

Automated Reasoning and Formal Verification

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intro

Slides will be on his webpage along with the recordings.

The exam will consist of a script and an oral exam on the topics of the whole course.

boolean/propositional logic

A propositional **formula** can be:

- \top, \perp
- Propositional **atoms** A_1, A_2, \dots, A_n
- A combination of other formulas. If φ_1 and ϕ_2 are formulas, so are:
 - $\neg\varphi_1$
 - $\varphi_1 \wedge \phi_2$
 - $\varphi_1 \vee \phi_2$
 - $\varphi_1 \rightarrow \phi_2$
 - $\varphi_1 \leftarrow \phi_2$
 - $\varphi_1 \leftrightarrow \phi_2$
 - $\varphi_1 \oplus \phi_2$

We define a function $Atoms(\varphi)$ representing the set $\{A_1, \dots, A_n\}$ of atoms in ϕ

A **clause** is a disjunction of literals $\bigvee_j l_j$ or $(A_1 \vee \neg A_2 \vee \dots)$

A **cube** is a conjunction of literals $\bigwedge_j l_j$ or $(A_1 \wedge \neg A_2 \wedge \dots)$

trees and DAGS

A tree is a natural representation of an expression, but in the worst cases it can grow exponentially. The same information about the formula can be conveyed by a *Directed Acyclic Graph*, which can grow linearly in size.

Total Truth Assignment

They can also be abbreviated as *Total Assignment*.

A total truth assignment $\mu : Atoms(\varphi) \mapsto \{\top, \perp\}$ represents *one* possible state of the formula.

Partial Truth Assignment

A partial truth assignment $\mu : \mathcal{A} \mapsto \{\top, \perp\}, \mathcal{A} \subset Atoms(\varphi)$ represents 2^k total assignments, where k is the number of unassigned literals.

μ defined for total and partial truth assignments can be seen as a set of literals (positive and negative ones) or a formula.

Set of models

$M(\varphi) \triangleq \{\mu \mid \mu \models \phi\}$ is the set of all models of ϕ .

Properties

- φ is *valid* if every μ models ϕ

- φ valid $\iff \neg\phi$ unsatisfiable

- $\alpha \models \beta \iff \alpha \rightarrow \beta$ valid

Deduction
theorem

- $\alpha \models \beta \iff \alpha \wedge \neg\beta$ not satisfiable

corollary

Equivalence and Equi-satisfiability

α and β are *equivalent* if $\forall \mu. \mu \models \alpha \iff \mu \models \beta$.

In other terms, $M(\alpha) = M(\beta)$.

Equi-satisfiability $M(\alpha) \neq \emptyset \iff M(\beta) \neq \emptyset$. This property is mostly used when applying transformations to formulas $\beta \triangleq T(\alpha)$.

Transformations can be *validity preserving* if they preserve the validity of the formula they're being applied to, or *satisfiability preserving* if they preserve its satisfiability.

Shannon's expansion

$$\exists v. \varphi := \phi|v = \perp \vee \phi|v = \top$$

The existential is a disjunction between two possible formulas. One where v is set to true, and one where it is set to false.

$$\forall v. \varphi := \phi|v = \perp \wedge \phi|v = \top$$

The universal one is similar, with a conjunction between the two.

Polarity of subformulas

Polarity is a metric defined for each subformula of a formula φ that tells us under how many nested negations it occurs. It can either be positive, negative, or both in some cases.

The recursive rules to determine the polarity are shown in the image below

- φ occurs positively in φ ;
- if $\neg\varphi_1$ occurs positively [negatively] in φ , then φ_1 occurs negatively [positively] in φ
- if $\varphi_1 \wedge \varphi_2$ or $\varphi_1 \vee \varphi_2$ occur positively [negatively] in φ , then φ_1 and φ_2 occur positively [negatively] in φ ;
- if $\varphi_1 \rightarrow \varphi_2$ occurs positively [negatively] in φ , then φ_1 occurs negatively [positively] in φ and φ_2 occurs positively [negatively] in φ ;
- if $\varphi_1 \leftrightarrow \varphi_2$ or $\varphi_1 \oplus \varphi_2$ occurs in φ , then φ_1 and φ_2 occur positively and negatively in φ ;

If we assume $\top = 1, \perp = 0$ we can also see the polarity of a subformula as "how much it contributes to the overall value of the formula".

2 Normal forms

2.1 Negative Normal Form - NNF

A negative normal form is a formula in which each negations has been pushed down to the atoms. This implies that every subformula in $NNF(\varphi)$ has positive polarity.

Properties

- Every formula can be made into negative normal form
- NNF transformation preserves equivalence

2.2 Conjunctive Normal Form - CNF

$$\bigvee_{i=1}^L \bigwedge_{j=1}^{K_i} l_{ij}$$
$$(l_{11} \wedge l_{12}) \vee (l_{21} \wedge l_{22} \wedge l_{23}) \vee (\dots) \vee \dots$$

Every formula can be converted in *Conjunctive Normal Form*, but there are different ways to do so.

2.2.1 Naive CNF conversion

The more intuitive and straightforward method consist of:

1. Expanding implications and equivalences
2. Pushing down negations like in NNF
3. Recursively applying DeMorgan's rule to get the CNF shape

This method produces a CNF that is equivalent to the original formula and has the same atoms. It is however rarely used in practical applications because it can be up to exponentially larger than the original formula.

2.2.2 Labeling CNF conversion

This is a more efficient *bottom-up* approach, which can be executed while parsing the expression. The main idea is that of introducing new variables that serve as "*labels*" for each subformula. The smaller formulas can be converted to CNF with the naive approach, and then assembled through the labels.

This method introduces new atoms, but $\exists(B_1, \dots, B_k).CNF_l(\varphi)$ equiv ϕ where B_1, \dots, B_k are the newly introduced variables. This means that ϕ and $CNF_l(\phi)$ are equisatisfiable.

The representation obtained from the CNF_l can be reduced further in size by using polarization to change some implications around.

3 Basic SAT-solving techniques

Example: A classic problem is that of checking a query under a (usually much larger) knowledge base. This problem can be reduced to SAT. $KB \models \alpha$ or $M(KB) \subseteq M(\alpha)$

$$KB \models \alpha \iff SAT(KB \vee \neg \alpha) = false$$

3.1 Intro - Unit propagation

Resolution rule Deduction of a new clause from a pair of clauses with exactly one incompatible variable (which is called the "resolvent").

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

Removal of valid clauses If a clause is valid (always true) it can be removed from the formula.

Clause subsumption If a clause appears on its own and inside another clause, we can remove the second, bigger, clause.

Unit resolution Having a clause composed of a single literal forces said literal to be true. This means we can remove all instances of the negated literal.

Unit subsumption Like clause subsumption but with a literal instead of a clause

These unit propagation rules can happen in a chain. After modifying the formula once we can create new unary clauses for example.

3.2 Resolution algorithm

Algorithm 1: Resolution algorithm

```

Assume input is in CNF;
 $\varphi$  is a set of clauses;
//Search for a refutation of  $\varphi$ ;
repeat
  | apply resolution rule to pairs of clauses;
until a false clause is generated  $\vee$  the rule is not applicable ;

```

This algorithm is correct and complete, but operates in exponential memory and is time inefficient.

3.3 Tableaux

Search assignments satisfying φ by applying *elimination rules* on its connectors.
Try to be clever and put put smaller clauses first in the branching order.

4 LAB 1

4.1 DIMACS

Dimacs is the standard format for representing SAT problems.

File structure

- Comments start with 'c'
- header of the form 'p cnf 7 8'
 - 7: number of variables
 - 8: number of clauses
- 1 clause for line of form "1 -2 0"
 - trailing 0 is a constant
 - 1 first variable
 - -2 second variable, negated

command `mathsat -input=dimacs -model file.cnf`

`-model` tells it to provide a model, otherwise it gives a yes/no answer