

# Lecture 3: Algorithms for SAT

#### Reminders

- Homework 0 was due on Monday
- Homework 1 due Monday, Feb 10, 11:59PM
- OH schedule:
  - Thomas: Sunday 3-4pm
  - Cindy: Tuesday 8-9pm
  - o Ishaan: Wednesday 9:30-10:30pm
  - All OH held on OHQ

#### Grading

- Homework: 44%
- Final Project: 38%
- Quizzes: 10%
- Attendance: 8%





## Academic Integrity

- Work on assignments individually (except final project)
  Discussion encouraged, but work should be yours
- OK: high-level discussions
  - "Can you help me understand the DPLL algorithm?"
- OK: low-level discussions
  - "How do I time my program in OR-Tools?"
- Be careful: mid-level discussions
  - Not OK: "How exactly do I write this constraint?"



#### **Health Logistics**

- If you have a reasonable suspicion that you have Covid or sickness, don't come
  - Email me before class and we'll work something out



#### Recap

#### Last week

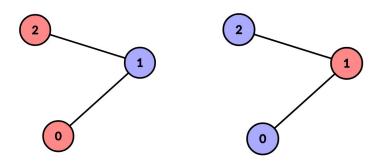
- Using SAT solvers in Python (PycoSAT)
- Encode other problems (graph coloring) as SAT

#### This week

• Build up an algorithm to solve SAT

#### **Symmetry Breaking**

- Solving UNSAT graph coloring problems takes a very long time... why?
- Must rule out every symmetric coloring
- Ex: equivalent colorings



#### **Symmetry Breaking**

- Key idea: add constraints that rule out equivalent symmetric colorings
- Basic way to do this: pick some vertices (ideally a dense subgraph) and fix their colors

# **DEMO Part 2**

We have *n* men and *n* women. Each man and woman submits a preference list ranking everyone of the opposite sex (descending).

Goal: find a **matching** of men to women.

A man and woman who both prefer each other to their matched partners are a **blocking pair**.

A matching is **stable** if it has no blocking pairs.

 $m_{ip}$ : if man *i* is matched to  $p^{th}$  woman or later on his list  $w_{ip}$ : if woman *i* is matched to  $p^{th}$  man or later on her list

 $\begin{bmatrix} W_1 > W_2 \end{bmatrix} M_1 \longrightarrow W_1 \quad \begin{bmatrix} M_1 > M_2 \end{bmatrix}$  $\begin{bmatrix} W_1 > W_2 \end{bmatrix} M_2 \longrightarrow W_2 \quad \begin{bmatrix} M_1 > M_2 \end{bmatrix}$ 

| m <sub>1, 1</sub> | т <sub>1, 2</sub> | m <sub>2, 1</sub> | m <sub>2, 2</sub> | W <sub>1, 1</sub> | W <sub>1, 2</sub> | W <sub>2, 1</sub> | W <sub>2, 2</sub> |
|-------------------|-------------------|-------------------|-------------------|-------------------|-------------------|-------------------|-------------------|
| т                 | F                 | т                 | т                 | т                 | F                 | т                 | т                 |

 $m_{ip}$ : if man *i* is matched to  $p^{th}$  woman or later on his list  $w_{ip}$ : if woman *i* is matched to  $p^{th}$  man or later on her list

• C1: every man is matched

#### $\{m_{i1} \mid 1 \le i \le n\}$

(plus symmetric constraints for women for this and the following constraints)

- $m_{ip}$ : if man *i* is matched to  $p^{th}$  woman or later on his list  $w_{ip}$ : if woman *i* is matched to  $p^{th}$  man or later on her list
- C2: if a man gets his p<sup>th</sup> or later choice, it's also his (p − 1)<sup>th</sup> or later choice

$$\left\{m_{ip} \Rightarrow m_{i(p-1)} \mid 1 \le i \le n, 2 \le p \le n\right\}$$

 $m_{ip}$ : if man *i* is matched to  $p^{th}$  woman or later on his list  $w_{ip}$ : if woman *i* is matched to  $p^{th}$  man or later on her list

 C3: if man *i* is matched to woman *j*, then she is matched to him also

$$\left\{m_{ip} \land \overline{m_{i(p+1)}} \Rightarrow w_{jq} \land \overline{w_{j(q+1)}} \mid 1 \le i, j \le n\right\}$$

- p = position of woman j in man i's list
- q = position of man i in woman j's list

- $m_{ip}$ : if man *i* is matched to  $p^{th}$  woman or later on his list  $w_{ip}$ : if woman *i* is matched to  $p^{th}$  man or later on her list
- **C4:** if man *i* is matched to someone worse than woman *j*, her match must be better than him

$$\left\{m_{i(p+1)} \Rightarrow \overline{w_{jq}} \mid 1 \le i, j \le n\right\}$$

- p = position of woman j in man i's list
- q = position of man i in woman j's list

#### Why Stable Matchings?

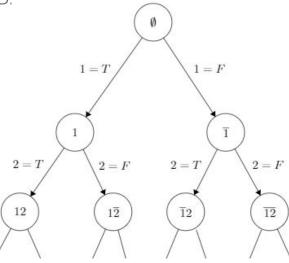
- Gale-Shapley algorithm solves SM problem in linear time. Why use SAT solvers?
- SMTI: stable matching problem where preference lists may be incomplete and contain ties
- SM-C: stable matching problem with couples
- Our encoding easily generalizes to SMTI, SM-C
- Theorem: SMTI and SM-C are NP-complete.

# **SAT is Hard!**

#### **Naive Search for SAT**



- Naive algorithm: try every possible assignment until we find a satisfying assignment or exhaust the search space
- Can interpret this as a DFS: (search tree)



#### **Overarching Class Themes**

- Accept the fact that the problems we will look at are very hard and "exponential runtime"
  - Take solace in the fact that for many inputs, the problem won't take exponential time
- Every speed-up counts
  - Take careful consideration of the balance between runtime and complexity
- There will never be a "right answer"
  - Often, the best thing to do for a problem depends on the problem itself and its data!

# Simplify the Search Space

Find a *minimal satisfying assignment* for the following formula:

 $\varphi = (x_1 \vee \overline{x_2} \vee \overline{x_3}) \land (\overline{x_2} \vee x_4) \land (\overline{x_1} \vee x_3 \vee x_5) \land (\overline{x_2} \vee \overline{x_1}) \land (\overline{x_2} \vee x_6 \vee x_7)$ 

Find a *minimal satisfying assignment* for the following formula:

 $\varphi = (x_1 \vee \overline{x_2} \vee \overline{x_3}) \land (\overline{x_2} \vee x_4) \land (\overline{x_1} \vee x_3 \vee x_5) \land (\overline{x_2} \vee \overline{x_1}) \land (\overline{x_2} \vee x_6 \vee x_7)$ 

#### $x_2 = \text{FALSE} \quad x_3 = \text{TRUE}$



#### **Trimming the Search Space**

• If a formula is satisfiable (has a satisfying assignment to variables), then in the assignment, each clause must individually evaluate to TRUE.

 $\varphi = C_1 \wedge C_2 \wedge \ldots \wedge C_n$ 



### **Trimming the Search Space**

- When we set *x* = *T*, what happens to the clauses containing *x*?
- **Observation 1:** Any clause containing the positive literal *x* becomes satisfied, so we no longer need to consider those clauses
  - In logic:  $(T \vee 1 \vee 2 \vee \cdots) = T$
  - Significance: we should remove all clauses containing *x*



#### **Trimming the Search Space**

- When we set x = T, what happens to the clauses containing  $\overline{x}$ ?
- **Observation 2:** Any clause containing the negative literal  $\overline{x}$  needs to be satisfied by a different literal, so we can ignore  $\overline{x}$  in that clause
  - In logic:  $(F \lor 1 \lor 2 \lor \cdots) = (1 \lor 2 \lor \cdots)$
  - Significance: we should remove  $\overline{x}$  from all clauses containing it



#### We are honing in on whatever is left that is unassigned and not yet evaluated to *TRUE*.

## **The Splitting Rule**



- The previous observations are called the **splitting rule**
- After repeatedly applying the splitting rule to formula  $\varphi$ :
  - If there are **no clauses left**, then all clauses have been satisfied, so  $\varphi$  is satisfied
    - $\varphi = \emptyset$  denotes that there are no clauses left
  - If φ ever contains an empty clause, then all literals in that clause are False, so we made a mistake
    - $\epsilon$  denotes the empty clause
    - $\epsilon \in \varphi$  denotes that  $\varphi$  contains an empty clause

## The Splitting Rule

- The splitting rule allows us to create a smarter recursive **backtracking** algorithm
- Backtracking: repeatedly make a guess to explore partial solutions, and if we hit "dead end" (contradiction) then undo the last guess





#### **Backtracking Notation**

- For a CNF φ and a literal x, define φ|x ("φ given x") to be a new CNF produced by:
  - Removing all clauses containing *x*
  - Removing  $\overline{x}$  from all clauses containing it
- Conditioning is "commutative":  $\varphi |x_1| x_2 = \varphi |x_2| x_1$



#### Backtracking (Pseudocode)

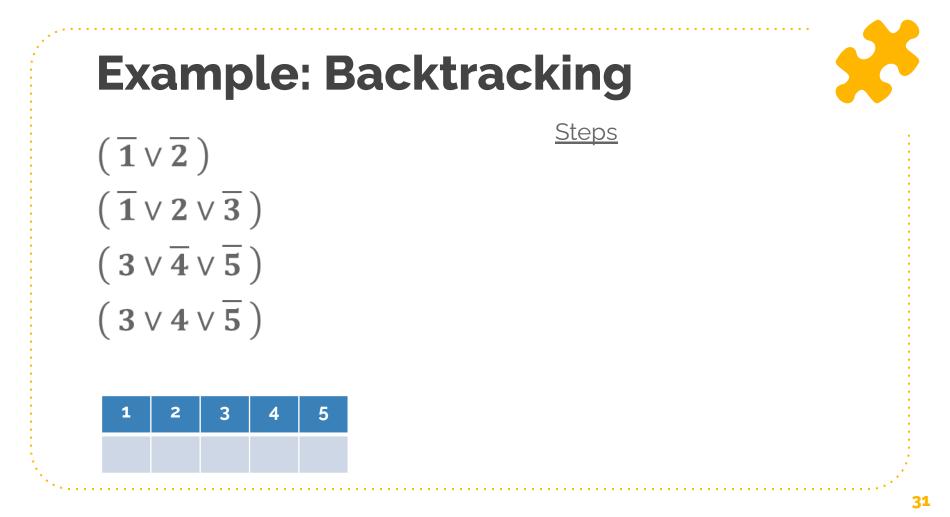
# check if  $\varphi$  is satisfiable **backtrack(\varphi):** 

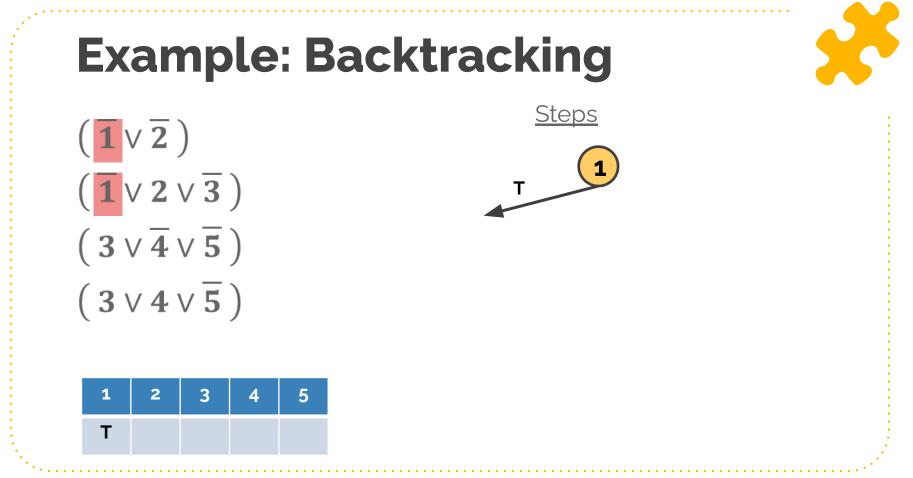
if  $\varphi = \emptyset$ : return True

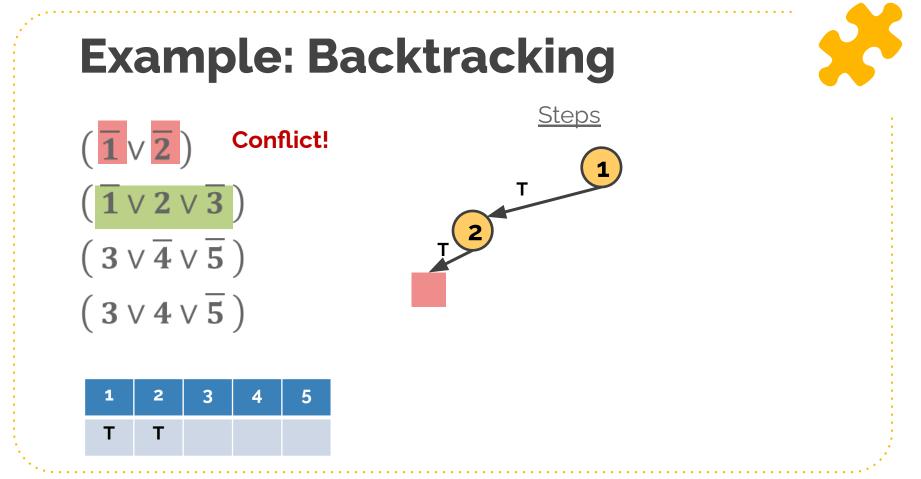
if  $\epsilon \in \varphi$ : return False

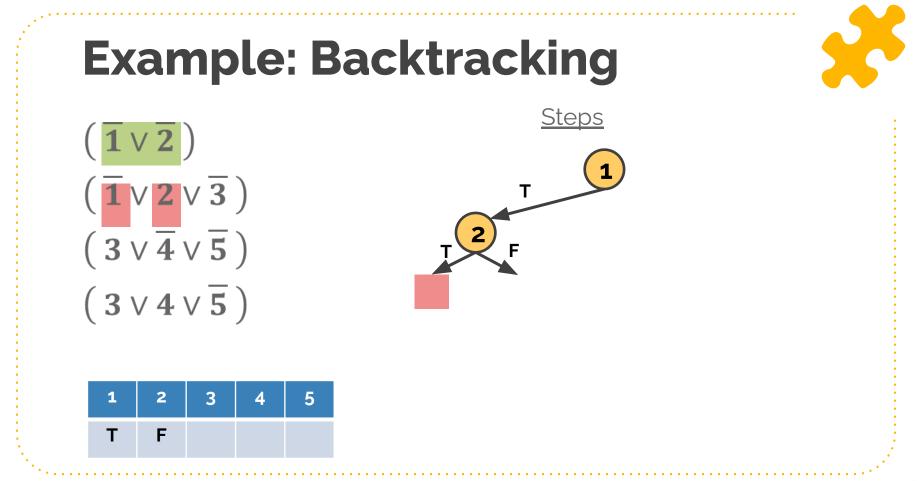
**let**  $x = pick_variable(\varphi)$ 

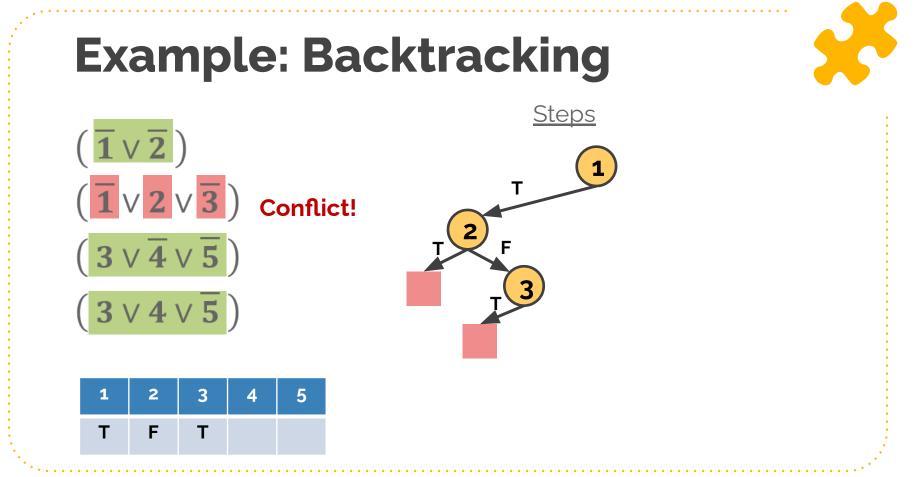
return backtrack( $\varphi \mid x$ ) OR backtrack( $\varphi \mid \overline{x}$ )

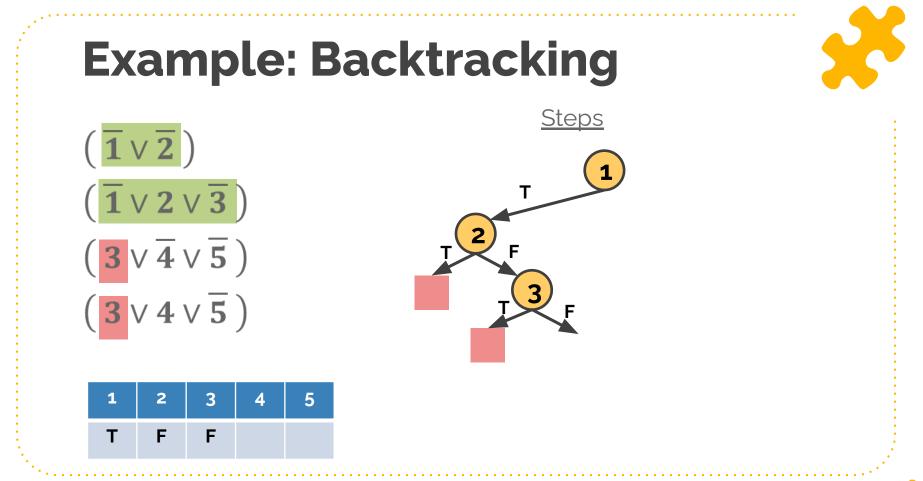


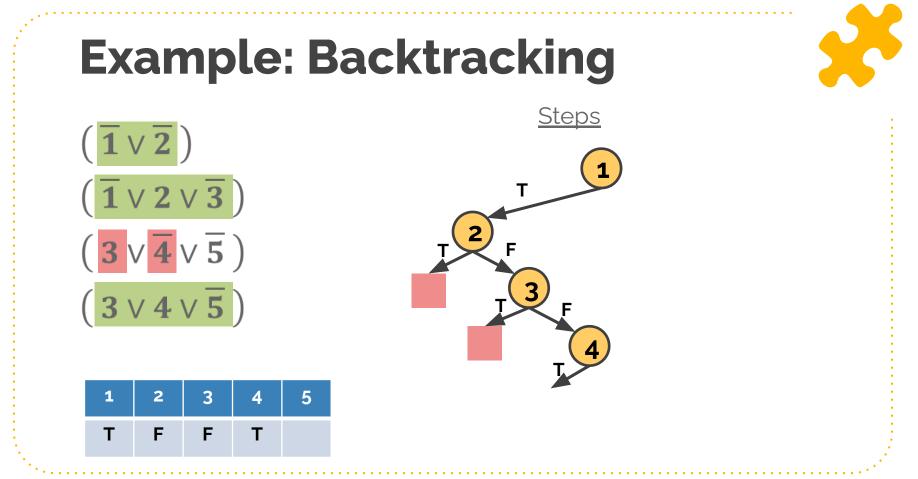


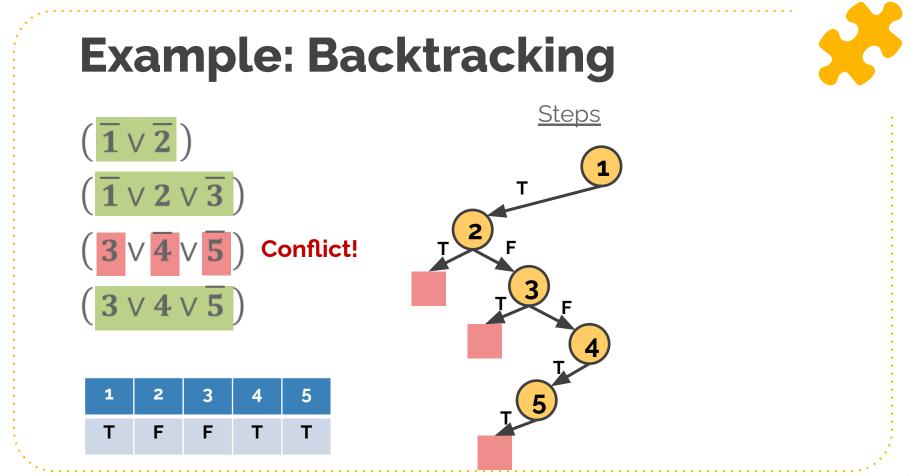


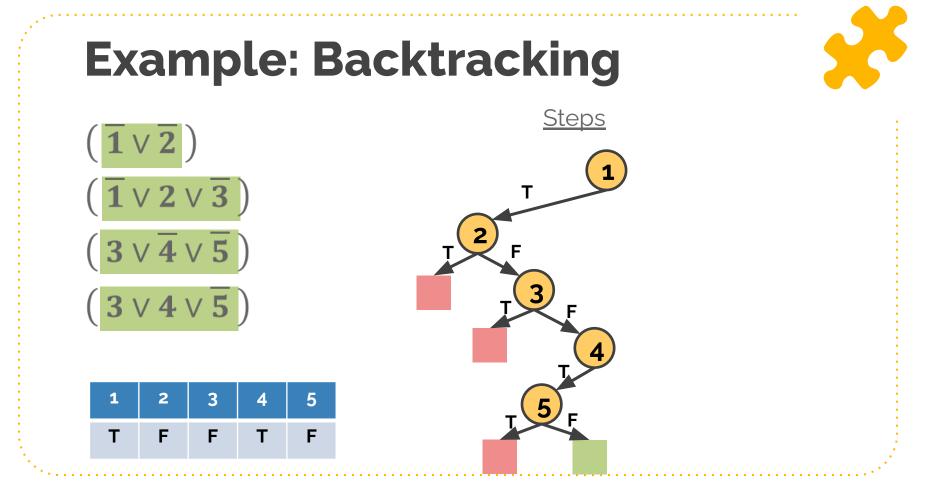












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#### **Towards Implementation: Efficient Splitting**

- How do we compute  $\varphi|x$ ?
  - Goals:
    - Support fast searching for empty clauses
    - Support fast backtracking
    - Fast to actually compute  $\varphi|x$

### Naïve Idea 1

- Transform φ into φ | x by deleting satisfied clauses and False literals from φ
  - Deletion not too expensive if we use linked lists
  - Can quickly recognize an empty clause (linked list will be empty), but need to check all clauses
  - Big issue: how do we backtrack?

### Naïve Idea 2

- Simple fix: instead of modifying  $\varphi$  directly, create a copy first and modify that
  - Easy backtracking just restore the old formula
  - Big issue: too expensive (time and memory) to copy formula every time we split
    - What if we have hundreds of thousands, even millions of clauses?





### Towards a smarter scheme

- Don't modify or copy the formula!
- **Key observation:** We must only backtrack once a clause has become empty *after* the Splitting Rule has been applied!



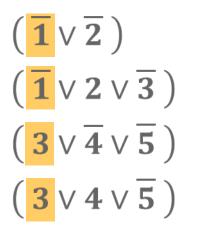
# **1** Watched Literal Scheme

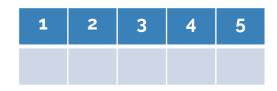
- **Observation:** a clause can only become empty if it has just one unassigned literal remaining
  - Ideally, only need to check these clauses
- Each clause "watches" one literal and maintains watching invariant: the watched literal is True or unassigned
  - If the watched literal becomes False, watch another
  - If there are no more True/unassigned literals to watch, then the clause must be empty



### **Example: 1 Watched Literal**



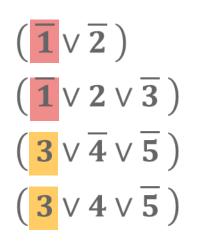


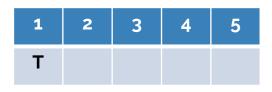




# Example: 1 Watched Literal

<u>Steps</u>

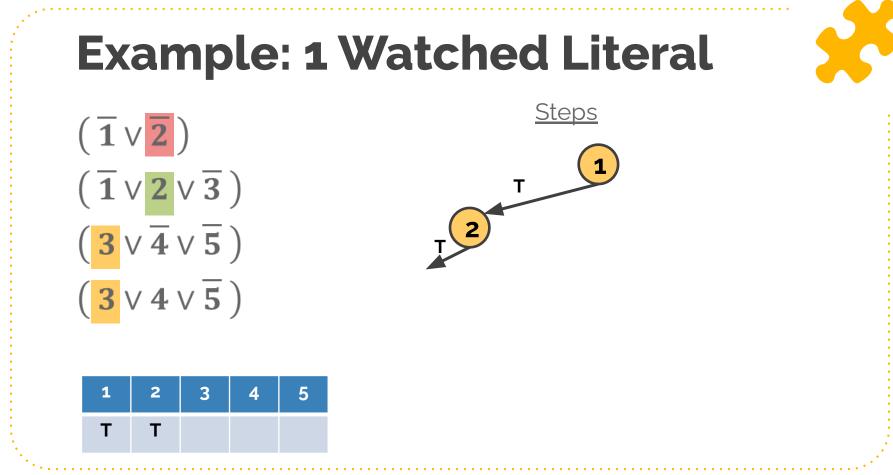






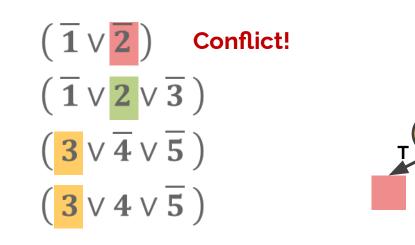


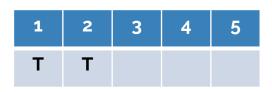
#### **Example: 1 Watched Literal** <u>Steps</u> $(\overline{1} \vee \overline{2})$ $\left( \begin{array}{c} \overline{1} \lor \overline{2} \lor \overline{3} \end{array} \right)$ $(\mathbf{3} \lor \mathbf{\overline{4}} \lor \mathbf{\overline{5}})$ $(3 \lor 4 \lor \overline{5})$ 1 2 3 5 4 Т

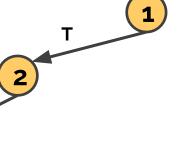




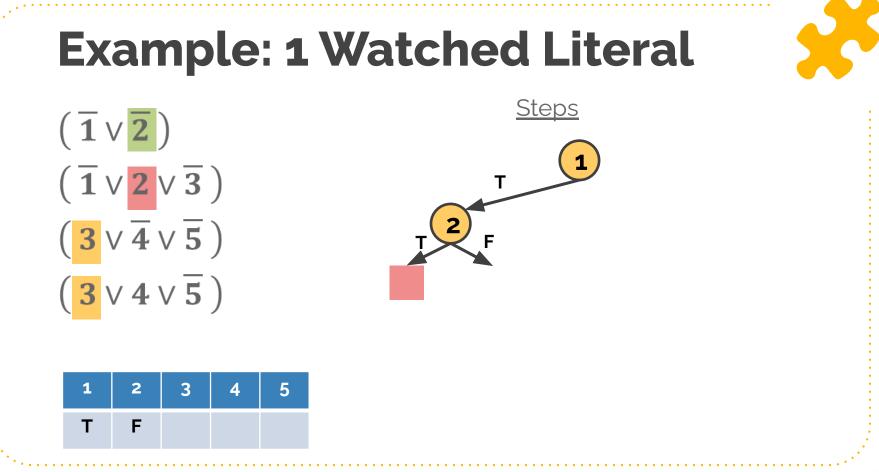
## **Example: 1 Watched Literal**

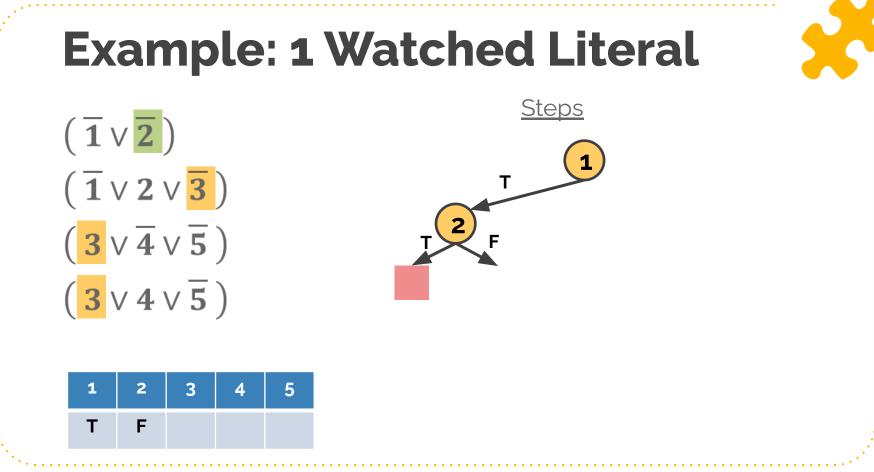


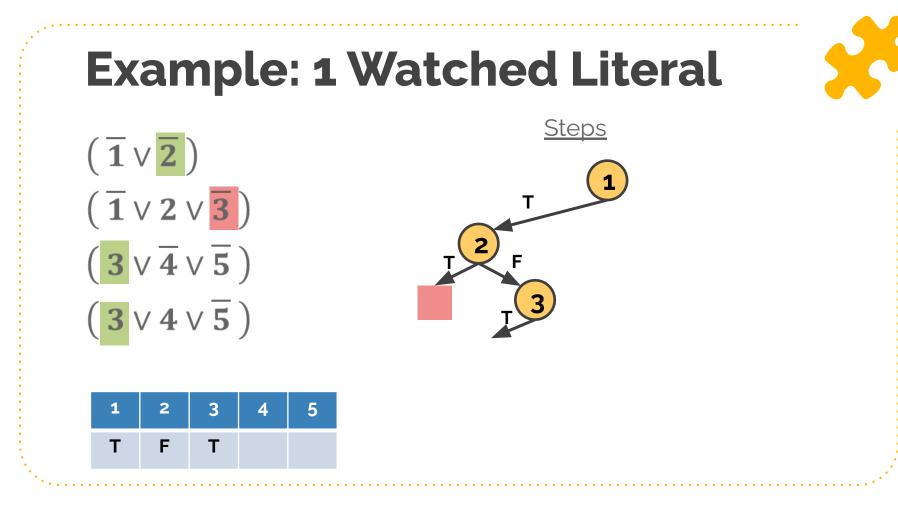


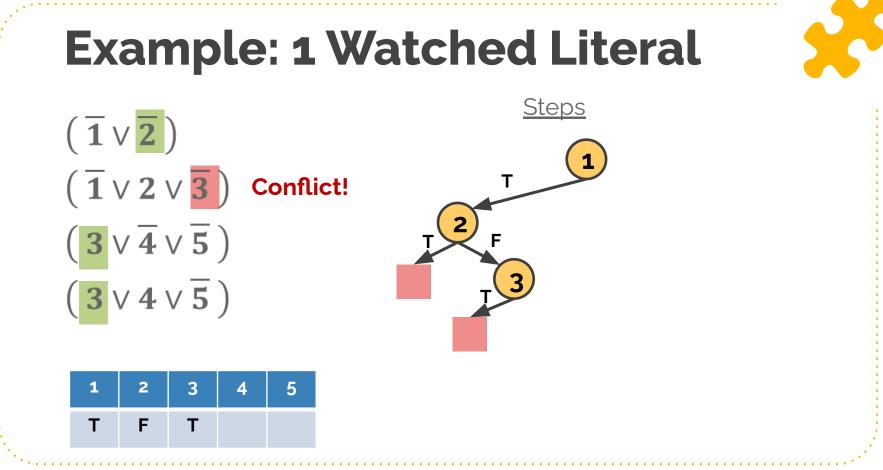


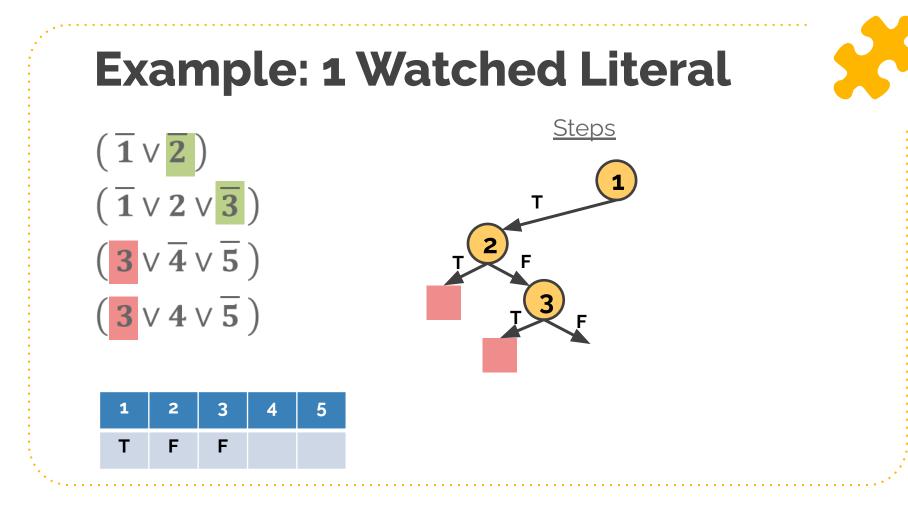
Steps

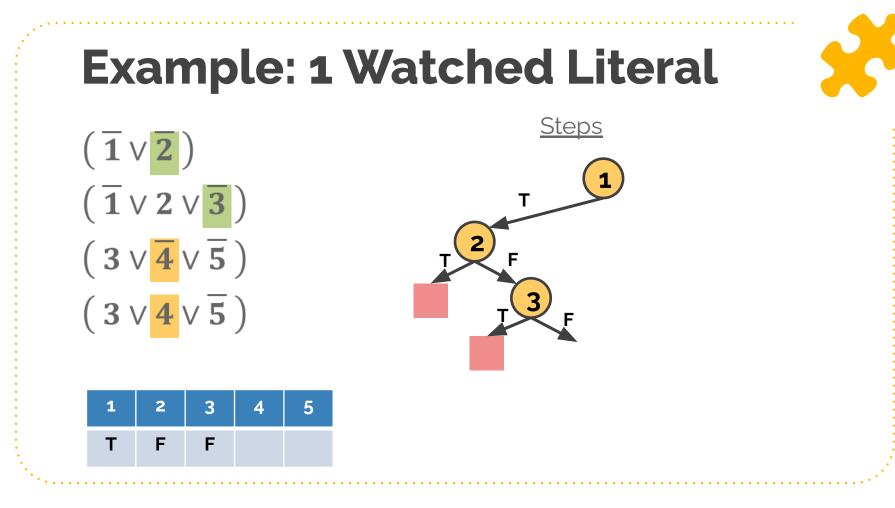


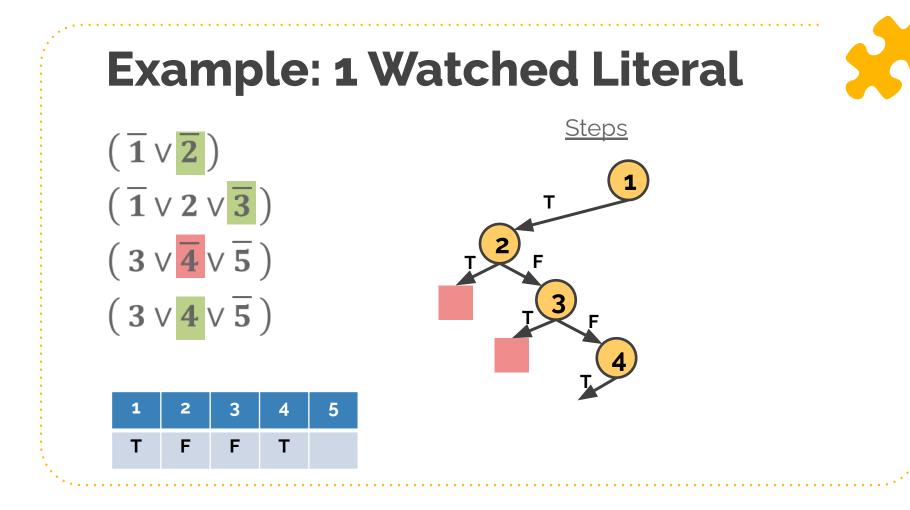


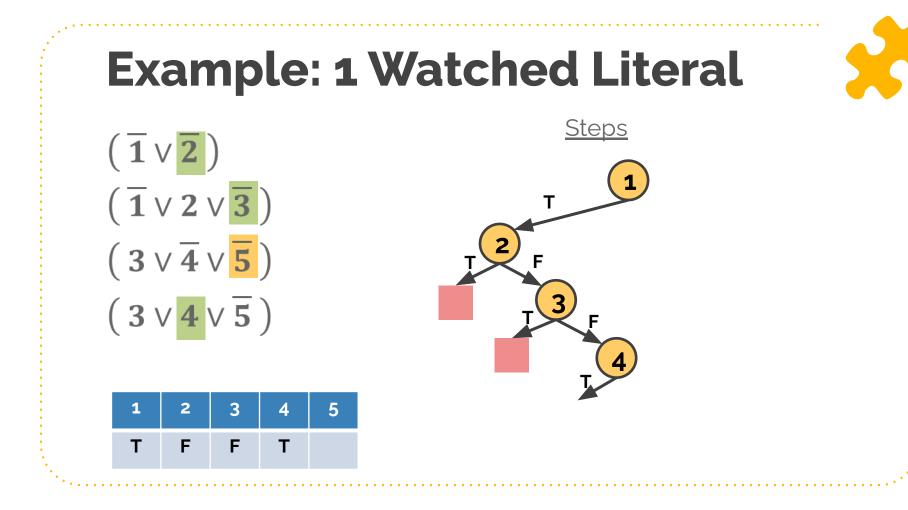


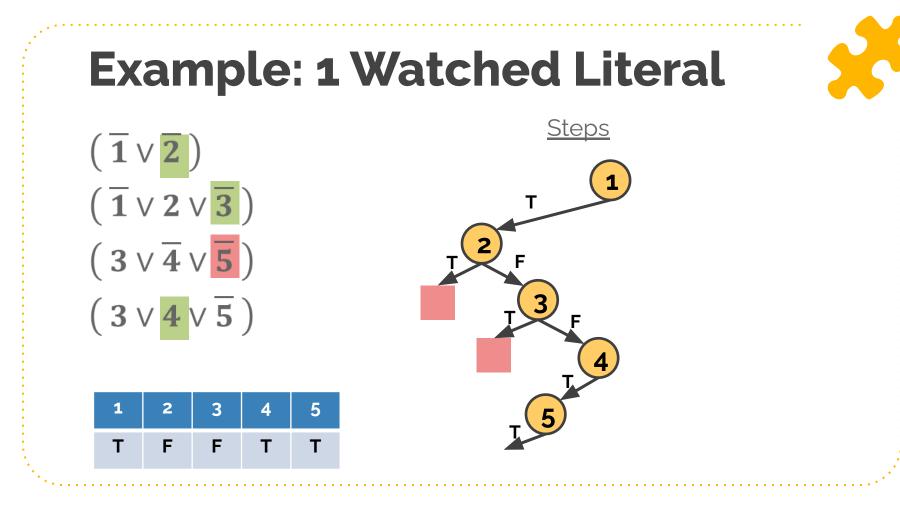


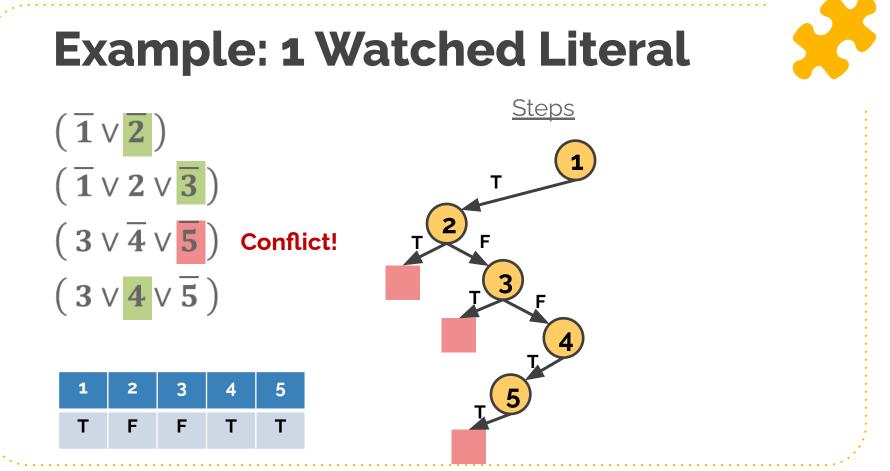


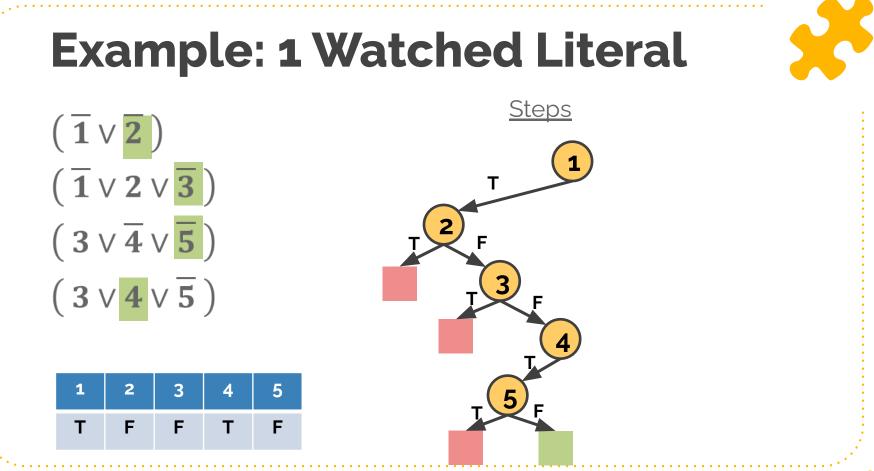












Find a *satisfying assignment* for the following formula:

 $\varphi = (x_1 \vee \overline{x_2} \vee \overline{x_3} \vee x_4) \land (\overline{x_1} \vee \overline{x_3}) \land (x_3) \land (x_4 \vee \overline{x_5} \vee \overline{x_7}) \land (x_3 \vee x_5 \vee x_6 \vee \overline{x_7}) \land (\overline{x_5} \vee \overline{x_6})$ 

Find a *satisfying assignment* for the following formula:

 $\varphi = (x_1 \vee \overline{x_2} \vee \overline{x_3} \vee x_4) \land (\overline{x_1} \vee \overline{x_3}) \land (x_3) \land (x_4 \vee \overline{x_5} \vee \overline{x_7}) \land (x_3 \vee x_5 \vee x_6 \vee \overline{x_7}) \land (\overline{x_5} \vee \overline{x_6})$ 

### $x_1 = \text{FALSE}$ $x_2 = \text{FALSE}$ $x_3 = \text{TRUE}$ $x_4 = \text{TRUE}$ $x_5 = \text{FALSE}$ $x_6 = \text{TRUE}$ $x_7 = \text{TRUE}$

# \*

# **Unit Propagation (UP)**

- A unit clause is a clause containing only one literal
- Unit propagation rule: for any unit clause  $\{\ell\}$ , we must set  $\ell = T$
- Applying unit propagation can massively speed up the backtracking algorithm in practice
  - Combining with the splitting rule can lead to a "domino effect" of cascading unit propagation

# The DPLL Algorithm

- Davis-Putnam-Logemann-Loveland (1962)
- Improved upon naive backtracking (search) with unit propagation (inference)
- Still the basic algorithm behind most state-of-the-art SAT solvers today!



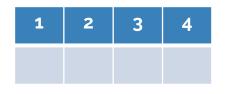


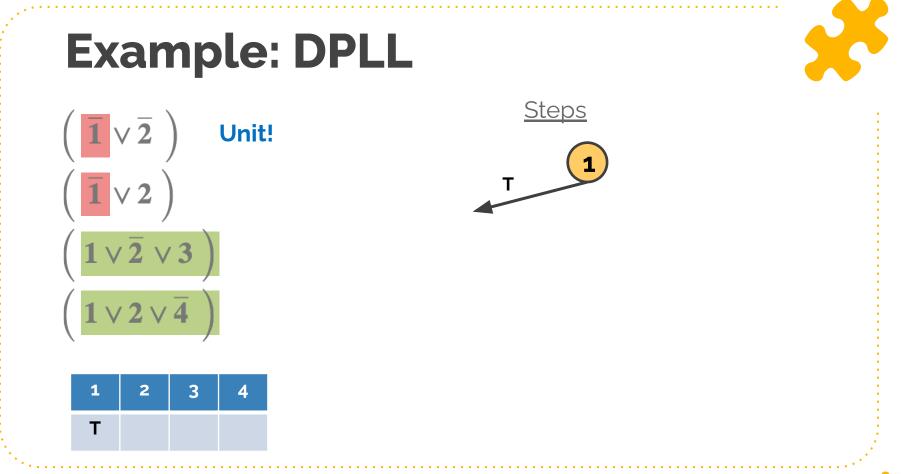
### **DPLL (Pseudocode)**

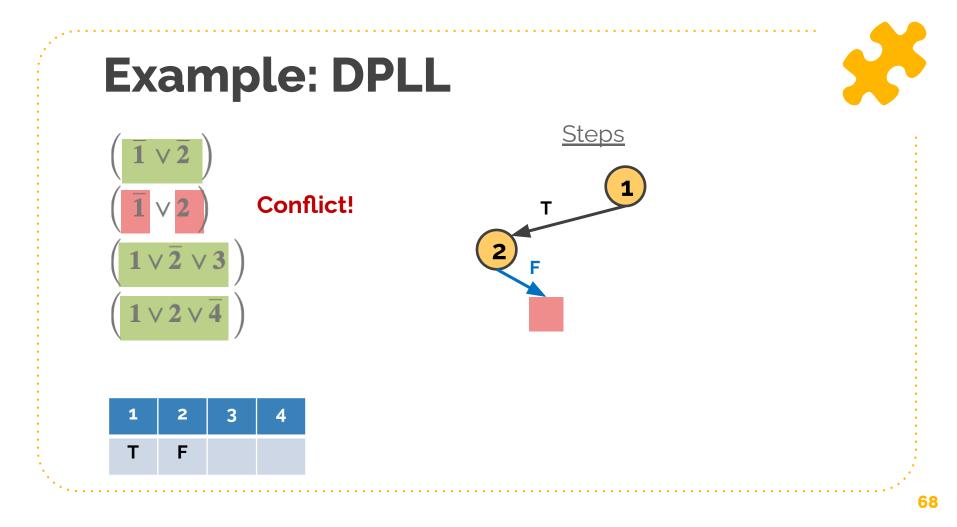
 $dpll(\varphi):$ if  $\varphi = \emptyset$ : return TRUE if  $\epsilon \in \varphi$ : return FALSE if  $\varphi$  contains unit clause  $\{\ell\}$ : return dpll( $\varphi | \ell$ ) let  $x = pick_variable(\varphi)$ return dpll( $\varphi | x$ ) OR dpll( $\varphi | \overline{x}$ )

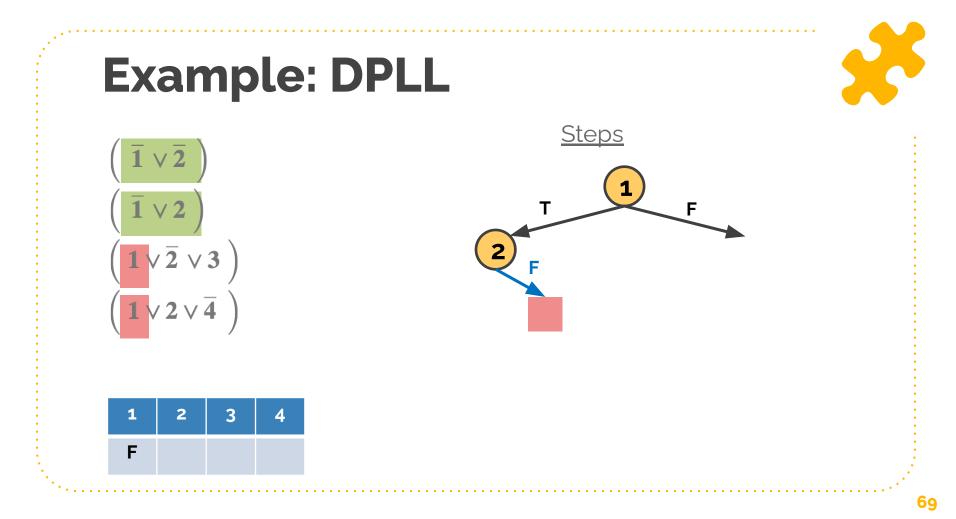


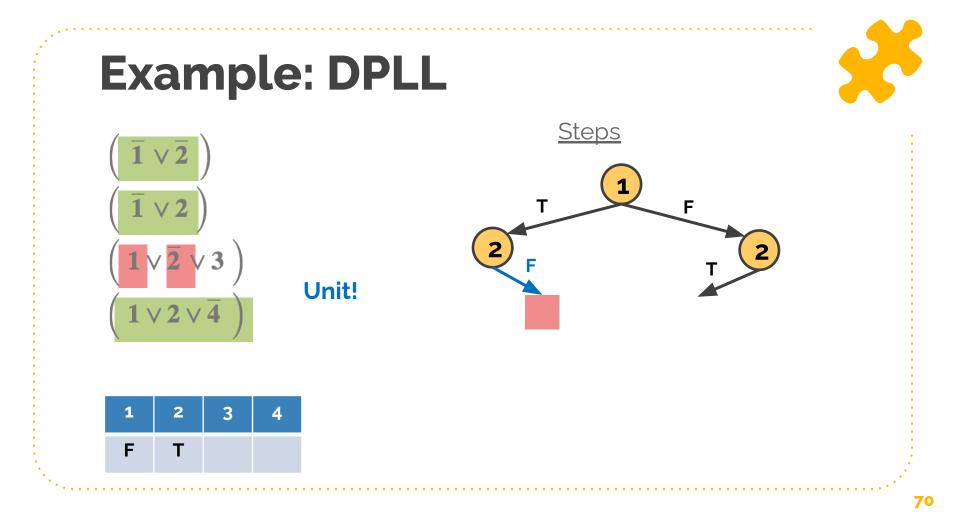


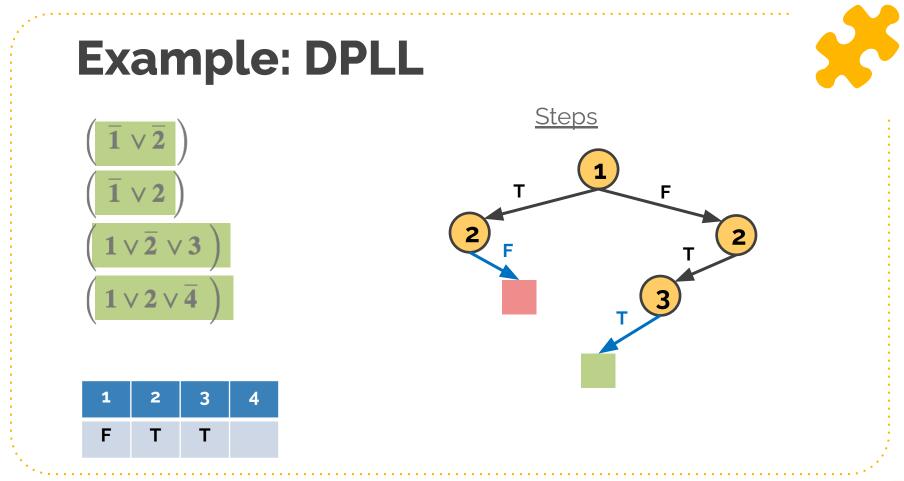














# **Engineering Matters**

- Since the main DPLL subroutine might run exponentially many times, every speedup counts
- DPLL spends by far the most time on UP
  - How can we speed this up?
- Although DPLL has a natural recursive formulation, recursion is slow lots of overhead
  - We can make DPLL **iterative** using a stack



# 2 Watched Literals (2WL)

- **Key observation:** a clause can only be unsatisfied or unit if it has at most one non-False literal
  - Optimize unit propagation: only visit those clauses
- Each clause "watches" two literals and maintains watching invariant: the watched literals are not False, unless the clause is satisfied
  - If a watched literal becomes False, watch another
- If can't maintain invariant, clause is unit (can propagate)



# 2 Watched Literals (2WL)

- Still use watchlists (list of all clauses watching each lit)
- Best part: since backtracking only unassigns variables, it can never break the 2WL invariant
  - Don't need to update watchlists

$$\left(\begin{array}{ccc} \mathbf{1} \lor \mathbf{2} \lor \mathbf{\overline{3}} \end{array}\right) \xrightarrow{\text{Set 1} = T} \left(\begin{array}{ccc} \mathbf{\overline{1}} \lor \mathbf{2} \lor \mathbf{\overline{3}} \end{array}\right) \xrightarrow{\text{Set 2} = F} \left(\begin{array}{ccc} \mathbf{\overline{1}} \lor \mathbf{2} \lor \mathbf{\overline{3}} \end{array}\right)$$

## How should we branch?



- Order of assigning variables greatly affects runtime
- Want to find a satisfying assignment quicker and find conflicts (rule out bad assignments) quicker
- **Ex:**  $\{1\overline{2}34, \overline{12}3, 12\overline{3}5, 23\overline{5}, 3\overline{45}, \dots, 67, \overline{67}, \overline{67}, \overline{67}\}$

• If we assign 6 first, then we can find conflicts right away

# **Decision Heuristics**



- Static heuristics: variable ordering fixed at the start
- **Dynamic heuristics:** variable ordering is updated as the solver runs
  - More effective, but also more expensive
- Basic examples of decision heuristics:
  - Random ordering
  - Most-frequent static ordering
  - Most-frequent dynamic ordering



"Intelligence is knowing it is a one-way street, wisdom is still looking both ways before crossing."

### References



A. Biere, Handbook of satisfiability. Amsterdam: IOS Press, 2009.

N. Eén and N. Sörensson, "An Extensible SAT-solver," *Theory and Applications of Satisfiability Testing Lecture Notes in Computer Science*, pp. 502–518, 2004.