Institute for Formal Models and Verification Johannes Kepler University Linz



VL Logik (LVA-Nr. 342208)

Winter Semester 2014/2015

# Propositional Logic: Evaluating the Formulas

Version 2014.2

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# Satisfiability Checking

### Definition (Satisfiability Problem of Propositional Logic (SAT))

Given a formula  $\phi$ , is there an assignment  $\nu$  such that  $[\phi]_{\nu} = 1$ ?

- oldest NP-complete problem (see next slides)
  - checking a solution (is an assignment satisfying a formula?) is easy (polynomial effort)
  - finding a solution is difficult (probably exponential in the worst case, what is easy compared to satisfiability checking in other logics)
- many practical applications (used in industry)
- efficient SAT solvers (solving tools) are available
- other problems can be translated to SAT:

problem	formulation in propositional logic
$\phi$ is valid	$ eg \phi$ is unsatisfiable
$\phi$ is refutable	to $\neg \phi$ is satisfiable
$\phi \Leftrightarrow \psi$	to $ eg(\phi \leftrightarrow \psi)$ is unsatisfiable
$\phi_1,\ldots,\phi_n\models\psi$	$\phi_1 \wedge \ldots \wedge \phi_n \wedge \neg \psi$ is unsatisfiable

# A Glimpse of Complexity Theory

characterization of computational *hardness* of a problem *Turing Machine*: machine model for abstract "run time" or"memory usage" allows more abstract versions of "run time", "memory usage" the focus is on worst-case *asymptotic* time and space usage

#### Definition

problem is in O(f(n)) iff exists constant *c* and an algorithm which needs  $c \cdot f(n)$  steps (in the worst case on a Turing machine) for an input of size *n* 

- logarithmic  $\mathcal{O}(\log n)$ , e.g. binary search on sorted array of size n
- linear  $\mathcal{O}(n)$ , e.g. linear search in list with *n* elements
- quadratic  $\mathcal{O}(n^2)$ , e.g. generate list of pairs of *n* elements
- exponential  $\mathcal{O}(2^n)$ , e.g. produce all subsets of a set of *n* elements

### Definition

polynomial problems: exists fixed *k* such that worst-case run time is in  $O(n^k)$  the class of polynomial problems is called **P** 

# SAT and the Complexity Class NP

### Definition

A decision problem asks whether an input belongs to a certain class.

**Prime**: decide whether a number given as input is prime. **SAT**: decide whether formula given as input is satisfiable.

Basic idea of **NP** is to use a "guess" and "check" approach, where "guessing" is non-deterministic, e.g. just a "good" choice has to exist.

### Definition

The class NP contains all decision problems which can be decided by a "guessing" and "checking" algorithm in polynomial time in the input size.

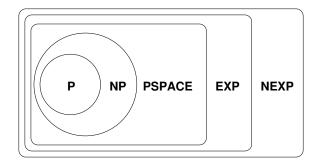
Clearly both Prime and SAT belong to NP.

### Theorem (Cook'71)

Any decision problem in NP can be reduced (encoded) polynomially into SAT.

Actually, **Prime** can also be solved polynomially (rather complicated). More on this topic in the "complexity" course.

## **Complexity Hierarchy**



- P polynomial time
- NP non-deterministic polynomial time
- **PSPACE** polynomial space
  - EXP exponential time
  - NEXP non-deterministic exponential time

except for  $\textbf{P}\neq\textbf{EXP}$  and  $\textbf{NP}\neq\textbf{NEXP}$  nothing is known about strict inclusion

## One Simple Algorithm for Satisfiability Checking

1 Algorithm: evaluate

```
Data: formula \phi
Result: 1 iff \phi is satisfiable
```

```
<sup>2</sup> if \phi contains a variable x then
          pick v \in \{\top, \bot\}
 3
          /* replace x by truth constant v, evaluate resulting formula */
 4
          if evaluate(\phi[x|v]) then return 1;
 5
          else return evaluate(\phi[x|\overline{v}]);
 6
    else
 7
          switch \phi do
 8
                 case \top return 1;
 9
                 case \perp return 0;
10
                 case \neg \psi return ! evaluate(\psi)
                                                                 /* true iff \psi is false */ :
11
                 case \psi' \wedge \psi''
12
                       return evaluate(\psi') && evaluate(\psi'') /* true iff both \psi' and \psi'' are true */
13
                 case \psi' \vee \psi''
14
                       return evaluate(\psi') || evaluate(\psi'') /* true iff <math>\psi' or \psi'' is true */
15
```

## Reasoning with (Propositional) Calculi

- goal: automatically reason about (propositional) formulas, i.e., mechanically show validity/unsatisfiability
- *basic idea:* use syntactical manipulations to prove/refute a formula
- elements of a calculus:
  - axioms: trivial truths/trivial contradictions
  - *rules*: inference of new formulas
- approach: construct a proof/refutation, i.e., apply the rules of the calculus until only axioms are inferred. If this is not possible, then the formula is not valid/unsatisfiable.

#### examples of calculi:

- sequence calculus: shows validity
- resolution calculus: shows unsatisfiability

# Sequent Calculus: Sequents

#### Definition

A sequent is an expression of the form

$$\phi_1,\ldots,\phi_n\vdash\psi$$

where  $\phi_1, \ldots, \phi_n, \psi$  are propositional formulas. The formulas  $\phi_1, \ldots, \phi_n$  are called *assumptions*,  $\psi$  is called *goal*.

#### remarks:

- *intuitively*  $\phi_1, ..., \phi_n \vdash \psi$  means goal  $\psi$  follows from  $\{\phi_1, ..., \phi_n\}$ ■ *special case n* = 0:
  - written as  $\vdash \psi$
  - $\blacksquare$  meaning: we have to prove that  $\psi$  is valid
- *notation*: for sequent φ<sub>1</sub>,..., φ<sub>n</sub> ⊢ ψ, we write K...φ<sub>i</sub> ⊢ ψ if we are only interested in assumption φ<sub>i</sub>
- the assumptions are orderless not ordered

axiom "goal in assumption":

If the goal is among the assumptions, the goal can be proved.

GoalAssum 
$$\overline{\textit{K}\ldots,\psi \vdash \psi}$$

axiom "contradiction in assumptions":

If the assumptions are contradicting, anything can be proved.

$$\overline{\mathbf{K}\ldots,\phi,\neg\phi\ \vdash\ \psi}$$

rule "add valid assumption":

$$\operatorname{ValidAssum} \frac{K \dots, \phi \vdash \psi}{K \dots \vdash \psi}$$
 if  $\phi$  is valid

rules "contradiction":

$$\underbrace{K \dots \neg \psi \vdash \bot}_{K \dots \vdash \psi} \qquad \qquad \underset{P \neg \neg}{K \dots \vdash \neg \phi}$$

rules "elimination of double negation":

$$\mathbb{P}_{\neg_{d}} \frac{K \dots \vdash \psi}{K \dots \vdash \neg \neg \psi} \qquad \qquad \mathbb{A}_{\neg_{d}} \frac{K \dots, \phi \vdash \psi}{K \dots, \neg \neg \phi \vdash \psi}$$

rules "conjunction":

$$\overset{A \land \wedge}{\underbrace{K \ldots, \phi_1, \phi_2 \vdash \psi}} \qquad \overset{P \land \wedge}{\underbrace{K \ldots \vdash \psi_1 \quad K \ldots \vdash \psi_2}}_{K \ldots \vdash \psi_1 \land \psi_2}$$

rules "disjunction":

$$\underset{\mathsf{P}^{-\vee}}{\overset{K\ldots,\neg\psi_{1}}{\vdash} \psi_{2}} \overset{K\ldots,\phi_{1}}{\leftarrow} \psi \overset{K\ldots,\phi_{2}}{\leftarrow} \psi}_{K\ldots,\phi_{1}} \vee \psi_{2} \vdash \psi}$$

Rules for other connectives like implication " $\rightarrow$ " and equivalence " $\leftrightarrow$ " are constructed accordingly.

### Some Remarks on Sequent Calculus

- premises of a rule: sequent(s) above the line
- *conclusion* of a rule: sequent below the line
- axiom: rule without premises
- non-deterministic rule: P-V
- further non-determinism: decision which rule to apply next
- rules with case split: P-A, A-V
- **proof of formula**  $\psi$ 
  - 1. start with  $\vdash \psi$
  - apply rules from bottom to top as long as possible, i.e., for given conclusion, find suitable premise(s)
  - 3. if finally all sequents are axioms then  $\psi$  is valid
  - note: there are many variants of the sequent calculus

### One Algorithm for Calculating with Sequent Calculus

#### 1 Algorithm: entails

**Data**: set of assumptions  $\mathcal{A}$ , formula  $\psi$ **Result**: **1** iff  $\mathcal{A}$  entails  $\psi$ , i.e.,  $\mathcal{A} \models \psi$ 

```
2 if \psi = \neg \neg \psi' then return entails (\mathcal{A}, \psi');
```

- $a \text{ if } \neg \neg \phi \in \mathcal{A} \text{ then return } entails (\mathcal{A} \setminus \{ \neg \neg \phi \} \cup \{ \phi \}, \psi);$
- 4 if  $\phi_1 \land \phi_2 \in \mathcal{A}$  then return *entails*  $(\mathcal{A} \setminus \{\phi_1 \land \phi_2\} \cup \{\phi_1, \phi_2\}, \psi)$ ;
- 5 if ( $\psi \in \mathcal{A}$ ) or ( $\phi, \neg \phi \in \mathcal{A}$ ) then return 1;
- 6 if  $\mathcal{A} \cup \{\psi\}$  contains only literals then return **0**;
- $_{7}\,$  switch  $\psi$  do

8 case  $\perp$ 9 if  $\neg \phi \in \mathcal{A}$  then return *entails*  $(\mathcal{A} \setminus \{\neg \phi\}, \phi)$ ; 10 if  $\phi_1 \lor \phi_2 \in \mathcal{A}$  then 11 if *! entails*  $(\mathcal{A} \setminus \{\neg \phi_1 \lor \phi_2\} \cup \{\phi_1\}, \bot)$  then return 0; 12 else return *entails*  $(\mathcal{A} \setminus \{\neg \phi_1 \lor \phi_2\} \cup \{\phi_2\}, \bot)$ ; 13 case *x* where *x* is a variable return *entails*  $(\mathcal{A} \cup \{\neg \phi\}, \bot)$ ; 14 case  $\neg \psi'$  return *entails*  $(\mathcal{A} \cup \{\psi'\}, \bot)$ ;

- 15 case  $\psi_1 \vee \psi_2$  return *entails*  $(\mathcal{A} \cup \{\neg \psi_1\}, \psi_2)$ ;
- 16 **Case**  $\psi_1 \wedge \psi_2$  return *entails*  $(\mathcal{A}, \psi_1)$  && *entails*  $(\mathcal{A}, \psi_2)$ ;

### Proving XOR stronger than OR with the Sequent Calculus

proof direction

### Refuting XOR stronger than AND with the Sequent Calculus

*counter example to validness:*  $a = \bot, b = \top$ 

### Soundness and Completeness

For any calculus important properties are, first *soundness*, i.e. the question "Can only valid formulas be shown as valid?" and second *completeness*, i.e. the question "Is there a proof for every valid formula?".

#### Soundness

If a formula is shown to be valid in the Gentzen Calculus, then it is valid.

*Proof sketch:* consider each rule individually and show that from valid premises only valid conclusions can be drawn.

### Completeness

Every valid formula can be proven to be valid in the Gentzen Calculus.

*Proof sketch:* Show that the algorithm terminates and that there is at least one case where it returns false if the formula is not valid.

# Proving Formulas in Normal Form

- In practice, formulas of arbitrary structure are quite challenging to handle
  - tree structure
  - simplifications affect only subtrees
- We have seen that CNF and DNF are able to represent every formula
  - so why not use them as input for SAT?

#### Conjunctive Normal Form

- refutability is easy to show
- CNF can be efficiently calculated (polynomial)

#### Disjunctive Normal Form

- satisfiability is easy to show
- complexity is in getting the DNF
- CNF and DNF can be obtained from the truth tables
  - exponential many assignments have to be considered
- alternative approach
  - structural rewritings which are (satisfiability) equivalence preserving

### Transformation to Normal Form

1. Remove 
$$\leftrightarrow$$
,  $\rightarrow$ ,  $\oplus$  as follows:  
 $\phi \leftrightarrow \psi \Leftrightarrow (\phi \rightarrow \psi) \land (\psi \rightarrow \phi), \phi \rightarrow \psi \Leftrightarrow \neg \phi \lor \psi,$   
 $\phi \oplus \psi \Leftrightarrow (\phi \lor \psi) \land (\neg \phi \lor \neg \psi)$ 

- Transform formula to negation normal form (NNF) by application of laws of De Morgan and elimination of double negation
- 3. Transform formula to CNF (DNF) by laws of distributivity

#### Example

Transform  $\neg(a \leftrightarrow b) \rightarrow (\neg(c \land d) \land e)$  to an equivalent formula in CNF.

- 1. a) remove equivalences:  $\Leftrightarrow \neg((a \rightarrow b) \land (b \rightarrow a)) \rightarrow (\neg(c \land d) \land e)$ b) remove implications:  $\Leftrightarrow \neg \neg((\neg a \lor b) \land (\neg b \lor a)) \lor (\neg(c \land d) \land e)$
- 2. NNF:  $\Leftrightarrow$   $((\neg a \lor b) \land (\neg b \lor a)) \lor ((\neg c \lor \neg d) \land e)$
- 3.  $\Leftrightarrow ((\neg a \lor b) \lor ((\neg c \lor \neg d) \land e))) \land ((\neg b \lor a) \lor ((\neg c \lor \neg d) \land e))) \\ \Leftrightarrow (\neg a \lor b \lor \neg c \lor \neg d) \land (\neg a \lor b \lor e) \land (\neg b \lor a \lor \neg c \lor \neg d) \land (\neg b \lor a \lor e)$

### Some Remarks on Normal Forms

- The presented transformation to CNF/DNF is exponential in the worst case (e.g., transform  $(a_1 \land b_1) \lor (a_2 \land b_2) \lor \cdots \lor (a_n \land b_n)$  to CNF).
- For DNF transformation, there is probably no better algorithm.
- For CNF transformation, there are polynomial algorithms.
  - Basic idea: introduce labels for subformulas.
  - Also works for formulas with sharing (circuits).
  - Also known as "Tseitin Encoding".
- CNF is usually not easier to solve, but easier to handle:
  - compact data structures: a CNF is simply a list of lists of literals.
- CNF very popular in practice: standard input format DIMACS
- For solving satisfiability of CNF formulas, there are many optimization techniques as well as dedicated algorithms.

### Resolution

the *resolution calculus* consists of the single resolution rule

$$\frac{x \lor C \qquad \neg x \lor D}{C \lor D}$$

- *C* and *D* are (possibly empty) clauses
- the clause  $C \lor D$  is called *resolvent*
- variable x is called pivot
- usually antecedent clauses  $x \lor C$  and  $\neg x \lor D$  are assumed not to be tautological, that is  $x \notin C$  and  $x \notin D$ .
- in other words:

 $(\neg x \rightarrow C), (x \rightarrow D) \models C \lor D$ 

- resolution is *sound* and *complete*.
- the resolution calculus works only on formulas in CNF
- if the empty clause can be derived then the formula is *unsatisfiable*
- if no new clause can be generated by application of the resolution rule then the formula is *satisfiable*

#### Example

one application of resolution

$$\frac{x \lor y \lor \neg z}{y \lor \neg z \lor y' \lor \neg z}$$

derivation of empty clause:

derivation of tautology:

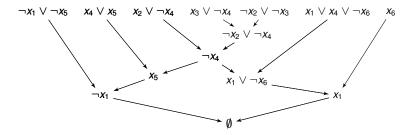
 $\frac{x \lor a \qquad \neg x \lor \neg a}{a \lor \neg a}$ 

### **Resolution Example**

#### We prove unsatisfiability of

$$\{(\neg x_1 \lor \neg x_5), (x_4 \lor x_5), (x_2 \lor \neg x_4), (x_3 \lor \neg x_4), (\neg x_2 \lor \neg x_3), (x_1 \lor x_4 \lor \neg x_6), (x_6)\}$$

as follows:



### **DPLL** Overview

The DPLL algorithm is ...

- old (invented 1962)
- easy (basic pseudo-code is less than 10 lines)
- popular (well investigated; also theoretical properties)
- usually realized for formulas in CNF
- using binary constraint propagation (BCP)
- in its modern form as *conflict drive clause learning (CDCL)* basis for state-of-the-art SAT solvers

# Binary Constraint Propagation (BCP)

### Definition (Binary Constraint Propagation (BCP))

Let  $\phi$  be a formula in CNF containing a unit clause *C*, i.e.,  $\phi$  has a clause

- ${\it C}=({\it I})$  which consists only of literal  ${\it I}.$  Then  ${\it BCP}(\phi,{\it I})$  is obtained from  $\phi$  by
  - removing all clauses with I
  - removing all occurrences of 1
  - one application of BCP can trigger other applications of BCP
  - BCP( $\phi$ ) denotes all possible applications of BCP( $\phi$ , I) until fixpoint
  - if  $BCP(\phi)$  produces the empty clause, then the formula  $\phi$  is unsatisfiable
  - if  $BCP(\phi)$  produces the empty CNF, then the formula  $\phi$  is satisfiable

#### Example

$$\phi = \{ (\neg a \lor b \lor \neg c), (a \lor b), (\neg a \lor \neg b), (a) \}$$

1. 
$$\phi' = BCP(\phi, a) = \{(b \lor \neg c), (\neg b)\}$$

2. 
$$\phi'' = BCP(\phi', \neg b) = \{(\neg c)\}$$

3. 
$$\phi'' = BCP(\phi', c) = \{\} = \top$$

# **DPLL Algorithm**

```
1 Algorithm: evaluate
    Data: formula \phi in CNF
    Result: 1 iff \phi satisfiable
 2 while 1 do
           \phi = \mathsf{BCP}(\phi)
 3
           if \phi == \top then return 1;
 4
           if \phi == \bot then
 5
                  if stack.isEmpty() then return 0;
 6
                  (I, \phi) = \text{stack.pop}(I)
 7
                  \phi = \phi \wedge I
 8
           else
 9
                  select literal / occurring in \phi
10
                  stack.push(\overline{l}, \phi)
11
                  \phi = \phi \wedge I
12
```

### Some Remarks on DPLL

- DPLL is the basis for most state-of-the-art SAT solvers
  - like Lingeling http://fmv.jku.at/lingeling
  - simpler or more established solvers: MiniSAT, PicoSAT, Cleaneling, ...
- DPLL alone is not enough powerful optimizations required for efficiency:
  - learning and non-chronological back-tracking (CDCL)
  - reset strategies and phase-saving
  - compact lazy data-structures
  - variable selection heuristics
  - usually combined with preprocessing before search
  - and inprocessing algorithms interleaved with search
- variants of DPLL are also used for other logics:
  - quantified propositional logic (QBF)
  - satisfiability modulo theories (SMT)
- challenge to parallelize
  - some succesfull attempts: ManySAT, Plingeling, Penelope, Treengeling, ...