

Practical SAT Solving: Look-ahead Techniques

Marijn J.H. Heule
Warren A. Hunt Jr.

The University of Texas at Austin

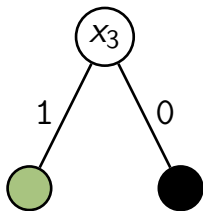
Davis Putnam Logemann Loveland [DP60,DLL62]

- Simplify (Unit Propagation)
- Split the formula
 - Variable Selection Heuristics
 - Direction heuristics

$$\mathcal{F}_{\text{DPLL}} := (x_1 \vee x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2 \vee x_3) \wedge \\ (\neg x_1 \vee \neg x_2 \vee x_3) \wedge (x_1 \vee x_3) \wedge (\neg x_1 \vee \neg x_3)$$

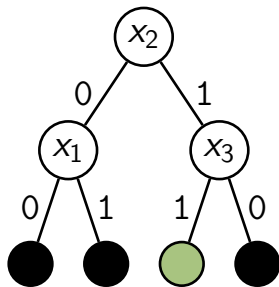
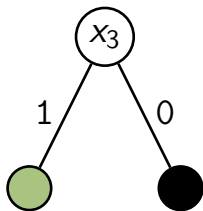
DPLL: Example

$$\mathcal{F}_{\text{DPLL}} := (x_1 \vee x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2 \vee x_3) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge (x_1 \vee x_3) \wedge (\neg x_1 \vee \neg x_3)$$



DPLL: Example

$$\mathcal{F}_{\text{DPLL}} := (x_1 \vee x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2 \vee x_3) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge (x_1 \vee x_3) \wedge (\neg x_1 \vee \neg x_3)$$



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

Look-ahead: Example

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

Look-ahead: Example

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

$$\varphi = \{x_2=0\}$$

Look-ahead: Example

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

Look-ahead: Example

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

Look-ahead: Example

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

- Very expensive

Look-ahead: Properties

- Very expensive
- Effective compared to cheap heuristics

- Very expensive
- Effective compared to cheap heuristics
- Detection of failed literals (and more)

- Very expensive
- Effective compared to cheap heuristics
- Detection of failed literals (and more)
- Strong on random k -SAT formulae

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

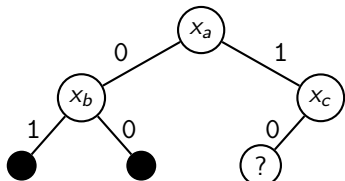
- Number of satisfied clauses

- Number of satisfied clauses
- Number of implied variables

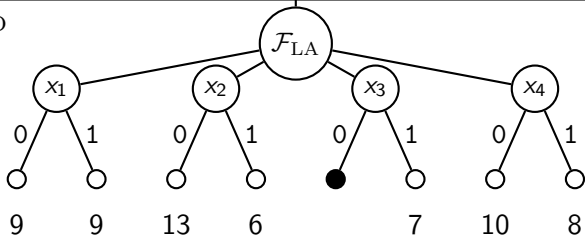
- Number of satisfied clauses
- Number of implied variables
- New (reduced, not satisfied) clauses
 - Smaller clauses more important
 - Weights based on occurring

Look-ahead: Architecture

DPLL



LOOKAHEAD



```
1:  $\mathcal{F} := \text{SIMPLIFY}(\mathcal{F})$ 
2: if  $\mathcal{F}$  is empty then return satisfiable
3: if  $\emptyset \in \mathcal{F}$  then return unsatisfiable
4:  $\langle \mathcal{F}; l_{\text{decision}} \rangle := \text{LOOKAHEAD}(\mathcal{F})$ 
5: if  $\text{DPLL}(\mathcal{F}(l_{\text{decision}} \leftarrow 1)) = \text{satisfiable}$  then
6:   return satisfiable
7: end if
8: return  $\text{DPLL}(\mathcal{F}(l_{\text{decision}} \leftarrow 0))$ 
```

$$\begin{aligned}\mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)\end{aligned}$$

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

$$\varphi = \{x_2=0\}$$

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

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

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

(local) constraint resolvents

a.k.a. hyper binary resolvents:

$(x_2 \vee x_3)$ and $(x_2 \vee \neg x_6)$

Look-ahead: Necessary assignments

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

Look-ahead: Necessary assignments

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

$$\varphi = \{x_1=1\}$$

Look-ahead: Necessary assignments

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

$$\varphi = \{x_1=1, x_2=1\}$$

Look-ahead: Necessary assignments

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

$$\varphi = \{x_1=1, x_2=1, x_3=1\}$$

Look-ahead: Necessary assignments

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

Look-ahead: Necessary assignments

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

Look-ahead: Necessary assignments

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

$$\varphi = \{x_1=0\}$$

Look-ahead: Necessary assignments

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

$$\varphi = \{x_1=0, x_6=0\}$$

Look-ahead: Necessary assignments

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

Look-ahead: Necessary assignments

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

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

Look-ahead: Autarky detection

$$\begin{aligned}\mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)\end{aligned}$$

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

$$\varphi = \{x_1=1\}$$

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

$$\varphi = \{x_1=1, x_2=1\}$$

Look-ahead: Autarky detection

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

$$\varphi = \{x_1=1, x_2=1, x_3=1\}$$

Look-ahead: Autarky detection

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

$\mathcal{F}_{\text{learning}}$ satisfiability equivalent to $(x_5 \vee \neg x_6)$

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

$\mathcal{F}_{\text{learning}}$ satisfiability equivalent to $(x_5 \vee \neg x_6)$

Could reduce computational cost on UNSAT

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

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>

Look-ahead: 1-Autarky learning

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

Look-ahead: 1-Autarky learning

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

$$\varphi = \{x_2=0\}$$

Look-ahead: 1-Autarky learning

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

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

Look-ahead: 1-Autarky learning

$$\begin{aligned} \mathcal{F}_{\text{learning}} := & (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ & (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ & (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6) \end{aligned}$$

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

Look-ahead: 1-Autarky learning

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

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

Look-ahead: 1-Autarky learning

$$\mathcal{F}_{\text{learning}} := (\neg x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_1 \vee \neg x_2 \vee x_3) \wedge \\ (\neg x_1 \vee x_2) \wedge (x_1 \vee x_3 \vee x_6) \wedge (\neg x_1 \vee x_4 \vee \neg x_5) \wedge \\ (x_1 \vee \neg x_6) \wedge (x_4 \vee x_5 \vee x_6) \wedge (x_5 \vee \neg x_6)$$

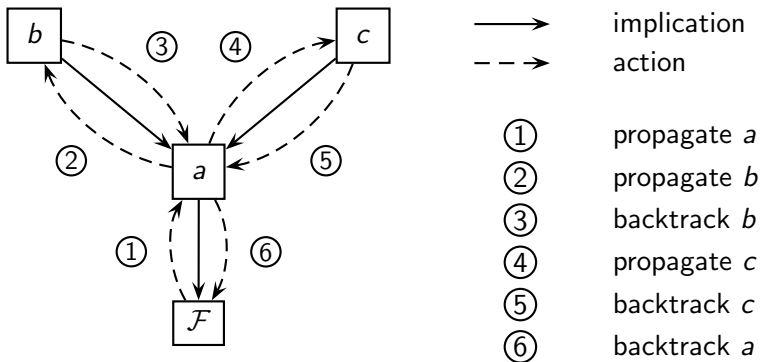
$$\varphi = \{x_2=0, x_1=0, x_6=0, x_3=1\}$$

(local) 1-autarky resolvents:

$$(\neg x_2 \vee \neg x_4) \text{ and } (\neg x_2 \vee \neg x_5)$$

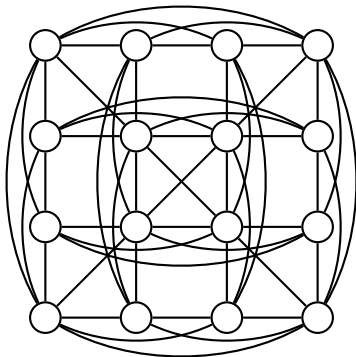
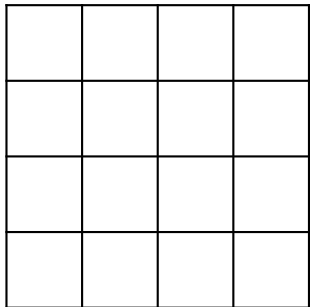
Tree-based Look-ahead

Given a formula F which includes the clauses $(a \vee \bar{b})$ and $(a \vee \bar{c})$,
tree-based look-ahead can reduce the costs of look-aheads.



Homework puzzle

Populate the matrix with the following numbers:
10, 40, 50, 12, 21, 31, 35, 43, 46, 56, 74, 75, 83, 87, 89, 98
such that no digit repeats in any row, column or diagonal.



Practical SAT Solving: Look-ahead Techniques

Marijn J.H. Heule
Warren A. Hunt Jr.

The University of Texas at Austin