

CS6410 Software Verification

#3. Tseitin Transform, Modeling systems, BDD

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Converting to CNF

- Every formula can be converted to CNF:
 - In **exponential** time and space with the same set of variables.
 - In **linear** time and space if new variables are added.
 - In this case the original and converted formulas are “equi-satisfiable”.
 - This technique is called Tseitin's encoding.

source: decision-procedures.org



Converting to CNF: the exponential way

$\text{CNF}(\phi) \{$

case

ϕ is a literal: return ϕ

ϕ is $\psi_1 \wedge \psi_2$: return $\text{CNF}(\psi_1) \wedge \text{CNF}(\psi_2)$

ϕ is $\psi_1 \vee \psi_2$: return $\text{Dist}(\text{CNF}(\psi_1), \text{CNF}(\psi_2))$

$\}$

$\text{Dist}(\psi_1, \psi_2) \{$

case

ψ_1 is $\phi_{11} \wedge \phi_{12}$: return $\text{Dist}(\phi_{11}, \psi_2) \wedge \text{Dist}(\phi_{12}, \psi_2)$

ψ_2 is $\phi_{21} \wedge \phi_{22}$: return $\text{Dist}(\psi_1, \phi_{21}) \wedge \text{Dist}(\psi_1, \phi_{22})$

else: return $\psi_1 \vee \psi_2$

Converting to CNF: the exponential way

- Consider the formula

$$\phi = (x_1 \wedge y_1) \vee (x_2 \wedge y_2)$$

- $\text{CNF}(\phi) =$

$$(x_1 \vee x_2) \wedge$$

$$(x_1 \vee y_2) \wedge$$

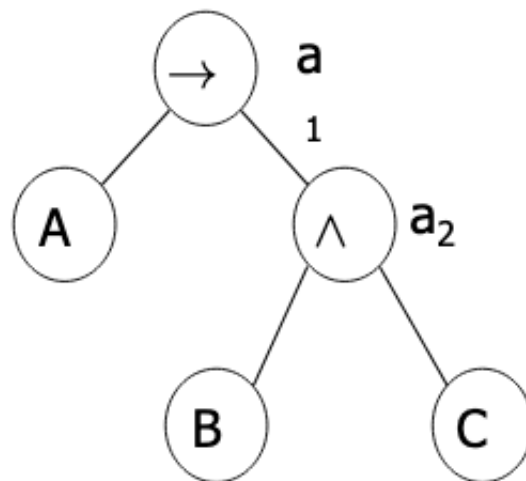
$$(y_1 \vee x_2) \wedge$$

$$(y_1 \vee y_2)$$

- Now consider: $\phi_n = (x_1 \wedge y_1) \vee (x_2 \wedge y_2) \vee \dots \vee (\underline{x_n} \wedge \underline{y_n})$
- Q: How many clauses $\text{CNF}(\phi)$ returns ?
- A: 2^n

Converting to CNF: Tseitin's encoding

- Consider the formula $\phi = (A \rightarrow (B \wedge C))$
- The parse tree:



- Associate a new auxiliary variable with each gate.
- Add constraints that define these new variables.
- Finally, enforce the root node.

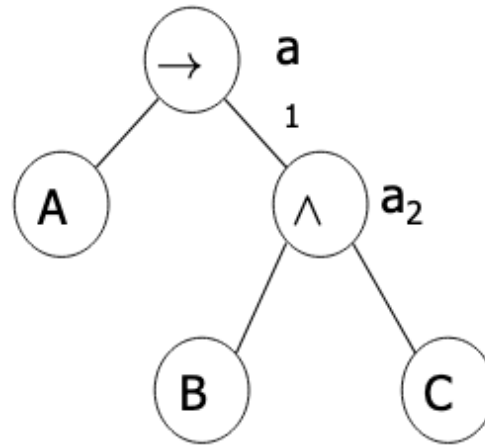
Converting to CNF: Tseitin's encoding

- Need to satisfy:

$$(a_1 \leftrightarrow (A \rightarrow a_2)) \wedge$$

$$(a_2 \leftrightarrow (B \wedge C)) \wedge$$

$$(a_1)$$



- Each such constraint has a CNF representation with 3 or 4 clauses.

source: decision-procedures.org



Converting to CNF: Tseitin's encoding

- Need to satisfy:

$$(a_1 \leftrightarrow (A \rightarrow a_2)) \wedge$$

$$(a_2 \leftrightarrow (B \wedge C)) \wedge$$

$$(a_1)$$

- First: $(a_1 \vee A) \wedge (a_1 \vee \neg a_2) \wedge (\neg a_1 \vee \neg A \vee a_2)$
- Second: $(\neg a_2 \vee B) \wedge (\neg a_2 \vee C) \wedge (a_2 \vee \neg B \vee \neg C)$



Converting to CNF: Tseitin's encoding

- Let's go back to

$$\phi_n = (x_1 \wedge y_1) \vee (x_2 \wedge y_2) \vee \cdots \vee (\underline{x_n} \wedge \underline{y_n})$$

- With Tseitin's encoding we need:

- $2n$ auxiliary variables a_1, \dots, a_{2n} .
- Each adds 3 constraints.
- Top clause: $(a_1 \vee \cdots \vee a_n)$

- Hence, we have

- $6n + 1$ clauses, instead of 2^n .
- $4n$ variables rather than $2n$.

Before how to solve these formulas

- Q: Suppose we can solve the satisfiability problem... how can this help us?
- A: There are numerous problems in the industry that are solved via the satisfiability problem of propositional logic
 - Logistics...
 - Planning...
 - Electronic Design Automation industry...
 - Cryptography...
 - ... (every NP-P problem...)

Modeling with PL

Example: placement of wedding guests

- Three chairs in a row: 1,2,3
- We need to place Aunt, Sister and Father
- Constraints:
 - Aunt doesn't want to sit near Father
 - Aunt doesn't want to sit in the left chair
 - Sister doesn't want to sit to the right of Father
- Q: Can we satisfy these constraints?

Example

- Denote: Aunt = 1, Sister = 2, Father = 3
- Introduce a propositional variable for each pair (person, place).
- x_{ij} = person i is sited at place j , for $1 \leq i, j \leq 3$

- Constraints:

- Aunt doesn't want to sit near father:

$$(x_{11} \rightarrow \neg x_{32}) \wedge (x_{12} \rightarrow \neg x_{31} \wedge \neg x_{33}) \wedge (x_{13} \rightarrow \neg x_{32})$$

- Aunt doesn't want to sit in the left chair

$$\neg x_{11}$$

- Sister doesn't want to sit to the right (immediate) of the Father

$$(x_{31} \rightarrow \neg x_{22}) \wedge (x_{32} \rightarrow \neg x_{23})$$

Example

- More constraints
 - Each person is placed

$$(x_{11} \vee x_{12} \vee x_{13}) \wedge (x_{21} \vee x_{22} \vee x_{23}) \wedge (x_{31} \vee x_{32} \vee x_{33})$$

- No person is placed in more than one place

$$\begin{aligned} & (x_{11} \rightarrow \neg x_{12} \wedge \neg x_{13}) \wedge (x_{12} \rightarrow \neg x_{11} \wedge \neg x_{13}) \wedge (x_{13} \rightarrow \neg x_{11} \wedge \neg x_{12}) \wedge \\ & (x_{21} \rightarrow \neg x_{22} \wedge \neg x_{23}) \wedge (x_{22} \rightarrow \neg x_{21} \wedge \neg x_{23}) \wedge (x_{23} \rightarrow \neg x_{21} \wedge \neg x_{22}) \wedge \\ & (x_{31} \rightarrow \neg x_{32} \wedge \neg x_{33}) \wedge (x_{32} \rightarrow \neg x_{31} \wedge \neg x_{33}) \wedge (x_{33} \rightarrow \neg x_{31} \wedge \neg x_{32}) \end{aligned}$$



Example 3: assignment of frequencies

- n radio stations
- For each assign one of k transmission frequencies, $k < n$.
- E -- set of pairs of stations, that are too close to have the same frequency.
- Q: which graph problem does this remind you of?

Example 3 (cont'd)

- $x_{i,j}$ – station i is assigned frequency j , for

$$1 \leq i \leq n, 1 \leq j \leq k$$

- Every station is assigned at least one frequency:

$$\bigwedge_{i=1}^n \bigvee_{j=1}^k x_{ij}$$

- Every station is assigned not more than one frequency:

$$\bigwedge_{i=1}^n \bigwedge_{j=1}^{k-1} (x_{ij} \rightarrow \bigwedge_{j < t \leq k} \neg x_{it})$$

- Close stations are not assigned the same frequency.

For each $(i,j) \in E$,

$$\bigwedge_{t=1}^k (x_{it} \rightarrow \neg x_{jt})$$



Example 2 (Lewis Carroll)

- (1) All the dated letters in this room are written on blue paper;
- (2) None of them are in black ink, except those that are written in the third person;
- (3) I have not filed any of them that I can read;
- (4) None of them, that are written on one sheet, are undated;
- (5) All of them, that are not crossed, are in black ink;
- (6) All of them, written by Brown, begin with "Dear Sir";
- (7) All of them, written on blue paper, are filed;
- (8) None of them, written on more than one sheet, are crossed;
- (9) None of them, that begins with "Dear Sir", are written in the third person.

Therefore, I cannot read any of Brown's letters.

- Is this statement **valid** ?

Example 2 (cont'd)

- $p = \text{"the letter is dated"}$

- $q = \text{"the letter is written on blue paper"}$

(1) All the dated letters in this room are written on blue paper;

$$p \rightarrow q$$

- $r = \text{"the letter is written in black ink"}$

- $s = \text{"the letter is written in the third person"}$

(2) None of them are in black ink, except those that are written in the third person;

$$\neg s \rightarrow \neg r$$

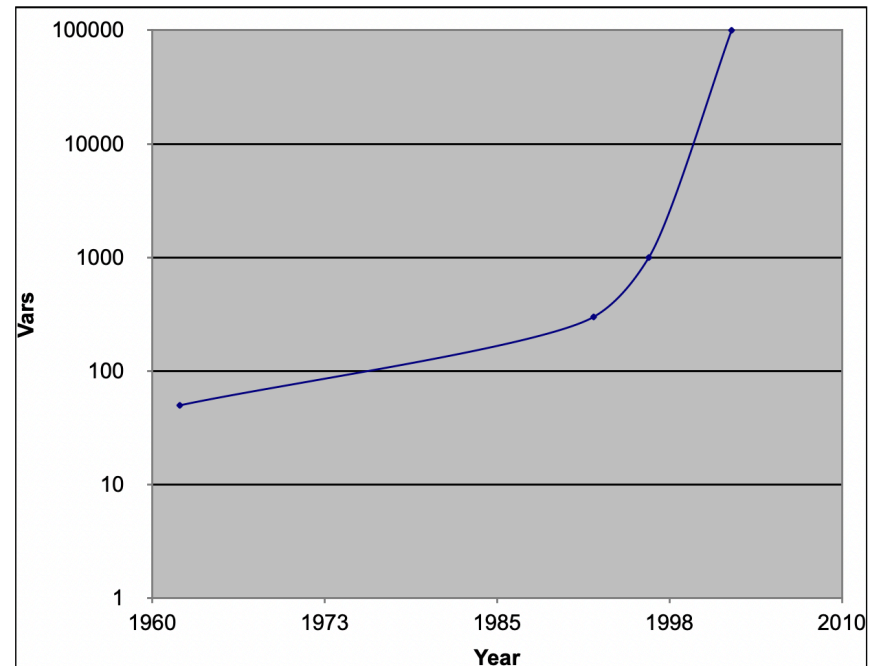
- ...

Overview

- Deciding Propositional Logic
 - SAT tools
 - BDDs

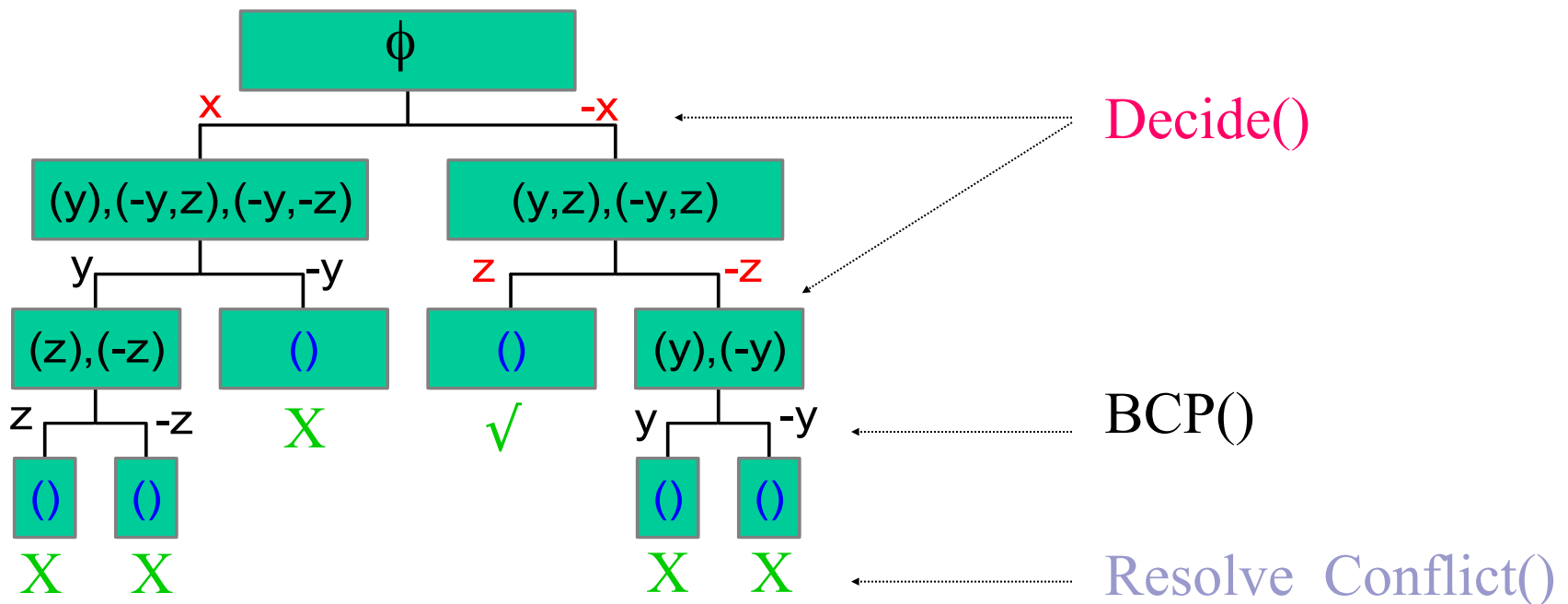
Modern SAT Solvers

- Modern SAT solvers can solve many real-life CNF formulas with hundreds of thousands or even millions of variables in a reasonable amount of time.

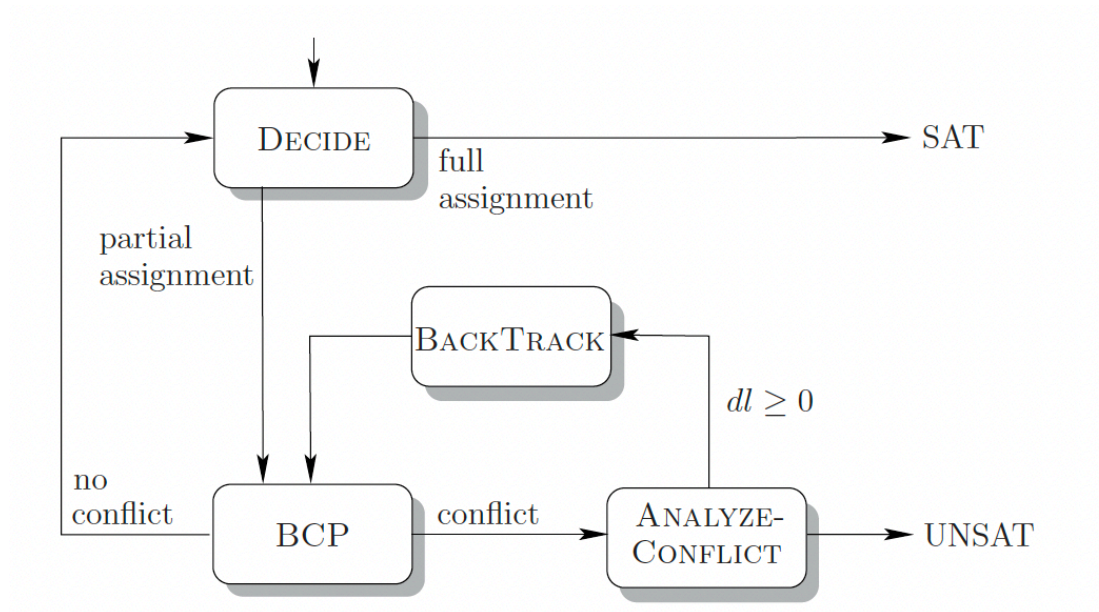


A High-level Basic SAT algorithm

- Given ϕ in CNF: $(x \vee y \vee z) \wedge (-x \vee y) \wedge (-y \vee z) \wedge (-x \vee -y \vee -z)$



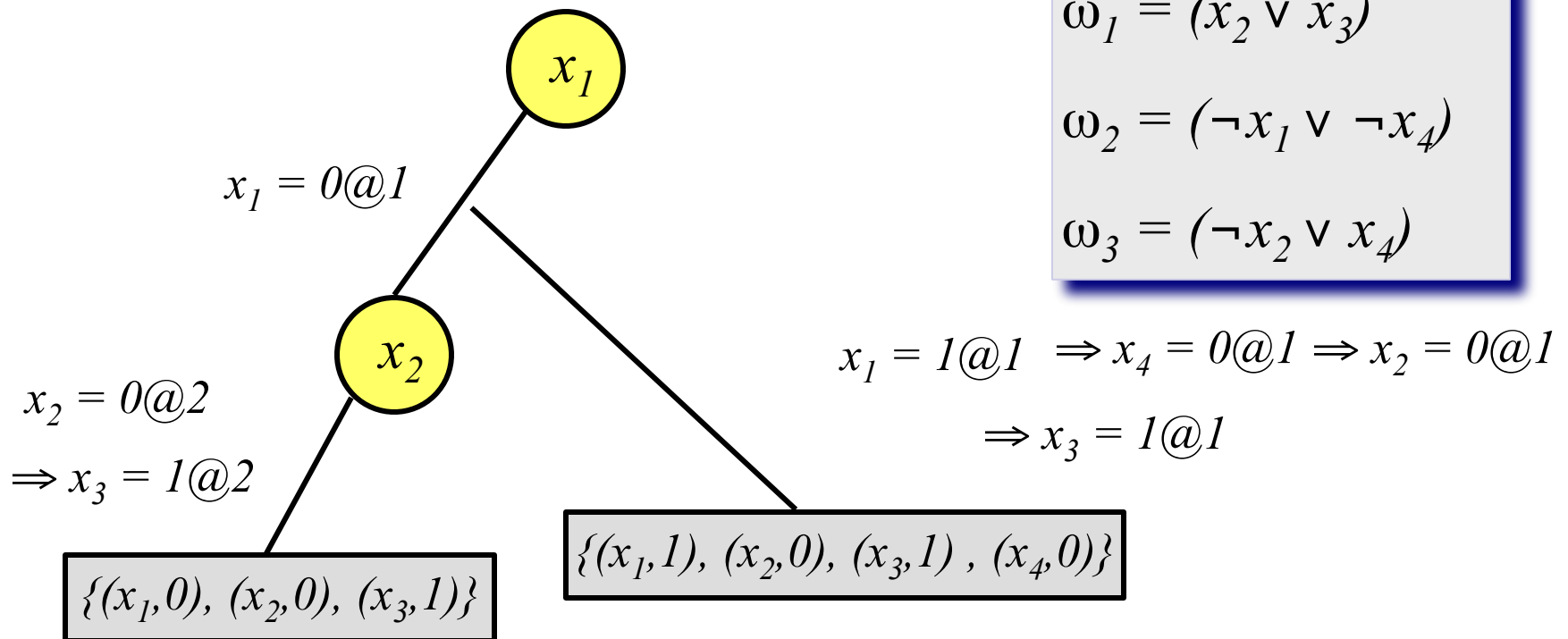
More realistic DPLL



Basic Backtracking Search

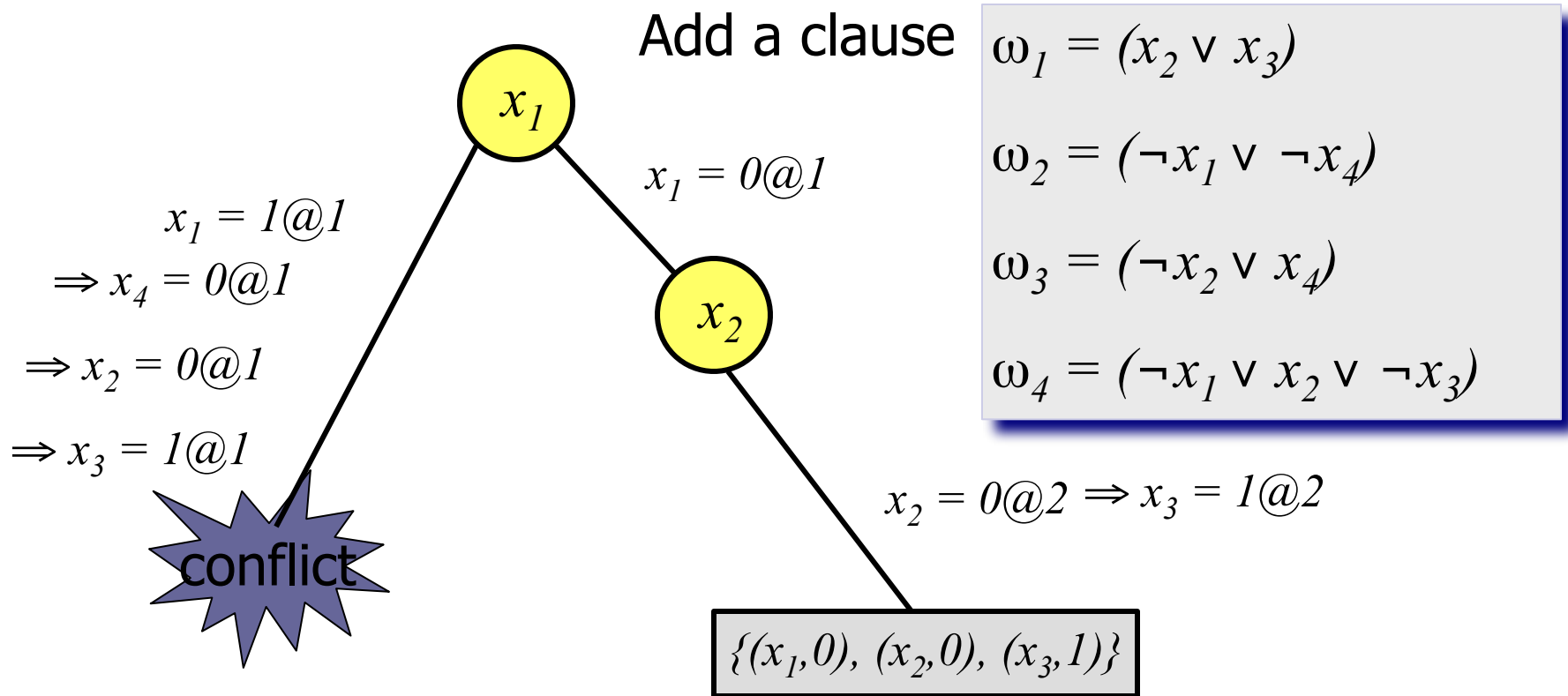
- Organize the search in the form of a decision tree
 - Each node corresponds to a decision
 - Definition: Decision Level (DL) is the depth of the node in the decision tree.
 - Notation: $x = v@d$
 $x \in \{0,1\}$ is assigned to v at decision level d

Backtracking Search in Action



No backtrack in this example,
regardless of the decision!

Backtracking Search in Action



Status of a clause

- A clause can be
 - **Satisfied**: at least one literal is satisfied
 - **Unsatisfied**: all literals are assigned but none are satisfied
 - **Unit**: all but one literals are assigned but none are satisfied
 - **Unresolved**: all other cases
- Example: $C = (x_1 \vee x_2 \vee x_3)$

x_1	x_2	x_3	C
1	0		Satisfied
0	0	0	Unsatisfied
0	0		Unit
	0		Unresolved

Decision heuristics - DLIS

DLIS (Dynamic Largest Individual Sum) –
choose the assignment that increases the most the number of
satisfied clauses

- For a given variable x :
 - C_{xp} – # unresolved clauses in which x appears positively
 - C_{xn} – # unresolved clauses in which x appears negatively
 - Let x be the literal for which C_{xp} is maximal
 - Let y be the literal for which C_{yn} is maximal
 - If $C_{xp} > C_{yn}$ choose x and assign it TRUE
 - Otherwise choose y and assign it FALSE
- Requires l (#literals) queries for each decision.

Decision heuristics - JW

Jeroslow-Wang method

Compute for every clause ω and every variable l (in each phase):

- $J(l) := \sum_{l \in \omega, \omega \in \varphi} 2^{-|\omega|}$
- Choose a variable l that maximizes $J(l)$.
- This gives an exponentially higher weight to literals in shorter clauses.

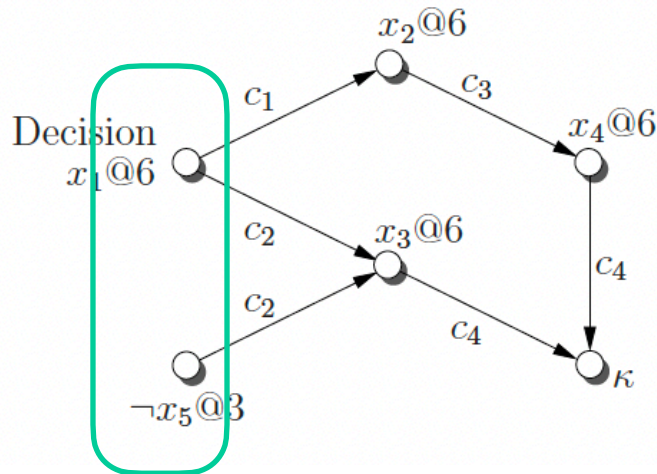
Pause... ||

- We will see other (more advanced) decision Heuristics soon.
- These heuristics are integrated with a mechanism called **Learning with Conflict-Clauses**, which we will learn next.

Implication Graph

- The process of BCP is best illustrated with an implication graph.
- An implication graph represents the current partial assignment and the reason for each of the implications.
- An implication graph is a labeled directed acyclic graph $G(V,E)$, where:
 - V represents the literals of the current partial assignment.
 - $E = \{(v_i, v_j) \mid v_i, v_j \in V, \}$ denotes the set of directed edges where each edge (v_i, v_j) is labeled with **Antecedent** (v_j).
 - G can also contain a single conflict node labeled with κ and incoming edges $\{(v, \kappa)\}$ labeled with c for some conflicting clause c .

Example

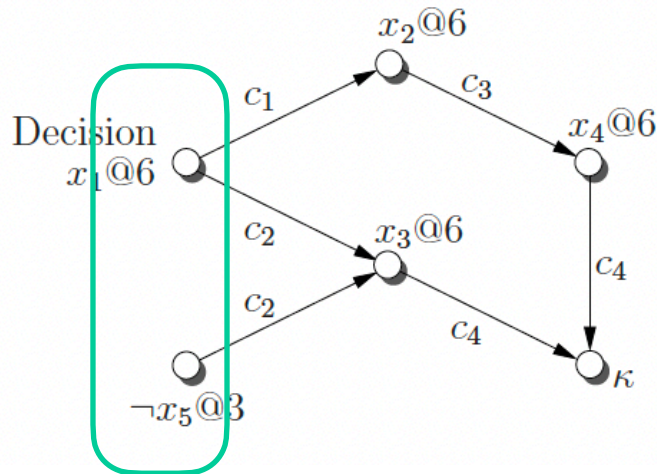


$$\begin{aligned}
 c_1 &= (\neg x_1 \vee x_2) , \\
 c_2 &= (\neg x_1 \vee x_3 \vee x_5) , \\
 c_3 &= (\neg x_2 \vee x_4) , \\
 c_4 &= (\neg x_3 \vee \neg x_4) , \\
 c_5 &= (x_1 \vee x_5 \vee \neg x_2) , \\
 c_6 &= (x_2 \vee x_3) , \\
 c_7 &= (x_2 \vee \neg x_3) , \\
 c_8 &= (x_6 \vee \neg x_5) .
 \end{aligned}$$

sufficient to create the conflict

In fact, This is a partial implication graph,
A subgraph which
illustrates the BCP at a specific decision level

Example



$$\begin{aligned}
 c_1 &= (\neg x_1 \vee x_2) , \\
 c_2 &= (\neg x_1 \vee x_3 \vee x_5) , \\
 c_3 &= (\neg x_2 \vee x_4) , \\
 c_4 &= (\neg x_3 \vee \neg x_4) , \\
 c_5 &= (x_1 \vee x_5 \vee \neg x_2) , \\
 c_6 &= (x_2 \vee x_3) , \\
 c_7 &= (x_2 \vee \neg x_3) , \\
 c_8 &= (x_6 \vee \neg x_5) .
 \end{aligned}$$

sufficient to create the conflict

We learn the *conflict clause* $c_9 : (\neg x_1 \vee x_5)$ Prunes the space

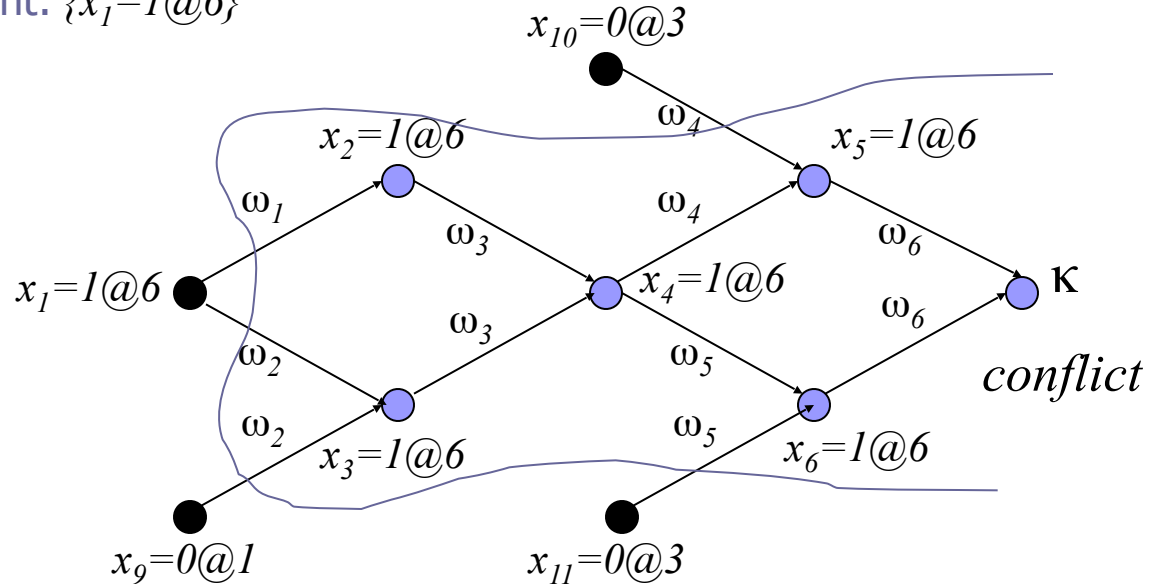
Reflects the fact that this is the solver's way to learn from its past mistakes.

Implication graphs and learning: option #1

Current truth assignment: $\{x_9=0@1, x_{10}=0@3, x_{11}=0@3, x_{12}=1@2, x_{13}=1@2\}$

Current decision assignment: $\{x_1=1@6\}$

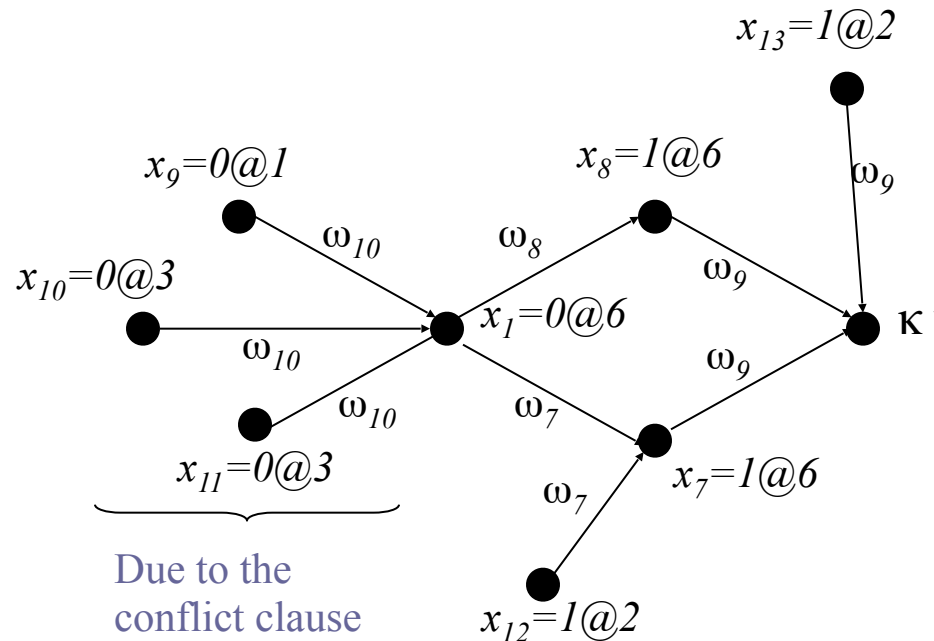
$\omega_1 = (\neg x_1 \vee x_2)$
 $\omega_2 = (\neg x_1 \vee x_3 \vee x_9)$
 $\omega_3 = (\neg x_2 \vee \neg x_3 \vee x_4)$
 $\omega_4 = (\neg x_4 \vee x_5 \vee x_{10})$
 $\omega_5 = (\neg x_4 \vee x_6 \vee x_{11})$
 $\omega_6 = (\neg x_5 \vee \neg x_6)$
 $\omega_7 = (x_1 \vee x_7 \vee \neg x_{12})$
 $\omega_8 = (x_1 \vee x_8)$
 $\omega_9 = (\neg x_7 \vee \neg x_8 \vee \neg x_{13})$



We learn the *conflict clause* $\omega_{10} : (:x_1 \vee x_9 \vee x_{11} \vee x_{10})$

Implication graph, flipped assignment option #1

$$\begin{aligned}\omega_1 &= (\neg x_1 \vee x_2) \\ \omega_2 &= (\neg x_1 \vee x_3 \vee x_9) \\ \omega_3 &= (\neg x_2 \vee \neg x_3 \vee x_4) \\ \omega_4 &= (\neg x_4 \vee x_5 \vee x_{10}) \\ \omega_5 &= (\neg x_4 \vee x_6 \vee x_{11}) \\ \omega_6 &= (\neg x_5 \vee x_6) \\ \omega_7 &= (x_1 \vee x_7 \vee \neg x_{12}) \\ \omega_8 &= (x_1 \vee x_8) \\ \omega_9 &= (\neg x_7 \vee \neg x_8 \vee \neg x_{13}) \\ \omega_{10} &= (:x_1 \vee x_9 \vee x_{11} \vee x_{10})\end{aligned}$$

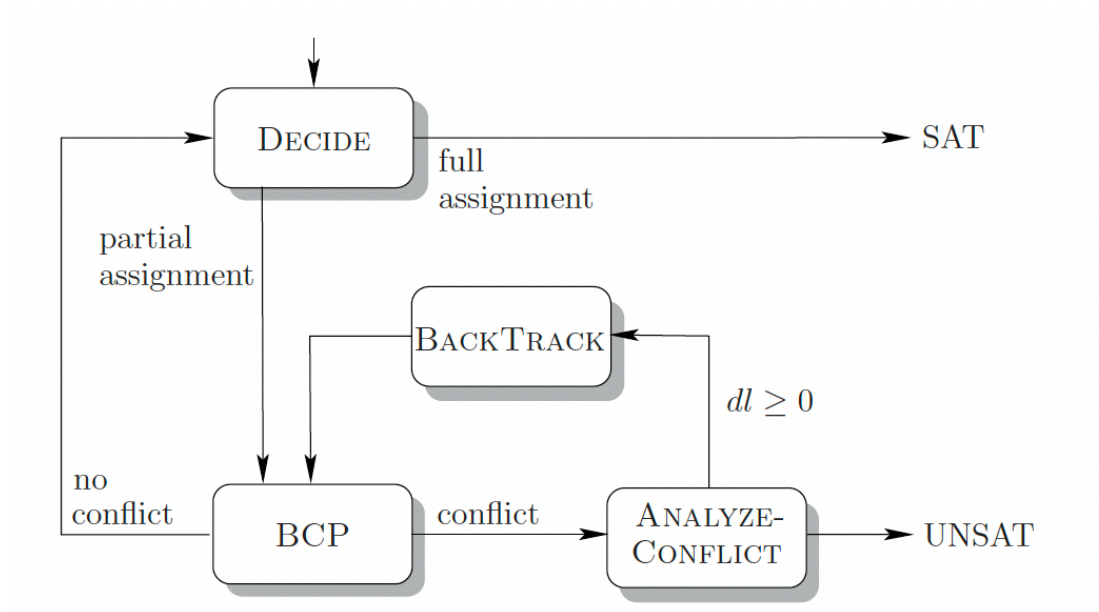


No decision here

Another conflict clause: $\omega_{11}: (:x_{13} \vee :x_{12} \vee x_{11} \vee x_{10} \vee x_9)$

where should we backtrack to now ?

More realistic DPLL



Non-chronological backtracking

Which assignments caused the conflicts ?

$$x_9 = 0@1$$

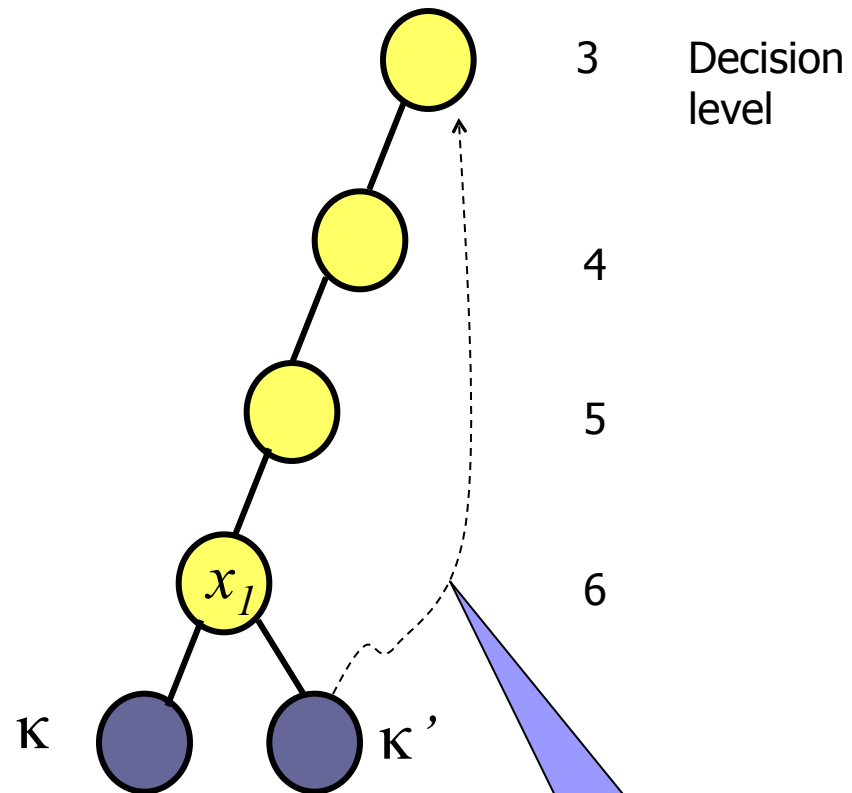
$$x_{10} = 0@3$$

$$x_{11} = 0@3$$

$$x_{12} = 1@2$$

$$x_{13} = 1@2$$

*These assignments
Are sufficient for
Causing a conflict.*



Backtrack to $DL = 3$

Back to the logistics

- Assignment 1
 - Practice: Work out exercise at the end of Chapter 1 in the CoC book. Due Tuesday.
 - Reading: M. Davis, G. Logemann, and D. Loveland. A machine program for theorem-proving. Communications of the ACM, 5(7):394–397, July 1962
- Next Class:
 - Backtracking, Decide heuristics
- Class rescheduling.