1 Introduction

Description Logics (DLs) are a family of formal logics designed for representing and inferring new knowledge from defined knowledge encapsulated in ontologies. However, some ontologies like SNOMED or LOINC [4] contain a vast number of axioms, making it challenging to extend them without generating unwanted inferences. Proofs, which are essentially trees consisting of axioms as vertices connected by direct consequence operators, can be useful in understanding the cause of these unwanted inferences. Recently, a new Java library called Eveelibs [3] has been developed to generate proofs for DLs up to ALCH. To achieve this, Evee-libs uses Lethe [5], a consequence-based reasoning procedure, to generate new axioms. However, the underlying direct consequence operator, which is Lethes underlying calculus, plays a critical role in determining the resulting proofs. Therefore, in this student project report, I will present an implementation of the alternative calculus proposed in [3] and compare the generated proofs to those of Evee-libs. This comparison will provide valuable insights into the effectiveness of different calculi in generating proofs for ontologies, which could ultimately lead to improved methods for extending and reasoning about ontologies.

2 The Description Logic \mathcal{ALCH}

2.1 Syntax

The DL \mathcal{ALCH} allows the operators concept negation (\neg) , concept intersection (\sqcap) , concept union (\sqcup) , value restriction (\forall) , existential restriction (\exists) , the top concept (\top) , the bottom concept (\bot) and finally general concept inclusions (GCIs) (\sqsubseteq) . With those operators and two additional sets of concept names (N_C) and role names (N_R) , one can then build all complex concepts of \mathcal{ALCH} $(C_{\mathcal{ALCH}})$ as follows:

```
\begin{array}{ll} \text{if } C \in N_C \\ \text{if } C \in N_{\mathcal{ALCH}} \\ \text{if } C \in N_{\mathcal{ALCH}} \\ \text{if } C_1, C_2 \in N_{\mathcal{ALCH}} \\ \text{if } C_1, C_2 \in N_{\mathcal{ALCH}} \\ \text{if } C \in N_{\mathcal{ALCH}} \\ \text{if } C \in N_{\mathcal{ALCH}} \\ \text{and } r \in N_R \\ \text{if } C \in N_{\mathcal{ALCH}} \text{ and } r \in N_R \\ \text{and negated concept names are called literals} \end{array} \right. \\ \text{then } C \in N_{\mathcal{ALCH}} \\ \text{then } C_1 \sqcup C_2 \in N_{\mathcal{ALCH}} \\
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Now that complex concepts are defined we can also formalize the notion of a TBox.

Definition 1 (TBox). An \mathcal{ALCH} TBox \mathcal{T} is a finite set of general concept inclusions and role inclusions i.e. $\mathcal{T} \subseteq \{C_1 \sqsubseteq C_2 \mid C_1, C_2 \in N_{\mathcal{ALCH}}\} \cup \{r_1 \sqsubseteq r_2 \mid r_1, r_2 \in N_R\}$

Definition 2 (ABox). An ABox \mathcal{A} is a finite set of concept and role assertions i.e. $\mathcal{A} \subseteq \{a: C \mid a \in N_I, C \in N_{\mathcal{ALCH}}\} \cup \{(a,b): r \mid a,b \in N_I, r \in N_R\}$

Definition 3 (Ontology). An Ontology \mathcal{O} is a pair $(\mathcal{T}, \mathcal{A})$ of a TBox \mathcal{T} and an $ABox \mathcal{A}$

In order to later define a normal form for an ontology, we need to define the notion of a concept to occur positively or negatively in an ontology.

Definition 4. Let $\mathcal{O} = (\mathcal{T}, \mathcal{A})$ some ontology and C some concept then we say that a concept C occurs positively (negatively) in \mathcal{O} if C occurs positively (negatively) in some GCI of \mathcal{T} .

- C occurs positively in itself.
- If C occurs positively (negatively) in C' then C occurs positively (negatively) in C' \(\partial DC \cap D', C' \sup D, C \sup D', \exists r.C', \forall r.C' \) and negatively (positively) in \(\neg C' \).

2.2 Semantics

The semantics of the DL \mathcal{ALCH} are defined by a first order logic interpretation $I := (\Delta^I, \cdot^I)$, where Δ^I is the *domain* of the interpretation and \cdot^I the interpretation function which assigns to:

- every concept name $A \in N_C$ a subset $A^I \subseteq \Delta^I$
- every role name $r \in N_R$ a binary relation $r^I \subseteq \Delta^I \times \Delta^I$

First Concepts are then interpreted as follows:

Definition 5 (Interpretation). Let $A \in N_C, C, D \in N_{ALCH}, r \in N_R$ then

```
\begin{array}{lll} \top^I & \coloneqq & \Delta^I \\ \bot^I & \coloneqq & \emptyset \\ (\neg C)^I & \coloneqq & C^I \\ (C \sqcap D)^I & \coloneqq & C^I \cap D^I \\ (C \sqcup D)^I & \coloneqq & C^I \cup D^I \\ (\exists r.C)^I & \coloneqq & \{a \in \Delta^I \mid \exists b \in \Delta^I : (a,b) \in r^I \text{ and } b \in C^I \} \\ (\forall r.C)^I & \coloneqq & \{a \in \Delta^I \mid \forall b \in \Delta^I : if(a,b) \in r^I, \text{ then } b \in C^I \} \end{array}
```

GCIs and role inclusions are used to express constrains which have to be satisfied by an interpretation. To express that an interpretation satisfies a GCI or a role inclusion the satisfaction operator (\models) is used.

Definition 6. For
$$C, D \in N_{\mathcal{ALCH}}: I \models C \sqsubseteq D: \iff C^I \subseteq D^I$$
 For $r, s \in N_R: I \models r \sqsubseteq s: \iff r^I \subseteq s^I$

This definition can be extended to TBoxes and Ontologies. Usually also the ABox has to be satisfied by an interpretation, but since the algorithm only reasons over the TBox, I will not define it here.

Definition 7. For
$$\mathcal{T}TBox\ I \models \mathcal{T} : \iff I \models C \sqsubseteq D \text{ for all } C \sqsubseteq D \in \mathcal{T}$$
 $I \models \mathcal{O} : \iff I \models \mathcal{T} \text{ and } I \models \mathcal{A}$

Another very important concept is that ontologies can entail some GCI or role inclusion. This is particular important since we want to find proofs for such entailments.

Definition 8. Let \mathcal{O} ontology, $C \sqsubseteq D$ GCI and $r \sqsubseteq s$ role inclusion. Then

- $\mathcal{O} \models C \sqsubseteq D$ iff for all interpretations I if $I \models \mathcal{O}$ then $I \models C \sqsubseteq D$
- $\mathcal{O} \models r \sqsubseteq s$ iff for all interpretations I if $I \models \mathcal{O}$ then $I \models r \sqsubseteq s$

3 The Algorithm

In this section I will describe how my algorithm works and therefore start with what a proof formally is.

3.1 Proofs

Proofs capture the logical derivation structure of an GCI, which is entailed by some Ontology and therefor we define a proof for some $\mathcal{O} \models \alpha$ as defined in [1] as finite, acyclic, directed, labeled hypergraphs G = (V, E, l) where V is a finite set of vertices, E is a finite set of hyperedges (S,d), where $S \subseteq V$ and $d \in V$ and l a labeling function that assigns GCIs to vertices. Note that leaves must be labeled by Ontology GCIs and α has to be the label for the root vertex. Hyperedges are also restricted further such that $\{l(v) \mid v \in S\} \models l(d)$ has to hold.

3.2 Normalization

Before we dive into the rule based procedure, we first need to be sure, that our ontology has the right form.

Definition 9 (Normal form). An ontology \mathcal{O} is said to be in normal form, iff all axioms in \mathcal{O} are of the form $\bigcap A_i \sqsubseteq \bigcup B_j, A \sqsubseteq \exists r.B, \exists r.A \sqsubseteq A, A \sqsubseteq \forall r.B$ or $r \sqsubseteq s$, where A_i, B_j, A, B are concept names and r, s are role names. [4]

An \mathcal{ALCH} ontology \mathcal{O} can be easily transformed into this normal form by first removing all negative occurrences of the form $\forall r.C$ and replacing them with $\neg \exists r. \neg C$. $\operatorname{st}(\mathcal{O})$ needs then to be further normalized by removing all concepts of the form $\forall r.C \sqsubseteq A$ and replacing them with the logical equivalent concept $\neg \exists r. \neg C$. Then we use the structural transformation operator st to transform the ontology into a new ontology $\operatorname{st}(\mathcal{O})$. $\operatorname{structural\ transformation\ st}: N_C \to N_C$. st is defined as follows:

$$\begin{array}{ll} \operatorname{st}(A) \coloneqq A & \operatorname{st}(C \sqcap D) \coloneqq [C] \sqcap [D] \\ \operatorname{st}(\top) \coloneqq \top & \operatorname{st}(C \sqcup D) \coloneqq [C] \sqcup [D] \\ \operatorname{st}(\bot) \coloneqq \top & \operatorname{st}(\exists r.C) \coloneqq \exists r.[C] \\ \operatorname{st}(\neg C) \coloneqq \neg [C] & \operatorname{st}(\exists r.C) \coloneqq \exists r.[C] \end{array}$$

where for a concept C [C] is just a new atom that corresponds to C. Now we obtain a new structurally transformed ontology $\operatorname{st}(\mathcal{O})$ form \mathcal{O} by first adding all role inclusions of \mathcal{O} into it and then adding

- $\operatorname{st}(C) \sqsubseteq [C]$ for every negative $C \in \mathcal{O}$
- $[D] \sqsubseteq \operatorname{st}(D)$ for every positive $D \in \mathcal{O}$
- $[C] \sqsubseteq [D]$ for every axiom $C \sqsubseteq D$ occurring $C \in \mathcal{O}$

Finally the following normalization rules have to be applied to $st(\mathcal{O})$ until no rule can be applied anymore.

$$\bullet \ \frac{[C\sqcap D]\sqsubseteq [C]\sqcap [D]}{[C\sqcap D]\sqsubseteq [C], \quad [C\sqcap D]\sqsubseteq [D]}$$

$$\bullet \ \frac{[C] \sqcup [D] \sqsubseteq [C \sqcup D]}{[C] \sqsubseteq [C \sqcup D], \quad [D] \sqsubseteq [C \sqcup D]}$$

$$\bullet \ \frac{[\neg C] \sqsubseteq \neg [C]}{[\neg C] \sqcap [C] \sqsubseteq \bot}$$

$$\bullet \ \frac{\neg [C] \sqsubseteq [\neg C]}{\top \sqsubseteq [\neg C] \sqcap [C]}$$

3.3 Inference rules

Definition 10 (Inference Rules). On a normalized ontology we can now apply the following inference rules.

$$\mathbf{R}_{\mathbf{A}}^{+} \frac{}{H \square A} : A \in H \tag{1}$$

$$\mathbf{R}_{\mathbf{A}}^{-} \frac{H \sqsubseteq N \sqcup A}{H \sqsubseteq N} : \neg A \in H \tag{2}$$

$$\mathbf{R}_{\sqcap}^{\mathbf{n}} \frac{\{H \sqsubseteq N_i \sqcup A_i\}_{i=1}^n}{H \sqsubseteq \bigsqcup_{i=1}^n N_i \sqcup M} : \prod_{i=1}^n A_i \sqsubseteq M \in \mathcal{O}$$
 (3)

$$\mathbf{R}_{\exists}^{+} \frac{H \sqsubseteq N \sqcup A}{H \sqsubseteq N \sqcup \exists r.B} : A \sqsubseteq \exists r.B \in \mathcal{O}$$

$$\tag{4}$$

$$\mathbf{R}_{\exists} \frac{H \sqsubseteq M \sqcup \exists r.K, \quad K \sqsubseteq N \sqcup A}{H \sqsubseteq M \sqcup B \sqcup \exists r.(K \sqcap \neg A)} : \exists s.A \sqsubseteq B \in \mathcal{O} \quad r \sqsubseteq_{\mathcal{O}} s$$
 (5)

$$\mathbf{R}_{\exists}^{\perp} \frac{H \sqsubseteq M \sqcup \exists r.K, \quad K \sqsubseteq \bot}{H \sqsubseteq M} \tag{6}$$

$$\mathbf{R}_{\forall} \frac{H \sqsubseteq M \sqcup \exists r.K, \quad H \sqsubseteq N \sqcup A}{H \sqsubseteq M \sqcup N \sqcup \exists r.(K \sqcap B)} : A \sqsubseteq \forall s.B \in \mathcal{O} \quad r \sqsubseteq_{\mathcal{O}} s$$
 (7)

where M,N are disjunctions of concept names and H,K are conjunctions of literals.

Algorithm 1 Main Loop

```
procedure MAIN(ontology, goal)

Normalizer.normalize(ontology)

proofHandler ← new ProofHandler(qoal)

rules ← getRules(ontology, proofHandler)

while notFinished(proofHandler) do

for (rule in rules) do

setNewRule(rule.apply())

end for
end while
end procedure
```

The algorithm's main loop consists of two stages: normalization and rule application. The normalization stage implements the algorithm as outlined in Section 3.2. The normalized ontology and a new instance of a ProofHandler are then used to create instances of each rule type. Every rule, except for 3.5.1, 3.5.2 and 6 which do not need ontology axioms, need these two as arguments. The ontology is preprocessed, such that all relevant ontology axioms are saved in a way which allows for an easier rule application and prevents searching for relevant axioms every time the rule is applied. The ProofHandler provides active axioms from which rules are able to create new axioms and processes them, as described in Section 3.4, to generate new inferences. The created rules are then applied one after the other until the halting condition is met. The halting condition is set to true either when the goal concept is reached or when all rules are unable to generate a new axiom in an iteration.

3.4 Proof Handler

The ProofHandler is the main component of the algorithm. It is responsible for handling new produced axioms, provides the rules with these new axioms, checks whether the halting condition is met and creates a proof in case the algorithm finds one. The ProofHandler is initialized with the goal axiom. The subconcept of this axiom is added to the list of active axioms and the super concept is saved in order to check whether the goal was found. The handling of newly produced axioms contains the following steps. First the ProofHandler is provided with all axioms which were used to generate, the new axiom the newly generated axiom and the name of the rule which calls the method. In case of rules which also use ontology axioms the provided axioms are split into premises and ontology premises. This information is then used to create a new Inference and add the new axiom to the list of active Axioms. Since some rules do generate some axioms multiple times the ProofHandler also checks whether the axiom is already in the list of active axioms and depending on the outcome of this check returns the information whether a new Inference was created. While doing so it is also checked whether the super concept of the new axiom is the goal axiom. In case the algorithm stops and the goal was found the ProofHandler also provides a method to create a proof. In order to prevent to just return all Inferences which were created during the algorithm the ProofHandler uses the Java library EVEE to extract a meaningful proof. EVEE not only provides data structures like Inferences and Proofs but also has different algorithms to create different types of proofs. In this project I use the algorithm which extracts a proof with the lowest depth possible.

3.5 Inference Rules

In this subsection, I will describe how each inference rule is implemented and what data structures they use.

3.5.1 The rule
$$\mathbf{R}_{\mathbf{A}}^+ := \frac{1}{H \square A} : A \in H$$

The rule 3.5.1 is one of the simplest rules, since it does not use any active axioms and also no axioms from the ontology. The only thing that has to be taken care of is creating new axioms from active concepts. In order to do that, the implementation iterates through all active concepts and searches for conjunctions or concept names, which are seen as conjunctions of size 1. In case a conjunction H is found, all concept names A within the conjunction are saved to a list. Each element A is then used as superconcept of a new axiom with its origin H as subconcept. After the handling of all active concepts, the implementation empties the list of active concepts in order to reduce redundant axiom creation.

3.5.2 The rule
$$\mathbf{R}_{\mathbf{A}}^- := \frac{H \sqsubseteq N \sqcup A}{H \sqsubseteq N} : \neg A \in H$$

This rule also does not use any axioms from the ontology but does use active axioms. The implementation first searches for axioms with a conjunction H as subconcept, that contains negative literals. In case such an axiom H was found the implementation searches for each negative literal $\neg A$ in H for the positive literal A in the superconcept $N \sqcup A$ of the axiom. If the superconcept contains the positive literal the implementation creates a new axiom with the same subconcept as the origin axiom and the superconcept of the origin axiom without the positive literal. Note that used active axioms are not removed from the list of active axioms, which allows the possibility of redundant axiom creation. In order to tackle this problem I created an alternative rule implementation which checks before handling an active axiom whether it was already used by this rule. To achieve this, the rule stores all used axioms in a set. This set can become very large and in worst case can contain all active axioms. That's why I made it an alternative implementation in order to allow for a time-optimized version of the algorithm and a space-optimized version of the algorithm.

$$\textbf{3.5.3} \quad \textbf{The rule } \mathbf{R}^{\mathbf{n}}_{\sqcap} \coloneq \frac{\{H \sqsubseteq N_i \sqcup A_i\}_{i=1}^n}{H \sqsubseteq \bigsqcup_{i=1}^n N_i \sqcup M} : \textstyle \bigcap_{i=1}^n A_i \sqsubseteq M \in \mathcal{O}$$

The rule 3.5.3 is the first rule which uses axioms from the ontology. On initialization of the rule first the ontology, which is passed to it, gets filtered for axioms of the form $\prod_{i=1}^n A_i \subseteq M$ These axioms get saved into a list of tuples where the first element of the tuple is the conjunction $\prod_{i=1}^{n} A_i$ stored as a set $\{A_1 \dots A_n\}$ and the second element is the entire axiom $\prod_{i=1}^n A_i \subseteq M$, which will not only be used to get its superconcept for creation of a new axiom, but also in order to pass it as ontology premise to the ProofHandler. On application of the rule the following function $m := N_{\mathcal{ALCH}} \times N_C \to \operatorname{list}(\{N \mid N \subseteq N_C\})$ is defined by using the current active axioms. First all active axioms get filtered for axioms of the form $H \subseteq N \sqcup A$, where H is a conjunction of literals, N a disjunction of concept names and A an atom. Every axiom is then used to define the function m in the following way: For each active axiom $\alpha = H \sqsubseteq N \sqcup A : m(H,A) := \langle \{N_c \mid N_c \in N\} \mid \rangle$. Note that there can be multiple active axioms with the same subconcept H and the same atom A but with different N_1 and N_2 and therefore m maps to a list of sets. After the function is defined it will be used to compose derived axioms. In order to do so the implementation first iterates through all ontology axioms and checks whether the LHS of a given axiom is subsumed by A. If this is the case, m can be used to find for all A_i in the matched ontology axiom the corresponding N_i and in the end the new axiom $H \subseteq \bigcup_{i=1}^n N_i \sqcup M$ is created. Note that due to m mapping to a list multiple axioms can be created.

$$\textbf{3.5.4} \quad \textbf{The rule } \mathbf{R}^+_\exists \coloneq \frac{H \sqsubseteq N \sqcup A}{H \sqsubseteq N \sqcup \exists r.B} : A \sqsubseteq \exists r.B \in \mathcal{O}$$

The rule (4) also uses axioms from the ontology so on creation of the rule implementation the ontology gets filtered for axioms of the form $A \sqsubseteq \exists r.B \in \mathcal{O}$. When the rule implementation is applied, it matches to every applicable ontology axioms all active axioms for axioms of the form $H \sqsubseteq N \sqcup A$ and creates a new axiom $H \sqsubseteq N \sqcup \exists r.B$ in case such a match is found.

$$\mathbf{3.5.5} \qquad \mathbf{R}_{\exists}^{-} : - \qquad \frac{H \sqsubseteq M \sqcup \exists r.K, \quad K \sqsubseteq N \sqcup A}{H \sqsubseteq M \sqcup B \sqcup \exists r.(K \sqcap \neg A)} : \\ A \sqsubseteq \forall s.B \in \mathcal{O} \quad r \sqsubseteq_{\mathcal{O}} s$$

The rule (5) not only uses axioms from the ontology but also uses role inclusions that follow from the ontology. The role inclusions are calculated separately since rule (7) also uses them. So the only thing that has to be done on creation is, that axioms of the form $\exists s.A \sqsubseteq B$ need to be filtered and afterwards fitting role inclusions are searched for and safed together in a list. On rule application the rule implementation then iterates through all found ontology axioms and role inclusions and searches for active axioms of the form $K \sqsubseteq N \sqcup A$. If such an axiom is found one then searches for active axioms of the form $H \sqsubseteq M \sqcup \exists r.K$. by using the just found subconcept K. If such an axiom is found the new axiom

 $H \sqsubseteq M \sqcup B \sqcup \exists r.(K \sqcap \neg A)$ is created.

3.5.6 The rule
$$\mathbf{R}_{\exists}^{\perp} := \frac{H \sqsubseteq M \sqcup \exists r.K, \quad K \sqsubseteq \bot}{H \sqsubseteq M}$$

The rule (6) does not use axioms from the original ontology. Usually there are way more axioms that contain existential restrictions in ontologies, which I used for testing, than axioms with the bottom concept on the right-hand side, and therefore the implementation first filters for axioms of the form $K \sqsubseteq \bot$ safes them in a list l. Afterwards the rule implementation filters for rules of the form $H \sqsubseteq M \sqcup \exists r.K$ and checks if an axiom of this form is found, if the K already occures in some LHS of an axiom in l. From this found rule then the new axiom $H \sqsubseteq M$ can then be created immediately.

3.5.7 The rule R_{\forall}

The rule (7) is no the final rule now again uses from the ontology following role inclusions and also axioms from the ontology of the form $A \sqsubseteq \forall s.B$. Therefore, on creation of the rule implementation first the ontology gets filtered for such axioms, where the role s is on the right-hand side of some implied role inclusion. On application of the rule implementation first searches for active axioms of the form $H \sqsubseteq M \sqcup \exists r.K$ where r has to match some RHS of some matched role inclusion. The found axioms are then used to define a function $m_l := N_{\rightarrow}$

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Example 1. Let \mathcal{O} = \{B \sqsubseteq \exists r.D, A \sqsubseteq \forall s.E, \exists r.E \sqsubseteq C, r \sqsubseteq s\} then a proof for A \sqcap B \sqsubseteq C would look like the following: (((), A \sqcap B \sqsubseteq B, \mathbf{R}_{\mathbf{A}}^+), ((A \sqcap B \sqsubseteq B, B \sqsubseteq \exists r.D), A \sqcap B \sqsubseteq \exists r.D, \mathbf{R}_{\exists}^+), ((A \sqcap B \sqsubseteq \exists r.D, A \sqsubseteq \forall s.E, r \sqsubseteq s), A \sqcap B \sqsubseteq \exists r.(D \sqcap E), \mathbf{R}_{\forall}), (((), (D \sqcap E) \sqsubseteq E, \mathbf{R}_{\mathbf{A}}^+), ((A \sqcap B \sqsubseteq \exists r.(D \sqcap E), (D \sqcap E) \sqsubseteq E, \exists r.E \sqsubseteq C), A \sqcap B \sqsubseteq \exists r.(D \sqcap E \sqcap \neg E) \sqcup C, \mathbf{R}_{\exists}^-), (((), D \sqcap E \sqcap \neg E \sqsubseteq E), D \sqcap E \sqcap \neg E \sqsubseteq L, \mathbf{R}_{\mathbf{A}}^-), ((A \sqcap B \sqsubseteq \exists r.(D \sqcap E \sqcap \neg E) \sqcup C, D \sqcap E \sqcap \neg E \sqsubseteq L), A \sqcap B \sqsubseteq C, \mathbf{R}_{\exists}^+))
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Notes that in the algorithm not only these concepts would be created but also other rules. For example even in the first step not only $A \sqcap B \sqsubseteq B$ would have been created but also $A \sqcap B \sqsubseteq A$.

4 Results

4.1 Setup

In order to compare the performance of the ALCH-Reasoner in terms of time consumption and also readability of the proofs, I used Lethe [5], as a reference. Lethe is integrated into EVEE and I use the LetheBasedALCHProofGenerator

Task	Time (ms)	#Axioms	Size largest Premise	$\# ext{RuleApplications}$
00001	192	19	3	20
00003	5913	23	4	27
00008	133	15	3	18
00009	119	25	4	27
00012	24	7	2	7

Table 1: Performance of the ALCH-Reasoner

Task	Time (ms)	# A xioms	Size largest Premise	$\# ext{RuleApplications}$
00001	2786	13	2	13
00003	502	5	2	5
00008	2728	8	2	8
00009	1348	11	2	11
00012	159	2	1	2

Table 2: Performance of the Lethe

to access it. Lethe is a uniform interpolation (UI) tool which uses a resolution based calculus to compute proofs for different DLs up to \mathcal{ALCH} .

The dataset I use to compare the two approaches is a dataset created in [2] It basically consists of justifications for axioms in the Bioportal-Repository, which where renamed, resulting in so-called justification patterns. Some of these patterns are quite challenging to compute proofs for, which is why I chose only a small subset of the dataset. Also, some of the tasks use domain specifications which I do not support.

4.2 Proofs Shape Comparison

When comparing the proofs generated by the ALCH-Reasoner and the Lethe-Reasoner, one can see that the proofs differ quite a bit. Also, the number of axioms and rule applications which are used in the proofs of Lethe are always less than the once of the ALCH-Reasoner, as can be seen in table 1 and 2.

I see 2 reasons for this. The first reason is that the Lethe-Reasoner uses a different set of rules than the ALCH-Reasoner. While for example the ALCH-Reasoner uses only the $\mathbf{R}^{\mathbf{n}}_{\sqcap}$ rule on a normalized task 1 ontology, the Lethe-Reasoner uses quite a few different types of elimination rules. Also, the interaction between the rules $\mathbf{R}^{-}_{\mathbf{A}}, \mathbf{R}^{-}_{\exists}$, $\mathbf{R}^{\perp}_{\exists}$, as seen in figure 5, is quite unintuitive and also blows up the proof. This behavior becomes very clear when comparing it to figure 6

The second reason is that the structural transformation introduces new concept names for the same concept. For example for the very simple ontology $\mathcal{O} = \{A \sqsubseteq B, B \sqsubseteq C\}$ and the GCI $\alpha = A \sqsubseteq C$ one first gets the normalized ontology $\operatorname{st}(\mathcal{O}') = \{[A] \sqsubseteq [B], A \sqsubseteq [A], [B] \sqsubseteq B, [B] \sqsubseteq [C], B \sqsubseteq [B], [C] \sqsubseteq C\}$ the proof in 3 of the GCI α there exist A and [A] which reference to the same

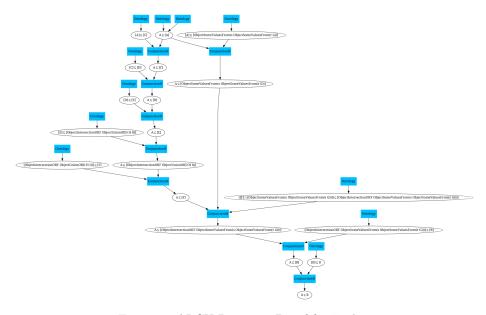


Figure 1: ALCH-Reasoner Proof for Task 1

concept. This results in the proof being unnecessarily long and little less readable, since we always need an additional step to switch from A to [A] using the $\mathbf{R}^{\mathbf{n}}_{\sqcap}$ rule. Note that this only introduces one more step in the proof and could be prevented by querying for the $[A] \sqsubseteq [B]$ instead. But the structural transformation introduces also unnecessary in order to unfold the structural transformed concepts into there semantic valuable representation. For example $A \sqsubseteq [B \sqcap C]$ must first be transformed into $A \sqsubseteq [B] \sqcap [C]$ using the $\mathbf{R}^{\mathbf{n}}_{\sqcap}$ rule before it can interact with axioms like $[B] \sqcap [C] \sqsubseteq [B]$. This behavior can also be found in figure 4. On could also think about removing such steps after the reasoning in order to improve the readability of proofs.

4.3 Proofs Time Performance Comparison

It is interesting to see that in most cases the ALCH-Reasoner is actually around 10 times faster. When taking a look at task 3 one can see that the rule \mathbf{R}_{\exists}^- is used quite often. I analyzed the behavior of the ALCH-Reasoner and found out that the rule \mathbf{R}_{\exists}^- is actually applied way more time than any other rule. This is due to the fact that the ALCH-Reasoner does not prune the search and therefore has to check every option again and again. This behavior is also the case for the rule $\mathbf{R}_{\sqcap}^{\mathbf{n}}$ which produces a lot of combinations in case there is a deep disjunction on the right and side of a concept. I also checked this behavior when trying to handle task 13 which contains the axiom $A \equiv ((C \sqcap D) \sqcup (C \sqcap E) \sqcup (D \sqcap E) \sqcup (F \sqcap E) \sqcup (C \sqcap F) \sqcup (D \sqcap F))$. Unfortunately, this task is too big for the ALCH-Reasoner to handle.

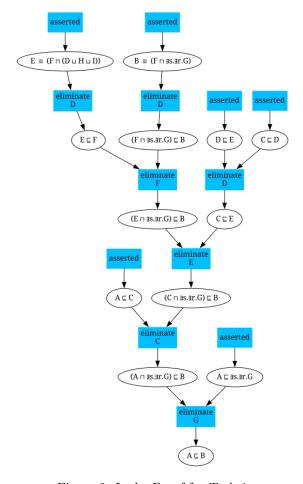


Figure 2: Lethe Proof for Task 1

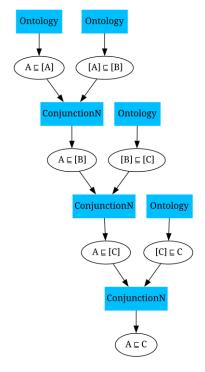


Figure 3: ALCH simple proof

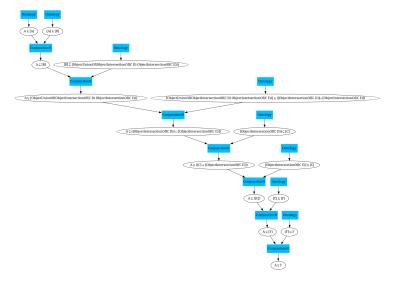


Figure 4: ALCH unfolding example

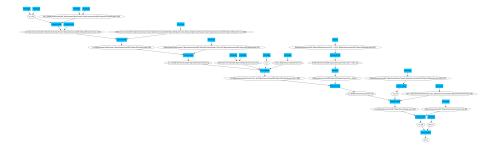


Figure 5: ALCH-Reasoner task 3

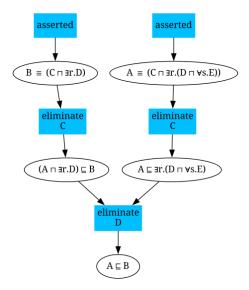


Figure 6: Lethe task 3

5 Conclusion

In conclusion is the ALCH-Reasoner quite fast in most cases and also produces in most cases quite readable proofs. The ALCH-Reasoner even outperforms Lethe on small, examples which do not contain big conjunctions or disjunctions nor very complex existential restrictions. Although there are some things which could be improved. First one could investigate how to efficiently remove unnecessary reasoning steps. And secondly there could be way more pruning when searching for axioms of specific forms to improve the performance. If someone wants to tackle these problems the code and also datasets can be found under 1 A major flaw probably can not be improved upon is the unintuitive behavior between the rules ${\bf R}_{\bf A}^{-}, {\bf R}_{\exists}^{-}$. All in all is my implementation is quite fast and also quite readable.

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 $^{^{1}} https://github.com/StBreuer/ALCH-Reasoner.git \\$