Integrity Constraints in OWL

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Abstract

In many data-centric semantic web applications, it is desirable to use OWL to encode the Integrity Constraints (IC) that must be satisfied by instance data. However, challenges arise due to the Open World Assumption (OWA) and the lack of a Unique Name Assumption (UNA) in OWL’s standard semantics. In particular, conditions that trigger constraint violations in systems using the Closed World Assumption (CWA), will generate new inferences in standard OWL-based reasoning applications. In this paper, we present an alternative IC semantics for OWL that allows applications to work with the CWA and the weak UNA. Ontology modelers can choose which OWL axioms to be interpreted with our IC semantics. Thus application developers are able to combine open world reasoning with closed world constraint validation in a flexible way. We also show that IC validation can be reduced to query answering under certain conditions. Finally, we describe our prototype implementation based on the OWL reasoner Pellet.

Introduction

The Web Ontology Language (OWL) (Smith, Welty, and McGuiness 2004; Motik, Patel-Schneider, and Grau 2009) is an expressive ontology language based on Description Logics (DL) (Baader et al. 2003)\footnote{Throughout the paper we use the terms OWL, OWL 2, and DL interchangeably.} with sound and complete reasoning algorithms. The semantics of OWL addresses distributed knowledge representation scenarios where complete knowledge about the domain cannot be assumed. Further, the semantics has the following characteristics:\footnote{Note that these characteristics do not just pertain to OWL.}

- the presence of the OWA; i.e., a statement cannot be inferred to be false on the basis of failures to prove it.
- the absence of the UNA: i.e., two different names may refer to the same object.

However, these characteristics can make it difficult to use OWL for data validation purposes in real-world applications where complete knowledge can be assumed for some or all parts of the domain, as shown by the following examples.

Example 1. Suppose we have a KB $K$ containing information about products in a company’s inventory as follows:

$K = \{ \text{Product} (p) \}$

One might add the following axiom to express the constraint that “every product is produced by a producer”:

$\alpha : \text{Product} \sqsubseteq \exists\text{hasProducer}.\text{Producer}$

In this example, due to the OWA, not having a known producer for $p$ does not cause a logical inconsistency. Therefore, we cannot use $\alpha$ to detect (or prevent) that a product is added to the KB without the producer information.

Example 2. Suppose the inventory KB $K$ looks like this:

$K = \{ \text{Product} (p), \text{hasProducer} (p, m_1), \text{hasProducer} (p, m_2) \}$

One might add the following axiom to express the constraint “a product has at most one producer”:

$\alpha : \text{Product} \sqsubseteq \leq 1\text{hasProducer}$.\top$

Since $m_1$ and $m_2$ are not explicitly defined to be different, they will be inferred to be same due to the cardinality restriction. However, in many cases, the intention to use the above cardinality restriction is not to draw this inference, but to detect an inconsistency. When the information about instances are coming from multiple sources we cannot always assume that explicit inequalities will be present.

Besides the examples above, we have also identified several other requirements for ICs from the OWL community through a user survey. We observed a need to use OWL as an Integrity Constraint (IC) language using closed world semantics. That is, we observed a desire to adopt the OWA without the UNA for parts of the domain where incomplete knowledge exists. Simultaneously, we would like to use the Closed World Assumption (CWA)\footnote{With CWA, a statement is inferred to be false if it is not known to be true, which is the opposite of OWA.} with the UNA otherwise. This calls for the ability to combine the open world reasoning of OWL with closed world constraint validation.

In this paper, we demonstrate how to extend OWL with ICs. First, we describe an alternative IC semantics for OWL, which enables developers to augment OWL ontologies with IC axioms. Standard OWL axioms in the ontologies are used to compute inferences with the open world semantics and ICs are used to validate instance data with the closed

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world semantics. Our goal is to enable efficient data validation with OWL, especially in settings where OWL KBs are integrated with relational databases and ICs are needed to enforce the named individuals have some known values.

Then, we show that IC validation can be reduced to query answering when the KB expressivity is within $\text{SROIQ}$ or the constraint expressivity is within $\text{SROIL}$. The queries generated from ICs can be expressed with SPARQL, allowing existing OWL reasoners to be used for IC validation easily.

### Related Work

Research on integrating ICs with OWL has been conducted in multiple directions. One approach to achieve this combination is to couple OWL with rule-based formalisms and to express ICs as rules without heads (e.g., (Eiter et al. 2008; Motik 2007)). With this approach, ontology developers have to deal with both the rule language and the ontology language (OWL) to model the domain.

ICs can also be expressed using the epistemic query language EQL-Lite (Calvanese et al. 2007) where one can pose epistemic FOL queries against standard FOL KBs. Although the data complexity of answering EQL-Lite queries in DL-Lite is LOGSPACE, it would require substantially more effort to support EQL-Lite in DL KBs with full expressivity and the complexity results are still unknown.

Another approach is the epistemic extension of DLs (Donini et al. 1998; Donini, Nardi, and Rosati 2002) where ICs are represented as epistemic DL axioms and the satisfaction of ICs is defined as epistemic axiom entailments. However, this approach adopts strict UNA which is not compatible with OWL because standard OWL axioms can be used to infer that two different names identify the same individual. While existing research has focused on epistemic extensions for the relatively inexpressive $\text{ALC}$, it is still not clear how to combine epistemic logics with more expressive DLs.

The closest related work to ours is the approach that reuses OWL as an IC language. One notable effort work along this line is a proposal by Motik et al. (Motik, Horrocks, and Sattler 2007) which is based on a minimal Herbrand model semantics of OWL. Given an OWL KB $\mathcal{K}$ consisting of a TBox and an ABox, and a constraint TBox, an axiom in the constraint TBox is satisfied by $\mathcal{K}$ if all minimal Herbrand models satisfy it. However, as shown in the following cases, this approach may result in counterintuitive results or a significant modeling burden.

First, unnamed individuals can satisfy constraints, which is not desirable for closed world data validation.

#### Example 3. Consider a KB $\mathcal{K}$ that contains a product instance $p$ and its unknown producer, and a constraint $\alpha$ that every product should have a known producer:

$$\mathcal{K} = \{ A \equiv \exists \text{hasProducer}. \text{Producer}, \text{Product}(p), A(p) \}$$

$$\alpha : \text{Product} \sqsubseteq A$$

Since $p$ has a producer in every minimal Herbrand model of $\mathcal{K}$, $\alpha$ is satisfied, even though the producer is unknown.

Second, if a constraint needs to be satisfied only by named individuals, then a special concept $O$ has to be added into the original IC axiom, and every named individual should be asserted as an instance of $O$. This adds a significant maintenance burden on ontology developers, in particular when modification to the original TBox or ABox is not allowed or too costly. Further, the next example shows that the intuition behind the constraint may not be correctly captured.

#### Example 4. Suppose we have a KB $\mathcal{K}$ where there are two possible producers for a product $p$ and a constraint $\alpha$:

$$\mathcal{K} = \{ B \equiv \exists \text{hasProducer}. \{ m_1, m_2 \}, \text{Product}(p), B(p), \text{Producer}(m_1), \text{Producer}(m_2), O(p), O(m_1), O(m_2) \}$$

$$\alpha : \text{Product} \sqsubseteq \exists \text{hasProducer}. (\text{Producer} \sqcap O)$$

The intuition behind the constraint $\alpha$ is that the producer of every product should be known. Even though we do not know the producer of $p$ is $m_1$ or $m_2$, $\alpha$ is still satisfied by the semantics of (Motik, Horrocks, and Sattler 2007) because in every minimal Herbrand model $p$ has a producer that is also an instance of $\text{Producer}$ and $O$.

Third, disjunctions and ICs may interact in unexpected ways.

#### Example 5. Consider the following KB $\mathcal{K}$ where there are two categories for products and a constraint $\alpha$ defined on one of the categories:

$$\mathcal{K} = \{ \text{Product} \sqsubseteq \text{Category}_1 \sqcup \text{Category}_2, \text{Product}(p) \}$$

$$\alpha : \text{Category}_1 \sqsubseteq \exists \text{categoryType}. \top$$

Since we are not sure that $p$ belongs to $\text{Category}_1$, it is reasonable to assume that the constraint will not apply to $p$ and it will not be violated. However, $\alpha$ is violated with the (Motik, Horrocks, and Sattler 2007) semantics because there is a minimal model where $p$ belongs to $\text{Category}_1$ but it does not have a $\text{categoryType}$ value.

In this paper, we present a new IC semantics for OWL that can correctly capture the intended closed world constraint semantics and thus overcomes the above issues.

### Preliminaries

#### Description Logics $\text{SROIQ}$

In this section, we describe the syntax and semantics of the Description Logic $\text{SROIQ}$ (Horrocks, Kutz, and Sattler 2006), which provides the logical foundation of OWL 2.

Let $N_C, N_R, N_I$ be non-empty and pair-wise disjoint sets of atomic concepts, atomic roles and named individuals respectively. The $\text{SROIQ}$ role $R$ is an atomic role or its inverse $R^-$. Concepts are defined inductively as follows:

$$C \leftarrow A \mid C \sqcap \neg C \mid C_1 \sqcap C_2 \mid \geq nR.C \mid \exists R.\text{Self} \mid \{ a \}$$

where $A \in N_C$, $a \in N_I$, $C^{(i)}$ a concept, $R$ a role.

We use the following standard abbreviations for concept descriptions:

- $\bot = C \sqcap \neg C$, $\top = \neg \bot$, $\neg (\neg C \sqcap \neg D) = C \sqcup D$
- $\neg \bot = \exists R.C = \{ a_{1}, \ldots, a_{n} \} = \{ a_{1} \} \sqcup \ldots \sqcup \{ a_{n} \}$
- $\text{A SROIQ-interpretation } I = (\Delta, \tau)$, where $\Delta$ is the domain, and $\tau$ is the interpretation function which maps $A \in N_C$ to a subset of $\Delta$, $R \in N_R$ to a subset of $\Delta \times \Delta$, $a \in N_I$ to an element of $\Delta$. The interpretation can be extended.
to inverse roles and complex concepts as follows:
\[(R^-)^T = \{\langle y, x \rangle \mid \langle x, y \rangle \in R^T\}, (-C)^T = \Delta \setminus C^T,\]
\[(C \sqcap D)^T = C^T \cap D^T,\]
\[\{nR.C)^T = \{x \mid \#\{y \mid \langle x, y \rangle \in R^T, y \in C^T\} \geq n\}\]
\[(\exists R.Self)^T = \{x \mid \langle x, x \rangle \in R^T\}, \{a\}^T = \{a^T\}.\]
where \# denotes the cardinality of a set.

A SROIQ knowledge base \(K\) is a collection of TBox (terminology) and RBox (role) SROIQ axioms which are listed in Table 1, and Abox (assertion) axioms \((C(a), R(a, b), a = b, a \neq b)\) where their semantics is given by encoding them as TBox axioms \(\{a\} \subseteq C, \{a\} \equiv \exists R.\{b\}, \{a\} \subseteq \{b\}, \{a\} \subseteq \neg\{b\}\).

<table>
<thead>
<tr>
<th>Type</th>
<th>Axiom</th>
<th>Condition on (I)</th>
</tr>
</thead>
<tbody>
<tr>
<td>TBox</td>
<td>(R_1 \sqsubseteq R_2)</td>
<td>(R_1^I \sqsubseteq R_2^I)</td>
</tr>
<tr>
<td></td>
<td>(R_1 \ldots R_n \sqsubseteq R)</td>
<td>(R_1^I \ldots R_n^I \sqsubseteq R^I)</td>
</tr>
<tr>
<td></td>
<td>Ref(R)</td>
<td>(\forall x \in I : \langle x, x \rangle \notin R^I)</td>
</tr>
<tr>
<td></td>
<td>Irr(R)</td>
<td>(\forall x \in I : \langle x, x \rangle \notin R^I)</td>
</tr>
<tr>
<td></td>
<td>Dist(R_1, R_2)</td>
<td>(R_1^I \cap R_2^I = \emptyset)</td>
</tr>
</tbody>
</table>

Table 1: Axiom satisfactions in the SROIQ-interpretation

We say that an interpretation \(I\) satisfies a SROIQ axiom \(\alpha\), denoted \(I \models \alpha\) if the corresponding condition on \(I\) in Table 1 is satisfied. \(I\) is a model of \(K\) if it satisfies all the axioms in \(K\). We define \(Mod(K)\) to be the set of all models of \(K\). We say that \(K\) entails \(\alpha\), written as \(K \models \alpha\), if \(I \models \alpha\) for all models \(I \in Mod(K)\).

Distinguished Conjunctive Query (DCQ)

We now describe the syntax and semantics of distinguished conjunctive query (DCQ). Let \(N_V\) be a non-empty set of variable names disjoint from \(N_I, N_C,\) and \(N_R\). A query atom is an ABox axiom where variables can be used in place of individuals. Formally, it is defined as follows:
\[q \leftarrow C(x) \mid R(x, y) \mid \neg R(x, y) \mid x = y \mid x \neq y\]
where \(x, y \in N_I \cup N_V\), \(C\) is a concept, and \(R\) is a role. A conjunctive query (CQ) is a conjunction of query atoms:
\[Q \leftarrow q \mid Q_1 \land Q_2\]

A DCQ is a CQ containing only distinguished variables.4

The semantics of DCQ is given in terms of interpretations defined in the previous subsection. We define an assignment \(\sigma : N_V \rightarrow N_I\) to be a mapping from the variables used in the query to named individuals in the KB. We use \(\sigma(Q)\) to denote the application of an assignment \(\sigma\) to a query \(Q\) such that the variables in the query are replaced with individuals according to the mapping. We say a KB \(K\) entails a DCQ \(Q\) with an assignment \(\sigma\), written as \(K \models^\sigma Q\), if:
\[K \models^\sigma q\quad \text{iff} \quad K \models \sigma(q)\]
\[K \models^\sigma Q_1 \land Q_2\quad \text{iff} \quad K \models^\sigma Q_1 \text{ and } K \models^\sigma Q_2\]

We define the answers to a query, \(A(Q, K)\), to be the set of all assignments for which the KB entails the query. That is, \(A(Q, K) = \{\sigma \mid K \models^\sigma Q\}\). We say that a query is true w.r.t. a KB, denoted \(K \models Q\), if there is at least one answer for the query, and false otherwise.

IC Semantics

There has been a significant amount of research to define the semantics of ICs for relational databases, deductive databases, and knowledge representation systems in general. There are several proposals based on KB consistency or KB entailment. Against these approaches, Reiter (Reiter 1988) argued that ICs are epistemic in nature and are about “what the knowledge base knows”. He proposed that ICs should be epistemic first-order queries that will be asked against a standard KB that does not contain epistemic axioms.

We agree with Reiter’s assessment of the epistemic nature of ICs. We believe this is an appropriate semantics for ICs. We will now describe an alternative IC semantics for OWL axioms, which is similar to how the semantics of epistemic DL ALCK (Donini et al. 1998) and MKNF DL ALCK\langle\forall\exists\rangle (Donini, Nardi, and Rosati 2002) are defined. Then, we discuss how our IC semantics addresses the issues explained in the introduction and the related work sections, and how it enables OWL to be an IC language.

Formalization

We define IC-interpretation as a pair \(I, U\) where \(I\) is a SROIQ interpretation defined over the domain \(\Delta^I\) and \(U\) is a set of SROIQ interpretations. The IC-interpretation function \(I, U\) maps concepts to a subset of \(\Delta\), roles to a subset of \(\Delta \times \Delta\), and individuals to an element of \(\Delta\) as follows:
\[C^I = \{x \mid x \in N_I \text{ s.t. } \forall J \in U, x^J \in C^J\}\]
\[R^I = \{\langle x^I, y^I \rangle \mid x, y \in N_I \text{ s.t. } \forall J \in U, \langle x^J, y^J \rangle \in R^J\}\]
where \(C\) is an atomic concept and \(R\) is a role. According to this definition, \(C^I\) is the interpretation of named individuals that are instances of \(C\) in every (conventional) interpretation from \(U\). \(R^I\) can be understood similarly.

IC-interpretation \(I, U\) is extended to inverse roles and complex concepts as follows:
\[(R^-)^I = \{\langle y^I, x^I \rangle \mid \langle x^I, y^I \rangle \in R^I\}\]
\[(C \sqcap D)^I = C^I \cap D^I, \quad (-C)^I = N_I \setminus C^I,\]
\[(\geq nR.C)^I = \{x^I \mid x \in N_I \text{ s.t. } \#\{y^I \mid \langle x^I, y^I \rangle \in R^I\} \geq n\}\]
\[(\exists R.Self)^I = \{x^I \mid x \in N_I \text{ s.t. } \langle x^I, x^I \rangle \in R^I\}\]
\[\{a\}^I = \{a^I\}\]

We can see that the IC-interpretation \(I, U\) is using the closed-world assumption. For example, the elements of \(C^I\) are the interpretation of named individuals that should be in the interpretation set of \(C^I\) for all \(I \in U\). Any named individual that can not be proven to be an instance of \(C\) is assumed to be an instance of \(\neg C\) since \((-C)^I\) is the complement of \(C^I\) w.r.t. \(N_I\).

Note that, although the IC interpretations have some similarities to the epistemic interpretations of ALCK and
is defined as:

\[ \mathcal{K} \models IC \alpha \iff \forall \mathcal{I}, \mathcal{U}, \mathcal{U} \models \alpha, \text{ where } \mathcal{U} = \text{Mod}_{ME}(\mathcal{K}) \]

We define an extended KB as a pair \((\mathcal{K}, \mathcal{C})\) where \(\mathcal{K}\) is a SROIQ KB interpreted with standard semantics and \(\mathcal{C}\) is a set of SROIQ axioms interpreted with constraint semantics. We say that \((\mathcal{K}, \mathcal{C})\) is valid if \(\forall \alpha \in \mathcal{C}, \mathcal{K} \models IC \alpha\), otherwise there is an IC violation.

**Discussion**

It is easy to verify that the IC semantics provides expected results for the examples presented in the introduction and related work sections, and enables OWL to be an IC language. In Example 1, we get an IC violation since the IC interpretation of Product contains \(p\) but the IC interpretation of (\(\exists \text{hasProducer. Producer}\)) is empty. We also get an IC violation for Example 2 because due to the weak UNA, \(m_1\) and \(m_2\) are interpreted as different individuals causing the IC interpretation of (\((\leq \text{hasProducer. } \top)\)) to be empty. In Example 3 and Example 4 there are also IC violations because the constraint requires the same named producer to exist in every ME model of the KB which is not the case. Since our IC semantics is targeted at named individuals, one does not need to use the concept \(O\) as in Example 4.

In Example 5, the IC interpretation of \(\text{Category}_1\) is empty, therefore the constraint does not apply to \(p\) and there is no violation.

The following example shows how the weak UNA allows the individuals that are not asserted to be equal to be treated differently for constraint validation purposes.

**Example 6.** Consider a KB \(\mathcal{K}\) and a constraint \(\alpha\):

\[ \mathcal{K} = \{C(c), R(c, d_1), R(c, d_2), D(d_1), D(d_2)\} \]

\[ \alpha : C \sqsubseteq 2R.D \]

Individuals \(d_1\) and \(d_2\) are interpreted to be different in every ME model. Therefore, the IC-interpretation of (\(\geq 2R.D\)) includes \(c\) and the constraint \(\alpha\) is satisfied by \(\mathcal{K}\).

Now we illustrate another point regarding disjunctions in constraints.

**Example 7.** Suppose we have a KB \(\mathcal{K}\) and a constraint \(\alpha\):

\[ \mathcal{K} = \{C(a), (C_1 \sqcup C_2)(a)\} \]

\[ \alpha : C \sqsubseteq C_1 \sqcup C_2 \]

Constraint \(\alpha\) should be read as “every instance of \(C\) should be either a known instance of \(C_1\) or a known instance of \(C_2\)”. Since we do not know for sure whether \(\alpha\) belongs to \(C_1\) or \(C_2\), \(\alpha\) is expected to be violated by \(\mathcal{K}\). Indeed, according to our semantics we get \(C^{IC} \models \{a_2^2\}\) and \((C_1 \sqcup C_2)^{IC} \models \emptyset\). Therefore \(C^{IC} \not\subseteq (C_1 \sqcup C_2)^{IC}\) and we conclude that there is an IC violation.

If we want to represent the alternative constraint: “every instance of \(C\) should be an instance of \(C_1\) or \(C_2\)”, we can define a new name \(C'\) in the KB to substitute \(C_1\) and \(C_2\), thus having the new KB \(\mathcal{K}'\) and constraint \(\alpha'\) as follows:

\[ \mathcal{K}' = \{C(a), (C_1 \sqcup C_2)(a), C' \equiv C_1 \sqcup C_2\} \]

\[ \alpha' : C \sqsubseteq C' \]

There is no IC violation in this version because now the disjunction is interpreted as standard OWL axioms. As these
IC Validation

Previously we defined that the extended KB \( \langle K, C \rangle \) is valid if every IC axiom in \( C \) is IC-satisfied by \( K \). In this section, we describe how to do IC validation, i.e., check IC-satisfaction by translating constraint axioms to queries with the Negation As Failure (NAF) operator \( \text{not} \). We start by giving the formal semantics for DCQ\(_{\text{not}}\), then describe the translation rules from IC axioms to DCQ\(_{\text{not}}\), and finally provide a theorem showing that IC validation can be reduced to answering DCQ\(_{\text{not}}\) under certain conditions.

DCQ\(_{\text{not}}\)

In the Preliminaries section, we introduced standard DCQ. However, the expressivity of standard DCQ is not enough to capture the closed world nature of the IC semantics. For this reason, we add the \text{not} operator to DCQ to get DCQ\(_{\text{not}}\). The syntax of DCQ\(_{\text{not}}\) is defined as follows:

\[ Q \leftarrow q \mid Q_1 \land Q_2 \quad \text{not} \quad Q \]

The semantics of \text{not} is defined as:

\[ K \models_{\text{not}} Q \iff \neg \exists \sigma \text{ s.t. } K \models_{\sigma} Q \]

Recall that, assignment functions \( \sigma \) map variables only to named individuals so a brute force way to check if a \text{not} atom is entailed can be done by enumerating all possible assignments and checking for entailment.

Translation Rules: from ICs to DCQ\(_{\text{not}}\)

We now present the translation rules from IC axioms to DCQ\(_{\text{not}}\). The translation rules are similar in spirit to the Lloyd-Topor transformation (Lloyd 1987) but instead of rules we generate DCQ\(_{\text{not}}\). The idea behind the translation is to translate a constraint axiom into a query such that when the constraint is violated, the KB entails the query.

The translation contains two operators: \( T_c \) for translating concepts and \( T_r \) for translating axioms. \( T_r \) is a function that takes a concept expression and a variable as input and returns a DCQ\(_{\text{not}}\) as the result:

\[
T_c(C_a, x) := C_a(x) \\
T_r(\neg C, x) := \text{not} \; T_c(C, x) \\
T_r(C_1 \cap C_2, x) := T_c(C_1, x) \land T_c(C_2, x) \\
T_r(\exists nR.C, x) := \bigwedge_{1 \leq i \leq n} (R(x, y_i) \land T_c(C, y_i)) \land \bigwedge_{1 \leq i < j \leq n} \text{not} \; (y_i = y_j) \\
T_r(\exists R.\text{Self}, x) := R(x, x) \\
T_r(\{a\}, x) := (x = a)
\]

where \( C_a \) is an atomic concept, \( C(i) \) is a concept, \( R \) is a role, \( a \) is an individual, \( x \) is a variable, and \( y(i) \) is a fresh variable. \( T \) is a function that maps a SROIQ axiom to a DCQ\(_{\text{not}}\) as follows:

\[
T(C_1 \sqsubseteq C_2) := T_c(C_1, x) \land \text{not} \; T_c(C_2, x) \\
T(R_1 \sqsubseteq R_2) := R_1(x, y) \land \text{not} \; R_2(x, y) \\
T(R_1 \ldots R_n \sqsubseteq R) := R_1(x, y_1) \land \ldots \land R_n(y_{n-1}, y_n) \land \text{not} \; R(x, y_n)
\]

\( T(\text{Ref}(R)) := \text{not} \; R(x, x) \)
\( T(\text{Ir}(R)) := R(x, x) \)
\( T(\text{Dis}(R_1, R_2)) := R_1(x, y) \land R_2(x, y) \)

where \( C(i) \) is a concept, \( R(i) \) is a role, \( x \) and \( y(i) \) are variables.

Example 8. Given the constraint \( \alpha \) in Example 1, applying the above translation rules, we get:
\[ T(\alpha) := \text{Product}(x) \land \text{not} \; (\text{hasProducer}(x, y) \land \text{Producer}(y)) \]

Reducing IC Validation to DCQ\(_{\text{not}}\) Answering

Previously we stated our goal to reduce the problem of IC validation to query answering. However, the following example shows that when both the KB and the constraints use the full expressivity of SROIQ, we cannot use the query translation approach in a straightforward way.

Example 9. Suppose we have a KB \( K \) and a constraint \( \alpha \):
\[ K = \{ D(d), R(d, a), R(d, b), R(d, c), \{a\} \subseteq \{b, c\} \} \]
\[ \alpha : D \sqsubseteq \leq 2.R. \top \]

In all models of \( K \), \( a \) is either interpreted to be equivalent to \( b \) or \( c \). Therefore, in all interpretations \( D \) has less than two \( R \) values satisfying the constraint. However, the query translation will yield atoms in the form \( \text{not} \; (y_1 = y_2) \) which will be true for any individual pair in this KB. As a result, the answer set for this query will include \( d \) which incorrectly indicates an IC violation. This is because of the interaction between disjunctive (in)equality axioms in \( K \) and cardinality restrictions in ICs: axiom \( \{a\} \subseteq \{b, c\} \) asserts disjunctive equality between individuals, therefore the IC axiom containing cardinality restrictions is satisfied in different ways in different interpretations.

To avoid such interactions, we can either prohibit cardinality restrictions in ICs or prohibit disjunctive (in)equality in KBs. In SROIQ, there are only three ways to infer (in)equality between individuals: (1) explicit (in)equality axioms; (2) nominals (as seen above); and (3) cardinality restrictions. Obviously, explicit ABox assertions cannot be disjunctive so they are not problematic. By excluding nominals and cardinality restrictions from SROIQ, we get the DL SRI.

In Theorem 1, we show that IC validation via query answering is sound and complete for the cases where the expressivity of the extended KB is either \( \langle SRI, SROIQ \rangle \) or \( \langle SROIQ, SROIT \rangle \). Due to space limitations we only present the main theorem here and refer the reader to the technical report (Tao et al. 2010) for details.

Theorem 1. Given an extended KB \( \langle K, C \rangle \) with expressivity \( \langle SRI, SROIQ \rangle \) (\( \langle SROIQ, SROIT \rangle \) resp.), we have \( K \models_{IC} \alpha \) iff \( K \not\models T(\alpha) \) where \( \alpha \in C \).

Implementation

The most widely used query language on the Semantic Web is SPARQL (Prud’hommeaux and Seaborne 2008) which allows querying over OWL ontologies via OWL entailment regimes. It is known that SPARQL has the same expressive power as nonrecursive Datalog programs (Angles and
Gutierrez 2008) and can express DCQ\textsuperscript{not}. Therefore, based on the results from the previous section, we can reduce IC validation to SPARQL query answering if the KB is SRI or the ICs do not contain cardinality restrictions.

In Table 3, we present a mapping $M : Q \rightarrow P$ from a DCQ\textsuperscript{not} $Q$ to a SPARQL graph pattern $P$, such that $A(Q, K)$ will be equivalent to the result set obtained by evaluating $P = M(Q)$ over the corresponding OWL ontology using the OWL entailment. Note that, in our mapping we use the NOT EXISTS pattern which is being added in SPARQL 1.1\textsuperscript{5} and can also be encoded in SPARQL 1.0 using a combination of other operators (Angles and Gutierrez 2008).

<table>
<thead>
<tr>
<th>(Q)</th>
<th>(P = M(Q))</th>
</tr>
</thead>
<tbody>
<tr>
<td>(C(x))</td>
<td>(x \text{ rdf:type } C)</td>
</tr>
<tr>
<td>(A(x, y))</td>
<td>(x \text{ R } y)</td>
</tr>
<tr>
<td>(x = y)</td>
<td>(x \text{ owl:sameAs } y)</td>
</tr>
<tr>
<td>(x \neq y)</td>
<td>(x \text{ owl:differentFrom } y)</td>
</tr>
<tr>
<td>(Q_1 \land Q_2)</td>
<td>(\text{JOIN}(M(Q_1), M(Q_2)))</td>
</tr>
<tr>
<td>not (Q)</td>
<td>not EXISTS(M(Q))</td>
</tr>
</tbody>
</table>

Table 3: Mapping DCQ\textsuperscript{not} to SPARQL queries

The SPARQL representation of the query in Example 8 becomes:

\[
\text{ASK WHERE } \{
?x \text{ rdf:type Product, NOT EXISTS(\{?x hasProducer ?y, ?y rdf:type Producer.\})}
\]

We have built a prototype IC validator\textsuperscript{6} by extending the OWL 2 DL reasoner Pellet\textsuperscript{7}. The prototype reads OWL IC axioms, translates each IC first to a DCQ\textsuperscript{not} and then to a SPARQL query which is executed by the SPARQL engine in Pellet where a non-empty result indicates a constraint violation. Since the translation algorithm is reasoner independent this prototype can be used in conjunction with any OWL reasoner that supports SPARQL query answering.

We have used this prototype to validate ICs with several large ontologies such as the LUBM dataset. For testing, we removed several axioms from the LUBM ontology and declared them as ICs instead. The dataset is logically consistent but turning axioms into ICs caused some violations to be detected. Since each constraint is turned into a separate query there is no dependence between the validation time of different constraints. We have not performed extensive performance analysis for IC validation but as a simple comparison we compared the logical consistency checking time and the IC validation time. For LUBM(5), which has 100K individuals and 800K ABox axioms, logical consistency checking was on average 10 seconds whereas validating a single IC took on average 2 seconds. The naive approach in our prototype to execute each query separately would not scale well as the number of ICs increase. However, there are many improvement possibilities ranging from combining similar queries into a single query to running multiple queries in parallel.

\textsuperscript{3}http://www.w3.org/TR/sparql11-query/#negation
\textsuperscript{4}http://clarkparsia.com/pellet/oicv-0.1.2.zip
\textsuperscript{5}http://clarkparsia.com/pellet

Conclusions and Future Work

We described an IC semantics for OWL axioms that can be used for data validation purposes. Our IC semantics provide intuitive results for various different use cases. We presented translation rules from IC axioms to DCQ\textsuperscript{not}, showing that IC validation can be reduced to query answering when the KB expressivity is SRI or when the constraint expressivity is SROIQ. Our preliminary results with a prototype IC validator implementation show that existing OWL reasoners can be used for IC validation efficiently with little effort. Using SPARQL queries for IC validation makes our approach applicable to a wide range of reasoners. In the future, we will look at the performance of IC validation in real-world datasets and investigate IC validation algorithms for the full expressivity of SROIQ.

References

Prud’hommeaux, E., and Seaborne, A. 2008. SPARQL Query Language for RDF.