$

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

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

hyper binary resolvents:

$$(x_2 \vee \overline{x}_6)$$
 and $(x_2 \vee x_3)$

Which one is more useful?

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

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

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

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

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

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

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

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

$$\alpha = \{\mathbf{x}_1 = \mathbf{1}, \mathbf{x}_2 = \mathbf{1}, \mathbf{x}_3 = \mathbf{1}, \mathbf{x}_4 = \mathbf{1}\}$$

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

$$\alpha = \{\mathbf{x}_1 = \mathbf{0}, \mathbf{x}_6 = \mathbf{0}\}$$

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

$$\alpha = \{\mathbf{x}_1 = \mathbf{1}, \mathbf{x}_2 = \mathbf{1}, \mathbf{x}_3 = \mathbf{1}, \mathbf{x}_4 = \mathbf{1}\}$$

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

$$\alpha = \{\mathbf{x}_1 = \mathbf{0}, \mathbf{x}_6 = \mathbf{0}, \mathbf{x}_3 = \mathbf{1}\}$$

Stålmarck's Method

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

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

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

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

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

a pure literal is an autarky

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

- a pure literal is an autarky
- each satisfying assignment is an autarky

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

- a pure literal is an autarky
- each satisfying assignment is an autarky
- the remaining formula is satisfiability equivalent to the original formula

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

- a pure literal is an autarky
- each satisfying assignment is an autarky
- the remaining formula is satisfiability equivalent to the original formula

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

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

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

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

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

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

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

$$\alpha = \{\mathbf{x}_1 = 1, \mathbf{x}_2 = 1, \mathbf{x}_3 = 1, \mathbf{x}_4 = 1\}$$

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

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

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

Could reduce computational cost on UNSAT

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

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

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

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

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

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

(local) 1-autarky resolvents:

$$(\overline{x}_2 \vee \overline{x}_4)$$
 and $(\overline{x}_2 \vee \overline{x}_5)$

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 1 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