Lookahead Techniques

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Automated Reasoning and Satisfiability
October 5, 2020

DPLL Procedure

Look-ahead Architecture

Look-ahead Learning

Autarky Reasoning

Tree-based Look-ahead

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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

$$\mathcal{F}_{\text{DPLL}} := (x_1 \vee x_2 \vee \overline{x}_3) \wedge (\overline{x}_1 \vee x_2 \vee x_3) \wedge (\overline{x}_1 \vee \overline{x}_2 \vee x_3) \wedge (x_1 \vee x_3) \wedge (\overline{x}_1 \vee \overline{x}_3)$$

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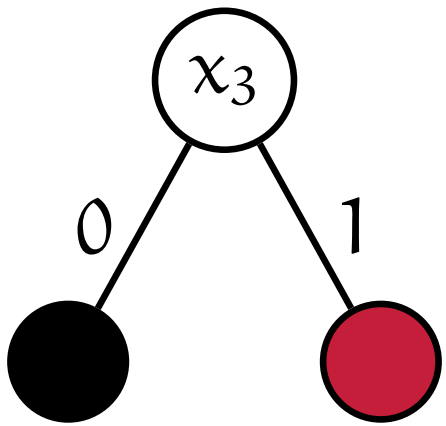

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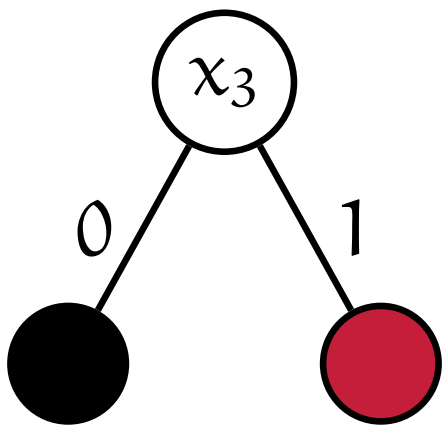

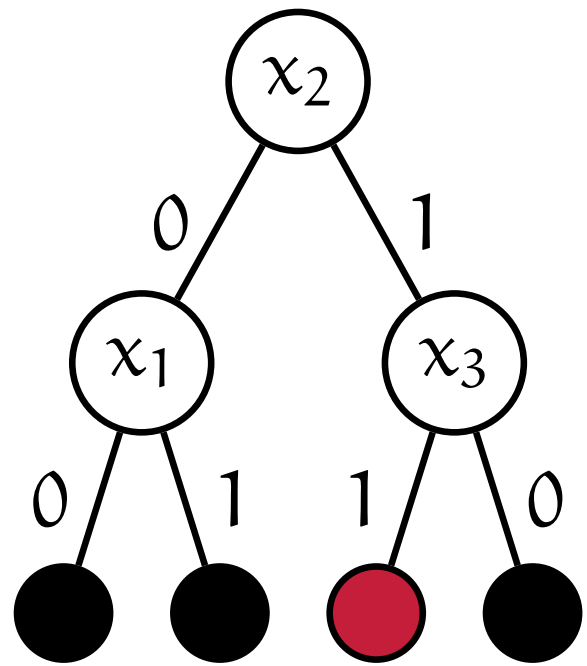

Slightly Harder Example

Construct a DPLL tree for:

$$(a \vee b \vee \overline{c}) \wedge (\overline{a} \vee \overline{b} \vee c) \wedge$$
$$(b \vee c \vee \overline{d}) \wedge (\overline{b} \vee \overline{c} \vee d) \wedge$$
$$(a \vee c \vee d) \wedge (\overline{a} \vee \overline{c} \vee \overline{d}) \wedge$$
$$(\overline{a} \vee b \vee d)$$

Look-ahead: Definition

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

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- Assign a variable to a truth value

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Look-ahead:

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- Backtrack

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- Strong on random k-SAT formulae
- Examples: march, OKsolver, kcnfs

DEMO

Look-ahead: Reduction heuristics

- Number of satisfied clauses

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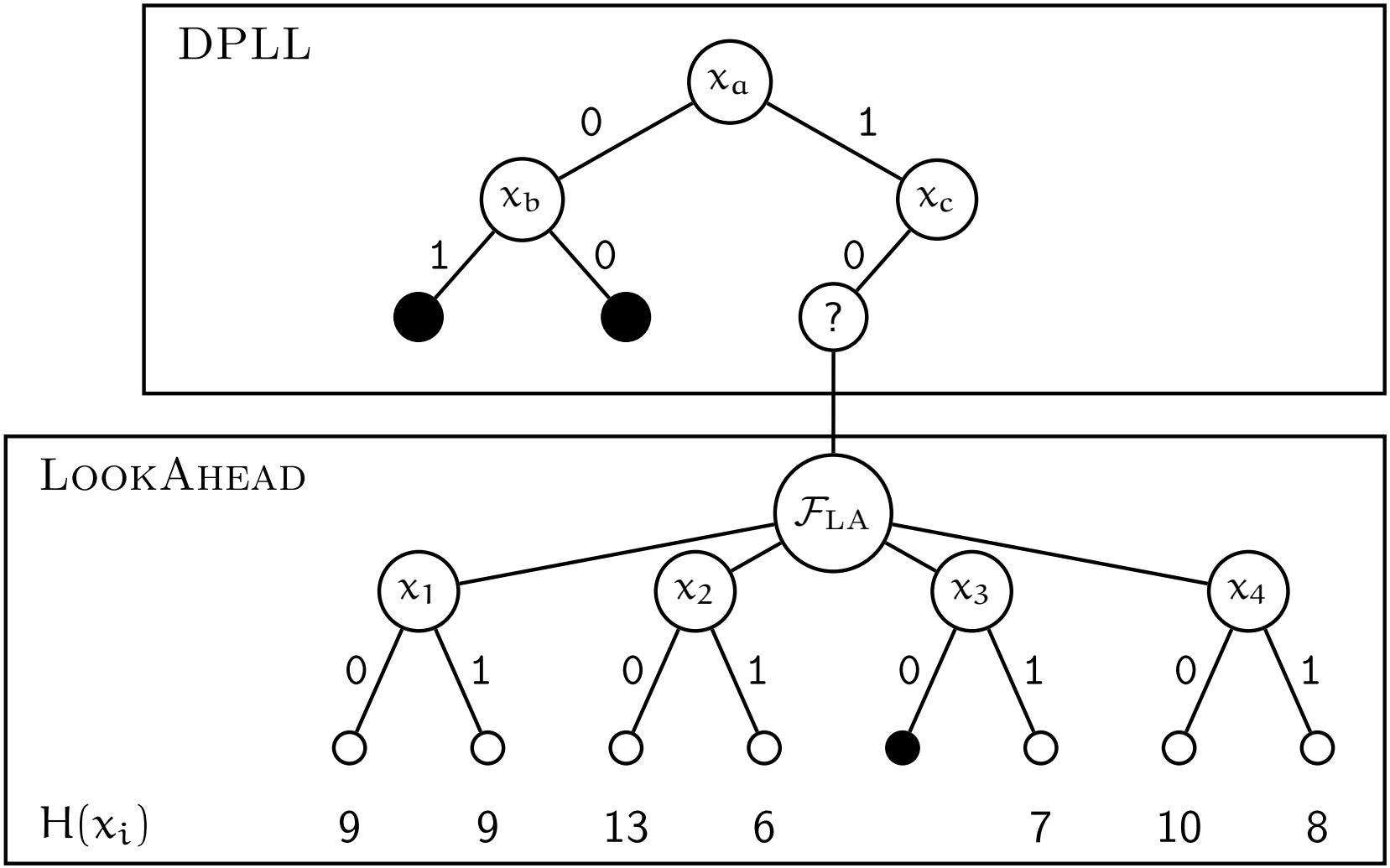
- Number of implied variables

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: $\mathcal{F} :=$ Simplify $(\mathcal{F})$
2: **if** $\mathcal{F}$ is empty **then return** `satisfiable`
3: **if** $\bot \in \mathcal{F}$ **then return** `unsatisfiable`
4: $\langle \mathcal{F};\ \mathsf{l}_{\text{decision}} \rangle :=$ LookAhead $(\mathcal{F})$
5: **if** $(\text{DPLL}(\ \mathcal{F}(\mathsf{l}_{\text{decision}} \leftarrow 1)) = \texttt{satisfiable})$ **then**
6: **return** `satisfiable`
7: **return** DPLL $(\mathcal{F}(\mathsf{l}_{\text{decision}} \leftarrow 0))$

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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

Failed Literals and Double Look-aheads

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.

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Hyper Binary Resolution [Bacchus 2002]

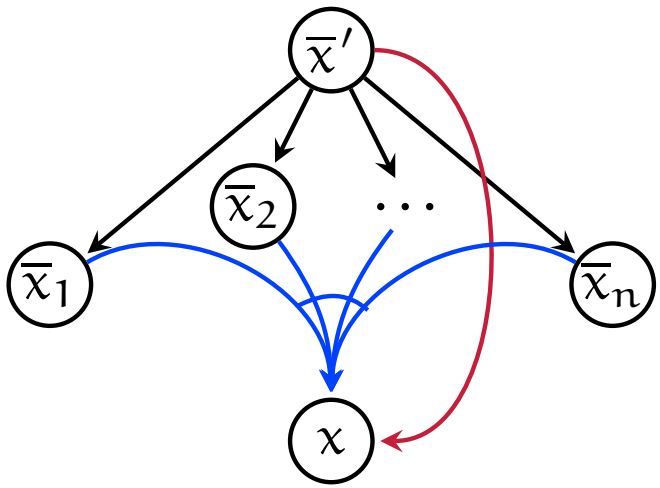
Definition (Hyper Binary Resolution Rule)

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

binary edge →

hyper edge →

hyper binary edge →



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 $x = x'$

Look-ahead: Hyper Binary Resolvents

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hyper binary resolvents:

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

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Which one is more useful?

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

Look-ahead: Necessary assignments

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

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

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

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

Stålmarck's Method

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

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

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

Afterwards the procedure is repeated using all pairs for variables x and y: Add $(\text{Simplify}(F|xy) \cap \text{Simplify}(F|x\overline{y}) \cap \text{Simplify}(F|\overline{x}y \cap \text{Simplify}(F|\overline{xy})) \setminus F$ to F.

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

Tree-based Look-ahead

Look-ahead: Autarky definition

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

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Look-ahead: Autarky definition

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

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

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$$\alpha = \{x_1 = 1\}$$

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

Look-ahead: Autarky detection

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

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$\mathcal{F}_{\text{learning}}$ satisfiability equivalent to $(x_5 \vee \overline{x}_6)$

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$\mathcal{F}_{\text{learning}}$ satisfiability equivalent to $(x_5 \vee \overline{x}_6)$

Could reduce computational cost on UNSAT

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

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SAT Game

by Olivier Roussel

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

Look-ahead: 1-Autarky learning

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

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

Look-ahead: 1-Autarky learning

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

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

Look-ahead: 1-Autarky learning

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

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

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

DPLL Procedure

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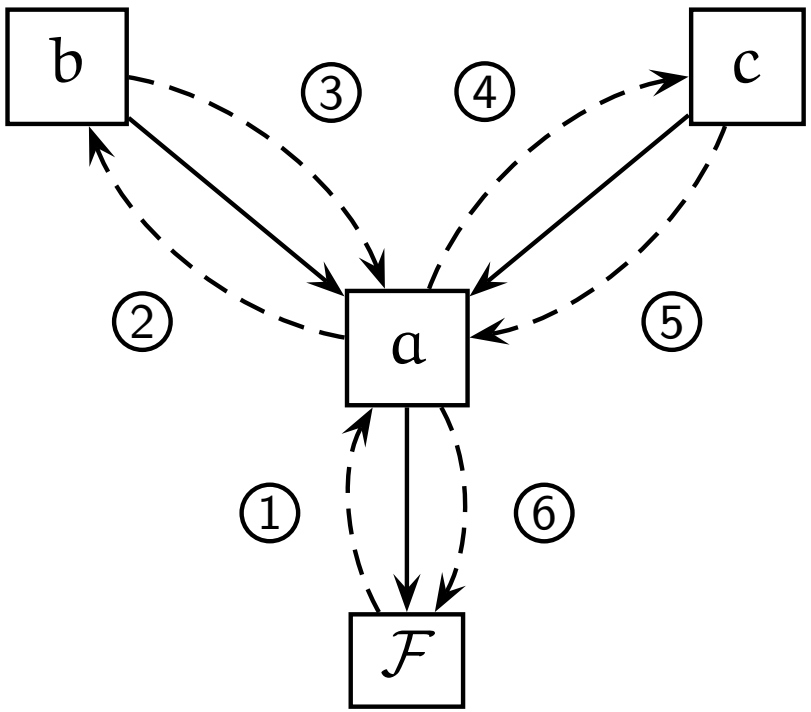
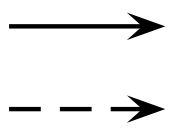
Tree-based Look-ahead

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





implication

action

① propagate a

② propagate b

③ backtrack b

④ propagate c

⑤ backtrack c

⑥ backtrack a

Lookahead Techniques

Marijn J.H. Heule

Carnegie Mellon University

http://www.cs.cmu.edu/~mheule/15816-f20/

https://cmu.zoom.us/j/93095736668

Automated Reasoning and Satisfiability
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