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Abstract
Many real world applications of ontologies call for reasoning with modular ontologies. Classical approaches to reasoning with ontology assume a single centralized, consistent ontology. However, in the case of multiple, autonomously developed ontology modules, it is usually neither possible nor desirable to integrate all involved modules into a single centralized ontology. In this paper, we propose a tableau-based reasoning algorithm based on Package-based Description Logics (P-DL), an ontology language that extends description logics with language features to support modularity. The algorithm adopts a federated approach to reasoning with modular ontologies wherein each ontology module has associated with it, a local reasoner. The local reasoners communicate with each other as needed in an asynchronous fashion.

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A Tableau-based Federated Reasoning Algorithm for Modular Ontologies*

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Abstract

Many real world applications of ontologies call for reasoning with modular ontologies. We describe a tableau-based reasoning algorithm based on Package-based Description Logics (P-DL), an ontology language that extends description logics with language features to support modularity. Unlike Classical approaches that assume a single centralized, consistent ontology, the proposed algorithm adopts a federated approach to reasoning with modular ontologies wherein each ontology module has associated with it, a local reasoner. The local reasoners communicate with each other as needed in an asynchronous fashion. Hence, the proposed approach offers an attractive approach to reasoning with multiple, autonomously developed ontology modules, in settings where it is neither possible nor desirable to integrate all involved modules into a single centralized ontology.

1 Introduction

Many real-world applications of ontologies, such as distributed collaboration and information integration, call for ontology languages with support for modularity in ontologies. Hence, there is a growing interest in ontology language features to support such modular ontologies as well as approaches to reasoning with multiple ontology modules including: Distributed Description Logics (DDL)[6], E-connections [9, 8] and Package-based Description Logics (P-DL) [5, 4]. Of particular interest in this context are algorithms for reasoning with multiple, distributed and autonomous ontology modules. For example, in a typical data integration application, a user submits queries expressed using terms in a user ontology that need to be answered from a set of autonomous data sources. In such a setting, the user query has to be translated from the user ontology (e.g. in the Gene Ontology (GO)) into queries expressed in terms of data source ontologies (e.g. the EC(Enzyme Commission) hierarchy or the MIPS Functional Catalogue). Such a translation requires reasoning with multiple semantically connected, but physically distributed ontologies.

Reasoning with ontologies in such a setting presents several challenges:

- The reasoning task involves not a single ontology, but a collection of ontologies about a domain of interest that are created and maintained by autonomous groups.
- In many cases, integrating distributed ontologies into one consistent centralized ontology is not possible for several reasons: the ontologies may be large and communication overhead is too expensive; the autonomous entities that control an ontology may be unwilling to share it in its entirety although be willing to answer queries to the ontology. In such a setting, it is not feasible to reduce the problem of reasoning over distributed ontology modules to the problem of reasoning over a single centralized ontology.
- In general, an ontology may reuse terms defined in another ontology. For example, current releases of GO, EC and MIPS ontologies all contain term correspondences from and to other ontologies. Mutual or cyclic reuse is also common.

Several authors have recently investigated distributed reasoning algorithms for modular ontologies. Serafini et. al. [11, 10] describe a tableau-based reasoning algorithm for DDL. The algorithm divides a reasoning problem w.r.t. a DDL TBox into several local reasoning problems answered by local modules. The basic idea behind this algorithm

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is to infer concept subsumption in one module from subsumptions in another module and inter-module bridge rules that relate concepts in one module to concepts in another module. For example, consider ontology modules $i$ and $j$ in which the concepts $A$, $B$ and $G$, $H$ respectively are defined, given the bridge rules $i : A \sqsupseteq j : G$, $i : B \sqsubseteq j : H$ and module $i$ entails $A \sqsubseteq B$, then it is possible for module $j$ to infer that $G \subseteq H$.

Grau et al. [8, 7] present a tableau-based reasoning procedure for $\mathcal{E}$-Connections. $\mathcal{E}$-connections divides roles into disjoint sets of local roles (connecting concepts in one module) and links (connecting concepts in different modules). For example, two modules about people ($L_1$) and pets ($L_2$) can be connected by a link owns, and $L_1$ can use such a link to build local concepts, e.g. $1 : DogOwner \sqsubseteq \existsowns.(2 : Dog)$. The tableau-based reasoning procedure for $\mathcal{E}$-Connections, implemented in the Pellet reasoner, generates a set of tableaux (trees) linked by $\mathcal{E}$-connection instances (cross-module role instances).

Bao et. al. [3] describe a distributed reasoning algorithm for P-DL with acyclic importing. This algorithm adopts a federated approach to reasoning using distributed storage of a global tableau. Some of local tableaux may share some nodes (i.e. “image” nodes) and communicate by sending messages to each other. Thus, search for a model of the ontology is distributed across the local tableaux.

However, existing approaches to reasoning with modular ontologies suffer from several limitations. Both DDL and $\mathcal{E}$-Connections, because of their limited expressivity, lack support for certain types of reasoning tasks. For example, DDLs have no support for inter-module role relations, whereas $\mathcal{E}$-connections lack inter-module subsumptions. Both DDL and P-DL reasoning algorithms do not allow mutual or cyclic references (bridge rules in DDL, term importing in P-DL) of concepts among ontology modules.

Current implementation of the $\mathcal{E}$-Connections reasoner, motivated by the “combined tableau” idea [8, 7], only “colors” local tableaux without separating them. Therefore, reasoning relies on one (combined) ABox thereby forcing the TBoxes of all modules to be loaded (through internalization) into the reasoner. The strategy actually loads all ontology modules into a single memory space thus makes a de facto ontology integration, which sacrifices many of the benefits of modular ontologies (e.g. scalability).

Against this background, we present an improved federated reasoning algorithm that overcomes many of these limitations and offers several advantages over existing approaches. It strictly avoids combining the local ontology modules in a centralized memory space using distributed reasoning with localized P-DL semantics thereby allowing local reasoning modules to operate in an asynchronous, peer-to-peer fashion. It supports reasoning with both inter-module subsumption and inter-module role relations and allows arbitrary references of concepts among ontology modules. The P-DL semantics also guarantees that the results of reasoning in the distributed setting are identical to those obtainable by applying a reasoner to an ontology constructed by integrating the different modules [4].

The rest of the paper is organized as follows. Section 2 briefly reviews the basic syntax and semantics of P-DL. Section 3 presents the details of the proposed reasoning algorithm. Section 4 establishes the soundness and completeness of the proposed algorithm. Section 5 concludes with a summary.

2 Package-based Description Logics

This section briefly reviews basic features of Package-based Description Logics (P-DL) as given in [5, 4]. In P-DL, an ontology is composed of a collection of modules called packages. Each term (name of a concept, property or individual) or axiom is associated with a home package:

Definition 1 (Package) Let $O = (\mathcal{S}, \mathcal{A})$ be an ontology, where $\mathcal{S}$ is the set of terms and $\mathcal{A}$ is the set of axioms over terms in $\mathcal{S}$. A package $P = (\Delta_S, \Delta_A)$ of the ontology $O$ is a fragment of $O$, such that $\Delta_S \subseteq \mathcal{S}$, $\Delta_A \subseteq \mathcal{A}$. A term $t \in \Delta_S$ or an axiom $t \in \Delta_A$ is called a member of package $P$, denoted as $t \in P$. $P$ is called the (only) home package of $t$, denoted as $\mathcal{H}(P)(t) = P$.

A package can use terms defined in other packages i.e., foreign terms: If a package $L_j$ uses a term $t$ with home package $L_i$ ($i \neq j$), then we say $t$ is imported into $L_j$. Formally, we have

Definition 2 (Foreign Term and Importing) A term $t$ that appears in a package $P$, but has a home package $Q$ that is different from $P$ is called a foreign term in $P$. We say that $P$ imports $Q$ : $t$ and denote it as $Q \xrightarrow{\mathcal{I}} P$. If any term defined in $Q$ is imported into $P$, we say that $P$ imports $Q$ and denote it as $Q \xrightarrow{\mathcal{I}} P$.

The importing closure $\mathcal{I}(P)$ of a package $P$ contains all packages that are directly or indirectly imported into $P$, such that:

- (direct importing) $R \rightarrow P \Rightarrow R \in \mathcal{I}(P)$
- (indirect importing) $Q \rightarrow R$ and $R \in \mathcal{I}(P) \Rightarrow Q \in \mathcal{I}(P)$

Definition 3 (Acyclic and Cyclic Importing) A P-DL ontology $\{P_i\}$ has acyclic importing relation if for any $i \neq j$, $P_j \in \mathcal{I}(P_i) \rightarrow P_i \notin \mathcal{I}(P_j)$, otherwise it has cyclic importing relation.

For example, an ontology $O$ with acyclic importing has two packages:
P<sub>Animal</sub>

(1a) 1 : Dog ⊑ 1 : Carnivore
(1b) 1 : Carnivore ⊑ 1 : Animal
(1c) 1 : Carnivore ⊑ ∀1 : eats.(1 : Animal)
(1d) 1 : Human ⊑ 1 : Animal

P<sub>Pet</sub>

(2a) 2 : Pet ⊑ 1 : Animal
(2b) 2 : PetDog ⊑ 1 : Dog ∩ 2 : Pet
(2c) 2 : PetDog ⊑ 2 : livesWith.(1 : Human)

By reusing terms defined in P<sub>Animal</sub>, the ontology is able to model both inter-module concept subsumption (e.g. axiom 2a) and role relations (e.g. axiom 2c). We denote the package extension to Description Logics (DL) as P. For example, ALC is the package-based version of DL ALC. In what follows, we will examine a restricted type of package extension which only allows import of concept names, denoted as P<sub>C</sub>. For example, the importing P<sub>Animal</sub> ⊑ (eats Pet) P<sub>Pet</sub> is not allowed in ALCPC.

For a package-based ontology ⟨{P<sub>i</sub>}, {P<sub>i</sub> → P<sub>j</sub>}:j≠i⟩, a distributed model is M = ⟨{IP<sub>i</sub>}, {rij}:i≠j⟩, where IP<sub>i</sub> = Δ<sub>i</sub> is the local model of package P<sub>i</sub>. r<sub>ij</sub> ⊆ Δ<sub>i</sub> × Δ<sub>j</sub> is the interpretation for the image domain relation P<sub>i</sub> → P<sub>j</sub>. For any image domain relation P<sub>i</sub> → P<sub>j</sub>, (x, y) ∈ r<sub>ij</sub> indicates an individual y ∈ Δ<sub>j</sub> is an “image” (or copy) of an individual x ∈ Δ<sub>i</sub>. Therefore, local models of P-DL can be partially overlapping, thus offer a tradeoff between the strong module disjointness assumption of DDL and E-connections, and complete overlapping of modules required by the OWL importing mechanics.

To ensure module transitive reusability and reasoning correctness, we require that every image domain relation has the following properties:

- It is one-to-one: for any x ∈ Δ<sub>i</sub>, there is at most one y ∈ Δ<sub>j</sub>, such that (x, y) ∈ r<sub>ij</sub>

- It is compositional consistent: r<sub>ij</sub> = r<sub>ik</sub> ∘ r<sub>jk</sub>, where ◦ denotes function composition. Therefore, semantic relations between terms in i and terms in k can be inferred even if k doesn’t directly import terms from i.

For a relation r<sub>ij</sub> and any individual d ∈ Δ<sub>i</sub>, r<sub>ij</sub>(d) denotes the set {d′ ∈ Δ<sub>j</sub> : (d, d′) ∈ r}. For a subset D ⊆ Δ<sub>i</sub>, r<sub>ij</sub>(D) denotes ∪<sub>d∈D</sub>r<sub>ij</sub>(d), is the image set of D.

A concept i : C is satisfiable w.r.t. a P-DL O = ⟨{P<sub>i</sub>}, {P<sub>i</sub> → P<sub>j</sub>}:i≠j⟩ if there exists a distributed model of O such that C<sub>i</sub> ̸= Ø. O entails subsumption i : C ⊑ j : D (i may or may not be the same as j), denoted as O |= i : C ⊑ j : D if r<sub>ij</sub>(C<sub>i</sub>) ⊆ D<sub>j</sub> holds in every model of O.

3 Distributed Reasoning for P-DL

We extend the tableau-based approach to distributed reasoning with P-DL modules introduced in [3] to a more general setting wherein arbitrary importing (e.g. cyclic or mutual importing) among packages is allowed, and the tableau search process is performed in parallel, asynchronous fashion. We demonstrate the strategy with the package-extended version of a representative DL ALC that allows importing of concepts between packages, i.e. ALCPC.

3.1 ALC Reasoning

We first briefly introduce the tableau algorithm for traditional DLs, e.g. ALC. A tableau is a representation of a model of a logic language, and in particular, of an ontology. Popular representation forms of a tableau include ABox and Completion Graph [2], while each of them can be transformed into the other. In this paper, we adopt the ABox representation since it is more explicit for incremental tableau storage needed for our algorithm.

An ABox contains a set of facts in the form of C(x), P(x, y), x = y or x ̸= y, where x, y are individuals, C is a concept name, and P is a property name. To test the satisfiability of a concept C w.r.t. a TBox T, an initial ABox A<sub>0</sub> is created as (C ∩ C<sub>T</sub>(x₀)), where C<sub>T</sub> is the internalization concept of T: C<sub>T</sub> = ⊓<sub>(C₁∈D₁):∈T</sub> (¬C<sub>i</sub> ∪ D<sub>i</sub>). Each individual x in any ABox of T will be an instance of C<sub>T</sub>.

New facts can be inferred from existing facts based on tableau expansion rules and added to the ABox. Assuming that all concepts are in Negation Normal Form (NNF), the ALC tableau expansion rules for traditional reasoning process (i.e. on a single ontology) are:

- ⊓-rule: if ABox A contains (C₁ ∩ C₂)(x) but not both C₁(x) and C₂(x), then A′ = A ∪ {C₁(x), C₂(x)}
- ⊔-rule: if ABox A contains (C₁ ∪ C₂)(x) but neither C₁(x) or C₂(x), then A₁ = A ∪ {C₁(x)}, A₂ = A ∪ {C₂(x)}
- ∃-rule: if A contains (∃R.C)(x) but no individual y such that C(y) and R(y, z) in A, then A′ = A ∪ {C(y), R(x, y)} where y is an individual name not occurring in original A.
- ∀-rule: if A contains (∀R.C)(x), R(x, y) but no C(y), then A′ = A ∪ {C(y)}

An ABox clash corresponds to the scenario: {C(x), ¬C(x)} ⊆ A (for any individual x and any concept name C). An ABox is consistent if it contains no clash, and is complete if no expansion rule can be applied on it. The given expansion rules for ALC is consistency.
preserving, i.e., if an ABox \( \mathcal{A} \) is consistent, then it has at least one descendent ABox \( \mathcal{A}' \) that derived from it by applying the inference rules on \( \mathcal{A} \) that is also consistent [1].

Note that the \( \cup \)-rule is nondeterministic in that it generates multiple possible new facts. The algorithm needs to try different choices i.e., search for different possible models. Once a chosen path leads to an inconsistency, the algorithm needs to backtrack to the ABox state before the choice, and try other remaining choices.

A concept \( C \) is said to be satisfiable w.r.t. a TBox \( \mathcal{T} \) if and only if the algorithm finds a consistent and complete ABox for both \( C \) and \( C_T \).

### 3.2 Incremental Distributed Tableau Storage

Tableau-based reasoning for modular ontologies [11, 10, 8, 7, 3] usually exploits multiple local tableaux instead of a single tableau. This supports the localized semantics requirement for modular ontologies [4], i.e., that there is no required global model. Thus, reasoning is carried out to obtain a set of connected local models for a modular ontology.

Each of the existing approaches assumes different properties of local models, and as a consequence, requires different procedures for constructing such local models and local tableaux. DDL and \( E \)-connections reasoning algorithms [11, 10, 8, 7] assume domains (the set of individuals) of local tableaux are disjoint, while P-DL reasoning algorithm [3] allows them to be partially overlapping. Advantages of the later approach include support for inter-module subsumption and transitive reusability of modules [3].

In this paper, we introduce the incremental storage for ABoxes to simplify the description of the algorithm. Note that a completion graph-based description of the algorithm is easy to obtain since there is no fundamental difference between the two representations.

We represent an ABox by a series of nodes, where each node contains one or more facts. The root node contains all the initial facts in the ABox. By applying the tableau expansion rules, starting with the root node, we can successively generate new inferred facts. The inferred facts are added as to the successor of the current node, called its expansion successor. The edge linking a node to its successor is called an expansion edge. Multiple choices for expansion (e.g. using the \( \cup \)-rule), result in multiple successors. Recursive application of the tableau expansion rules yields an ABox Tree, with each node in the tree representing an ABox that contains all the facts on the path to that node from the root node. When the algorithm terminates, each leaf in the tree corresponds to either an inconsistent ABox or a complete and consistent ABox (See Figure 1 for an example).

For each node \( n \), let \( \Delta(n) \) denoting the tree that \( n \) belongs to; \( f(n) \) is the set of facts in \( n \); and \( \mathcal{A}(n) \) is the ABox containing all the facts in \( n \) and all of its ancestors up to the root node. Thus, if a node \( n \) is a successor of node \( m \), \( \mathcal{A}(n) = \mathcal{A}(m) \cup f(n) \). Every package \( P_i \) participating in the reasoning process has associated with it, an initial node \( n_0^i \) containing a special fact \( C_T(.) \), the internalization concept of the TBox of \( P_i \). That is, each individual introduced into \( n_0^i \) or any of its local descendants must be an instance of \( C_T(.) \).

Given a P-DL ontology \( O = \{P_i\} \), we can obtain an ABox forest wherein each package has associated with it exactly one ABox tree. A distributed ABox \( \mathcal{A}_{d} \) (i.e., a distributed model) of \( O \) is represented by a set of complete and consistent leaf nodes \( \{n_i\} \), one from each ABox tree, where \( \mathcal{A}(n_i) \) is a local ABox (i.e. a local model), and \( \mathcal{A}_{d} = \bigcup_i \mathcal{A}(n_i) \).

Thus, each ABox tree is maintained by the corresponding local reasoner. The reasoning process is undertaken by a federation of such local reasoners. Since each ABox tree is only locally internalized, integration of the ontology modules into a centralized ontology or of local models into a centralized model is strictly avoided.

### 3.3 Distributed Tableau Expansion

To construct a distributed model for an \( \mathcal{ALCP}_C \) ontology, we start with a list of initial ABox nodes corresponding to each package in the ontology. New facts can be added to the ABox forest by applying tableau expansion rules similar to that of \( \mathcal{ALC} \). However, the traditional \( \mathcal{ALC} \) expansion procedure needs to be modified in several important aspects. We refer to the resulting expansion rules as \( \mathcal{ALCP}_C \) expansion rules.

First of all, new facts should be sent to a “destination” ABox tree to reduce the cost of detecting a clash. Since a concept can be imported into another package, it is possible that a fact \( C(x) \) is generated from an expansion in an ABox of a package that is not \( C \)'s home package. Therefore, \( C(x) \) and \( \neg C(x) \) can be generated in different local ABox trees in which case, a clash cannot be locally detected. However,
global check for such clashes is expensive. Hence, we adopt a strategy that is designed to minimize the cost of detecting clashes. We start by introducing some relevant definitions:

**Definition 4 (Concept Destination)** An atomic concept $C$ or its negation $\neg C$’s destination is $C$’s home package $\mathcal{HP}(C)$. A complex concept $C$’s destination is the tree in which it is generated. Destination of $C$ is denoted as $\delta(C)$.

Each generated fact $C(x)$ from any ABox tree node will be sent to an ABox tree of the destination of $C$, i.e., $\delta(C)$. The destination ABox tree of a fact $f$ is denoted by $\delta(f)$. Thus, all clashes can be detected locally. Note that there is no role importing in $\mathcal{LCCP}_C$, therefore a role fact $P(x,y)$ is always generated in (and stays in) the ABox tree of $P$’s home package.

We call a fact that is sent from one ABox tree to another a fact message, and we add a message edge from the sending node to the receiving node. In such cases, two copies of the fact are kept in the two nodes. For example, in Figure 4 at Time 4, $B_1(x)$ is generated in the ABox tree $T_A$ (of package $A$), but $B_1$ has home package $B$. Therefore, $B_1(x)$ is sent to the ABox tree $T_B$ of $B$, and finally results in a local clash that is detected locally in $T_B$.

Since a fact (e.g., $C(x)$) may be shared by two ABox trees (e.g. $T_i$, $T_j$), an individual name (e.g., $x$) may also appear in the two trees. We denote such a shared individual name in different ABox trees with prefixes such as $i:x$ and $j:x$. However, we assume those names can still be identified as variances for the same individual.

The termination of the algorithm can still be ensured using the subset blocking [2]: for an ABox tree node $n$, the application of the $\exists$-rule is blocked to an individual $x$ by an individual $y$ iff $\{D|D(x) \in A(n)\} \subseteq \{D'|D'(y) \in A(n)\}$.

Note that the algorithm we presented so far is equivalent to the completion graph-based $\mathcal{LCCP}_C$ reasoning algorithm in [3] when importing between packages is restricted to be acyclic. An ABox-based version of the Lemmas 1 and 2 of [3] can be obtained as: if a fact is sent from an ABox tree of $P_1$ to an ABox tree of $P_2$, $P_1$ must directly or indirectly reuse $P_2$, i.e., $P_2$ is in $P_1$’s importing closure. Therefore, if there is only acyclic importing among packages, there are only uni-directional message edges between ABox trees: once an ABox tree $t_1$ receives a fact from $t_2$, there is no path on the ABox forest (linked by message edges) from a node in $t_1$ ending in a node in $t_2$. Therefore, there is no risk of message loop for the acyclic concept importing case. However, in order to guarantee termination of the algorithm, we still need to find a way to prevent message looping in the presence of cyclic importing.

### 3.4 Handling Cyclic Importing

The distributed tableau expansion algorithm follows the “divide and conquer” strategy, i.e., the reasoning task is divided into smaller subtasks each undertaken by a local reasoner. If a fact $C(x)$ is sent from an ABox tree $T$ to another ABox tree branch of a package $p$, $T$ in fact submits a satisfiability query of $C$ to $p$ under the presence of other existing facts on the branch. The final result of reasoning in $T$ depends on the answer from $p$’s ABox tree. This strategy is guaranteed to terminate with acyclic importing since the messages between ABox trees are unidirectional. However, cyclic importing presents further difficulties in message exchange among ABox trees because it may lead to ABox trees waiting for each other in a cycle or a deadlock. How can we avoid such a deadlock?

To develop some intuition regarding this problem, we consider the logical meaning of edges in the ABox forest. If a fact $f$ is generated by applying expansion rules at a node $n$, $f$ is actually the logical consequence of some facts in the ABox $A(n)$. For example, in Figure 1, the fact $D(x)$ is one possible logical consequence of $(C \sqcup D)(x)$. Therefore, if a new fact $f$ that is a (direct or indirect) logical consequence of $A(n)$ is to be added on the ABox tree, it should be added as a child of node $n$. For example, in Figure 4 Time 5, a fact $A_2(x)$ is generated in the ABox tree $T_B$ while the destination of $A_2$ is $T_A$. However, since an ancestor of $A_2(x)$ has received a fact $B_1(x)$ from $T_A$, $A_2(x)$ is an indirect logical consequence of $B_1(x)$. Hence, $A_2(x)$ should be added to $T_A$ under the node containing $B_1(x)$. We refer to an ABox graph containing both expansion edges and message edges as an ABox graph.

Another intuition is that an ABox graph is a representation of global tableau, while each branch in a local ABox tree stands for a search choice in finding such a global tableau. Thus, when adding new facts to the graph, the distinction between the different search choices must be maintained. In other words, different reasoning subtasks should be kept separate.

The preceding considerations suggest the following strategy for avoiding message looping or deadlock in the presence of cyclic importing:

- Let each node $n$ maintain a contact list $lst(n)$ of nodes from other ABox trees.
- Initial contact list of a root node is initialized with the list of root nodes of other ABox trees.
- If a new node $n$ is generated under a node $m$ in the same ABox tree $\Delta(m)$, $lst(n) \leftarrow lst(m)$.
- If a node $n$ in an ABox tree $T$ generates a new fact $f$ such that $\delta(f) \neq T$, $f$ is sent to a new node $l$ under $m$ on the destination tree $\delta(f)$, where $m \in \text{lst}(n)$. $\text{lst}(l)$
is obtained by merging \(lst(n)\) and \(lst(m)\): if both \(n\) and \(m\) contain contacts from an ABox tree, discard one if it is an ancestor of the other on the ABox graph.

This strategy ensures that a node always has at most one contact node from each of the other ABox trees:

**Lemma 1** For a node \(n\) in an ABox tree \(T\), for any package \(p, p \neq \Delta(n)\), there is at most one node \(m\) in the ABox tree of \(p\) such that \(m\) is in \(n\)’s contact list: \(|\{m | m \in lst(n), \Delta(m) = p\}| \leq 1.\)

Proof sketch: If there are two nodes \(m_1\) and \(m_2\) from one ABox tree of \(p\) that are both in \(lst(n)\), there must be two paths from \(m_1\) to \(n_1\) and from \(m_2\) to \(n_2\), where \(n_1\) and \(n_2\) are on the path from \(n\) to its root. Without loss of generality, suppose \(n_2\) is a descendant of \(n_1\); there must be an ancestor of \(n_2\), say \(n_3\), such that \(n_3 \in lst(m_2)\). Then \(n_1 = n_3\) or \(n_2\) is a descendant of \(n_3\) and there must be a path from \(n_3\) through some ancestor of \(m_2\) to \(m_2\); however, such nodes must be \(m_1\)'s descendants and hence \(m_2\) must be a descendant of \(m_1\). According to our contact merging rule, only \(m_2\) will be kept in \(n\)’s contact list.

The contact list update rule described above ensures that the contact list contains only the most “recent” message sender from ABox trees of other packages. We also denote \(lst_i(n)\) as the contact of \(n\) on the ABox tree \(i\).

Immediately from Lemma 1 we have:

**Lemma 2** If a node has two ancestors \(n, m\) in an ABox graph, it must be the case that \(n\) is an ancestor of \(m\) on the ABox graph, or \(m\) is an ancestor of \(n\) on the ABox graph (not both).

This lemma implies that when adding new facts to the graph, the distinction between the different search choices (tree branches) must be maintained across all local reasoners. For example, the Figure 4 Time 6, \(B_2(x)\) in \(T_B\) has two ancestors \(B_1(x)\) and \(B_2(x)\) in \(T_A\), and \(B_1(x)\) is a local ancestor of \(B_2(x)\). Therefore, the set of all ancestor nodes of a node \(n\) on the ABox graph contains facts associated with a single search branch.

Thus, there is effectively, a virtual global ABox (directed) graph that corresponds to a conceptually integrated ontology. This graph can be decomposed into multiple smaller local ABox trees (by copying some nodes as needed). There must be no (directed) loop on the ABox graph, thereby ensuring the termination of the algorithm:

**Lemma 3 (Termination)** Let \(C_0\) be an \(\mathcal{ALCP}_C\) -concept description in NNF. There cannot be an infinite sequence of \(\mathcal{ALCP}_C\) rule applications.

We summarize the expansion rules for \(\mathcal{ALCP}_C\) in what follows, starting with some notation: for any node \(n\) on an ABox tree \(k\), \(A_k(n)\) is the ABox represented by \(n\); for any fact \(f, m(n, f)\) is a query from \(n\) for \(f\)’s existence in its destination, i.e., if \(f \in A_{\delta(f)}(\text{lst}_{\delta(f)}(n))\); \(r(n, f)\) is an action that sends a fact \(f\) to its destination, i.e., creates a new node containing \(f\) under \(\text{lst}_{\delta(f)}(n)\). When \(\delta(f) = n, m(n, f)\) is reduced to a local query that if \(f \in A_k(n)\), and \(r(n, f)\) is reduced to a local action that adds a new node containing \(f\) under \(n\). The \(\mathcal{ALCP}_C\) expansion rules are:

- \(\nabla\)-rule: if \(A_k(n)\) contains fact \((C_1 \cap C_2)(x)\), \(x\) is not blocked in \(A_k(n)\), then do \(r(n, C_i(x))\) if \(m(n, C_i(x))\) = false, for \(i = 1, 2\)

- \(\sqcup\)-rule: if \(A_k\) contains fact \((C_1 \cup C_2)(x)\), \(x\) is not blocked in \(A_k(n)\), but \(m(n, C_1(x)) \lor m(n, C_2(x))\) = false, then do \(r(n, C_1(x))\) or \(r(n, C_2(x))\)

- \(\exists\)-rule: if \(A_k\) contains fact \((\exists R.C)(x)\), \(x\) is not blocked in \(A_k(n)\), and \(R \in P_k\), for any \(R(x, z) \in A_k\), we have \(m(n, C(z))\) = false, then do \(r(n, R(x, y))\) and \(r(n, C(y))\) where \(y\) is a new individual name.

- \(\forall\)-rule: if \(A_k\) contains \((\forall R.C)(x), R(x, y)\), \(x\) is not blocked in \(A_k(n)\), \(R \in P_k\), and \(m(n, C(y))\) = false, then do \(r(n, C(y))\).

### 3.5 Asynchronous Federated Reasoning

We need several types of messages to complete our description of the search for a complete and consistent global tableau, in which the targets of all messages are all \(n\)’s contacts and \(n\)’s parent node on the local ABox tree \((\Delta(n)):\)

- If a local clash is found in \(A(n)\), mark \(n\) as \(\perp\), send clash messages.

- If all expansion successors of \(n\) are marked as \(\perp\), or any of the message successors is marked as \(\perp\), mark \(n\) as \(\perp\); send clash messages.

- If \(A(n)\) is locally complete, mark \(n\) as \(\top\), send model messages.

- If any expansion successors of \(n\) are marked as \(\top\), and all message successors is marked as \(\top\), mark \(n\) as \(\top\), send model messages.

The clash and model messages are explained in Figure 2. In the case of centralized tableau reasoning, given three nodes \(x, y, z\), where \(x\) is the expansion parent of \(y\) and \(z\), it must be the case that \(A(x)\) has a clash iff \(A(y)\) has a clash and \(A(z)\) has clash. However, this is not necessarily true in the distributed setting. For example (See Figure 2(a)), even when all of \(x\)’s successors are inconsistent, if all message recipients of \(x\) find no clash, it is possible for them to send
back to $x$ another fact and open a new branch under $x$ (also see the example in Figure 4 Time 7, node $B_1(x)$ in $T_A$, its only successor $A_2(x)$ contains a clash, while on Time 8 a new branch $A_3(x)$ is created by a message from $T_B$). On the other hand, unlike in the centralized setting where $A(x)$ is consistent iff $A(y)$ is consistent or $A(z)$ is consistent, in the distributed setting (See Fig 2(b), remote inconsistencies might be discovered on $x$’s copies in other ABox trees although $x$ is locally consistent.

Once a node sends a fact message to one or more ABox trees, we do not require that the node wait until an answer is received from other ABox trees. A search branch may be closed on the basis of a local clash or clash messages received from other ABox trees. Therefore, local reasoners for each of the ABox trees may work on different reasoning subtasks concurrently to make the best use of the computational resources available to each of them. Thus, the algorithm presented here also relaxes the waiting (after sending a fact) strategy adopted in [3], to be preformed in an asynchronous fashion.

Figure 4 (Time 1 -Time 7) illustrates the working of the algorithm presented in this paper. The example ontology contains two packages $A$ and $B$. Package $A$ contains axioms $A \sqsubseteq A_1 \sqcap \neg A_3, A_1 \sqsubseteq B_1, A_2 \sqsubseteq B_2$; package $B$ contains axioms $B_1 \sqsubseteq A_2 \sqcup A_3, B_2 \sqsubseteq A_3$. Thus, the two packages mutually import each other. The reasoning task is to check the consistency of the ontology. The result is that the ontology is inconsistent.

4 Complexity

It is easy to show that the worst case time complexity in the proposed distributed reasoner is no worse than that of a centralized reasoner on $ALC$, i.e. Exptime for consistency of $ALC$-ABoxes [2]. It can be shown by a conceptual reduction from $ALCP_C$ expansions to $ALC$ expansions: consider if all local ABox trees are generated and maintained at a single location, the given $ALCP_C$ expansion rules will be reduced to $ALC$ expansion rules. Thus, the total number of tableau expansions required to find an $ALCP_C$ model in all local reasoners is same as the number of tableau expansions required to find an $ALC$ model for an integrated ontology of all ontology modules.

However, since different reasoners may concurrently explore different tableau search choices in $ALCP_C$ expansions, in practice, the time requirements of the proposed algorithm can be much less than the time used by a centralized reasoner.

The space complexity for the algorithm may become a concern when the individual reasoners operate in an asynchronous fashion (i.e., the individual reasoners do not wait for responses to messages that they have sent to other reasoners before proceeding). In this mode of operation, the local reasoners depart from the strictly depth-first search (with PSPACE complexity) [1]. While this allows each reasoner to make the best use of its computational resources, the departure from strictly depth-first search can lead to a worst case space complexity that is EXPSPACE (same as that of $ALC$ ). Alternatively, each local reasoner can implement a mixed strategy of switching to the synchronous (i.e. waiting for responses before proceeding) mode when available memory is limited, thus requires only polynomial space.

In practice, the proposed $ALCP_C$ reasoning algorithm may be more memory efficient than a centralized reasoner, since each package is only locally internalized. Therefore, once a new individual $x$ is introduced in an ABox tree for package $i$, only $C_T(x)$ is added to the ABox tree, while in the centralized case a more complex fact $c_T(x)$ is added, where $c_T$ is the internalization concept for the combined TBox of all packages.

In conclusion, by localizing reasoning subtasks in multiple peer reasoners, the time and space required by each local reasoner can be much less than that required by a centralized reasoner working on an integrated version of the modular ontology. Thus, the distributed reasoning algorithm will allow reasoning with very large ontology that cannot be handled in a single computer.

5 Soundness and Completeness

The proposed algorithm reduces the problem of checking inconsistency of a model to a combination of checking inconsistency of local ABoxes. Given ABoxes $A, B, \ldots$, we denote by $(A,B, \ldots)$ the ABox obtained by merging the respective ABoxes, with shared facts and shared individual names are merged.

The $ALCP_C$ expansion rules expand one or more existing ABoxes as shown in the Figure 3. The first type of
expansion, called augmenting, infers new facts from existing facts in an ABox and adds them to the same ABox i.e. \( A' = A \cup \{ \alpha \} \). The second type of expansion, called searching, adds multiple possible inferred facts to the original ABox i.e. \( A_1 = A \cup \{ \alpha \} \). The third type of expansion, called reporting, adds the inferred new facts into another ABox, \( A \cup \{ \beta \} \). The first two types are also present in traditional tableau expansions for DLs, whereas reporting, i.e. adding the inferred new facts into another ABox, \( B' = B \cup \{ \alpha \} \). The first two types are also present in traditional tableau expansions for DLs, whereas reporting is unique to P-DL expansions [3].

**Lemma 4** If an ABox \( A \) is expanded by the \( \text{ALCP}_{C} \) tableau expansion rules using

1. augmenting: from \( A \) to \( A' \), then \( A \) is consistent iff \( A' \) is consistent;
2. searching: from \( A \) to \( A_1 \) or \( A_2 \), then \( A \) is consistent iff \( A_1 \) is consistent or \( A_2 \) is consistent;
3. reporting: from \( A, B \) to \( A, B' \), then \( (A, B) \) is consistent iff \( (A', B') \) is consistent;

**Lemma 5** If a set of locally complete ABoxes \( \{ A_i \} \), \( i = 1, ..., m \) is generated by the \( \text{ALCP}_{C} \) tableau expansion rules, \( (A_1, ..., A_m) \) is consistent iff \( \forall i, A_i \) is consistent.

The above lemma follows from the observation that \( \text{ALCP}_{C} \) expansions will send any concept fact to the ABox of its destination package, and inconsistency is detected when both \( C(x) \) and \( \neg C(x) \) appear in some local ABox. Global inconsistency necessarily must result in local consistency in some ABox; and a locally inconsistent ABox implies that the set of ABoxes taken together must also be inconsistent. Thus we have:

**Lemma 6 (Soundness)** Assume that \( S_0 \) is obtained from the finite set of ABoxes \( S \) by application of an \( \text{ALCP}_{C} \) transformation rule. Then \( S \) is consistent iff \( S_0 \) is consistent.

Completeness of the algorithm follows from the observation that a P-DL model can be induced by a set of distributed ABoxes. Given a set of complete and consistent ABoxes \( \{ A_i \}, i = 1, ..., m \) for a set of packages \( \{ P_i \} \) generated by \( \text{ALCP}_{C} \) expansion rules, we can easily build a distributed P-DL model as follows:

- For each ABox \( A_i \), the domain of local individuals \( \Delta^I_i \) is the set of all individuals occurring in it. The domain of all individuals in the distributed interpretation \( \Delta^I = \bigcup_{i=1}^{m} \Delta^I_i \).
- For each concept name \( C \in P_i \), \( C^{T_i} = \{ x | C(x) \in A_i \} \).
- For each role name \( R \in P_i \), \( R^{T_i} = \{ (x,y) | R(x,y) \in A_i \} \).
- If there is a fact message \( C(x) \) sent from \( A_i \) to \( A_j \), add a pair \( (i : x, j : x) \) to the image domain relation \( r_{ij} \).

Note that the one individual name may appear in different local domains, thus those local domains may be partially overlapping. Thus we have:

**Lemma 7 (Completeness)** Any complete and clash-free \( \text{ALCP}_{C} \) Global ABox \( \{ A_i \} \) has a model.

### 6 Summary and Discussion

We have presented a distributed tableau-based reasoning algorithm for the package-based extension of the DL language \( \text{ALCP}_{C} \). The proposed approach offers a practical approach that:

1. By strictly avoiding integration of modules into a single ontology, the need for transferring the entire contents of ontology modules to a central location is avoided.
2. Using a message-based inter-reasoner communication strategy, the algorithm allows arbitrary reuse among the ontology modules, such as the presence of mutual or cyclic importing among packages. Thus, it overcomes one of the important limitations of the approach presented earlier in [3] which allowed only acyclic importing among packages.
3. The reasoning process can be carried out in an asynchronous, peer-to-peer fashion, such that different local reasoners can concurrently work on different reasoning subtasks to improve the efficiency and scalability of reasoning.

4. Based on the P-DL formalism, the algorithm is able to tackle a broader range of reasoning tasks, including those involving both inter-module subsumption and inter-module role relations.

Work in progress is aimed at:

- Extending the proposed reasoning algorithm to work with more expressive DLs such as $SHIOP_C$ (i.e. package-based $SHIQ$ with concept importing).
- Extending the proposed approach to work with a globally inconsistent set of ontology modules. DDL [10] have presented an approach to prevent the propagation of local inconsistencies using a special “Hole” semantics. However, in general, it is possible for an ontology to be globally inconsistent although each local module is consistent. Furthermore, applying the hole semantics requires the identification of inconsistent modules before initiating the reasoning process, which may not always be possible. In this context, package-based variants of defeasible extensions to modular ontology languages are of interest to prevent both local and global inconsistencies.
- Detailed performance evaluation of implementations of the proposed algorithm in several practical application scenarios.

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References

Figure 4. Reasoning with Mutual Importing