



## SAT and SMT Solving

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### Approach

- ▶ most state-of-the-art SAT solvers use variation of Davis - Putnam - Logemann - Loveland (DPLL) procedure (1962)
- ▶ DPLL is sound and complete backtracking-based search algorithm
- ▶ can be described abstractly by transition system (Nieuwenhuis, Oliveras, Tinelli 2006)

### Definition (Abstract DPLL)

- ▶ **decision literal** is annotated literal  $l^d$
- ▶ **state** is pair  $M \parallel F$  for
  - ▶ list  $M$  of (decision) literals
  - ▶ formula  $F$  in CNF
- ▶ transition rules

$$M \parallel F \implies M' \parallel F' \text{ or } \text{FailState}$$

## Outline

- Summary of Last Week
- Conflict Analysis
- Conflict Driven Clause Learning
- Application: Test Case Generation

1

### Definition (DPLL Transition Rules)

- ▶ **unit propagation**  $M \parallel F, C \vee l \implies M l \parallel F, C \vee l$   
if  $M \models \neg C$  and  $l$  is undefined in  $M$
- ▶ **pure literal**  $M \parallel F \implies M l \parallel F$   
if  $l$  occurs in  $F$  but  $l^c$  does not occur in  $F$ , and  $l$  is undefined in  $M$
- ▶ **decide**  $M \parallel F \implies M l^d \parallel F$   
if  $l$  or  $l^c$  occurs in  $F$ , and  $l$  is undefined in  $M$
- ▶ **backtrack**  $M l^d N \parallel F, C \implies M l^c \parallel F, C$   
if  $M l^d N \models \neg C$  and  $N$  contains no decision literals
- ▶ **fail**  $M \parallel F, C \implies \text{FailState}$   
if  $M \models \neg C$  and  $M$  contains no decision literals
- ▶ **backjump**  $M l^d N \parallel F, C \implies M l' \parallel F, C$   
if  $M l^d N \models \neg C$  and  $\exists$  clause  $C' \vee l'$  such that
  - ▶  $F, C \models C' \vee l'$  backjump clause
  - ▶  $M \models \neg C'$  and  $l'$  is undefined in  $M$ , and  $l'$  or  $l'^c$  occurs in  $F$  or in  $M l^d N$

## Definition

basic DPLL  $\mathcal{B}$  consists of unit propagation, decide, fail, and backjump

## Theorem (Termination)

there are *no infinite derivations*  $\parallel F \implies_{\mathcal{B}} S_1 \implies_{\mathcal{B}} S_2 \implies_{\mathcal{B}} \dots$

## Theorem (Correctness)

for derivation with *final state*  $S_n$ :

$$\parallel F \implies_{\mathcal{B}} S_1 \implies_{\mathcal{B}} S_2 \implies_{\mathcal{B}} \dots \implies_{\mathcal{B}} S_n$$

- ▶ if  $S_n = \text{FailState}$  then  $F$  is *unsatisfiable*
- ▶ if  $S_n = M \parallel F'$  then  $F$  is *satisfiable* and  $M \models F'$

4

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6

## Definition

polarity of subformula  $\varphi$  in  $\psi$  is + if number of negations above  $\varphi$  in  $\psi$  is even, and - otherwise

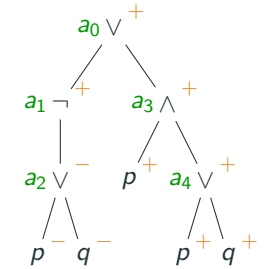
## Example (Efficient Transformations to CNF)

▶  $\varphi = \neg(p \vee q) \vee (p \wedge (p \vee q))$

▶ use fresh propositional variable for every connective

$$a_0: \neg(p \vee q) \vee (p \wedge (p \vee q)) \quad a_1: \neg(p \vee q)$$

$$a_2: p \vee q \quad a_3: p \wedge (p \vee q)$$



▶ Tseitin: add clause  $a_0$  plus  $(a_i \leftrightarrow \dots)$  for every subformula

$$\varphi \approx a_0 \wedge (a_0 \leftrightarrow a_1 \vee a_3) \wedge (a_1 \leftrightarrow \neg a_2) \wedge (a_2 \leftrightarrow p \vee q) \wedge (a_3 \leftrightarrow p \wedge a_2)$$

▶ Plaisted & Greenbaum:  $(a_i \rightarrow \dots)$  if polarity of  $a_i$  is + and  $(a_i \leftarrow \dots)$  if -

$$\varphi \approx a_0 \wedge (a_0 \rightarrow a_1 \vee a_3) \wedge (a_1 \rightarrow \neg a_2) \wedge (a_2 \leftarrow p \vee q) \wedge (a_3 \rightarrow p \wedge a_4) \wedge (a_4 \rightarrow p \vee q)$$

▶ replace  $\leftrightarrow$  and  $\rightarrow$  by 2 or 3 clauses each

5

## Backjump: Idea

- ▶ backjump clause  $C' \vee I'$  is entailed by formula (magically detected)
- ▶ prefix  $M$  of current literal list entails  $\neg C'$

## Backjump to Definition

- ▶ **backjump**  $M I^d N \parallel F, C \implies M I' \parallel F, C$   
if  $M I^d N \models \neg C$  and  $\exists$  clause  $C' \vee I'$  such that
  - ▶  $F, C \models C' \vee I'$  backjump clause
  - ▶  $M \models \neg C'$  and  $I'$  is undefined in  $M$ , and  $I'$  or  $I'^c$  occurs in  $F$  or in  $M I^d N$

## Example

$$\underbrace{1^d 2}_{M} \underbrace{3^d}_{I} \underbrace{4^d 5}_{N} \parallel \bar{1} \vee 2, \bar{1} \vee \bar{3} \vee 4 \vee 5, \bar{2} \vee \bar{4} \vee \bar{5}, 4 \vee \bar{5}, \bar{4} \vee 5, \bar{1} \vee \bar{5} \vee 6, \bar{2} \vee \bar{5} \vee \bar{6}$$

$$\xrightarrow{\text{backjump}} 1^d 2 \bar{5} \parallel \bar{1} \vee 2, \bar{1} \vee \bar{3} \vee 4 \vee 5, \bar{2} \vee \bar{4} \vee \bar{5}, 4 \vee \bar{5}, \bar{4} \vee 5, \bar{1} \vee \bar{5} \vee 6, \bar{2} \vee \bar{5} \vee \bar{6}$$

$$M = 1^d 2 \quad I = 3 \quad N = 4^d \bar{5} \quad C = \bar{4} \vee 5 \quad C' = \bar{1} \quad I' = \bar{5}$$

- ▶  $1^d 2 3^d 4^d \bar{5} \models \neg(\bar{4} \vee 5)$
- ▶ backjump clause  $C' \vee I' = \bar{1} \vee \bar{5}$  satisfies  $F, C \models C' \vee I'$
- ▶  $1^d 2 \models 1$ , and  $5$  is undefined in  $1^d 2$  but occurs in  $F$

7

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8

## Desirable Properties of Backjump Clauses

- ▶ small
- ▶ should trigger progress

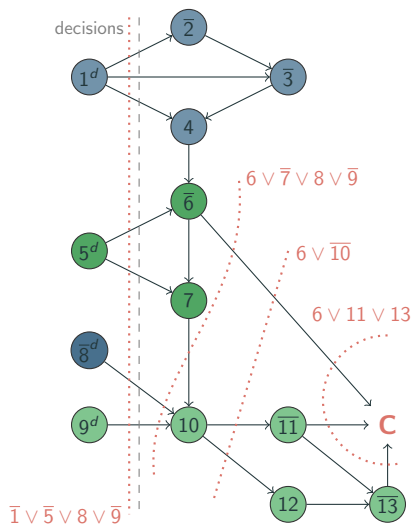
## How to Determine Backjump Clauses?

- ▶ implication graph
- ▶ resolution

9

## Example: Implication Graph

$$\varphi = (\bar{1} \vee \bar{2}) \wedge (\bar{1} \vee 2 \vee \bar{3}) \wedge (\bar{1} \vee 3 \vee 4) \wedge (\bar{4} \vee \bar{5} \vee \bar{6}) \wedge (\bar{5} \vee 6 \vee 7) \wedge (\bar{7} \vee 8 \vee \bar{9} \vee 10) \wedge (\bar{10} \vee \bar{11}) \wedge (\bar{10} \vee 12) \wedge (\bar{12} \vee \bar{13}) \wedge (6 \vee 11 \vee 13)$$



level	literal	reason
1	1	decision
	$\bar{2}$	$\bar{1} \vee \bar{2}$
	3	$\bar{1} \vee 2 \vee \bar{3}$
	4	$\bar{1} \vee 3 \vee 4$
2	5	decision
	$\bar{6}$	$\bar{4} \vee \bar{5} \vee \bar{6}$
	7	$\bar{5} \vee 6 \vee 7$
3	8	decision
	9	decision
	10	$\bar{7} \vee 8 \vee \bar{9} \vee 10$
	11	$\bar{10} \vee \bar{11}$
	12	$\bar{10} \vee 12$
	13	$\bar{12} \vee \bar{13}$

10

next

## What to Learn from That?

### Definitions

- ▶ **cut** of implication graph has at least all decision literals on the left, and at least the conflict node on the right
- ▶ literal  $l$  in implication graph is **unique implication point (UIP)** if all paths from last decision literal to conflict node go through  $l$
- ▶ **first UIP** is UIP closest to conflict node

### Key Observations

- ▶ if  $l_1 \rightarrow l'_1, \dots, l_k \rightarrow l'_k$  are cut edges then  $l'_1 \vee \dots \vee l'_k$  is entailed clause
- ▶ last decision literal is UIP

### Example

- ▶  $\bar{1} \vee \bar{5} \vee 8 \vee \bar{9}$
- ▶  $6 \vee 11 \vee 13$
- ▶  $6 \vee \bar{10}$
- ▶  $6 \vee \bar{7} \vee 8 \vee \bar{9}$
- ▶ UIPs are 9 and 10
- ▶ first UIP is 10

11

## Definition (Implication Graph)

Consider DPLL derivation to  $\| F \implies_B^* M \| F$ .

Implication graph is a directed acyclic graph constructed as follows:

- ▶ add node labelled  $l$  for every decision literal  $l$  in  $M$
- ▶ repeat until there is no change:
  - if  $\exists$  clause  $l_1 \vee \dots \vee l_m \vee l'$  in  $F$  such that there are already nodes  $l_1^c, \dots, l_m^c$ 
    - ▶ add node  $l'$  if not yet present
    - ▶ add edges  $l_i^c \rightarrow l'$  for all  $1 \leq i \leq m$  if not yet present
- ▶ if  $\exists$  clause  $l'_1 \vee \dots \vee l'_k$  in  $F$  such that there are nodes  $l_1^c, \dots, l_k^c$ 
  - ▶ add conflict node labeled  $C$
  - ▶ add edges  $l_i^c \rightarrow C$

potential backjump clause

## Lemma

if edges intersected by cut are  $l_1 \rightarrow l'_1, \dots, l_k \rightarrow l'_k$  then  $F \models l_1^c \vee \dots \vee l_k^c$

12

## How to Derive Backjump Clause by Resolution

- ▶ let  $C_0$  be the conflict clause
- ▶ let  $l$  be last assigned literal such that  $l^c$  is in  $C_0$
- ▶ while  $l$  is no decision literal:
  - ▶  $C_{i+1}$  is resolvent of  $C_i$  and clause  $D$  that led to assignment of  $l$
  - ▶ let  $l$  be last assigned literal such that  $l^c$  is in  $C_{i+1}$

## Observation

every  $C_i$  corresponds to cut in implication graph

## Example

- ▶  $C_0 = 6 \vee 11 \vee 13$
- ▶  $C_1 = 6 \vee 11 \vee \overline{12}$
- ▶  $C_2 = 6 \vee 11 \vee \overline{10}$
- ▶  $C_3 = 6 \vee \overline{10}$
- ▶  $C_4 = 6 \vee \overline{7} \vee 8 \vee \overline{9}$

10

14

## Remarks

- ▶ keeping track of implication graph is too expensive in practice
- ▶ compute clauses associated with cuts by resolution instead

## Definition (Resolution)

$$\frac{C \vee l \quad C' \vee \neg l}{C \vee C'}$$

(assuming literals in clauses can be reordered)

## Example

$$\frac{6 \vee 11 \vee 13 \quad \overline{12} \vee \overline{13}}{6 \vee 11 \vee \overline{12}}$$

13

## Observations

- ▶ adding backjump clause to formula helps to prune search space (learning)
- ▶ if progress is too slow according to some measure, restart procedure: due to learned clauses heuristics will lead to different proof search

## Definition (DPLL with Learning and Restarts)

DPLL with learning and restarts  $\mathcal{R}$  extends system  $\mathcal{B}$  by following three rules:

- ▶ learn  $M \| F \implies M \| F, C$   
if  $F \models C$  and all atoms of  $C$  occur in  $M$  or  $F$
- ▶ forget  $M \| F, C \implies M \| F$   
if  $F \models C$
- ▶ restart  $M \| F \implies \| F$

15

## Theorem (Termination)

any derivation  $\parallel F \Rightarrow_{\mathcal{R}} S_1 \Rightarrow_{\mathcal{R}} S_2 \Rightarrow_{\mathcal{R}} \dots$  is finite if

- ▶ it contains *no infinite subderivation of learn and forget steps*, and
- ▶ *restart is applied with increasing periodicity*

## Theorem (Correctness)

for derivation with final state  $S_n$ :

$$\parallel F \Rightarrow_{\mathcal{R}} S_1 \Rightarrow_{\mathcal{R}} S_2 \Rightarrow_{\mathcal{R}} \dots \Rightarrow_{\mathcal{R}} S_n$$

- ▶ if  $S_n = \text{FailState}$  then  $F$  is *unsatisfiable*
- ▶ if  $S_n = M \parallel F'$  then  $F$  is *satisfiable* and  $M \models F$

16

## Problem

given software system with  $n$  parameters, generate set of test cases which covers all problematic situations while being as small as possible

## Pairwise Testing

- ▶ well-practiced software testing method
- ▶ **observation**: most faults are caused by interaction of at most two parameters

## Example (Testing on Mobile Phones)

some combinations may be infeasible

property	values		storage	cores	camera	SIM	OS
storage	32GB, 64GB, 128GB	1	128GB	4	12MP	single	Android
cores	2, 4, 8	2	32GB	2	8MP	single	Android
camera	8MP, 12MP, 16MP	3	64GB	2	12MP	dual	iOS
SIM	single, dual	4	32GB	4	16MP	dual	iOS
OS	Android, iOS	5	64GB	8	16MP	single	Android
		6	128GB	8	8MP	dual	iOS
		7	128GB	2	12MP	dual	Android
		8	32GB	8	16MP	single	iOS
		9	64GB	4	8MP	single	iOS

(a) testing model for mobile phones

(b) test case set with pairwise coverage

18

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17

## Encode Test Set of Fixed Size in SAT

- ▶ have  $n$  parameters, and parameter  $i$  has  $C_i$  values
- ▶ for all  $m$  test cases use variables  $x_{ij}$  meaning that parameter  $i$  has value  $j$
- ▶ parameter  $j$  has exactly one value

$$\text{one\_value}(x_{j1}, \dots, x_{jC_j}) = \bigvee_{1 \leq k \leq C_j} x_{jk} \wedge \bigwedge_{1 \leq k < k' \leq C_j} \neg x_{jk} \vee \neg x_{jk'}$$

- ▶ in test case every parameter has one value

$$\text{test\_case}(x_{11}, \dots, x_{nC_n}) = \bigwedge_{1 \leq j \leq n} \text{one\_value}(x_{j1}, \dots, x_{jC_j})$$

- ▶ constraints on test case can be expressed by formula  $\text{constraints}(x_{11}, \dots, x_{nC_n})$
- ▶ use overall encoding assuming set of parameter pairs  $P$

$$\bigwedge_{1 \leq i \leq m} \text{test\_case}(\bar{x}^i) \wedge \text{constraints}(\bar{x}^i) \wedge \bigwedge_{(j,k),(j',k') \in P} \bigvee_{1 \leq i \leq m} x_{jk}^i \wedge x_{j'k'}^i$$

- ▶ Minimal test set can be found by repeating approach with smaller  $m$

19