# **Automated Reasoning**

- A Comprehensive Collection -

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# A document presented for the Automated Reasoning

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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**

# **Conflict Driven Clause Learning (CDCL)**

### 1.1 The Boolean Satisfiability Problem

#### Propositional Variable

**Definition 1.1.** A **propositional variable** is an input variable (that can either be true or false) of a truth function. Propositional variables are the *basic building-blocks* of propositional formulas, used in propositional logic and higher-order logics.

**Example 1.1.** For a statement variable, a lowercase letter is usually used, for example: p, q, r, . . . , and so on or lowercase Greek letters, for example:  $\phi$ ,  $\psi$ ,  $\chi$  and so on.

**Remark 1.1.** The citing of a propositional variable can be interpreted as an assertion that the proposition represented by that symbol is true. That is:

"p" means "p is true".

#### **Propositional Function (Formula)**

**Definition 1.2.** A **propositional function** (or **formula**)  $P(x_1, x_2, ...)$  is an operation which acts on the objects denoted by the object variables (here, propositional variables)  $x_1, x_2, ...$  in a particular universe to return a truth value which depends on:

- (1) The values of  $x_1, x_2, \ldots$
- (2) The nature of P.

The boolean satisfiability problem (SAT) is the following: given a formula F on propositional variables, does there exists an assignment  $\mathcal{A}$  on theses variables, such that  $\mathcal{A}(F) = 1$ .

Given a formula F over a set of propositional variables  $\{x_1, x_2, \dots, x_n\}$ ,

$$\exists \mathcal{A}: \{x_1, x_2, \dots, x_n\} \rightarrow \{0, 1\}: \mathcal{A}(F) = 1.$$

### Boolean Satisfiability Problem (SAT)

**Definition 1.3.** 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:

- $x_i \in X \text{ for } i = 1, ..., n;$
- •

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

#### 1.1.1 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  $\{x\} \cup A$ ...

### 1.2 Principles of CDCL

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

#### 1.2.1 The trail

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

### 1.2.2 Conflict clauses and backjumping

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

### 1.2.3 The implication graph

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

#### 1.2.4 The algorithm

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

## 1.3 Implementation

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

#### 1.3.1 Clauses

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

#### 1.3.2 Literals

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

# 1.4 Results

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