Bidirectional Constraints for Exchanging Data: Beyond Monotone Queries

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Abstract

In this paper, we propose to use the language of bidirectional constraints to specify schema mappings in the context of data exchange. These constraints impose restrictions over both the source and the target data, and have the potential to minimize the ambiguity in the description of the target data to be materialized. We start by making a case for the usefulness of bidirectional constraints to give a meaningful closed-world semantics for st-tgd, which is motivated by Clark’s predicate completion and Reiter’s formalization of the closed-world assumption of a logical theory. We then formally study the use of bidirectional constraints in data exchange. In particular, we pinpoint the complexity of the existence-of-solutions and the query evaluation problems in several different scenarios, including in the latter case both monotone and non-monotone queries.

1 Introduction

Data exchange is the process of taking data structured under a source schema and transforming it into data structured under a target schema. A fundamental building block in this process is the notion of a schema mapping, which is a high-level specification that describes how data from the source should be mapped into the target.

In relational databases, schema mappings are usually specified by using a logical language. In particular, most of the research on data exchange [Lenzerini, 2002; Fagin et al., 2005a] has focused on mappings specified by the so-called source-to-target tuple-generating dependencies (st-tgds), which have been also widely used in practical applications [Haas et al., 2005; Bernstein et al., 2006]. An st-tgd is a first-order logic sentence of the form:

$$\forall \bar{x} \forall \bar{y} \left( \varphi_{S}(\bar{x}, \bar{y}) \rightarrow \exists \bar{z} \psi_{T}(\bar{x}, \bar{z}) \right),$$

where $\varphi_{S}(\bar{x}, \bar{y})$ and $\psi_{T}(\bar{x}, \bar{z})$ are conjunctions of relational atoms from the source and target schemas, respectively. For example, consider a source schema consisting of a relation $\text{Student}(\text{name}, \text{course})$ storing students’ names and the names of the courses where they are enrolled in. Moreover, consider a target schema consisting of a relation $\text{Enroll}(\text{name}, \text{cid})$, storing students’ names and the ids of the courses where they are enrolled in, and of a relation $\text{Course}(\text{cid}, \text{name})$, storing for every course its id and name. In this case, the exchange is driven by the following st-tgd:

$$\forall \bar{x} \forall \bar{y} \left( \text{Student}(\bar{x}, \bar{y}) \rightarrow \exists \bar{z} \left( \text{Enroll}(\bar{x}, \bar{z}) \land \text{Course}(\bar{z}, \bar{y}) \right) \right). \quad (1)$$

In data exchange, one is given a source database $I$ and a schema mapping $\mathcal{M}$, and then the main problem is to find a target database $J$ that is a solution for $I$ under $\mathcal{M}$, that is, a valid translation of $I$ according to the constraints imposed by $\mathcal{M}$. For example, consider a mapping $\mathcal{M}$ specified by st-tgd (1) and the following source database $I_1$:

<table>
<thead>
<tr>
<th>Student</th>
<th>name</th>
<th>course</th>
</tr>
</thead>
<tbody>
<tr>
<td>John</td>
<td></td>
<td>AI</td>
</tr>
<tr>
<td>Albert</td>
<td></td>
<td>Databases</td>
</tr>
</tbody>
</table>

Then the following target database $J_1$ is a possible solution for $I_1$ under the constraints imposed by $\mathcal{M}$:

<table>
<thead>
<tr>
<th>Enroll</th>
<th>name</th>
<th>cid</th>
</tr>
</thead>
<tbody>
<tr>
<td>John</td>
<td></td>
<td>CS301</td>
</tr>
<tr>
<td>Albert</td>
<td></td>
<td>CS303</td>
</tr>
</tbody>
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<table>
<thead>
<tr>
<th>Course</th>
<th>cid</th>
<th>cname</th>
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<tbody>
<tr>
<td>CS301</td>
<td>AI</td>
<td>Databases</td>
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<tr>
<td>CS303</td>
<td></td>
<td>Databases</td>
</tr>
</tbody>
</table>

In fact, $I_1$ together with $J_1$ satisfy st-tgd (1) according to the standard first-order logic semantics. Nevertheless, $J_1$ is not the only target database that satisfies this condition. Consider, for example, the following target database $J_2$:

<table>
<thead>
<tr>
<th>Enroll</th>
<th>name</th>
<th>cid</th>
</tr>
</thead>
<tbody>
<tr>
<td>John</td>
<td></td>
<td>CS301</td>
</tr>
<tr>
<td>Albert</td>
<td></td>
<td>CS303</td>
</tr>
<tr>
<td>Albert</td>
<td></td>
<td>CS400</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Course</th>
<th>cid</th>
<th>cname</th>
</tr>
</thead>
<tbody>
<tr>
<td>CS301</td>
<td>AI</td>
<td>Databases</td>
</tr>
<tr>
<td>CS303</td>
<td></td>
<td>Databases</td>
</tr>
<tr>
<td>CS400</td>
<td>Algorithms</td>
<td></td>
</tr>
</tbody>
</table>

Again, we have that $I_1$ together with $J_2$ satisfy st-tgd (1) according to the standard first-order logic semantics. Thus, the database $J_2$, although less natural than $J_1$, is also a solution for $I_1$ according to the constraints imposed by $\mathcal{M}$. This sort of anomaly is caused by the semantics of the implication; the formula used to exchange data is not restricting the possibility of adding arbitrary tuples to $\text{Enroll}$ and $\text{Course}$. Moreover, this formula does not impose any restrictions on the courses’ ids; if we replace CS301 by CS401 in the database $J_2$, we also obtain a solution for $I_1$ under $\mathcal{M}$. Thus, target databases are allowed to contain labeled nulls in this context, which represent values that can be replaced by any constant. In fact, the following target database is a solution for $I_1$ under $\mathcal{M}$, where $\perp_1$ and $\perp_2$ are distinct null values:

$$\forall \bar{x} \forall \bar{y} \left( \text{Student}(\bar{x}, \bar{y}) \rightarrow \exists \perp_1 \perp_2 \left( \text{Enroll}(\bar{x}, \perp_1) \land \text{Course}(\perp_1, \perp_2) \right) \right).$$
The semantics of st-tgds has raised several issues in data exchange. One of the most prominent is the problem of answering target queries. Given a source database, a schema mapping, and a query \( Q \) over the target schema, what should be the answer for \( Q \) after the exchange? Given that there are many possible solutions for the source instance, this semantics is not immediately clear.

The certain answers semantics has been adopted as the standard semantics for query answering in data exchange. More precisely, a tuple \( t \) is said to be a certain answer for a target query \( Q \) given a source database \( I \) and a mapping \( \mathcal{M} \), if \( t \) is an answer for \( Q \) over every possible solution for \( I \) under the constraints imposed by \( \mathcal{M} \). Although this semantics gives intuitively correct results for positive queries [Fagin et al., 2003; 2005a], it has a less natural behavior for non-monotone queries [Arenas et al., 2004; Libkin, 2006; Hernich, 2013]. For instance, consider again the source database \( I_1 \) and the mapping \( \mathcal{M} \) specified by st-tgd (1), and assume that \( Q \) is the following target query:

\[
\neg \exists z (\text{Enroll}(x, z) \land \text{Course}(z, \text{Programming})).
\] (2)

This query is asking whether John is not enrolled in Programming. Given the data in \( I_1 \), the natural answer to \( Q \) would be true. However, the actual answer to this query is false as there exists a possible solution for \( I_1 \) under \( \mathcal{M} \) where John is enrolled in Programming.

To tackle this problem, many authors have proposed to restrict the notion of data exchange solution based on some minimality criteria [Fagin et al., 2005b; Libkin, 2006; Afrati and Kolaitis, 2008; Hernich et al., 2011; Libkin and Sirangelo, 2011; Gottlob et al., 2011; Hernich, 2012; 2013], which has led to different alternatives for the semantics of non-monotone queries in this context. Regardless of the usefulness of these proposals, each one of them departs from the well-understood semantics of first-order logic for st-tgds, as initially proposed in [Fagin et al., 2005a]. We are convinced that having a clean semantics for schema mappings based on the semantics of first-order logic has been instrumental in the adoption of this technology. Thus, as opposed to the approaches mentioned above, our main goal in this work is to propose and study a mapping-specification language that can be defined in terms of the usual syntax and semantics of first-order logic (FO), and whose semantics is suitable for both monotone and non-monotone queries.

In [Arenas et al., 2014b], we proposed to use the so-called bidirectional constraints to specify schema mappings. A bidirectional constraint is an FO-sentence of the form \( \forall \sigma \exists \tau (s(\vec{x}) \leftrightarrow \beta_\sigma(\vec{x})) \), where \( s(\vec{x}) \) and \( \beta_\sigma(\vec{x}) \) are FO-formulas over the source and target schemas, respectively. Thus, these dependencies allow expressing in a natural way what pieces of data should and should not be in the target, reducing the ambiguity when describing what target instances should be materialized. In this paper, we embark on a study of the fundamental properties of these constraints regarding data exchange and non-monotone queries, aiming to meet the requirements mentioned in the previous paragraph. It is important to mention that although this type of dependencies has been previously considered in the literature [Melnik et al., 2008; Arenas et al., 2014b], the properties studied in this paper have not been investigated.

Our contributions. First, in Section 3 we make a case for the usefulness of bidirectional constraints to give a meaningful closed-world semantics for st-tgds. Our proposal is motivated by the classical notion of predicate completion [Clark, 1977] and Reiter’s formalization of the closed-world assumption of a logical theory [Reiter, 1977]. Essentially, we prove that for every schema mapping \( \mathcal{M} \) specified by a set of st-tgds, one can compute a closure of \( \mathcal{M} \) specified by a set of bidirectional constraints which, in a precise sense, completes \( \mathcal{M} \) assuming as negative any piece of information for which there is no evidence according to \( \mathcal{M} \). For example, for the case of st-tgd (1), the closure is very simple and is given by the following dependency:

\[
\forall x \forall y (\text{Student}(x, y) \leftrightarrow \exists z (\text{Enroll}(x, z) \land \text{Course}(z, y))).
\] (3)

It is worth noticing that under this bidirectional constraint, we obtain true as the answer to the query (2) given the source instance \( I_1 \) defined previously, which is the expected answer. Besides, it is worth noticing that although the closure was obtained in this case by just replacing an implication by a double implication, this process is much more involved in general.

Second, we study bidirectional constraints in depth by considering two fundamental data exchange problems: the existence of solutions problem in Section 4, and the query evaluation problem in Section 5. We pinpoint the complexity of these problems in several different scenarios and, in particular, for several different classes of monotone and non-monotone queries in Section 5. We also describe several restricted scenarios for which these problems become tractable.

2 Preliminaries

We assume that data is represented in the relational model. A relational schema \( R \), or just schema, is a finite set \( \{R_1, \ldots, R_n\} \) of relation symbols, with each \( R_j \) having a fixed arity \( k_j \). Let \( C \) and \( N \) be countably infinite sets with no elements in common. We refer to the elements in \( C \) as constants and to the elements in \( N \) as labeled nulls, or just nulls. An instance \( I \) of \( R \) assigns to each relation symbol \( R_j \) a finite \( k_j \)-ary relation \( R_j^I \subseteq (C \cup N)^{k_j} \). We denote by \( \text{Inst}(R) \) the set of all instances of \( R \). Finally, slightly abusing notation, we use \( \text{C}(\cdot) \) to denote a built-in predicate such that \( \text{C}(a) \) holds if and only if \( a \) is a constant (that is, \( a \in C \)).

Query languages. We assume some familiarity with first-order logic (FO), and we focus on queries expressed by using formulas in this logic. Besides FO, the main query languages considered in this paper are the languages of conjunctive queries (CQ), unions of conjunctive queries (UCQ), and the languages obtained from them by adding the equality predicate, the inequality predicate and the negation operator (e.g. UCQ, CQ and UCQ*). We also consider the class of monotone queries, denoted by \( \text{Mon} \). This class contains all queries \( Q \) over a schema \( R \) that satisfy the following property: given two instances \( J_1, J_2 \) of \( R \) such that \( J_1 \subseteq J_2 \), it holds that \( Q(J_1) \subseteq Q(J_2) \). Finally, we assume that all query languages allow the use of built-in predicate \( \text{C} \).
Schema mappings. As is customary in the data exchange literature, if we refer to a schema $S$ as a source schema, then $\text{Inst}(S)$ is defined to be the set of all instances of $S$ that are constructed by using only elements from $C$, and if we refer to a schema $T$ as a target schema, then $\text{Inst}(T)$ is defined as usual (that is, the instances of $T$ are constructed by using elements from both $C$ and $N$). In this article, we use $S$ to refer to a source schema and $T$ to refer to a target schema. Besides, in this section we assume that $S$ and $T$ are fixed source and target schemas with no relation symbols in common.

Schema mappings are used to define a semantic relationship between a source and a target schema. A schema mapping, or just mapping, $\mathcal{M}$ from $S$ to $T$ is a set of pairs $(I,J)$, where $I$ is an instance of $S$, and $J$ is an instance of $T$. That is, a mapping $\mathcal{M}$ is just a subset of $\text{Inst}(S) \times \text{Inst}(T)$. Given an instance $I$ of $S$, a mapping $\mathcal{M}$ associates to $I$ a set of possible solutions for $I$, denoted by $\text{SOL}_M(I)$, given by the set $\{J \in \text{Inst}(T) \mid (I,J) \in \mathcal{M}\}$. Notice that the set $\text{SOL}_M(I)$ represents the possible translations of $I$ according to $\mathcal{M}$.

In practice, schema mappings are represented by using logical formulas. Again, we focus on using fragments of FO to specify mappings. Given a finite set $\Sigma$ of FO-sentences over vocabulary $S \cup T$, we say that a mapping $\mathcal{M}$ from $S$ to $T$ is specified by $\Sigma$ if for every pair of instances $(I,J) \in \text{Inst}(S) \times \text{Inst}(T)$, it holds that $(I,J) \in \mathcal{M}$ if and only if $I \cup J \models \Phi$ for every $\Phi \in \Sigma$, where $I \cup J$ represents the instance of $S \cup T$ obtained by taking the union of the relations in $I$ and $J$ (recall that $S$ and $T$ have no relation symbols in common). For convenience, we write the last statement as $(I,J) \models \Sigma$. Therefore, we usually refer to a mapping as a triple $\mathcal{M} = (S,T,\Sigma)$, and to the set of solutions as $\text{SOL}_M(I) = \{J \in \text{Inst}(T) \mid (I,J) \models \Sigma\}$.

Dependencies. Schema mappings from $S$ to $T$ are usually defined by source-to-target dependencies [Fagin et al., 2005a], that is, by formulas of the form $\forall \bar{x} (\varphi(\bar{x}) \rightarrow \psi(\bar{x}))$, where $\varphi(\bar{x})$ is an FO-formula over $S$, $\psi(\bar{x})$ is an FO-formula over $T$, and $\bar{x}$ is the tuple of free variables of both formulas. We usually drop the outermost universal quantification when specifying these constraints, and thus we only write $\varphi(\bar{x}) \rightarrow \psi(\bar{x})$ for the previous formula. Depending on which fragments of FO we use to define formulas $\varphi(\bar{x})$ and $\psi(\bar{x})$, we obtain a wide range of possible fragments of source-to-target dependencies. Given fragments $\mathcal{L}_1$ and $\mathcal{L}_2$ of FO, an $\mathcal{L}_1$-to-$\mathcal{L}_2$ dependency is a formula of the above form in which $\varphi(\bar{x})$ is an $\mathcal{L}_1$-formula over $S$ and $\psi(\bar{x})$ is an $\mathcal{L}_2$-formula over $T$. In [Fagin et al., 2005a], the language of CQ-to-CQ dependencies was chosen as the preferred formalism for specifying schema mappings, calling them source-to-target tuple-generating dependencies (st-tgds). In this article, we also consider full $\mathcal{L}$-to-$\mathcal{L}$ dependencies, which are formulas in which the target part is a conjunctive query without existential quantifiers.

In this paper, we study mappings specified by sets of formulas of the following form

$$\forall \bar{x} \left( \varphi(\bar{x}) \leftrightarrow \psi(\bar{x}) \right), \quad \text{(4)}$$

where $\varphi(\bar{x})$ is an FO-formula over $S$, $\psi(\bar{x})$ is an FO-formula over $T$, and $\bar{x}$ is the tuple of free variables of both formulas. Such a formula is called a bidirectional constraint [Arenas et al., 2014b]. We usually drop the outermost universal quantification, and we write $\varphi(\bar{x}) \leftrightarrow \psi(\bar{x})$ for formula (4). We say that $\Phi$ is an $\langle \mathcal{L}_1, \mathcal{L}_2 \rangle$-dependency between $S$ and $T$, if $\Phi$ is of the form (4) with $\varphi(\bar{x})$ in $\mathcal{L}_1$ and $\psi(\bar{x})$ in $\mathcal{L}_2$. When the source and target schemas are clear from the context, we will only talk about $\langle \mathcal{L}_1, \mathcal{L}_2 \rangle$-dependencies. We also consider full $(\mathcal{L}, \mathcal{C}Q)$-dependencies, defined as before.

Query answering in data exchange. An important aspect in data exchange is how to answer queries over a target schema. The most accepted semantics for query answering is the certain answers semantics: given a mapping $\mathcal{M}$ from $S$ to $T$, an instance $I$ of $S$ and a query $Q$ over $T$, the certain answers of $Q$ with respect to $I$ under $\mathcal{M}$ is the set $\text{CERTAIN}_M(Q,I) = \bigcap_{J \in \text{SOL}_M(I)} Q(J)$. In other words, a tuple $\bar{t} \in \text{CERTAIN}_M(Q,I)$ if and only if $\bar{t} \in Q(J)$ for every solution $J$ for $I$ under $\mathcal{M}$.

3 Closing Mappings through Bidirectional Constraints

The problem of answering non-monotone queries is of great interest in the data exchange area. In particular, the issue of defining a meaningful semantics for such queries has received a lot of attention, mainly because an open-world approach was followed in the initial definition of the notion of solution in data exchange [Fagin et al., 2005a]. Many authors have proposed restrictions on this notion based on some minimality criteria [Fagin et al., 2005b; Libkin, 2006; Afrati and Kolaitis, 2008; Hernich et al., 2011; Libkin and Sirangelo, 2011; Gottlob et al., 2011; Hernich, 2012; 2013], which has led to some different alternatives for the semantics of non-monotone queries in this context.

Arguably, the main advantage of the data exchange framework proposed in [Fagin et al., 2005a] is that mappings are specified by using st-tgds, whose syntax and semantics are defined in a clean way by relying on the well-understood syntax and semantics of first-order logic. We are convinced that having such a clean framework has been instrumental in the adoption of this technology. Thus, as opposed to the approaches mentioned in the previous paragraph, our main goal in this work is to propose and study a mapping-specification language that can be defined in terms of the usual syntax and semantics of first-order logic, and whose semantics is suitable for both monotone and non-monotone queries.

Bidirectional constraints allow expressing in a natural way what pieces of data should and should not be in the target. This, together with the fact that their syntax and semantics is inherited from first-order logic, makes these constraints quite appealing to meet the aforementioned requirements. Moreover, as shown next, bidirectional constraints can be used to give a closed-world semantics to a mapping specified by st-tgds, as these constraints are expressive enough to encode an extension of the classical notion of predicate completion [Clark, 1977] and, at the same time, an adaptation of Reiter’s formalization of the closed-world assumption [Reiter, 1977] to the case of st-tgds. This gives more evidence that bidirectional constraints form a good mapping-specification language that deserves careful consideration.
Let \( M = (S, T, \Sigma) \) be a mapping specified by a set \( \Sigma \) of st-tgds. Then a mapping \( M^* \) from \( S \) to \( T \) is said to be a closure of \( M \) if for every st-tgd \( \varphi(x) \rightarrow \psi(x) \in \Sigma \), every source instance \( I \) and every tuple \( t \) of constants:

\[
\text{if } t \in \text{CERTAIN}_M(\psi(x), I), \quad \text{then } t \in \text{CERTAIN}_{M^*}(\psi(x), I); \quad \text{and}
\]

\[
\text{if } t \notin \text{CERTAIN}_M(\psi(x), I), \quad \text{then } t \in \text{CERTAIN}_{M^*}(\neg\psi(x), I).
\]

To the best of our knowledge, this formalization of the notion of a closure of a mapping is new. Notice that this notion is an extension of Clark’s predicate completion [Clark, 1977], as in \( M^* \) we complete the definition of the heads of the dependencies in \( \Sigma \). Besides, this notion is also inspired by Reiter’s formalization of the closed-world assumption [Reiter, 1977], as if \( \psi(t) \) cannot be inferred from \( \Sigma \), then \( \neg\psi(t) \) is added into the theory.

A natural question is whether a closure of a mapping \( M \) given by st-tgds can be expressed by using also st-tgds. The following proposition gives a negative answer to this question, even if we are allowed to use FO-TO-CQ dependencies.

**Proposition 1.** There is a mapping \( M = (S, T, \Sigma) \), where \( \Sigma \) is a set of st-tgds, for which no mapping \( M^* = (S, T, \Sigma^*) \) is a closure, assuming that \( \Sigma^* \) is a set of FO-TO-CQ dependencies.

This proposition follows from the fact that for every mapping \( M = (S, T, \Sigma) \) specified by a set \( \Sigma \) of FO-TO-CQ dependencies, every instance \( I \) of \( S \) and every conjunctive query \( Q(\bar{x}) \) over \( T \), it holds that \( \text{CERTAIN}_{M^*}(\neg Q(\bar{x}), I) = \emptyset \). In fact, this property holds even if \( \Sigma^* \) is restricted to contain a single copying dependency \( A(x) \rightarrow A'(x) \). Thus, the class of FO-TO-CQ dependencies is not expressive enough to capture the certain answers semantics for the negation of conjunctive queries, which is required for expressing the closure of a mapping.

Proposition 1 tells us that some restrictions on the target instances considered as solutions have to be imposed in order to obtain the closure of a mapping. Target dependencies have been used in data exchange as a primary way to restrict the spaces of solutions of source instances [Arenas et al., 2014a], so they form a natural alternative to solve this problem. Unfortunately, the following proposition shows that this alternative does not work. Notice that in this proposition we use notation \( M = (S, T, \Sigma, \Gamma) \) to represent a mapping with target dependencies, where \( \Sigma \) is a set of FO-TO-CQ dependencies and \( \Gamma \) is a set of FO-sentences over \( T \). Moreover, we assume that \( J \in \text{Sol}_M(I) \) if and only if \( (I, J) \models \Sigma \) and \( J \models \Gamma \). Again, this result holds even for a mapping \( M \) specified by a single copying dependency \( A(x) \rightarrow A'(x) \).

**Proposition 2.** There is a mapping \( M = (S, T, \Sigma) \), where \( \Sigma \) is a set of st-tgds, for which no mapping \( M^* = (S, T, \Sigma^*, \Gamma) \) is a closure, assuming that \( \Sigma^* \) is a set of FO-TO-CQ dependencies and \( \Gamma \) is a set of FO-sentences over \( T \).

Bidirectional constraints come to the rescue at this point, as it is possible to show that every mapping given by st-tgds has a closure specified by a set of bidirectional constraints. In the rest of this section, we outline the proof of this result. In fact, we provide an algorithm that given a mapping \( M \) specified by a set of st-tgds, computes a mapping \( \text{COMP}(M) \) specified by a set of bidirectional constraints such that \( \text{COMP}(M) \) is a closure of \( M \).

Our algorithm is based on query rewriting. Next we review the terminology and results about this concept needed in our algorithm. Let \( M = (S, T, \Sigma) \), where \( \Sigma \) is a set of st-tgds, and assume that \( Q \) is a query over \( T \). Then a query \( Q' \) over \( S \) is a rewriting of \( Q \) over the source if for every \( I \in \text{Inst}(S) \), it holds that \( Q'(I) = \text{CERTAIN}_M(Q, I) \). That is, to obtain the set of certain answers of \( Q \) over \( I \), we just have to evaluate its rewriting \( Q' \) over instance \( I \).

The computation of a rewriting of a conjunctive query is a basic step in the algorithm presented in this section. This problem has been extensively studied in the literature [Levy et al., 1995; Abiteboul and Duschka, 1998] and, in particular, in the data integration context [Halevy, 2000; 2001; Lenzerini, 2002]. In fact, it is known that there exists an exponential-time algorithm \text{QUERYREWITING} that gives a mapping \( M = (S, T, \Sigma) \), with \( \Sigma \) a set of st-tgds, and a conjunctive query \( Q \) over \( T \), computes a query \( Q' \) over \( S \) such that \( Q' \) is in UCQ and \( Q' \) is a rewriting of \( Q \) over the source.

We are ready to present our algorithm to compute a closure of a mapping. In this algorithm, if \( \bar{x} = (x_1, \ldots, x_k) \), then \( C(\bar{x}) \) is a shorthand for \( C(x_1) \land \cdots \land C(x_k) \), where \( C(\cdot) \) is the unary built-in predicate introduced in the preliminaries which distinguishes between constants and nulls.

**Algorithm \text{CLOSE}(M)**

**Input:** \( M = (S, T, \Sigma) \), where \( \Sigma \) is a set of st-tgds.

**Output:** \( \text{COMP}(M) = (S, T, \Delta) \), where \( \Delta \) is a set of bidirectional constraints and \( \text{COMP}(M) \) is a closure of \( M \).

1. Start with \( \Delta \) as the empty set.
2. For every st-tgd \( \varphi(\bar{x}) \rightarrow \psi(\bar{x}) \in \Sigma \), do the following:
   (a) Use \text{QUERYREWITING}(M, \psi(\bar{x})) to compute a query \( \alpha(\bar{x}) \) that is a rewriting of \( \psi(\bar{x}) \) over the source.
   (b) Add dependency \( \alpha(\bar{x}) \leftrightarrow (\psi(\bar{x}) \land C(\bar{x})) \) to \( \Delta \).
3. Return \( \text{COMP}(M) = (S, T, \Delta) \).

The following theorem establishes the correctness of algorithm \text{CLOSE}. Its proof is direct, and it is based on some general properties regarding the relationship between the solutions under a mapping and the solutions under its closure.

**Theorem 1.** Let \( M = (S, T, \Sigma) \), where \( \Sigma \) is a set of st-tgds. Then \( \text{CLOSE}(M) \) computes a closure of \( M \) in exponential time in the size of \( \Sigma \), which is specified by a set of \( \text{UCQ}, \text{CQ} \)-dependencies.

We conclude this section by pointing out that the previous result provides evidence of the richness of the language of bidirectional constraints. In what follows, we continue our investigation of this language by studying two fundamental problems in data exchange, namely the existence of solutions and the query evaluation problems.
4 The Existence of Solutions Problem

As pointed out before, one of the main problems to be solved in the data exchange area is the issue of computing a valid translation of a source instance according to a mapping. The natural decision problem associated to this issue is what has been called the existence of solutions problem: to decide, given a mapping $M$ and a source instance $I$, whether $\text{SOL}_M(I) \neq \emptyset$. For the case of a mapping specified by a set of st-tgds, this problem is trivial as every source instance has a solution. However, for more expressive mapping languages the situation can be different, in particular for the case of bidirectional constraints as shown in the following example.

**Example 1.** Take a simple setting $M = (S, T, \Sigma)$, with $\Sigma = \{A(x) \rightarrow R(x), B(x) \rightarrow R(x)\}$. Then for the source instance $I = \{B(1)\}$, we have that the instance $J = \{R(1)\}$ is a solution under $M$. Now consider the setting $M' = (S, T, \Delta)$, with $\Delta$ consisting of the following bidirectional constraints:

$$A(x) \leftrightarrow R(x) \text{ and } B(x) \leftrightarrow R(x).$$

As opposed to the previous case, $I$ does not have any solution under $M'$.

In this section, we embark on a study of the complexity of the existence of solutions problem for different types of bidirectional constraints. Formally, given a class $C$ of bidirectional constraints, this problem is defined as follows:

<table>
<thead>
<tr>
<th>Problem: $\text{EXISTENCEOFSOLUTIONS}(C)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Input: Mapping $M = (S, T, \Sigma)$, where $\Sigma \subseteq C$, and an instance $I$ of $S$</td>
</tr>
<tr>
<td>Question: Is $\text{SOL}_M(I) \neq \emptyset$?</td>
</tr>
</tbody>
</table>

The starting point of our investigation is the class of $\langle\text{CQ}, \text{CQ}\rangle$-dependencies, which is the natural counterpart of the class of st-tgds. Moreover, we consider the class of $\langle\text{UCQ}^-, \text{CQ}\rangle$-dependencies, as they are needed in the algorithm in Section 3 to compute the closure of a mapping given by st-tgds. Finally, we consider the restrictions of these two classes to the case of full dependencies, as full source-to-target dependencies are widely used in theory and practice.

It is important to notice that the previous definition corresponds to the combined complexity [Vardi, 1982] of the existence of solutions problem. As mapping specifications are usually much smaller than source instances, it is also natural in this context to consider the data complexity [Vardi, 1982] of this problem, which is defined by assuming that a mapping $M = (S, T, \Sigma)$ is fixed, and then considering the input to be the source instance alone. This problem is denoted by $\text{EXISTENCEOFSOLUTIONS}(M)$.

The following theorem summarizes both the combined complexity results obtained in our investigation and, for the sake of completeness, the data complexity results obtained in [Arenas et al., 2014b]. The results in this theorem should be read as follows. Let $C$ be a class of bidirectional constraints. When we refer to the combined complexity of the existence of solutions problem for $C$, we are talking about the complexity of $\text{EXISTENCEOFSOLUTIONS}(C)$. On the other hand, when we say that the data complexity of the existence of solutions problem for $C$ is in a complexity class $\mathcal{N}$, we mean that $\text{EXISTENCEOFSOLUTIONS}(M) \in \mathcal{N}$ for every mapping $M$ defined by a set of bidirectional constraints in $C$. Moreover, when we say that the data complexity of the existence of solutions problem for $C$ is $\mathcal{N}$-complete, we mean that the data complexity of the existence of solutions problem for $C$ is in $\mathcal{N}$, and there exists a mapping $M_0$ defined by a set of bidirectional constraints in $C$ such that $\text{EXISTENCEOFSOLUTIONS}(M_0)$ is $\mathcal{N}$-hard.

**Theorem 2.** Let $C$ be any of the classes of bidirectional constraints shown in the header of Table 1. Then the data and combined complexity of the existence of solutions problems for $C$ are as stated in Table 1.

<table>
<thead>
<tr>
<th>Data Complexity</th>
<th>Combined Complexity</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\text{full } \langle\text{CQ}, \text{CQ}\rangle$-dependencies</td>
<td>$\text{full } \langle\text{UCQ}^-, \text{CQ}\rangle$-dependencies</td>
</tr>
<tr>
<td>in PTIME</td>
<td>in PTIME</td>
</tr>
<tr>
<td>$\Pi^2_2$-complete</td>
<td>$\Pi^2_2$-complete</td>
</tr>
<tr>
<td>$\text{NP}$-complete</td>
<td>$\text{NP}$-complete</td>
</tr>
</tbody>
</table>

The upper bounds in Table 1 for the combined complexity of the existence of solutions problem are derived directly. The lower bound in this table for the case of $\langle\text{CQ}, \text{CQ}\rangle$-dependencies, and thus also for the case of $\langle\text{UCQ}^-, \text{CQ}\rangle$-dependencies, is shown via a reduction from TILING [Papadimitriou, 1994], a well-known $\text{NEXPTIME}$-complete problem. The input of this problem is a set of tile types $T = \{t_0, \ldots, t_m\}$, relations $H \subseteq T \times T$ and $V \subseteq T \times T$ (which represent horizontal and vertical adjacency constraints between tile types), and an integer $n$ given in unary. Then the question to answer is whether there exists a tiling of a $2^n \times 2^n$ square with tile types in $T$, starting with tile type $t_0$ in the origin and satisfying the constraints imposed by $H$ and $V$. The lower bound in Table 1 for the case of full $\langle\text{CQ}, \text{CQ}\rangle$-dependencies, and thus also for the case of full $\langle\text{UCQ}^-, \text{CQ}\rangle$-dependencies, is shown via a reduction from the problem of determining if a quantified Boolean formula of the form $\forall \bar{x} \exists \bar{y} \varphi(\bar{x}, \bar{y})$ is true, where $\varphi$ is a propositional formula in 3-CNF and $\bar{x}, \bar{y}$ form a partition of the variables mentioned in $\varphi$. This problem is known to be $\Pi^2_2$-complete [Stockmeyer, 1976; Wrathall, 1976].

5 The Query Evaluation Problem

A fundamental problem in data exchange is the computation of certain answers. In this section, we study in depth the complexity of this problem for the case of bidirectional constraints. In particular, we are interested in the complexity of answering non-monotone queries, thus going beyond the more classical scenario of positive queries.

Given a class $C$ of bidirectional constraints and a class $Q$ of queries, the problem that we study is defined as follows:
Table 2: The data complexity of the certain answers problem.

<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>coNP-complete</td>
<td>coNP-complete</td>
<td>coNP-complete</td>
<td>coNP-complete</td>
<td>coNP-complete</td>
<td>coNP-complete</td>
<td>undecidable</td>
</tr>
<tr>
<td>coNP-complete</td>
<td>coNP-complete</td>
<td>coNP-complete</td>
<td>coNP-complete</td>
<td>coNP-complete</td>
<td>coNP-complete</td>
<td>undecidable</td>
</tr>
</tbody>
</table>

Table 3: The combined complexity of the certain answers problem.

<table>
<thead>
<tr>
<th>Dependencies</th>
<th>CQ</th>
<th>UCQ^[n]</th>
<th>UCQ^[1]</th>
<th>UCQ^[2]</th>
<th>UCQ^-</th>
</tr>
</thead>
<tbody>
<tr>
<td>coNEXPTIME-complete</td>
<td>coNEXPTIME-complete</td>
<td>coNEXPTIME-complete</td>
<td>coNEXPTIME-complete</td>
<td>coNEXPTIME-complete</td>
<td></td>
</tr>
<tr>
<td>coNEXPTIME-complete</td>
<td>coNEXPTIME-complete</td>
<td>coNEXPTIME-complete</td>
<td>coNEXPTIME-complete</td>
<td>coNEXPTIME-complete</td>
<td></td>
</tr>
</tbody>
</table>

Theorem 3. Let \( \mathcal{C} \) be either the class of full \( \langle \text{UCQ}\^[n], \text{CQ}\rangle \)-dependencies or the class of \( \langle \text{UCQ}\^[n], \text{CQ}\rangle \)-dependencies.

1. If \( Q \in \{ \text{CQ}, \text{MON}, \text{UCQ}\^[n], \text{UCQ}\^[2], \text{UCQ}^-, \text{FO} \} \), then the data complexity of the certain answers problem for \( \mathcal{C} \) and \( Q \) is as stated in Table 2.

2. If \( Q \in \{ \text{CQ}, \text{UCQ}^\#, \text{UCQ}\^[1], \text{UCQ}\^[2], \text{UCQ}^- \} \), then the combined complexity of the certain answers problem for \( \mathcal{C} \) and \( Q \) is as stated in Table 3.

Now we briefly discuss our results, starting with the data complexity of the certain answers problem. The coNP-completeness of the problem for \( \langle \text{UCQ}\^[n], \text{CQ}\rangle \)-dependencies is consistent with the NP-completeness of the existence of solutions problem for this type of dependencies (shown in Table 1). In fact, the lower bound for the case of conjunctive queries, which is also a lower bound for all the other query languages considered in Table 2, is proved by slightly modifying a reduction from 3-COLORABILITY used in [Arenas et al., 2014b] to prove the NP-hardness (in data complexity) of the existence of solutions problem for the case of \( \langle \text{CQ}, \text{CQ}\rangle \)-dependencies. The upper bound for UCQ\^[n] queries, which is also an upper bound for the query languages in the first four columns of Table 2, is shown by using a modified version of the chase procedure [Maier et al., 1979], inspired by the techniques used in [Fagin et al., 2005a] and [Fuxman et al., 2006]. In the case of full \( \langle \text{CQ}, \text{CQ}\rangle \)-dependencies, the chase procedure is again used to prove the tractability of the problem for UCQ\^[1] queries, and the lower bound for UCQ\^[2] queries is proved by reducing 3-CNF-SAT to the complement of our problem. It is important to mention that the tractability for the case of monotone queries requires of some techniques tailored to them, since this class of queries has a semantic definition, as opposed to the other query languages in this article that have syntactic definitions. Notice that UCQ\^[#] queries are monotone, and thus the tractability for this class is obtained from the tractability for monotone queries. Finally, if we consider FO queries, then the problem becomes undecidable. To prove this result we consider the embedding problem for finite semigroups, which is an undecidable problem [Evans, 1951; 1953; 1978; Gurevich, 1966] that has as input a finite partial semigroup, and asks whether it can be embedded to a finite semigroup. More precisely, we prove the undecidability for the case of FO by constructing a fixed mapping \( M \) and a fixed FO query \( Q \), and then providing a reduction from the embedding problem for finite semigroups to the complement of
CERTAINANSWERS(\(M, Q\)).

In the case of combined complexity, the proofs are similar. We just would like to point out here that for UCQ\(^{-}[1]\) queries, the upper bound follows directly from the techniques used in the study of the data complexity for the same class, but not fixing the mapping and the query, while the EXPTIME-hardness is proved by providing a reduction from the evaluation problem for Database programs containing a single rule, which is an EXPTIME-complete problem [Gottlob and Papadimitriou, 2003; Kolaitis et al., 2006]. Besides, for the case of CQ\(^{-}[2]\) queries, the lower bound is shown by providing a reduction from the complement of the TILING problem defined in Section 4.

The computation of certain answers has been extensively studied for the case of st-tgds [Fagin et al., 2003; Arenas et al., 2004; Libkin, 2006; Hernich, 2013] and also for a setting including target-to-source dependencies [Fuxman et al., 2006]. The latter case could be considered as a generalization of our scenario, as a bidirectional constraint consists of an st-tgd and a target-to-source dependency. However, this derived target-to-source dependency may have constraints in the right-hand side, making our setting incomparable with the framework of [Fuxman et al., 2006] where only tuple-generating dependencies are considered. Besides this, neither the combined complexity of the query evaluation problem nor the case of non-monotone queries were considered in [Fuxman et al., 2006]. It is worth mentioning, though, that some of the techniques introduced in [Fuxman et al., 2006] proved to be useful in our case, and inspired some of our results.

5.1 LAV, GAV, and some tractable cases

As it can be seen in Tables 2 and 3, most of the results on the complexity of the query evaluation problem are negative. Thus, it is natural to ask what kind of conditions could be imposed in order to get tractability. In this section, we study the computation of certain answers in two restricted scenarios that have been widely studied in the data integration/exchange literature [Lenzerini, 2002], and have been extensively investigated for differential query answering [Fuxman et al., 2006].

Theorem 4. CERTAINANSWERS(\(M, Q\)) is in PTIME, for every mapping \(M\) specified by a set of true-GAV \(\{UCQ^n, CQ\}\)-dependencies and every query \(Q\) in FO.

As it can be seen in Tables 2 and 3, most of the results on the complexity of the query evaluation problem are negative. Thus, it is natural to ask what kind of conditions could be imposed in order to get tractability. In this section, we study the computation of certain answers in two restricted scenarios that have been widely studied in the data integration/exchange literature [Lenzerini, 2002] and have been widely used in practice, namely the local-as-view (LAV) and the global-as-view (GAV) scenarios.

A set \(\Delta\) of \(\{UCQ^n, CQ\}\)-dependencies is said to be LAV if the left-hand side of every dependency in \(\Delta\) is a conjunctive query mentioning a single relational atom. Similarly, \(\Delta\) is said to be GAV if the right-hand side of every dependency in \(\Delta\) is a single relational atom (not including existential quantifiers). A further restriction is to consider true-LAV and true-GAV dependencies [Arocena et al., 2010]. In true-LAV, the left-hand side of every dependency must be a conjunctive query mentioning a single relational atom which has no repeated variables, and every source relation must appear in a dependency. Similarly, in true-GAV, the right-hand side of every dependency must be a single relational atom with no repeated variables (and no existential quantifiers), and every target relation should appear in the right-hand side of a dependency.

Example 2. Dependency (3) in Section 1 is true-LAV, \(\Gamma = \{A(x, y) \land B(y) \rightarrow R(x, x, y)\}\) is GAV, but not true-GAV. The set \(\Delta\) in Example 1 is true-LAV and true-GAV.

Unfortunately, the LAV and true-LAV restrictions do not reduce the complexity of the query evaluation problem. More specifically, all the lower bounds in Table 2 hold for the case of true-LAV (and, thus, for the case of LAV). On the other hand, the query answering problem can be easily solved for the case of true-GAV. In fact, if \(M\) is defined by a set of true-GAV dependencies, then it can be proved that for every instance \(I\) such that \(\text{Sol}_{M}(I) \neq \emptyset\), it holds that \(\text{Sol}_{M}(I) = 1\). Hence, the certain answers can be obtained by just executing the input query over a single solution.

Proposition 3. CERTAINANSWERS(\(M, Q\)) is in PTIME, for every mapping \(M\) specified by a set of true-GAV \(\{UCQ^n, CQ\}\)-dependencies and every query \(Q\) in FO.

Theorem 4 is proved by providing a polynomial time algorithm for CERTAINANSWERS(\(M, Q\)) which uses a subroutine for the problem of verifying whether a truth assignment satisfies a propositional formula. Given that the latter can be solved in polynomial time, we conclude that CERTAINANSWERS(\(M, Q\)) can be solved in polynomial time.

6 Concluding Remarks and Future Work

We have considered the use of bidirectional constraints to specify schema mappings, showing their usefulness to define the closure of a schema mapping. We have studied two fundamental problems in data exchange (existence of solutions and query evaluation) for these new mapping specifications, obtaining a wide range of results for different scenarios.

Several issues deserve further investigation. First, some properties of the notion of closure of a mapping need to be investigated in more depth, in particular, whether the closure of a mapping is unique, and if this is not the case, what is the relationship between the different closures of a mapping. Second, a deeper study of the connection between our proposal and some existing semantics for non-monotone queries needs to be conducted, in particular in the data exchange context [Libkin, 2006; Hernich, 2013].

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References


