Concept and Role Forgetting in \(\mathcal{ALC}\) Ontologies

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Abstract. Forgetting is an important tool for reducing ontologies by eliminating some concepts and roles while preserving sound and complete reasoning. Attempts have previously been made to address the problem of forgetting in relatively simple description logics (DLs) such as DL-Lite and extended \(\mathcal{EL}\). The ontologies used in these attempts were mostly restricted to TBoxes rather than general knowledge bases (KBs). However, the issue of forgetting for general KBs in more expressive description logics, such as \(\mathcal{ALC}\) and OWL DL, is largely unexplored. In particular, the problem of characterizing and computing forgetting for such logics is still open.

In this paper, we first define semantic forgetting about concepts and roles in \(\mathcal{ALC}\) ontologies and state several important properties of forgetting in this setting. We then define the result of forgetting for concept descriptions in \(\mathcal{ALC}\), state the properties of forgetting for concept descriptions, and present algorithms for computing the result of forgetting for concept descriptions. Unlike the case of DL-Lite, the result of forgetting for an \(\mathcal{ALC}\) ontology does not exist in general, even for the special case of concept forgetting. This makes the problem of how to compute forgetting in \(\mathcal{ALC}\) more challenging. We address this problem by defining a series of approximations to the result of forgetting for \(\mathcal{ALC}\) ontologies and studying their properties and their application to reasoning tasks. We use the algorithms for computing forgetting for concept descriptions to compute these approximations. Our algorithms for computing approximations can be embedded into an ontology editor to enhance its ability to manage and reason in (large) ontologies.

1 Introduction

The amount of semantically annotated data available on the Web is growing rapidly. Often, the formal model used for representing such information is an ontology in some description logic. As more ontologies are used for annotating data on the Web, and as the populated ontologies become larger and more comprehensive, it becomes increasingly important for the Semantic Web [4] to be able to construct and manage such ontologies. Examples of large ontologies currently in use include the Systematised Nomenclature

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of Medicine Clinical Terms (SNOMED CT) containing 380K concepts, GALEN, the Foundational Model of Anatomy (FMA), the National Cancer Institute (NCI) Thesaurus containing over 60K axioms, and the OBO Foundry containing about 80 biomedical ontologies.

While it is expensive to construct large ontologies, it is even more expensive to host, manage and use a large, comprehensive ontology when a smaller ontology would suffice. Therefore, tools to reduce large ontologies to smaller ontologies that meet the needs of specific applications aid and encourage the use of existing ontologies. However, as the tool evaluation study in [5] shows, existing tools, such as Protégé [28], NeOn [29] and TopBraid [30], are far from satisfactory for this purpose.

Ontology engineers thus face the task of reducing existing, large ontologies to smaller, better focussed ontologies by hiding or forgetting irrelevant concepts and roles while preserving required reasoning capabilities. Such ontology reductions can be applied to ontology extraction, ontology summary, ontology integration, and ontology evolution.

We consider two typical scenarios, in ontology extraction and ontology summary, respectively.

**Ontology extraction**: To avoid building new ontologies from scratch, it is preferable to reuse existing ontologies whenever possible. But often, only part of an existing ontology is required. For example, if we had an ontology of medical terms, such as SNOMED, but were only interested in infectious diseases, it would be desirable to forget all other terms in the ontology before starting to reason about infectious diseases.

**Ontology summary**: As argued in [1, 18], a key problem faced by ontology engineers is the task of constructing a summary of an existing ontology so that other users can decide whether or not to use it. This task involves first identifying the key concepts (and roles) in the ontology and then hiding or forgetting all other concepts (and roles). For example, an astronomical ontology of the solar system might have planets as key concepts and asteroids and comets as less important concepts.

However, an ontology is often represented as a logical theory, and the removal of one term may influence other terms in the ontology. Thus, more advanced methods for dealing with large ontologies and reusing existing ontologies are desired.

Forgetting (or uniform interpolation) has previously been studied for propositional and first-order logic and logic programming[15, 16, 6], where it has proved a useful technique for reducing the size of a logical theory while preserving sound and complete reasoning in the resulting smaller theory.

However, description logic (DL) [3] is a different, important, knowledge representation framework, which is the basis for ontology languages such as OWL, that are widely used in the Semantic Web. In this context, an ontology is a knowledge base (KB) in a particular description logic, where a knowledge base consists of a terminology box (TBox) and an assertion box (ABox).

Although most description logics are equivalent to fragments of first-order logic (FOL), the forgetting for first-order logic introduced in [16] is not directly applicable to description logics for at least two reasons. First, the correspondence between DLs and FOL is useless in investigating forgetting for DLs because the result of forgetting in a theory of the first-order logic (FOL) may only be expressible in second-order logic.
Second, it is preferable to perform forgetting in description logics directly rather than transforming an ontology into a first-order theory, forgetting and then transforming back to an ontology.

Attempts have previously been made to address the problem of forgetting in relatively simple description logics (DLs) such as DL-Lite [21, 14] and extended $\mathcal{E}L$ [12]. The ontologies used in these attempts were mostly restricted to KBs with empty ABoxes. Forgetting also generalizes previous work on conservative extensions [9, 7, 17] and the modularity defined in [8, 10, 13]. A definition of forgetting for TBoxes in the more expressive DL $\mathcal{ALC}$ was given in [7].

However, the issue of forgetting for general KBs, with nonempty ABoxes, in more expressive DLs, such as $\mathcal{ALC}$ and OWL DL, is largely unexplored. In particular, the problem of characterizing and computing forgetting for such logics is still open.

In this paper we first give a semantic definition of forgetting for ontologies in the description logic $\mathcal{ALC}$ and state several important properties of forgetting. We choose $\mathcal{ALC}$ to study in this paper because it allows all boolean operations and most expressive DLs are based on it. Others have argued [19] that practical ontologies such as SNOMED would benefit from a more expressive DL. We then define the result of forgetting for concept descriptions in $\mathcal{ALC}$, state the properties of forgetting for concept descriptions, and present algorithms for computing the result of forgetting for concept descriptions. (Forgetting for concept descriptions in $\mathcal{ALC}$ has previously been investigated under the name of uniform interpolation in [20].) Unlike the case of DL-Lite, the result of forgetting for an $\mathcal{ALC}$ ontology does not exist in general, even for the special case of concept forgetting. This makes the problem of how to compute forgetting in $\mathcal{ALC}$ more challenging. We address this problem by defining a series of approximations to the result of forgetting for $\mathcal{ALC}$ ontologies and studying their properties and their application to reasoning tasks. We use the algorithms for computing forgetting for concept descriptions to compute these approximations. Our algorithms for computing approximations can be embedded into an ontology editor to enhance its ability to manage and reason in (large) ontologies.

Our work significantly extends previous work in at least two ways: (1) We make the first attempt to study forgetting for an expressive description logic, instead of DL-Lite and variants of $\mathcal{EL}$. (2) Ontologies in this paper are KBs with nonempty ABoxes, rather than TBoxes only in previous work. In addition, our definitions and results hold for forgetting about both concepts and roles.

Due to space limitation, proofs are omitted in this paper but can be found at http://www.cit.gu.edu.au/~kewen/Papers/alc_forget_long.pdf.

2 Description Logic $\mathcal{ALC}$

In this section, we briefly recall some preliminaries of $\mathcal{ALC}$, the basic description logic which contains all boolean operators. Further details of $\mathcal{ALC}$ and other DLs can be found in [3].

First, we introduce the syntax of concept descriptions for $\mathcal{ALC}$. To this end, we assume that $N_C$ is a set of concept names (or concept), $N_R$ is a set of role names (or roles) and $N_I$ is a set of individuals.
Elementary concept descriptions consist of both concept names and role names. So a concept name is also called atomic concept while a role name is also called atomic role. Complex concept descriptions are built inductively as follows: $A$ (atomic concept); $\top$ (universal concept); $\bot$ (empty concept); $\neg C$ (negation); $C \cap D$ (conjunction); $C \cup D$ (disjunction); $\forall R.C$ (universal quantification) and $\exists R.C$ (existential quantification).

Here, $A$ is an (atomic) concept, $C$ and $D$ are concept descriptions, and $R$ is a role.

An interpretation $I$ of $\mathcal{ALC}$ is a pair $(\Delta^I, \cdot^I)$ where $\Delta^I$ is a non-empty set called the domain and $\cdot^I$ is an interpretation function which associates each (atomic) concept $A$ with a subset $A^I$ of $\Delta^I$ and each atomic role $R$ with a binary relation $R^I \subseteq \Delta^I \times \Delta^I$. The function $\cdot^I$ can be naturally extended to complex descriptions:

\[
\begin{align*}
\top^I &= \Delta^I \\
(\neg C)^I &= \Delta^I - C^I \\
(C \cup D)^I &= C^I \cup D^I \\
(\forall R.C)^I &= \{ a \in \Delta^I : \forall b, (a, b) \in R^I \text{ implies } b \in C^I \} \\
(\exists R.C)^I &= \{ a \in \Delta^I : \exists b, (a, b) \in R^I \text{ and } b \in C^I \}
\end{align*}
\]

An $\mathcal{ALC}$ assessorial box (or ABox) is a finite set of assertions. An assertion is a concept assertion of the form $C(a)$ or a role assertion of the form $R((a, b))$, where $a$ and $b$ are individuals, $C$ is a concept name, $R$ is a role name.

An interpretation $I$ satisfies a concept assertion $C(a)$ if $a^I \in C^I$, a role assertion $R((a, b))$ if $(a^I, b^I) \in R^I$. If an assertion $\alpha$ is satisfied by $I$, it is denoted $I \models \alpha$. An interpretation $I$ is a model of an ABox $\mathcal{A}$, written $I \models \mathcal{A}$, if it satisfies all assertions in $\mathcal{A}$.

A inclusion axiom (simply inclusion, or axiom) is of the form $C \sqsubseteq D$ ($C$ is subsumed by $D$), where $C$ and $D$ are concept descriptions. The inclusion $C \equiv D$ ($C$ is equivalent to $D$) is an abbreviation of two inclusions $C \sqsubseteq D$ and $D \sqsubseteq C$. A terminology box, or TBox, is a finite set of inclusions. An interpretation $I$ satisfies an inclusion $C \sqsubseteq D$ if $C^I \subseteq D^I$. $I$ is a model of a TBox $T$, denoted $I \models T$, if $I$ satisfies every inclusion of $T$. $T \models C \sqsubseteq D$ if for any $I$, $I \models T$ implies $I \models C \sqsubseteq D$.

Formally, a knowledge base (KB) is a pair $(T, \mathcal{A})$ of a TBox $T$ and an ABox $\mathcal{A}$. An interpretation $I$ is a model of $\mathcal{K}$ if $I$ is a model of both $T$ and $\mathcal{A}$, denoted $I \models \mathcal{K}$. If $\alpha$ is an axiom or an assertion, $\mathcal{K} \models \alpha$ if every model of $\mathcal{K}$ is also a model of $\alpha$. Two KBs $\mathcal{K}$ and $\mathcal{K}'$ are equivalent, written $\mathcal{K} \equiv \mathcal{K}'$, if they have the same models. “$\equiv$” can be similarly defined for ABoxes and TBoxes.

The signature of a concept description $C$, written $\text{sig}(C)$, is the set of all concept and role names in $C$. Similarly, we can define $\text{sig}(\mathcal{A})$ for an ABox $\mathcal{A}$, $\text{sig}(T)$ for a TBox $T$, and $\text{sig}(\mathcal{K})$ for a KB $\mathcal{K}$.

### 3 Forgetting in $\mathcal{ALC}$ Ontologies

In this section, we will first give a semantic definition of what it means to forget about a set of variables in an $\mathcal{ALC}$ KB, and then state and discuss several important properties of forgetting that justify the definition chosen. This is the first study of forgetting both concepts and roles for arbitrary knowledge bases in $\mathcal{ALC}$.
As explained earlier, given an ontology $K$ on signature $S$ and $V \subseteq S$, in ontology engineering it is often desirable to obtain a new ontology $K'$ on $S - V$ such that reasoning tasks on $S - V$ are still preserved in $K'$. As a result, $K'$ is weaker than $K$ in general. This intuition is formalized in the following definition.

Definition 3.1 (KB-forgetting). Let $K$ be a KB in $\mathcal{ALC}$ and $V$ be a set of variables. A KB $K'$ over the signature $\text{sig}(K) - V$ is a result of forgetting about $V$ in $K$ if

(KF1) $K \models K'$;
(KF2) for each concept inclusion $C \subseteq D$ in $\mathcal{ALC}$ not containing any variables in $V$, $K \models C \subseteq D$ implies $K' \models C \subseteq D$;
(KF3) for each membership assertion $C(a)$ or $R(a,b)$ in $\mathcal{ALC}$ not containing any variables in $V$, $K \models C(a)$ (resp., $K \models R(a,b)$) implies $K' \models C(a)$ (resp., $K' \models R(a,b)$);

Condition (KF3) extends previous definitions [7] to allow for nonempty ABoxes in KBs.

To illustrate the above definition of semantic forgetting and how forgetting can be used in ontology extraction, consider the following example of designing an $\mathcal{ALC}$ ontology about flu.

Example 3.1. Suppose we have searched the Web and found an ontology about human diseases (such a practical ontology could be very large):

\begin{align*}
\text{Disease} & \sqsubseteq \forall \text{attacks.Human}, \\
\text{Human} & \equiv \text{Male} \sqcup \text{Female}, \\
\text{Human} \sqcap \text{Infected} & \sqsubseteq \exists \text{shows.Symptom}, \\
\text{Disease} & \equiv \text{Infectious} \sqcup \text{Noninfectious}, \\
\text{Influenza} \sqcup \text{HIV} \sqcup \text{TB} & \sqsubseteq \text{Infectious}.
\end{align*}

We want to construct a (smaller) ontology only about flu by reusing the above ontology. This is done by forgetting about the undesired concepts \{\text{Disease}, \text{Noninfectious}, \text{HIV}, \text{TB}\}. As a result, the following ontology is obtained:

\begin{align*}
\text{Influenza} & \sqsubseteq \text{Infectious}, \\
\text{Infectious} & \sqsubseteq \forall \text{attacks.Human}, \\
\text{Human} & \equiv \text{Male} \sqcup \text{Female}, \\
\text{Human} \sqcap \text{Infected} & \sqsubseteq \exists \text{shows.Symptom},
\end{align*}

The next example shows that the result of forgetting in an $\mathcal{ALC}$ ontology may not exist in some cases.

Example 3.2. Let $K = (\mathcal{T}, \mathcal{A})$ be an $\mathcal{ALC}$ KB where $\mathcal{T} = \{ A \sqsubseteq B, B \sqsubseteq C, C \sqsubseteq \forall R.C, C \sqsubseteq D \}$, and $\mathcal{A} = \{ B(a), R(a,b) \}$.

Take $K_1 = (\mathcal{T}_1, \mathcal{A}_1)$ where $\mathcal{T}_1 = \{ A \sqsubseteq C, C \sqsubseteq \forall R.C, C \sqsubseteq D \}$ and $\mathcal{A}_1 = \{ C(a), R(a,b) \}$. Then $K_1$ is a result of forgetting about concept $B$ in $K$.

However, there does not exist a result of forgetting about $\{B, C\}$ in $K$. To understand this, we note that the result of forgetting about $\{B, C\}$ in $K$ should include the following inclusions:

\begin{align*}
A & \subseteq D, A \subseteq \forall R.D, A \subseteq \forall R.\forall R.D, \ldots,
\end{align*}
\[ D(a), (\forall R.D)(a), (\forall R.\forall R.D)(a), \ldots, \text{ and} \]
\[ B(a, b), D(b), (\forall R.D)(b), (\forall R.\forall R.D)(b), \ldots \]

In fact, there is no finite ALC KB which is equivalent to the above infinite set of inclusions.

If the result of forgetting about \( V \) in \( K \) is expressible as an ALC KB, we say \( V \) is forgettable from \( K \).

In the rest of this section, we state and discuss some important consequences of this definition of forgetting for KBs in ALC. These properties provide evidence that the definition is appropriate.

**Proposition 3.1.** Let \( K \) be a KB in ALC and \( V \) a set of variables. If both \( K' \) and \( K'' \) in ALC are results of forgetting about \( V \) in \( K \), then \( K' \equiv K'' \).

This proposition says that the result of forgetting in ALC is unique up to KB equivalence. Given this result, we write forget\((K, V)\) to denote any result of forgetting about \( V \) in \( K \) in ALC. In particular, forget\((K, V) = K'\) means that \( K' \) is a result of forgetting about \( V \) in \( K \).

In fact, forgetting in TBoxes is independent of ABoxes as the next result shows.

**Proposition 3.2.** Let \( T \) be an ALC TBox and \( V \) a set of variables. Then, for any ALC ABox \( A \), \( T' \) is the TBox of forget\(((T, A), V)\) iff \( T' \) is the TBox of forget\(((T, \emptyset), V)\).

For simplicity, we write forget\((T, V)\) for forget\(((T, \emptyset), V)\) and call it the result of TBox-forgetting about \( V \) in \( T \).

The following result, which generalizes Proposition 3.1, shows that forgetting preserves implication and equivalence relations between KBs.

**Proposition 3.3.** Let \( K_1, K_2 \) be two KBs in ALC and \( V \) a set of variables. Then

- \( K_1 \models K_2 \) implies forget\((K_1, V) \models \) forget\((K_2, V)\);
- \( K_1 \equiv K_2 \) implies forget\((K_1, V) \equiv \) forget\((K_2, V)\).

However, the converse of Proposition 3.3 is not true in general. Consider \( K \) and \( K_1\) in Example 3.2, it is obvious that forget\((K, \{B\}) \equiv \) forget\((K_1, \{B\})\). However, \( K \) and \( K_1 \) are not equivalent.

Consistency and query answering are two major reasoning tasks in description logics. It is a key requirement for a reasonable definition of forgetting to preserve these two reasoning forms.

**Proposition 3.4.** Let \( K \) be a KB in ALC and \( V \) a set of variables. Then

1. \( K \) is consistent iff forget\((K, V) \) is consistent;
2. for any inclusion or assertion \( \alpha \) not containing variables in \( V \), \( K \models \alpha \) iff forget\((K, V) \models \alpha \).

The next result shows that the forgetting operation can be divided into steps, with a part of the signature forgotten in each step.
Proposition 3.5. Let $K$ be a KB in ALC and $V_1, V_2$ two sets of variables. Then we have
\[
\text{forget}(K, V_1 \cup V_2) \equiv \text{forget}(\text{forget}(K, V_1), V_2).
\]

To compute the result of forgetting about $V$ in $K$, it is equivalent to forget the variables in $V$ one by one, i.e., forgetting can be computed incrementally.

4 Forgetting in ALC Concept Descriptions

Forgetting in a concept description has been investigated under the name of uniform interpolation in [20]. In this section, we reformulate the definition of the forgetting about concept and role names in ALC concept descriptions (briefly, c-forgetting) and introduce some results that will be used in the next section. From the viewpoint of ontology management, the issue of forgetting in concept descriptions is less important than that for KBs and TBoxes. However, we will show later that c-forgetting can be used to provide an approximation algorithm for KB-forgetting in ALC, as well as its theoretical importance.

Intuitively, the result $C'$ of forgetting about a set of variables from a concept description $C$ should be weaker than $C$ but as close to $C$ as possible. For example, after the concept Male is forgotten from a concept description for “Male Australian student” Australians $\cap$ Students $\cap$ Male, then we should obtain a concept description Australians $\cap$ Students for “Australian student”. More specifically, $C'$ should be a concept description that defines a minimal concept description among all concept descriptions that subsume $C$ and are syntactically irrelevant to $V$ (i.e., variables in $V$ do not appear in the concept description).

Definition 4.1 (c-forgetting). Let $C$ be a concept description in ALC and $V$ a set of variables. A concept description $C'$ on the signature $\text{sig}(C) - V$ is a result of c-forgetting about $V$ in $C$ if the following conditions are satisfied:

(CF1) $\models C \subseteq C'$.
(CF2) For every ALC concept description $C''$ with $\text{sig}(C'') \cap V = \emptyset$, $\models C \subseteq C''$ implies $\models C' \subseteq C''$.

The above (CF1) and (CF2) correspond to the conditions (2) and (3) of Theorem 8 in [20]. A fundamental property of c-forgetting in ALC concept descriptions is that the result of c-forgetting is unique under concept description equivalence.

Proposition 4.1. Let $C$ be a concept description in ALC and $V$ a set of variables. If two concept descriptions $C'$ and $C''$ in ALC are results of c-forgetting about $V$ in $C$, then $\models C' \equiv C''$.

As all results of c-forgetting are equivalent, we write $\text{forget}(C, V)$ to denote an arbitrary result of c-forgetting about $V$ in $C$.

Example 4.1. Suppose the concept “Research Student” is defined by $C = \text{Student} \cap (\text{Master} \sqcup \text{PhD}) \sqcap \exists \text{supervised}. \text{Professor}$ where “Master”, “PhD” and “Professor” are
all concepts; “supervised” is a role and $\text{supervised}(x, y)$ means that $x$ is supervised by $y$. If the concept description $C$ is used only for students, we may wish to forget about Student: $\text{forget}(C, \text{Student}) = (\text{Master} \sqcup \text{PhD}) \sqcap \exists \text{supervised. Professor}$. If we do not require that a supervisor for a research student must be a professor, then the filter “Professor” can be forgotten: $\text{forget}(C, \text{Professor}) = \text{Student} \sqcap (\text{Master} \sqcup \text{PhD}) \sqcap \exists \text{supervised.}$

A concept description $C$ is **satisfiable** if $C^I \neq \emptyset$ for some interpretation $I$ on $\text{sig}(C)$. $C$ is unsatisfiable if $\models C \equiv \bot$. By Definition 4.1, c-forgetting also preserves satisfiability of concept descriptions.

**Proposition 4.2.** Let $C$ be a concept description in $\text{ALC}$, and $V$ be a set of variables. Then $C$ is satisfiable iff $\models \text{forget}(C, V)$ is satisfiable.

Similar to forgetting in KB, the c-forgetting operation can be divided into steps.

**Proposition 4.3.** Let $C$ be a concept description in $\text{ALC}$ and $V_1, V_2$ two sets of variables. Then we have

$$\models \text{forget}(\text{forget}(C, V_1), V_2) \equiv \text{forget}(\text{forget}(C, V_1), V_2).$$

Given the above result, when we want to forget about a set of variables, they can be forgotten one by one. Also, the ordering of c-forgetting operation is irrelevant to the result.

**Corollary 4.1.** Let $C$ be a concept description in $\text{ALC}$ and let $V = \{V_1, \ldots, V_n\}$ be a set of variables. Then, for any permutation $(i_1, i_2, \ldots, i_n)$ of $\{1, 2, \ldots, n\}$,

$$\models \text{forget}(\text{forget}(C, V_{i_1}), V_{i_2}), \ldots, V_{i_n}) \equiv \text{forget}(\text{forget}(C, V_{i_1}), V_{i_2}), \ldots, V_{i_n}).$$

The following result, which is not obvious, shows that c-forgetting distributes over union $\sqcup$.

**Proposition 4.4.** Let $C_1, \ldots, C_n$ be concept descriptions in $\text{ALC}$. For any set $V$ of variables, we have

$$\models \text{forget}(C_1 \sqcup \ldots \sqcup C_n, V) \equiv \text{forget}(C_1, V) \sqcup \ldots \sqcup \text{forget}(C_n, V).$$

However, c-forgetting for $\text{ALC}$ does not distribute over intersection $\sqcap$. For example, if the concept description $C = A \sqcap \neg A$, then $\text{forget}(C, A) = \bot$, since $\models C \equiv \bot$. But $\text{forget}(A, A) \sqcap \text{forget}(\neg A, A) \equiv \top$.

An important reason for this is that c-forgetting does not distribute over negation. Actually, we have

$$\models \neg \text{forget}(C, V) \sqsubseteq \neg C \sqsubseteq \text{forget}(\neg C, V).$$

These subsumptions may be strict, e.g., if $C$ is $A \sqcap B$ and $V$ is $\{A\}$, then $\neg \text{forget}(C, V)$ is $\neg B$, but $\text{forget}(\neg C, V)$ is $\top$.

The next result shows that c-forgetting does distribute over quantifiers. Since c-forgetting does not distribute over negation, the two statements in the following proposition do not necessarily imply each other. The proof uses tableau reasoning for $\text{ALC}$ and is surprisingly complex.
Proposition 4.5. Let $C$ be a concept description in $\mathcal{ALC}$, $R$ be a role name and $V$ be a set of variables. Then

- $\text{forget}(\forall R.C, V) = \top$ for $R \in V$, and $\text{forget}(\forall R.C, V) = \forall R.\text{forget}(C, V)$ for $R \notin V$;
- $\text{forget}(\exists R.C, V) = \top$ for $R \in V$, and $\text{forget}(\exists R.C, V) = \exists R.\text{forget}(C, V)$ for $R \notin V$.

These results suggest a way of computing c-forgetting about set $V$ of variables in a complex $\mathcal{ALC}$ concept description $C$. That is, to forget about each variable $V$ in $V$ one after another, and to distribute the c-forgetting computation to subconcepts of $C$.

In what follows, we introduce an algorithm for computing the result of c-forgetting through rewriting of concept descriptions (syntactic concept transformations) [20]. This algorithm consists of two stages: (1) $C$ is first transformed into an equivalent disjunctive normal form (DNF), which is a disjunction of conjunctions of simple concept descriptions; (2) the result of c-forgetting about $V$ in each such simple concept description is obtained by removing some parts of the conjunct.

Before we introduce disjunctive normal form (DNF), some notation and definitions are in order. We call an (atomic) concept $A$ or its negation $\neg A$ a literal concept or simply a literal. A pseudo-literal with role $R$ is a concept description of the form $\exists R.F$ or $\forall R.F$, where $R$ is a role name and $F$ is an arbitrary concept. A generalized literal is either a literal or a pseudo-literal.

Definition 4.2. A concept description $D$ is in disjunctive normal form (DNF) if $D = \bot$ or $D = \top$ or $D$ is a disjunction of conjunctions of generalized literals $D = D_1 \sqcup \cdots \sqcup D_n$, where each $D_i \neq \bot$ ($1 \leq i \leq n$) is a conjunction $\bigwedge L$ of literals, or of the form

$$\forall R.\bigwedge \left[ \forall R.U_R \cap \bigwedge_k \exists R.(E_R^{(k)} \cap U_R) \right]$$

where $R$ is the set of role names that occur in $D_i$, and each $U_R$ and each $E_R^{(k)} \cap U_R$ is a concept description in DNF.

We note that, to guarantee the correctness of the algorithm, the above DNF for $\mathcal{ALC}$ is more complex than we have in classical logic and DL-Lite. See Example 4.2 for an example of a concept description in DNF.

Each concept description in $\mathcal{ALC}$ can be transformed into an equivalent one in DNF by the following two steps: (1) first transform the given concept description into a disjunction of conjunctions of pseudo-literals using De Morgan’s laws, distributive laws and necessary simplifications, and then (2) for each conjunction in the resulting concept description, perform the following three transformations in order:

$$C \leadsto \forall R.\top \cap C$$
$$\forall R.C_1 \cap \exists R.C_2 \leadsto \forall R.C_1 \cap \exists R.(C_1 \cap C_2)$$
$$\forall R.C_1 \cap \cdots \cap \forall R.C_n \leadsto \forall R.(C_1 \cap \cdots \cap C_n).$$

The first transformation above is to transform a concept description containing only existential quantifier into the normal form. For example, if $C$ is concept name, $\exists R.C,$
Algorithm 1 (Compute c-forgetting)

Input: An \( \mathcal{ALC} \) concept description \( C \) and a set \( V \) of variables in \( C \).

Output: \( \text{forget}(C, V) \).

Method:

\begin{enumerate}
  \item Transform \( C \) into its DNF \( D \). If \( D \) is \( \top \) or \( \bot \), return \( D \); otherwise, let \( D = D_1 \sqcup \ldots \sqcup D_n \) as in Definition 4.2.
  \item For each conjunct \( E \) in each \( D_i \), perform the following transformations:
    \begin{itemize}
      \item if \( E \) is a literal of the form \( A \) or \( \neg A \) with \( A \in V \), replace \( E \) with \( \top \);
      \item if \( E \) is a pseudo-literal in the form of \( \forall R.F \) or \( \exists R.F \) with \( R \in V \), replace \( E \) with \( \top \);
      \item if \( E \) is a pseudo-literal in the form of \( \forall R.F \) or \( \exists R.F \) where \( R \notin V \), replace \( F \) with \( \text{forget}(F, V) \), and replace each resulting \( \forall R.(\top \sqcup F) \) with \( \top \).
    \end{itemize}
  \item Return the resulting concept description as \( \text{forget}(C, V) \).
\end{enumerate}

which is not in normal form, can be transformed into the normal form \( \forall R. \top \sqcap \exists R.C \).

While the second is to assemble several quantifications with the same role name into a single one, the third is crucial for guaranteeing the correctness of our algorithm.

Once an \( \mathcal{ALC} \) concept description \( D \) is in the normal form, the result of c-forgetting about a set \( V \) of variables in \( D \) can be obtained from \( D \) by simple symbolic manipulations (ref. Algorithm 1).

According to Algorithm 1, an input concept description must first be transformed into the normal form before the steps for forgetting are applied. For instance, if we want to forget \( A \) in the concept description \( D = A \sqcap \neg A \sqcap B \), \( D \) is transformed into the normal form, which is \( \bot \), and then obtain \( \text{forget}(D, A) = \bot \). We note that \( B \) is not a result of forgetting about \( A \) in \( D \).

Example 4.2. Given a concept \( D = (A \sqcup \exists R. \neg B) \sqcap \forall R.(B \sqcup C) \), we want to forget about concept name \( B \) in \( D \). In Step 1 of Algorithm 1, \( D \) is firstly transformed into its DNF \( D' = [A \sqcap \forall R.(B \sqcup C)] \sqcup [\forall R.(B \sqcup C) \sqcap \exists R.(\neg B \sqcap C)] \). Note that \( \exists R.(\neg B \sqcap C) \) is transformed from \( \exists R.\neg [B \sqcap (B \sqcup C)] \). Then in Step 2, each occurrence of \( B \) in \( D' \) is replaced by \( \top \), and \( \forall R.(\top \sqcup F) \) is replaced with \( \top \). We obtain \( \text{forget}(D, \{B\}) = A \sqcup \exists R.C \). To forget about role \( R \) in \( D \), Algorithm 1 replaces each pseudo-literals in \( D' \) of the form \( \forall R.F \) or \( \exists R.F \) with \( \top \), and returns \( \text{forget}(D, \{R\}) = \top \).

Obviously, the major cost of Algorithm 1 is from transforming the given concept description into its DNF. For this reason, the algorithm is exponential time in the worst case. However, if the concept description \( C \) is in DNF, Algorithm 1 takes only linear time (w.r.t. the size of \( C \)) to compute the result of c-forgetting about \( V \) in \( C \). And the result of c-forgetting is always in DNF.

Theorem 4.1. Let \( V \) be a set of concept and role names and \( C \) a concept description in \( \mathcal{ALC} \). Then Algorithm 1 always returns \( \text{forget}(C, V) \).

The proof of this theorem uses the tableau for \( \mathcal{ALC} \).
Approximate Forgetting in $\mathcal{ALC}$ Ontologies

As we showed in Section 3, the result of forgetting for an $\mathcal{ALC}$ KB might not exist. However, if our goal is to determine whether a given set of axioms $\Sigma$ are logical consequences of the result of forgetting for a KB $K$, we show in this section how to determine this without computing the result of forgetting at all. Instead, given an upper bound on the size of the axioms in $\Sigma$, we show how to compute a finite KB $K'$ such that $K' \models \Sigma$ if and only if $K \models \Sigma$. Here, we assume $\Sigma$ does not contain any of the variables that are being forgotten. The finite KB $K'$ is called an approximation to the result of forgetting for $K$. In fact, we show how to compute a sequence of approximations that can determine whether axioms of increasingly large size are logical consequences of the result of forgetting. These approximations are computed using Algorithm 1 for c-forgetting. We can thus obtain the benefits of forgetting, by computing a smaller KB, and performing reasoning, without having to actually compute the (non-existent) result of forgetting.

As a special case, we first introduce an approximation to TBox-forgetting. Example 3.2 shows that, for some TBox $T$, $\text{forget}(T, V)$ may not be expressible as a finite $\mathcal{ALC}$ TBox. Thus, it is natural to consider a sequence of (finite) TBoxes that approximate the result of forgetting in $T$ in the sense that the sequence is non-decreasing in terms of logical implication and the limit of the sequence is the result of forgetting. Such a sequence is constructed by using results developed for c-forgetting in Section 4.

We note that, for an inclusion $C \subseteq D$ in $T$, $\text{forget}(C, V) \subseteq \text{forget}(D, V)$ may not be a logical consequence of $T$ and thus may not be in $\text{forget}(T, V)$. However, if we transform $T$ into an equivalent singleton TBox $\{T \subseteq C_T\}$, where $C_T = \bigcap_{C \subseteq D \in T}(\neg C \cup D)$, then inclusion $\alpha_0$ of the form $T \subseteq \text{forget}(C_T, V)$ is a logical consequence of $T$. In general, the singleton TBox $\{\alpha_0\}$ is not necessarily equivalent to forget($T$, $V$). However, it can be a starting point of a sequence whose limit is forget($T$, $V$). Note that $T$ is also equivalent to $\{T \subseteq C_T \cap \forall R.C_T\}$ for an arbitrary role name $R$ in $T$. Hence, inclusion $\alpha_1$ of the form $T \subseteq \text{forget}(C_T \cap \forall R.C_T, V)$ is a logical consequence of $T$, and it can be shown that TBox $\{\alpha_1\}$ is logically stronger then $\{\alpha_0\}$. That is, $\text{forget}(T, V) \models \{\alpha_1\} \models \{\alpha_0\}$. Let $\alpha_2$ be $T \subseteq \text{forget}(C_T \cap \forall R.C_T \cap \forall R \forall R.C_T, V)$, then we have $\text{forget}(T, V) \models \{\alpha_2\} \models \{\alpha_1\} \models \{\alpha_0\}$. In this way, we can construct a sequence of TBoxes with increasing logical strength, whose limit is forget($T$, $V$).

For $n \geq 0$, define

$$C_T^{(n)} = \bigcap_{k=0}^{n} \bigcap_{R_1, \ldots, R_k \in R} \forall R_1 \cdots \forall R_k.C_T$$

where $C_T = \bigcap_{C \subseteq D \in T}(\neg C \cup D)$ and $R$ is the set of role names in $K$.

We now define a sequence of TBoxes, which essentially provides an approximation to the result of TBox-forgetting.

Definition 5.1. Let $T$ be an $\mathcal{ALC}$ TBox and $V$ be a set of variables. For each $n \geq 0$, the TBox

$$\text{forget}^n(T, V) = \{T \subseteq \text{forget}(C_T^{(n)}, V)\}$$

is called the $n$-forgetting about $V$ in $T$. 

Note that the above n-forgetting for TBoxes is defined in terms of forgetting in concept descriptions (c-forgetting).

**Example 5.1.** Consider the TBox $T$ in Example 3.2, we have $C_T = (\neg A \sqcup B) \cap (\neg B \sqcup C) \sqcap (\neg C \sqcup \forall R.C) \sqcap (\neg C \sqcup D)$, and $C_T^{(0)} = C_T, C_T^{(1)} = C_T \sqcap \forall R.C_T, \ldots, C_T^{(n)} = C_T \sqcap \forall R.C_T^{(n-1)} (n \geq 2)$.

Let $V = \{B, C\}$. For each $n \geq 0$, the forget$^n(T, V)$ can be computed as follows.

\[
\begin{align*}
&\text{forget}^0(T, V) = \{ \top \sqsubseteq \neg A \sqcup D \}, \text{ which is equivalent to } \{ A \sqsubseteq D \}. \\
&\text{forget}^1(T, V) = \{ \top \sqsubseteq \neg A \sqcup (D \sqcap \forall R.D) \}, \text{ which is } \{ A \sqsubseteq D, A \sqsubseteq \forall R.D \}. \\
&\ldots. \\
&\text{forget}^n(T, V) = \{ A \sqsubseteq D, A \sqsubseteq \forall R.D, \ldots, A \sqsubseteq \forall R. \Bigwedge_{i=1}^{n} R(D) \}.
\end{align*}
\]

We call $\left( \bigwedge_{i=1}^{n} C_i \right)(a)$ the conjunction of assertions $C_1(a), \ldots, C_n(a)$ while $\left( \bigvee_{i=1}^{n} C_i \right)(a)$ is called the disjunction of these assertions.

Before we can perform forgetting on a given ABox, we need to preprocess it and thus transform it into a normal form. To this end, we give the following definition.

**Definition 5.2.** An ABox $A$ in $\mathcal{ALC}$ is complete if for any individual name $a$ in $A$ and assertion $C(a)$ with $A \models C(a)$, we have $\models C' \sqsubseteq C$, where $C'$ is the conjunction of all the concept assertions about $a$ in $A$.

For example, ABox $A = \{ \forall R.A(a), R(a, b) \}$ is incomplete, because $A \models A(b)$ whereas no assertion $C(b)$ exists in $A$ such that $\models C \sqsubseteq A$. After adding assertions $A(b), \exists R.A(a)$ and $\top(a)$ into $A$, the resulting ABox is complete.

In complete ABoxes, concept entailment can be reduced to concept subsumption, and is independent of role assertions.

However, there exist incomplete ABoxes that are not equivalent any (finite) complete ABox. For example, the ABox $\{ R(a, b), R'(b, a) \}$ has infinitely many logical consequences of the form $\exists R. \exists R'. C \sqcup \neg C(a)$ where $C$ is an arbitrary concept description. This kind of situations are caused by certain cycles in ABoxes. We say an ABox is acyclic if there exists no cycle of the form $R_1(a, a_1), R_2(a_1, a_2), \ldots, R_n(a_i, a)$ in $A$ in the ABox. Many practical ABoxes are acyclic or can be transformed to equivalent acyclic ABoxes.

Note all the role assertions in an acyclic ABox form tree-shape relations between individuals. We call an individual without any predecessor a root individual, and that without any successor a leaf individual.

Algorithm 2 is developed to transform a given acyclic $\mathcal{ALC}$ ABox into an equivalent complete ABox. The correctness of the algorithm shows that any acyclic ABox can be transformed to an equivalent complete ABox in $\mathcal{ALC}$.

Note that Algorithm 2 always terminates.

**Lemma 5.1.** Given an acyclic $\mathcal{ALC}$ ABox $A$, Algorithm 2 always returns a complete ABox $A'$ that is equivalent to $A$.

With the notion of complete ABox, we can extend n-forgetting in TBoxes and define n-forgetting for an $\mathcal{ALC}$ KB as follows.

For the remainder of this section, we assume that the ABox of every KB $K$ is acyclic.
Algorithm 2 (Complete an acyclic ABox)

Input: An acyclic $\mathcal{ALC}$ ABox $\mathcal{A}$.
Output: An equivalent complete $\mathcal{ALC}$ ABox $\mathcal{A'}$.

Method:

Step 1. Starting from root individuals, for each individual $a$ and each role assertion $R(a, b)$ in $\mathcal{A}$, add $\forall R.a \in b$ to the conjunction of all the concept assertions about $a$ in $\mathcal{A}$.

Step 2. For each individual $a$ in $\mathcal{A}$, add $\forall (a)$ to $\mathcal{A}$.

Step 3. Starting from leaf individuals, for each individual $b$ and each role assertion $R(a, b)$ in $\mathcal{A}$, add assertion $(\exists R.E)(a)$ to $\mathcal{A}$, where $E(b)$ is the conjunction of all the concept assertions about $b$ in $\mathcal{A}$.

Step 4. Return the resulting ABox.

\[\text{Fig. 2. Transform an acyclic ABox into a complete ABox.}\]

Definition 5.3. Let $\mathcal{K} = (T, \mathcal{A})$ be an $\mathcal{ALC}$ KB and $\mathcal{V}$ a set of variables. For each $n \geq 0$, the KB forget$^n(\mathcal{K}, \mathcal{V}) = (T', \mathcal{A'})$ is called the result of $n$-forgetting about $\mathcal{V}$ in $\mathcal{K}$, where $T' = \text{forget}^n(T, \mathcal{V}) = \{\top \subseteq \text{forget}(C_i^{(n)}, \mathcal{V})\}$ and $\mathcal{A'}$ is obtained from $\mathcal{A}$ through the following steps:

1. For each individual name $a$ in $\mathcal{A}$, add $C_i^{(n)}(a)$ to $\mathcal{A}$.
2. Apply Algorithm 2 to obtain a complete ABox, still denoted $\mathcal{A}$.
3. For each individual name $a$ in $\mathcal{A}$, replace $C(a)$ with $(\text{forget}(C, \mathcal{V}))(a)$, where $C(a)$ is the conjunction of all the concept assertions about $a$ in $\mathcal{A}$.
4. Remove each $R(a, b)$ from $\mathcal{A}$ where $R \in \mathcal{V}$.

The basic idea behind Definition 5.3 is to transform the given KB into a new KB such that forgetting can be done in its ABox and TBox, separately, in terms of c-forgetting for individual assertions and inclusions.

Example 5.2. Consider the KB $\mathcal{K} = (T, \mathcal{A})$ in Example 3.2 and let $\mathcal{V} = \{B, C\}$.

For each $n \geq 0$, let $\mathcal{A}_n$ be the ABox of forget$^n(\mathcal{K}, \mathcal{V})$. We will elaborate the computation of $\mathcal{A}_0$ as follows: Note that $\mathcal{A}$ is acyclic. First $C_T^{(0)}(a)$ and $C_T^{(0)}(b)$ are added into $\mathcal{A}$, where $C_T^{(0)}$ is the same as in Example 5.1. After applying Algorithm 2 to $\mathcal{A}$, the resulting ABox is equivalent to

\[
\{ \langle B \sqcap C \sqcap \forall R.C \sqcap D \rangle(a), R(a, b), \{ (\neg A \sqcup B) \sqcap C \sqcap \forall R.C \sqcap D \rangle(b), \\
\exists R.(\neg A \sqcup B) \sqcap C \sqcap \forall R.C \sqcap D \rangle(a) \}
\]

By applying c-forgetting to the conjunctions of concept assertions about, respectively, $a$ and $b$, we obtain $\mathcal{A}_0 = \{D(a), R(a, b), D(b)\}$.

Similarly, we can compute $\mathcal{A}_1, \ldots, \mathcal{A}_n$ as:

\[
\mathcal{A}_1 = \{D(a), (\forall R.D)(a), R(a, b), D(b), (\forall R.D)(b)\},
\]

\[
\mathcal{A}_n = \{D(a), (\forall R.D)(a), \ldots, (\forall R.R \cdots \forall R.D)(a), R(a, b), \\
D(b), (\forall R.D)(b), \ldots, (\forall R.R \cdots \forall R.D)(b)\}.
\]
The following result shows that \( n \)-forgetting preserves logical consequences of the original KB.

Given a concept description \( C \), let \( |C| \) be the number of all different subconcepts of \( C \). For a TBox \( T \), define \( |T| = \sum_{C \subseteq D \in T} (|C| + |D|) \). Similarly, for an ABox \( A \), define \( |A| = \sum_{C(a) \in A} |C| \). Then for a KB \( K = (T, A) \), \( |K| = |T| + |A| \).

**Proposition 5.1.** Let \( K \) be an ALC KB and \( V \) be a set of variables. Then \( \text{forget}^n(K, V) \) satisfies the following conditions:

1. \( K \models \text{forget}^n(K, V) \).
2. Let \( C \) and \( D \) be two concept descriptions containing no variable in \( V \). If \( n \geq 2^{(|C|+|D|+|K|)} \), then \( K \models C \subseteq D \) iff \( \text{forget}^n(K, V) \models C \subseteq D \).
3. Let \( C \) be a concept description containing no variable in \( V \), and \( a \) an individual name in \( K \). If \( n \geq 2^{(|C|+|K|)} \), then \( K \models C(a) \) iff \( \text{forget}^n(K, V) \models C(a) \).
4. Let \( R \) be a role name not in \( V \), and \( a, b \) two individual names in \( K \). Then \( K \models R(a, b) \) iff \( \text{forget}^n(K, V) \models R(a, b) \).

Recall from the definition of KB-forgetting that, with respect to inclusions and assertions not containing variables in \( V \), \( K \) is logically equivalent to \( \text{forget}(K, V) \). Proposition 5.1 tells us that if we know which inclusions and assertions not containing variables in \( V \) we wish to reason about in advance, then we can derive a value for \( n \), compute \( \text{forget}^n(K, V) \), and use the fact that, with respect to these inclusions and assertions, \( \text{forget}^n(K, V) \) is logically equivalent to \( K \) and hence to \( \text{forget}(K, V) \). In this way, we can use \( \text{forget}^n(K, V) \) as a practical approximation to \( \text{forget}(K, V) \).

The above proposition shows that, for any \( n \geq 0 \), each \( \text{forget}^n(K, V) \) is logically weaker than \( \text{forget}(K, V) \). Also, as the number \( n \) is sufficiently large, \( \text{forget}^n(K, V) \) preserves more and more consequences of \( K \). Therefore, the sequence of KBs \( \{\text{forget}^n(K, V)\}_{n \geq 0} \) is non-decreasing w.r.t. semantic consequence as the next proposition shows.

**Proposition 5.2.** Let \( K \) be an ALC KB and \( V \) a set of variables. Then, for any \( n \geq 0 \), we have \( \text{forget}^{n+1}(K, V) \models \text{forget}^n(K, V) \).

Based on the above two results, we can show the main theorem of this section as follows, which states that the limit of the sequence of \( n \)-forgettings captures the result of forgetting.

**Theorem 5.1.** Let \( K \) be an ALC KB and \( V \) a set of variables. Then

\[
\text{forget}(K, V) = \bigcup_{n=0}^{\infty} \text{forget}^n(K, V).
\]

So, by Theorem 5.1, we can compute \( \text{forget}(K, V) \), if it exists, using algorithms introduced in the paper.

**Corollary 5.1.** Let \( K \) be an ALC KB and \( V \) be a set of variables. \( V \) is forgettable from \( K \) if and only if there exists \( N \geq 0 \) such that \( \text{forget}^n(K, V) \equiv \text{forget}^N(K, V) \) for all \( n \geq N \). In this case, \( \text{forget}(K, V) = \text{forget}^N(K, V) \).
As we can see from Example 3.2, the sizes of consequences (assertions and inclusions) of $K$ not containing variables in $V$ do not have an upper bound. If it does not exist, we can always choose $n$ large enough to “approximate” $\text{forget}(K, V)$. However, two issues are still unclear to us: First, the computation of $\text{forget}^{n+1}(K, V)$ is not based on $\text{forget}^n(K, V)$. Next, it would be interesting to find a way to measure how close $\text{forget}^n(K, V)$ is from $\text{forget}(K, V)$.

6 Conclusion

We have presented a theory and methods for forgetting for knowledge bases in the expressive description logic $\text{ALC}$. This is the first work that deals with the combination of concepts and roles, nonempty ABoxes, and $\text{ALC}$. Because the result of forgetting may not exist for $\text{ALC}$ knowledge bases, we define a sequence of finite knowledge bases that approximate the result of forgetting and serve as a basis for query answering over the ontology with forgotten concepts and roles. We provide algorithms for computing these approximations, using algorithms for forgetting for concept descriptions, and note their correctness. Proofs of our results may be found in an online report.

Many interesting problems remain unsolved. It would be useful to clarify the decision problem of whether the result of forgetting for an $\text{ALC}$ knowledge base exists. It would be useful to extend the results of this paper to even more expressive description logics. It would be interesting to find lower bounds on the complexity of forgetting for description logics. It would be useful to find an incremental algorithm for computing approximations. It would be useful implement our algorithms and incorporate them into ontology editors such as Protégé [28]. See http://www.cit.gu.edu.au/~kewen/DLForget for our progress on this task.

References