Automated Reasoning

- A Comprehensive Collection -

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Introduction

Welcome to the seminar on Automated Reasoning. This document is a compilation of various seminar materials designed to provide a comprehensive overview of the field. Here, we explore the fundamental concepts, methodologies, and applications of automated reasoning in computer science and logic.

This is an example of referencing Section

Chapter 1

Preliminaries

Truth Function

Definition 1.1. Let $\mathbb{B} = \{\mathsf{T}, \mathsf{F}\}$ be the *boolean domain*. Let $k \in \mathbb{N}$. A mapping

$$f: \mathbb{B}^k \to \mathbb{B}$$

is called a truth function.

Remark 1.1 (Truth Functions of Connectives). The logical connectives are assumed to be **truth-functional**. Hence, they are represented by certain **truth functions**.

Logical Negation The *logical not connective* defines the truth function $f \neg$ as follows:

$$f^{\neg}(\mathsf{F}) = \mathsf{T}$$

 $f^{\neg}(\mathsf{T}) = \mathsf{F}$

Logical Conjunction The *conjunction connective* defines the truth function f^{\wedge} as follows:

$$f^{\wedge}(T,T) = T$$

 $f^{\wedge}(T,F) = F$
 $f^{\wedge}(F,T) = F$
 $f^{\wedge}(F,F) = F$

Logical Disjunction The *disjunction connective* defines the truth function f^{\vee} as follows:

$$f^{\vee}(T,T) = T$$

 $f^{\vee}(T,F) = T$
 $f^{\vee}(F,T) = T$
 $f^{\vee}(F,F) = F$

Count of Truth Functions

Proposition 1.1. There are $2^{(2^k)}$ distinct truth functions on k variables.

Proof. Let $f : \mathbb{B}^k \to \mathbb{B}$ be a truth function for $k \in \mathbb{N}$. Then

(Cardinality of Cartesian Product of Finite Sets)

$$\#(\mathbb{B}^{k}) = \#(\mathbb{B} \times \cdots \times \mathbb{B})$$

$$= \#\mathbb{B} \#\mathbb{B} \cdots \#\mathbb{B}$$

$$= k \text{ times}$$

$$= 2 \cdot 2 \cdots 2$$

$$= 2^{k}.$$

(Cardinality of Set of All Mappings.)

$$\#(T^S) := \{ f \subseteq S \times T : f \text{ is a mapping} \} = (\#T)^{(\#S)} \implies \#(\mathbb{B}^{(\mathbb{B}^k)}) = 2^{(2^k)} \}$$

Unary Truth Functions

Corollary 1.1.1. *There are 4 distinct unary truth functions:*

- *The constant function* f(p) = F
- The constant function f(p) = T
- The identity function f(p) = p
- The logical not function $f(p) = \neg p$

Proof. From Count of Truth Functions there are $2^{(2^1)} = 4$ distinct truth functions on 1 variable. These can be depicted in a truth table as follows:

 \circ_1 : Whether p = T or p = F, $\circ_1(p) = T$. Thus \circ_1 is the constant function $\circ_1(p) = T$.

 \circ_2 : We have

(1)
$$p = T \implies \circ_2(p) = T$$

(2)
$$p = F \implies \circ_2(p) = F$$

Thus \circ_2 is the *identity function* $\circ_2(p) = p$.

∘₃: We have

(1)
$$p = T \implies \circ_3(p) = F$$

(2)
$$p = F \implies \circ_3(p) = T$$

Thus \circ_3 is the *logical not function* $\circ_3(p) = \neg p$.

 \circ_4 : Whether p = T or p = F, $\circ_4(p) = F$. Thus \circ_1 is the constant function $\circ_4(p) = F$.

Binary Truth Functions

Corollary 1.1.2. *There are* 16 *distinct unary truth functions:*

• Two constant operations:

-
$$f_{\mathsf{T}}(p,q) = \mathsf{T}$$

$$-f_{\mathsf{F}}(p,q)=\mathsf{F}$$

• Two projections:

-
$$Proj_1(p,q) = p$$

-
$$Proj_2(p,q) = q$$

• Two negated projections:

$$- \overline{\mathsf{Proj}_1}(p,q) = \neg p$$

$$- \overline{\mathsf{Proj}_2}(p,q) = \neg q$$

- *The conjunction:* $p \land q$
- *The disjunction:* $p \lor q$
- Two conditionals:

$$-p \implies q$$

$$-q \implies p$$

- The biconditional (iff): $p \iff q$
- The exclusive or (xor): $\neg(p \iff q)$
- Two negated conditionals:

$$-\neg(p \implies q)$$

$$-\neg(q \implies p)$$

- The NAND $p \uparrow q$
- The NOR $p \downarrow q$

Proof. From Count of Truth Functions there are $2^{(2^2)} = 16$ distinct truth functions on 2 variable. These can be depicted in a truth table as follows:

p	T	Т	F	F
q	Т	F	Τ	F
$f_{F}(p,q)$	F	F	F	F
$p \downarrow q$	F	F	F	Τ
$\neg (p \iff q)$	F	F	Τ	F
$\overline{Proj_1}(p,q)$	F	F	Τ	Τ
$\neg(p \implies q)$	F	Τ	F	F
$\overline{Proj_2}(p,q)$	F	Τ	F	Τ
$\neg (p \iff q)$	F	Т	Т	F
$p \uparrow q$	F	Τ	Τ	Τ
$p \wedge q$	Т	F	F	F
$p \iff q$	Т	F	F	Τ
$Proj_2(p,q)$	Т	F	Τ	F
$p \implies q$	Т	F	Τ	Τ
$Proj_1(p,q)$	Т	Τ	F	F
$p \longleftarrow q$	Т	Τ	F	Τ
$p \vee q$	Т	Τ	Τ	F
$f_{T}(p,q)$	Т	Т	Т	Т

Chapter 2

Conflict Driven Clause Learning (CDCL)

Propositional Function (Formula)

Definition 2.1. A **propositional function** (or **formula**) is defined inductively as follows:

(Basic Step) Any propositional variable $x \in V$ is a formula.

(Inductive Step) If *F* and *G* are formulas, then the following are also formulas:

- $-\neg F$
- $-F \wedge G$
- $-F\vee G$
- $-F \rightarrow G$
- $F \leftarrow G$

Formally, the set of all Φ is the smallest set satisfying:

$$\Phi = V \cup \{\neg F : F \in \Phi\} \cup \{\circ(F, G) : F, G \in \Phi \text{ and } \circ \in \{\land, \lor, \rightarrow, \leftrightarrow\}\}$$

Boolean Satisfiability Problem (SAT)

Definition 2.2. Let $X = \{x_1, x_2, \dots, x_n\}$ be a set of propositional variables. Let L be a set of one or more propositional formulas constructed using only:

- $x_i \in X$ for $i = 1, \ldots, n$;
- the $2^{(2^1)} = 4$ unary logical connectives;
- the $2^{(2^2)} = 16$ binary logical connectives;

The problem is to find truth values for all $x \in X$ such that all the formulas in L are true. Such a problem is a boolean satisfiability problem.

I present the CDCL algorithm and its implementation based on existing literature. This algorithm is used to solve SAT problems efficiently...

2.0.1 SAT solving basics

The CDCL algorithm is a mix of two older approaches to SAT solving: DPLL and Resolution...

Backtracking and unit propagation as in DPLL solvers

When we provide a SAT instance to a DPLL solver it builds up a search tree of assignments...

Resolution

If a formula *F* contains the clauses $\{\neg x\} \cup A$ and $\{x\} \cup A$...

2.1 Principles of CDCL

Now that we have seen how Backtracking and Resolution work we are ready to merge these approaches...

2.1.1 The trail

When applying CDCL rather than exploring a search tree of assignments...

2.1.2 Conflict clauses and backjumping

Consider our previous example again. When we want to continue building up our trail...

2.1.3 The implication graph

A nice way to illustrate the functionality of CDCL are implication graphs...

2.1.4 The algorithm

To get a clearer view Algorithm 4.1 shows the pseudocode for the CDCL algorithm...

2.2 Implementation

Let us now look at how the algorithm is implemented in real life...

2.2.1 Clauses

We use a monolithic array MEM to hold the original formula's clauses as well as the newly learned clauses...

2.2.2 Literals

Assume the variables are x_1, x_2, \ldots, x_n . We represent x_k by k...

2.3 Results

It is important to say that CDCL is a sound and complete algorithm for the propositional satisfiability problem...

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