Towards A Non Monotonic Description Logics Model

Martín O. Moguillansky  
mom@cs.uns.edu.ar

Marcelo A. Falappa  
mfalappa@cs.uns.edu.ar

Laboratorio de Investigación y Desarrollo en Inteligencia Artificial (LIDIA)  
Departamento de Ciencias e Ingeniería de la Computación (DCIC)  
Universidad Nacional del Sur (UNS)  
Consejo de Investigaciones Científicas y Técnicas (CONICET)  
Av. Alem 1253 - (B8000CPB) Bahía Blanca - Argentina

Abstract

In order to deal with the Ontology Change problem and considering an environment where Description Logics (DLs) are used to describe ontologies, the question of how to integrate distributed ontologies appears to be in touch with Belief Revision since DL terminologies may define same concept descriptions of a not necessarily same world model. A possible alternative to reason about these concepts is to generate unique concept descriptions in a different terminology. This new terminology needs to be consistently created, trying to deal with the minimal change problem, and moreover, yielding a non-monotonic layer to express ontological knowledge in order to be further updated with new distributed ontologies.

Keywords: Belief Revision, Description Logics, Non Monotonic Reasoning, Ontology Change.

1 INTRODUCTION

In order to reason about different ontologies, probably allocated in different places round the web, we will consider translated OWL ontologies into description logics (DLs). In DLs, the concept of Knowledge Base (KB) is composed of two main parts, TBoxes or Terminologies and ABoxes or Assertions. In this paper we focuss our investigation on how to reason about terminologies. Here many possibilities come through.

Just think about two distinct terminologies modeling each two different worlds, but containing a same subset of concepts (referred as Ontology Integration in [Flo06]). Or just two distinct terminologies modeling the same world, where naturally a common subset of concepts will be described as part of both terminologies (referred as Ontology Merging in [Flo06]). Furthermore, it might be probably impossible to get two concepts defined by different persons with exactly the same logic intention. Here is where the theory change arises as relevant protagonist in order to join consistently two terminologies redefining or reinforcing sub-concepts. The remainder of this paper is disposed as follows. The next section gives a brief description of the DL formalism, continued by section 3 with the analogous description of the theory change model, section 4 contributes to the formalization of merging DL terminologies, and describes an example operation of two different terminologies. Finally section 5 concludes and explains the related and future work in the area.

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2 THE DL BASIC FORMALISM

A Knowledge Representation (KR) system based on Description Logics (DL) provides a formalization to specify the knowledge base (KB) contents, a way to reason about it, and a process to infer implicit knowledge. A KB is composed by two components. A TBox to manage the terminology of the application world and an ABox containing the assertions about named individuals in terms of the previous concepts.

A terminology is composed by atomic concepts which denote sets of individuals and atomic roles to manage relationships between individuals. Besides, complex concepts and roles are built from the atomics using given constructors. Reasoning tasks are dedicated to determine whether a description is satisfiable (i.e. non-contradictory), or whether one description is more general than another one, that is, whether the first subsumes the second.

For an ABox, the problem is to verify the consistency of each set of assertions (i.e. test if there is a model for the set) and find out whether a particular individual is an instance of a concept description in the TBox depending on the assertions in the ABox. The environment will interact with the KR by querying the KB and finally by adding and retracting concepts, roles and assertions.

2.1 Description Languages

Description Languages are defined by the constructors they provide. In this paper we will consider a subset of the large DL constructors set investigated so far. The basic Description Language introduced by [SSS91] is the $\mathcal{AL}$ (Attribute Language). Let $A$ be an atomic-concept, $R$ an atomic-role, and $C, D$ complex concepts, the grammar for the $\mathcal{AL}$ language is defined as follows,

$$\begin{align*}
C, D & \rightarrow A | \top | \bot | \neg A | C \sqcap D | \forall R.C | \exists R.\top
\end{align*}$$

To define the formal semantics of $\mathcal{AL}$-concepts we use interpretations ($\lambda$) that consist of a non-empty set $\Delta^3$ (the domain of the interpretation) and an interpretation function $\lambda^3$, that assigns to every atomic concept $A$ a set $A^3 \subseteq \Delta^3$ and to every atomic role $R$ a binary relation $R^3 \subseteq \Delta^3 \times \Delta^3$. The interpretation function $\lambda = (\Delta^3, \lambda^3)$ is extended to concept descriptions as follows,

$$\begin{align*}
\top^3 &= \Delta^3 \\
\bot^3 &= \emptyset \\
(\neg A)^3 &= \Delta^3 \setminus A^3 \\
(C \sqcap D)^3 &= C^3 \cap D^3 \\
(\forall R.C)^3 &= \{a \in \Delta^3 | \forall b.(a, b) \in R^3 \rightarrow b \in C^3\} \\
(\exists R.\top)^3 &= \{a \in \Delta^3 | \exists b.(a, b) \in R^3\}
\end{align*}$$

In [BN02] is exhaustively detailed all language extensions depending on the constructors allowed in it, and namely $\mathcal{AL}[\cup][\exists][\cap][\forall][\neg][\parallel][\bot]$ In Table I a brief summary is given.

Example 1: Considering a language constructor $\cap$ and given a role completedCourse, a student of computer science is considered advanced if he has passed 15 courses out of a total of 25, $\geq 15$ completedCourse $\sqcap \leq 25$ completedCourse. For $\parallel$, the number restrictions are concerned with roles limited to a certain concept. In this case, one can also say that a student should pass at least 6 logic courses and 9 computational courses to be considered advanced,

$$\begin{align*}
\geq 6 \text{ completedCourse.LogicCourse} & \sqcap \\
\geq 9 \text{ completedCourse.ComputationalCourse} & \sqcap \\
& \leq 25 \text{ completedCourse}
\end{align*}$$

\footnote{The use of $\cap$ stands for complement. For $\cap$ and $\parallel$, $n$ varies over the nonnegative integers, and $\|X\|$ stands for the cardinality of the set $X$.}
I also role descriptions. An interpretation indicates how concepts or roles are related to each other following the Terminological axioms.

### 2.2 Terminologies

**Terminological axioms** indicate how concepts or roles are related to each other following the inclusion form, $C \sqsubseteq D$ (or $R \sqsubseteq S$), or the equality form, $C \equiv D$ (or $R \equiv S$), where $C$ and $D$ are concepts ($R$ and $S$ are roles).

An interpretation $I$ satisfies an inclusion $C \sqsubseteq D$ if $C^3 \sqsubseteq D^3$, and it satisfies an equality $C \equiv D$ if $C^3 = D^3$. Now given a set of axioms $\mathcal{T}$, an interpretation $I$ satisfies $\mathcal{T}$ iff $I$ satisfies each element of $\mathcal{T}$. If $I$ satisfies an axiom in $\mathcal{T}$, then we say that it is a *model* of this axiom in $\mathcal{T}$. Then two axioms or two sets of axioms are equivalent if they have the same models.

**Definitions** are used to describe complex concepts and made abstraction of them using a single name. An atomic concept on the left side of an equality defines the complex description explained on its right side.

A set of definitions $\mathcal{T}$ is called a terminology or a TBox if a symbolic name is defined only once. A terminology $\mathcal{T}$ contains a cycle iff there exists an atomic concept in $\mathcal{T}$ that uses itself [BN02]; otherwise $\mathcal{T}$ is called acyclic. An acyclic terminology $\mathcal{T}$ can be expanded iteratively through each definition in it, replacing each occurrence of a name on the right hand side with the concepts that it stands for. Now, we say that a terminology $\mathcal{T}$ is definitorial if it is acyclic, and we call to its semantics descriptive semantics. Those semantics that are motivated by the use of intuitively cyclic definitions are called fixpoint semantics. We will not consider fixpoint semantics in this paper.

### 2.3 Role Constructors

Binary relations between concepts are modeled by roles. If every role name is considered a role description or atomic role, and if $R$ and $S$ are roles descriptions, then $R \sqcap S$ (intersection), $R \sqcup S$ (union), $\neg R$ (complement), $R \circ S$ (composition), $R^+$ (transitive closure), and $R^-$ (inverse) are also role descriptions. An interpretation $I$ is adapted to the inverse role description as follows.

**Inverse** ($I^{-1}$): $(R^-)^3 = \{(b, a) \in \Delta^3 \times \Delta^3 \mid (a, b) \in R^3\}$

**Example 2**: For instance a hasParent role is obtained by applying the inverse role constructor to a given hasChild role. □
2.4 Properties for Reasoning

Given a terminology $\mathcal{T}$, if there is some interpretation of a concept that satisfies the axioms in $\mathcal{T}$ (a model of $\mathcal{T}$), then the concept denotes a nonempty set for the interpretation, furthermore this concept is known to be satisfiable w.r.t. $\mathcal{T}$. Otherwise it is called unsatisfiable. Formally,

Satisfiability [BN02]: A concept $C$ is satisfiable w.r.t. $\mathcal{T}$ if there exists a model $I$ of $\mathcal{T}$ such that $C^I$ is nonempty. In such a case we say that $I$ is a model of $C$.

Checking (un)satisfiability of concepts might be considered a key inference given that a number of other important inferences for concepts can be reduced to it. For instance, in order to check whether a domain model is correct, or to optimize concepts, we may want to know whether one concept is more general than another. This is called the subsumption problem. A concept $C$ is subsumed by a concept $D$ if in every model of $\mathcal{T}$, $C$ is a subset of $D$.

Subsumption [BN02]: A concept $C$ is subsumed by a concept $D$ w.r.t. $\mathcal{T}$ if $C^I \subseteq D^I$ for every model $I$ of $\mathcal{T}$. In such a case we write $C \sqsubseteq_{\mathcal{T}} D$ or $\mathcal{T} \models C \sqsubseteq D$.

A new kind of reasoning algorithms in DLs raised from the approach of considering satisfiability checking as the main inference. These algorithms are known as Tableaux Algorithms and can be understood as a specialized tableaux calculi.

3 THE BELIEF DYNAMIC MODEL

A belief base is a knowledge state represented through a set of sentences not necessarily closed under logical consequence. We also know that a belief set is a set of sentences of a determined language, closed under logical consequence. In general, a belief set is infinite being this the main reason of the impossibility to deal with this kind of sets in a computer. Instead, it is possible to characterize the properties that must satisfy each of the change operations on finite representations of a knowledge state.

The classic operations in the theory change are expansions, contractions, and revisions. An Expansion operation noted with “+”, adds a new belief to the epistemic state, without guaranteeing its consistency after the operation. A Contraction operation, noted with “−”, eliminates a belief $\alpha$ from the epistemic state and those beliefs that make possible its deduction or inference. The sentences to eliminate might represent the minimal change on the epistemic state. Finally, a Revision operation (“∗”) inserts sentences to the epistemic state, guaranteeing consistency (if it was consistent before the operation)[AGM85] [Gar98]. This means that a revision adds a new belief and perhaps it eliminates others to avoid inconsistencies. Other non-classical operations exists, like Merge operation [Fuh96][FKS02] noted with “◦”, that fusions belief bases or sets assuring a consistent resultant epistemic state, and a Consolidation operation (“! ”) that restores consistency to the epistemic state [Han97].

3.1 Kernel Contractions

The Kernel Contraction operator is applicable to belief bases and belief sets. It consist of a contraction operator capable of the selection and elimination of those beliefs in $K$ that contribute to infer $\alpha$.

Definition 3.1.1 - [Han94]: Let $K$ be a set of sentences and $\alpha$ a sentence. The set $K^\cup \alpha$, called set
of kernels is the set of sets $K'$ such that (1) $K' \subseteq K$, (2) $K' \vdash \alpha$, and (3) if $K'' \subset K'$ then $K'' \not\vdash \alpha$. The set $K^{\perp \perp} \alpha$ is also called set of $\alpha$-kernels and each one of its elements are called $\alpha$-kernel.

For the success of a contraction operation, we need to eliminate, at least, an element of each $\alpha$-kernel. The elements to be eliminated are selected by an Incision Function.

**Definition 3.1.2 - [Han94]:** Let $K$ be a set of sentences and "$\sigma$" be an incision function for it such that for any sentence $\alpha$ it verifies, (1) $\sigma(K^{\perp \perp} \alpha) \subseteq \bigcup (K^{\perp \perp} \alpha)$ and (2) If $K' \in K^{\perp \perp} \alpha$ and $K' \neq \emptyset$ then $K' \cap \sigma(K^{\perp \perp} \alpha) \neq \emptyset$.

Once the incision function was applied, we must eliminate from $K$ those sentences that the incision function selects, i.e. the new belief base would consist of all those sentences that were not selected by $\sigma$.

**Definition 3.1.3 - [Han94]:** Let $K$ be a set of sentences, $\alpha$ a sentence, and $K^{\perp \perp} \alpha$ the set of $\alpha$-kernels of $K$. Let "$\sigma$" be an incision function for $K$. The operator "$-\sigma$", called kernel contraction determined by "$\sigma$", is defined as, $K -_{\sigma} \alpha = K \setminus \sigma(K^{\perp \perp} \alpha)$.

Finally, an operator "$-$" is a kernel contraction operator for $K$ if and only if there exists an incision function "$\sigma$" such that $K - \alpha = K -_{\sigma} \alpha$ for all sentence $\alpha$.

### 3.2 Merging Belief Bases

Fuhrmann defined in [Fuh96] a partial meet merge operation as a union of two bases, not necessarily closed under logic consequence, and a later consistency restoring applying a bottom contraction. Inspired on it we propose a merge operation over two bases, defined by means of the Kernel Contraction operator, and determined by an Incision Function, as follows.

**Definition 3.2.1 - Merge:** Let "$-$" be a kernel contraction for the union of two belief bases $K_1 \cup K_2$, determined by an incision function "$\sigma$". Then the Merge for Belief Bases operator $\circ$ is defined as, $K_1 \circ K_2 = (K_1 \cup K_2) -_{\sigma} \perp$.

### 4 CONSISTENT TERMINOLOGY INTEGRATION

In order to reason about two presumably distributed and potentially inconsistent terminologies, a non monotonic operation for integration of both terminologies in a new consistent one is required. For achieving a formal definition of such an operator the following features might be desirable.

- A function that maps a concept name defined in one terminology to another concept name in the second terminology in order to recognize different concept names referring to a same concept of the real world.
- A theory change framework would be an interesting environment for revising beliefs and merging knowledge bases defined as part of the given terminologies.
- Conveniently, a concept defined in a given terminology will be expanded in order to generate its correspondent belief base in the theory change. This means that each concept description in it is expressed as a conjunction of basic concept descriptions in the same terminology.
- A translation method for expressing DL concepts as part of a first order logic language might be needed in order to apply the theory change operations.
- Contractions might retract minimal information. This means that an operation for merging terminologies should keep as much as possible knowledge in a consistency restoring process. This property reflects the minimal change requirement of the theory change.
4.1 Basic Definitions

Formal definitions might be interesting to specify reflecting some of the previous features to be the basis of a terminology integration operator.

**Definition 4.1.1 - Concept Id.** Let $T$ be a terminology composed of $n$ distinct definitions, the $i^{th}$ concept description $D$ defined in the terminology $T$ will be identified as $D^T_i$, where $1 \leq i \leq n$.

**Definition 4.1.2 - Names Mapping** Let $\xi$ be a mapping function from a concept name defined in a terminology $T_1$ to a different concept name defined in a terminology $T_2$. Then $D^T_1i = \xi(D^T_2j)$ indicates that $D^T_1i$ and $D^T_2j$ are name identifiers of a unique concept of the real world.

**Definition 4.1.3 - Consistent Terminology Integration** Let $\Join\Delta$ be the operator for a consistent terminology integration $T_1 \Join\Delta T_2$. Such an operation will be composed of two consecutive sub-operations, terminology unification and terminology consolidation.

Intuitively, a terminology unification consist of copying every concept description from each terminology to a new one except for the application of a non monotonic concept conjunction. In such a case, those concepts defined in both terminologies that refers to a same concept in the real framework, identified by a names mapping function $\xi$, are consistently unified in a unique new concept. The following sub-operation is defined as terminology consolidation. Here the idea is to restore consistency to the unified terminology $T$, capturing and solving inconsistencies that yields the unification process in concept descriptions that refers to a yet unified concept in its right hand side. A tentative unification operation was provided in [MF06] although this version did not consider a more complex case of study that we capture in this paper with the consolidation operation.

4.2 From Description Logics to First Order Logic Languages

DLs are not rule based languages but they may be translated into fragments of first order logics (FOL) in order to make an easier mapping to rule languages. By this, efficient logic programming based reasoners and deductive systems may be defined to make inference about the knowledge representation originally specified by a description language. Table 2, originally specified in [GHV03], summarizes the translation rules above introduced.

<table>
<thead>
<tr>
<th>DL</th>
<th>FOL</th>
</tr>
</thead>
<tbody>
<tr>
<td>$D \equiv C$</td>
<td>$\forall x. C(x) \iff D(x)$</td>
</tr>
<tr>
<td>$C \subseteq D$</td>
<td>$\forall x. C(x) \rightarrow D(x)$</td>
</tr>
<tr>
<td>$C$</td>
<td>$C(x)$</td>
</tr>
<tr>
<td>$C \cap D$</td>
<td>$C(x) \land D(x)$</td>
</tr>
<tr>
<td>$\neg C$</td>
<td>$\neg C(x)$</td>
</tr>
<tr>
<td>$\exists R.C$</td>
<td>$\exists x. (R(y, x) \land C(x))$</td>
</tr>
<tr>
<td>$\forall R.C$</td>
<td>$\forall x. (R(y, x) \rightarrow C(x))$</td>
</tr>
<tr>
<td>$\geq n R$</td>
<td>$\exists y_1, \ldots, y_n. (R(x, y_1) \land \ldots \land R(x, y_n)) \land (\bigwedge_{1 \leq i &lt; j &lt; n} y_i \neq y_j)$</td>
</tr>
<tr>
<td>$\leq n R$</td>
<td>$\forall y_1, \ldots, y_{n+1}. (R(x, y_1) \land \ldots \land R(x, y_{n+1})) \rightarrow (\bigvee_{1 \leq i &lt; (n+1), i &lt; j \leq (n+1)} y_i = y_j)$</td>
</tr>
</tbody>
</table>

Table 2: DL - FOL equivalence.

The following definitions describe how our proposal manages DLs concepts descriptions as fragments in the theory change.

**Definition 4.2.1 - KB of a Concept Description** Let $C \equiv C_1 \sqcap \ldots \sqcap C_n$ be an expanded concept

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2 We adopt a $ALC\bar{E}\bar{N}$ description language, where the classical $AL$ attribute language is extended by complement, existential quantification, and number restriction concept constructors.
description, then the set $K(C)$ will be the knowledge base for the concept description $C$ such that $\phi_{C_1}, \ldots, \phi_{C_n} \in K(C)$, where $\phi_{C_i}$ is the first order logic translation of the concept $C_i$.

**Observation 4.2.2:** A DL conjunction $C_1 \sqcap C_2$ is interpreted in the theory change as $K(C_1) \cup K(C_2)$.

**Proof:** Let $C \equiv C_1 \sqcap C_2$ be a concept description, and let $C_1 \equiv A_1 \sqcap \ldots \sqcap A_n$, and $C_2 \equiv B_1 \sqcap \ldots \sqcap B_n$ be their respective atomic concept descriptions. Then the expanded concept description for $C$ is $C \equiv A_1 \sqcap \ldots \sqcap A_n \sqcap B_1 \sqcap \ldots \sqcap B_n$. Finally using definition [4.2.1] follows that $K(C) = K(C_1) \cup K(C_2)$. $\square$

### 4.3 Non Monotonic Concept Conjunction ($\sqcap$)

Let “$\sqcap$” be a *Non Monotonic Concept Conjunction* DL operator used as part of the terminology unification process and furthermore in the specification of the terminology consolidation, such that $D = D_1 \sqcap D_2$ defines a new *satisfiable* concept description $D$ from consistently unifying concepts $D_1$ and $D_2$ by the application of the operator $\sqcap$.

A consistent conjunction of two such a concept descriptions may intuitively be thought as a merge of the belief bases representing each concept. Afterwards, the resultant belief base should be translated back to the original description language in order to express the result as a new concept description. A translation method from DLs to belief bases in the theory change is necessary, then following the translation rules from a description language to a subset of first order logic rules in table 2, we may obtain the conversions of concept descriptions to belief bases in the theory change.

**Definition 4.3.1 - Non Monotonic Concept Conjunction:** Let $D_1$ and $D_2$ be two concept descriptions and let $K(D_1)$ and $K(D_2)$ be their correspondent belief bases (not necessarily closed under logical consequence). A *non monotonic concept conjunction* operation $D_1 \sqcap D_2$, is interpreted as a merge operation for belief bases, such that $D_1 \sqcap D_2 \iff K(D_1) \circ K(D_2)$.

**Observation 4.3.2:** If the conjunction $D_1 \sqcap D_2$ is satisfiable\(^3\) then the non monotonic conjunction $D_1 \sqcap D_2$ is equivalent to $D_1 \sqcap D_2$.

**Proof:** Suppose we have two concept descriptions $D_1$ and $D_2$, then by definition [4.2.1] their respective knowledge bases are $K(D_1)$ and $K(D_2)$.

($\Rightarrow$) Consider the non monotonic concept conjunction $D_1 \sqcap D_2$, by definition [4.3.1] it may be though as a merge operation of their knowledge bases such that $K(D_1) \circ K(D_2)$. Using definition [3.2.1] (merge) we have $(K(D_1) \cup K(D_2)) \neg \sigma \bot$. Then by definition [3.1.3] (kernel contraction) it is equivalent to $(K(D_1) \cup K(D_2)) \sigma((K(D_1) \cup K(D_2)) \downarrow \bot)$.

Now suppose that *(by hypothesis)* $D_1 \sqcap D_2$ is satisfiable, then by definition [3.1.1] (set of kernels) it follows that there is no $K' \subseteq (K(D_1) \cup K(D_2))$ such that $K' \vdash \bot$. Thus, we have no beliefs to select from $(K(D_1) \cup K(D_2))$, and finally by observation [4.2.2] we have $D_1 \sqcap D_2$ (as we wanted to prove). The opposite ($\iff$) direction can be similarly proved. $\square$

The following algorithm describes the operation $D \equiv D_1 \sqcap D_2$ defined in definition [4.3.1] and optimizes it considering the observation [4.3.2]

### 4.4 Consistent Merging of Terminologies ($\bowtie$)

As previously specified, an operation $\mathcal{T} = \mathcal{T}_1 \bowtie \mathcal{T}_2$, consist of two sub-operations; *unification* of both terminologies $\mathcal{T}_1$ and $\mathcal{T}_2$ in a new one $\mathcal{T}$, and the respective *consolidation* of $\mathcal{T}$ for consistency restoring. In what follows we provide the correspondent algorithms for both sub-operations.

**Observation 4.4.1:** If no loss of knowledge is desired in a concept unification process – *i.e.* that the

\(^3\)A conjunction $D_1 \sqcap D_2$ is satisfiable if there exists a model $\mathcal{T}$ of $\mathcal{T}$ such that $(D_1 \sqcap D_2)^T$ is non empty.
Algorithm 1 Non Monotonic Concept Conjunction $D \equiv D_1 \sqcap D_2$

Input: Two expanded concepts $D_1, D_2$.

Output: $D$.

if $D_1 \sqcap D_2$ is un-satisfiable then
    $D \leftarrow K(D_1) \circ K(D_2)$.
else
    $D \leftarrow D_1 \sqcap D_2$
end if

Algorithm 2 Terminology Unification

Input: Two terminologies $T_1$ and $T_2$.

Output: A unified terminology $T$.

for all $D$ in $T_1$ or $T_2$ do
    if exists a mapping $\xi(D)$ for concept description $D$ then
        $\mathcal{D}^T \leftarrow D \sqcap \xi(D)$ is a new concept in $T$
    else
        $\mathcal{D}^T \leftarrow D$ is a new concept in $T$
    end if
end for

Algorithm 3 Terminology Consolidation

Input: The unified terminology $T$.

Output: The consolidated terminology $\mathcal{T}$.

for all $\mathcal{D}^T$ do
    for all $C^T \equiv D \sqcap C_{tail}$, where $C_{tail}$ may be thought as $C_1 \sqcap C_2 \sqcap ... \sqcap C_n$ do
        if $\mathcal{D}^T \sqcap C_{tail}$ is satisfiable then
            Replace the original definition $C^T \equiv D \sqcap C_{tail}$ by $\mathcal{C}^T \equiv \mathcal{D}^T \sqcap C_{tail}$
        else
            Let $S$ be $\sigma((K(\mathcal{D}^T) \cup K(C_{tail})) \bot \bot \bot)$ where the selection’s scope is restricted to $K(\mathcal{D}^T)$.
            Generate $\mathcal{D}_1^T \leftarrow K(\mathcal{D}^T) \setminus S$ in $\mathcal{T}$.
            Redefine $\mathcal{D}^T \equiv \mathcal{D}_1^T \sqcap S_{DL}$ in $\mathcal{T}$, where $S_{DL}$ is the description language translation of $S$.
            Replace the original definition $C^T \equiv D \sqcap C_{tail}$ by $\mathcal{C}^T \equiv \mathcal{D}_1^T \sqcap C_{tail}$.
        end if
    end for
end for

Observation 4.4.2: Note that the consolidation algorithm adds new concept descriptions noted as $\mathcal{C}$ and $\mathcal{D}$ and these new definitions are considered in later iterations.

4.5 Worked Example

In this section we show how two different terminologies might be consistently merged in a new one following the previous definitions.

The terminology expressed in table 4 shows among other definitions, some main characteristics of mammals and oviparous animals, and particularly defines monotremes to be a conjunction of both...
animal classes. When we try to merge terminologies $\mathcal{T}_1$ and $\mathcal{T}_2$ we find that the concept descriptions for mammals and oviparous in $\mathcal{T}_1$ yield the following contradiction,

$$\text{hasVoluntaryMovement} \sqcap \neg \text{hasVoluntaryMovement}$$

### 4.5.1 Method Application

In what follows we will develop the integration of both terminologies following the previous $\bowtie$ operation definition in order to see more clearly how the consistency problem is solved. Let consider the following mapping instances, $\text{Mammal}^{\mathcal{T}_1} = \xi(\text{Mammal}^{\mathcal{T}_2})$ and $\text{Oviparous}^{\mathcal{T}_1} = \xi(\text{Oviparous}^{\mathcal{T}_2})$.

Then, the unified terminology $\mathcal{T} = \mathcal{T}_1 \bowtie \mathcal{T}_2$ will have the following concept descriptions,

<table>
<thead>
<tr>
<th>Concept</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bird</td>
<td>$\text{Animal} \sqcap \text{Bipedal} \sqcap \text{Oviparous} \sqcap \text{hasFeathers} \sqcap = 2$\text{hasWings}$</td>
</tr>
<tr>
<td>Mammal</td>
<td>$\text{Animal} \sqcap \forall \text{giveBirth}.\text{LiveBirth}$</td>
</tr>
<tr>
<td>Oviparous</td>
<td>$\text{Animal} \sqcap \forall \text{giveBirth}.\text{Egg}$</td>
</tr>
<tr>
<td>Bipedal</td>
<td>$\text{hasFoot} = 2$</td>
</tr>
<tr>
<td>LiveBirth</td>
<td>$\text{hasHeartBeat} \sqcap \text{hasVoluntaryMovement}$</td>
</tr>
<tr>
<td>Egg</td>
<td>$\text{hasHeartBeat} \sqcap \neg \text{hasVoluntaryMovement}$</td>
</tr>
</tbody>
</table>

Table 3: A terminology $\mathcal{T}_1$ (TBox) with concepts about animals.

<table>
<thead>
<tr>
<th>Concept</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Platypus</td>
<td>$\text{Aquatic} \sqcap \text{Monotreme}$</td>
</tr>
<tr>
<td>Monotreme</td>
<td>$\text{Mammal} \sqcap \text{Oviparous}$</td>
</tr>
<tr>
<td>Mammal</td>
<td>$\text{Animal} \sqcap \geq 2$\text{hasMammaryGlands}$</td>
</tr>
<tr>
<td>Oviparous</td>
<td>$\text{Animal} \sqcap \geq 1$\text{layEggs}$</td>
</tr>
</tbody>
</table>

Table 4: A terminology $\mathcal{T}_2$ (TBox) with concepts about animals.

When merging the two belief bases $K(\text{Mammal}^{\mathcal{T}_1})$ and $K(\text{Mammal}^{\mathcal{T}_2})$, no inconsistency arises,

<table>
<thead>
<tr>
<th>Concept</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bird</td>
<td>$\text{Bird}^{\mathcal{T}_1}$</td>
</tr>
<tr>
<td>Platypus</td>
<td>$\text{Platypus}^{\mathcal{T}_2}$</td>
</tr>
<tr>
<td>Monotreme</td>
<td>$\text{Monotreme}^{\mathcal{T}_2}$</td>
</tr>
<tr>
<td>Mammal</td>
<td>$\text{Mammal}^{\mathcal{T}_1} \sqcap \text{Mammal}^{\mathcal{T}_2}$</td>
</tr>
<tr>
<td>Oviparous</td>
<td>$\text{Oviparous}^{\mathcal{T}_1} \sqcap \text{Oviparous}^{\mathcal{T}_2}$</td>
</tr>
<tr>
<td>Bipedal</td>
<td>$\text{Bipedal}^{\mathcal{T}_1}$</td>
</tr>
<tr>
<td>LiveBirth</td>
<td>$\text{LiveBirth}^{\mathcal{T}_1}$</td>
</tr>
<tr>
<td>Egg</td>
<td>$\text{Egg}^{\mathcal{T}_1}$</td>
</tr>
</tbody>
</table>

Table 5: The unified terminology $\mathcal{T}$ from the originals $\mathcal{T}_1$ and $\mathcal{T}_2$.

When merging the two belief bases $K(\text{Mammal}^{\mathcal{T}_1})$ and $K(\text{Mammal}^{\mathcal{T}_2})$, no inconsistency arises,

<table>
<thead>
<tr>
<th>Concept</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Animal</td>
<td>$\forall \text{giveBirth}$</td>
</tr>
<tr>
<td>hasMammaryGlands</td>
<td>$\forall \text{giveBirth}$.\text{hasMammaryGlands}$</td>
</tr>
</tbody>
</table>

The resultant unified concept will be translated from the previous belief base to the correspondent description language as, $\text{Mammal}^{\mathcal{T}} \equiv \text{Animal} \sqcap \geq 2$\text{hasMammaryGlands} \sqcap \forall \text{giveBirth}$. 
\[(\text{hasHeartBeat} \cap \text{hasVoluntaryMovement})\]. A similar situation occurs with the belief base for concept description \textit{Oviparous}, and its correspondent translation to DLs, \(\text{Oviparous}^T \equiv \text{Animal} \cap \geq 1\text{layEggs} \cap \forall \text{giveBirth}.(\text{hasHeartBeat} \cap \neg \text{hasVoluntaryMovement})\).

\[
K(\text{Oviparous}^T)
\]

\[
\begin{align*}
\text{Animal}(X) \\
\text{layEggs}(X, y_1) \\
\text{giveBirth}(Y, X) &\rightarrow \text{hasHeartBeat}(X) \land \neg \text{hasVoluntaryMovement}(X)
\end{align*}
\]

We developed so far the terminology unification, first sub-operation of the terminology integration previously defined. The following step is the application of the terminology consolidation in order to verify and restore consistency to the resultant unified terminology. Here, the concept description for \textit{Monotreme} yields the inconsistency.

\[
\text{Monotreme}^T \equiv \text{Animal} \cap \geq 2\text{hasMammaryGlands} \cap \\
\forall \text{giveBirth}.(\text{hasHeartBeat} \cap \text{hasVoluntaryMovement}) \cap \\
\text{Animal} \cap \geq 1\text{layEggs} \cap \forall \text{giveBirth}.(\text{hasHeartBeat} \cap \neg \text{hasVoluntaryMovement})
\]

Clearly, this concept is unsatisfiable, so from the set \(K(\text{Mammal}^T) \cup K(\text{Oviparous}^T)) \Downarrow \bot\) of \(\bot\)-kernels, an appropriate incision function \(\sigma\) would select rules only from \(K(\text{Mammal}^T)\) in order to avoid the inconsistency in this concept. In such a case the correspondent sentence would be

\[
giveBirth(Y, X) \rightarrow \text{hasHeartBeat}(X) \land \text{hasVoluntaryMovement}(X)
\]

that comes from the DL concept \(\forall \text{giveBirth}.(\text{hasHeartBeat} \cap \text{hasVoluntaryMovement})\) so a new concept \(\text{Mammal}^T_1\) is defined to be part of the terminology \(T\) without considering the previous selection, such that \(\text{Mammal}^T_1 \equiv \text{Animal} \cap \geq 2\text{hasMammaryGlands}\). Then the definitive concept for the definition of \(\text{Mammal}\) in \(T\) would be,

\[
\text{Mammal}^T \equiv \text{Mammal}^T_1 \cap \forall \text{giveBirth}.(\text{hasHeartBeat} \cap \text{hasVoluntaryMovement})
\]

Finally, following the consolidation algorithm, the concept description for \textit{Monotreme} in terminology \(T\) would be, \(\text{Monotreme}^T \equiv \text{Mammal}^T_1 \cap \text{Oviparous}^T\).

Note that following the proposed method for terminology integration we do not only eliminate the inconsistency when merging both terminologies, but also keep all information as part of the resultant terminology, by identifying and splitting the problematic concept in two interrelated, revisiting the hierarchy technique of the object oriented paradigm.

\section{5 CONCLUSIONS, RELATED AND FUTURE WORK}

A union operation of terminologies probably yields contradictions on concept descriptions and further inconsistency in the resultant terminology. The use of a belief revision framework to define terminologies in order to meet a consistent merge operation is proposed and generates a new \textit{Non-monotonic Description Logics} model as a powerful theory to be applied on future Semantic Web researches.

\textbf{Ontology Change} \cite{Flo06} expresses the necessity of modifying the knowledge described in ontologies responding to different given interests. For instance, ontology changes may arise due to some change originated in the world being modeled, or on users’ needs; and/or due to previously unknown knowledge, or bugs found after design steps.

\textsuperscript{4}The reader is invited to refer the conclusions of this work to understand this selection.
<table>
<thead>
<tr>
<th>Concept</th>
<th>Definition</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bird</td>
<td>Animal ∩ Bipedal ∩ Oviparous ∩ hasFeathers ⊑ 2hasWings</td>
</tr>
<tr>
<td>Platypus</td>
<td>Aquatic ∩ Monotreme</td>
</tr>
<tr>
<td>Monotreme</td>
<td>Mammal ∩ Oviparous</td>
</tr>
<tr>
<td>Mammal</td>
<td>Animal ⊑ 2hasMammaryGlands</td>
</tr>
<tr>
<td>Mammal₁</td>
<td>Animal ⊑ ∀giveBirth.LiveBirth</td>
</tr>
<tr>
<td>Oviparous</td>
<td>Animal ⊑ ∀giveBirth.Egg ⊑ 1layEggs</td>
</tr>
<tr>
<td>Bipedal</td>
<td>= 2hasFoot</td>
</tr>
<tr>
<td>LiveBirth</td>
<td>hasHeartBeat ∩ hasVoluntaryMovement</td>
</tr>
<tr>
<td>Egg</td>
<td>hasHeartBeat ⊑ ¬hasVoluntaryMovement</td>
</tr>
</tbody>
</table>

Table 6: The resultant terminology $\mathcal{T} = \mathcal{T}_1 \bowtie \mathcal{T}_2$ with concepts about animals.

Many authors are nowadays focusing their efforts in some closely related areas under terms like, ontology translation, evolution, integration and merging. For instance, in [Flo06] an AGM compatible contraction operator for DLs is described, and based on it a DL revision operator is defined in [QLB06], from the generalized AGM postulates in [Flo06]. These two works are some of the newest results obtained so far by considering theory change and DLs as an hybrid framework for belief revision.

Our proposed approach means an interesting alternative in the area, whose most notorious difference relays in the fact of the focalization on non-prioritized change operations where the AGM postulates are not fully met. From a naming convention point of view, although most of our work might be thought as Ontology Merging, we enclose it in the so called Ontology Integration, following the terminology specified in [Flo06], due to some slight overlapping that we consider exists between these terms.

The incision function ($\sigma$) deserves a special discussion due to its responsibility on the information management of knowledge bases which makes to keep or restore consistency regarding the selections it does. For instance, several merge operators have been proposed in the theory change bibliography, but in this work we proposed a different one based on a kernel contraction determined by an incision function. This is due to our intentions to reduce every belief revision and merge operation to an “intelligent” definition of the incision function, so placing it on top of our future investigations.

Another situation is given by concept descriptions to be consistently merged “retracting” those simple concepts descriptions that presumably generate the inconsistency in the new concept description. This selection is the one being made by the incision function, but here always might be more than one possible election to avoid inconsistency. More formally, $C \leftarrow K(C_1) \cup K(C_2) - \sigma \perp$ might be solved selecting knowledges from any of the two involved concept descriptions. Here is where Epistemic Entrenchment methods arise, in order to properly select, w.r.t. the local environment, the most convenient knowledge and to give only one possible concept description as the operation resultant.

Motivated by the latter examples, a deeper investigation on Epistemic Entrenchment methods could be useful to semi-automate the well functioning of a reasonable incision function ($\sigma$) to cut the $\alpha$-kernels obtained by the use of merge operations. This means that the incision function ($\sigma$) will select those sub-concepts with less epistemic entrenchment to be cut off the resulting definition.

To achieve this, we will investigate confidence levels on terminologies to be incorporated by the ontology integration requester depending on his confidence about the origin where the ontology came from (author). Furthermore, it would be interesting also to consider confidence levels on concept descriptions incorporated by the ontology designer. By this, we intend to have some extra information

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5The Epistemic Entrenchment method specifies a way to measure the level of importance of a sentence $\alpha$ to belong to the epistemic state.
at the time of evaluating which is the most relevant knowledge to be held after an integration operation.

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REFERENCES


