

**CSE 417 Autumn 2025**

# **Lecture 26: How SAT solvers work**

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# **Review of last lecture**

# SAT (Satisfiability)

**literals**: variables or their negation

$$a, \neg b, x, \neg x, y$$

**clause**: OR of literals

$$(a \vee \neg b), (x \vee \neg y \vee z)$$

**conjunction normal form (CNF)**: AND of clauses

$$(a \vee \neg b) \wedge (x \vee \neg y \vee z)$$

# SAT (Satisfiability)

**Input:** A CNF formula  $f(x_1, \dots, x_n)$  (equivalently a set of clauses)

**Goal:** Does there exist  $x_1, \dots, x_n$  such that  $f(x_1, \dots, x_n)$  is true?

**SAT is NP-hard:** We don't believe we can solve it quickly in general.

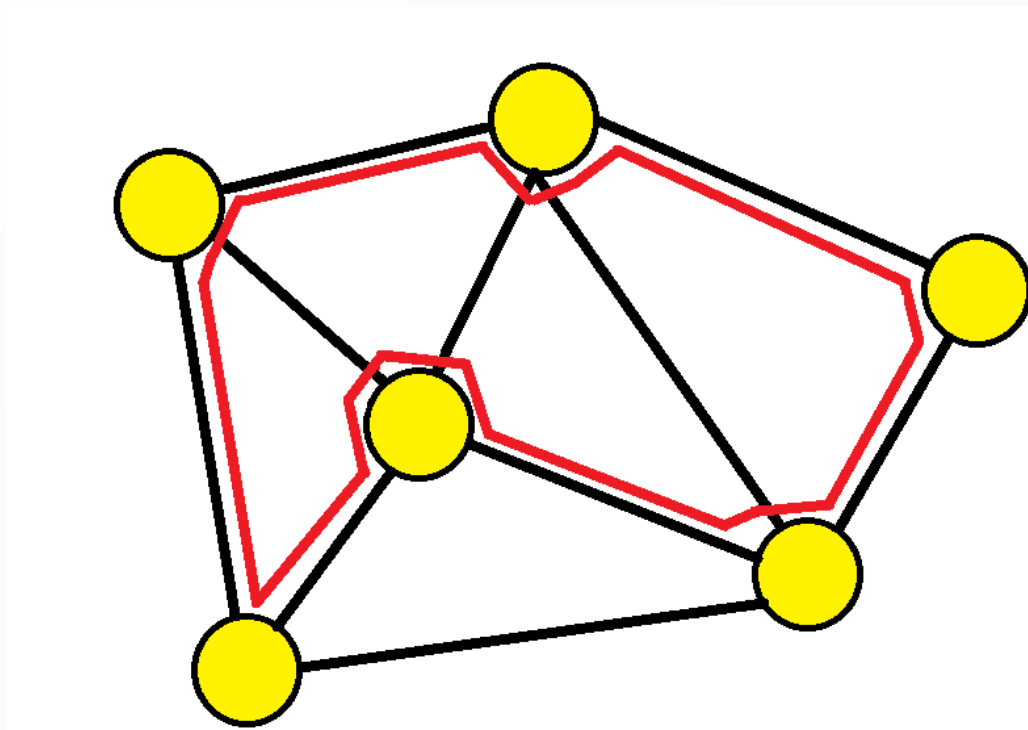
However, in the last ~20 years, we've gotten very good!

Real world problems with  $> 1,000,000$  variables/clauses are OK!

# Hamiltonian path

**Input:** An undirected graph with vertices  $V$  and edges  $E$

**Goal:** Is there a path that uses every vertex exactly once?



Also an NP-hard problem!

# Hamiltonian path

Define  $p_{v,i}$  to mean “vertex  $v$  is the  $i$ th vertex on the path”.

- Each vertex appears exactly once on the path.
- Each position on the path has exactly one vertex.
- If two vertices are adjacent on the path, then there is an edge between them.

# “Exactly one” constraints

We just saw this with “every vertex gets exactly one color”!

To translate “exactly one of  $x_1, x_2, \dots, x_n$ ”:

- At least one of  $x_1, x_2, \dots, x_n$ :

$$(x_1 \vee x_2 \vee \dots \vee x_n)$$

- No two of  $x_1, x_2, \dots, x_n$ :

$$(\neg x_i \vee \neg x_j) \text{ for every pair of } i \text{ and } j$$

Requires  $O(n^2)$  constraints.

# Hamiltonian path

Define  $p_{v,i}$  to mean “vertex  $v$  is the  $i$ th vertex on the path”.

- Each vertex appears exactly once on the path.

“exactly one of  $p_{v,1}, \dots, p_{v,n}$ ” for every vertex  $v$

- Each position on the path has exactly one vertex.

“exactly one of  $p_{1,i}, \dots, p_{n,i}$ ” for every position  $i$



# Hamiltonian path

Define  $p_{v,i}$  to mean “vertex  $v$  is the  $i$ th vertex on the path”.

- If two vertices are adjacent on the path, then there is an edge between them.

$$(p_{u,i} \wedge p_{v,i+1}) \Rightarrow “(u, v) \text{ is an edge}”$$

Use contrapositive: whenever  $(u, v)$  is not an edge, include clause

$$(\neg p_{u,i} \vee \neg p_{v,i+1})$$

# Tseitin transformations

To translate longer Boolean sentences efficiently, introduce helper variables! For example, if you have  $x_1 \oplus x_2 \oplus \cdots \oplus x_n$ , let

- $z_2 \Leftrightarrow x_1 \oplus x_2$
- $z_3 \Leftrightarrow z_2 \oplus x_3$
- ...
- $z_n \Leftrightarrow z_{n-1} \oplus x_n$

Each  $a \Leftrightarrow b \oplus c$  takes 4 clauses to convert to CNF (from concept check).

Use  $n - 1$  new variables and represent this sentence in  $O(n)$  clauses!

# Tseitin transformations

For *any* boolean operation  $R$  (could be XOR, AND, OR, etc.),

$$a \Leftrightarrow b R c$$

takes at most 8 clauses to convert to CNF (since there are only 8 possible clauses at all with 3 variables).

Doing this is called a **Tseitin transformation**.

# Program verification

**Input:** A computer program written in some language and a formal specification

**Goal:** Does the program meet the spec for all inputs?

# Program verification

```
division(int x, int y) {  
    int r = x;  
    int q = 0;  
  
    while (r >= y) {  
        r = r - y;  
        q++;  
    }  
}
```



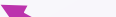
code

```
assert x == y * q + r;  
assert r >= 0 && r < Math.abs(y);
```

spec

# Program verification

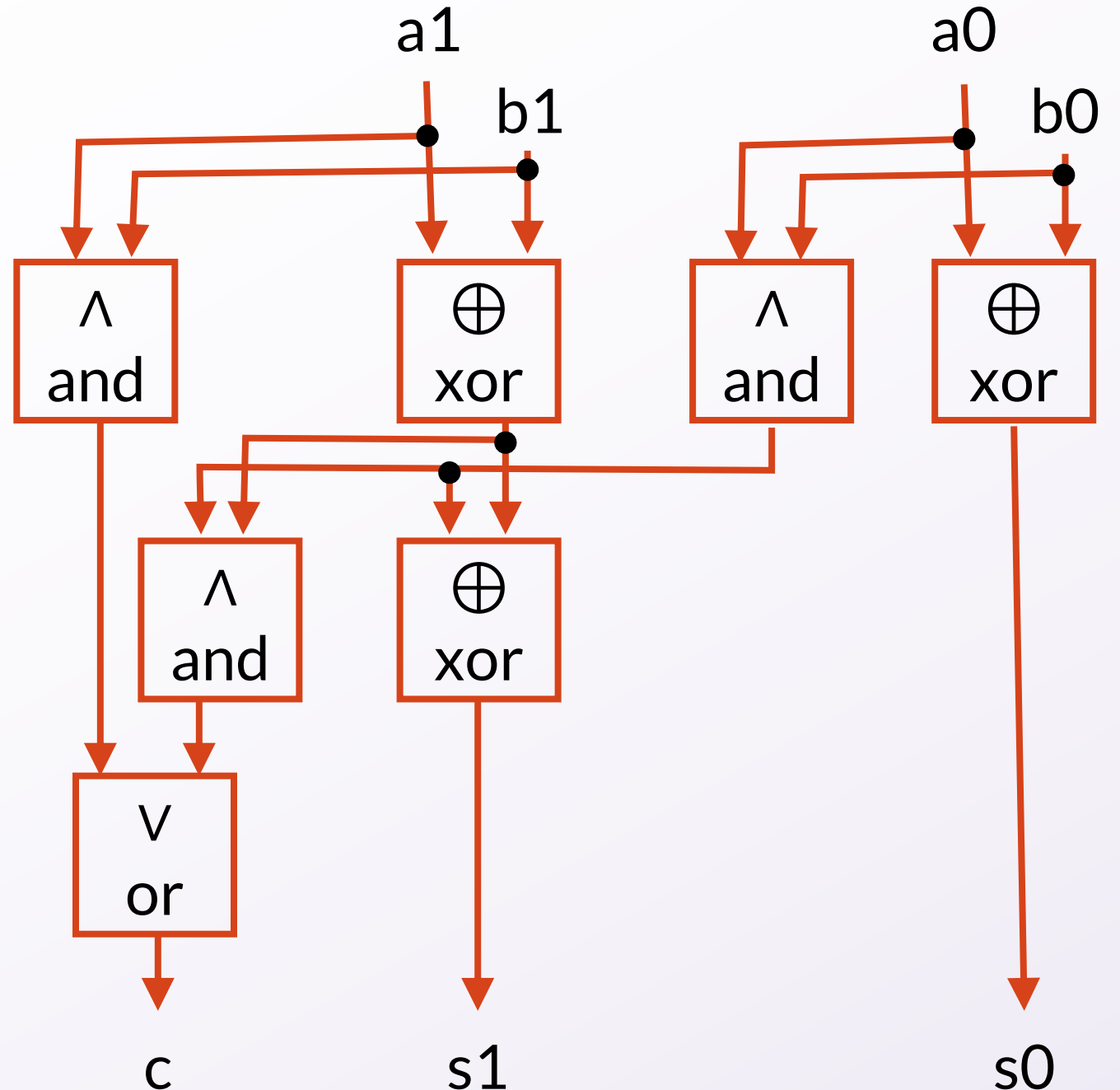
The basic idea is to convert our program into **single static assignment form**, where it will only have:

- Basic functions like  $+$ ,  $>$ , and  $\_? \_ : \_$   Kind of like OR, AND, XOR, etc?
- Assignment to variables  Kind of like Tseitin transformations, except not just bools!
- “Assume” statements
- “Assert” statements  “The spec is satisfied for all small inputs that don't result in long loops.”

# Circuits

In a class on digital design (CSE 369/EE 271), you would learn how to implement all these basic functions with circuits!

$$\begin{array}{r} \phantom{+} \phantom{a1} \phantom{a0} \\ + \phantom{a1} \phantom{a0} \\ \hline c \phantom{a1} \phantom{a0} \end{array}$$



# Some things you can do with circuits

|   |   |
|---|---|
| Addition of $n$ -bit binary numbers                     | $O(n)$ gates  |
| Multiplication of $n$ -bit binary numbers               | $O(n^2)$ gates with simple implementation, improvable |
| Comparison of $n$ -bit numbers                          | $O(n)$ gates  |
| If-then-else for $n$ -bit numbers                       | $O(n)$ gates  |
| Anything that you can compute on a computer in $T$ time | at most $O(T \log T)$ gates                           |

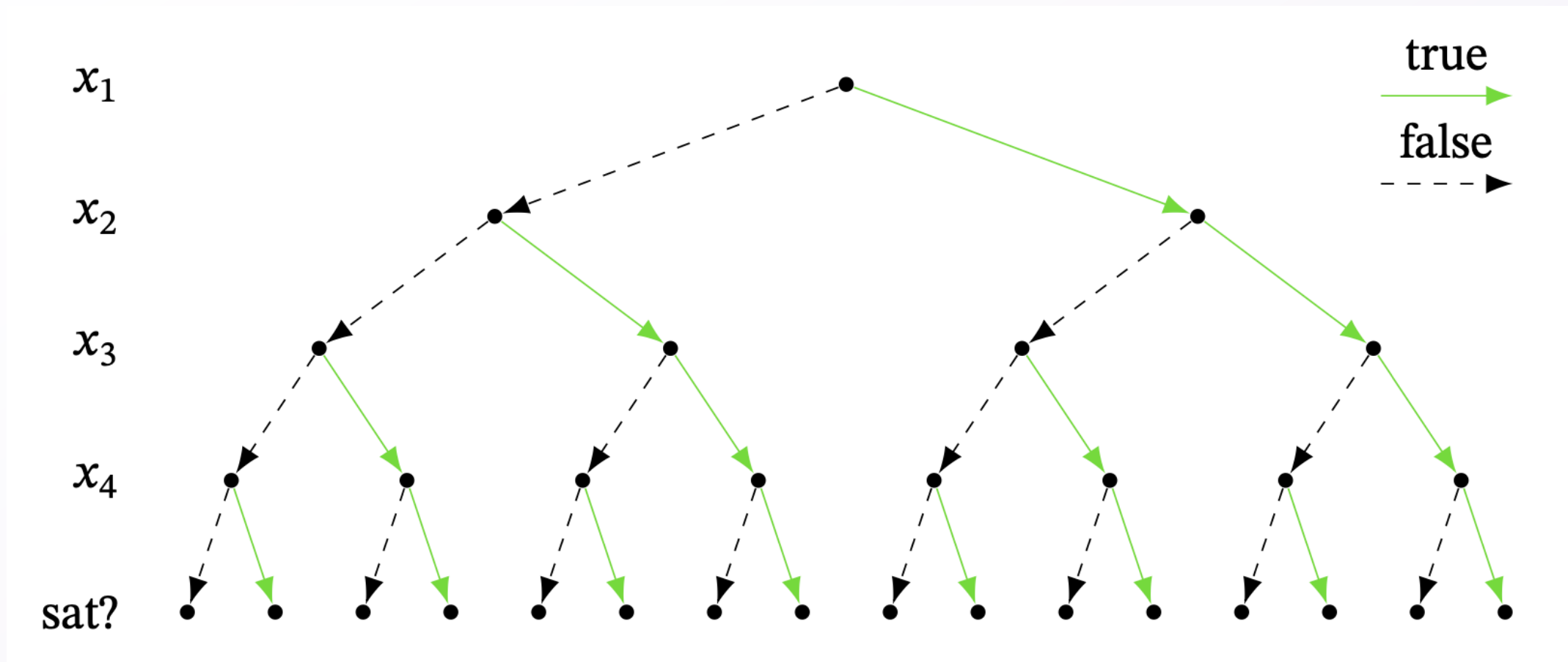
By Tseitin transformations: # new clauses  $\approx$  # new variables = # gates!



# The DPLL algorithm

# First idea: brute force

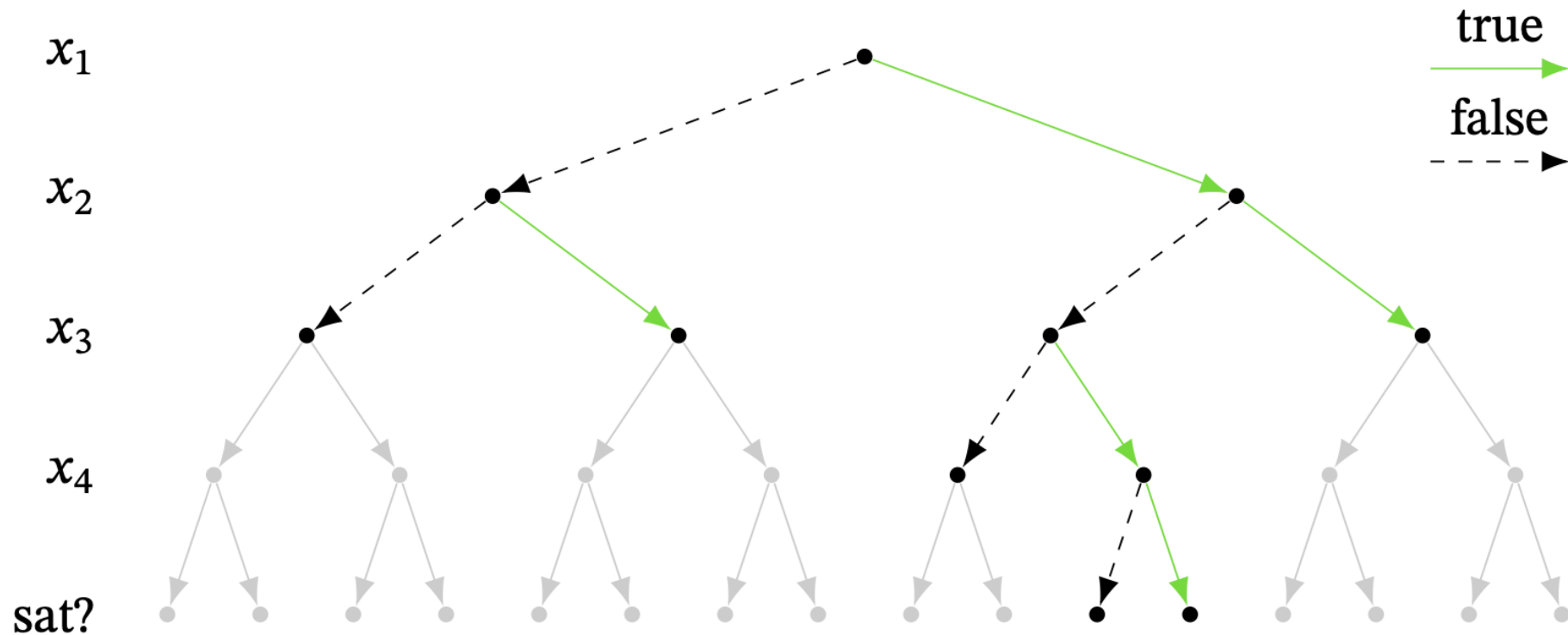
To solve SAT via brute force, we try every assignment.



# Second idea: check unsatisfiability earlier

At every node, check if we've already made the things unsatisfiable.

$(x_1 \vee x_2)$      $(\neg x_2)$      $(x_2 \vee x_3)$      $(x_2 \vee \neg x_3 \vee x_4)$      $(\neg x_3 \vee \neg x_4)$ .



# Even faster: Your ideas

$$(x_1 \vee x_2) \quad (\neg x_2) \quad (x_2 \vee x_3) \quad (x_2 \vee \neg x_3 \vee x_4) \quad (\neg x_3 \vee \neg x_4).$$

“As a human, I would find the clauses with only one variable, which could easily tell whether the variable should be true or false.”

“I would say at least one of  $x_1$  and  $x_2$  needs to be true. Then the second says that  $x_2$  has to be false, so then plugging that back in,  $x_1$  needs to be true.”

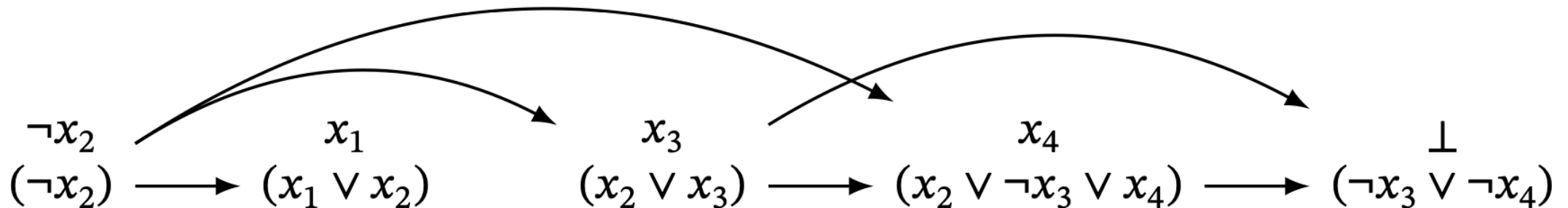
“If  $x_2$  is false then  $(x_1 \vee x_2)$  forces  $x_1$  true and  $(x_2 \vee x_3)$  forces  $x_3$  true.”

# Unit propagation

A **unit** is a clause with one literal.

If you know a unit, simplify other clauses with this knowledge!

$$(x_1 \vee x_2) \quad (\neg x_2) \quad (x_2 \vee x_3) \quad (x_2 \vee \neg x_3 \vee x_4) \quad (\neg x_3 \vee \neg x_4).$$



# Unit propagation

**Input:** set of units  $U$  and a clause  $C$

1. **if** every literal in  $C$  is made false by  $U$ ,
2.     **return** unsatisfiable
3. **else if** every literal in  $C$  except one is made false by  $U$ ,
4.     **return** the unfalsified literal in  $C$

# Unit propagation

**Input:** set of units  $U$  and set of clauses  $\Delta$

1. Repeat the following until an entire iteration passes without propagating a new unit:
2.     **for each** clause  $C \in \Delta$ ,
3.         Unit propagate with  $U$  and  $C$ , possibly updating  $U$ .
4.     **if** unit propagation returned “unsatisfiable”,
5.         **return** “unsatisfiable”
6. **return** the updated set of units  $U$

# DPLL algorithm (Davis–Putnam–Logemann–Loveland, 1961)

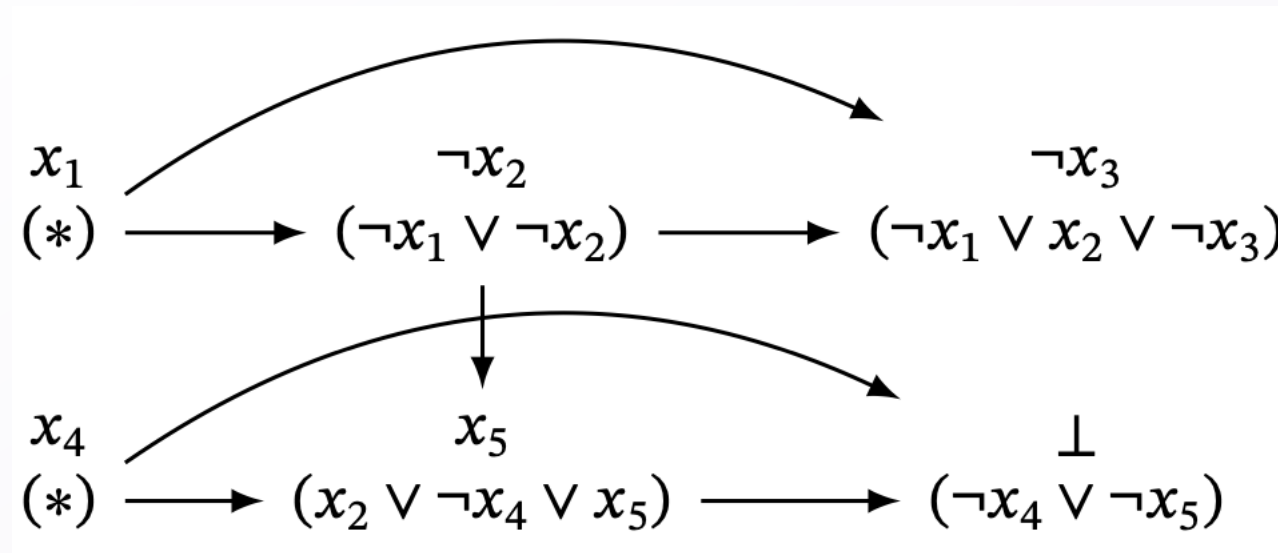
The following defines a recursive function **DPLL**( $U, \Delta$ )

1. Run unit propagation and update  $U$ .
2. **if** unit propagation learns contradiction, **return** false
3. **else**,
4.     **if**  $U$  does not set every variable,
5.         Pick an unset variable  $x$ .
6.         **return** **DPLL**( $U \cup \{x\}, \Delta$ ) OR **DPLL**( $U \cup \{\neg x\}, \Delta$ )
7.     **else, return** true



# DPLL algorithm

$$(\neg x_1 \vee \neg x_2) \quad (\neg x_1 \vee x_2 \vee \neg x_3) \quad (x_2 \vee \neg x_4 \vee x_5) \quad (\neg x_4 \vee \neg x_5)$$

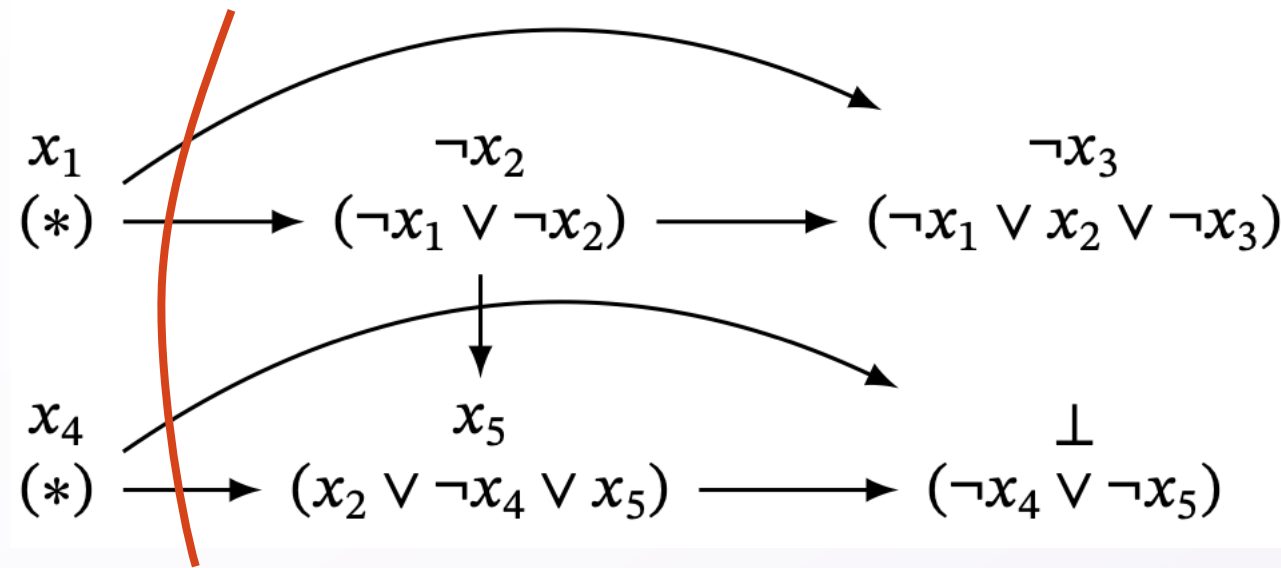


Next, try  $x_4$  being false.

In this case, these clauses are now all satisfied!

# The CDCL algorithm

# Cuts in an implication graph



This cut says: if we know  $x_1$  and  $x_4$ , then we get a contradiction.

Therefore,  $(\neg x_1 \vee \neg x_4)$  must be true.

**Learning** this clause lets unit propagation deduce  $\neg x_4$  from  $x_1$ !

# Clause learning view of DPLL

DPLL recursively calls:

**return**  $\text{DPLL}(U \cup \{x\}, \Delta)$  OR  $\text{DPLL}(U \cup \{\neg x\}, \Delta)$

Instead, it is equivalent for the solver to:

- Decide to check only  $\text{DPLL}(U \cup \{x\}, \Delta)$ .
- Upon reaching contradiction,
  - Let  $C$  be the clause with  $\neg x$  for every decision  $x$ .
  - Add  $C$  to  $\Delta$  and revert  $U$  to before the last decision.

call this the DPLL clause



# Clause learning view of DPLL

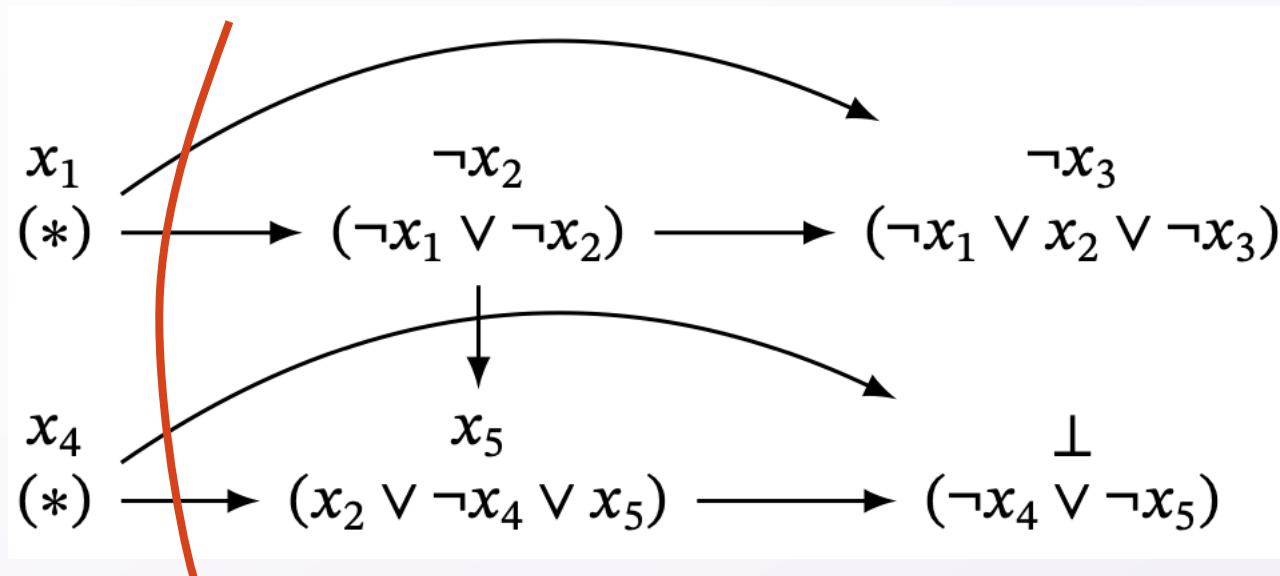
1. Do the following in a loop:
2.     Run unit propagation and update  $U$ .
3.     **if** unit propagation learns a contradiction,
4.         **if**  $U$  still contains a decision, **learn the DPLL clause** and  
            revert  $U$  to before the last decision.
5.         **else, return** unsatisfiable
6.     **else,**
7.         **if** there is an unset variable, pick one and add it to  $U$ .
8.         **else, return** satisfiable.

# Cuts in an implication graph

Take any cut separating the decisions from the contradiction.

The **conflict clause** associated with this cut has the negation of everything immediately before the cut.

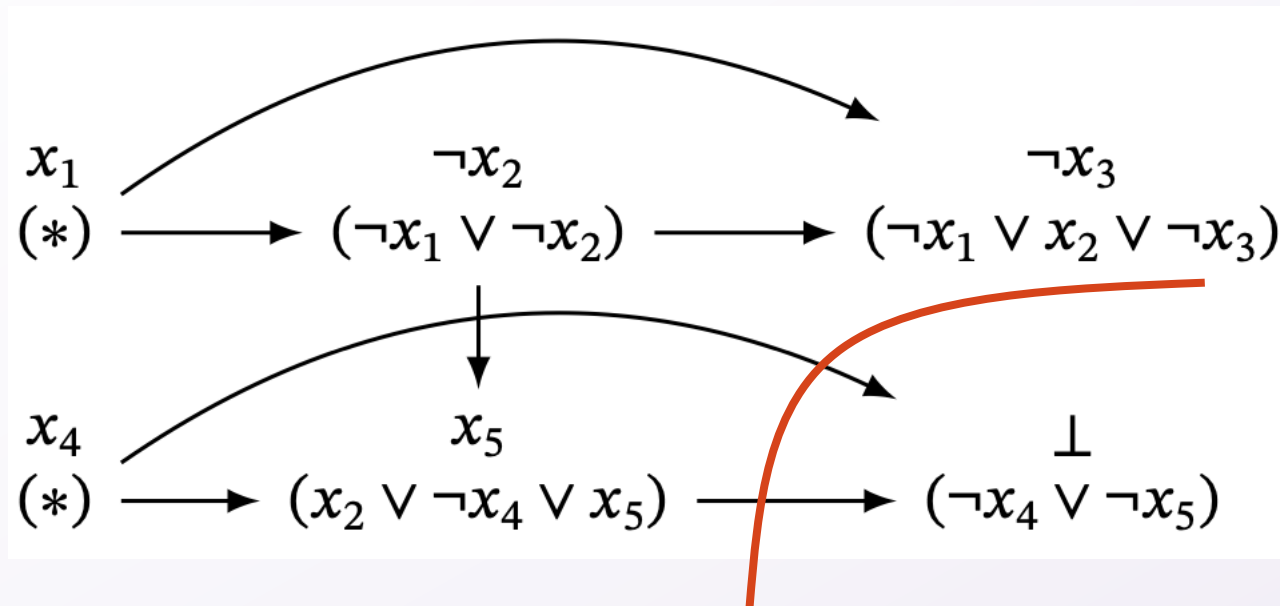
$$(\neg x_1 \vee \neg x_4)$$



# Cuts in an implication graph

Take any cut separating the decisions from the contradiction.

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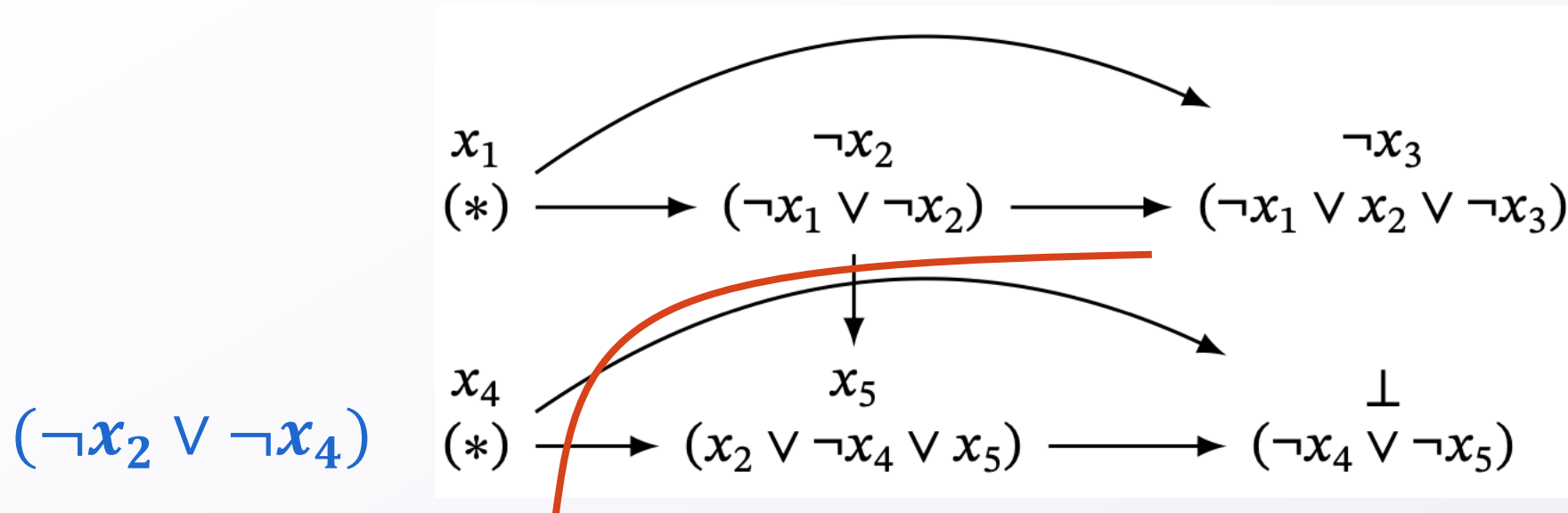


$$(\neg x_4 \vee \neg x_5)$$

# Cuts in an implication graph

Take any cut separating the decisions from the contradiction.

The **conflict clause** associated with this cut has the negation of everything immediately before the cut.

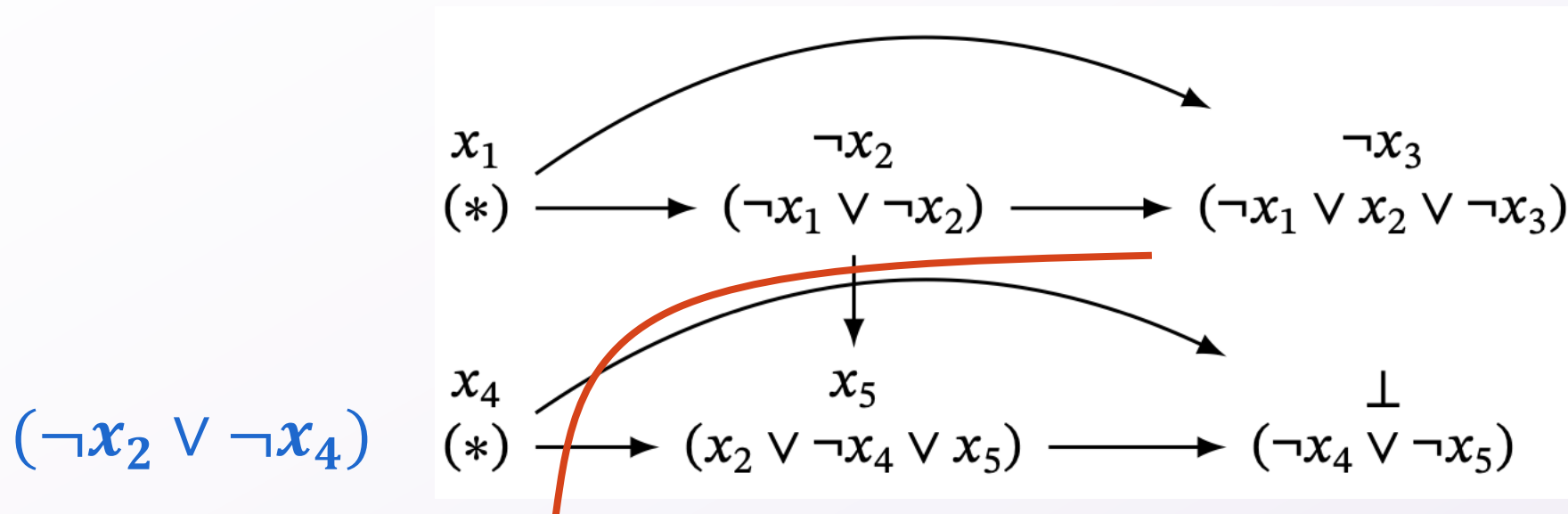




# Cuts in an implication graph

A conflict clause is **asserting** if it can be used for unit propagation with fewer decisions.

Any asserting clause can be learned!



# CDCL (conflict-driven clause learning)

1. Do the following in a loop:
2.     Run unit propagation and update  $U$ .
3.     **if** unit propagation learns a contradiction,
4.         **if**  $U$  still contains a decision, **learn any asserting clause**  
and revert  $U$  to before the last decision.
5.         **else, return** unsatisfiable
6.     **else,**
7.         **if** there is an unset variable, pick one and add it to  $U$ .
8.         **else, return** satisfiable.

# More improvements

- Improved heuristics for choosing the next variable to branch on
- Improved heuristics for choosing which conflict clause to learn
- Faster unit propagation with watched literals
- Random restarts
- Clause deletion

Improving SAT solvers remains an active area of research.

# Final reminders

HW6 (Greedy) resubmissions close tonight @ 11:59pm!

HW7 (Flows) due tonight @ 11:59pm!

I have OH now-12:30pm:

- Meet at front of classroom, we'll walk over together
- CSE (Allen) 214 if you're coming later

Nathan has online OH 12-1pm:

- <https://washington.zoom.us/my/nathanbrunelle>