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An Approach for Optimizing \mathcal{AE} -Approximation of \mathcal{AC} -Concepts

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Abstract

An approximation of an \mathcal{AUC} -concept by an \mathcal{AUE} -concept can be computed in double exponential time [4]. Consequently, one needs powerful optimization techniques for approximating an entire unfoldable TBox. Addressing this issue we identify a special form of \mathcal{AUC} -concepts, which can be divided into parts s.t. each part can be approximated independently. This independent approximation in turn facilitates caching during the computation of approximation.

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1 Motivation

This report presents some preliminary results on optimization techniques for the computation of approximations. Approximation is a new non-standard inference service in Description Logics (DLs) introduced in [4]. Approximating a concept, defined in one DL, means to translate this concept to another concept, defined in a second typically less expressive DL, such that both concepts are as closely related as possible with respect to subsumption. Like other non-standard inferences such as computing the least common subsumer (lcs) or matching of concepts, approximation has been introduced to support the construction and maintenance of DL knowledge-bases (see [8, 5]). Approximation has a number of different applications some of which we will mention here, see [4] for others.

Computation of commonalities of concepts. Typically, the lcs is employed to accomplish this task. In case the DL \mathcal{L} provides concept disjunction, the lcs is just the disjunction of C_1 and C_2 ($C_1 \sqcup C_2$). Thus, a user inspecting this concept does not learn anything about the commonalities between C_1 and C_2 . By using approximation, however, one can make the commonalities explicit to some extent by first approximating C_1 and C_2 in a sublanguage of \mathcal{L} which does not provide disjunction, and then computing the lcs of the approximations in \mathcal{L} .

Translation of knowledge-bases. Approximation can be used to (automatically) translate a knowledge-base written in an expressive DL into a another (semantically closely related) knowledge-base in a less expressive DL. The translation may become necessary to port knowledge-bases between different knowledge representation systems or to integrate different knowledge-bases.

We investigate the case of translating an unfoldable \mathcal{ACC} -TBox into an \mathcal{ACE} -TBox by computing the approximation of each concept defined in the \mathcal{ACC} -TBox. In [4], a first in-depth investigation of the approximation inference has been presented. Particularly, a double-exponential time algorithm has been devised to approximate \mathcal{ACC} -concepts by \mathcal{ACE} -concepts. Consequently, approximating an entire TBox requires substantial optimizations. We address this problem by identifying a form of \mathcal{ACC} -concept descriptions whose conjuncts can be approximated independently. This approach does not only speed-up the computation of a single approximation, but also allows to re-use an obtained approximation in subsequent approximations. The obtained approximations can directly be inserted into the approximations of such a conjunction. The identification of conjunctions with conjuncts that can be approximated independently is thereby a prerequisite for applying caching techniques to approximation.

Syntax	Semantics	ALE	ALC
Т	Δ	X	X
$C\sqcap D$	$C^I \cap D^I$	X	X
$\exists r.C$	$\{x \in \Delta \mid \exists y : (x, y) \in r^I \land y \in C^I\}$	X	X
$\forall r.C$	$\{x \in \Delta \mid \forall y : (x, y) \in r^I \to y \in C^I\}$	X	X
$\neg A, A \in N_C$	$\Delta \setminus A^I$	X	X
上	\emptyset	X	X
$C \sqcup D$	$C^I \cup D^I$		X
$\neg C$	$\Delta \setminus C^I$		X

Table 1: Syntax and semantics of concept descriptions.

2 Preliminaries

As usual concept descriptions are inductively defined based on a set of concept constructors, starting with a set N_C of concept names and a set N_R of role names. In this report, we consider concept descriptions built from the constructors shown in Table 1. Note that in \mathcal{ALC} every concept description can be negated whereas in \mathcal{ALE} negation is only allowed in front of concept names. For a DL \mathcal{L} , such as \mathcal{ALE} and \mathcal{ALC} , a concept description formed with the constructors allowed in \mathcal{L} is called \mathcal{L} -concept description in the following.

As usual, the semantics of a concept description is defined in terms of an interpretation $\mathcal{I} = (\Delta, \cdot^I)$. The domain Δ of \mathcal{I} is a non-empty set and the interpretation function \cdot^I maps each concept name $A \in N_C$ to a set $A^I \subseteq \Delta$ and each role name $r \in N_R$ to a binary relation $r^I \subseteq \Delta \times \Delta$. The extension of \cdot^I to arbitrary concept descriptions is defined inductively, as shown in Table 1.

For the sake of simplicity, we assume that the set N_R of role names is the singleton $\{r\}$. However, all definitions and results can easily be generalized to arbitrary sets of role names. We also assume that each conjunction in an \mathcal{ALE} -concept description contains at most one value restriction of the form $\forall r.C'$ (this is w.l.o.g. due to the equivalence $\forall r.E \sqcap \forall r.F \equiv \forall r.(E \sqcap F)$).

A TBox is a finite set of concept definitions of the form $A \doteq C$, where $A \in N_C$ and C is a concept description. In addition, we require that TBoxes are unfoldable, i.e. they are acyclic and do not contain multiple definitions (see, e.g., [9]). Concept names occurring on the left-hand side of a definition are called defined concepts. All other concept names are called primitive concepts. In TBoxes of the DL \mathcal{ALE} , negation may only be applied to primitive concepts. An interpretation \mathcal{I} is a model of the TBox \mathcal{T} iff it satisfies all its concept definitions, i.e., $A^{\mathcal{I}} = C^{\mathcal{I}}$ for all definitions $A \doteq C$ in \mathcal{T} .

One of the most important traditional inference services provided by DL systems is computing the subsumption hierarchy. The concept description C is subsumed by the description D ($C \sqsubseteq D$) iff $C^I \subseteq D^I$ holds for all interpretations \mathcal{I} ; C and D are equivalent ($C \equiv D$) iff $C \sqsubseteq D$ and $D \sqsubseteq C$. Subsumption and

equivalence in \mathcal{ALC} is PSPACE-complete [10] and NP-complete in \mathcal{ALE} [6].

2.1 \mathcal{ALE} -Approximation for \mathcal{ALC}

In order to approximate \mathcal{ALC} -concept descriptions by \mathcal{ALE} -concept descriptions, we need to compute the lcs in \mathcal{ALE} .

Definition 1 Given \mathcal{L} -concept descriptions C_1, \ldots, C_n with $n \geq 2$ for some description logic \mathcal{L} , the \mathcal{L} -concept description C is the least common subsumer (lcs) of C_1, \ldots, C_n ($C = \mathsf{lcs}(C_1, \ldots, C_n)$ for short) iff (i) $C_i \sqsubseteq C$ for all $1 \leq i \leq n$, and (ii) C is the least concept description with this property, i.e., if C' satisfies $C_i \sqsubseteq C'$ for all $1 \leq i \leq n$, then $C \sqsubseteq C'$.

As already mentioned, in \mathcal{ALC} the lcs trivially exists since $lcs(C, D) \equiv C \sqcup D$. For \mathcal{ALE} the existence is not obvious. It was shown in [2], that the lcs of two or more \mathcal{ALE} -concept descriptions always exists, that its size may grow exponentially in the size of the input descriptions, and that it can be computed in exponential time.

Intuitively, to approximate an \mathcal{ACC} -concept description from "above" means to compute an \mathcal{ACE} -concept description that is more general than the input concept description but minimal w.r.t. subsumption.

Definition 2 Let \mathcal{L}_1 and \mathcal{L}_2 be two DLs, and let C be an \mathcal{L}_1 - and D be an \mathcal{L}_2 concept description. Then, D is called an upper \mathcal{L}_2 -approximation of C (D =
approx $_{\mathcal{L}_2}(C)$ for short) iff (i) $C \sqsubseteq D$, and (ii) D is minimal with this property,
i.e., $C \sqsubseteq D'$ and $D' \sqsubseteq D$ implies $D' \equiv D$ for all \mathcal{L}_2 -concept descriptions D'.

Although defined in [4] lower approximations are not yet further investigated. In this report, we restrict our investigations to upper \mathcal{ALE} -approximations of \mathcal{ALC} -concept descriptions. Therefore, whenever we speak of approximations, we mean upper \mathcal{ALE} -approximations. Thus, having defined approximation we turn now to how to actually compute them.

2.2 The Approximation Algorithm

Before a defined concept from a TBox can be approximated it has to be unfolded w.r.t. the underlying TBox to make the information captured in the concept definitions explicit. To this end, every defined concept is replaced by the concept description on the right-hand side of its concept definition until no defined concept occurs in the concept description. It is well known that this process can cause an exponential blow-up of the concept description, see [9]. To recapitulate the approximation algorithm presented in [4], we need to introduce \mathcal{ACC} -normal forms and some notation for accessing the different parts of a concept.

For an unfolded concept description C the role-depth rd(C) is inductively defined as follows:

$$\begin{split} rd(N) &:= 0 & \text{, where } N \in N_C \cup \{\bot, \top\} \\ rd(\neg C) &:= rd(C) \\ rd(C_1 \ \rho \ C_2) &:= \max\{rd(C_1), rd(C_2)\} & \text{, where } \rho \in \{\sqcap, \sqcup\} \\ rd(Qr.C) &:= 1 + rd(C) & \text{, where } Q \in \{\exists, \forall\} \end{split}$$

A role-level of a concept C is the set of all concept descriptions occurring on the same role-depth in C. The topmost role-level of a concept description is also called its top-level.

We call a concept description top-level \sqcup -free if it is in negation normal form (NNF), i.e., negation is pushed inwards until in front of a concept name, and does not contain any disjunction on top-most role-level. Some notation is needed to access the different parts of an \mathcal{ALE} -concept description or a top-level \sqcup -free \mathcal{ALC} -concept description C:

- prim(C) denotes the set of all (negated) concept names and the bottom concept occurring on the top-level of C;
- $\operatorname{\mathsf{val}}_{\mathsf{r}}(C) := C_1 \sqcap \cdots \sqcap C_n$, if there exist value restrictions of the form $\forall r.C_1, \ldots, \forall r.C_n$ on the top-level of C; otherwise, $\operatorname{\mathsf{val}}_{\mathsf{r}}(C) := \top$;
- $ex_r(C) := \{C' \mid \text{ there exists } \exists r.C' \text{ on the top-level of } C\}.$

Equipped with these we can define the \mathcal{AC} -normal form where conjuncts are distributed over the disjuncts. An arbitrary \mathcal{AC} -concept description is transformed into a concept description with at most one disjunction on top-level of every concept of each role-level.

Definition 3 An \mathcal{ACC} -concept description C is in \mathcal{ACC} -normal form iff

- 1. if $C \equiv \bot$, then $C = \bot$; if $C \equiv \top$, then $C = \top$;
- 2. otherwise, C is of the form $C = C_1 \sqcup \cdots \sqcup C_n$ with

$$C_i = \mathop{\textstyle\prod}_{A \in \operatorname{prim}(C_i)} A \sqcap \mathop{\textstyle\prod}_{C' \in \operatorname{ex}_{\operatorname{r}}(C_i)} \exists r.C' \sqcap \forall r.\operatorname{val}_{\operatorname{r}}(C_i),$$

 $C_i \not\equiv \bot$, and $\mathsf{val}_\mathsf{r}(C_i)$ and every concept description in $\mathsf{ex}_\mathsf{r}(C_i)$ is in \mathcal{ALC} -normal form, for all $i = 1, \ldots, n$.

Every disjunct of a concept in \mathcal{AU} -normal form is top-level \sqcup -free. Obviously, every \mathcal{AU} -concept description can be turned into an equivalent concept description in \mathcal{AU} -normal form. Unfortunately, this may take exponential time, as the example $(A_1 \sqcup A_2) \sqcap \cdots \sqcap (A_{2n-1} \sqcup A_{2n})$ shows whose \mathcal{AU} -normal form is of size exponential in n.

```
Input: \mathcal{ALC}-concept description C
Output: upper \mathcal{ALE}-approximation of C

1. If C \equiv \bot, then \operatorname{c-approx}_{\mathcal{ALE}}(C) := \bot; if C \equiv \top, then \operatorname{c-approx}_{\mathcal{ALE}}(C) := \top

2. Otherwise, transform C into \mathcal{ALC}-normal form C_1 \sqcup \cdots \sqcup C_n and return \operatorname{c-approx}_{\mathcal{ALE}}(C) := \Box
\Box \bigcap_{A \in \bigcap_{i=1}^n \operatorname{prim}(C_i)} A \sqcap \bigcap_{(C'_1, \dots, C'_n) \in \operatorname{ex}_r(C_1) \times \dots \times \operatorname{ex}_r(C_n)} \exists r.\operatorname{lcs}\{\operatorname{c-approx}_{\mathcal{ALE}}(C'_i \sqcap \operatorname{val}_r(C_i)) \mid 1 \leq i \leq n\} \sqcap \forall r.\operatorname{lcs}\{\operatorname{c-approx}_{\mathcal{ALE}}(\operatorname{val}_r(C_i)) \mid 1 \leq i \leq n\}
```

Figure 1: The recursive algorithm $\operatorname{c-approx}_{\mathcal{ALE}}(C)$.

The approximation algorithm displayed in Figure 1 checks if the input is a concept equivalent to \top or \bot —in this case the approximation is trivial—otherwise it proceeds recursively on the \mathcal{AC} -normal form of the input and extracts the commonalities of all disjuncts. Unfortunately, the algorithm needs double-exponential time for arbitrary \mathcal{AC} -concepts in the worst case. Despite of its high complexity, our prototypical implementation of the algorithm showed a quite promising performance in respect to run-time and resulting concept sizes when applied it to concept descriptions derived from a chemical process engineering application, for details see [4].

3 Optimizing \mathcal{ALE} -Approximations

A TBox can be translated by computing the approximation of every unfolded concept description appearing on the right-hand side of a concept definition in the TBox. Each defined concept has to be unfolded and transformed into \mathcal{AU} -normal form before the approximation algorithm can be applied. Unfortunately, both of these steps cause exponential growth of the concept description.

For standard reasoning tasks [1, 7] and also for the computation of the lcs [3] the first source of complexity can often be alleviated by *lazy unfolding*. Here the idea is to replace a defined concept in a concept description only if examination of that part of the description is necessary. Lazy unfolding unfolds all defined concepts appearing on the top-level of the concept description under consideration while defined concepts on deeper role-levels remain unchanged as long as possible.

When computing the lcs the main benefit of lazy unfolding is that in some cases defined concepts can be used directly in the lcs concept description. If, for example a defined concept C appears in all input concept descriptions on the same role-level, the concept definition of C does not need to be processed, but C can be inserted into the lcs directly, see [3] for details. In the case of approximation, however, this effect of lazy unfolding can not be utilized even if a defined concept is obviously common to all disjuncts. For example, in $(A \sqcap C) \sqcup (C \sqcap (\neg B))$ the concept name C cannot be used directly as a name in the approximation because the \mathcal{ALC} -concept description C stands for must be approximated. Thus unfolding a concept completely can in principle not be avoided for approximation.

The double-exponential time complexity of the approximation algorithm, however, suggests another approach to optimization. Instead of approximating an input concept C as a whole a significant amount of time could be saved by splitting C into its conjuncts and approximating them separately. If, for instance, C consists of two conjuncts of size n then the approximation of C takes some $a^{b^{2n}}$ steps while the conjunct-wise approach would just take $2a^{b^n}$. Unfortunately, splitting an arbitrary input concept at conjunctions leads to incorrect approximations, as examples show [4]. In the following section we will therefore introduce a class of so-called nice \mathcal{AC} -concepts for which the conjunct-wise approximation still produces the correct result.

3.1 Nice Concepts

In the following we assume that all concept descriptions are unfolded and in NNF. For an \mathcal{AU} -concept description C and $i \in \mathbb{N}$ the quantor set $Q_r(C,i)$ denotes the set of quantors used on the role-level i of C (referring to role r). Hence, for $0 \le i \le rd(C)$ the quantor set $Q_r(C,i)$ is a nonempty subset of $\{\forall,\exists\}$. Similarly, the name set $N_r(C,i)$ denotes the atomic concepts used on a specific role-level. Formally, Q and N are defined as follows.

Definition 4 Let $C := \bigsqcup_{i=1}^k C_i$ be an \mathcal{ALC} -concept description in \mathcal{ALC} -normal form. For $d \in \mathbb{N}$, the sets $Q_r(C, d)$ and $N_r(C, d)$ are inductively defined by:

$$\bullet \ Q_r(C,0) := \{\exists \mid \bigcup_{i=1}^k \mathsf{ex_r}(C_i) \neq \emptyset\} \cup \{\forall \mid \sqcap_{i=1}^k \mathsf{val_r}(C_i) \not\equiv \top\}$$

$$N_r(C,0) := \bigcup_{i=1}^k \mathsf{prim}(C_i)$$

$$\bullet \ Q_r(C,d+1) := \bigcup_{i=1}^k \ \bigcup_{C' \in \mathsf{ex}_\mathsf{r}(C_i)} Q_r(C',d) \cup \bigcup_{i=1}^k Q_r(\mathsf{val}_\mathsf{r}(C_i),d)$$

$$N_r(C,d+1) := \bigcup_{i=1}^k \ \bigcup_{C' \in \mathsf{ex}_\mathsf{r}(C_i)} N_r(C',d) \cup \bigcup_{i=1}^k N_r(\mathsf{val}_\mathsf{r}(C_i),d).$$

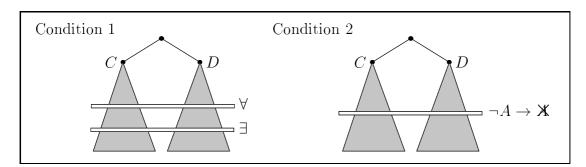


Figure 2: What nice concepts look like

For a concept C not in \mathcal{ALC} -normal form, Q and N are defined in terms of the \mathcal{ALC} -normal form of C. For example the unfolded concept $C = (\exists r.(A \sqcap B) \sqcap \forall r.(D \sqcup (\exists r.\neg E)))$ has the set of quantors $Q_r(C,0) = \{\forall,\exists\}, Q_r(C,1) = \{\exists\}$ and $Q_r(C,i) = \emptyset$ for $i \geq 2$. For the name set, we have $N(C,0) = \emptyset$, $N(C,1) = \{A,B,D\}$, and $N(C,2) = \{\neg E\}$.

We are now ready to specify in detail what nice concepts are. In general, an approximation $approx_{\mathcal{AE}}(C \sqcap D)$ cannot be split at the conjunction because of possible interactions between existential and value restrictions on the one hand and inconsistencies induced by negation on the other. For example, the approximation $approx_{\mathcal{AE}}(\exists r. \top \sqcap (\forall r. A \sqcup \exists r. A))$ yields $\exists r. A$ while the split version $approx_{\mathcal{AE}}(\exists r. \top) \sqcap approx_{\mathcal{AE}}(\forall r. A \sqcup \exists r. A)$ only produces $\exists r. \top$. Similarly, the conjunction $A \sqcap (\neg A \sqcup B)$ cannot be approximated separately.

We now call those concepts nice for which this simplified strategy still produces the correct result and for which a simple syntactic discrimination rule exists. Firstly, the role quantors occurring in nice concepts are restricted to one type per role level. Hence, on every role level of the syntax tree of a nice concept either no ∀-restrictions or no ∃-restrictions occur. Secondly, a concept name and its negation may not occur on the same same level of the syntax tree of a nice concept. Consider Figure 2 for an illustration of these rules.

Formally, we can define nice concepts by means of the syntactical operators from Definition 4.

Definition 5 Let C be an \mathcal{AC} -concept description in NNF. Then C is nice iff for every $d \in \mathbb{N}$ it holds that

- 1. $|Q_r(C,d)| \leq 1 \ and$
- 2. $N_r(C,d)$ does not contain a concept name and its negation.

It remains to be shown that nice concepts as defined above in fact have the desired property. In preparation for this we firstly present a simple set-theoretic result which later on will allow us to reduce the number of existential restrictions computed in an approximation of a nice concept.

Lemma 6 Let $m, n \in \mathbb{N}$. For every $i \in \{1, ..., m\}$ and $j \in \{1, ..., n\}$, let A_i and B_j be arbitrary finite sets, let $U_{ij} := A_i \cup B_j$, and let $u_{ij} \in U_{ij}$. Denote by U the set of all u_{ij} , i.e., $U := \{u_{ij} \mid 1 \leq i \leq m, 1 \leq j \leq n\}$. Then one of the following claims holds: either, for every i there exist elements $a_i \in A_i$ with $\{a_i \mid 1 \leq i \leq m\} \subseteq U$; or, for every j there exist $b_j \in B_j$ with $\{b_j \mid 1 \leq j \leq n\} \subseteq U$.

For all $j \in \{1, ..., m\}$ and all $j \in \{1, ..., n\}$ consider arbitrary $u_{ij} \in U_{ij}$. Assume that the second claim for the sets $B_1, ..., B_n$ does not hold. Then there is one j' with $B_{j'} \cap U = \emptyset$, otherwise $b_{j'}$ could be chosen from this intersection to satisfy the claim. Since $u_{ij'} \in A_i \cup B_{j'}$ for all i it follows that $u_{ij'} \in A_i$ for all i, satisfying the first claim for $A_1, ..., A_m$.

The choice of sets in the unions $A_i \cup B_j$ in the above lemma corresponds to tuples in the product $\{A_1, \ldots, A_m\} \times \{B_1, \ldots, B_n\}$. The claim can be generalized to n-ary products where every union corresponds to a tuple from $\{S_{11}, \ldots, S_{1k_1}\} \times \cdots \times \{S_{n1}, \ldots, S_{nk_n}\}$. The following lemma provides the more general result.

Lemma 7 For every $1 \le i \le n$ and $1 \le j_i \le k_i$, let S_{ij_i} be an arbitrary set. For every tuple \bar{t} in the set $T := \{1, \ldots, k_1\} \times \cdots \times \{1, \ldots, k_n\}$, denote by $U_{\bar{t}}$ the union $\bigcup_{i=1}^n S_{i\bar{t}(i)}$ (with $\bar{t}(i)$ denoting the ith component of \bar{t}). For every $\bar{t} \in T$, let $u_{\bar{t}} \in U_{\bar{t}}$. Let $U := \{u_{\bar{t}} \mid \bar{t} \in T\}$. Then there exists an index $i \in \{1, \ldots, n\}$ and elements $s_{ij} \in S_{ij}$ for $1 \le j \le k_i$ such that the set $\{s_{ij} \mid 1 \le j \le k_i\}$ is a subset of U.

Consider arbitrary elements $u_{\bar{t}}$. Assume that no index i has the claimed property. Then for every i there exists some j'_i such that $S_{ij'_i} \cap U = \emptyset$, otherwise appropriate elements from every $S_{ij'_i}$ could be found. Consider the union $U_{\bar{t}}$ where $\bar{t}(i) = j'_i$ for every i. Since $u_{\bar{t}} \in U_{\bar{t}}$ it holds that there is some i' with $u_{\bar{t}} \in S_{i'j'_{i'}}$. Clearly, $u_{\bar{t}} \in U$, since $\bar{t} \in T$. But then $S_{i'j'_{i'}} \cap U \neq \emptyset$, in contradiction to the assumption.

By means of the above lemma we can show that the least common subsumer of sets of nice \mathcal{ALE} -concepts of a certain form can be simplified. The following example motivates the relevant case. The \mathcal{ALC} -normal form of a nice concept of the form $(C_1 \sqcup \cdots \sqcup C_m) \sqcap (D_1 \sqcup \cdots \sqcup D_n)$ with no further disjunction on the toplevel of all C_i and D_j results in a disjunction of the form $\coprod_{i,j} C_i \sqcap D_j$. Assume that all subconcepts C_i , D_j have only existential restrictions on toplevel. For the approximation of this concept, computing the resulting existential restrictions requires to compute the least common subsumer of (the approximation of) every combination of existential restrictions from the relevant disjuncts. Thus, for every pair (i,j), every existential restriction $E_{ij} \in ex_r(C_i \sqcap D_j)$ is approximated, then the lcs over all $\{approx_{\mathcal{ALE}}(E_{ij}) \mid i,j\}$ is computed. Since $ex_r(C_i \sqcap D_j)$ equals the union $ex_r(C_i) \cup ex_r(D_j)$, the previous lemma can be employed to restrict the lcs to a much smaller set. The following lemma provides the exact proof.

Lemma 8 For $1 \leq i \leq 2$, let C_i and D_i be \mathcal{ALE} -concept descriptions such that $C_1 \sqcap C_2 \sqcap D_1 \sqcap D_2$ is a nice concept. Then it holds that $\mathsf{lcs}\{C_i \sqcap D_j \mid i, j \in \{1,2\}\} \equiv \mathsf{lcs}\{C_1,C_2\} \sqcap \mathsf{lcs}\{D_1,D_2\}$.

Proof by induction over the maximum role-depth d of all C_i , D_i .

- d = 0
 - Then $C_i = \prod_{A \in \mathsf{prim}(C_i)} A$ and $D_j = \prod_{A \in \mathsf{prim}(D_j)} A$ for all i, j. The definition of nice guarantees that no inconsistencies can be introduced by a combination of an atomic concept and its negation. Hence, the $\mathsf{lcs}\{C_i \sqcap D_j \mid i, j \in \{1, 2\}\}$ then yields $\prod_{A \in S} A$ where S is the intersection of all sets of primitive concepts of the form $\mathsf{prim}(C_i \sqcap D_j)$. Hence, S equals $\bigcap \{\mathsf{prim}(C_i) \cup \mathsf{prim}(D_j) \mid i, j \in \{1, 2\}\}$. By distributing the intersection over the union, S can be expressed as the union $(\mathsf{prim}(C_1) \cap \mathsf{prim}(C_2)) \cup (\mathsf{prim}(D_1) \cap \mathsf{prim}(D_2))$. The conjunction $\prod_{A \in S} A$ is therefore equivalent to $\prod_{A \in \mathsf{prim}(C_1) \cap \mathsf{prim}(C_2)} A \sqcap \prod_{A \in \mathsf{prim}(D_1) \cap \mathsf{prim}(D_2)} A$. By definition of the lcs, this conjunction is equivalent to the conjunction of $\mathsf{lcs}\{C_i \mid 1 \leq i \leq 2\}$ and $\mathsf{lcs}\{D_j \mid 1 \leq j \leq 2\}$.
- *d* > 0

Depending on the role quantors on the outermost role level two cases are distinguished. In the first case, all C_i and D_i are of the form $\bigcap_{A \in \mathsf{prim}(C_i)} A \bigcap \forall r.C_i'$ and $\bigcap_{A \in \mathsf{prim}(D_j)} A \bigcap \forall r.D_j'$, respectively. Then, $\mathsf{lcs}\{C_i \bigcap D_j \mid i,j \in \{1,2\}\}$ is defined as

$$\prod_{A \in S} A \sqcap \forall r.\mathsf{lcs}\{C_i' \sqcap D_j' \mid i, j \in \{1, 2\}\},\$$

where S again equals $\bigcap \{ \mathsf{prim}(C_i) \cup \mathsf{prim}(D_j) \mid i, j \in \{1, 2\} \}$. Analogously to the case of d = 0, the set S can be expressed as the union of $\mathsf{prim}(C_1) \cap \mathsf{prim}(C_2)$ and $\mathsf{prim}(D_1) \cap \mathsf{prim}(D_2)$. Due to the induction hypothesis the lcs in the value restriction is equivalent to $\mathsf{lcs}\{C_i \mid 1 \leq i \leq 2\} \cap \mathsf{lcs}\{D_j \mid 1 \leq j \leq 2\}$. Since a conjunction in a value restriction may be split into a conjunction of value restrictions, we obtain

$$\begin{array}{c} \underset{A \in \mathsf{prim}(C_1) \cap \mathsf{prim}(C_2)}{\sqcap} A \sqcap \underset{A \in \mathsf{prim}(D_1) \cap \mathsf{prim}(D_2)}{\sqcap} A \\ \\ \sqcap \forall r. \mathsf{lcs}\{C_i' \mid 1 \leq i \leq 2\} \sqcap \forall r. \mathsf{lcs}\{D_j' \mid 1 \leq j \leq 2\}. \end{array}$$

According to the definition of the lcs, this expression can be written as

$$lcs\{C_i \mid 1 \le i \le 2\} \sqcap lcs\{D_j \mid 1 \le j \le 2\}.$$

In the second case, all C_i and D_j are of the form $\bigcap_{A \in \mathsf{prim}(C_i)} A \bigcap \bigcap_{C'_i \in \mathsf{ex}_r(C_i)} \exists r. C'_i$ and $\bigcap_{A \in \mathsf{prim}(D_j)} A \bigcap \bigcap_{D'_j \in \mathsf{ex}_r(D_j)} \exists r. D'_j$, respectively. The least common sub-

sumer $lcs\{C_i \sqcap D_j \mid i, j \in \{1, 2\}\}$ then yields

$$\begin{array}{l} \underset{A \in S}{ \textstyle \sqcap } A \\ \sqcap \qquad \qquad \qquad \qquad \qquad \qquad \\ E_1 \in \operatorname{ex_r}(C_1) \cup \operatorname{ex_r}(D_1), \\ E_2 \in \operatorname{ex_r}(C_1) \cup \operatorname{ex_r}(D_2), \\ E_3 \in \operatorname{ex_r}(C_2) \cup \operatorname{ex_r}(D_1), \\ E_4 \in \operatorname{ex_r}(C_2) \cup \operatorname{ex_r}(D_2) \end{array}$$

with S as before. It is shown in Lemma 7 that every set $\{E_i \mid 1 \leq i \leq 4\}$ in the above conjunction is a superset of a set either of the form $\{C_1', C_2'\}$ with all $C_i' \in \mathsf{ex}_\mathsf{r}(C_i)$ or of the form $\{D_1', D_2'\}$ with all $D_j' \in \mathsf{ex}_\mathsf{r}(D_j)$. Due to the monotonicity of the lcs (w.r.t. subsumption) every expression $\mathsf{lcs}\{E_i \mid 1 \leq i \leq 4\}$ is therefore more general than either $\mathsf{lcs}\{C_1', C_2'\}$ or $\mathsf{lcs}\{D_1', D_2'\}$ for appropriate existential restrictions C_1', C_2' or D_1', D_2' . Conversely, every set of the form $\{C_1', C_2'\}$ and $\{D_1', D_2'\}$ occurs in the above conjunction as one choice of E_i , $1 \leq i \leq 4$. Hence, the above existential restrictions can be simplified, yielding

$$\begin{split} & \underset{A \in S}{\sqcap} A \\ & \sqcap \underset{E_1 \in \mathsf{ex_r}(C_1)}{\sqcap} \quad \underset{E_2 \in \mathsf{ex_r}(C_2)}{\sqcap} \mathsf{lcs}\{E_1, E_2\} \\ & \sqcap \underset{E_3 \in \mathsf{ex_r}(D_1)}{\sqcap} \quad \underset{E_4 \in \mathsf{ex_r}(D_2)}{\sqcap} \mathsf{lcs}\{E_3, E_4\}. \end{split}$$

Analogously to the previous case the set S of atomic concepts can be written as a union, yielding the following conjunction

$$\begin{array}{c|c} & \prod\limits_{A \in \mathsf{prim}(C_1) \cap \mathsf{prim}(C_2)} A \sqcap \prod\limits_{A \in \mathsf{prim}(D_1) \cap \mathsf{prim}(D_2)} A \\ \sqcap \prod\limits_{E_1 \in \mathsf{ex_r}(C_1)} \prod\limits_{E_2 \in \mathsf{ex_r}(C_2)} \mathsf{lcs}\{E_1, E_2\} \\ \sqcap \prod\limits_{E_3 \in \mathsf{ex_r}(D_1)} \prod\limits_{E_4 \in \mathsf{ex_r}(D_2)} \mathsf{lcs}\{E_3, E_4\}. \end{array}$$

By definition of the lcs, this is equivalent to

$$lcs\{C_i \mid 1 \le i \le 2\} \sqcap lcs\{D_j \mid 1 \le j \le 2\}.$$

The above claim can again be generalized to larger conjunctions. Let $1 \leq i \leq n$ and $1 \leq j \leq k_i$ and let C_{ij} be \mathcal{ALE} -concepts whose overall conjunction is nice. For every tuple $\bar{t} \in \{1, \ldots, k_1\} \times \cdots \times \{1, \ldots, k_n\} =: T$ denote by $C_{\bar{t}}$ the conjunction $\bigcap_{i=1}^n C_{i\bar{t}(i)}$. Then the least common subsumer $\mathsf{lcs}\{C_{\bar{t}} \mid \bar{t} \in T\}$ is equivalent to the conjunction $\bigcap_{i=1}^n \mathsf{lcs}\{C_{ij} \mid 1 \leq j \leq k_i\}$. The proof is analogous to the one shown above.

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We are now ready to prove that approximating nice concepts, as defined in Definition 5, can be simplified to a conjunction of approximations. For the sake of simplicity we restrict our attention to binary conjunctions. The proof for n-ary conjunctions is analogous.

Theorem 9 Let $C \sqcap D$ be a nice \mathcal{ALC} -concept description. Then $approx_{\mathcal{ALE}}(C \sqcap D) \equiv approx_{\mathcal{ALE}}(C) \sqcap approx_{\mathcal{ALE}}(D)$.

Proof by induction over the sum n of the nesting depths of \sqcap and \sqcup on every role-level in C and D.

• n = 0

Then no conjunction or disjunction occurs at any position in C or D, implying that C and D are nice \mathcal{ALE} -concepts. Hence, $approx_{\mathcal{ALE}}(C) \equiv C$ and $approx_{\mathcal{ALE}}(D) \equiv D$. Since $C \sqcap D$ is also an \mathcal{ALE} -concept, we also know that $approx_{\mathcal{ALE}}(C \sqcap D) \equiv C \sqcap D$. Consequently, $approx_{\mathcal{ALE}}(C \sqcap D) \equiv approx_{\mathcal{ALE}}(C) \sqcap approx_{\mathcal{ALE}}(D)$.

• *n* > 0

Due to the definition of nice, three cases have to be distinguished.

1.
$$C = \prod_{i=1}^{k} C_i$$
 and $D = \prod_{j=1}^{l} D_j$

Then the approximation under consideration is $approx_{\mathcal{AE}}((\bigcap_{i=1}^k C_i) \bigcap (\bigcap_{j=1}^l D_j))$ which can be flattened. The nesting depth of the argument concept thus has decreased by 1 and we still have a nice concept. According to the induction hypothesis, the result is therefore equivalent to

$$\bigcap_{i=1}^{k} approx_{\mathcal{ALE}}(C_i) \cap \bigcap_{i=1}^{l} approx_{\mathcal{ALE}}(D_j)$$

which in turn is equivalent to

$$approx_{\mathcal{ALE}}(\bigcap_{i=1}^k C_i) \cap approx_{\mathcal{ALE}}(\bigcap_{j=1}^l D_j).$$

2. $C = \bigsqcup_{i=1}^k C_i$ and $D = \bigsqcup_{j=1}^l D_j$

To compute $approx_{\mathcal{ALE}}(C \sqcap D)$, the approximation algorithm at first transforms the input concept into \mathcal{ALC} -normal form. The \mathcal{ALC} -normal form of C is of the form $\bigsqcup_{i'=1}^{k'} C'_{i'}$ with no disjunction on the topmost role-level of every $C'_{i'}$. Similarly, the \mathcal{ALC} -normal form of D yields an expression of the form $\bigsqcup_{j'=1}^{l'} D'_{j'}$. Hence, the \mathcal{ALC} -normal form of $C \sqcap D$ is of the form

$$\bigsqcup_{1 < i' < k; \ 1 < j' < l'} (C'_{i'} \sqcap D'_{j'})$$

The approximation $approx_{\mathcal{AC}}(C \sqcap D)$ then by definition equals

$$\operatorname{lcs}\{\operatorname{approx}_{\mathcal{AE}}(C'_{i'} \sqcap D'_{j'}) \mid 1 \leq i' \leq k', 1 \leq j' \leq l'\}.$$

As the maximum nesting depth in all of the occurring approximation expressions has decreased, we may exploit the induction hypothesis, obtaining $lcs\{approx_{\mathcal{ALE}}(C'_{i'}) \cap approx_{\mathcal{ALE}}(D'_{j'})\} \mid 1 \leq i' \leq k', 1 \leq j' \leq l'\}$. According to Lemma 8, the lcs can be split into two lcs-expressions of the form

$$\mathsf{lcs}\{\mathit{approx}_{\mathit{ALE}}(C'_{i'}) \mid 1 \leq i' \leq k'\} \sqcap \mathsf{lcs}\{\mathit{approx}_{\mathit{ALE}}(D'_{j'}) \mid 1 \leq j' \leq l'\}$$

which by definition of the approximation are equivalent to the conjunction of two approximations, namely

$$approx_{\mathcal{ALE}}(\bigsqcup_{i'=1}^{k'}C'_{i'}) \cap approx_{\mathcal{ALE}}(\bigsqcup_{j'=1}^{l'}D'_{j'}),$$

which is equivalent to $approx_{\mathcal{ALE}}(C) \cap approx_{\mathcal{ALE}}(D)$, i.e., the separate approximation of the input concepts.

3.
$$C = \bigsqcup_{i=1}^{k} C_i \text{ and } D = \prod_{i=1}^{l} D_i$$

3. $C = \bigsqcup_{i=1}^k C_i$ and $D = \prod_{j=1}^l D_j$ Similar to the previous case. The \mathcal{ALC} -normal form of the input concept yields an expression of the form $\bigsqcup_{i'=1}^{k'} (C_i \sqcap \bigcap_{j'=1}^{l'} C'_{j'})$. The approximation $approx_{\mathcal{ALE}}(C \sqcap D)$ therefore equals $lcs\{approx_{\mathcal{ALE}}(C'_{i'} \sqcap \bigcap_{i'=1}^{l'} D'_{i'})\}$ $1 \leq i' \leq k'$. According to the induction hypothesis the approximation can be split into $\operatorname{lcs}\{approx_{\mathcal{ACE}}(C'_{i'}) \cap approx_{\mathcal{ACE}}(\bigcap_{i'=1}^{l'} D'_{i'}) \mid 1 \leq i' \leq k'\}.$ Lemma 8 states that this lcs is equivalent to

$$\operatorname{lcs}\{\operatorname{approx}_{\operatorname{ACE}}(C'_{i'}) \mid 1 \leq i' \leq k'\} \cap \operatorname{lcs}\{\operatorname{approx}_{\operatorname{ACE}}(\bigcap_{j'=1}^{l'} D'_{j'})) \mid 1 \leq i' \leq k'\}.$$

The first lcs-expression is equivalent to the approximation of a disjunction while the second one contains k' equal concepts. We thus obtain

$$approx_{\mathcal{A\!L\!E}}(\bigsqcup_{i'=1}^{k'}C'_{i'}) \cap approx_{\mathcal{A\!L\!E}}(\bigcap_{j'=1}^{l'}D'_{j'})).$$

Hence, we end up with the conjunction of the separate approximations $approx_{\mathcal{ALE}}(C) \cap approx_{\mathcal{ALE}}(D).$

Due to Theorem 9 it is now possible to split the computation of approximations into independent parts. Although this does of course not change the complexity class of the approximation algorithm it is still a significant benefit for practical applications. The improved approximation algorithm is displayed in Figure 3. The algorithm requires the unfolded input concept to be in NNF. In the first step the c-approx_{ALE} function checks whether the approximation is trivial. If it is not the next step is to check whether the concept is nice. For nice

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Input: unfolded \mathcal{AU}-concept description C already in NNF
Output: upper ALE-approximation of C
c-approx_{ALE}
      1. If C \equiv \bot, then \operatorname{c-approx}_{\mathcal{ALE}}(C) := \bot;
            if C \equiv \top, then \operatorname{c-approx}_{\mathcal{ALE}}(C) := \top
      2. If nice-concept-p(C) then return c-approx<sub>ALE</sub>(C) := c-nice-approx<sub>ALE</sub>(C)
     3. Otherwise, transform the top-level of C into \mathcal{AU}-normal form C_1 \sqcup \cdots \sqcup C_n
            and return
            c-approx_{\mathcal{ALE}}(C) :=
             \bigcap_{\substack{A \in \bigcap_{i=1}^n \mathsf{prim}(C_i)}} A \quad \sqcap \quad \forall r.\mathsf{lcs}\{\mathsf{c-approx}_{\mathcal{ALE}}(\mathsf{val_r}(C_i)) \mid 1 \leq i \leq n\} \quad \sqcap \\ \bigcap_{\substack{(C_1', \dots, C_n') \in \mathsf{ex_r}(C_1) \times \dots \times \mathsf{ex_r}(C_n)}} \exists r.\mathsf{lcs}\{\mathsf{c-approx}_{\mathcal{ALE}}(C_i' \sqcap \mathsf{val_r}(C_i)) \mid 1 \leq i \leq n\} 
c-nice-approx_{ALE}
      1. If C \equiv \bot, then c-nice-approx<sub>\mathcal{ALE}</sub>(C) := \bot;
            if C \equiv \top, then c-nice-approx<sub>ALE</sub>(C) := \top
      2. If C = C_1 \sqcap \cdots \sqcap C_n, then return
            \mathsf{c-nice-approx}_{\mathcal{ALE}}(C) := \bigcap_{i=1}^n \mathsf{c-nice-approx}_{\mathcal{ALE}}(C_i)
     3. Otherwise, return
            \mathsf{c\text{-}nice\text{-}approx}_{\mathcal{ALE}}(C) :=
         \underset{A \in \bigcap_{i=1}^n \operatorname{prim}(C_i)}{\sqcap} A \qquad \square \qquad \forall r.\mathsf{lcs} \big\{ \operatorname{c-nice-approx}_{\mathcal{ALE}}(\operatorname{val_r}(C_i)) \mid 1 \leq i \leq n \big\} \ \ \square
     \underset{(C_1',\dots,C_n')\in \mathsf{ex_r}(C_1)\times\dots\times\mathsf{ex_r}(C_n)}{\sqcap}\exists r.\mathsf{lcs}\{\mathsf{c-nice-approx}_{\mathcal{ALE}}(C_i'\sqcap\mathsf{val_r}(C_i))\mid 1\leq i\leq n\}
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Figure 3: The improved algorithm c-approx_{ALE} and c-nice-approx_{ALE}.

concepts the $\mathsf{c\text{-}nice\text{-}approx}_{\mathcal{ALE}}$ function is invoked. For all other concepts the \mathcal{ACC} -normal form is computed lazily, i.e., the conjunctions are distributed over the disjunctions only for the current top-level. Then the $\mathsf{c\text{-}approx}_{\mathcal{ALE}}$ algorithm proceeds as before. The $\mathsf{c\text{-}nice\text{-}approx}_{\mathcal{ALE}}$ function for nice concepts works similar. Having treated the trivial cases, the second step is to test if the concept is a conjunction. In that case the approximation is obtained by splitting the concept conjunct-wise and making a recursive call for each conjunct. For all other nice concepts the approximation is computed as in $\mathsf{c\text{-}approx}_{\mathcal{ALE}}$, besides the recursive calls refer to $\mathsf{c\text{-}nice\text{-}approx}_{\mathcal{ALE}}$.

Observe that the test condition for nice concepts can be checked in linear time once the concept description is unfolded and in NNF. Since unfolding and the transformation into NNF are always necessary before applying c-approx_{ALE}, the test for the nice-property adds very little extra costs to the approximation.

Moreover, the conjunct-wise approximation of nice concepts C of the form $(C_1 \sqcup D_1) \sqcap \cdots \sqcap (C_n \sqcup D_n)$ avoids the exponential blow-up caused by the \mathcal{AC} -normalization of C. Hence, this optimization improves the computation times for many concepts which lead the standard algorithm to perform particularly badly.

3.2 Approximating Nice Concepts in TBoxes

If an entire \mathcal{ALC} -TBox is to be translated into an \mathcal{ALE} -TBox, the concept description from the right-hand side of each concept definition has to be replaced by its approximation. For practical applications it is obviously not feasible to perform such a translation in a naive way, simply because of the size of application TBoxes. The idea for optimizing this procedure is to re-use the approximation of a defined concept when approximating concept descriptions that in turn make use of this defined concept. More precisely, if we have already obtained the approximation of C and want for example to compute the approximation of $(D \sqcap \exists r.C)$, we would like to be able to insert the concept description approx(C)directly into the right place in the concept description of $\mathsf{approx}(D \sqcap \exists r.C)$. Unfortunately, this approach does not work for arbitrary All-concept descriptions due to possible interactions between different parts of the concept description. Nice concepts, however, are defined to rule out this kind of interaction. Hence, besides speeding-up the computation of a single approximation, the property of being a nice concept also is a prerequisite for caching and the re-use of already computed approximations. For example, if the defined concepts C_1, C_2, C_3 from the following TBox (with A, B and D being primitive concepts)

$$\mathcal{T} = \{ C_1 = (\exists r. \neg A) \sqcup (\exists r. B),$$

$$C_2 = \exists r. (\forall r. D \sqcup \neg E) \sqcap C_1 \sqcap \neg B,$$

$$C_3 = \neg (\forall r. \exists r. (\neg D \sqcap A) \sqcup \neg C_1 \sqcup \neg C_2) \}$$

are to be approximated and C_1 is approximated first, then this concept description can be re-used in subsequent approximations. If unfolded and transformed into NNF the concepts C_2 and C_3 are nice concepts. So the approximation of C_2 is the conjunction of $\operatorname{approx}(\exists r.(\forall r.D \sqcup \neg E))$ and $\operatorname{approx}(C_1)$ and $\operatorname{approx}(B)$, where the already computed approximation of C_1 can be employed directly. For C_3 we can re-use both approximations of C_1 and C_2 directly and have only to compute the approximation of $\exists r. \forall r.(\neg D \sqcup \neg A)$. Thus, the cost for approximating the entire TBox is reduced heavily.

4 Conclusion and Future Work

In this report we have presented some first steps towards optimizing the computation of approximations. The main idea is to identify concepts that can be decomposed into parts which then can be approximated independently. These so called nice concepts are structured in such a way that the top-level conjuncts cannot interact with one another therefore each conjunct can be approximated separately. Detecting nice concepts and approximating each of their conjuncts independently should be especially powerful in the context of translating entire \mathcal{ALC} -TBoxes into \mathcal{ALE} -TBoxes because it enables the direct re-use of already computed approximations and caching. Unfortunately, the conditions for nice concepts are very strict.

It is an open problem whether the rather strict conditions for nice concepts can be relaxed. To determine if independent approximation of nice concepts is a real benefit for practical applications, requires an implementation of modular approximation. Moreover, it is unknown if nice concepts occur frequently in application TBoxes.

Another open problem is whether the given conditions for nice concepts can be extended to the case where \mathcal{ALCN} -concept descriptions are approximated by \mathcal{ALEN} -concept descriptions.

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