

# Algorithms for Computational Logic

## Introduction

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## Outline

- 1 Introduction to Boolean Satisfaction
- 2 Boolean Reasoning

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  - Propositional Logic
  - The Satisfiability Problem
  - Some Fragments of Propositional Logic

- 2 Boolean Reasoning
  - Unit Propagation
  - Resolution
  - Proof Systems

### Proposition

A *proposition* is an assertion that can be:

- assigned a truth value (**true** or **false**)
- written using *atomic propositions* (or *atoms*) and *logic connectors*

An atom is a proposition written using a unique symbol.

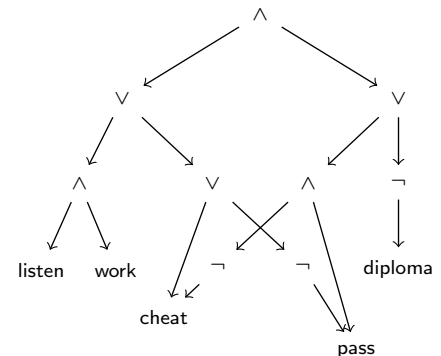
- Atomic propositions:
  - ▶ “Adam follows the lecture”, “Adam works at home”, “Adam cheats at the exam”, “Adam passes the exam”
- Propositions:
  - ▶ “if Adam does not listen the lecture **and** does not work at home **then** he will not pass the exam **unless** he cheats

## Formulae (syntax)

A non-atomic proposition (*Formula*)  $\varphi$  is either:

- an atom
- the negation  $\neg\psi$  of another proposition  $\psi$
- the concatenation of two or more propositions  $\varphi_1$  and  $\varphi_2$  by a logical connector  $\{\wedge, \vee, \rightarrow, \oplus, \dots\}$

$((\text{"listen lecture"} \wedge \text{"work at home"}) \vee \text{"cheat"} \vee \neg \text{"pass exam"}) \wedge$   
 $((\neg \text{"cheat"} \wedge \text{"pass exam"}) \vee \neg \text{"get diploma"})$



## Models (interpretations)

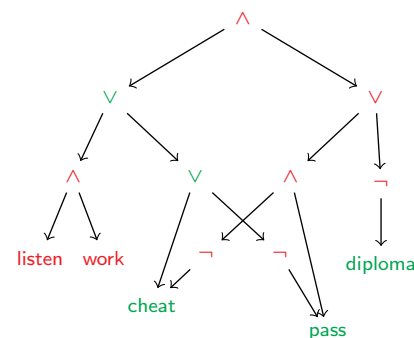
A *model*  $\mathcal{A}$  is a mapping from atoms in  $\mathcal{X}$  to  $\{\text{true}, \text{false}\}$ . We write  $\mathcal{A} \models x$  for “Atom  $x$  is true in model  $\mathcal{A}$ ”

A proposition  $\varphi$  written using atoms in  $\mathcal{X}$  can be interpreted (given a truth value) using a model  $\mathcal{A}$  on  $\mathcal{X}$ :

- if  $\varphi$  is the negation of a proposition  $\psi$ , then  $\mathcal{A} \models \varphi$  if and only if  $\mathcal{A} \not\models \psi$
  - if  $\varphi$  is a conjunction  $\varphi_1 \wedge \varphi_2$ , then  $\mathcal{A} \models \varphi$  if and only if  $\mathcal{A} \models \varphi_1$  **and**  $\mathcal{A} \models \varphi_2$
  - if  $\varphi$  is a disjunction  $\varphi_1 \vee \varphi_2$ , then  $\mathcal{A} \models \varphi$  if and only if  $\mathcal{A} \models \varphi_1$  **or**  $\mathcal{A} \models \varphi_2$
- Ex: “listen lecture”  $\wedge$  “work at home”  $\wedge$   $\neg$  “cheat”  $\wedge$   $\neg$  “pass exam”  $\wedge$   $\neg$  “get diploma”

$$((\text{"listen lecture"} \wedge \text{"work at home"}) \vee \text{"cheat"} \vee \neg \text{"pass exam"}) \wedge ((\neg \text{"cheat"} \wedge \text{"pass exam"}) \vee \neg \text{"get diploma"})$$

	"listen lecture"	"work at home"	"cheat"	"pass exam"	"get diploma"
$\mathcal{A}_1$	false	false	true	true	true



## SAT

- **data:** A Boolean formula  $\phi$
  - **question:** Does there exist an interpretation that satisfies  $\phi$ ?
- 
- A formula is *satisfiable* iff there exists an interpretation that satisfies it
  - A formula  $\varphi$  is *unsatisfiable* iff there is no interpretation that satisfies it
    - ▶ Write it UNSAT ( $\varphi$ )
  - A formula is *valid* / a *tautology* iff all interpretations satisfy it
    - ▶ Equivalent to UNSAT ( $\neg\varphi$ )
  - A formula  $\psi$  is an *implicate* of  $\varphi$  iff all interpretations satisfying  $\varphi$  also satisfy  $\psi$ 
    - ▶ Equivalent to UNSAT ( $\varphi \wedge \neg\psi$ )
  - A formula  $\psi$  is an *implicant* of  $\varphi$  iff  $\varphi$  is an *implicate* of  $\psi$

- Linux package upgrade
  - ▶ The Eclipse foundation uses Daniel le Berre's SAT solver **SAT4j** to solve this problem
  - ▶ Equinox/p2/CUDFResolver
- (Re-)Attribution of the TV radiospectrum by the Federal Communications Commission (FCC) in 2017
  - ▶ The radiofrequency allocation problem corresponds to *Graph Coloring*
    - ★ Vertices are broadcasters, colors are frequencies
    - ★ Easy to encode as SAT
  - ▶ Reverse auction: the FCC buys frequencies and starts with high quotes that decrease at each round
    - ★ Stops when it is *not* possible to assign frequencies to broadcasters who opted out
  - ▶ Critical to *prove* unsatisfiability (the auction yielded \$20 billion)

- SAT is in **NP**, the interpretation  $\sigma$  that satisfies it is a polynomial certificate

## Théorème de Cook-Levin

SAT is **NP**-complete

- ▶ At least as hard as any problem in **NP**
- ▶ If SAT is in **P** then **P** = **NP**

- Fragments of SAT are particular case defined by the *language*
  - ▶ Using only negation ( $\neg$ ), disjunction ( $\vee$ ) and conjunction ( $\wedge$ ) is not restrictive

- Disjunctive normal form:
  - ▶ Disjunction of conjunctions (sum) of literals (products)
  - ▶ Ex:  $(\neg a \wedge b \wedge c) \vee (\neg b \wedge \neg c) \wedge (a \wedge \neg b)$
- Every product is an *implicant*, and corresponds to an interpretation
- Satisfiability of a DNF is easy

- Conjunctive normal form:
  - ▶ **Conjunction** of **disjunctions** of literals (clauses)
  - ▶ Ex:  $(\neg a \vee b \vee c) \wedge (\neg b \vee \neg c) \wedge (a \vee \neg b)$
- For any formula  $\varphi$ , there is a CNF formula  $\varphi'$  such that
  - ▶  $\text{SAT}(\varphi) \iff \text{SAT}(\varphi')$
  - ▶  $|\varphi'| \in \mathcal{O}(|\varphi|^c)$  for some constant  $c$
- Every clause is an *implicate*
- **Validity of a CNF is easy**

- Horn clause:
  - ▶ Clause with at most one positive literal
  - ▶ Ex:  $(\neg a \vee \neg c \vee b) \wedge (\neg b \vee \neg c) \wedge (\neg b \vee a)$
  - ▶ Equivalent to implications
    - ★  $(a \wedge c \Rightarrow b) \wedge (b \wedge c \Rightarrow \text{false}) \wedge (b \Rightarrow a)$

- Comments
- Header [#variables(=5)] [#clauses(=7)]
- Variables are numbered 1 to  $n$
- One line per clause '0' is a delimiter
- positive (negative) numbers are positive (negative) literals
  - ▶  $(\neg x_1 \vee x_3 \vee \neg x_5 \vee x_4)$

```
c This line is a comment.
p cnf 5 7
-1 3 -5 4 0
2 -3 0
1 5 0
-3 -4 0
-1 2 4 0
-2 0
2 -3 -5 0
```

- Typename/classes
  - ▶ **Variable**: used for indexing  $\rightarrow$  e.g., int from 0 to  $n - 1$
  - ▶ **Literal**: used for indexing  $\rightarrow$  e.g., int from 0 to  $2n - 1$
  - ▶ **TruthValue**: three possibility (true, false, undef)  $\rightarrow \{1, 0, -1\}$
  - ▶ **Clause**: iterable list of literals
- Functions on variables
  - ▶  $\text{pos}(\text{Variable}:x) \mapsto \text{Literal } x$  (e.g.,  $2x + 1$ )
  - ▶  $\text{neg}(\text{Variable}:x) \mapsto \text{Literal } \neg x$  (e.g.,  $2x$ )
- Functions on literals
  - ▶  $\text{sign}(\text{Literal}:l) \mapsto \{\text{false}, \text{true}\}$  (e.g.,  $l\%2$ )
  - ▶  $\text{not}(\text{Literal}:l) \mapsto \neg l$  (e.g.,  $l^{\wedge}1$ )
  - ▶  $\text{var}(\text{Literal}:l) \mapsto x$  (e.g.,  $l/2$ )



## • Data structures

- ▶ **model** [**Variable** :  $x$ ]  $\mapsto$  **TruthValue** stores the current truth value of  $x$
- ▶ **clauses** [**Literal** :  $l$ ]  $\mapsto$  [**Clause**,...] list of clauses containing literal  $l$
- ▶ **unit-literals** stack of true literals (efficient **push(Literal:l)** and **Literal:back()** and **pop-back()**)

## • Functions

- ▶ **val(Variable:x)**  $\mapsto$  **TruthValue** truth value of variable  $x$
- ▶ **falsified(Literal:l)**  $\mapsto$  Boolean literal is falsified in **model**
- ▶ **satisfied(Literal:l)**  $\mapsto$  Boolean literal is satisfied in **model**

## • IN/OUT

- ▶ Functions **from-dimacs(int:d)**  $\mapsto$  **Literal** and **to-dimacs(Literal:l)**  $\mapsto$  **int**
- ▶ Functions **read-dimacs()** and **write-dimacs()**

## 1 Introduction to Boolean Satisfaction

- Propositional Logic
- The Satisfiability Problem
- Some Fragments of Propositional Logic

## 2 Boolean Reasoning

- Unit Propagation
- Resolution
- Proof Systems

- A clause forbids exactly one tuple

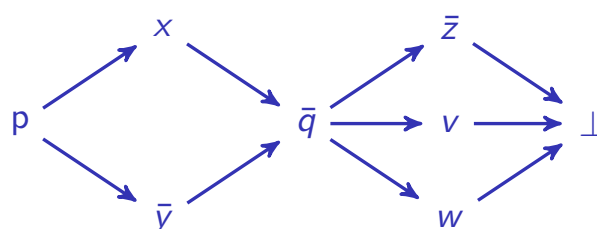
$$(\bar{x} \vee y \vee z \vee \bar{v} \vee \bar{w}) \iff \neg(x \wedge \bar{y} \wedge \bar{z} \wedge v \wedge w)$$

- What can we deduce by looking at just *one* clause?
- Nothing unless it is a unit clause ( $p$ ): then we deduce that the literal  $p$  is true
  - ▶  $x$  is true if  $p = x$
  - ▶  $x$  is false if  $p = \bar{x}$
- If the clause has two (independent) literals, any one can be false, providing that the other is true
- Incomplete proof system (e.g.  $(x \vee a) \wedge (\bar{x} \vee a) \wedge (\bar{y} \vee \bar{a}) \wedge (y \vee \bar{a})$ )

- However it *propagates*: if we have the unit literal  $p$ , a clause containing  $\bar{p}$  can be reduced, and maybe become unit, triggering more unit propagation

$$(\bar{x} \vee y \vee z \vee \bar{v} \vee \bar{w}) \wedge (\bar{p} \vee x) \wedge (\bar{p} \vee \bar{y}) \wedge (q \vee \bar{z}) \wedge (q \vee v) \wedge (p) \wedge (q \vee w) \wedge (\bar{q} \vee \bar{x} \vee y)$$

- $(p)$  is a unit clause
- $(x)$  and  $(\bar{y})$
- $(\bar{q})$  is a unit clause
- $(\bar{z})$ ,  $(v)$  and  $(w)$  are unit clauses
- Unit propagation produces an empty clause



- Unit propagation solves *Horn*-SAT
- If a *Horn*-SAT formula has no unit clause, then every clause has at least one negative literal
  - ▶ The model with all variables false satisfies the formula
- Otherwise, unit propagate until reaching an inconsistency or a subformula without unit clauses

- A clause can either be:
  - ▶ Satisfied iff it contains at least one true literal
  - ▶ Falsified iff it contains only false literals
  - ▶ Unit iff it contains a single unknown literal, and  $n - 1$  false literals
  - ▶ Unresolved iff it contains no true literal and at least two unknown literals



## Unit propagation algorithm (counters)

Organise clauses per literals ( $Clauses(l)$  is the set of clauses containing literal  $l$ )

keep an initially null counter  $\#f_i$  of false literals for each clause  $c_i$

Put all unit clauses (*true literals*) in a list

**while** *There is a non-processed true literal  $l$*  **do**

    mark  $l$  as processed

**foreach**  $c_i \in Clauses(l)$  **do**

        increment  $\#f_i$

// at most once per literal:  $O(s)$

**if**  $\#f_i = |c_i|$  **then** return *FAIL*

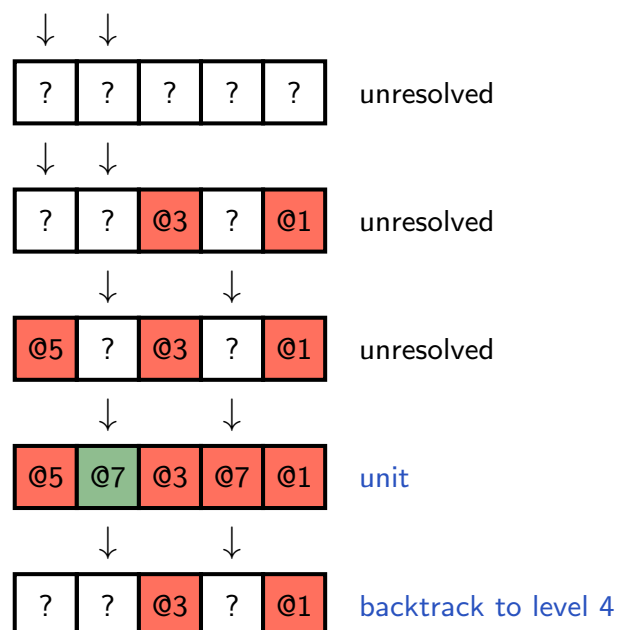
**if**  $\#f_i = |c_i| - 1$  **then**

            find the last literal and add it to the list of true literals

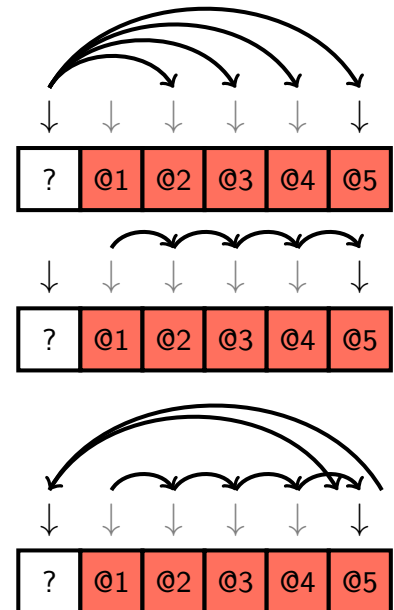
//  $\Theta(|c_i|)$  at most once per clause:  $O(s)$

- Let  $\varphi$  have  $n$  variables and  $m$  clauses, and let  $s$  be the total number of literals  $s = \sum_{i=1}^m |c_i|$
- Worst case: every variable  $x$  is unit propagated ( $x$  if  $|Clauses(x)| \geq |Clauses(\bar{x})|$ , and  $\bar{x}$  otherwise)
- Overall linear time  $\Theta(s)$  **amortized down a branch**

- Invariant *Watch* only two *non-false* literals per clause
  - $Watch(l)$  is the list of clauses that *watches* literal  $l$
- Non-watched literals can become false, it *cannot* make the clause *unit* or *falsified* as long as two unknown literals remain
- When a watched literal become false, a **replacement must be found**
- When no replacement can be found, the clause is either unit or falsified
- Nothing to do when backtracking*: the literals watched at level  $i$  cannot be false at level  $i - 1$



- Scan the clause from first to last literal: possibly  $\Theta(|c_i|)$  scans each costing  $\Theta(|c_i|)$ 
  - Quadratic
- Store the initial position of the watch and scan *forward*
  - Linear but we must update the position of the watchers when *backtracking*
- Circular list: scan *forward*, but past the end and back to the *current* position
  - The clause is scanned at most twice: linear and no need to do anything when backtracking!



- Let  $n$  be the number of variables,  $m$  be the number of clauses,  $s = \sum_{i=1}^m |c_i|$  be the overall size of the formula,  $k$  be the number of true literals after unit propagation
- Consider first the clauses that unit propagated
  - They contain only variables among the  $k$  true literals
  - In order to propagate them, every literal must be explored (to increment the counter of find a new watched): it takes linear time in both cases call that  $O(K)$
- Consider now the  $m'$  clauses that did not unit propagate (and let  $s'$  be their total size)
  - The counters algorithm increments the counters of every clause containing one of the  $k$  true literals
    - ★ The average number of clauses per literal is  $\frac{s'}{n}$  so  $\Theta\left(\frac{ks'}{n}\right)$  time in average
  - Overall:  $\Theta(O(K) + \frac{ks'}{n})$  time
  - The watched algorithm increments finds a new watched literal for each of the clauses that watch it
    - ★ A literal is watched by  $\frac{m'}{n}$  of these clauses in average
    - ★ The probability that a random literal is not false is  $\frac{n-k}{n}$ , so the expected number of literals to scan to find a valid one to watch is  $\frac{n}{n-k}$
  - Overall:  $\Theta(O(K) + \frac{km'}{n-k})$  time

## ● Structure

▶ **watches** [**Literal** :  $l$ ]  $\mapsto$  [**Clause**, ...]

list of clauses watching literal  $l$

▶ **int:to-propagate**

the first non-unit-propagated literal in **unit-literals**

## ● Functions

▶ **get-rank**(**Clause**: $c$ , **Literal**: $l$ )  $\mapsto$  {0, 1}

0 if  $l$  is the first watched in  $c$ , 1 otherwise

▶ **get-index**(**Clause**: $c$ , {0, 1}: $r$ )  $\mapsto$  **int**

index of the  $(r + 1)$ -th watched in  $c$

▶ **set-watcher**(**Clause**: $c$ , **Literal**: $l$ , {0, 1}: $r$ )

set  $l$  as  $(r + 1)$ -th watcher of  $c$

▶ **assign**(**Literal**: $l$ )

push  $l$  onto **unit-literals** and set **model** [**var**( $l$ )]

## Unit propagation algorithm (watched literals)

**Algorithm:** **unit-propagate**()

```
while to-propagate < |unit-literals| do
  l ← not(unit-literals [to-propagate])
  if not unit-propagate(l) then
    return false
  to-propagate ← to-propagate + 1
return true
```

**Algorithm:** **unit-propagate**( $l$ )

**Input:** A non-unit propagated false literal  $l$

**Output:** false in case of a contradiction, true otherwise

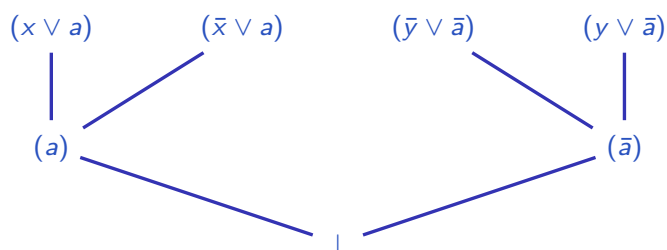
```
foreach c ∈ clauses[l] do
  r ← get-rank(c, l); start ← i ← get-index(c, r)
  p ← c[get-index(c, 1-r)]
  if not satisfied(p) then
    while true do
      i ← i + 1
      if i = |c| then i ← 0
      if i = start then break
      if c[i] ≠ p then
        if not falsified(c[i]) then
          set-watcher(c, c[i], r)
          break
    if i = start then
      if falsified(p) then return false
      assign(p)
return true
```

- Resolution rule:

[DP60,R65]

$$\frac{(\alpha \vee x) \quad (\beta \vee \bar{x})}{(\alpha \vee \beta)}$$

- Complete proof system for propositional logic: If the formula  $\varphi$  is not satisfiable, then there is sequence of resolution steps that produce the *empty clause*  $\perp$



- **Self-subsuming** resolution (with  $\alpha' \subseteq \alpha$ ):

[e.g. SP04,EB05]

$$\frac{(\alpha \vee x) \quad (\alpha' \vee \bar{x})}{(\alpha)}$$

## Theorem

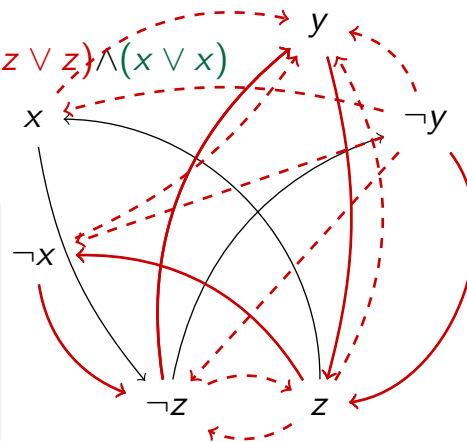
## Resolution solves 2-SAT in polynomial time

- Resolution is a complete refutation system for SAT (and hence for 2-SAT)
- Resolvent clauses have at most 2 literals
  - ▶ There are at most  $n^2$  binary clauses

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

### Algorithm

- $x \vee y$  is equivalent to  $\neg x \Rightarrow y$  and  $\neg y \Rightarrow x$
- Add transitive edges
  - ▶ If there is an inconsistency, then the formula is not satisfiable
  - ▶ If not, it is satisfiable, because the choice  $x \Rightarrow \neg x$  closes a cycle only if there is a path  $\neg x \Rightarrow x$



- SAT is in **NP**: if an instance is satisfiable, it is possible to prove it efficiently
  - ▶ Just show a model and check clause by clause that it is correct (it is a **certificate**)
- What about the question "is  $\varphi$  *unsatisfiable*?", or "is  $\varphi$  a *tautology*?"
  - ▶ There might not exist short certificates for problems in **coNP**, but we can provide a *long* one
- Proof system: maps to every *unsatisfiable* formula  $\varphi$  a refutation **R**
  - ▶ There is a polynomial algorithm (in  $|R|$ ) to check the refutation proof

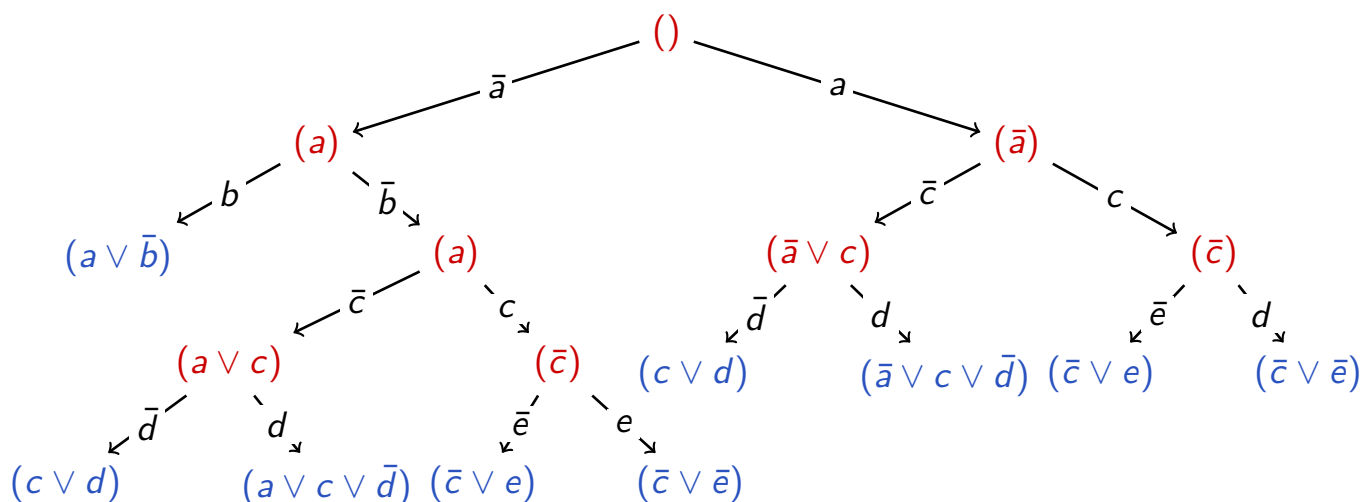
★ *Pebbling* formulas



$$\varphi = (a \vee \neg b) \wedge (\neg a \vee c \vee \neg d) \wedge (a \vee c \vee \neg d) \wedge (\neg c \vee \neg e) \wedge (\neg c \vee e) \wedge (c \vee d)$$

$c_1$	$=$	$(\neg c \vee e)$	$\in \varphi$
$c_2$	$=$	$(\neg c \vee \neg e)$	$\in \varphi$
$c_3$	$=$	$(\neg c)$	resolvent of $c_1$ and $c_2$
$c_4$	$=$	$(a \vee c \vee \neg d)$	$\in \varphi$
$c_5$	$=$	$(\neg a \vee c \vee \neg d)$	$\in \varphi$
$c_6$	$=$	$(c \vee \neg d)$	resolvent of $c_4$ and $c_5$
$c_7$	$=$	$(c \vee d)$	$\in \varphi$
$c_8$	$=$	$(c)$	resolvent of $c_6$ and $c_7$
$c_9$	$=$	$()$	resolvent of $c_3$ and $c_8$

$$\varphi = (a \vee \neg b) \wedge (\neg a \vee c \vee \neg d) \wedge (a \vee c \vee \neg d) \wedge (\neg c \vee \neg e) \wedge (\neg c \vee e) \wedge (c \vee d)$$



- *Soundness*: if there exists a resolution refutation then the formula is unsatisfiable
  - ▶ Resolution is a *sound* proof system simply because the resolution step is sound
- *Completeness*: if a formula is unsatisfiable then there exists a resolution refutation of that formula
  - ▶ Tree search is obviously a complete proof system
  - ▶ To every search tree we can associate a resolution proof
  - ▶ Therefore resolution is a *complete* proof system

- What does make a proof system good? (besides soundness and completeness)
- A good proof system is one that allows shorter proofs
  - ▶ If refutations are polynomial size in general, then  $\text{NP} = \text{coNP}$
- For any tree search refutation, there is a resolution refutation of same size
- There exist formulas with short resolution refutation but *exponential* tree search refutations

## Pigeon Hole Principle

If  $m > n$  there is no injective mapping of  $m$  objects onto  $n$

$PHP^{m \rightarrow n} :$	$(x_{1,1} \vee x_{1,2} \vee \dots \vee x_{1,n}) \wedge$	Pigeon 1 needs a hole
	$\dots$	
	$(x_{m,1} \vee x_{m,2} \vee \dots \vee x_{m,n}) \wedge$	Pigeon $m$ needs a hole
	$\bigwedge_{1 \leq i < j \leq m} (x_{i,1} \vee x_{j,1}) \wedge$	Hole 1 can contain at most 1 pigeon
	$\dots$	
	$\bigwedge_{1 \leq i < j \leq m} (x_{i,n} \vee x_{j,n})$	Hole $n$ can contain at most 1 pigeon

- Resolution refutations of the pigeon hole principle are exponential
- Using induction, for instance, one can make a linear size refutation