A Polynomial Approach for finding Most Specific Concept w.r.t General $\mathcal{EL}$-TBoxes

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Master Thesis
A Polynomial Approach for finding Most Specific Concept w.r.t General $\mathcal{EL}$-TBoxes

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Objectives of work
The task to be solved in the Master's thesis is to implement and optimize a polynomial-time procedure that can test for the existence of the most specific concept and compute it in case it exists. Moreover, we provide a proof showing the correctness of this procedure. Finally, providing experimental results to show the compare it with the previous approach.

Focus of work
• Defining a polynomial-time procedure to check the existence of msc.
• Proving correctness and completeness of the defined procedure.
• Providing an implementation for the defined procedure.
• Providing an evaluation to compare the new approach with the previous approach.

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Abstract

Description logics (DLs) [Baa99; BHS04; Baa+17a; Tur13; KH17] are a family of formal knowledge representation languages used in artificial intelligence to describe and reason about the concepts of an application domain. In particular, they are providing formalization for ontologies and the semantic web. The most specific concept (msc) [BKM98] is an inference task that can support the bottom-up construction of knowledge bases in description logics. The most specific concept of an individual is the least concept that has this individual as an instance.

In description logics which contain existential restrictions, the most specific concept does not always exist in the case of acyclic ABoxes. However, the latest results [ZT13] show that the existence of the msc w.r.t. an individual can be decided in polynomial time. Also, the role-depth of these most specific generalizations is polynomially bounded by the size of the input, which yields a decision procedure for the existence problem. The polynomial bound can be used to compute the msc if it exists. Otherwise, the computed concept can still serve as an approximation[PT11]. However, computing the msc could take at least exponential time if the msc is exponentially large. Also, the previous approach does not make it clear how to compute the most specific concept in practice.

First, we revisit the previous approach for constructing the msc of an individual w.r.t. a general $\mathcal{EL}$-TBox. We present a new method for tree unravelling of an interpretation and introduce a characteristic concept w.r.t. the least tree unravelling. Moreover, we provide a new approach and an algorithm to decide the existence of the msc in polynomial time without relying on computing the concept. Then, one can compute the actual msc in exponential time. Finally, we provide an experimental evaluation to state that the concept constructed from the new approach has a smaller bound than the bound of the concept constructed from the previous approach.

Keywords— description logic, ontologies, most specific concept, polynomial time algorithms
Statement of authorship

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Dresden, 2nd September 2022

Mohamed Ali Abdelmajid Labib Nadeem

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1 Introduction

In philosophy, an ontology [Hof21] is the philosophical study of being in general, where concepts such as existence, being, becoming, and reality are studied. It includes the questions of how entities are grouped into basic categories and which of these entities exist on the most fundamental level. In computer science, an ontology [SS09] represents an abstract, simplified view of the relevant entities (objects, concepts, and relations) that exist in the domain of interest.

Ontology engineering is a field that studies methods for building an ontology. It aims to describe the knowledge of software applications, enterprises and business procedures for a particular domain. Supporting such applications require that the knowledge is presented precisely, so that it can be processed by automated tools. Ontology languages, such as OWL [Gra+08] is used to overcome this problem by defining a formal syntax and semantics to describe the knowledge of a particular domain.

Description Logics (DLs) are logic-based knowledge representation languages [Baa+17b] which are used to provide the formal foundation of the ontology language OWL for application domains such as the Semantic Web, biology and medicine, and engineering domains [HSG15]. It separates domain knowledge into a terminological part (called TBox) and an assertional part (called ABox), while the combination of both TBox and ABox is called a knowledge base (KB). The TBox describes the relations between concepts of an application domain, and the ABox describes the individual objects of an application domain and their relations. Concepts are built using concept and role names, and individual objects are built using concept and role assertions.

DLs are less expressive than first-order logic, but the core reasoning problems like the subsumption and the instance problem are usually decidable. Therefore, studying those problems is highly important in scientific and industrial fields. In applications of DL systems, it turned out that building and maintaining large DL knowledge bases can be facilitated by procedures for other, non-standard inference problems [BK06]. The non-standard inferences were introduced to maintain and support building large DL knowledge bases. An example of the non-standard inferences is generating a new concept from an individual (Also called the most specific concept of an individual).

Computing the most specific concept [Baa03a] is mainly used in the bottom-up construction of the knowledge base. This methodology allows the user to give an example of an individual belonging to the concept to be defined instead of defining the relevant concepts of an application domain from scratch. Thus, the knowledge engineer can use the computed concept as a starting point for defining the concept.

Moreover, since large ontologies might contain errors [Baa+21] often detected when DL reasoners compute unwanted consequences, it has gained interest in detecting errors in the data such that the unwanted consequences no longer follow, while keeping the other
1 Introduction

consequences are preserved. An optimal repair is the repair with the least number of other consequences are removed. Most of the previous approaches of the ontology repairing rely on the syntactic nature of the ontology without relying on the semantic nature. The recent work \cite{Baa+22} was addressed to compute the optimal repair of ABoxes, where the unwanted consequences are described as concept assertions. It also shows that the existence of such a type of repair can be decided by computing the most specific concept.

Unfortunately, the most specific concept does not always exist in $\mathcal{EL}$. It was investigated before by \cite{Baa03a} in the presence of a cyclic description logic $\mathcal{EL}$-TBox (which allows for conjunctions, existential restrictions and the top concept), where the bottom-up construction of the knowledge base was used.

Example 1. consider the following knowledge base \cite{KM02}:

\[
\mathcal{K}_1 := (\emptyset, \mathcal{A}_1), \text{ with } \mathcal{A}_1 := \{ r(a, a), C(a) \}
\]

The msc of the individual $a$ does not exist, as the TBox $\mathcal{T}_1 = \emptyset$ is empty, while the ABox $\mathcal{A}_1$ is cyclic \cite{Baa03a}, due to the fact there is no least concept such that the individual $a$ is an instance. The reason is that the concept $C$ has more specific concept $\exists r.C$, and $\exists r.C$ has more specific concept $\exists \exists r.C$ and so on, while this cycle in $\mathcal{A}_1$ is not covered by any cycle in $\mathcal{T}_1$. However, consider the modified knowledge base:

\[
\mathcal{K}_2 := (\mathcal{T}_2, \mathcal{A}_2), \text{ with } \mathcal{A}_2 := \{ r(a, a), C(a) \}, \text{ and } \mathcal{T}_2 := \{ C \sqsubseteq \exists r.C \}
\]

Since the cycle in $\mathcal{A}_2$ is covered by $\mathcal{T}_2$, we obtain that $C$ is the msc of $a$.

The first approach to decide the existence of the most specific concept \cite{ZT13} shows that deciding the existence of the msc w.r.t. an individual can be done in polynomial time, where the construction of the generalization concept is computed up to a given $k$, a bound on the maximal nestings of quantifiers. However, this approach rely on computing the concept to decide the existence of the msc. Therefore, in practice it is still necessary to provide an efficient algorithm to decide the existence of the msc without relying on computing the concept.

Through the thesis, we provide an algorithm for deciding the most specific concept and prove its correctness and completeness. We show that the algorithm decides the existence of the msc in a polynomial run-time. Also, we provide an experimental evaluation to state the difference between the previous and new approach in terms of the bounded role-depth of the concept.

1.1 Thesis Structure

We begin with the preliminaries for the notation and terminology used throughout the thesis in Chapter 2. Then we start with the first part, which is about the first attempt to decide the existence of the most specific concept in Chapter 3. In Chapter 4, we introduce the least tree unravelling as a new approach for tree unravelling of an interpretation and show the difference between the new approach for tree unravelling and the previous one. Furthermore, we continue with the least tree unravelling of an interpretation and employ it for deciding the existence of the most specific concept. At the same time, we introduce a representation of interpretations as a graph and define a simulation between graphs.

Chapter 5 constitutes the final part of this thesis, in which we present the implementation. In the second part of this chapter, we prove that the algorithm runs in polynomial time. We also prove the correctness and completeness of the introduced algorithm.

Finally, in Chapter 6 we show experimental results to show the difference between the newly introduced approach and the previous approach in terms of the size of the bounded role-depth of the concept and the time required to decide the existence of the msc.

To lead the future work, In the final Chapter 7, we summarize our findings and point to some future work.
2 Preliminaries

This chapter introduces the general notation and terminology we use throughout this thesis. It is based on [Baa+17b] with some extensions. Note that these preliminaries are for description logics in general. In contrast, the background of more specific topics (e.g. the tree unravelling and the $k$-characteristic concept) will be given later when needed.

The chapter starts with an introduction of the description logic $\mathcal{EL}$. Then it introduces the syntax and semantics of the DL $\mathcal{EL}$ and continues with the definition of canonical models and simulation relations between two interpretations. In the end, it introduces the formal definition of the most specific concept of an individual.

2.1 The Description Logic $\mathcal{EL}$

We start with defining the syntax of $\mathcal{EL}$ and the knowledge base (KB). Then we define the semantics of $\mathcal{EL}$. A knowledge base is defined with the help of a set of constructors, starting with a set $N_C$ of concept names, a set $N_R$ of role names and a set $N_I$ of individual names. The sets $N_C$, $N_R$ and $N_I$ are pairwise disjoint and countable infinite.

**Definition 2.1.1 ($\mathcal{EL}$-concept).** Let $A \in N_C$ and $r \in N_R$, $\mathcal{EL}$-concepts $C$ are then built according to the following syntax rule:

$$C ::= \top \mid A \mid C \cap D \mid \exists r.C$$

where $\top$ is the top concept, and $D$ is an $\mathcal{EL}$-concept.

To be able measure the complexity of concepts, we need a notion of role-depth of concepts.

**Definition 2.1.2 (role-depth).** Let $C$ be a $\mathcal{EL}$-concept. The role–depth of the concept $C$ ($rd(C)$) is inductively defined as follows:

- $rd(A) = rd(T) = 0$, for all $A \in C$.
- $rd(C \cap D) = max\{rd(C), rd(D)\}$, for all concepts $C, D$.
- $rd(\exists r.C) = 1 + rd(C)$, for all $r \in N_R$ and concepts $C$.

To illustrate this, consider the following example.

**Example 2.** Consider a concept $C := \exists r.(\exists s.A \cap \exists r.\exists s.T)$. Then $rd(C) = 3$.

**Definition 2.1.3 (General concept inclusion).** Let $C, D$ be $\mathcal{EL}$-concepts. A general concept inclusion (GCI) is an expression of the form $C \sqsubseteq D$. $C \equiv D$ is called a concept definition and it is an abbreviation for $C \sqsubseteq D, D \sqsubseteq C$. 
A terminological TBox $\mathcal{T}$ is a finite set of concept definitions. A general TBox $\mathcal{T}$ is a finite set of GCIs.

**Definition 2.1.4 (Acyclic TBox).** ([Baa+17a]) Let $\mathcal{T}$ be a TBox. We call $\mathcal{T}$ is an acyclic TBox if one of the following hold:

- There is no concept name in $\mathcal{T}$ that uses itself, and
- no concept name occurs more than once on the left-hand side of a concept definition in $\mathcal{T}$.

Otherwise, $\mathcal{T}$ is cyclic.

**Definition 2.1.5 (Assertion).** Let $a, b \in N_l, r \in N_R$ and $C \in N_C$, then:

- $C(a)$ is called a concept assertion.
- $r(a, b)$ is called role assertion. $a$ is called $r$-predecessor of $b$, and $b$ is called $r$-successor of $a$.

An ABox $\mathcal{A}$ is a finite set of concept and role assertions. The set $N_{l,\mathcal{A}}$ denotes the set of individual names occurring in ABox $\mathcal{A}$.

**Definition 2.1.6 (Acyclic ABox).** Let $\mathcal{A}$ be an ABox. We call $\mathcal{A}$ is an acyclic ABox if there exists no individual $a$ such that $a$ occurs as $r$-successor of $b$ and as $r'$-predecessor of $c$, for all $r, r' \in N_R$ and $b, c \in N_{l,\mathcal{A}}$. Otherwise, $\mathcal{A}$ is cyclic.

A knowledge base $\mathcal{K}$ consists of a TBox and an ABox $(\mathcal{K} = (\mathcal{T}, \mathcal{A}))$. The set $\text{sub}(X)$ denotes the set of sub-concepts occurring in $X$, where $X$ is a concept, ABox, TBox or a knowledge base. The set $N_{R,\mathcal{K}}$ denotes the set of role names occurring in $\mathcal{K}$.

The semantics of $\mathcal{EL}$-concepts are defined in terms of interpretations $\mathcal{I} = (\Delta^I, \mathcal{I})$ [Baa+07].

**Definition 2.1.7 (Interpretation).** An interpretation $\mathcal{I} = (\Delta^I, \mathcal{I})$ consists of a non-empty set $\Delta^I$ of individuals and a mapping $\mathcal{I}$ that maps

- every concept $A \in N_C$ to a set $A^\mathcal{I} \subseteq \Delta^I$.
- every role $r \in N_R$ to a binary relation $r^\mathcal{I} \subseteq \Delta^I \times \Delta^I$.
- every individual name $a \in N_l$ to an element $a^\mathcal{I} \in \Delta^I$.

Given an interpretation $\mathcal{I}$, both syntax and semantics of $\mathcal{EL}$-concepts, TBox and ABox are summarized in Table 2.1. An interpretation $\mathcal{I}$ is said to satisfy a GCI $C \sqsubseteq D$ if $C^\mathcal{I} \subseteq D^\mathcal{I}$. An interpretation $\mathcal{I}$ is called a model of a TBox $\mathcal{T}$ if it satisfies all GCIs in $\mathcal{T}$. Let $C(a)$ be a concept assertion. Then an interpretation $\mathcal{I}$ is said to satisfy $C(a)$ if $a^\mathcal{I} \in C^\mathcal{I}$. Let $r(a, b)$ be a role assertion. Then an interpretation $\mathcal{I}$ is said to satisfy $r(a, b)$ if $(a^\mathcal{I}, b^\mathcal{I}) \in r^\mathcal{I}$. An interpretation $\mathcal{I}$ is called a model of an ABox $\mathcal{A}$ if it satisfies all concept and role assertions. An interpretation $\mathcal{I}$ is said to be a model of a knowledge base $\mathcal{K}$ if it satisfies both $\mathcal{T}$ and $\mathcal{A}$. Since we are only considering $\mathcal{EL}$, we sometimes write 'concept' and 'TBox' instead of $\mathcal{EL}$-concept' and 'EL-TBox', respectively.

**Subsumption and instance checking** are considered essential reasoning tasks that are generalized to computing the most specific concept of an individual. These two reasoning tasks in $\mathcal{EL}$ can be decided in polynomial time [BBL05].

**Definition 2.1.8 (Instance checking and subsumption).** Let $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ be a knowledge base, let $C, D$ be a $\mathcal{EL}$-concepts and $a$ an individual name that occurs in $\mathcal{A}$, then:

- $a$ is an instance of $C$ w.r.t. $\mathcal{J} (\mathcal{A} \models \mathcal{J} C(a))$ iff $a^\mathcal{J} \in C^\mathcal{J}$ holds for all models $\mathcal{J}$ of $\mathcal{K}$.
- $C$ is subsumed by $D$ w.r.t. $\mathcal{J} (\mathcal{C} \sqsubseteq \mathcal{D})$ iff $C^\mathcal{J} \subseteq D^\mathcal{J}$ holds for all models $\mathcal{J}$ of $\mathcal{J}$. 
2 Preliminaries

Table 2.1 The syntax and semantics of $\mathcal{EL}$-concept, TBox and ABox w.r.t. an interpretation $\mathcal{I}$.

<table>
<thead>
<tr>
<th>name of constructor</th>
<th>Syntax</th>
<th>Semantics</th>
</tr>
</thead>
<tbody>
<tr>
<td>concept name $A \in N_C$</td>
<td>$A$</td>
<td>$A^I \subseteq A^0$</td>
</tr>
<tr>
<td>role name $r \in N_R$</td>
<td>$r$</td>
<td>$r^I \subseteq A^I \times A^I$</td>
</tr>
<tr>
<td>top-concept</td>
<td>$\top$</td>
<td>$A^I$</td>
</tr>
<tr>
<td>conjunction</td>
<td>$C \sqcap D$</td>
<td>$C^I \sqcap D^I$</td>
</tr>
<tr>
<td>existential restriction</td>
<td>$\exists r.C { x \in A^I \mid \exists y : (x,y) \in r^I \land y \in C^I }$</td>
<td></td>
</tr>
<tr>
<td>GCI</td>
<td>$C \sqsubseteq D$</td>
<td>$C^I \subseteq D^I$</td>
</tr>
<tr>
<td>concept definition</td>
<td>$C \equiv D$</td>
<td>$C^I = D^I$</td>
</tr>
<tr>
<td>individual name</td>
<td>$a$</td>
<td>$a^I \in A^I$</td>
</tr>
<tr>
<td>concept assertion</td>
<td>$A(a)$</td>
<td>$a^I \in A^I$</td>
</tr>
<tr>
<td>role assertion</td>
<td>$r(a,b)$</td>
<td>$(a^I, b^I) \in r^I$</td>
</tr>
</tbody>
</table>

2.2 The Most Specific Concept

Roughly speaking, the most specific concept of a given individual $a$ is a generalization of $a$ into a complex concept such that the concept is the most specific one of which an individual is an instance. The most specific concept of an individual $a$ is defined based on subsumption and instance checking.

Definition 2.2.1 (Most specific concept). Let $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ be a knowledge base and $a \in N_{IA}$ be an individual name. A concept $C$ is the most specific concept of $a$ w.r.t. $\mathcal{K}$ if the following statements hold:

- $\mathcal{K} \models C(a)$.
- $\mathcal{K} \models D(a)$ implies $C \subseteq_T D$.

As explained in [Baa03a] the msc does not always exist in $\mathcal{EL}$, in case we have a cyclic $\mathcal{A}$. However, if there is no msc, then one can approximate the msc concept by adding a bounded role-depth $k$ of the concepts $C$ and $D$. Then concept $C$ is called a role-depth bounded msc of $a$ w.r.t. $\mathcal{K}(k$-msc$_{\mathcal{K}}(a))$. Both msc and $k$-msc are unique up to equivalence in $\mathcal{EL}$.

2.3 The Canonical Model and The Simulation Relation

The computation of the msc is based on the characterizations of instance checking and subsumption. Thus, such characterizations w.r.t. general TBoxes was introduced in [LW10b] by employing both canonical models and simulation relations. A homomorphism between the syntax trees of concepts was used to characterize subsumptions in the knowledge base with an empty TBox [BKM98]. Also, simulation relation plays the same role as homomorphism, and if applied on the canonical models, then it can also be used to characterize subsumptions. As shown in [ZT13], the decision of the existence of msc of an individual $a$ w.r.t. $\mathcal{K}$ requires establishing a simulation relation between the canonical model and the constructed concept up to some bound $k$. Therefore, we will introduce the canonical model and simulation that we will use to decide the existence of msc.

Definition 2.3.1 (Canonical model of a concept). Let $\mathcal{T}$ be a TBox, $\mathcal{I}$ a model of $\mathcal{T}$, and $C$ a concept. The canonical model $\mathcal{I}_{C,\mathcal{T}}$ of $C$ and $\mathcal{T}$ is defined as follows:

- $A^{\mathcal{I}_{C,\mathcal{T}}} := \{ d_C \} \cup \{ d_{C'} \mid \exists r.C' \in sub(C) \cup sub(\mathcal{T}) \}$. 

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The canonical model

Let $C$ be a concept and $d$ as shown in [LW10a]. If $d \leq C$, then $d \vdash C$, for all $r \in N_R$. The canonical model $\mathcal{I}$ of $\mathcal{K}$ is defined as follows:

- $A^d_C := \{ d \in A \mid d \subseteq C \}$, for all $A \subseteq C$.
- $r^d_C := \{ (d, d') \mid d' \subseteq C \}$, for all $r \in N_R$.

This definition can be extended to be the canonical model of a knowledge base.

Definition 2.3.2 (Canonical model of a knowledge base). Let $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ be a knowledge base. The canonical model $\mathcal{I}_\mathcal{K}$ of $\mathcal{K}$ is defined as follows:

- $A^\mathcal{K} := \{ d \in A \mid d \subseteq \mathcal{N}_\mathcal{A} \} \cup \{ C \mid \mathcal{A}(a) \}$.
- $A^\mathcal{K} := \{ d \in A^\mathcal{K} \mid \mathcal{K}(a) = A(a) \} \cup \{ C \mid C \subseteq A \}$, for all $A \subseteq C$.
- $r^\mathcal{K} := \{ (d, d') \mid C \subseteq D' \subseteq \mathcal{N}_\mathcal{A} \}$, for all $r \in N_R$.
- $\sigma^\mathcal{K} := d_0$, for all $a \in \mathcal{N}_\mathcal{A}$.

Let $\mathcal{I}$ be an interpretation, $d \in A$. Then, $(\mathcal{I}, d)$ is said to be a pointed interpretation, where $\mathcal{I}$ is rooted on $d$. The simulation relations are defined between pointed interpretations.

Definition 2.3.3 (Simulation). Let $(\mathcal{I}_1, d)$ and $(\mathcal{I}_2, d')$ be interpretations. The binary relation $S \subseteq A^\mathcal{K} \times A^\mathcal{K}$ is called a simulation from $\mathcal{I}_1$ to $\mathcal{I}_2$ if all the following statements hold:

- $S_1$: For every pair $(e_1, e_2) \in S$ it holds: $e_1 \in A^\mathcal{K}$ implies $e_2 \in A^\mathcal{K}$, for all $A \subseteq C$.
- $S_2$: if $(e_1, e_2) \in S$ and $(e_1, f_1) \in R^d$, then there exists some $f_2 \in A^\mathcal{K}$ such that $f_1, f_2) \in S$ and $(e_2, f_2) \in R^d$.

A pointed interpretation $(\mathcal{I}_1, d)$ is said to be simulated by a pointed interpretation $(\mathcal{I}_2, d')$ (written as $(\mathcal{I}_1, d) \leq (\mathcal{I}_2, d')$) if there exists a simulation $S \subseteq A^\mathcal{K} \times A^\mathcal{K}$ with $(d, d') \subseteq S$. Note that the relation $\leq$ is transitive and reflexive. A pointed interpretation $(\mathcal{I}_1, d)$ is said to be simulation equivalent to a pointed interpretation $(\mathcal{I}_2, d')$ (written as $(\mathcal{I}_1, d) \equiv (\mathcal{I}_2, d')$) if both $(\mathcal{I}_1, d) \leq (\mathcal{I}_2, d')$ and $(\mathcal{I}_2, d') \leq (\mathcal{I}_1, d)$ hold.

As shown [LW10b; ZT13], the canonical model has essential properties.

Lemma 2.3.4. Let $C$ be a concept and $\mathcal{T}$ be a TBox.

1. $d \cap E^{\mathcal{K}}$ for all $d \subseteq C$.
2. $\mathcal{I}^\mathcal{T}$ is a model of $\mathcal{T}$.
3. $(\mathcal{I}^\mathcal{T}, d_0) \simeq (\mathcal{I}^\mathcal{T}, d_0)$, for all concepts $C'$ and all $d_0 \in E^{\mathcal{K}} \cap E^{\mathcal{K}}$.
4. For all models $\mathcal{I}$ of $\mathcal{T}$ and all $d \subseteq A$, the following statements are equivalent:
   - $d \subseteq C$.
   - $(\mathcal{I}^\mathcal{T}, d) \leq (\mathcal{I}, d)$.
5. For every concept $C$ and $D$, the following statements are equivalent:
   - $C \subseteq D$.
   - $(\mathcal{I}^\mathcal{T}, d) \leq (\mathcal{I}^\mathcal{T}, d)$.

Case (5) of the previous lemma allows us to give a characterization of the subsumption task, as shown in [LW10a]. If $C \subseteq D$, then $C \leq D$, for all models $\mathcal{I}$ of $\mathcal{T}$. Hence, $C \leq D$, and $d \subseteq D$. Moreover, we have that $(\mathcal{I}^\mathcal{T}, d_0) \leq (\mathcal{I}^\mathcal{T}, d_0)$. Similarly, in order to characterize the instance problem, consider the following lemma.

Lemma 2.3.5. Let $\mathcal{K}$ be a knowledge base. $\mathcal{I}_\mathcal{K}$ has the following properties:

1. $\mathcal{I}_\mathcal{K}$ is a model of $\mathcal{K}$.
2. For every concept $C$, the following statements are equivalent:
2 Preliminaries

\begin{itemize}
  \item $\mathcal{K} \models C(\alpha)$.
  \item $d_\alpha \in C^\mathcal{K}$.
\end{itemize}

By definition of 2.1.8, we have that $\alpha$ is an instance of $C$ w.r.t. $\mathfrak{I}$ $(\mathfrak{I} \models_{\mathfrak{I}} C(\alpha))$ iff $d_\alpha \in C^\mathfrak{I}$ holds for all models $\mathfrak{I}$ of $\mathcal{K}$. Therefore, if $\mathfrak{I}_\mathcal{K}$ is the canonical model of $\mathcal{K}$, then for every concept $C$ we have that $\mathcal{K} \models C(\alpha)$ iff $d_\alpha \in C^\mathcal{K}$.
3 The Previous Approach for Computing The Msc

In this chapter, we revisit the previous approach [ZT13] for computing the msc. We introduce the tree unravelling of an interpretation \( I \). Then we present the construction of a concept from tree unravelling up to some bound \( k \) and show that it is sufficient to generate candidate concepts up to upper bound \( k \). Then we show that an msc exists iff there exists a simulation equivalent relation between the pointed interpretation \((I_K, d_a)\) w.r.t. an individual \( a \) and the canonical model of one of the generated concepts.

3.1 The Tree Unravelling

Every interpretation \( I \) can be represented as a tree by taking an individual \( d \) as a root node and using as nodes all paths \( \sigma = d r_1 d_1 r_2 d_2 r_3 ... \), where the nodes \{\( d, d_1, d_2, ... \)\} are elements from \( \Delta^I \) and \{\( r_1, r_2, ... \)\} are role names. (\( \sigma \) is a path in interpretation \( I \) if the individuals \( d_i \) and \( d_i+1 \) are connected via \( r_i+1 \), for all \( i \in \mathbb{N} \)). The nodes of the tree correspond to individual \( d_i \) in the path \( \sigma \), and the edges of the tree correspond to role \( r_i \) in the path \( \sigma \).

Definition 3.1.1 (Tree unravelling of an interpretation). Let \( I \) be an interpretation and \( d \in \Delta^I \) w.r.t. \( N_C \) and \( N_R \). The tree unravelling \( I_d \) of \( I \) in \( d \) is defined as follows:

\[
\begin{align*}
\Delta^I_d &= \{ dr_1 d_1 r_2 d_2 r_3 ... d_n | (d_i, d_{i+1}) \in r_{i+1}^I \land 0 \leq i < n \land d_0 = d \}.
\end{align*}
\]

\[
\begin{align*}
A^I_d &= \{ \sigma d' | \sigma d' \in \Delta^I_d \land d' \in A^I \}, \text{ for all } A \in N_C.
\end{align*}
\]

\[
\begin{align*}
r^I_d &= \{ (\sigma, \sigma' d') | (\sigma, \sigma' d') \in \Delta^I_d \times \Delta^I_d \}, \text{ for all } r \in N_R.
\end{align*}
\]

\(|\sigma|\) denotes the number of role names occurring in \( \sigma \), where \( \sigma \in \Delta^I \). Given \( \sigma = dr_1 d_1 r_2 d_2 r_3 ... d_n \), \( \text{tail}(\sigma) = d_n \) is the last individual occurring in \( \sigma \) (also called the tail of \( \sigma \)).

Example 3. Consider The following knowledge base \( \mathcal{K} \):
\[
\mathcal{K} = (\emptyset, A), \text{ with } A = \{ r(a, b), s(a, b), r(a, c), r(c, b) \}.
\]

Now we consider the pointed interpretation \((I_K, d_a)\) as shown in Figure 3.1a. We can see in Figure 3.1b that \((I_K, d_a)\) is unravelled into a finite tree. However, if the interpretation contains a loop, then the tree unravelling of the interpretation is infinite.

An interpretation \( I \) is unravelled into an infinite tree in case it contains a loop (cyclic path). Since the concept can only be constructed from a finite tree, therefore it is essential to unravel the tree up to depth \( k \). \( I^I_{d_a} \) denotes the finite sub-tree with root \( d_a \) of the tree unravelling \( I_d \) containing all elements up to depth \( k \).
We recall some of the essential lemmas that are obtained from [ZT13]. First, we start with an

### 3.3 Characterizing The Existence of The Msc

We recall some of the essential lemmas that are obtained from [ZT13]. First, we start with an

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Figure 3.2 The interpretation \((I_{X_1}, d_a)\) with two loops, the tree unravelling \(I_{d_a}\) of \((I_{X_1}, d_a)\) and the tree unravelling \(I_{d_a}^2\) constructed w.r.t. Example 4.

Lemma 3.3.1. ([ZT13]) Let \(I_{d_a}\) be the tree unravelling of \((I_X, d_a)\) in \(d_a\) and \(C\) be \(k\)-msc\(X_1\)(a). It holds that \(I_{d_a}^k \subsetneq (I_{C,T}, d_C)\).

To illustrate the previous lemma, if the concept \(C\) is a \(k\)-msc\(X_1\)(a), then the canonical model of \(C\) simulates the tree unravelling up to \(k\).

The decision of whether the concept \(C\) is an msc or not can be obtained from the following lemma.

Lemma 3.3.2. ([ZT13]) The concept \(C\) is the most specific concept of \(a\) w.r.t. \(K\) iff \((I_X, d_a) \simeq (I_C, d_C)\).

Since the concept \(C\) can be constructed from a finite tree interpretation, it is required to define an upper bound to which the interpretation is unravelled. The upper bound is defined as follows.

Lemma 3.3.3. ([ZT13]) Let \(m := \max\{ \text{rd}(F) \mid F \in \text{sub}(K)\}\) and \(n := |N_{I_A}|\). If msc\(X_1\)(a) exists, then \((I_X, d_a) \subsetneq I_{d_a}^{n^2 + m + 1}\).

We are not only interested in computing the \(k\)-msc\(X_1\)(a) with some bound \(k\), but also in deciding the existence of msc\(X_1\)(a) for all \(k \in \mathbb{N}\) and compute msc\(X_1\)(a) if it exists. This can be characterized by the following lemma.

Lemma 3.3.4. ([ZT13]) Let \(K\) be a knowledge base. msc\(X_1\)(a) exists iff there exists a \(k\) such that the canonical model of \(X^k(I_X, d_a)\) w.r.t. \(T\) simulates \((I_X, d_a)\).

To decide the existence of the msc of an individual \(a\) w.r.t. \(K\), first we compute the upper bound \(k\) w.r.t. Lemma 3.3.3. Then the set \(M\) of \(L\)-characteristic concepts of \((I_X, d_a)\) such that \(M = \{X^L(I_X, d_a) \mid L \leq k\}\). Finally, for each concept \(C \in M\), \(C\) is the msc iff the canonical model of concept \(C\) w.r.t. \(T\) simulates \((I_X, d_a)\) w.r.t. the individual \(a\) \((I_X, d_a) \subsetneq (I_{C,T}, d_C)\). We illustrate this with the following example.
3 The Previous Approach for Computing The Msc

Figure 3.3 The interpretation $(I_K, d_a)$, the simulation relation from $(I_K, d_a)$ to $(I_C, T, d_C)$, where $C = X^0(I_K, d_a)$ w.r.t. Example 5.

Example 5. Consider the following knowledge base $\mathcal{K} = (\mathcal{T}, \mathcal{A})$:

- $\mathcal{T} = \{C \sqsubseteq \exists s.D, D \sqsubseteq \exists s.C\}$.
- $\mathcal{A} = \{s(a, b), s(b, a), C(a), D(b)\}$.

In Figure 3.3 it is sufficient to generate concept with $k = 0$, as the canonical model of the concept $X^0(I_K, d_a)$ w.r.t. $\mathcal{T}$ simulates $(I_K, d_a)$. Thus $X^0(I_K, d_a)$ is msc$_\mathcal{K}(a)$. Since computing the concept $C$ could be done in exponential time in case the concept $C$ is exponentially large, therefore in practice deciding the existence of the msc with the previous method could be done in exponential time.

In the next chapter, we introduce a new method for unravelling an interpretation. Then we employ this method to provide a more efficient way to decide the existence of the msc in a polynomial-time algorithm, which yields an efficient computation of the msc if it exists.
4 Towards Polynomial Computation

In this chapter, we revisit the sufficient condition from [Baa03a] for characterizing the existence of an msc in the terminological TBoxes. Then we introduce a new method for a tree unravelling of an interpretation \( I \). Moreover, we show the difference between the previous and the new method for the unravelling of an interpretation \( I \). Moreover, we show that the newly introduced method for tree unravelling has a tighter upper bound. Thus, the characteristic concept can be constructed w.r.t. this new method. Also, we use this newly introduced method in a polynomial approach for deciding the existence of the msc. Furthermore, we show the correctness and completeness of this Algorithm. Finally, we show that it decides the existence of the msc w.r.t. an individual in polynomial time.

The most specific concept always exists if the ABox is acyclic. However, it need not to exist in the case of cyclic ABox. As in Example 1 the most specific concept w.r.t. \( \alpha \) does not exist in \( \mathcal{K}_1 \), due to the fact that \( \mathcal{K}_1 \) is cyclic. However, it exists in \( \mathcal{K}_2 \) even that \( \mathcal{K}_2 \) is cyclic. One can observe that not all cyclic paths cause this problem. The result from [Baa03a], which was defined w.r.t. terminological cycles, shows that the msc exists iff every cyclic path in \( \mathcal{A} \) is simulated by at least one cyclic path in \( \mathcal{T} \). However, we cannot ensure that it is still holds in the case of EL-TBoxes. Intuitively, the goal is to extend this result to the case where we have a general EL-TBoxes.

In Lemma 3.3.2, a concept \( C \) is the msc of an individual \( \alpha \) iff there exists a simulation equivalence between \( (\mathcal{J}_\mathcal{K}, \Delta) \) and the canonical model of \( C \) w.r.t. \( \mathcal{T} \), where the concept \( C \) is constructed from the tree unravelling of \( (\mathcal{J}_\mathcal{K}, \Delta) \) up to some bound \( k \).

By Definition 2.3.2, if the interpretation \( \mathcal{J}_\mathcal{K} \) contains cyclic paths, then these cyclic paths might generated by \( \mathcal{T} \) or \( \mathcal{A} \). Hence, \( (\mathcal{J}_\mathcal{K}, \Delta) \) w.r.t. an individual \( \alpha \) contains all cyclic paths from which \( \alpha \) is reachable. Therefore, it is sufficient to check, whether \( (\mathcal{J}_\mathcal{K}, \Delta) \) contains cycles or not. These cycles in the interpretation lead to the tree unravelling growing infinitely, while the tree unravelling of the Definition 3.1.1 cannot distinguish the existence of cycles in the interpretation. One direction to overcome such a problem is to use an upper bound \( k \) to which the tree is unravelled. Another direction is to define a new method for unravelling, a so-called least tree unravelling.

4.1 The Least Tree Unravelling

Roughly speaking, a least tree unravelling of an interpretation \( \mathcal{J} \) is a tree unravelling, where each path \( \sigma = d_1 \cdots d_m \) contains only distinct individuals from \( \Delta^\mathcal{J} \). If the \( \Delta^\mathcal{J} \) is finite, then the tree constructed w.r.t. the least tree unravelling is also finite.

Definition 4.1.1 (Least tree unravelling of interpretation). Let \( \mathcal{J} \) be an interpretation and \( \Delta^\mathcal{J} \) w.r.t. \( \Delta_C \) and \( \Delta_R \). The least tree unravelling \( \mathcal{J}_d \) of \( I \) in \( d \) is defined as follows:
w.r.t. the least tree unravelling coincides with the characteristic concept $C$ will have only tree depth equal to 1 (see Figure 4.1).

Proof.\(\Rightarrow\) Let the tree unravelling. Thus, \((K, a, C) \simeq (C', a, C')\) and the concept $C'$ constructed w.r.t. the least tree unravelling coincides with the characteristic concept $C'$ constructed w.r.t. the tree unravelling. Therefore, the concept $C'$ constructed w.r.t. the least tree unravelling and the concept $C$ are identical. Moreover, \((C, a, C) \simeq (C, a, C')\). By Lemma 3.3.2, we obtain that \((C, a, C) \simeq (C, a, C)\) and the concept $C'$ is a msc$_K(a)$.

\(\Leftarrow\) The other direction can be proved analogously.
We still need to ensure whether Corollary 4.1.2 holds if \( \mathcal{A} \) is cyclic. In the next section, we will introduce terminology that is essential in showing that the concept constructed w.r.t. the least tree unravelling coincides with the concept constructed w.r.t the tree unravelling. In the next section, we introduce a mechanism to convert an interpretation into a so-called \( \mathcal{EL} \)-graph.

### 4.2 An \( \mathcal{EL} \)-Graph

We start by defining interpretations in terms of graphs. Furthermore, we define a mechanism to convert a graph into an interpretation. Thus, the obtained interpretation can be used to construct a characteristic concept w.r.t. the least tree unravelling.

**Definition 4.2.1 (\( \mathcal{EL} \)-Graph).** (From [Baa03a]) An \( \mathcal{EL} \)-graph is a directed graph \( G = (V, E, \text{Ind}, Y) \), such that:

- \( V \) is a set of nodes.
- \( E \subseteq V \times N_R \times V \) is a set of edges labeled by role names.
- \( Y : V \to 2^{NC} \) is a function that labels each node \( v \in V \) with sets of concepts \( L \subseteq NC \).
- \( \text{Ind} : V \to 2^N \) is a function that assigns each node \( v \in V \) to a set of individual names.

We denote the root node of \( G \) by \( v_0 \). A set \( pe(E, v) \) is the set of all edges with a node \( v \) as a predecessor.

**Definition 4.2.2 (Sub-graph).** Let \( G_1 = (V_1, E_1, \text{Ind}_1, Y_1) \) be an \( \mathcal{EL} \)-graph. The \( \mathcal{EL} \)-graph \( G_2 = (V_2, E_2, \text{Ind}_2, Y_2) \) is said to be a sub-graph of \( G_1 \) \((G_2 \subseteq G_1)\) iff the following statements hold:

- \( V_2 \subseteq V_1 \).
- \( E_2 \subseteq E_1 \) such that for every \( e = (v, r, d) \in E_2 \) we have \( v, d \in V_2 \).
- \( Y_2(v) \subseteq Y_1(v) \), for all \( v \in V_2 \).
- \( \text{Ind}_2(v) \subseteq \text{Ind}_1(v) \), for all \( v \in V_2 \).

The interpretation \( J \) can be translated into an \( \mathcal{EL} \)-graph by the following definition.

**Definition 4.2.3 (Interpretation as a graph).** (From [Baa03a]) Let \( J = (\Delta^3, \Delta^q) \) be an interpretation and \( d \) be an element in \( \Delta^q \). The graph \(^1\) \( G_J = (V_J, E_J, \text{Ind}_J, Y_J) \) of the interpretation \( J \) is defined as follows:

- \( V_J \) are the elements of \( \Delta^3 \), with \( v_0 = d \).
- \( E_J = \{(x, r, y) \mid r \in N_R, (x, y) \in r^3\} \).
- \( Y_J(v) = \{L \in NC \mid v \in L^3\} \), for all \( v \in \Delta^3 \).
- \( \text{Ind}_J(v) = \{a \in N_I \mid v \in a^3\} \), for all \( v \in \Delta^3 \).

Given a graph \( G = (V, E, \text{Ind}, Y) \), a node \( v' \in V \) is said to be directly reachable from a node \( v \in V \) (denoted by \( \text{reach}_{\text{direct}}(v, v') \)), if there exists a path \( \sigma \) from \( v \) to \( v' \) and \( |\sigma| = 1 \) (\(|\sigma| \) denotes the number of role names occurring in \( \sigma \)). More precisely, \( v' \in V \) is directly reachable from \( v \in V \) iff there exists an edge \( (v, r, v') \in E \). A node \( v' \in V \) is said to be reachable from a node \( v \in V \) (denoted by \( \text{reach}(v, v') \)), if there exists a path \( \sigma = v_{r1} \ldots v_{rn} v' \) from \( v \) to \( v' \).

A path \( \sigma = v_0 r_1 v_1 \ldots r_n v_n \) is said to be reachable from a node \( v \in V \), if the node \( v_0 \) is reachable from \( v \) (denoted by \( \text{reach}(v, \sigma) \)). We denote by \( \text{head}(\sigma) \) the first element of the path \( \sigma = v_0 r_1 v_1 \ldots r_n v_n \), while \( \text{tail}(\sigma) \) denotes the last element of the path \( \sigma = v_0 r_1 v_1 \ldots r_n v_n \).

**Proposition 4.2.4.** ([LPP12]) Let \( G = (V, E, \text{Ind}, Y) \) be an \( \mathcal{EL} \)-graph, \( v, v' \) nodes of \( V \). It is decidable in polynomial time whether \( v' \) is reachable from \( v \).

\(^1\) We write graph to refer to \( \mathcal{EL} \)-graph.
In order to construct a characteristic concept, it is essential to provide a mechanism to translate a graph $G$ into an interpretation $I_G$, which is defined as follows.

**Definition 4.2.5.** [Interpretation from graph] Let $G = (V, E, Ind, Y)$ be a graph, $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ a knowledge base. The interpretation $I_G$ w.r.t. $N_R, N_C$ and $N_{I_A}$ is defined as follows:

- $\Delta^G = V$.
- $\mathcal{A}^G = \{d_v \mid A \in Y(v) \land v \in V\}$, for all $A \in N_C$.
- $r^G = \{(d_v, d_{v'}) \mid (v, r, v') \in E\}$, for all $r \in N_R$.
- $\sigma^G = \{d_a \mid a \in Ind(v) \land v \in V\}$, for all $a \in N_{I_A}$.

As we are interested in characterizing the cycles of $I$, we need to distinguish between the cyclic paths and the acyclic paths of $G_I$. This can be obtained from the following definition.

**Definition 4.2.6** (Cyclic path). Let $G_0 = (V_0, E_0, Ind_0, Y_0)$ be a graph. A cyclic path $G_0$ is a path of $G_0$ such that:

- $G_0 = \{x_0r_1x_1...r_nx_n \mid (x_i, r_i, x_{i+1}) \in E_0 \land x_0 \in \{x_0, x_1, ..., x_{n-1}\} \land 0 \leq i < n \land n \leq |V_0|\}$.

$|G_0|$ denotes the number of role names occurring in $G_0$. The set $G_{cycles}$ denotes the set of all cyclic paths $G_0$ of $G_I$.

To be able to distinguish between the cycles of $I$ that are generated from $\mathcal{T}$ and $\mathcal{A}$, we need to construct sub-graphs $G'_{I_A}$ and $G''_{I_T}$ of $G_I$.

**Definition 4.2.7** (ABox graph). Let $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ be a knowledge base, $a$ an individual, $G_J = (V_J, E_J, Ind_J, Y_J)$ be the graph of $(I_X, d_0)$. The ABox graph $G_{I_A} = (V_{I_A}, E_{I_A}, Ind_{I_A}, Y_{I_A})$ is a sub-graph of $G_J$ such that:

- $V_{I_A} = \{v \in V_J \mid Ind(v) \in N_{I_A}\}$.
- $E_{I_A} = \{(x, r, y) \mid (x, y) \in r^{(X, d_0)} \land x, y \in V_{I_A}\}$.
- $Y_{I_A}(v) = \{L \in N_C \mid v \in L^{(X, d_0)}\}$, for all $v \in V_{I_A}$.
- $Ind_{I_A}(v) = Ind_J(v)$, for all $v \in V_{I_A}$.

$V_A$ denotes the set of nodes that are generated by $\mathcal{A}$.

The graph $G_{I_T}$ can be defined similarly.

**Definition 4.2.8** (TBox graph). Let $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ be a knowledge base, $a$ an individual, $G_J = (V_J, E_J, Ind_J, Y_J)$ be the graph of $(I_X, d_0)$. The TBox graph $G_{I_T} = (V_{I_T}, E_{I_T}, Ind_{I_T}, Y_{I_T})$ is a sub-graph of $G_J$ such that:

- $V_{I_T} = \{v \in V_J \mid Ind(v) \notin N_{I_A}\}$.
- $E_{I_T} = \{(x, r, y) \mid (x, y) \in r^{(X, d_0)} \land x, y \in V_{I_T}\}$.
- $Y_{I_T}(v) = \{L \in N_C \mid v \in L^{(X, d_0)}\}$, for all $v \in V_{I_T}$.
- $Ind_{I_T}(v) = Ind_J(v)$, for all $v \in V_{I_T}$.

$V_T$ denotes the set of nodes that are generated by $\mathcal{T}$.

We observe that $G_{I_A}$ and $G_{I_T}$ are unique w.r.t. the same model $(I_X, d_0)$. Moreover, it is clear to see that $G_{I_T}$ and $G_{I_A}$ are sub-graphs of $G_J$, where $G_{I_T}$ and $G_{I_A}$ are disjoint (see Figure 4.2).

From the fact that $G_{I_A}$ and $G_{I_T}$ are sub-graphs of $G_J$, where $G_J$ is the graph of $(I_X, d_0)$, one can observe that adding all paths $\sigma \in G_J$ that are reachable from any node $v$ in $G_{I_A}$ allows us to obtain back $G_J$.

**Definition 4.2.9** (Canonical graph). Let $G_1 = (V_1, E_1, Ind_1, Y_1)$ be the graph of $(I_X, d_0)$, $G_{I_A}$ the ABox graph of $G_1$, $G_2 = (V_2, E_2, Ind_2, Y_2)$ a sub-graph of $G_{I_A}$. We define as canonical graph of $G_2$ w.r.t. $G_1$ as the graph $G_{2, G_1} = (V_{2, G_1}, E_{2, G_1}, Ind_{2, G_1}, Y_{2, G_1})$ such that:
Example 6. Consider the following knowledge base $K = (\mathcal{T}, \mathcal{A})$: 

- $\mathcal{T} = \{C \subseteq \exists r.D\}$.
- $\mathcal{A} = \{s(a, b), C(a)\}$.

The previous definition adds only those nodes and edges that are generated by $G_{\mathcal{T}, \mathcal{A}}$ (see Figure 4.3). As a consequence of the previous definition, we obtain that the canonical graph $G_{\mathcal{T}, \mathcal{A}}$, of $G_{\mathcal{T}}$ w.r.t. $G_1$ gives us back $G_1$. Thus, we have that $(G_{\mathcal{T}, \mathcal{A}}, \mathcal{V}_0)$ is identical to $(G_1, \mathcal{V}_0)$. Moreover, if $\mathcal{A}$ is acyclic, then constructing a concept $C$ w.r.t. the least tree unravelling of $\mathcal{J}_{G_{\mathcal{T}, \mathcal{A}}}$, we have that by the Corollary 4.1.2 ($a_{\mathcal{C}, \mathcal{T}} \simeq (J_{\mathcal{X}, d_0})$).

From the definitions 4.2.2, 4.2.9 and 4.2.7, we can obtain the following Lemma that characterizes the relation between two sub-graphs of the ABox graph $G_{\mathcal{A}}$ constructed w.r.t. the same graph.

Lemma 4.2.10. Let $G = (V, E, Ind, Y)$ be the $\mathcal{E}\mathcal{L}$-graph of $(J_{\mathcal{X}, d_0})$, $G_1 = (V_1, E_1, Ind_1, Y_1), G_2 = (V_2, E_2, Ind_2, Y_2)$ be two sub-graphs of the ABox graph $G_{\mathcal{A}}$ such that $G_1 \subseteq G_2$. Then $G_{G_1, G} = (V_{G_1, G}, E_{G_1, G}, Ind_{G_1, G}, Y_{G_1, G}) \subseteq G_{G_2, G} = (V_{G_2, G}, E_{G_2, G}, Ind_{G_2, G}, Y_{G_2, G})$.

Proof. Assume that $G_1 \subseteq G_2$. Then by the Definition 4.2.2, we have the following:

- $V_1 \subseteq V_2$.
- $E_1 \subseteq E_2$.
- $Y_1(v) \subseteq Y_2(v)$, for all $v \in V_1$.
- $Ind_1(v) \subseteq Ind_2(v)$, for all $v \in V_1$.

Also, from the construction of the Definition 4.2.9, we have that the canonical graph $G_{G_1, G}$ is a sub-graph of $G$ such that:
4 Towards Polynomial Computation

The canonical graph $G_{G_1,G}$ is constructed analogously. Then by construction of the Definition 4.2.9 if there exists a node $v$ such that $v \in V_{G_1,G} \land v \notin V_1$, then $v$ must be reachable from a node $v' \in V_1$ where $v' \in V_1$. Since $V_1 \subseteq V_2$, then the node $v'$ is also in $V_2$, and $v$ must be in $G_{G_2,G}$ and $v \notin V_2$.

Similarly, if there exists an edge $e = (x,r,y) \in E_{G_1,G}$, then $e = (x,r,y) \in E_{G_2,G}$. Due to the fact that $x,y \in V_{G_1,G}$ and $x,y \in V_{G_2,G}$. From $V_{G_1,G} \subseteq V_{G_2,G}$ we have that $V_{G_1,G}$ is also a subset of $V_{G_2,G}$ ($Y_{G_1,G} \subseteq Y_{G_2,G}$).

We can also characterize the edges that are added in the canonical graphs of two sub-graphs of the ABox graph $G_A$ w.r.t. the same graph $G_2$. This can be characterized as follows.

Lemma 4.2.11. Let $G = (V,E,Ind,Y)$ be the EL-graph of $(I_3,d_3)$, $V_1$ be the set of nodes in $G$ that are generated by $T$, $G_1 = (V_1,E_1,Ind_1,Y_1)$, $G_2 = (V_2,E_2,Ind_2,Y_2)$ be two sub-graphs of the ABox graph $G_A$. For $v_1 \in V_{G_1,G},v_2 \in V_{G_2,G}$ if $Y_{G_1,G}(v_1) \subseteq Y_{G_2,G}(v_2)$, $(v_1,r,v) \in E_{G_1,G}$ and $v \in V_2$, then $(v_2,r,v) \in E_{G_2,G}$ where $G_{G_1,G} = (G_1,G_2,Ind_{G_1,G},Y_{G_1,G})$ and $G_{G_2,G} = (G_2,G_1,Ind_{G_2,G},Y_{G_2,G})$ are the canonical graphs of $G_1$ and $G_2$ w.r.t. $G$.

Proof. We assume that ontology is in normal form. An ontology is in a normal form if every existential restriction occurs in axioms of the following form:

\[
\begin{align*}
&\exists r.A \sqsubseteq B. \text{ Or,} \\
&A \sqsubseteq \exists r.B.
\end{align*}
\]

By construction of the Definition 4.2.9 we have that all nodes $v \in V_T$ that are added to $G_{G_1,G}$ must be reachable from a node $v_1 \in V_1$. Therefore, there must exists an edge $(v_1,r,v) \in E_{G_1,G}$ where $v \in V_T$. Also by construction of the Definition 2.3.2, the edge $(v_1,r,v) \in E_{G_1,G}$ exists in $G_{G_1,G}$ iff $(v_1,v) \in r^{I_3}$. Moreover, we can obtain the set of concept names $M$ such that every $C \in M$ we have $v_1 \in C^{I_3}$. Similarly, we can obtain the set of concepts $M'$ such that every $C' \in M'$ we have $v_2 \in C'^{I_3}$. If $Y_{G_1,G}(v_1) \subseteq Y_{G_2,G}(v_2)$, then $M \subseteq M'$. Hence, we have that every concept $C \in M$, we have that $v_2 \in C^{I_3}$. Hence, as a consequence of the construction of the Definition 2.3.2 we have that $v_2 \in r^{I_3}$ and $(v_1,v) \in r^{I_3}$, for the same role name $r$. Hence, the node $v \in V_T$ must be reachable from the node $v_2 \in V_{G_2,G}$. Therefore, there must exists an edge $(v_2,r,v) \in E_{G_2,G}$.

As graphs are used to represent the interpretations, then it is necessary to define the simulation explicitly in terms of graphs.

4.3 The Simulation Over Graphs

Definition 4.3.1. [Simulation over graphs] Let $G_1 = (V_1,E_1,Ind_1,Y_1)$, $G_2 = (V_2,E_2,Ind_2,Y_2)$ be EL-graphs, $v \in V_1$ be a node, $v' \in V_2$ be a node. A binary relation $S \subseteq V_1 \times V_2$ is called a simulation from $(G_1,v)$ to $(G_2,v')$ if all the following conditions are satisfied:

(S1) If $(v_1,v_2) \in S$, then $Y_1(v_1) \subseteq Y_2(v_2)$.

(S2) If $(v_1,v_2) \in S$, and $(v_1,r,v') \in E_1$, then there exists some $v'_2 \in V_2$ such that $(v'_1,v'_2) \in S$, and $(v_2,r,v'_2) \in E_2$. 

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We write \( S : (G_1, v) \preceq (G_2, v') \) to express that there exists a simulation \( S \) from \( G_1 \) to \( G_2 \) such that \( (v, v') \in S \). We write \( S : (G_1, v) \simeq (G_2, v') \) to express that there exists a simulation equivalence between \( G_1 \) and \( G_2 \) \((G_1, v) \preceq (G_2, v') \) and \( (G_2, v') \preceq (G_1, v) \).2

It is easy to see that the set of all simulations from \( G_1 \) to \( G_2 \) is closed under arbitrary unions, if \( G_1 \) and \( G_2 \) are finite. Consequently, there always exists a greatest simulation from \( G_1 \) to \( G_2 \). If \( G_1 \) and \( G_2 \) are finite, then this greatest simulation can be computed in polynomial time [HM01]. As a consequence of this fact, we obtain the following proposition.

**Proposition 4.3.2.** ([Baa03b]) Let \( G_1, G_2 \) be two finite \( \mathcal{E} \mathcal{L} \)-graphs, \( v \) a node of \( G_1 \) and \( v' \) a node of \( G_2 \). It is decidable in polynomial time whether there exists a simulation \( S : (G_1, v) \preceq (G_2, v') \) such that \( (v, v') \in S \).

**Definition 4.3.3** (Simulation over paths). Let \( G \) be a graph constructed w.r.t. an interpretation \( J \), \( \sigma, \sigma' \) be two paths occurring in \( G \). A path \( \sigma = x_0r_1... \) is said to be simulated by \( \sigma' = x'_0r'_1r'_2... \) iff there exists a simulation \( S \) such that \((x_i, x'_i) \in S \), for all \( 0 \leq i \) and \( x_0 = x'_0 \). We write \( S : \sigma \preceq \sigma' \) to express that there exists a simulation \( S \) from \( \sigma \) to \( \sigma' \) (see Figure 4.4).

**Definition 4.3.4** (Simulation over edges). Let \( G = (V, E, \text{Ind}, Y) \) be an \( \mathcal{E} \mathcal{L} \)-graph, \( e = (v, r, d) \), \( e' = (v', r, d') \) be two edges occurring in \( E \). \( e \) is said to be simulated by \( e' \) \((e \preceq e') \) if the following conditions are satisfied:

- \( Y(v) \subseteq Y(v') \).
- There exists a simulation \( S \) such that \((d, d') \in S \).

### 4.4 Computing The Msc w.r.t. The Least Tree Unravelling

Now we are ready to construct the most specific concept w.r.t. the least tree unravelling. Instead of generating all the candidate concepts up to upper bound \( k \) and check whether the simulation equivalence holds between the canonical model of one of the candidates w.r.t. \( \mathcal{T} \) and \((J_X, d_0)\), we employ the Lemmas 3.3.2 and 4.1.2 to decide the existence of the msc w.r.t. an individual.

Intuitively, we construct a sub-interpretation \((\sigma', d_0)\) of \((J_X, d_0)\) up to equivalence, such that \((\sigma', d_0) \preceq (J_X, d_0)\) and \((J_X, d_0) \preceq (\sigma', d_0)\). We construct a sub-graph \( \text{msc}G = (V_{\text{msc}G}, E_{\text{msc}G}, \text{Ind}_{\text{msc}G}, Y_{\text{msc}G}) \) of ABox graph \( G_A \) by recursively copying nodes and concepts related to them from the graph \( \text{can}M = (V, E, \text{Ind}, Y) \) of \((J_X, d_0)\) starting from the root node \( v_0 \). \( v_0 \) is the node that represents the individual \( a \). We define the construction of \( \text{msc}G \) in terms of an algorithm. The algorithms that we will be used in the following sections of this chapter, will be captured in Chapter 5.

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2We may just write \((G_1, v) \preceq (G_2, v') \) instead of \( S : (G_1, v) \preceq (G_2, v') \), if the simulation relation is not important in the context.
4.4.1 Deciding The Existence of The Msc

From Now we use the graph $mscG = (V_{mscG}, E_{mscG}, Ind_{mscG}, Y_{mscG})$ to represent the graph that we want to construct, the graph $canM = (V, E, Ind, Y)$ to represent the graph of $(\mathcal{K}, d_a)$ and the graph $temp_i = (V, E, Ind, Y_i)$ to be the canonical graph of $mscG$ w.r.t. $canM$ after the termination of the recursive call of the Algorithm 2.

We start by taking $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ and the individual $a$ as inputs. Then $canM$ is constructed w.r.t. $(\mathcal{K}, d_a)$.

We initialize $mscG_0 = (V_{mscG_0}, E_{mscG_0}, Ind_{mscG_0}, Y_{mscG_0})$ as follows:

- $V_{mscG_0} = \{v_0\}$, where $v_0$ is the root node of $canM$.
- $E_{mscG_0} = \emptyset$.
- $Y_{mscG_0}(v_0) = Y(v_0)$.
- $Ind_{mscG_0}(v_0) = a$.

Then, recursively we visit every node $v \in V_A$ (recall the Definition 4.2.7) and add the edges if needed. More precisely, for $0 \leq i \leq |V_A|$, we do the following:

- we build $temp_i = (V, E, Ind, Y_i)$ to be the canonical graph of $mscG_i$ w.r.t. $canM$.
- for each edge $(v, r, d) \in pe(E, v)$, we check whether there exists an edge $(v, r, d') \in pe(E_i, v)$ such that $(v, r, d)$ is simulated by $(v, r, d')$.

We can define $mscG_i = (V_{mscG_i}, E_{mscG_i}, Ind_{mscG_i}, Y_{mscG_i})$, where $i > 0$ as follows:

- $V_{mscG_i} = V_{mscG_{i-1}} \cup \{Ind(d) \in N_{IA} \mid \text{there exists } v \in V_{mscG_{i-1}} \text{ and for all } (v, r, d') \in pe(E_{i-1}, v) \text{ such that } (v, r, d) \in pe(E, v) \text{ w.r.t. } canM \land (v, r, d) \not\in (v, r, d') \}$.
- $E_{mscG_i} = \{(v, r, d) \in pe(E, v) \mid v, d \in V_{mscG_i}\}$.
- $Y_{mscG_i}(v) = Y(v)$, for all $v \in V_{mscG_i}$.
- $Ind_{mscG_i}(v) = Ind(v)$, for all $v \in V_{mscG_i}$.

In the previous definition, only nodes that are contained in $V_A$ are added to $V_{mscG}$. Since the number of nodes of the graph $G$ is finite, then also the size of $V_A$ is finite. Also, the number of edges in $E$ is finite. Therefore, we ensure that the Algorithm terminates.

After Algorithm terminates, we obtain that $mscG$ is a sub-graph of ABox graph $G_A$. The reason is that $mscG$ contains only nodes from $V_A$. Moreover, we observe that there exists always a simulation equivalence between $(temp_i, v_0)$ and $(canM, v_0)$. This can be characterized by the following lemmas.

**Lemma 4.4.1.** Let $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ be a knowledge base, $canM = (V, E, Ind, Y)$ be the graph of $(\mathcal{K}, d_a)$, and $mscG = (V_{mscG}, E_{mscG}, Ind_{mscG}, Y_{mscG})$ be a graph constructed after the termination of Algorithm 1. It holds that $(canM, v_0) \sim (temp_i, v_0)$, where $temp_i = (V, E_i, Ind_i, Y_i)$ is the canonical graph of $mscG$ w.r.t. $canM$.

**Proof.** Let $G_A = (V_A, E_A, Ind_A, Y_A)$ be an ABox graph constructed w.r.t. $canM$. Let $G_T = (V_T, E_T, Ind_T, Y_T)$ be a TBox graph constructed w.r.t. $canM$.

We start defining a relation $S_i \subseteq V \times V$ for $i \in \{0, 1, \ldots, |V_A|\}$, where $temp_i = (V, E_i, Ind_i, Y_i)$ is the canonical graph of $mscG$ w.r.t. $canM$ that is constructed at each recursive call. The relation $S_i$ is defined by following the steps of the Algorithm 1:

1. As shown in Line 4, the graph $mscG$ is initialized as an empty graph. Then the root node $v_0$ in $canM$ is added to $mscG_0$ together with its set of concepts. Therefore, $S_0 = \{(v_0, v_0)\}$, where $v_0, v'_0$ are the root nodes of $canM$ and $mscG_0$ respectively.
2. Then, all nodes in $V_A$ are visited only once, starting from the root node $v_0$ that contains the individual $a$ (as shown in Algorithm 2):
For every
Assume
We prove by induction over i.

Claim:

Base Induction: \( i = 0 \)

Assume \((\mathcal{V}, E, \text{Ind}_i, Y)\) is the canonical graph of \( \text{mscG}_i \) w.r.t. \( \text{canM} \), where \( \text{mscG}_i \) is the graph that is obtained at the recursive call \( i \).

Given a node \( v \), a set \( \text{pe}(E, v) = \{(v, r, v') \in E\} \) is the set of edges from \( \text{canM} \) that contains \( v \) as a predecessor, a set \( \text{pe}(E_i, v) = \{(v, r, v') \in E_i\} \) is the set of edges from \( \text{temp} \) that contains \( v \) as a predecessor.

An edge \( e \in \text{pe}(E, v) \) is added to \( \text{mscG}_i \) if there exists no edge \( e' \in \text{pe}(E_i, v) \) such that \( e \) is simulated by \( e' \) (see Line 10 of Algorithm 2).

An edge \( e = (v, r, v') \in E \) is said to be simulated by an edge \( e' = (d, r, d') \in E_i \) iff the following hold:

\[- Y(v) \subseteq Y(d) \text{ and } Y(v') \subseteq Y(d'). \]
\[- \text{There exists a simulation } S \text{ such that } (v', d') \in S. \]

Then we have one of two cases:

(A) There exists an edge \( e' = (v, r, d') \in \text{pe}(E_i, v) \) that simulates the edge \( e = (v, r, v') \in \text{pe}(E, v) \) and \( e' \in V_{Y} \), then there exists a simulation \( S \) such that \( (v', d') \in S \). We then have that \( S_i = S_{i-1} \cup S \).

(B) Otherwise, \( e = (v, r, v') \in \text{pe}(E, v) \) is added to \( \text{mscG}_i \). Hence, \( S_i = S_{i-1} \cup \{(v', v')\} \).

3. Algorithm 2 terminates after all nodes are visited and for every edge \( e \in E \), there is an edge \( e' \in E_i \) such that \( e \) is simulated by \( e' \).

As a consequence of Lemma 4.2.10, we have that \( \text{temp}_i \subseteq \text{temp}_{i+1} \), for all \( 0 \leq i < |V_A| \). Therefore, if there exists a simulation from a node \( v \in V \) to a node \( v' \in \text{temp}_i \), then there exists also a simulation from \( v \in V \) to \( v' \in \text{temp}_{i+1} \), for all \( 0 \leq i < |V_A| \). Let \( \text{temp}_i = (\mathcal{V}, E_i, \text{Ind}_i, Y_i) \) be the canonical graph of \( \text{mscG} \) w.r.t. \( \text{canM} \) that is constructed after the Algorithm 2 terminates.

Claim: For every \( i \in \{0, \ldots, |V_A|\} \), \( S_i \) satisfies:

(S'1) For every \( (v, v') \in S_i \), \( Y(v) \subseteq Y(v') \).

(S'2) If \( i > 0 \): for every \( (v, v') \in S_{i-1} \) and \( (v, r, e) \in E \), there exists \( (v', r, e') \in E_i \) \( (e, e') \in S_i \).

Proof of Claim: We prove by induction over \( i \).

Base Induction: \( i = 0 \). Then, \( S_0 = (v_0, v_0) \), where \( v_0, v_0 \) are the root nodes of \( \text{canM} \) and \( \text{temp}_0 \).

The Algorithm 1 copies \( v_0 \) to \( \text{mscG} \) together with its concepts. Therefore, (A) is satisfied. (B) is also satisfied, because \( i = 0 \).

Induction Step: Assume \( i > 0 \) and that claim holds for \( S_{i-1} \).

(S'1) Take some \( (v, v') \in S_i \). If \( (v, v') \in S_{i-1} \), then (S'1) holds from the induction hypothesis.

Assume \( (v, v') \in S_i \). This means we are in one of two cases:

(I) From (A) we have that \( S_i = S_{i-1} \cup S \) and \( (v, v') \in S \), or

(II) From (B) we have that \( S_i = S_{i-1} \cup \{(v, v')\} \), and \( Y(v) = Y_i(v) \).

In case (I), \( S \) is a simulation, and thus \( Y(v) \subseteq Y_i(v') \) by (S'1). In case (II), \( Y(v) = Y_i(v) \) follows directly.

(S'2) Take some \( (v, v') \in S_{i-1} \). If \( (v, v') \in S_{i-2} \), then (S'2) holds from the induction hypothesis.

Assume \( (v, v') \in S_{i-1} \). Then there exists an edge \( (v, r, d) \in E \). Then there are two cases depending on whether \( (v, r, d) \in E \) is simulated by some \( (v', r, d') \in E_i \) or not:

(I) From (A) we have that \( S_i = S_{i-1} \cup S \) where \( (d, d') \in S \) and \( (v', r, d') \in E_i \). Or

(II) From (B) we have that \( S_i = S_{i-1} \cup \{(d, d)\} \), and \( (v', r, d) \in E_i \).

In both cases, we have that (S'2) is satisfied.
We write $S_t$ to denote the relation $S_t \subseteq V \times V_t$, where $t = |V_A|$.

From the claim we have that the relation $S_t \subseteq V \times V_t$ defines a simulation from $canM = (V,E,Ind,Y)$ to $temp_t = (V_t,E_t,Ind_t,Y_t)$ as follows:

- Take $(v,v') \in S_t \setminus S_{t-1}, (v,r,e) \in E$. Then $e$ must be from $S_t$ because we already processed all successors of nodes. We have one of two cases.

  (I) $e$ was not processed in the last step but in an earlier step. That means $S_t \subseteq S_{t-1}$, where $(e,e') \in S_t$. The Algorithm 2 would add $(v',r,e')$ to $E_{mscG}$ and $(e,e') \in S_t$, unless it is already there. Since this is the last step of the Algorithm 2, then $(v',r,e')$ must be already in $E_{mscG}$ and thus also in $E_t$. It follows that (S2) is also satisfied for $(v,v')$.

  (II) $e$ was processed in the last step. Because we always process one node in each step, then $e = v$. This means that $(v, r, v') \in E$. If $(v', r', e') \in E_t$ such that $(canM, v) \subseteq (temp_{t-1}, e')$ through the simulation $S$, then $S_t = S_{t-1} \cup S$ and thus $(v, e') \in S_t$. Then also (S2) is satisfied for $(v, v')$. Otherwise, $E_t = E_{t-1} \cup \{(v', r, v)\}$ and $S_t = S_{t-1} \cup \{(v', v)\}$ so that (S2) is also satisfied.

Hence, the relation $S_t \subseteq V \times V_t$ defines a simulation from $canM$ to $temp_t$, where $temp_t$ is the canonical graph of $mscG$ w.r.t. $canM$ that is obtained after Algorithm 2 terminates.

\[ \square \]

**Lemma 4.4.2.** Let $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ be a knowledge base, graph $canM = (V,E,Ind,Y)$ be the graph of $(\mathcal{J}_K, d_0)$, and graph $mscG = (V_{mscG}, E_{mscG}, Ind_{mscG}, Y_{mscG})$ be the graph obtained after the termination of Algorithm 1. It holds that $(canM, v_0) \subseteq (temp_t, v_0)$, where $temp_t = (V_t,E_t,Ind_t,Y_t)$ is the canonical graph of $mscG$ w.r.t. $canM$.

**Proof.** Again, after the recursive call of the Algorithm 2, The $mscG$ graph is obtained such that the $mscG$ graph is a sub-graph of $canM$, and also $J_{mscG} \subseteq J_{canM}$ (recall the Definition 4.2.5). As consequence also the canonical graph $temp = (V_t, E_t, Ind_t, Y_t)$ of $mscG$ w.r.t. $canM$ is a sub-graph of the $canM$ graph. Hence, also $J_{temp} \subseteq J_{canM}$. Therefore, there exists a simulation from $temp$ to $canM$, due to that each node $v \in V_t$ has a corresponding node $v' \in V$ with $Y_t(v) \subseteq Y(v')$ and each edge $(v_1, r, v_2) \in E_t$ has a corresponding edge $(v_1', r, v_2') \in E$ with $Y_t(v_1) \subseteq Y(v_1')$ and $Y_t(v_2) \subseteq Y(v_2')$. Hence, it holds that $(temp, v_0) \subseteq (canM, v_0)$ after the Algorithm 1 terminates.

Since there exists always a simulation equivalence between $(temp, v_0)$ and $(canM, v_0)$ after Algorithm 1 terminates and the fact that $mscG$ contains only nodes from $V_A$, it is not a sufficient condition to decide the existence of the msc w.r.t. $\mathcal{A}$. It is still essential to check whether $mscG$ contains a cycle or not. If there exists a cyclic path $\sigma$ in $mscG$, then $\sigma$ is not simulated by any path from $temp_{t'}$. This allows us to obtain that if $mscG$ is cyclic, then this cyclic path is generated by $\mathcal{A}$. Therefore, there exists no msc w.r.t. $\mathcal{A}$. The reason is that $J$ that is obtained from $mscG$ would contain a loop. Therefore, the tree unravelling of $J$ is infinite. One can employ the least tree unravelling of $J$ to construct the characteristic concept $C$. Then we can check whether $C$ is the msc or not by Lemma 3.3.2.

**Example 7.** Consider the knowledge base $\mathcal{K} = (\mathcal{T}, \mathcal{A})$:

- $\mathcal{T} = \{ A \subseteq \exists r.A \}$.
- $\mathcal{A} = \{ s(a,b), r(b,c), r(b,d), r(d,d), A(d), A(b), B(c) \}$.
Algorithm 2 starts with two initial graphs $mscG_0$ and $canM$. $mscG_0$ is initialized with only root node $a$ in Algorithm 1. The graph $mscG_0$ contains only $a$ and no edges. Then Algorithm 2 visits node $a$, and builds $temp_0$ to be the canonical graph of $mscG_0$ w.r.t. $canM$. Then it checks every edges $e \in pe(E, a)$ in $canM$ whether there exists an edge $e' \in pe(E, a)$ in $temp_0$ such that $e$ is simulated by $e'$. Since the node $a$ contains only one edge $e = (a, s, b)$ where $Ind(a), Ind(b) \in N_{I, A}$, therefore there is no simulation to it by any edge $e$ from $temp_0$. Hence, the edge $e = (a, s, b)$ is added to $mscG_1$.

Then Algorithm 2 visits the next node which is $b$ and builds $temp_1$ of $mscG_2$ w.r.t. $canM$. Since $E_b$ in $canM$ contains three edges $e_1 = (b, r, a), e_2 = (b, r, c), e_3 = (b, r, b)$, it checks whether any of these edges can be simulated by an edge from $E_b$ in $temp_1$. In this case we have that $e_2 = (b, r, d)$ in $E_b$ of $canM$ is simulated by $e' = (b, r, A)$ in $E_b$ of $temp_1$. Therefore, the Algorithm skips adding this edge to $mscG_1$. We have also that $e_3 = (b, r, b)$ in $E_b$ of $canM$ is simulated by $e' = (b, r, A)$ in $E_b$ of $temp_1$. Hence, it is also skipped. However, the edge $e_1 = (b, r, c)$ is not simulated by any edge from $E_b$ in $temp_1$. Hence, $e_1$ is added to $mscG_2$. Now we have only node $c$ in $mscG$ that should be visited before the recursive call terminates. In the last recursive call since the node $c$ has no successors, the algorithm 2 terminates. In Figure 4.5, we can see the graphs generated of Example 7 by Algorithm 1.

Finally, there is still needed to check whether there exists a simulation equivalence between $(temp_1, v_0)$ and $(canM, v_0)$. We also need to check whether $mscG$ contains a cycle. This two checks are done after the recursive call of Algorithm 2 and it returns to continue with remaining steps in Algorithm 1 (see Figure 4.6).

![Figure 4.5](image_url) **Figure 4.5** $canM$ constructed w.r.t. $(I_X, d_a)$, the graphs $mscG_0$, $mscG_1$, $mscG_2$ obtained at the end of each recursive call.

![Figure 4.6](image_url) **Figure 4.6** $temp_1$ constructed w.r.t. $mscG_2$ and $canM$, the simulation equivalence between $(temp_1, v_0)$ and $(canM, v_0)$.
As we can see in Figure 4.6 that after Algorithm 1 terminates, there exists a simulation equivalence between \((\text{temp}, v_0)\) and \((\text{canM}, v_0)\).

However, there could be the case that there exists the msc of \(a\), even though \(\text{mscG}\) contains a cycle. To illustrate this consider the following example.

**Example 8.** Consider the knowledge base \(\mathcal{K} = (\mathcal{T}, \mathcal{A})\):

- \(\mathcal{T} = \{A \subseteq \exists r. A\}\).
- \(\mathcal{A} = \{r(a, b), r(b, a), r(a, c), r(c, d), A(a)\}\).

![Figure 4.7 canM of \((I\mathcal{K}, d_a)\) and mscG w.r.t. Example 8.](image)

In the previous example, we have that \(\text{mscG}\) contains a cyclic path, although there exists the msc of \(a\). The concept \(C = \exists r. (\exists r. A)\) is the msc of \(a\).

Therefore, it is not sufficient to check directly whether \(\text{mscG}\) is cyclic. It is still necessary to find the least sub-graph of \(\text{mscG}\) such that every path \(\sigma = v_r v_1 r v_2 \ldots\) in \(\text{mscG}\) is not simulated by any path \(\sigma' = v_r' v_1' r v_2' \ldots\) in \(\text{temp}_t\), with \(v_i \neq v'_i\) for some \(i > 0\).

![Figure 4.8 \(\sigma \preceq \sigma'\), where \(\sigma\) in \(\text{mscG}\) and \(\sigma'\) in \(\text{temp}_t\), and the resulting sub-graph of \(\text{mscG}\).](image)

In Figure 4.8, we remove the path \(\sigma\) in \(\text{mscG}\) since it is simulated by the path \(\sigma'.\) Hence, we obtain that the final \(\text{mscG}\) is the least graph such that \((\text{temp}_t, v_0)\) simulates \((\text{canM}, v_0)\). Then, if \(\text{mscG}\) is acyclic, then there exists the msc of \(a\). One can construct the characteristic concept \(C\) w.r.t. 3 of \(\text{mscG}\).

Based on this construction of \(\text{mscG}\), we observe the following.

**Lemma 4.4.3.** Let \(\mathcal{K} = (\mathcal{T}, \mathcal{A})\) be a knowledge base, \(\text{canM} = (V, E, \text{Ind}, Y)\) be the graph of \((I\mathcal{K}, d_a)\), \(\text{mscG} = (V_{\text{mscG}}, E_{\text{mscG}}, \text{Ind}_{\text{mscG}}, Y_{\text{mscG}})\) be a graph constructed after the termination of Algorithm 1, and \(\text{temp}_t = (V_t, E_t, \text{Ind}_t, Y_t)\) be the canonical graph of \(\text{mscG}\) w.r.t. \(\text{canM}\). Then For every \((v, r, v_1), (v, r, v_2) \in E_t\) satisfy one of the following conditions:
\[ \textbf{Algorithm 1 ensures all edges } e \text{ (cal graph of } mscG \text{ w.r.t. } canM. Then, every infinite path, } \text{Ind} \text{, the algorithm. This can be characterized by the following lemma.} \]

**Proof.** In Line 6 of Algorithm 1 it calls Algorithm 4, where Algorithm 4 intuitively finds a least graph of \( mscG \) such that the simulation equivalence relation between \( (temp_1, v_0) \) and \( (canM, v_0) \) still holds. By construction of the Algorithm 4 in Line 5, every edge \( (v, r, v') \in E_{mscG} \) is removed from \( mscG \) if there exists an edge \( (v, r, v'') \in E_t \) such that \( (v, r, v') \) is simulated by \( (v, r, v'') \). Therefore, Algorithm 1 ensures all edges \( e \in E_{mscG} \) cannot be simulated by any edge \( e' \in E_t \). This means given two edges \( (v, r, v_1), (v, r, v_2) \in E_t \) we have one of the following cases:

(A) \( (temp_t, v_1) \preceq (temp_t, v_2) \).

(B) Or, \( (temp_t, v_1) \preceq (temp_t, v_2) \).

- Case (A) coincides with the our second condition.
- Case (B). Assume that \( (temp_t, v_1) \preceq (temp_t, v_2) \). Due to the fact that all edges \( e = (v, r, v_1) \in E_{mscG} \), is not simulated by any edge \( e' = (v, r, v_2) \in E_t \). Then, if \( e = (v, r, v_1) \) is simulated by edge \( e' = (v, r, v_2) \), then \( e = e' \). Therefore, also \( v_1 = v_2 \). \[
\]

As a consequence of the construction of Algorithm 1, one can observe that every edge \( e = (v, r, v') \in E_{mscG} \) is not simulated by any edge \( e' = (v, r, v'') \in E_t \) such that \( v' \neq v'' \), where \( temp_t = (V_t, E_t, Ind_t, Y_t) \) is the canonical graph of \( mscG \) w.r.t. \( canM \) that is obtained after termination of the algorithm. This can be characterized by the following lemma.

**Lemma 4.4.4.** Let \( canM = (V_{canM}, E_{canM}, Ind_{canM}, Y_{canM}) \) be the graph of \( (\mathcal{A}, d_a) \), \( mscG = (V_{mscG}, E_{mscG}, Ind_{mscG}, Y_{mscG}) \) the graph obtained after Algorithm 1 terminates, \( temp_t = (V_t, E_t, Ind_t, Y_t) \) the canonical graph of \( mscG \) w.r.t. \( canM \). Then, every infinite path \( \sigma = v_0r_1v_1r_2v_2... \in E_{mscG} \) is not simulated by an infinite path \( \sigma' = v_0'v_1'v_2'... \in temp_t \) such that \( v_i' \in V_t, (mscG, v_i) \preceq (temp_t, v_i') \), and \( v_i \neq v_i' \) for some \( i \geq 0 \).

**Proof.** By construction of Algorithm 1:

1. As shown in Line 4, the graph \( mscG \) is initialized as an empty graph. Then the root node \( v_0 \) in \( canM \) is added to \( mscG \) together with its set of concepts.

2. Then, all nodes in \( V.A \) are visited only once, starting from the root node \( v_0 \) that contains the individual \( a \) (as shown in Algorithm 2):

   - At each recursive call, a node \( v \) is the current node, \( temp_i = (V_i, E_i, Ind_i, Y_i) \) is constructed to be the canonical graph of \( mscG \) w.r.t. \( canM \), where \( mscG_i \) is the graph that is obtained at the recursive call \( i \).
   - Given a node \( v \), a set \( pe(E, v) = \{(v, r, v') \in E\} \) is the set of edges from \( canM \) that has \( v \) as a predecessor, a set \( pe(E_i, v) = \{(v, r, v') \in E_i\} \) is the set of edges from \( temp_i \), that has \( v \) as a predecessor.
   - In Line 15 of Algorithm 2 an edge \( e \in pe(E, v) \) is added to \( mscG \) if there exists no edge \( e' \in pe(E_i, v) \) such that \( e \) is simulated by \( e' \) (recall the Definition 4.3.4).

3. Let \( temp_i = (V_i, E_i, Ind_i, Y_i) \) be the canonical graph of \( mscG \) w.r.t. \( canM \) after Algorithm 2 terminates.

4. Algorithm 2 terminates after all nodes are visited and for every edge \( e \in E \), there is an edge \( e' \in E_t \) such that \( e \) is simulated by \( e' \). Also we can observe that every edge \( e \in pe(E_{mscG}, v) \) is not simulated by any edge \( e' \in pe(E_t, v) \), for every node \( v \in V_{mscG} \). Otherwise, the edge \( e \) would not be added to \( mscG \). Moreover, every edge \( e = (v, r, v') \in E_{mscG} \) we have
that \( v, v' \in V_A \) (\( V_A \) is a set of nodes occurring in \( \text{canM} \) such that the nodes are generated by \( A \)). Due to the fact that \( e \) is not simulated by any of the edges that are added from the canonical graph of \( \text{mscG} \) w.r.t. \( \text{canM} \) (recall the Definition 4.2.9).

Now Suppose that there exists a cyclic path in \( \text{mscG} \). This means that there exists an infinite path \( \sigma = v_0 r_1 v_1 r_2 v_2 ... \in \text{mscG} \) such that every edge \((v_{i−1}, r_i, v_i) \in E_{\text{mscG}}\) is not simulated by an edge \((v_{i−1}, r_i, v_i') \in E_t\), where \( v_{i−1} \in V_t \) and \( v_i \neq v_i' \), for all \( i > 0 \). Otherwise, the edge \((v_{i−1}, r_i, v_i)\) would not exist in \( E_{\text{mscG}} \). Hence, \( \sigma = v_0 r_1 v_1 r_2 v_2 ... \in \text{mscG} \) is not simulated by any infinite path \( \sigma' = v_0 r_1' v_1' r_2' v_2' ... \in \text{temp}_t \) such that \( v_i' \notin V_t \) and \( v_i \neq v_i' \) for some \( i > 0 \).

We recall that Algorithm 1 constructs \( \text{mscG} \) graph such that \( (\text{temp}_t, v_0) \simeq (\text{canM}, v_0) \). Also it checks whether \( \text{mscG} \) contains a cyclic path or not. If \( \text{mscG} \) contains a cyclic path, then there exists no msc of individual \( a \). Otherwise, there exists an msc of \( a \), and one then use convert \( \text{mscG} \) into an interpretation \( I_{\text{mscG}} \) and use the interpretation to compute the msc.

Since deciding the existence of the msc with this approach does not rely on computing the concept, it is easy to see that it can be decided in polynomial time. Then, the constructed graph can be used to construct a possibly exponentially large concept. Also we can observe that role-depth of \( (I_{\text{mscG}, d_a}) \) is bounded by \(|V_{\text{mscG}}|\). Therefore, it is sufficient to construct up to \(|V_{\text{mscG}}|\) to decide the existence of the msc.

4.5 proofs

4.5.1 Correctness

Theorem 4.5.1 (CORRECTNESS). If the Algorithm 1 returns true, then there exists a most specific concept w.r.t. individual \( a \).

Proof. Assume that Algorithm 1 returns true. By the Definition 4.2.5, the interpretation \( I_{\text{canM}} \) is constructed w.r.t. the \( \text{canM} \) graph, and the interpretation \( I_{\text{mscG}} \) is constructed w.r.t. the \( \text{mscG} \) graph. A concept \( C \) is the characteristic concept obtained w.r.t. least tree unravelling of the interpretation \( I_{\text{mscG}, k} \), where \( k = |\Delta_{\text{mscG}}| \). This means that by Line 8, we have the following:

• There exists a simulation equivalence between \( I_{\text{canM}} \) and the canonical model of the concept \( C \) w.r.t. \( J \) (\( I_{\text{canM}} \simeq (J_{\Delta, C}, d_C) \)).
• Graph \( \text{mscG} \) contains no cycle; Meaning that there is no path \( v_1 e_1 ... e_n v_n \) with \( v_n = v_1 \), for all \( e \in E \), and for all \( v \in V \).

Since the \( \text{mscG} \) graph contains no cycle, then \( I_{\text{mscG}} \) contains no cycle and by the definition of least tree unravelling, every path \( \sigma = d r_1 d_1 ... r_m d_m \) of \( \text{mscG} \), we have that \( \{d, d_1, ..., d_m\} \) are distinct individuals; \( \text{Tail}(\sigma) \neq \{d, d_1, ..., d_m\} \), where path \( \sigma = d d_1 r_1 ... r_m d_n \) for all paths \( \sigma \in \mathcal{Y}_{\text{mscG}, d} \). Moreover, By Lemmas 4.4.2 and 4.4.1 we have that \( (\text{temp}, v_0) \simeq (\text{canM}, v_0) \), where \( \text{temp} \) is the canonical graph of \( \text{mscG} \) w.r.t. \( \text{canM} \). While \( \text{mscG} \) is acyclic, it can be un unravelled into a finite tree. Then the concept \( C \) is constructed w.r.t. the least tree unravelling of \( I_{\text{mscG}} \). We have that \( (J_{\Delta, C}, d_C) \simeq (J_{\Delta}, d) \). By Lemma 3.3.2, we obtain that the concept \( C \) is the msc of the individual \( a \).

4.5.2 Completeness

Theorem 4.5.2 (COMPLETENESS). If there exists a most specific concept w.r.t. individual \( a \), then Algorithm 1 returns true.

Proof. Assume that there exists a most specific concept w.r.t. individual \( a \). Now suppose by contradiction that Algorithm 1 returns false. Then by Line 8, either:
(A) There exists no simulation equivalence between \(\text{canM}\) and the canonical graph \(\text{temp}\) of \(\text{mscG}\) w.r.t. \(\text{canM}\), where \(\text{canM}\) is the graph of \((J_X, d_\sigma)\) and \(\text{mscG}\) is the graph obtained after the Algorithm 1 terminates.

(B) Or, the constructed \(\text{mscG} = (V_{\text{mscG}}, E_{\text{mscG}}, \text{Ind}_{\text{mscG}}, Y_{\text{mscG}})\) Graph has a cycle.

Case (A). Suppose that \((\text{canM}, v_0) \not\sim (\text{temp}, v_0)\). Then it is a contradiction for Lemmas 4.4.2 and 4.4.1. After the Algorithm 1 terminates, there must exists a simulation equivalence between \((\text{canM}, v_0)\) and \((\text{temp}, v_0)\).

Case (B). Suppose that \(\text{mscG}\) is cyclic. By construction of Algorithm 1:

1. As shown in Line 4, the graph \(\text{mscG}\) is initialized as an empty graph. Then the root node \(v_0\) in \(\text{canM}\) is added to \(\text{mscG}\) together with its set of concepts.

2. Then, all nodes in \(V_A\) are visited only once, starting from the root node \(v_0\) that contains the individual \(a\) (as shown in Algorithm 2):
   - At each recursive call, a node \(v\) is the current node, \(\text{temp}_i = (V_i, E_i, \text{Ind}_i, Y_i)\) is constructed to be the canonical graph of \(\text{mscG}\) w.r.t. \(\text{canM}\), where \(\text{mscG}\) is the graph that is obtained at the recursive call \(i\).
   - Given a node \(v\), a set \(pe(E, v) = \{(v, r, v') \in E\}\) is the set of edges from \(\text{canM}\) that has \(v\) as a predecessor, a set \(pe(E_i, v) = \{(v, r, v') \in E_i\}\) is the set of edges from \(\text{temp}_i\) that has \(v\) as a predecessor.
   - In Line 15 of Algorithm 2, an edge \(e \in pe(E, v)\) is added to \(\text{mscG}\) if there exists no edge \(e' \in pe(E_i, v)\) such that \(e\) is simulated by \(e'\) (recall the Definition 4.3.4).
   - If there exists an edge \(e' = (v, r, d') \in pe(E_i, v)\) that simulates the edge \(e = (v, r, v') \in pe(E, v)\), then we skip adding edge \(e\) to \(\text{mscG}\). Otherwise, \(e\) is added to \(\text{mscG}\).

3. Let \(\text{temp}_i = (V_i, E_i, \text{Ind}_i, Y_i)\) be the canonical graph of \(\text{mscG}\) w.r.t. \(\text{canM}\) after Algorithm 2 terminates.

4. Algorithm 2 terminates after all nodes are visited and for every edge \(e \in E\), there is an edge \(e' \in E_i\) such that \(e\) is simulated by \(e'\). Also we can observe that every edge \(e \in pe(E_{\text{mscG}}, v)\) is not simulated by any edge \(e' \in pe(E_i, v)\) for every node \(v \in V_{\text{mscG}}\). Otherwise, the edge \(e\) would not be added to \(\text{mscG}\). Moreover, every edge \(e = (v, r, v') \in E_{\text{mscG}}\), we have that \(v, v' \in V_A\) (\(V_A\) is a set of nodes occurring in \(\text{canM}\) such that the nodes are generated by \(A\)). Due to the fact that \(e\) is not simulated by any of the edges that are added from the canonical graph of \(\text{mscG}\) w.r.t. \(\text{canM}\) (recall the Definition 4.2.9).

So, if there is a cyclic path \(\sigma = v_0r_1v_1r_2v_2...\) in the \(\text{mscG}\) graph, then \(\sigma = v_0r_1v_1r_2v_2...\) contains elements from \(V_A\). Furthermore, from 4.4.4 we have that this path \(s\) is not simulated by any cyclic path \(s' = v_0r_1v_1'r_2v_2'...\) in \(\text{temp}_i\) such that \(v_i' \in Y_i\), \((\text{mscG}, v_i) \not\subseteq (\text{temp}_i, v_i')\), and \(v_i \neq v_i'\) for some \(i > 0\).

Now suppose that a concept \(C_0\) is \(\text{mscG}(a)\). Then by Lemma 3.3.2 we have that \((J_{C_0}, d_{C_0}) \simeq (J_X, d_\sigma)\). Let \(G_0 = (V_{G_0}, E_{G_0}, \text{Ind}_{G_0}, Y_{G_0})\) be a graph constructed w.r.t. \((J_{C_0}, d_{C_0})\). Since \(\text{canM} = (V, E, \text{Ind}, Y)\) is the graph of \((J_X, d_\sigma)\), we have that \((G_0, v_0) \simeq (\text{canM}, v_0)\).

Also by Lemmas 4.4.2 and 4.4.1 we have that \((\text{canM}, v_0) \simeq (\text{temp}_i, v_0)\), where \(\text{temp}_i = (V_i, E_i, \text{Ind}_i, Y_i)\) is the canonical graph of \(\text{mscG}\) w.r.t. \(\text{canM}\). Let \(S, S'\) be simulations such that:

i) \(S : (\text{temp}_i, v_0) \subset (\text{canM}, v_0)\)

ii) \(S' : (\text{canM}, v_0) \subset (\text{temp}_i, v_0)\)

Since \((\text{temp}_i, v_0) \subset (\text{canM}, v_0)\) and \((\text{canM}, v_0) \simeq (G_0, v_0)\), we have that there exists a simulation equivalence between \(G_0\) and \(\text{temp}_i\) i.e., \((G_0, v_0) \simeq (\text{temp}_i, v_0)\). Let \(S_1, S_2\) be simulations such that:

1) \(S_1 : (\text{temp}_i, v_0) \subset (G_0, v_0)\)
2) $S_2 : (G_0, v_0) \preceq (\text{temp}_t, v_0)$.

Due to the fact that $\text{mscG}$ is cyclic, then there must exists an infinite path $\sigma = v_0r_1v_1r_2v_2... \in \text{mscG}$. Hence, from 1) there must exists an infinite path $\sigma' = v_0r_1'i_1r_2'i_2... \in G_0$ such that $(v_i, v_j') \in S_1$ for all $i > 0$ and for some $j > 0$, $v_j' \in V_T$ and $v_{j-1} \notin V_T$. Due to the fact that the concept $C_0$ can only be constructed from a finite tree untravelling. Therefore, the cyclic path comes only from the nodes in $V_T$.

Moreover, from 2) there must exists a path $\sigma'' = v_0r_1v_1'r_2v_2'... \in \text{temp}_t$ such that such that $(v_i', v_{i}'') \in S_2$ for all $i > 0$. We conclude that $\sigma \preceq_{S_1} \sigma' \preceq_{S_2} \sigma''$. Also, $(\text{mscG}, v_j) \preceq_{S_1} (G_0, v_j) \preceq_{S_2} (\text{temp}_t, v_j')$ for all $j > 0$.

By Lemma 4.4.4, we one of two cases:

A') $v_j = v_j''$.

B') Or, $v_j'' \notin V_T$ and $v_j \neq v_j''$.

Case A'). Assume that $v_j = v_j''$. Since $(v_j', v_j'+1) \in S_2$, we have that $Y_{G_0}(v_j'+1) \subseteq Y_i(v_j''+1)$. By Lemma 4.4.3 we have that $v_j''+1$ has $v_j'$ as a successor. Assume that $v_j'' = v_j' \in V_T$. Then, we have that $v_j = v_j'' = v_j'$. We have that the cyclic path $\sigma$ in $\text{mscG}$ contains only nodes $v$ such that $v \in V_A$. Then it is a contradiction that $v_j = v_j'' = v_j'$, where $v_j' \in V_T$.

Case B'). Assume that $v_j'' \notin V_T$ and $v_j \neq v_j''$. Then, by Lemma 4.4.4 we have either one of the following conditions:

i') $v_j = v_j''$.

ii') Or, $(\text{mscG}, v_j) \not\preceq (\text{temp}_t, v_j')$.

• Case i'). Suppose that $v_j = v_j''$. Then, it is a contradiction for the assumption that $v_j \neq v_j''$.

• Case ii'). Suppose that $(\text{mscG}, v_j) \not\preceq (\text{temp}_t, v_j'')$. Then it also contradicts the assumption that $\sigma \preceq \sigma''$. Due to the fact that $(\text{mscG}, v_j) \preceq (\text{temp}_t, v_j)$ for all $j > 0$.

Therefore, it is a contradiction for Case B).

Hence, we have that there is a contradiction for Case (B) that $\text{mscG}$ is acyclic. Since we have a contradiction for Cases (A) and (B), we conclude that if there exists a most specific concept w.r.t. individual $a$, then Algorithm 1 must return true.

4.5.3 Polynomial Time Complexity

Theorem 4.5.3 (COMPLEXITY). Let $\mathcal{K} = (\mathcal{T}, A)$ be a knowledge base. the Algorithm 1 decides the existence of the most specific concept w.r.t. an individual $a$ in polynomial time.

Proof. We start with Algorithm 1, which constructs the $\text{mscG} = (V_{mscG}, E_{mscG}, \text{Ind}_{mscG}, Y_{mscG})$ graph from $\text{canM} = (V, E, \text{Ind}, Y)$. Starting from the root node in $\text{canM}$, Algorithm 2 is called recursively, to visit nodes $v_1 \in V$ and adds edges $e_1 \in E$ to $E_{mscG}$. At each recursive call, The $\text{temp}_i = (V_i, E_i, \text{Ind}_i, Y_i)$ graph is the canonical graph of $\text{mscG}$ w.r.t. $\text{canM}$ with $i$ is the number of recursive calls of the Algorithm 2.

Then we go over all edges $e_1 = (v_1, r, v_1') \in E \land v_1 \in V$, and check whether there exists an edge $e_2 = (v_2, r, v_2') \in E \land v_2 \in V$, such that $e_1$ is simulated by $e_2$. It is known that constructing such a simulation between two graphs can be computed in polynomial time in the size of the two graphs $\text{canM}$ and $\text{temp}_i$. [HHK95]. Therefore, Algorithm 2 can be computed in polynomial time w.r.t. the size of two graphs $O(|V| \times |E|)^3$, given that the simulation has the complexity of $O(|V| \times |E|)$. 38
Assuming that the conM graph is a Complete graph, where all vertices are connected directly with an edge, then $|E| = |V|^2$. The worst-case time complexity will be $\mathcal{O}(|V|^2)$. Assuming that the conM graph is a path; Meaning that each node has only one successor, then $|E| = 1$. The best-case time complexity will be $\mathcal{O}(1)$. Since both best and worst-case time complexities belong to the polynomial time complexity class, the Algorithm 1 decides the existence of the most specific concept w.r.t. the individual $a$ in polynomial time.
5 Implementation

In this chapter, we specify an algorithm for deciding the existence of the msc w.r.t. an individual, and it computes the concept if it exists. Moreover, we present an example to describe the construction of the msc.

5.1 The Algorithm

In this section, we describe all the algorithms that we were using to compute the msc if it exists. We use a popular reasoner to reason about ontology. ELK [KKS12] is a specialized reasoner for the lightweight ontology language OWL 2EL. It is used to compute the concept related to each individual. Also, it is used in building the canonical model of $\mathcal{A}$. The program is available at [Nad], where Java programming [AGH05] is used. The reason to choose the programming language is that the OWL API [HB09] is available in Java. OWL API is a Java API that allows us to create and manipulate OWL ontologies.

Let $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ be a knowledge base, $a$ be an individual. Since we assume that $\mathcal{T}$ is normalized TBox, then we normalize $\mathcal{T}$. Then, we check whether the individual $a$ exists in $\mathcal{A}$. If $a \notin N_{\mathcal{I}}$, then return false. Otherwise, we build the canonical model $\mathcal{I}_K$ of $\mathcal{K}$. Then we construct the pointed interpretation $(\mathcal{I}_K, d_a)$ from $\mathcal{I}_K$. Now the interpretation $(\mathcal{I}_K, d_a)$ are used as inputs for the Algorithm 1.

The Algorithm 1 starts with constructing a graph $canM = (V, E, Ind, Y)$ from $(\mathcal{I}_K, d_a)$ (See Line 1 of Algorithm 1). Then initializes a graph $mscG = (V_{mscG}, E_{mscG}, Ind_{mscG}, Y_{mscG})$ as an empty graph. Furthermore, the root node containing the individual $a$ is added to $mscG$. The algorithm traverses all nodes of $canM$ and adds edges and nodes to $mscG$ if needed. For this reason, the algorithm uses two lists visited and non-visited to keep track of the visited and the non-visited nodes in $canM$. Recursively visit nodes of $canM$ starting from the root node. This recursive call is done using the Algorithm 2.

The Algorithm 2 takes initially the root node $v$, two graphs $canM$, $mscG$ and the lists visited, non-visited as inputs. It checks whether the node $v$ is already visited before or not. If yes, then it terminates. Otherwise, it adds the node $v$ to visited to make sure that the algorithm will terminate after visiting all nodes only once. Then it builds a graph $temp_i = (V_i, E_i, Ind_i, Y_i)$ to be the canonical graph of $mscG$ w.r.t. $canM$ (recall the Definition 4.2.9), where $i$ is the number of the recursive calls of Algorithm 2. Let $pe(E, v)$ be the set of edges in $canM$ with the node $v$ as predecessor, and $pe(E_i, v)$ be the set of edges in $mscG$ with the node $v$ as predecessor. It goes over the edges of $pe(E, v)$ and check for every edge $e = (v, r, v') \in pe(E, v)$ whether there exists an edge $e' = (v, r, d) \in pe(E, v)$ such that $e$ is simulated by $e'$ (recall the Definition 4.3.4). If no, then the algorithm adds $e$ to $mscG$ with its nodes. Furthermore, it checks whether the
successor node \( v' \) has been visited before or not. If yes, then it is ignored. Otherwise, it is added to the non-visited list. After checking every edge \( e \in pe(E,v) \) and adding all edges that are not simulated by any edge \( e' \in pe(E,v) \), the algorithm picks a new node to be visited in its next recursive call. For that, the algorithm picks the first element in the non-visited list. Picking the first element of the list ensures that the nodes of \( canM \) are traversed in the breadth-first order.

Algorithm 3 is the algorithm that checks given two edges \( e, e' \) whether there exists a simulation from \( e \) to \( e' \). It takes an edge \( e = (v_1, r, v_1') \) from \( canM \) and an edge \( e' = (v_2, r, v_2') \) from \( temp_i \). Then it checks whether \( Y(v_1) \subseteq Y(v_2) \) and \( Y(v_1') \subseteq Y(v_2') \). If yes, then it checks whether there exists a simulation from \( canM \) to \( temp \) starting from \( v_1' \) in \( canM \) and \( v_2' \) in \( temp \). If it also returns TRUE, then \( e' \) simulates \( e \) and returns TRUE. If any of these checks fail, then \( e' \) does not simulate \( e \), and it will return FALSE.

After the recursive call of the Algorithm 2, it is still essential to construct the least graph of \( mscG \), where every edge \( e \) in \( E_{mscG} \) is not simulated by any edge \( e' \) in \( E_{temp} \). Algorithm 4 is the algorithm which performs this operation. It generates all edges of nodes recursively starting from the root node \( v_0 \) and does the following:

- The set \( (E_{mscG}, v_i) \) is the set containing all edges \( e \) of \( mscG \) with root node \( v_i \). The set \( (E_{temp}, v_i) \) is the set containing all edges \( e \) of \( temp \) with root node \( v_i \).
- For each edge \( e \in (E_{mscG}, v_i) \), we check whether there exists a edge \( e' \in (E_{temp}, v_i) \) such that \( e \preceq e' \) and \( e \neq e' \).
- If the edge \( e \) is simulated by \( e' \), then we remove the edge \( e \) from \( E_{mscG} \).

After Algorithm 4 removes all simulated edges, we still need to remove the nodes \( v \in V_{mscG} \) such that \( v \) is not reachable from \( v_0 \). Then, we obtain the least graph \( mscG \) such that \((temp_i, v_0) \simeq (canM, v_0) \).

Finally, the Algorithm 1 checks whether there exists a simulation equivalence between \((canM, v_0)\) and \((temp_i, v_0)\), where \( temp_i \) is the canonical graph of \( mscG \) w.r.t. \( canM \) that is obtained after Algorithm 2 terminates. Also, it checks whether \( mscG \) contains a cycle. If both checks return TRUE, then Algorithm 1 returns TRUE and \( mscG \) is the most specific concept w.r.t. the individual \( a \). Otherwise, there is no \( msc \) w.r.t. \( a \).

---

### Algorithm 1: BuildMsc function.

For an individual \( a \) and an ontology \( O \). The function call is \( \text{BuildMsc}(a,O) \).

<table>
<thead>
<tr>
<th>Input:</th>
<th>a, O /* individual a and ontology O */</th>
</tr>
</thead>
<tbody>
<tr>
<td>Output:</td>
<td>Boolean</td>
</tr>
</tbody>
</table>

1. \((V, E, Ind, Y) \leftarrow \text{CanonicalModel}(a)\) /* set \((J_X, d_X)\) to \( canM \) */
2. \( \text{non-visited} \leftarrow \{v_0 \in V \mid v_0 \text{ is root node} \} \) /* initialize the set of non visited nodes to root node \( v_0 \) of \( canM \) */
3. \( \text{visited} \leftarrow \emptyset \) /* initialize the set of visited nodes to empty */
4. \((V_{mscG}, E_{mscG}, Ind_{mscG}, Y_{mscG}) \leftarrow (\{v_0 \in V \}, \emptyset, (v_0; Y(v_0))) \) /* intialize \( mscG \) with \( v_0 \in V \) and its concepts * /
5. \( \text{visitNode}() \) /* recursively add nodes and edges from \( canM \) to \( mscG \) */
6. \( \text{LeastGraph}(mscG, canM) \) /* construct the least graph of \( mscG \) */
7. \((V_t, E_t, Ind_t, Y_t) \leftarrow \text{CanonicalGraph}(mscG, canM) \) /* construct the canonical graph \( temp \) */
8. \( \text{if} \ (temp_i, v_0) \simeq (canM, v_0) \text{ AND Cyclic}(mscG) = \text{FALSE} \) /* check whether \( mscG \) is cyclic and there exists a simulation equivalence between \((temp_i, v_0)\) and \((canM, v_0)\) */
9. \( \text{then} \)
10. \( \text{return} \ \text{TRUE} \) /* \( msc \) exists */
11. \( \text{else} \)
12. \( \text{return} \ \text{FALSE} \) /* No \( msc \) exists */

---
Algorithm 2: visitNode recursive function. For a node $v$, a set non-visited of non visited nodes, a set visited of visited nodes, a graph $mscG = (V_{mscG}, E_{mscG}, Ind_{mscG}, Y_{mscG})$ and graph $canM = (V, E, Ind, Y)$. The initial function call is visitNode$(v_0, \text{non-visited}, \text{visited}, mscG, canM)$.

**Input:** $v$, non-visited, visited, mscG, canM /* node $v$, sets non-visited and visited of visited and non visited nodes, Graphs mscG and canM */

**Output:** mscG /* the constructed mscG */

1. if $v \notin \text{visited}$ /* check whether $v$ is not visited before */
2. then
3. 
4. 
5. for $(v, r, d) \in pe(E, v) \land d \in V_A$ /* loop over the edges $e \in pe(E, v)$, and successor node $d$ is an individual */
6. do
7. 
8. 
9. if simulatedBy$((v, r, d), (v, r, d'), canM, temp) = \text{TRUE} /* check whether $(v, r, d) \preceq (v, r, d') */$
10. then
11. 
12. if simulated = \text{FALSE} /* check whether there is no simulation with any edge $e' \in pe(E_{temp}, v)$ to $e \in pe(E, v)$ */
13. then
14. 
15. if $d \notin \text{visited}$ /* check whether node $d$ is already visited before */
16. then
17. 
18. else
19. if non-visited $\neq \emptyset$ /* non-visited contains nodes that still need to be visited */
20. then
21. 
22. 
23. return mscG /* return the constructed graph mscG */
Algorithm 3: SimulatedBy function. For an edge \((v, r, d) \in pe(E, v)\), an edge \((v, r, d') \in pe(E_{\text{temp}}, v)\), a graph \(canM = (V, E, \text{Ind}, Y)\) and \(temp = (V_{\text{temp}}, E_{\text{temp}}, \text{Ind}_{\text{temp}}, Y_{\text{temp}})\). The function call is SimulatedBy((\(v, r, d\)), (\(v, r, d'\)), canM, temp).

**Input:** (\(v, r, d\)), (\(v, r, d'\)), canM, temp 

**Output:** Boolean

1. If \(Y(d) \subseteq Y_{\text{temp}}(d')\) then
2. If \((canM, d) \preceq (\text{temp}, d')\) then
3. Return TRUE /* \((v, r, d) \preceq (v, r, d')\) */
4. Else
5. Return FALSE /* \((v, r, d) \npreceq (v, r, d')\) */
6. Else
7. Return FALSE /* \((v, r, d) \npreceq (v, r, d')\) */

Algorithm 4: LeastGraph function. For a graph \(mscG = (V_{\text{mscG}}, E_{\text{mscG}}, \text{Ind}_{\text{mscG}}, Y_{\text{mscG}})\), a graph \(canM = (V, E, \text{Ind}, Y)\). The function call is LeastGraph(mscG,canM).

**Input:** mscG, canM /* graphs mscG and canM */

**Output:** mscG /* the least graph of mscG */

1. For \(v \in V_{\text{mscG}}\) do
2. \(\text{temp} \leftarrow \text{CanonicalGraph}(\text{mscG}, \text{canM})\)
3. For \(e \in (E_{\text{mscG}}, v)\) do
4. For \(e' \in (E_{\text{temp}}, v)\) do
5. If \(e \preceq e' \text{ AND } e \neq e'\) then
6. \(E_{\text{mscG}} \leftarrow E_{\text{mscG}} \setminus \{e\}\) /* remove the simulated edges from \(E_{\text{mscG}}\) */
7. Else
8. For \(d \in V_{\text{mscG}}\) do
9. If \(pe(E_{\text{mscG}}, d) = \emptyset\) /* check whether \(d\) has no successors */
10. Then
11. If \((v', r, d) \notin E_{\text{mscG}}\) /* check whether \(d\) has no predecessor */
12. Then
13. \(V_{\text{mscG}} \leftarrow V_{\text{mscG}} \setminus \{d \in V_{\text{mscG}} \text{ no successors}\}/* remove the node that is not reachable and has no successors */
14. Return mscG
6 Evaluation

In this chapter, we provide an experimental evaluation of a different set of benchmarks. We state the hardware setup that was used to evaluate the benchmarks. Then, we provide the selected set of instances from the different benchmarks. Finally, we provide experimental results to state the difference between the previous and the new approach within a timeout of 1,600 seconds.

6.1 Experimental Setting

6.1.1 Hardware Setup

All experiments have been performed on a machine equipped with 11th Gen Intel(R) Core(TM) i7-11370H with four 3.30GHz cores and 16GB of RAM operated by Windows 11 (21H2 version). Each task has been given a 1,600-second timeout. This timeout includes the time required to decide the existence of the msc and the time required to manipulate the input ontology.

6.1.2 Benchmarks

We considered a set of instances from three benchmark sets. The set consists of 155 instances from BioPortal which includes ontologies from applications in biology and medicine [Noy+09], 109 instances from OWL reasoner evaluation (ORE) 2015 competition [Par+17], and 2091 instances from the Manchester OWL repository [Mat+14]. This set is filtered to contain only those instances such that each instance contains at least one individual with at least one successor. We write ‘ORE’, ‘BIO’ and ‘MOR’ to express the instances from OWL reasoner evaluation (ORE) 2015 competition, BioPortal and the Manchester OWL repository respectively. We write ‘ALL’ to express the full set of instances. For each instance, we select the individual with the most number of successors as our input.

6.2 Experimental Results

In this section, we provide experimental results to compare the new approach with the previous approach. First, we show the number of decided and undecided instances w.r.t. a timeout of 1,600 seconds. Then, we show the time required to decide the existence of the msc. Finally, we show the difference between the maximal role-depth k that would have been used for Lemma 3.3.3, and the role-depth actually explored by our method.
In Table 6.1, the leftmost “Benchmark” column indicates each benchmark set, while “instance” indicates the number of instances. Sub-columns “Decided” and “Undecided” represent the number of solved and unsolved instances within the timeout of 1,600 seconds. The sub-column “Total” represents the total number of the selected instances per each set. The row “ALL” represents the full set of the selected instances.

Table 6.1 The number of decided and undecided instances with a 1,600 seconds timeout.

<table>
<thead>
<tr>
<th>Benchmark</th>
<th>Decided</th>
<th>Undecided</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>ORE</td>
<td>95</td>
<td>14</td>
<td>109</td>
</tr>
<tr>
<td>BIO</td>
<td>148</td>
<td>7</td>
<td>155</td>
</tr>
<tr>
<td>MOR</td>
<td>2003</td>
<td>88</td>
<td>2091</td>
</tr>
<tr>
<td>ALL</td>
<td>2246</td>
<td>109</td>
<td>2355</td>
</tr>
</tbody>
</table>

We acknowledge that all experimental results and all drawn conclusions are based on a single benchmark set.

Table 6.2 The comparison between the maximal role-depth of the previous and new approach.

<table>
<thead>
<tr>
<th>Benchmark Set</th>
<th>min</th>
<th>median</th>
<th>mean</th>
<th>max</th>
</tr>
</thead>
<tbody>
<tr>
<td>Previous</td>
<td>2315.00</td>
<td>61779602.00</td>
<td>344157725.99</td>
<td>2075238298.00</td>
</tr>
<tr>
<td>New</td>
<td>1.00</td>
<td>1.67</td>
<td>1.00</td>
<td>1.00</td>
</tr>
<tr>
<td>Previous</td>
<td>3.00</td>
<td>995144.50</td>
<td>25039500.90</td>
<td>2101075370.00</td>
</tr>
<tr>
<td>New</td>
<td>1.00</td>
<td>2.50</td>
<td>1.89</td>
<td>11.00</td>
</tr>
<tr>
<td>Previous</td>
<td>3.00</td>
<td>327.00</td>
<td>13954266.52</td>
<td>2082188164.00</td>
</tr>
<tr>
<td>New</td>
<td>1.00</td>
<td>2.00</td>
<td>1.80</td>
<td>21.00</td>
</tr>
<tr>
<td>Previous</td>
<td>3.00</td>
<td>2250.50</td>
<td>28651480.83</td>
<td>2101075370.00</td>
</tr>
<tr>
<td>New</td>
<td>1.00</td>
<td>1.50</td>
<td>1.80</td>
<td>21.00</td>
</tr>
</tbody>
</table>

In Table 6.2, the leftmost “Benchmark” column indicates each benchmark set, and “Role-depth” indicates the role-depth explored by Algorithm 1 and respectively the role-depth bound from Lemma 3.3.3. Sub-columns “min”, “median”, “mean”, and “max” present information about instances which did not include time out for a given setup in seconds. Their meaning is self-explanatory, i.e. “min” and “max” stand for minimal and maximal role-depth required to decide the existence of the msc in an instance within a setup. “Mean” and “median” denote the mean and median of role-depth required needed to solve all instances within a setup. As mentioned before, timed-out instances are not included in this analysis. Rows “Previous” and “New” indicate the computation of the role-depth in the previous- and the new approach per each benchmark set.

As we can see that the role-depth of the interpretation with the new approach is significantly smaller than the one we obtained from the previous one.

Figure 6.1 shows how often there existed an MSC, respectively did not exist one, w.r.t. the individual, for the benchmarks “ORE” and “BIO”. It decides whether there exists or not within the timeout. Therefore, we exclude those instances that exceeded the timeout. In the “ORE” benchmark set of Figure 6.1a, we can see that Algorithm 1 was able to find the msc of an individual in all instances without exceeding the timeout. Therefore, the number of instances where the msc of an individual exists was 95. However, in the “BIO” benchmark set of Figure 6.1b,
a) ORE

**Figure 6.1** Bar plots indicating the number of instances with msc and No msc w.r.t. “ORE” and “BIO” within the timeout.

we have that the number of instances with no msc exists was 13, while the number of instances with the msc was 135.

b) BIO

a) MOR

**Figure 6.2** Bar plots indicating the number of instances with msc and No msc w.r.t. “MOR” and “ALL” within the timeout.

Figure 6.2 indicates the existence of the msc w.r.t. an individual for both “MOR” and “ALL” benchmarks. It decides whether there exists or not within the timeout. Therefore, we exclude those instances that exceeded the timeout. In the “MOR” benchmark set of Figure 6.2a, the number of instances with msc exists was 1837, while the number of instances with no msc exists was 166. Also, in the “ALL” benchmark set, we have that the number of instances with msc exists was 2067, while the number of instances with no msc exists was 179. Table 6.3 summarizes the number of instances with msc and no msc found that was computed with Algorithm 1.

<table>
<thead>
<tr>
<th>Number of Instances</th>
<th>Benchmark</th>
<th>Msc</th>
<th>No Msc</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>ORE</td>
<td>95</td>
<td>0</td>
<td>95</td>
<td></td>
</tr>
<tr>
<td>BIO</td>
<td>135</td>
<td>13</td>
<td>148</td>
<td></td>
</tr>
<tr>
<td>MOR</td>
<td>1837</td>
<td>166</td>
<td>2003</td>
<td></td>
</tr>
<tr>
<td>ALL</td>
<td>2067</td>
<td>179</td>
<td>2246</td>
<td></td>
</tr>
</tbody>
</table>

Table 6.3 The number of instances with msc and with no msc exists.

Figure 6.3 indicates the time required to decide the existence of the msc w.r.t. an individual
6 Evaluation

a) ORE

Figure 6.3 Line plots indicating the time required to decide the existence of the msc. In both plots, the instances are sorted in ascending order by the time required, where x-axes indicate the time (in seconds) and y-axes indicate the number of instances.

Within 1,600 seconds, where the line highlighted in “red” indicates the time for instances with the msc and the line highlighted in “blue” indicates the time for instances without the msc. Since all instances of “ORE” have the msc, therefore we can see in Figure 6.3a that there is no line highlighted “blue”. In Figure 6.3b, we see that the number of instances without the msc required more time to be computed than the number of instances with the msc in “BIO” instances.

b) BIO

Figure 6.4 Line plots indicating the time required to decide the existence of the msc. In both plots, the instances are sorted in ascending order by the time required, where x-axes indicate the time (in seconds) and y-axes indicate the number of instances.

Figure 6.4 indicates the time required to decide the existence of the msc w.r.t. an individual within 1,600 seconds, where the line highlighted in “red” indicates the time for instances with the msc and the line highlighted in “blue” indicates the time for instances without the msc. Figure 6.4a of “MOR” instances shows that the number of instances without the msc exists is smaller than the number of instances with the msc. However, the time consumed in the instances with no msc w.r.t. an individual is larger than the time consumed in the instances with msc. This is due to the time required to find the least graph and remove the simulated paths in the cyclic graphs.

Figure 6.4b of “ALL” instances shows that the number of instances without the msc is smaller than the number of instances with the msc. However, the time consumed in the instances with no msc w.r.t. an individual is larger than the time consumed in the instances with msc. This is due to the time required to find the least graph and remove the simulated paths in the cyclic graphs.

Figures 6.5 and 6.6 indicate the computed bounded role-depth of the previous- and the new approach, where the line highlighted in “red” indicates the role-depth of the previous approach.
6 Evaluation

Figure 6.5 Line plots indicating the bounded role-depth of the previous- and new approach. In both plots, the instances are sorted in ascending order by the bounded role-depth, where x-axes indicate the bounded role-depth and y-axes indicate the number of instances.

and the line highlighted in “blue” indicates the role-depth of the new approach.

As we can see that the overall bounded role-depth of the new approach has a significantly smaller number than the one we obtained from the previous approach. Therefore, deciding the existence of the msc w.r.t. an individual using the new approach is much better in terms of performance and time required than deciding the existence of the msc w.r.t. an individual using the previous approach. Thus, constructing the concept w.r.t. the new bound is much smaller in terms of the size of the concept than constructing the concept w.r.t. the previous bound.

Figure 6.6 Line plots indicating the bounded role-depth of the previous- and new approach. In both plots, the instances are sorted in ascending order by the bounded role-depth, where x-axes indicate the bounded role-depth and y-axes indicate the number of instances.

We can summarize that the new approach provides an efficient mechanism for deciding the existence of the msc. Also, it has a smaller bound than the previous bound. Finally, the new approach does not rely on constructing the concept. Thus, it is polynomial even if constructing the concept could take exponential time in case that concept was exponentially large. Although computing the concept could take exponential time, it is still better to compute it with our new method. The reason is that the role-depth computed by our new method is significantly smaller than the one obtained from Lemma 3.3.3. Thus, computing the concept from the new approach would take less time than computing it from the previous approach even that the
concept was exponentially large.
7 Conclusion and Future Work

Computing the most specific concept is a non-standard inference task that supports bottom-up construction of the knowledge base. It is the least concept which has an individual as an instance.

We have revisited the previous approach from [Baa03a] which computes the most specific concept w.r.t. an individual in the presence of terminological TBoxes.

We have presented the previous approach for deciding the existence of the msc w.r.t. an individual from [ZT13] in the presence of general TBoxes. We have shown that in practice, the previous approach does not explain how to decide the existence of the msc practically without constructing the msc explicitly, which could be exponentially large.

In Chapter 4, we have introduced a new approach for deciding the existence of the msc w.r.t. an individual without relying on constructing the concept. The main idea of the new approach is to construct a sub-interpretation of the canonical model such that there exists a simulation equivalence between the canonical graph of the sub-interpretation and the canonical model. We observed that it is only necessary to construct the sub-interpretation in the cyclic ABoxes, as the msc always exists in the acyclic ABoxes.

Moreover, in Chapter 5, we have presented an implementation of our approach. We have introduced an efficient method to decide the existence of the msc practically. We have provided correctness and completeness proofs of our newly introduced approach. Also, we have shown that it has polynomial run-time.

Finally, in Chapter 6, we have provided an experimental evaluation to state the difference between both approaches in terms of bounded role-depth. Also, we have shown the time performance of the new approach for deciding the existence of the msc. We have observed that our new approach has smaller maximal role-depth than the one obtained from the previous approach from [ZT13]. Also, the concept computed with our new approach is smaller the one computed from the previous approach. We have acknowledged that the method takes a longer time in the instances the msc does not exist than the instance with msc exists.

7.1 Future Work

The Future work on the practical side is to implement the newly introduced approach in the context of ontology repairing. This can be achieved by integrating the new approach with the procedure provided by [Kri] and integrate it with the existing tools [ET12].

On the theoretical side, we would like to extend the results towards knowledge bases formulated in more expressive Horn-DLs than $\mathcal{EL}$. 

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Bibliography


<table>
<thead>
<tr>
<th>Reference</th>
<th>Description</th>
</tr>
</thead>
</table>