



FACULTY OF COMPUTER SCIENCE

INSTITUTE OF THEORETICAL COMPUTER SCIENCE

CHAIR OF AUTOMATA THEORY

PROF. DR.-ING. FRANZ BAADER

MASTER'S THESIS

# A Tableau Algorithm for the Numerical Description Logic $\mathcal{ALCSCC}$

Ryny Khy

**Matriculation Number:** 4751049

born on 30. November 1994 in Landshut

supervised by  
Dr.-Ing. Stefan Borgwardt  
Filippo De Bortoli M.Sc.

reviewed by  
Dr.-Ing. Stefan Borgwardt  
Prof. Sebastian Rudolph

December 5, 2020



# Declaration of Authorship

**Author:** Ryny Khy

**Matriculation Number:** 4751049

**Title:** A Tableau Algorithm for the Numerical Description Logic  $\mathcal{ALCSCC}$

I hereby declare that I have written this final thesis independently and have listed all used sources and aids. I am submitting this thesis for the first time as a piece of assessed academic work. I understand that attempted deceit will result in the failing grade “not sufficient” (5.0).

---

Place and Date

---

Author's signature



### **Abstract**

In the research field of Description Logics (DLs) checking satisfiability of  $\mathcal{ALCQ}$  has been investigated thoroughly and is therefore well-known. The DL  $\mathcal{ALCSCC}$  extend  $\mathcal{ALCQ}$  with constraints over role successor using quantifier-free fragment (QF) of Boolean Algebra (BA) and Presburger Arithmetic (PA). Checking satisfiability of this DL has been proven to be decidable and PSpace-complete. We provide in this work a tableau algorithm for checking satisfiability of  $\mathcal{ALCSCC}$  and its correctness proof.



# Contents

<b>1</b>	<b>Introduction</b>	<b>9</b>
1.1	Motivation . . . . .	9
1.2	Related Works . . . . .	11
<b>2</b>	<b>Preliminaries</b>	<b>13</b>
2.1	QFBAPA . . . . .	13
2.2	$\mathcal{ALCSCC}$ . . . . .	15
<b>3</b>	<b>Tableau for <math>\mathcal{ALCSCC}</math></b>	<b>19</b>
3.1	Transforming an ABox into a formula . . . . .	20
3.2	Solution of a formula . . . . .	21
3.3	The Tableau Algorithm . . . . .	24
<b>4</b>	<b>Correctness</b>	<b>27</b>
4.1	Termination . . . . .	27
4.2	Soundness and Completeness . . . . .	30
<b>5</b>	<b>Conclusion</b>	<b>35</b>





# Chapter 1

## Introduction

### 1.1 Motivation

In traditional databases stored data objects do not have any relation with each other unless explicitly stated. However we can extract additional information about these objects if we use database systems which employ *semantics*. Imagine we want to add data entries for two people (Anna and Beth) to a traditional database. Anna is a teacher at the local school. Beth is a student of her class. In our database we save their names, the class that Beth attends as well as the class that Anna teaches. As we do not explicitly encode their student-teacher relationship the traditional database does not know about it. If we use an ontology-based system we can deduce this information by making use of semantics. These semantics are described by a set of rules (axioms). In our example one axiom would be that if a teacher teaches a class and students attend the same class, then there is a student-teacher relationship between the teacher and these students. By applying this axiom the ontology-based database can automatically deduce that Anna is the teacher of Beth. Popular use-cases for ontology-based systems are databases for biological and medical research [4]. As an example ontologies can be used to automatically fill in missing information about patients which are helpful in diagnostics. Another major use-case for ontologies is the *Semantic Web*, which is an extension of the World Wide Web with standards given by the World Wide Web Consortium (W3C)<sup>1</sup>. These standards allow a more effective way of combining information from different sources.

An Ontology (in the field of computer science) can be viewed as formal representation of a certain domain of interest. The relationships between entities in an ontology-based database are formulated by a fragment of first-order logic (FOL). This fragment of FOL is called *Description Logic (DL)* and is a family of knowledge representation systems. DLs mainly consist of concepts, which correspond to unary relations in FOL, and relations between the concepts, which

---

<sup>1</sup><https://www.w3.org/standards/semanticweb/>, last accessed on December 5, 2020

correspond to binary relations in FOL. To create more complex (compound) concepts we can combine concepts by using operators like  $\sqcap$ ,  $\sqcup$ ,  $\sqsubseteq$ ,  $\exists$  and  $\forall$ . For example the statement "All Humans who have children are parents" can be formalized in DL as  $Human \sqcap \exists hasChild. \top \sqsubseteq Parent$ , where *Human* and *Parent* are concept names and *hasChild* is a role name. This statement can also be formalized with a numerical restriction:  $Human \sqcap \geq 1 hasChild. \top \sqsubseteq Parent$ . A knowledge base consists of a *TBox*, which contains the axioms, and an *ABox*, which contains assertions about certain individual names (objects).

The process of determining whether some statement can be concluded from a set of information is called *reasoning*. Reasoning can be done by adding this statement in negated form (as an axiom or assertion) to the set of information (TBox or ABox) and then checking whether the updated knowledge base is now *unsatisfiable*. If it is unsatisfiable the statement can be concluded from the information set. Being able to check the satisfiability of DL statements is therefore a valuable tool to conduct reasoning in ontology systems. The DL *ALCQ* [5][9] has been investigated thoroughly and therefore we know a lot about its satisfiability. This DL allows conjunctions ( $\sqcap$ ), disjunctions ( $\sqcup$ ), negations ( $\neg$ ) and number restrictions ( $\leq nr C$  and  $\geq nr C$ , where  $n$  is a number,  $r$  a role name and  $C$  a concept name). In [5] Hollunder and Baader proved that checking satisfiability of a *ALCQ* concept without a TBox is in PSpace and otherwise in ExpTime.

The DL *ALCSCC* [1] extends *ALCQ* with *set constraints* and *cardinality constraints* over role successors, which use the logic of *QFBAPA* (quantifier-free fragment of Boolean Algebra with Presburger Arithmetic)[7]. Instead of the quantifiers  $\exists$  and  $\forall$  we use set expressions (Boolean Algebra) and numerical constraints (Presburger Arithmetic). For example  $Human \sqcap \geq 1 hasChild. \top \sqsubseteq Parent$  is written in *ALCSCC* as  $Human \sqcap succ(|hasChild| \geq 1) \sqsubseteq Parent$ . This DL is more expressive than *ALCQ* because every quantified restriction of the form  $\leq nr.C$  or  $\geq nr.C$  can be written in *ALCSCC* as  $succ(|r \sqcap C| \leq 1)$  or  $succ(|r \sqcap C| \geq 1)$  respectively. However a constraint like  $succ(|r| = |s|)$  can not be formulated in *ALCQ* [1]. Because of this extension checking satisfiability over *ALCSCC* becomes more complicated. Nevertheless in [1] Baader has shown that the satisfiability problem for *ALCSCC* is still PSpace-complete.

In this work we present a tableau algorithm for *ALCSCC*. While a tableau algorithm leads to a runtime complexity worse than PSpace-complete the benefit of a tableau algorithm is that we gain a satisfied interpretation (a correct assignment without contradiction) of the concepts, which is also called *witness*. A tableau algorithm consists of completion rules which are applied to the assertions of the ABox. By applying these rules new assertions that can be derived from the original assertions are added to the ABox. If we can no longer apply any rules and the ABox contains a contradiction, then the ABox is unsatisfiable. Otherwise it is satisfiable. The main difficulty in creating the completion rules for *ALCSCC* is that unlike in *ALCQ*, the number of successor is not bounded. By adding role successors the cardinalities in a constraint can vary. For example if we have a constraint  $succ(|r| = |s|)$  then the bound for the number of  $s$ -successors is equal to the number of  $r$ -successors we already have. During a

tableau algorithm we can add, merge or replace  $r$ -successors which changes the bound for the number of  $s$ -successors. To deal with the changing cardinalities of  $\mathcal{ALCSCC}$  we transform the assertions with cardinalities to a QFBAPA formula and use a QFBAPA solver to determine whether the formula is satisfiable or not. If the formula is not satisfiable, there must be a contradiction. If the solver returns a solution, we add new assertions accordingly. Since we use a solver that is capable of returning every possible solution, there can be an infinite number of solutions. We can show that we can shorten each of these solutions to a bound number of role successors without losing any information.

## 1.2 Related Works

The tableau algorithm is a popular tool to solve the satisfiability problem for description logics. Hence a lot of research has been done trying to formulate tableau algorithms for different description logic languages. In [5] Hollunder and Baader present a tableau algorithm for checking the satisfiability of an ABox in the DL  $\mathcal{ALCQ}$ . This satisfiability problem is PSpace-hard. The presented algorithm consists of five rules which can be applied to the ABox in non-deterministic order. Two of these rules are decomposing rules for  $\sqcap$  and  $\sqcup$ . Furthermore there is a rule that decides whether a successor of an individual name also contributes in any other numerical assertion, which helps to determine the exact number of role successors. The last two rules add and replace individual names to the ABox according to the numerical restrictions. This can lead to an endless loop of adding and replacing individual names. To avoid endless loops the author introduces a concept of *safeness*, which has a similar purpose as blocking in other tableau algorithms: Individual names can only be replaced if they fulfill the safeness criteria. In [10] Tobies presented an optimized tableau algorithm for  $\mathcal{ALCQ}$ , which runs in PSpace. This optimization is achieved by saving an integer which denotes the number of successors already introduced to satisfy a restriction  $\geq nr.C$ , instead of keeping all  $n$  possible successors.

In [6] Horrocks et al. published tableau algorithms for  $\mathcal{SHIF}$  concepts and  $\mathcal{SI}$  concepts. The DL  $\mathcal{SI}$  extends the DL  $\mathcal{ALC}$  with transitive and inverse role names. The DL  $\mathcal{SHIF}$  further extends  $\mathcal{SI}$  with role hierarchy and functional restriction. For both DLs the tableau algorithm has to have a blocking technique to avoid infinite chains of introducing elements with the same properties which can be caused by transitive or inverse roles. For  $\mathcal{SI}$  the tableau algorithm does not only look at the successors but also the predecessors of the considered individual names when dealing with a  $\forall$ -assertion. In case of  $\exists$ -assertions the algorithm first determines whether the considered individual name  $x$  is blocked or not. It is blocked if an ancestor (a non-direct predecessor) is blocked or if an ancestor has "similar" assertions as  $x$ . This algorithm runs in PSpace. For  $\mathcal{SHIF}$  the tableau algorithm has to ensure that the considered individual name  $x$  is not pair-wise blocked for any rule. Being pair-wise blocked means that for a predecessor  $y$  of  $x$  there are two ancestors of  $x$ , such that they behave "similar" to  $y$  and  $x$ . In [3] a DL called  $\mathcal{SHQ}$  (also known  $\mathcal{ALCQH}_{R^+}$  from the  $\mathcal{SI}$  family

is presented. This DL does not have inverse roles, but a role hierarchy and numerical restrictions. The difficulty of creating a tableau algorithm for this DL is that with numerical restrictions an infinite chain of adding individual names can prevent the termination of the algorithm. Hence a blocking technique is also needed to ensure termination: An individual name  $x$  blocks another individual name  $y$  if  $x$  was introduced before  $y$  and if any assertion about  $x$  also holds for  $y$ . To deal with the number restrictions a reasoner about sets of linear inequations is used.

Regarding the DL  $\mathcal{ALCSCC}$ , which is considered in this work, Baader provides a solution for the satisfiability problem without a TBox in [1], which is PSpace-complete: For a part of the ABox we guess the values (true or false) of the top-level atoms (concepts). This can already lead to a *false* result, which would mean that the ABox is unsatisfiable. If not, then the ABox is formulated into a QFBAPA formula. Then the formula is extended with constraints over the *Venn regions* of the concepts. For the new formula we test whether it returns true or false with a satisfiability algorithm for QFBAPA. This satisfiability algorithm runs in NP. If the algorithm returns true we are done. If it returns false, we create a concept for every guessed Venn region. Then the algorithm is applied on these new concepts recursively. If it return false, the ABox is unsatisfiable, otherwise satisfiable.

## Chapter 2

# Preliminaries

In order to be able to follow the construction of the tableau algorithm for  $\mathcal{ALCSCC}$  this section provides a number of important definitions on QFBAPA and  $\mathcal{ALCSCC}$ .

### 2.1 QFBAPA

The logic QFBAPA [7] combines boolean algebra (BA) over a sets of symbols with Presburger arithmetic (PA). *Terms* in boolean algebra over a symbol set  $T$  are comprised of conjunctions ( $\cap$ ) and/or disjunctions ( $\cup$ ) of symbols. These symbols can also be used in negated form ( $s^\neg$ ,  $s \in T$ ). *Terms* in Presburger arithmetic are additions of natural numbers. In QFBAPA we construct *set terms* using boolean algebra and create *cardinality terms* over the cardinalities of those set terms with the help of Presburger arithmetic. As multiplications can be constructed as chained additions, we also allow multiplication in cardinality terms. QFBAPA allows the construction of inclusion and comparison constraints over set and cardinality terms. This is defined as:

**Definition 1** (QFBAPA). Let  $T$  be a finite set of symbols

- set terms over  $T$  are:
  - empty set  $\emptyset$  and universal set  $\mathcal{U}$
  - every set symbol in  $T$
  - if  $s, t$  are set terms then so are  $s \cap t$ ,  $s \cup t$  and  $s^\neg$
- set constraints over  $T$  are
  - $s \subseteq t$  and  $s \not\subseteq t$
  - $s = t$  and  $s \neq t$

where  $s, t$  are set terms

- cardinality terms over  $T$  are:
  - every number  $n \in \mathbb{N}$
  - $|s|$  if  $s$  is a set term
  - if  $k, l$  are cardinality terms then so are  $k + l$  and  $n \cdot k$ ,  $n \in \mathbb{N}$
- cardinality constraints over  $T$  are:
  - $k = l$  and  $k \neq l$
  - $k < l$  and  $k \geq l$
  - $k \leq l$  and  $k > l$
  - $n \text{ dvd } k$  and  $n \neg \text{dvd } k$  ( $n \text{ dvd } k$ :  $n$  divides  $k$ )

where  $k, l$  are cardinality terms and  $n \in \mathbb{N}$

A QFBAPA formula  $\phi$  consists of disjunctions ( $\vee$ ) and/or conjunctions ( $\wedge$ ) of (possible negated) cardinality constraints, where every set symbol is represented as a set variable.

Set constraints of the form  $s \subseteq t$  can be expressed as cardinality constraints like  $|s \cap t^c| \leq 0$ . Analogously  $s \not\subseteq t$  can be expressed as  $|s \cap t^c| > 0$ . Set constraints of the form  $s = t$  can be written as  $|s \cap t^c| \leq 0$  and  $|s^c \cap t| \leq 0$ . Analogously for  $s \neq t$ . As all set constraints can be written as cardinality constraints we will not use them any further. For better readability we write  $k \leq l$  instead of  $l \geq k$ ,  $k + 1 \leq l$  instead of  $k < l$  and  $k \leq l$  and  $l \leq k$  instead of  $k = l$ .

As an example for a QFBAPA formula consider the symbols  $T = \{l, a, n, e, f\}$  and the constraints  $|l| = 2$ ,  $|l| = |a|$ ,  $|e \cap f^c| = 0$ ,  $|n \cap f^c| = 0$ . A formula can be written as:

$$|l| = 2 \wedge |l| = |a| \wedge |e \cap f^c| = 0 \wedge |n \cap f^c| = 0 \quad (2.1)$$

To satisfy this formula we have to create a semantic that satisfies all cardinality constraints (since all constraints are connected with  $\wedge$ ). In this case we need two elements which are in the semantic of  $l$  and therefore also two elements in the semantic of  $a$ .

The semantics of QFBAPA, called substitutions, are defined as follows:

**Definition 2** (Substitutions of QFBAPA). A substitution  $\sigma$  over a symbol set  $T$  is a mapping that assigns

- $\mathcal{U}$  to a finite set  $\sigma(\mathcal{U})$
- every symbol  $a$  in  $T$  to  $\sigma(a) \subseteq \sigma(\mathcal{U})$
- $\emptyset$  to  $\sigma(\emptyset) = \emptyset$
- $\sigma(s \cap t) := \sigma(s) \cap \sigma(t)$ ,  $\sigma(s \cup t) := \sigma(s) \cup \sigma(t)$
- $\sigma(s^c) := \sigma(\mathcal{U}) \setminus \sigma(s)$

- $\sigma(|s|) := |\sigma(s)|$
- $\sigma(k + l) := \sigma(k) + \sigma(l)$ ,  $\sigma(n \cdot k) := n \cdot \sigma(k)$

Given cardinality terms  $k, l$  we say that  $\sigma$  satisfies

- $k \leq l$  iff  $\sigma(k) \leq \sigma(l)$
- $n \text{ dvd } l$  iff  $\exists m \in \mathbb{N} : n \cdot m = \sigma(l)$

The substitution  $\sigma$  is a solution of a QFBAPA formula  $\phi$  if it evaluates the formula to  $\top$ . A QFBAPA formula is satisfiable if it has a solution  $\sigma$  and is unsatisfiable otherwise.

For our example (2.1) a possible solution  $\sigma$  is: Let  $\sigma(\mathcal{U}) = \{\text{leg1, leg 2, arm1, arm 2, nose, ear1, ear2}\}$  and:

- $\sigma(l) = \{\text{leg1, leg2}\}$
- $\sigma(a) = \{\text{arm1, arm2}\}$
- $\sigma(n) = \{\text{nose}\}$
- $\sigma(e) = \{\text{ear1, ear 2}\}$
- $\sigma(f) = \{\text{nose, ear1, ear2}\}$

This interpretation satisfies the formula because  $\sigma(|l|) = 2 = \sigma(|k|)$ ,  $\sigma(|n \cap f^-|) = 0$  and  $\sigma(|e \cap f^-|) = 0$ .

Now we can interpret our formula as:

- we have 2 legs
- we have as many legs as arms
- nose and two ears are both in the same set hence they belong to a common body part (face)

In [7] Kuncak and Rincard show that checking satisfiability of QFBAPA formulas is a NP-complete problem.

## 2.2 $\mathcal{ALCSCC}$

Next we define the parts and semantics of the description logic  $\mathcal{ALCSCC}$  [1]. Let  $\mathbf{C}$  be a set of concept names and  $\mathbf{R}$  a set of role names, such that  $\mathbf{C}$  and  $\mathbf{R}$  are disjoint.

**Definition 3** ( $\mathcal{ALCSCC}$ ).  $\mathcal{ALCSCC}$  concepts over  $\mathbf{C}$  and  $\mathbf{R}$  are defined inductively as:

- all concept names in  $\mathbf{C}$
- if  $C, D$  are  $\mathcal{ALCSCC}$  concepts over  $\mathbf{C}$  and  $\mathbf{R}$  then so are:
  - $\neg C$





We can define a model  $\mathcal{I}$  of the ABox defined in (2.2) by setting  $\Delta^{\mathcal{I}} = \{Anna^{\mathcal{I}}, Leg1^{\mathcal{I}}, Leg2^{\mathcal{I}}, Arm1^{\mathcal{I}}, Arm2^{\mathcal{I}}\}$  and

- $Female^{\mathcal{I}} = \{Anna^{\mathcal{I}}\}$
- $Leg^{\mathcal{I}} = \{Leg1^{\mathcal{I}}, Leg2^{\mathcal{I}}\}$
- $Arm^{\mathcal{I}} = \{Arm1^{\mathcal{I}}, Arm2^{\mathcal{I}}\}$
- $bodyPart^{\mathcal{I}} = \{(Anna, Leg1), (Anna, Leg2), (Anna, Arm1), (Anna, Arm2)\}$

By mapping  $Anna$  to  $Anna^{\mathcal{I}}$ ,  $Leg1$  to  $Leg1^{\mathcal{I}}$  and so on we see that this interpretation satisfies the ABox:  $Anna^{\mathcal{I}}$  is indeed in  $succ(|Legs \cap bodyParts| = 2)^{\mathcal{I}}$  because  $Leg1^{\mathcal{I}}, Leg2^{\mathcal{I}} \in bodyPart^{\mathcal{I}_{Anna}} \cap Leg^{\mathcal{I}_{Anna}}$ . Analogously for the second *succ*-assertion.

Next we define the *negated normal form* (NNF) for  $\mathcal{ALCSCC}$ . By transforming all concepts into *NNF* we avoid nested negations e.g.  $\neg(\neg(\neg(A \cup B)))$  which helps to formulate the rules for the tableau algorithm.

**Definition 5** (Negation Normal Form). A  $\mathcal{ALCSCC}$  concept is in *negation normal form* (NNF) if the negation sign  $\neg$  only appears in front of a concept name or above a role name. Let  $C$  be an arbitrary  $\mathcal{ALCSCC}$  concept. With  $NNF(C)$  we denote the concept which is obtained by applying the the following rules on  $C$  until none of them are applicable anymore.

- |   |  |
|---|--|
| 1 $\neg \top \rightarrow \perp$                       | 8 $\neg(k \leq l) \rightarrow l \leq k$                      |
| 2 $\neg \perp \rightarrow \top$                       | 9 $\neg(n \text{ dvd } k) \rightarrow n \neg \text{dvd } k$  |
| 3 $\neg \neg C \rightarrow C$                         | 10 $\neg(n \neg \text{dvd } k) \rightarrow n \text{ dvd } k$ |
| 4 $\neg(C \sqcap D) \rightarrow \neg C \sqcup \neg D$ | 11 $(s \cap t)^{\neg} \rightarrow s^{\neg} \cup t^{\neg}$    |
| 5 $\neg(C \sqcup D) \rightarrow \neg C \sqcap \neg D$ | 12 $(s \cup t)^{\neg} \rightarrow s^{\neg} \cap t^{\neg}$    |
| 6 $C^{\neg} \rightarrow \neg C$                       | 13 $(s^{\neg})^{\neg} \rightarrow s$                         |
| 7 $\neg succ(c) \rightarrow succ(\neg c)$             |  |

The rule  $C^{\neg} \rightarrow \neg C$  is necessary because  $C^{\neg}$  can be a result of  $s^{\neg}$ , where  $s$  is a set term.  $C^{\neg}$  can be transformed into  $\neg C$ : For every substitution  $\sigma$  for a concept  $C$  based on QFBAPA it holds that  $\sigma(C^{\neg}) = \sigma(\mathcal{U}) \setminus \sigma(C)$  and for every interpretation  $\mathcal{I}$  based on  $\mathcal{ALCSCC}$  it holds that  $(\neg C)^{\mathcal{I}} = \Delta^{\mathcal{I}} \setminus C^{\mathcal{I}}$ . Since  $\sigma(\mathcal{U}) \subseteq \Delta^{\mathcal{I}}$  we can conclude that every element in  $\sigma(C^{\neg})$  is also in  $(\neg C)^{\mathcal{I}}$ .

All rules used to obtain the NNF can be applied in linear time. For rules 1-5 this is shown in [5],[9]. For rules 6-11 this can be shown analogously because we only shift the negation signs. Rules 11, 12 and 13 work similarly to rules 4, 5 and 3 respectively.

Regarding the normal form we additionally replace every disjunction and conjunction in form as  $\sqcup$  and  $\sqcap$  in every  $succ(c)$  concept with  $\cup$  and  $\cap$ . We can do

this because for an arbitrary interpretation  $\mathcal{I}$  for each  $x, y \in \Delta^{\mathcal{I}}$  it holds that  $y \in (C \sqcap D)^{\mathcal{I}_x}$  iff  $y \in (C \cap D)^{\mathcal{I}_x}$ :

$$\begin{aligned}
y \in (C \sqcap D)^{\mathcal{I}_x} & \quad \Leftrightarrow \\
y \in (C \sqcap D)^{\mathcal{I}} \cap \mathcal{U}^{\mathcal{I}_x} & \quad \Leftrightarrow \\
y \in C^{\mathcal{I}} \cap D^{\mathcal{I}} \cap \mathcal{U}^{\mathcal{I}_x} & \quad \Leftrightarrow \\
y \in (C^{\mathcal{I}} \cap \mathcal{U}^{\mathcal{I}_x}) \cap (D^{\mathcal{I}} \cap \mathcal{U}^{\mathcal{I}_x}) & \quad \Leftrightarrow \\
y \in C^{\mathcal{I}_x} \cap D^{\mathcal{I}_x} & \quad \Leftrightarrow \\
y \in (C \cap D)^{\mathcal{I}_x} & 
\end{aligned}$$

Next we define the size of an  $\mathcal{ALCSCC}$  concept  $C$  inductively over concepts, set terms and cardinality constraints. This definition is necessary for the termination proof.

- $size(r) = size(C) = 1$  if  $r \in \mathbf{R}$ ,  $C \in \mathbf{C}$
- $size(n) = size(|k|) = 1$  if  $n \in \mathbb{N}$ ,  $k$  cardinality term
- $size(\neg C) = size(C) + 1$
- $size(k^-) = size(k) + 1$
- $size(C \sqcap D) = size(C \sqcup D) = size(C) + size(D) + 1$
- $size(k \cap l) = size(k \cup l) = size(k) + size(l) + 1$
- $size(|k|) = size(k)$
- $size(succ(c)) = \begin{cases} 1 + size(k) + size(l) & c = k \leq l \\ 1 + size(l) & c = n \text{ dvd } l \end{cases}$

## Chapter 3

# Tableau for $\mathcal{ALCSCC}$

The tableau algorithm is a popular tool to check satisfiability. Even though the complexity of tableau algorithms can grow exponentially one advantage of them is that they do not only check if an ABox is satisfiable but also return an interpretation that satisfies this ABox (*witness*) if one exists. A tableau algorithm consists of completion rules that are used to iteratively add new assertions to the ABox that are derived from pre-existing assertions. These rules are exhaustively applied to the ABox until there are no more applicable rules. For some completion rules like the rule for disjunctions ( $x : C \sqcup D$ ), the algorithm can decide which assertion is added to the ABox (either  $x : C$  or  $x : D$ ). If such a choice results in a *clash* the algorithm back tracks to the point of the decision and tries an alternative choice instead. If all choices end in a clash, then the ABox is unsatisfiable.

Before *clashes* can be defined we first need to introduce the concept of *induced interpretations*. Induced interpretations can be used to count the number of successors of any individual name after any rule application and hereby detect *violated assertions*.

**Definition 6** (Induced Interpretation). An interpretation  $\mathcal{I}(\mathcal{A})$  can be induced from an ABox  $\mathcal{A}$  through the following steps:

- for each individual name  $x \in I(\mathcal{A})$  we introduce  $x^{\mathcal{I}(\mathcal{A})}$  and add it to  $\Delta^{\mathcal{I}(\mathcal{A})}$
- for each  $x : C$  such that  $C$  is a concept name we add  $x^{\mathcal{I}(\mathcal{A})}$  to  $C^{\mathcal{I}(\mathcal{A})}$
- for each  $(x, y) : r$  such that  $r$  is a role name we add  $(x^{\mathcal{I}(\mathcal{A})}, y^{\mathcal{I}(\mathcal{A})})$  to  $r^{\mathcal{I}(\mathcal{A})}$

**Definition 7** (Violated assertion). Let  $\mathcal{A}$  be an ABox and  $x$  be an individual name in  $I(\mathcal{A})$ . An assertion  $x : succ(c)$  is *violated* if  $x^{\mathcal{I}(\mathcal{A})} \notin succ(c)^{\mathcal{I}(\mathcal{A})}$ .

Violated assertions can sometimes be resolved by applying further completion rules. However if we find a violated assertion and can no longer apply any rules there is a clash. Aside from unresolvable violated assertions there are other kinds of situations that are also labelled as clashes.

**Definition 8** (Clash). An ABox  $\mathcal{A}$  contains a *clash* if

- $\{x : \perp\} \subseteq \mathcal{A}$  or
- $\{x : C, x : \neg C\} \subseteq \mathcal{A}$  or
- $\{(x, y) : r, (x, y) : \neg r\} \subseteq \mathcal{A}$  or
- $x : succ(c) \in \mathcal{A}$  violated and no more rules are applicable

### 3.1 Transforming an ABox into a formula

One major difficulty of creating a tableau algorithm for  $\mathcal{ALCSCC}$  is its numerical arithmetic in the form of successor-assertions. The application of rules can change the number of successors which can in turn influence the number of successors that are demanded by certain constraints. Furthermore it introduces the problem of *nested* successor assertions.

**Definition 9** (Nested Level). Let  $\mathcal{A} = \{x : C\}$  be an ABox. An individual name lays in the  $i$ -th nested level if it is the  $i$ -th individual name in a role chain beginning from  $x$ , where the individual name  $x$  is in the 0th nested level. A direct successor of  $x$  is in the 1st nested level.

In some DLs we are able to describe the successor of a successor like  $\exists r.(\exists r.C)$ , such that the number of needed successors is fixed. However in  $\mathcal{ALCSCC}$  such boundaries can vary like in  $x : succ(|succ(|A| < |B|)| > |A|)$ . The number of successors needed to satisfy  $succ(|A| < |B|)$  depends on how many successors  $x$  already has in  $A$ . By applying rules the number of successors for  $x$  in  $A$  can change. Hence we use a QFBAPA solver whenever we want to add successors for an individual name  $x$ . To do this we first collect all *succ*-assertions regarding  $x$  and then transform them into a QFBAPA formula for the next nested level, which means we only consider the direct successors of  $x$ . We assume that the ABox is already in *NNF*. To create an example for a transformation we consider the following ABox:

**Example 1.**

$$\mathcal{A} = \{x : succ(1 \leq |succ(|A| \leq |B \cap r|)|), x : succ(|A| \leq |B|), x : C\}$$

with  $\mathbf{C} = \{A, B, C\}$  and  $\mathbf{R} = \{r\}$

In this example the first assertion states that  $x$  must have at least one successor  $y$  which in turn has at least as many successors in  $B \cap r$  as in  $A$ . The individual names for  $succ(|A| \leq |B \cap r|)$  are on a different nested level than the ones for  $succ(|A| \leq |B|)$ . The second assertion states that  $x$  has at least as many successors in  $B$  as in  $A$ .

We start by gathering all *succ*-assertion regarding  $x$  and then transform the cardinality constraints into a formula by carrying out the following steps:

- replace all role names  $r$  with  $X_r$
- replace all concepts names  $C$  with  $X_C$
- replace all  $\text{succ}(c)$  with  $X_{\text{succ}(c)}$
- connect all formulas with  $\wedge$
- include the conjunct  $\mathcal{U} = X_{r_1} \cup \dots \cup X_{r_n}, r_1, \dots, r_n \in \mathbf{R}$

We replace (compound) concepts and role names with set variables, so a solver can assign elements to them. The last step is important because sometimes it is not explicitly stated what kind of successor an individual name has. However a successor must always be "connected" to its predecessor by at least one role name. Through the last step we ensure that every element (successor) is assigned to a set variable, which represents a role name.

In our example we have five set variables:  $X_A, X_B, X_r, X_{\text{succ}(|A| \leq |B \cap r|)}$  and  $\mathcal{U}$ . The QFBAPA formula for Example 1 is:

$$\phi = 1 \leq |X_{\text{succ}(|A| \leq |B \cap r|)}| \wedge |X_A| \leq |X_B| \wedge \mathcal{U} = X_r \quad (3.1)$$

We can use the solver to get a possible solution, if there is one. We make two assumptions about the QFBAPA solver so we can use it in the tableau algorithm.

**Assumption 1.** We assume that every considered QFBAPA solver is correct which means

- it terminates for all QFBAPA formulas
- a formula is satisfiable iff it returns a solution

**Assumption 2.** Let  $\phi$  be an arbitrary QFBAPA formula. We assume that every considered QFBAPA solver is able to return all possible solutions of  $\phi$ .

## 3.2 Solution of a formula

Since we make Assumption 2 there can be infinitely many solutions. Actually Example 1 can have infinitely many solutions: We can always increase the amount of successors in  $B$  as long as we have fewer successors in  $A$ . However having that means that the tableau algorithm can work on an infinite space and/or might not terminate. Therefore we want to only consider solutions inside some pre-computed *upper bounds*. For an *Integer Linear Programming* (ILP) problem, which is described as a system of linear equalities, possible upper bounds are already investigated like in [8], which solution we are going to use. Therefore we want to transform our formula into a linear system of equalities of the form  $Ax = b$ , where  $A$  and  $b$  describe our cardinality constraints and  $x$  is the solution i.e. denotes the numbers of elements we have to assign to set variables to satisfy the formula.

First, we notice that every inequality in a QFBAPA formula can be rewritten as  $n_1 \cdot |X_1| \pm \dots \pm n_i \cdot |X_i| \lesseqgtr I$ ,  $\lesseqgtr \in \{\leq, \geq, =\}$ , where  $n_1, \dots, n_i, I \in \mathbb{Z}$  are constants. The numbers  $n_1, \dots, n_i$  are called *pre-factors*. Let  $c = \text{succ}(|A| \leq |B \cap r|)$ . We arrange  $\phi$  in (3.1) into

$$\phi' = |X_c| \geq 1 \wedge |X_A| - |X_B| \leq 0 \wedge |\mathcal{U} \cap X_r^-| = 0 \wedge |\mathcal{U}^- \cap X_r| = 0 \quad (3.2)$$

Then we change each inequalities into equalities by adding two slack variables  $I_1$  and  $I_2$ :

$$\begin{aligned} \phi'' &= |X_c| - I_1 = 1 \wedge |X_A| - |X_B| + I_2 = 0 \wedge \\ &|\mathcal{U} \cap X_r^-| = 0 \wedge |\mathcal{U}^- \cap X_r| = 0 \end{aligned} \quad (3.3)$$

Right now it is not clear whether the set variables are overlapping or not which is a problem because for a system of linear equations the variables are always disjunct. Therefore we consider *Venn regions*, which are of the form  $X_1^i \cap \dots \cap X_k^i$ . The subscript  $i$  denotes either 0 or 1.  $X_1^0$  denotes  $X_1^-$  and  $X_1^1$  denotes  $X_1$ . Since we have 5 set variable, we have  $2^5 = 32$  Venn regions. We already see that the number of Venn regions grows exponentially with the number of set variables. In [1] it is stated that there exists a number  $N$ , which is polynomial in the size of  $\phi$ , such that at most  $N$  Venn regions are not empty if there exists a solution.

**Lemma 1** (Lemma 3 from [1]). For every QFBAPA formula  $\phi$  a number  $N$ , which is polynomial in the size of  $\phi$ , can be computed in polynomial time such that for every solution  $\sigma$  of  $\phi$  there exists a solution  $\sigma'$  of  $\phi$  such that

- $|\{v | v \text{ is a Venn region and } \sigma'(v) \neq \emptyset\}| \leq N$
- $\{v | v \text{ is a Venn region and } \sigma'(v) \neq \emptyset\} \subseteq \{v | v \text{ is a Venn region and } \sigma(v) \neq \emptyset\}$

Hence we guess (in non-deterministic polynomial time) a number  $N$  of Venn regions, which are non-empty. In Example 1 we can already guess that any Venn region within  $X_r^-$  or  $\mathcal{U}^-$  must be empty, because every element must be in  $\mathcal{U}$  and since  $\mathcal{U} = X_r$  they must be all in  $X_r$ . Hence we can drop 24 Venn regions. Therefore we construct  $A$  and  $b$  such that instead of assigning elements to set variables we assign them to the remaining 8 Venn regions. That means for the vector  $x$  the entry  $x_k$ ,  $1 \leq k \leq 8$ , denotes the number of elements in the  $k$ -th Venn region. Since we have four equations and two slack variables the matrix  $A$  has four rows and 10 columns with  $a_{ij}$ ,  $1 \leq i \leq 4$  and  $1 \leq j \leq 10$ , denoting the sum of the pre-factors of the set variables, in which the  $j$ -th Venn region is included, occurring in the  $i$ -th equation. Let the vectors for the Venn regions be in the following order:

- $v_1 = X_A \cap X_B \cap X_c \cap X_r \cap \mathcal{U} \cap I_1^- \cap I_2^-$
- $v_2 = X_A \cap X_B \cap X_c^- \cap X_r \cap \mathcal{U} \cap I_1^- \cap I_2^-$
- $v_3 = X_A \cap X_B^- \cap X_c \cap X_r \cap \mathcal{U} \cap I_1^- \cap I_2^-$

- $v_4 = X_A \cap X_B^- \cap X_c^- \cap X_r \cap \mathcal{U} \cap I_1^- \cap I_2^-$
- $v_5 = X_A^- \cap X_B \cap X_c \cap X_r \cap \mathcal{U} \cap I_1^- \cap I_2^-$
- $v_6 = X_A^- \cap X_B \cap X_c^- \cap X_r \cap \mathcal{U} \cap I_1^- \cap I_2^-$
- $v_7 = X_A^- \cap X_B^- \cap X_c \cap X_r \cap \mathcal{U} \cap I_1^- \cap I_2^-$
- $v_8 = X_A^- \cap X_B^- \cap X_c^- \cap X_r \cap \mathcal{U} \cap I_1^- \cap I_2^-$

The last two vectors denote the slack variables

- $v_9 = X_A^- \cap X_B^- \cap X_c^- \cap X_r^- \cap \mathcal{U}^- \cap I_1 \cap I_2^-$
- $v_{10} = X_A^- \cap X_B^- \cap X_c^- \cap X_r^- \cap \mathcal{U}^- \cap I_1^- \cap I_2$

We create now the linear system of equation:

$$\begin{pmatrix} 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & -1 & -1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \\ x_9 \\ x_{10} \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \end{pmatrix}$$

Note that  $a_{2,1}$  and  $a_{2,2}$  are 0 because the pre-factors of  $|X_A|$  and  $|X_B|$  in the second equation are 1 and  $-1$  and the Venn regions  $v_1$  and  $v_2$  are included both in  $X_A$  and  $X_B$ . If  $x_i = 0$ , then the  $i$ -th Venn region is empty. The last two rows of  $A$ , which present the equations  $|\mathcal{U} \cap X_r^-| = 0$  and  $|\mathcal{U}^- \cap X_r| = 0$ , are lines of zeros because we omit the Venn regions, in which  $\mathcal{U} \cap X_r^-$  and  $\mathcal{U}^- \cap X_r$  are included. However we do not loose information because the information is that those Venn regions must be empty.

Going back to the upper bound problem: The following result from [8] can be used to establish an upper bound for the solution of the *ILP* in NP:

**Theorem 1** (Theorem 1 from [8]). Let  $A \in \mathbb{Z}^m \times \mathbb{Z}^n$  be a matrix and  $b \in \mathbb{Z}^m$  a vector. If  $x \in \mathbb{N}^n$  is a solution of  $Ax = b$ , then there exists a solution  $x'$  such that all entries are integers between 0 and  $n \cdot (m \cdot \max_{i,j} \{|a_{ij}|, |b_i|\})^{2 \cdot m+1}$ .

We take a look now in the proof of this theorem to understand how the solution is decreased. We distinguish between two cases. Let  $M = m \cdot \max_{i,j} \{|a_{ij}|\}^m$ ,  $F = \{i | x_i > M\}$  and  $v_i$  be the  $i$ -th column of  $A$

- If there exist integers  $\alpha_i$ , for all  $i \in F$ , such that  $\sum_{i \in F} \alpha_i \cdot v_i = 0$  and  $\exists i : \alpha_i > 0$  then  $x' = x - d$ ,  $d_j = \alpha_j$  if  $j \in M$  else  $d_j = 0$ ,  $1 \leq j \leq n$ .

- Else: There must be a vector  $h \in \{0, \pm 1, \pm 2, \dots, \pm M\}^m$  such that  $h^T v_i \geq 1, i \in F$ . We premultiply  $A$  and  $b$  with  $h^T$  and show that  $x$  is already in the bound:

$$h^T Ax = h^T b$$

Therefore we are able to set an upper bound a priori for the solutions the QFBAPA solver returns, which is important for the termination proof.

In our example the upper bound for all  $x_i$  is  $10 \cdot (4 \cdot \max\{|1|, |-1|\})^{2 \cdot 4 + 1} = 2621440$ , which means that we can discard any solution in which a Venn region has more than 2621440 elements.

### 3.3 The Tableau Algorithm

Finally we can present the tableau algorithm for an ABox in  $\mathcal{ALCSCC}$ . Before we handle the numerical arithmetic of  $\mathcal{ALCSCC}$  we want to decompose compound concepts first. That means we want to consider conjunctions and disjunctions first and then consider the transformation to a QFBAPA formula. Hence we divide the algorithm in two parts: a boolean part, where the decomposition of compound concepts takes place, and a numerical part, where a part of the ABox is transformed into a QFBAPA formula. In the second part we calculate an upper bound and let a solver return a possible solution within this bound, in case the ABox is satisfiable. The boolean part has a higher priority than the numerical part.

**Definition 10** (Tableau for  $\mathcal{ALCSCC}$ ). The completion rules for a  $\mathcal{ALCSCC}$  ABox  $\mathcal{A}$  in  $NNF$  are the following.

Boolean part:

- $\sqcap$ -rule:  $\mathcal{A}$  contains  $x : C_1 \sqcap C_2$  but not both  $x : C_1$  and  $x : C_2$   
 $\rightarrow \mathcal{A} := \mathcal{A} \cup \{x : C_1, x : C_2\}$
- $\sqcup$ -rule:  $\mathcal{A}$  contains  $x : C_1 \sqcup C_2$  but neither  $x : C_1$  nor  $x : C_2$   
 $\rightarrow \mathcal{A} := \mathcal{A} \cup \{x : C_1\}$  or  $\mathcal{A} := \mathcal{A} \cup \{x : C_2\}$

Numerical part:

- *successor*-rule:  $\mathcal{A}$  contains for an individual name  $x$  at least one violated assertion of the form  $x : succ(c)$ , this rule has not been applied for  $x$  yet and no boolean rules are applicable:
  - gather all assertions of the form  $x : succ(c)$  into a set  $\mathcal{S}$
  - transform  $\mathcal{S}$  into a QFBAPA formula  $\phi$  as in Section 3.1
  - calculate the upper bound as in Theorem 1

If the QFBAPA solver returns *unsatisfiable*, then  $\mathcal{A} := \mathcal{A} \cup \{x : \perp\}$

If the QFBAPA solver returns *satisfiable*, then select one solution  $\sigma$  with in the upper bound. For each  $e \in \sigma(\mathcal{U})$ , we introduce a new individual name  $y_e$  and



- if  $e \in X_C$  we set  $\mathcal{A} := \mathcal{A} \cup \{y_e : C\}$
- if  $e \in X_{succ(c)}$  we set  $\mathcal{A} = \mathcal{A} \cup \{y_e : succ(c)\}$
- if  $e \in X_r$ ,  $r \in \mathbf{R}$ , we set  $\mathcal{A} := \mathcal{A} \cup \{(x, y_e) : r\}$
- if  $e \notin X_C$  we set  $\mathcal{A} := \mathcal{A} \cup \{y_e : \neg C\}$
- if  $e \notin X_{succ(c)}$  we set  $\mathcal{A} := \mathcal{A} \cup \{y_e : NNF(\neg succ(c))\}$
- if  $e \notin X_r$ ,  $r \in \mathbf{R}$ , we have  $\mathcal{A} := \mathcal{A} \cup \{(x, y_e) : \neg r\}$

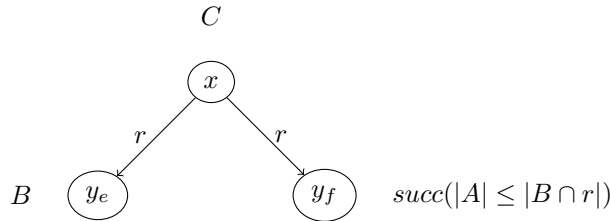
A *complete* ABox is an ABox to which no more rules of the tableau algorithm are applicable.

We present a possible run of this algorithm over the following ABox:

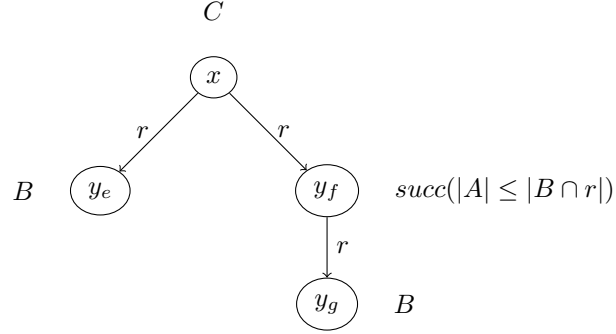
**Example 2.**

$$\mathcal{A} = \{x : succ(1 \leq |succ(|A| \leq |B \cap r|)|) \sqcap succ(|A| \leq |B|) \sqcap C\}$$

We are able to apply to apply the  $\sqcap$ -rule and hence we apply them first and derive the ABox in Example 1. Then neither the  $\sqcap$ - nor the  $\sqcup$ -rule is applicable. Therefore we apply the *successor*-rule: We collect every *succ*-assertion regarding  $x$  to a set  $\mathcal{S} := \{x : succ(1 \leq |succ(|A| \leq |B \cap r|)|), x : succ(|A| \leq |B|)\}$  and convert  $\mathcal{A}$  to the QFBAPA formula in (3.1). The upper bound for that formula is 2621440 (see Section 3.2). If the formula have been unsatisfiable, we would add  $x : \perp$  to our ABox. this would have led us to the end of the tableau algorithm because this "choice" is the only one possible since we do not have any disjunction. However since the formula is satisfiable we obtain a solution from the solver because of Assumption 1. We see that the formula is satisfiable with  $X_{succ(|A| \leq |B \cap r|)} = \{f\}$ ,  $X_A = \{\}$ ,  $X_B = \{e\}$  and  $X_r = \{e, f\}$ . Since we have  $\mathcal{U} = \{X_r\}$  every element must be in  $X_r$  i.e. every successor is a  $r$ -successor. The considered QFBAPA formula is capable of returning this solution because of Assumption 2. Obviously in this solution every Venn-region has less element than 2621440. We then introduce for  $e$  and  $f$  two individual names  $y_e$  and  $y_f$  and the assertion  $y_e : X_B$ ,  $(x, y_e) : r$ ,  $y_f : succ(|A| \leq |B \cap r|)$  and  $(x, y_f) : r$  to  $\mathcal{A}$ .



Then for  $y_f$  we have to apply the *successor*-rule because no boolean rule is applicable. Since  $y_f : succ(|A| \leq |B \cap r|)$  is the only *succ* assertion we only have to add an  $r$ -successor of the concept  $B$ . We assume that the QFBAPA solver returns a solution  $X_B = X_r = \{g\}$  and hereby introduce an individual name  $y_g$  and add  $y_g : B$  and  $(x, y_g) : r$  to  $\mathcal{A}$ .



In this procedure we do not add any assertions of individual names which are not freshly introduce. Hence if we applied all possible rules on the individual name  $x$  and we do not have a clash, then all assertions of  $x$  remains satisfied until the tableau algorithm ends. In the next chapter we provide a formal proof of the its correctness.

# Chapter 4

## Correctness

For the correctness proof of the tableau algorithm we have to show the following:

- For every input the tableau algorithm terminates.
- If no more rules are applicable on a clash-free ABox  $\mathcal{A}$ , then  $\mathcal{A}$  is satisfiable.
- If  $\mathcal{A}$  is satisfiable, then the tableau algorithm terminates without a clash.

In all proofs we consider Assumption 1 and 2. First we prove that the algorithm always terminates.

### 4.1 Termination

We define a *derived* ABox as an ABox  $\mathcal{A}_2$  after a finite number of rule applications on an ABox  $\mathcal{A}_1$ . Each rule terminates:

- $\sqcap$ - and  $\sqcup$ -rule: Obviously both rule terminates because we decompose finite compound concepts.
- *successor*-rule: Since ABoxes are finite we can only have a finite subset  $\mathcal{S}$  of *successor*-assertions and therefore can always form a (finite) QFBAPA formula. By Assumption 1 the QFBAPA solver always terminates and we gain a finite solution. Therefore we always add a finite number of successor in this rule. Hence this rule application always terminates.

For the termination proof we map every ABox  $\mathcal{A}$  to an element  $\Psi(\mathcal{A})$  of a set  $Q$ . Each  $\Psi(\mathcal{A})$  consists of triples  $\psi_{\mathcal{A}}(x)$  for each individual name  $x$ . Let  $\prec$  be an strict partial ordering, which is a irreflexive (if  $a \prec b$  then  $b \not\prec a$ ) and transitive (if  $a \prec b$  and  $b \prec c$  then  $a \prec c$ ) relation (with  $a, b, c$  being comparable elements). If we show that  $\prec$  is well-founded, e.g. there is no infinite decreasing chain, and that for every ABox  $\mathcal{A}'$ , which is derivable from an ABox  $\mathcal{A}$ , we can only have  $\Psi(\mathcal{A}') \prec \Psi(\mathcal{A})$  then the termination of the algorithm can be concluded. Each triple in  $Q$  consists of a multiset of natural numbers and two natural













- $y_e : succ(c) \in \mathcal{A}_{i+1}$  iff  $e \in succ(c)^{\mathcal{I}}$
- $(x, y_e) : r \in \mathcal{A}_{i+1}$  iff  $(x^{\mathcal{I}}, e) \in r^{\mathcal{I}}$

Hence we can extend  $\mathcal{I}$  by  $y_e^{\mathcal{I}} = e$  which satisfies  $\mathcal{A}_{i+1}$ .

□



## Chapter 5

# Conclusion

The description logic  $\mathcal{ALCSCC}$  extends the well-known description logic  $\mathcal{ALCQ}$  with set constraints and cardinality constraints over role successors which are hard to deal with when checking satisfiability over  $\mathcal{ALCSCC}$ . We present a tableau algorithm which uses a QFBAPA solver to deal with the successors. It is shown that checking satisfiability for this DL is in PSpace, however the tableau algorithm runs in 2ExpSpace: We know that in each *successor*-rule application we can add in worse case exponential many successors as shown in Section 3.2. Each of the newly added successors is also capable of obtaining exponential many successors. The advantage of this approach is that we do not only return the statement about the satisfiability but also a satisfied interpretation, a model, for the concept.

For future works one can extend the algorithm for  $\mathcal{ALCSCC}$  concepts w.r.t. a TBox. In [1] it is stated that the satisfiability problem w.r.t. a TBox is in ExpTime. One approach can be to add all information we can concluded from the TBox first i.e. if we have  $C \sqsubseteq D$  (says that every individual name in  $C$  is also in  $D$ ) and  $x : C$  then  $x : D$  has to be added to the ABox, too. We have to do the same for every freshly introduced individual name, too. One can also research for a tableau algorithm which does not use a QFBAPA solver. For that introducing blocking technique is most likely required because like in for  $\mathcal{ALCQ}$  and the  $\mathcal{SI}$  families endless loops of adding and merging (or replacing) individual names are possible. Another interesting area to do research on is the pre-computed upper bound. In this work the bound is exponentially large which is the reason for the ExpSpace complexity. If there exists a smaller upper bound one can run this algorithm in a smaller complexity.



# Bibliography

- [1] F. Baader. A New Description Logic with Set Constraints and Cardinality Constraints on role Successors, 2017.
- [2] F. Baader and T. Nipkow. *Abstract Reduction Systems*, page 7–33. Cambridge University Press, 1998.
- [3] V. Haarslev. Combining tableau and algebraic methods for reasoning with qualified number restrictions in description logics. 05 2002.
- [4] R. Hoehndorf, P. N. Schofield, and G. V. Gkoutos. The role of ontologies in biological and biomedical research: a functional perspective. *Brief. Bioinform*, page 1069–1080, 2015.
- [5] B. Hollunder and F. Baader. Qualifying Number Restrictions Concept Languages. Technical report, Research Report RR-91-03, 1991.
- [6] I. Horrocks, U. Sattler, and S. Tobies. Practical reasoning for very expressive description logics. *CoRR*, cs.LO/0005014, 05 2000.
- [7] V. Kuncak and M. Rinard. Towards efficient satisfiability checking for Boolean Algebra with Presburger Arithmetic. In *Conference on Automated Deduction (CADE-21)*, volume 4603 of *LNCS*, 2007.
- [8] C. H. Papadimitriou. On the complexity of integer programming. *J. ACM*, 28(4):765–768, Oct. 1981.
- [9] S. Tobies. *Complexity Results and Practical Algorithms for Logics in Knowledge Representation*. PhD thesis, Rheinisch-Westfälischen Technischen Hochschule Aachen, 2001.
- [10] S. Tobies and H. Ganzinger. A PSpace Algorithm for Graded Modal Logic. In *In Proc. CADE 1999, Lecture Notes in Artificial Intelligence*, volume 1632, pages 674–674, 01 1999.