Conjunctive Query Containment over Trees^{*}

Henrik Björklund

Wim Martens

Thomas Schwentick

Technical University of Dortmund Germany

Abstract

The complexity of containment and satisfiability of conjunctive queries over finite, unranked, labeled trees is studied with respect to the axes *Child*, *NextSibling*, their transitive and reflexive closures, and *Following*. For the containment problem a trichotomy is presented, classifying the problems as in PTIME, coNP-complete, or Π_2^P -complete. For the satisfiability problem most problems are classified as either in PTIME or NP-complete.

1 Introduction

Conjunctive query containment for relational databases is one of the most thoroughly investigated problems in database theory. It is known to be essentially equivalent to conjunctive query evaluation and to Constraint Satisfaction in AI [11]. From the database point of view, the importance of conjunctive queries on relational structures lies in the fact that they are the most widely used queries in practice. More precisely, they correspond to the select-fromwhere queries from SQL that only use "and" as a Boolean connective.

Recently, conjunctive queries have also been studied over tree structures [9]. It is somewhat surprising that they have not been studied earlier, as they arise very naturally in various settings, such as data extraction and integration, computational linguistics, and dominance constraints [9]. Moreover, unary and binary conjunctive queries over trees form a very natural fragment of XPath

^{*} This work was supported by the DFG Grant SCHW678/3-1. The present paper is the full version of reference [2], which appeared in the Symposium on Data Base Programming Languages 2007.

Email addresses: henrik.bjoerklund@udo.edu (Henrik Björklund), wim.martens@udo.edu (Wim Martens), thomas.schwentick@udo.edu (Thomas Schwentick).

	Child	$Child^+$	$Child^*$	NextSibling	$NextSibling^+$	$NextSibling^*$	Following
Child	in P	Π_2^P	Π_2^P	coNP	coNP	coNP	Π_2^P
$Child^+$		coNP	coNP	Π_2^P	Π_2^P	Π_2^P	Π_2^P
$Child^*$			coNP	Π_2^P	Π_2^P	Π_2^P	Π_2^P
NextSibling				in P	coNP	coNP	Π_2^P
$NextSibling^+$					coNP	coNP	Π_2^P
$NextSibling^*$						coNP	Π_2^P
Following							coNP

Table 1

Complexities of Conjunctive Query Containment.

2.0 [1], and therefore also of XQuery [4]. Indeed, unary and binary conjunctive queries over trees correspond to Core XPath without *negation* and *union* (see, e.g., [8]), but with *path intersection*, as introduced in XPath 2.0 (see, e.g., [10,14]). Gottlob et al. already showed that unary conjunctive queries over trees can be translated to XPath 1.0 queries, albeit with an exponential blow-up [9], and the above-mentioned Core XPath queries with path intersection can be translated into conjunctive queries by identifying variables. Hence, our complexity upper bounds transfer to positive Core XPath expressions with path intersection, but without union.

In this paper, we consider conjunctive query containment on trees. We mainly focus on Boolean containment of conjunctive queries, i.e., given two conjunctive queries P and Q, is $L(P) \subseteq L(Q)$, where L(P) (resp., L(Q)) denotes the set of trees on which P (resp., Q) has a non-empty output. Conjunctive query containment over trees is a problem that needs to be solved for conjunctive query optimization. The latter is, for instance, important for XQuery engines, but is also relevant in the other settings mentioned above. Moreover, conjunctive query *satisfiability*, which we also study and which is a simplified form of containment, needs to be solved if one wants to decide well-definedness for important XQuery fragments [15]. There is a further relevant setting in which the set of trees under consideration is restricted by a schema and the containment question is asked relative to this schema. We give a brief overview of our results.

Containment. We obtain a similar classification as Gottlob et al. [9]. The most essential differences are that the PTIME membership results for conjunctive query evaluation translate to coNP membership results for containment and that NP-completeness results for evaluation translate to Π_2^P -completeness results for containment. The former translation is easy to obtain due to a polynomial size witness property for counter-examples (Lemma 10). For the latter translation, we build on some of the NP lower bound reductions by Gottlob et

	Child	$Child^+$	$Child^*$	NextSibling	$NextSibling^+$	$NextSibling^*$	Following
Child	in P	NP [10]	NP	in P	in P	in P	NP
$Child^+$		in P	in P	?	?	?	?
$Child^*$			in P	?	?	?	?
NextSibling				in P	NP	NP	NP
$NextSibling^+$					in P	in P	in P
$NextSibling^*$						in P	in P
Following							in P

Table 2

Complexities of Conjunctive Query Satisfiability.

al. for our Π_2^P lower bound proofs. They had to be significantly adapted, however, as unlike in the relational setting, conjunctive query containment on trees cannot be reduced in a straightforward manner to conjunctive query evaluation on a canonical model. Most of our complexity results on conjunctive query containment are summarized in Table 1. From the above-mentioned polynomial size witness property and the results by Gottlob et al. [9], we can also conclude that containment is in coNP for the fragments CQ(*Child*, *NextSibling*, *NextSibling*^{*}, *NextSibling*⁺), CQ(*Child*^{*}, *Child*⁺), and CQ(*Following*). Combined with the results from the table, this gives us a complete trichotomy of the complexity of conjunctive query containment with respect to all subsets of the axes we consider.

Unfortunately, as we can see from the table, conjunctive query containment on trees is quite a hard problem. We only identify two tractable fragments, that is, CQ(NextSibling) and CQ(Child). For the latter fragment, PTIME membership is already non-trivial. All other combinations of axes are at least coNP-hard.

Satisfiability. Conjunctive query satisfiability can be seen as a simplification of the containment problem. Indeed, Q is satisfiable if and only if $L(Q) \not\subseteq L(\text{false})$. Our results on satisfiability are summarized in Table 2. Interestingly, we see here that the dichotomy drawn by the evaluation and the containment problem shifts. For the satisfiability problem, we obtain significantly more tractable fragments than for the containment problem. Some cases, however, still remain NP-hard.

We note that the NP lower bound for satisfiability of $CQ(Child, Child^+)$ was already obtained by Hidders [10]. We give an alternative proof (Theorem 31).

Related Work. Most of the related work has already been mentioned. We note, however, that conjunctive query containment has also been investigated

for object-oriented database systems [5]. In particular, it is shown that conjunctive query containment is Π_2^P -complete. The classes of conjunctive queries studied in [5] are, however, incomparable to ours.

2 Preliminaries

2.1 Trees

By Σ we always denote a fixed but infinite set of labels. For a finite set S, we denote by |S| the number of elements of S. The trees we consider are rooted, ordered, finite, labeled, unranked trees, which are directed from the root downwards. That is, we consider trees with a finite number of nodes and in which nodes can have arbitrarily many children. We view a tree t as a relational structure over a finite number of unary labeling relations $a(\cdot)$, where each $a \in \Sigma$, and binary relations $Child(\cdot, \cdot)$ and $NextSibling(\cdot, \cdot)$. Here, a(u) expresses that u is a node with label a, and Child(u, v) (respectively, NextSibling(u, v)) expresses that v is a child (respectively, next sibling) of u. We assume that each node in a tree bears precisely one label, i.e., for each u, there is precisely one $a \in \Sigma$ such that a(u) holds in t.

Notice that, in contrast to standard practice, we have an infinite set of labels from which our (finite) trees can choose. This reflects how trees occur in an XML-context: an XML tree is a finite structure, but there is no restriction on how it should be labeled (if no schema is provided).

In addition to *Child* and *NextSibling*, we use their transitive closures (denoted $Child^+$ and $NextSibling^+$) and their transitive and reflexive closures (denoted $Child^*$ and $NextSibling^*$). We also use the *Following*-relation, which is inspired by XPath [6] and defined as

 $Following(u, v) = \exists x \exists y Child^*(x, u) \land NextSibling^+(x, y) \land Child^*(y, v).$

We denote the set of nodes of a tree t by Nodes(t). We define the *size of* t, denoted by |t|, as the number of nodes of t. We refer to the above-mentioned binary relations as *axes*.

A tree t' is a *subtree* of a tree t if t' is a tree and a substructure of t. In other words, t' is connected and the relations in t' are subsets of the (*Child*, *NextSibling*) relations in t. Furthermore, the labels of nodes are the same in t and t'. Our definition of subtree implies that siblings in t' are also siblings in t. Sometimes, we do not want this restriction on subtrees. Therefore, for a set \mathcal{R} of axes, we say that t' is an \mathcal{R} -subtree of t if t' is a tree and, for each axis R

in \mathcal{R} , R(x, y) in t' implies that R(x, y) holds in t. Just as in normal subtrees, t' preserves the labels of t.

2.2 Conjunctive Queries

Let $X = \{x, y, z, ...\}$ be a set of variables. A conjunctive query (CQ) over alphabet Σ is a positive existential first-order formula without disjunction over a finite set of unary predicates a(x) where each $a \in \Sigma$, and the binary predicates *Child*, *Child*⁺, *Child*^{*}, *NextSibling*, *NextSibling*⁺, *NextSibling*^{*}, and *Following*. In this paper, we will mainly focus on Boolean satisfaction of conjunctive queries. We will therefore consider conjunctive queries without free variables. As our queries do not contain free variables, we often omit the existential quantifiers to simplify notation. For a conjunctive query Q, we denote the set of variables appearing in Q by Var(Q). We use $CQ(R_1, \ldots, R_k)$ or $CQ(\mathcal{R})$ (where $\mathcal{R} = \{R_1, \ldots, R_k\}$) to denote the fragment of CQs that uses only the unary alphabet predicates and the binary predicates R_1, \ldots, R_k . We use the terminology on valuations of a query and query graphs from Gottlob et al. [9].

Definition 1 Let Q be a conjunctive query, and t a tree. A valuation of Q on t is a total function θ : $Var(Q) \rightarrow Nodes(t)$. A valuation is a satisfaction if it satisfies the query, that is, if every atom of Q is satisfied by the assignment. A tree t models Q ($t \models Q$) if there is a satisfaction of Q on t. The language L(Q) of Q is the set of all trees that model Q.

We say that a tree t is a minimal model of Q if $t \models Q$ and the number of nodes in t is minimal among all trees in L(Q).

The following example illustrates a conjunctive query.

Example 2 Consider the conjunctive query $Q = Child^+(x_1, x_2) \wedge Child^+(x_2, x_4) \wedge Child^+(x_1, x_3) \wedge Child^+(x_3, x_4) \wedge a(x_1) \wedge b(x_2) \wedge c(x_3) \wedge d(x_4) \wedge e(x_5)$. For readability, we often represent queries graphically. For example, Figure 1 depicts query Q. For readability of the figures, we often omit the variable names of the queries in the figures. Any tree t that models Q must have an a-labeled node u with a d-labeled descendant v such that the path from u to v contains a b-labeled node and a c-labeled node (in arbitrary order). Moreover, t must contain an e-labeled node somewhere.

Definition 3 Let Q be a conjunctive query over Σ with variables Var(Q). The query graph Q is the directed multigraph $G_Q = (V, E)$ with edge labels and node labels such that V = Var(Q), node x is labeled a if and only if a(x)is an atom in Q; and E contains the labeled directed edge $x \xrightarrow{R} y$ if and only if R(x, y) is an atom in Q.



Fig. 1. Graphical representation of the query in Example 2. Double arrows represent $Child^+$ relations.

We assume familiarity with standard graph-related terminology such as reachability, connected components, etc. Subgraphs of G_Q correspond to subqueries of Q. We will sometimes slightly abuse the terminology by using graph-related concepts when talking about queries. Thus "variable x is reachable from variable y in Q" means that x is reachable from y in G_Q . Similarly, "maximal connected component of Q" means a subquery corresponding to a maximal connected component of G_Q .

Sometimes, we use the notation $R^i(x, y)$, where R is an axes and $i \in \mathbb{N}$. This means that y can be reached from x using i steps of R, and is shorthand for $R(x, x_1) \wedge R(x_1, x_2) \wedge \cdots \wedge R(x_{i-1}, y)$, where x_1, \ldots, x_{i-1} are variables that do not appear anywhere else in the query.

The following decision problems for conjunctive queries are the main topic of interest for this paper.

Definition 4 • Containment: Given two conjunctive queries P and Q, is $L(P) \subseteq L(Q)$?

• Satisfiability: Given a conjunctive query Q, is $L(Q) \neq \emptyset$?

The above problems are in a sense both instances of the containment problem. That is, satisfiability for Q is testing whether $L(Q) \not\subseteq L(\text{false})$.

For the containment problem, many of our algorithms will search for a tree t such that $t \in L(P) - L(Q)$. If $t \in L(P) - L(Q)$, we call t a *counterexample*. Similarly, for the satisfiability problem, we will often search for a tree $t \in L(Q)$, which we call a *witness*.

As mentioned above, we consider conjunctive queries without free variables. The result of evaluating such a query on a tree is therefore boolean. In general one can also consider k-ary conjunctive queries, i.e., CQs with k free variables, returning a k-ary relation when evaluated on a tree. For two k-ary queries P and Q, P is contained in Q if, for every tree t, the relation returned by P is a subset of the relation returned by Q. Using a result of Miklau and Suciu [12], this problem reduces to containment for Boolean queries for all fragments that include the *Child*-axis. For instance, consider the left query $P(x_1, x_2, x_3)$ in Figure 2. By introducing, for each free variable x_i , a new variable x'_i and adding the atoms $Child(x_i, x'_i) \wedge X_i(x'_i)$ to the query, where X_i



Fig. 2. How to reduce from k-ary queries to 0-ary queries.

is a new label, the query $P'(x_1, x_2, x_3)$, depicted on the right of Figure 2, is obtained. It is now easy¹ to see that, for two queries $P(\overline{x})$ and $Q(\overline{x})^2$ with k free variables, P is contained in Q if and only if $L(P') \subseteq L(Q')$, where P'and Q' are obtained by adding the atoms $Child(x_i, x'_i) \wedge X_i(x'_i)$ to P and Q, respectively. For satisfiability, it is of course immediate that the complexities are the same for 0-ary and k-ary queries.

2.3 Basic Properties

In this section we list a few basic properties of conjunctive queries which are quite well-known and easy to prove. We use them further on in our proofs. If t and t' are trees, h is a function from t to t', and \mathcal{R} is a set of binary relations, we say that h is an \mathcal{R} -homomorphism if h(u) is defined for every node u in t, a(u) in t implies a(h(u)) in t', for each $a \in \Sigma$, and R(u, v) holds in t implies that R(h(u), h(v)) holds in t', for each $R \in \mathcal{R}$.

Observation 5 Let t be a tree and let $Q \in CQ(\mathcal{R})$ be a query such that $t \models Q$. If t' is a tree and there exists an \mathcal{R} -homomorphism $h : t \to t'$, then $t' \models Q$.

Observation 6 Conjunctive queries are monotonous. More precisely, let Q be a $CQ(\mathcal{R})$ and let $t \models Q$. Then $t' \models Q$ for all trees t' for which t is an \mathcal{R} -subtree of t'.

For the next observation, we extend the notion of \mathcal{R} -homomorphisms to queries. That is, if P and Q are in $CQ(\mathcal{R})$, we say that $h : Var(P) \to Var(Q)$ is an \mathcal{R} -homomorphism from P to Q if h is total, a(x) in P implies that a(h(x)) in Q for each $a \in \Sigma$, and $R(x_1, x_2)$ in P implies that $R(h(x_1), h(x_2))$ occurs in Q, for each $R \in \mathcal{R}$.

Observation 7 Let P and Q be in $CQ(\mathcal{R})$. If there exists a homomorphism from Q to P, then $L(P) \subseteq L(Q)$.

As we will see in the proof of Theorem 9, the other direction of Observation 7 does not always hold.

¹ The proof is analogous to the one of Proposition 1 in [12].

² We can assume w.l.o.g. that the free variables are the same in P and Q.

3 Containment

When we investigate whether query P is contained in query Q, i.e., $L(P) \subseteq L(Q)$, we will always assume that the graph of Q has only one maximal connected component.

Observation 8 Let P and Q be CQs and let Q_1, \ldots, Q_k be the maximal connected components of Q. Then $L(P) \subseteq L(Q)$ if and only if $L(P) \subseteq L(Q_1) \cap \cdots \cap L(Q_k)$.

3.1 PTIME Upper Bounds.

Theorem 9 Containment is in PTIME for CQ(Child) and CQ(NextSibling).

PROOF. The proof for CQ(*NextSibling*) is straightforward. For testing whether $L(P) \subseteq L(Q)$, we first test that both queries are satisfiable. This can be done in polynomial time by Theorem 27. If P is unsatisfiable, containment trivially holds. If Q is unsatisfiable while P is satisfiable, containment fails. Next, we simplify both queries by applying the chase for the relation *NextSibling*(A, B) with functional dependencies $A \to B$ and $B \to A$. (For details on how this procedure works, see the proof of Theorem 27.) After this, none of the queries has variables $x \neq y \neq z$ such that both *NextSibling*(x, y) and *NextSibling*(x, z) or *NextSibling*(y, x) and *NextSibling*(z, x) are atoms. In other words, each query is a collection of *linear* maximal connected subqueries. Further more, by Observation 8, we can assume that Q has only one maximal connected component.

We now claim that containment holds if and only if there is a homomorphism from Q to P. Since the queries are linear, testing if there is a homomorphism can be done in polynomial time. If there is such a homomorphism, containment trivially holds. If not, we construct a counter-example tree $t \in L(P) - L(Q)$ as follows. Let P_1, \ldots, P_k be the maximal connected components of P. Each such P_i has variables $x_1^i, \ldots, x_{n_i}^i$, binary atoms $NextSibling(x_j^i, x_{j+1}^i)$ for each $j \in \{1, \ldots, n_i - 1\}$, and a number of unary atoms. With each P_i we associate a string S_i of length n_i . Position j of S_i has label $a \in \Sigma$ if $a(x_j^i)$ is an atom of P_i . If there is no such atom, position j gets label #, where $\# \in \Sigma$ is a symbol that occurs in neither P nor Q.

The tree t has levels $0, 1, \ldots, k$. On each level, except level k, there is exactly one node that has children—all the others are leaves. On level 0, there is only the root, which has label #. All nodes on level i, for $i \in \{1, \ldots, k\}$ are children of the sole non-leaf node on level i - 1. Level i has n_i nodes, which are labeled,



Fig. 4. Example for which $L(P) \subseteq L(Q)$, but there is no homomorphism from Q to P. Each arrow denotes a *Child*-axis.

from left to right, with the symbols of S_i . The construction of t is depicted in Figure 3.

Clearly, t satisfies P. Also, if there was a satisfaction for Q on t, this satisfaction would immediately give a homomorphism from Q to P.

The proof for CQ(Child) is considerably more involved. A naive algorithm would try to find an embedding of Q into P and accept iff it can be found. However, Figure 4 illustrates that not finding an embedding of Q into P does not imply that $L(P) \not\subseteq L(Q)$.

Let P and Q be two queries in CQ(Child). We want to decide whether $L(P) \subseteq L(Q)$. First, we check if the two queries are satisfiable. This can be done in polynomial time by Theorem 27. If at least one of the queries is not satisfiable, we already have our answer, so we assume in the remainder of the proof that both are satisfiable.



Fig. 3. The tree construction for CQ(*NextSibling*) containment.

Any satisfiable query P in CQ(*Child*) can be transformed in polynomial time into an equivalent one that is *tree-shaped*, i.e., such that there are no variables $x \neq y \neq z$ such that both *Child*(x, z) and *Child*(y, z) are atoms of the query. This is achieved by applying the chase procedure for the *Child*(A, B) relation with the functional dependency $A \rightarrow B$ (see the proof of Theorem 27). Thus we can assume that both P and Q are tree-shaped.

We now test containment of L(P) in L(Q) by performing a series of tests.

First, we test whether there is a homomorphism from Q to P. As we can assume that the queries are tree-shaped, this can be done in polynomial time; see, e.g., [12].

If there is a homomorphism, we can conclude that $L(P) \subseteq L(Q)$. If there is no embedding and P has only one maximal connected component, we can conclude that $L(P) \not\subseteq L(Q)$, since none of P's minimal models model Q. However, if there is no homomorphism from Q to P and P has more than one maximal connected component, it is still possible that $L(P) \subseteq L(Q)$ holds. An example is given in Figure 4.

We try to find a counter-example to containment, that is, a tree that satisfies P but not Q. As usual, by Observation 8, we can assume that Q has only one maximal connected component.

Since Q is tree-shaped, it has a unique root variable r_Q . Let C_1, \ldots, C_k be the subqueries of Q such that the root r_{C_j} of each C_j is a child of r_Q (i.e., $Child(r_Q, r_{C_j})$ is an atom of Q). Also, let P_1, \ldots, P_m be the maximal connected components of P, where the root variable of each P_i is r_{P_i} . If r_Q has a label (i.e., $a(r_Q)$ is an atom of Q for some $a \in \Sigma$), we can easily find a counterexample tree: a root labeled with a new symbol # which has the roots of minimal models for P_1, \ldots, P_m as children. Since there is no homomorphism from Q to P, in particular, there is no homomorphism from Q to P_i for any i. Thus we assume that r_Q has no label.

In the following, we will reason about what criteria a counter-exampel tree must satisfy, and try to construct one that does satisfy them. If we succeed with the construction, it is clear that containment fails. On the other hand, if we find that it is impossible to construct a tree that satisfies the criteria, containment holds.

Let $n = |\operatorname{Var}(Q)|$. For each $j \in \{0, \ldots, n-1\}$ and $i \in \{1, \ldots, m\}$, let P_i^j be the query obtained by adding new variables z_1, \ldots, z_j to P_i , each of them labeled by the new symbol #, adding the atoms $Child(z_l, z_{l+1})$ for $1 \leq l < j$, and $Child(z_j, r_{P_i})$; see Figure 5. In particular, $P_i^0 = P_i$.

Now, for each $1 \leq i \leq m$, we define v_i to be the largest



Fig. 5. P_i^j .

number smaller than n + 1 such that there is no homomorphism from Q to $P_i^{v_i}$. Notice that if there is no homomorphism fro Q to P_i^n , there is no homomorphism from Q to P_i^l for any $l \in \mathbb{N}$, since Q has n variables, is connected, and only uses

the Child axes.

The intuition behind the above construction is the following. If $v_i < n$, then there is a homomorphism from Q to $P_i^{v_i+1}$. This means that for any counterexample tree t in L(P) - L(Q), and any satisfaction θ of P on t, it must be the case that the distance from the root of t to $\theta(r_{P_i})$ is at most v_i . Indeed, since the label # doesn't occur in Q and there is a homomorphism from Q to $P_i^{v_i+1}$, any tree which has a path of length $v_i + 1$ above a model for P also satisfies Q. For some P_i , we may have $v_i = 0$. This means that there is a homomorphism from Q to P_i^1 . Using the above arguments, for any counter-example tree t and any satisfaction θ for P on t, the root variable r_{P_i} of P_i must be assigned to the root of t by θ . Let r be the number of maximal connected components P_i of P such that $v_i > 0$. Without loss of generality, we assume that $v_i > 0$ for all i in $\{1, \ldots, r\}$ while $v_j = 0$ for all j in $\{r + 1, \ldots, m\}$. If there are two components, P_{i_1} and P_{i_2} such that $i_1, i_2 > r$ and the roots of P_{i_1} and P_{i_2} have different labels, they cannot both be assigned to the root of a counter-example tree. Thus no witness tree exists, and we can conclude that containment holds.

Next, for each $1 \leq j \leq k$ and each $1 \leq i \leq r$, we define S_j^i to be the subset of $\{0, \ldots, v_i - 1\}$ such that for each $l \in S_j^i$, there is a homomorphism h from C_j to P_i^l such that $h(r_{C_j})$ is the root of variable of P_i^l . The intuition behind this definition is the following. Suppose t is a tree in L(P) and θ is a satisfaction for P on t such that the distance t to $\theta(r_{P_i})$ is d, and $d - 1 \in S_j^i$. Then there is a satisfaction θ_{C_j} for C_j on t such that $\theta_{C_j}(r_{C_j})$ is a child of the root of t.

Now, for each C_j $(1 \le j \le k)$ we will try to find out whether concentrating on C_j can help us find a counter-example tree. To be more specific, we will try to construct a tree t in L(P) such that there is no satisfaction for C_j on t that assigns the root variable of C_j to a child of the root of t. The rationale is that we will also ensure that there is no satisfaction for Q on t that assigns the root variable of Q to any other node than the root of t. Together, this means that if we can find such a t, then containment fails.

For each $1 \leq i \leq r$, we pick a number x_i from $\{0, \ldots, v_i - 1\} - S_j^i$. If, for some i, there is no such number, i.e., $\{0, \ldots, v_i - 1\} = S_j^i$, then we set $x_i = -1$.

Given these values x_1, \ldots, x_r we try to construct a witness tree t as follows. For each $1 \leq i \leq r$ such that $x_i \geq 0$, we place the root r_{P_i} of a minimal model for P_i at depth $x_i + 1$ (where 0 is the depth of the root). Between the root and r_{P_i} , we place a non-branching path of nodes labeled #. For the remaining maximal subqueries of P, i.e., the P_i such that $i \in \{r+1, \ldots, m\}$ or $x_i = -1$, we take a minimal model of P_i and identify its root with the root of t. This may cause a conflict, if two or more of these minimal models already have fixed and different labels. If this is the case, the test for C_j fails. Otherwise, if at least one of them has a fixed label a, the root of t gets label a. If not, the we give the root label #. This construction is depicted in Figure 6. Clearly, $t \in L(P)$. For each P_i such that $i \leq r$, the distance from the root to r_{P_i} is smaller than v_i . This means that it is impossible for Q to match along any of the branches from the root, i.e., the root r_Q of Q has to be matched at the root of t if it can be matched at all.

We now test whether $t \models Q$. If it doesn't, containment clearly fails, and our test for C_j was successful. If it does, let θ_Q be a satisfaction for Q on t. We



Fig. 6. The construction of the candidate counter-example tree for C_i .

know that $\theta_Q(r_Q)$ must be the root of t. Also, θ_Q must assign all variables of C_j completely within a subtree of t corresponding to the the minimal model of some P_i with i > r or such that $x_i = -1$. Otherwise, $\theta_Q(r_{C-j})$ would have to be a child of the root of t that corresponds to a minimal model of $P_i^{x_i}$, for some $i \leq r$ and $x_i \geq 0$. But we know that $x_i \notin S_j^i$ for any $i \leq r$. Thus there is no $P_i^{x_i}$ such that C_j can be matched in a corresponding minimal model. This means that C_j can always be matched one step away from the root in any possible witness tree, and our test for C_j failed.

If we go through all subqueries C_j of Q without being able to construct a witness tree, we argue that containment holds. Indeed, suppose there were a tree in L(P) - L(Q) and let θ be a satisfaction of P on t. We summarize why this is impossible:

- (1) For each $i \in \{1, \ldots, m\}$ the distance from the root of t to $\theta(r_{P_i})$ can be at most v_i . This is because there is a homomorphism from Q to $P_i^{v_i+1}$.
- (2) For every vector (d_1, \ldots, d_m) of distances from the root of t to $\theta(r_{P_1}), \ldots, \theta(r_{P_m})$ such that $d_i \leq v_i$ for all i, and every $j \in \{1, \ldots, k\}$ we know that there is a satisfaction for C_j on t that assigns r_{C_i} to a child of the root of t.
- (3) Since we have assumed that r_Q has no label in Q, there is nothing to stop r_Q from being assigned to the root of t.
- (4) Thus there is also a satisfaction for Q on t, which is a contradiction.

We first show that if CQ P is not contained in CQ Q, then there is a polynomial size witness tree.

Lemma 10 Let P and Q be conjunctive queries. If $L(P) \not\subseteq L(Q)$ then there exists a tree t such that $t \models P$, $t \not\models Q$, and $|t| \le 2 \cdot |Var(P)| \cdot (|Var(Q)| + 4)$.

PROOF. Let t be a tree such that $t \models P$ and $t \not\models Q$. Let θ be a satisfaction of P on t, and let $T = \{\theta(x) | x \in \operatorname{Var}(P)\}$. Further, let S be the set of nodes v of t such that v is the least common ancestor of some nonempty subset of T. Now we remove all nodes from t that are not located on a path between two nodes in S. Thus we obtain a new tree t'. Clearly, $t' \models P$, θ is a satisfaction of P on t, and, by Observation 6, $t' \not\models Q$. Notice that $|S| < 2 \cdot |\operatorname{Var}(P)|$, and that t' only branches at nodes in S.

Now suppose that $|t'| > 2 \cdot |Var(P)| \cdot (|Var(Q)| + 4)$. We will prove that we can obtain a tree t_{small} in L(P) - L(Q) which is smaller than t'. There is a pair u, v of nodes in t' from S such that

- (1) u is an ancestor of v;
- (2) the path ρ from u to v has length at least |Var(Q)| + 3; and
- (3) no internal node on ρ belongs to S.

We now show how we obtain t_{small} from t'. First, we change the label of every node in Nodes(t') - T to a new Σ -symbol # that does not appear in Q. Call the obtained tree $t_{\#}$. Notice that such a Σ -symbol always exists because Qonly makes use of a finite subset of our infinite labeling alphabet. This clearly preserves satisfaction of P and non-satisfaction of Q. Next, we remove the parent of v from t', by making v a child of its grandparent, thereby obtaining tree t_{small} . We next show that t_{small} is indeed the tree we are looking for.

First of all, notice that t_{small} still models P, as θ is still a satisfaction of P on t_{small} . Furthermore, towards a contradiction, suppose there is a satisfaction θ_Q for Q on t_{small} . As the length of the path ρ' from u to v in t_{small} is at least $|\operatorname{Var}(Q)| + 2$, there is at least one interior node w of ρ' such that no variable of Q is assigned to w by θ_Q . Partition $\operatorname{Var}(Q)$ into the set Y of variables assigned by θ_Q to nodes of the subtree of t_{small} rooted at w, and the set X of those that are not. Then Q cannot contain a predicate Child(x, y) for any variables $x \in X$ and $y \in Y$. Now we can insert a node labeled # between w and its child, obtaining a tree isomorphic to $t'_{\#}$. It is straightforward to verify that θ_Q is a satisfaction for Q on this new tree, and thus also on $t'_{\#}$. This is a contradiction.



Fig. 7. Structure of query P in the proof of Theorem 12.

The above process can be repeated until we have a witness tree of size at most $2 \cdot |\operatorname{Var}(P)| \cdot (|\operatorname{Var}(Q)| + 4)$.

The above lemma puts conjunctive query containment in Π_2^P . Indeed, for testing whether $L(P) \not\subseteq L(Q)$, the algorithm would guess a tree t_{small} of size at most $2 \cdot |\operatorname{Var}(P)| \cdot (|\operatorname{Var}(Q)| + 4)$, test in NP whether $t_{\text{small}} \models P$ and test in coNP whether $t_{\text{small}} \not\models Q$. As Gottlob et al. showed that conjunctive query evaluation is in PTIME for CQ(*Child*, *NextSibling*, *NextSibling*^{*}, *NextSibling*⁺), CQ(*Child*^{*}, *Child*⁺), and CQ(*Following*) [9], the above algorithm gives us a coNP upper bound for containment for these fragments. We can therefore state the following theorem.

Theorem 11 (1) Containment is in Π_2^P for CQs.

(2) Containment is in coNP for CQ(Child^{*}, Child⁺), CQ(Following), and CQ(Child, NextSibling, NextSibling^{*}, NextSibling⁺).

3.3 coNP Lower Bounds.

For the coNP lower bounds, we will either reduce from the complement of the SHORTEST COMMON SUPERSEQUENCE problem; or from the SHORT-EST COMMON SUPERSTRING problem, both of which are known to be NPcomplete [13,7]. The SHORTEST COMMON SUPERSEQUENCE (respectively, SHORTEST COMMON SUPERSTRING) problem asks, given a set of strings S, and an integer k, whether there exists a string of length at most k which is a supersequence (respectively, superstring) of each string in S. Here, s is a supersequence of s_0 if s_0 can by obtained by deleting symbols from s, and sis a superstring of s_0 if s_0 can be obtained by deleting a prefix and a postfix of s.

Theorem 12 Containment is coNP-hard for $CQ(NextSibling^+)$, $CQ(NextSibling^*)$, $CQ(Child^+)$, $CQ(Child^*)$, and CQ(Following).

PROOF. All cases are proved by a reduction from the complement of SHORT-EST COMMON SUPERSEQUENCE. To this end, let S and k be an instance of SHORTEST COMMON SUPERSEQUENCE. We now define conjunctive queries *P* and *Q* such that $P \not\subseteq Q$ if and only if there exists a shortest common supersequence for *S* of length at most *k*. Let $S = \{s_1, \ldots, s_m\}$ where, for each $i = 1, \ldots, m, s_i = a_i^1 \cdots a_i^{n_i}$. Let # be a symbol not occurring in any string in *S*.

We first show how the proof works for $NextSibling^+$. The query P is defined as in Figure 7, where each arrow represents a $NextSibling^+$ -axis and # and each a_i^j is a Σ -symbol. The query Q now essentially states that each tree must have a string of siblings with at least k + 1 + 2 different nodes. Formally, we define Q as

 $NextSibling^+(x_1, x_2) \land \dots \land NextSibling^+(x_{k+2}, x_{k+3}).$

It is not difficult to see that $P \not\subseteq Q$ if and only if there exists a shortest common supersequence for S of length at most k. The proofs for $Child^+$ and *Following* are completely analogous. For $Child^*$ and $NextSibling^*$, we need to insert dummy #-symbols between all a_i^j labels in P, and adapt the query Q accordingly.

The proof of the next theorem is along the same lines as the previous one, but this time we reduce from the SHORTEST COMMON SUPERSTRING problem. The essential difference is that P now does not contain the leftmost and rightmost #-labeled symbol in Figure 7, the arrows in Figure 7 now denote *NextSibling*-axes, and that all the a_j^i -labeled nodes are connected to a common parent by *Child*-axes.

Theorem 13 Containment is coNP-hard for CQ(Child, NextSibling).

3.4 Π_2^P Lower Bounds

The Π_2^P lower bounds in this section will all be obtained by a reduction from $\forall \exists$ positive 1-in-3 SAT, which is formally defined as follows. A set C_1, \ldots, C_m of clauses is given, each of which has three Boolean variables from $\{x_1, \ldots, x_{n_x}\} \uplus \{y_1, \ldots, y_{n_y}\}$. No variable is negated. The question is whether, for every truth assignment for $\{x_1, \ldots, x_{n_x}\}$, there exists a truth assignment for $\{y_1, \ldots, y_{n_y}\}$ such that each C_i contains precisely one true variable.

The proof of the following lemma is analogous to a standard proof showing that positive 1-in-3 SAT is NP-complete.

Lemma 14 $\forall \exists$ positive 1-in-3 SAT is Π_2^P -complete.

PROOF. Membership of the problem in Π_2^P is trivial. For Π_2^P -hardness, we

reduce from $\forall \exists 3SAT$. First, we convert a $\forall \exists 3SAT$ formula ϕ into a $\forall \exists 1\text{-in-}3$ SAT formula ϕ' . Second, we show how to get rid of negative literals.

Let $C = (x \lor y \lor z)$ be a clause of ϕ (here, x, y, z are literals, not variables). We introduce six new existentially quantified variables, a, b, c, d, e, f, to simulate C. To do this, we introduce the new clauses $(x \lor a \lor d), (y \lor b \lor d), (a \lor b \lor e), (c \lor d \lor f)$, and $(z \lor c)$. It is easy to verify that there is an assignment of truth values to the new variables that makes exactly one literal per clause true if and only if at least one of the literals x, y, z is true.

We show next how to make all literals positive. For each variable x that appears both positively and negatively, we replace all occurrences of $\neg x$ with a new existentially quantified variable \bar{x} , and add the clause $(x \lor \bar{x})$. This makes sure that exactly one of x and \bar{x} is assigned true.

Finally, we show how to ensure that each clause contains exactly three literals. Thereto, suppose that we have a clause $(x \lor y)$. We introduce four new existentially quantified variables f, a, b, c and rewrite $(x \lor y)$ as $(x \lor y \lor f)$, $(f \lor a \lor b)$, $(f \lor b \lor c)$, and $(a \lor b \lor c)$. The intuition is that f can never be chosen to be true and that, if f is false, we can choose b to be true.

Theorem 15 Containment is Π_2^P -complete for $CQ(Child, Child^+)$ and $CQ(Child, Child^*)$.

PROOF. We present a proof for $CQ(Child, Child^+)$ and discuss in the end how to adapt it for $CQ(Child, Child^*)$.

The proof is an adaptation of a proof by Gottlob et al., showing that the query complexity of evaluation for CQ(*Child*, *Child*⁺) is NP-hard [9]. We reduce from $\forall \exists$ positive 1-in-3 SAT, which is Π_2^P -complete according to Lemma 14.

For the readability of this proof, we will first assume that each tree node can carry multiple labels. We explain at the end of the proof how it can be modified to work for the standard definition of labeled trees, where each node has only one label.

Let $\forall \overline{x} \exists \overline{y}C_1, \ldots, C_m$ be an instance of $\forall \exists$ positive 1-in-3 SAT, where $\overline{x} = \{x_1, \ldots, x_{n_x}\}$ and $\overline{y} = \{y_1, \ldots, y_{n_y}\}$. We may assume that no clause contains a particular literal more than once. Let Φ denote the formula

$$\forall \overline{x} \exists \overline{y} C_1, \ldots, C_m, C_{m+1}, \ldots C_{m+n_x}.$$

Here, for each $i = 1, ..., n_x$, C_{m+i} denotes the clause (y'_i, x_i, y''_i) , where y'_i and y''_i are new existentially quantified variables. It is easy to see that there is a



Fig. 8. Illustration of the definition of query P in the proof of Theorem 15.

 $\forall \exists$ 1-in-3 SAT solution for the original formula if and only if there is one for Φ .

Let query P be defined as in Figure 8, where single lines represent the *Child* axis, double lines represent the *Child*⁺axis, the symbols inside the nodes are variables of P and the symbols to the left of nodes are the Σ -symbols.

For the query Q, we introduce variables a_i, b_i for each $i = 1, \ldots, m + n_x$ and in addition a variable $c_{k,l,i,j}$ whenever the k-th literal of C_i coincides with the *l*-th literal of C_j $(1 \le j \le m + n_x, i \ne j, 1 \le k, l \le 3)$.

The query Q consists of the following atoms:

- for each $i = 1, \ldots, m + n_x$, $A(a_i) \wedge B(b_i) \wedge Child^3(a_i, b_i)$;
- for each variable $c_{k,l,i,j}$, $L_k(c_{k,l,i,j}) \wedge Child^+(b_i, c_{k,l,i,j}) \wedge Child^{8+k+l}(a_j, c_{k,l,i,j})$; and,
- for each $i = m + 1, ..., m + n_x, X_{i-m}(a_i)$.

Before we show that the reduction is correct, we start with an observation. Consider the set of minimal models of P. It is easy to see that this set is not empty, and every minimal model of P has the shape of the tree in Figure 8 with the addition that, for every $i = 1, ..., n_x$, at least one of the nodes v_1, v_2, v_3 is labeled with X_i . Let T_P be the subset of the minimal models such that, for each X_i , precisely one of v_1, v_2, v_3 is labeled X_i . We refer to T_P as the set of intended models.

The following observation is immediate from the monotonicity of conjunctive queries (Observation 6) and the fact that each $t \in L(P)$ has a {*Child*}-subtree in T_P .³

Observation 16 The following statements are equivalent:

- $\forall t_P \in T_P : t_P \models Q$
- $\forall t \in L(P) : t \models Q.$

We show that the reduction is correct; that is,

$$\forall \overline{x} \exists \overline{y} C_1, \dots, C_m \Leftrightarrow L(P) \subseteq L(Q).$$

 (\Rightarrow) Assume that, for every truth assignment $\sigma_x : \{x_1, \ldots, x_{n_x}\} \to \{\text{true, false}\},$ there exists a truth assignment $\sigma_y : \{y_1, \ldots, y_{n_y}\} \to \{\text{true, false}\}$ such that each clause $C_i, 1 \leq i \leq m$, contains precisely one true literal under σ_x and σ_y . We show that $t_P \models Q$ for every $t_P \in T_P$. According to Observation 16, this implies that $L(P) \subseteq L(Q)$.

Let t_P be an arbitrary, but fixed, tree in T_P . Then there exists a satisfaction θ_P of P on t_P . From θ_P , we define a truth assignment $\sigma_{t_P} : \{x_1, \ldots, x_{n_x}\} \rightarrow \{\text{true, false}\}$ as follows:

- if $\theta_P(v_2)$ is labeled X_i , then we set $\sigma_{t_P}(x_i) = \text{true}$;
- otherwise, we set $\sigma_{t_P}(x_i) = \text{false.}$

By definition of P, σ_{t_P} assigns a truth value to every x_i , $1 \le i \le n_x$. Hence, there exists a $\sigma_y : \{y_1, \ldots, y_{n_y}\} \to \{\text{true, false}\}\$ such that each clause C_i , $1 \le i \le m$, contains precisely one true literal under σ_{t_P} and σ_y . From σ_y , we now construct a truth assignment $\sigma'_y : \{y_1, \ldots, y_{n_y}, y'_1, \ldots, y'_{n_x}, y''_1, \ldots, y''_{n_x}\} \to \{\text{true, false}\}\$ as follows:

³ Recall the definition of \mathcal{R} -subtrees from Section 2.

- for each $i = 1, \ldots, n_y, \sigma'_y(y_i) = \sigma_y(y_i);$
- if $\theta_P(v_2)$ is labeled X_i , then we set $\sigma'_y(y'_i) = \sigma'_y(y''_i) =$ false;
- otherwise, if $\theta_P(v_1)$ is labeled X_i , then we set $\sigma'_y(y'_i) =$ true and $\sigma'_y(y''_i) =$ false;
- otherwise, we set $\sigma'_y(y''_i)$ = true and $\sigma'_y(y'_i)$ = false.

It is easy to see that each clause C_1, \ldots, C_m contains precisely one true literal under σ_{t_P} and σ_y if and only if each clause $C_1, \ldots, C_m, C_{m+1}, \ldots, C_{m+n_x}$ contains precisely one true literal under σ_{t_P} and σ'_y .

We will show how σ_{t_P} and σ'_y induce a satisfaction θ_Q of Q on t_P . Let σ : {1,..., $m + n_x$ } \rightarrow {1,2,3} be defined as $\sigma(i) = k'$ if and only if the k'-th literal in C_i is true under σ_{t_P} and σ'_y . Notice that σ is total and well-defined. We first define a valuation θ_Q of Q on t_P and then show that all query atoms are satisfied. We set

- $\theta_Q(a_i) = \theta_P(v_{\sigma(i)})$ for each $i = 1, \dots, m + n_x$;
- $\theta_Q(b_i) = \theta_P(w_{\sigma(i),\sigma(i)})$ for each $i = 1, \dots, m + n_x$; and
- $\theta_Q(c_{k,l,i,j}) = \theta_P(w_{\sigma(i),5+k-l+\sigma(j)})$ for each variable $c_{k,l,i,j}$.

We now prove that θ_Q is a satisfaction of Q on t_P . Our choice of θ_Q implies that the variables a_i and b_i are mapped to nodes with labels A and B, respectively. Furthermore, $\theta_Q(b_i) = \theta_P(w_{\sigma(i),\sigma(i)})$ can be reached from $\theta_Q(a_i) = \theta_P(v_{\sigma(i)})$ with three child-steps. For every variable of the form $c_{k,l,i,j}$, we know that $\theta_Q(c_{k,l,i,j}) = \theta_P(w_{\sigma(i),5+k-l+\sigma(j)})$ is always a descendant of $\theta_P(w_{\sigma(i),\sigma(i)})$. If $\sigma(i) \neq k$, then $\theta_Q(c_{k,l,i,j}) = \theta_P(w_{\sigma(i),5+k-l+\sigma(j)})$ has label L_k because $4 \leq 5+k-l+\sigma(j) \leq 10$ and the nodes $\theta_P(w_{\sigma(i),4}), \ldots, \theta_P(w_{\sigma(i),10})$ all have (at least) the two labels $L_{k'}$ for which $\sigma(i) \neq k'$. If $\sigma(i) = k$, then $\sigma(j) = l$. By going 8+k-l steps downward from $\theta_P(v_{\sigma(j)})$, passing through $\theta_P(w_{k,k})$, we reach node $\theta_P(w_{k,5+k})$, which has label L_k . Since $\theta_Q(c_{k,l,i,j}) = \theta_P(w_{\sigma(i),5+k-l+\sigma(j)}) =$ $\theta_P(w_{k,5+k})$, the query atoms $Child^{8+k+l}(a_j, c_{k,l,i,j})$ are satisfied. For each i = $m+1, \ldots, m+n_x$, we have that $\sigma(i) = k$ if and only if $\theta_P(v_k)$ is labeled X_{i-m} . Hence, for each $i = m + 1, \ldots, m + n_x$, $\theta_Q(a_i) = \theta_P(v_{\sigma(i)})$ is labeled X_{i-m} .

(\Leftarrow) Assume that $t_P \models Q$ for every $t_P \in T_P$. We show that, for each truth assignment $\sigma_x : \{x_1, \ldots, x_{n_x}\} \to \{\text{true, false}\}$, there exists a truth assignment $\sigma_y : \{y_1, \ldots, y_{n_y}\} \to \{\text{true, false}\}$ such that each clause $C_i, 1 \leq i \leq m$, contains precisely one true literal under σ_x and σ_y .

Let $\sigma_x : \{x_1, \ldots, x_{n_x}\} \to \{\text{true, false}\}\ \text{be a truth assignment. We define the tree } t_x \text{ as the tree implied by the variables and$ *Child* $-axes in in Figure 8 with the additions that, for each <math>i = 1, \ldots, n_x$,

- if $\sigma_x(x_i) = \text{true}$, then only v_2 is labeled X_i ; and
- if $\sigma_x(x_i)$ = false, then only v_1 is labeled X_i .

Obviously, t_x is in T_P and therefore t_x models P. Hence, $t_x \models Q$.

Let θ be a satisfaction of Q on t_x . We show that θ induces a truth assignment $\sigma_y : \{y_1, \ldots, y_{n_y}\} \to \{\text{true, false}\}\$ such that each clause $C_i, 1 \leq i \leq m$ contains precisely one true literal under σ_x and σ_y . We first show that θ induces a truth assignment $\sigma'_y : \{y_1, \ldots, y_{n_y}, y'_1, \ldots, y'_{n_x}, y''_1, \ldots, y''_{n_x}\} \to \{\text{true, false}\}\$ such that each clause $C_i, 1 \leq i \leq m + n_x$ contains precisely one true literal under σ_x and σ'_y .

To this end, if $\theta(a_i) = v_k$, we interpret this as the k-th literal of clause C_i being chosen to be true. Obviously, under any valuation of Q on t_x , we select precisely one literal from each clause C_i in this way. Because of the constructions of t_x , we know that the literal x_i is selected for clause C_{m+i} if and only if $\sigma_x(x_i) =$ true. We have to verify that if a literal L occurs in two clauses C_i and C_j and we select L in C_i , we also select L in C_j . Let L be the k-th literal of C_i and the l-th literal of C_j , and let $\theta(a_i) = v_k$ (i.e., L is selected in C_i). Then $\theta(c_{k,l,i,j}) = w_{k,5+k}$ because that is the only node below $\theta(b_i) = w_{k,k}$ that has label L_k . The query contains the atom $Child^{8+k-l}(a_j, c_{k,l,i,j})$ for variable $c_{k,l,i,j}$. From node $w_{k,5+k}$, by 8 + k - l upward steps we arrive at node v_l . Hence $\theta(a_j) = v_l$, and we select L from clause C_j .

The truth assignment σ_y we are looking for is σ'_y restricted to $\{y_1, \ldots, y_{n_y}\}$.

To conclude the proof, we discuss how to deal with the multiple node labels. The idea is to replace each variable z of P that has k labels by k + 1 variables $\{z_0, z_1, \ldots, z_k\}$. In the construction from Figure 8, z_0 takes the place of z, while each $z_i, 1 \leq i \leq k$ carries one of the k labels, and is required to be a child of z_0 (*Child* (z_0, z_i)). In query Q, the same transformation is then used.

Finally, we describe what changes have to be made for the proof to work in the $CQ(Child, Child^*)$ case. In the reduction, we replace each pair of atoms $Child^+(v_0, X_i), Child^+(X_i, w_{2,1})$ of P (for $1 \le i \le n_x$) with the pair $Child^*(v_1, X_i)$, $Child^*(X_i, v_3)$. In Q, we can simply replace $Child^+$ with $Child^*$. The correctness proof is then analogous. \Box

Theorem 17 Containment is Π_2^P -hard for CQ(Child, Following).

PROOF. We adapt the proof of Theorem 15 by simulating $Child^+$ with Childand Following. To this end, we begin by equipping each of the variables u in query P defined in Figure 8 that has an outgoing $Child^+$ -axes by two "dummy" children z_1 and z_2 . These new variables are used nowhere else, and get a new Σ label # that doesn't appear in the queries P and Q of the proof of Theorem 15. Now, whenever $Child^+(u, v)$ is used in one of the queries, we can replace it by

$$Child(u, z_1) \land Child(u, z_2) \land Following(z_1, v) \land Following(v, z_2).$$



(b) Fragment X.

Fig. 9. Definition of query P in the proof of Theorem 18.

It is now enough to note that all variables in the queries P and Q that have no specified label are required by the queries to have children. Thus none of them can bind to a node in one of the minimal models of the modified P query that is labeled by #. \Box



Fig. 10. Gadgets for the definition of query Q in the proof of Theorem 18.

Theorem 18 Containment is Π_2^P -hard for $CQ(Child^+, Following)$ and for $CQ(Child^*, Following)$.

PROOF. We first explain the proof for $CQ(Child^+, Following)$ and argue later that it works analogously for $CQ(Child^*, Following)$. Let $\forall \overline{x} \exists \overline{y} C_1, \ldots, C_m$ be

an instance of $\forall \exists$ positive 1-in-3 SAT. Let $\overline{x} = \{x_1, \ldots, x_{n_x}\}$ and let $\overline{y} = \{y_1, \ldots, y_{n_y}\}$. We can assume that no clause contains a particular literal more than once.

We construct two queries, P and Q, over the labeling alphabet $\{A, B, C, D, D_{\text{true}}, D_{\text{false}}, L_1, L_2, L_3, X_1, \ldots, X_{n_x}, W, Z\}$ such that $L(P) \subseteq L(Q)$ if and only if $\forall \overline{x} \exists \overline{y} C_1, \ldots, C_m$ has a solution. The current proof builds further on a proof by Gottlob et al. that shows that the query complexity of evaluation for $CQ(Child^+, Following)$ is NP-hard (Theorem 5.2 in [9]).

The construction of query P is illustrated in Figure 9. Here, every double-lined edge represents a $Child^+$ -axis and every directed edge represents a Followingaxis. Figure 10 depicts the gadgets from which query Q will be constructed. For improved readability, we adopt the terminology of the proof by Gottlob et al. That is, we will refer to the nodes labeled L_1 , L_2 , and L_3 in the 1-in-3 gadget from Figure 10(a) by v_1 , v_2 , and v_3 , respectively. Moreover, we annotate the query fragment T in Figure 9(a) with numbers from 1 to 7. We call the node 1 (resp., 3, 6) the topmost position of variable v_1 (resp., v_2 , v_3).

Let t_{\min} be a minimal model of fragment T from Figure 9(a). That is, t_{\min} is essentially shaped as the structure given by the $Child^+$ axes in T. Gottlob et al. show that the following observation holds.

Observation 19 ([9]) Every satisfaction θ of the 1-in-3 gadget on t_{min} maps exactly one of the variables v_1, v_2 , and v_3 to its topmost position.

Given a clause C, we interpret a satisfaction θ in which variable v_k is mapped to its topmost position as the selection of the k-th literal from C to be true. Hence, the 1-in-3 gadget would ensure that, on t_{\min} , exactly one variable of clause C is selected to be true.

We now define the query P as in Figure 9(c). That is, P contains two copies of the fragment T, followed by a copy of the X-fragment from Figure 9(b). The ordering between the subqueries of P is enforced by *Following*-axes: the root of T's left copy has a *Following*-axis to the root of T's right copy, and the root of T's right copy has a *Following*-axis to the root of the X-fragment.

Intuitively, the purposes of the different parts of the query P are as follows. The left copy of the T-fragment in P, together with the 1-in-3 gadget, is used to verify that the truth assignments we consider for \overline{x} and \overline{y} actually make one literal per clause of $\forall \overline{x} \exists \overline{y} C_1, \ldots, C_m$ true. The second copy of T in P is needed to ensure consistency of variable assignments between clauses: if we pick a variable to be true in one clause, that variable must be true in all clauses. Finally, the fragment X is used in P to generate all possible truth assignments for the \overline{x} -variables. Roughly, we interpret x_i as "true" if X_i can be reached from the W-labeled node with a *Following*-step, and as "false"

$k \backslash l$	1	2	3
1	10	13	18
2	5	8	13
3	2	5	10

Table 3 The function NAND(k, l) [9].

otherwise (see Figure 9(b)). For example, all X_i -labeled descendants of the D_{true} node are interpreted as "true", and all X_i -labeled ancestors of the lower D_{false} node are interpreted as "false".

The query Q is defined much like the query in the proof of Gottlob et al., with the essential difference that we have to transfer the variable assignment that is generated in the X-fragment of P to the matching of L_1 , L_2 , and L_3 of the 1-in-3 gadget of Q onto the subtrees that satisfy the two copies of Tin P. This will be taken care of by the X-assignment gadgets in Q, which are illustrated in Figure 10(b).

Formally, query Q is defined as follows. Each clause C_i is represented by two copies of the 1-in-3 gadget of Figure 10(a), a left copy Q_i and a right copy Q'_i . The two sets of subqueries $Q_1, \ldots, Q_m, Q'_1, \ldots, Q'_m$ are connected as follows. Consider the function NAND(k, l) in Table 3, as defined by Gottlob et al. In a left and right copy of the tree t_{\min} that would match the left and right copy of T in P, we can enforce that two variables, x and y, labeled L_k and L_l in their respective subqueries in Q, cannot both match the topmost node labeled L_k , respectively L_l , in the left, respective right, copy of t_{\min} by adding an atom of the form $Following^{\text{NAND}(k,l)}(x, y)$ to the query Q. To see this, we exemplify the case where k = l = 1. Observe that, from the top L_1 node in the left copy of T in Q, one can reach the upper L_1 node in the right copy of T with 9 Following-steps, but not with 10. The lower L_1 node, however, can be reached with 10 Following-steps. Hence, NAND(1, 1) = 10. The other cases are analogous.

So, for each pair of clauses C_i , C_j , variable $x \in Var(Q)$ such that Q_i contains the atom $L_k(x)$, and variable $y \in Var(Q)$ such that Q'_j contains the atom $L_l(y)$, if

- the k-th literal of C_i also occurs in C_j and
- the k-th literal of C_i and the l-th literal of C_j are different,

then we add an atom $Following^{\text{NAND}(k,l)}(x, y)$ to Q. As in the proof by Gottlob et al., these query atoms make sure that if a literal is chosen to be true in one clause, it is chosen to be true in other clauses as well; and that both copies Q_i and Q'_i of the query gadget of each clause make the same choice of selected literal.

Finally, we need to make sure that the assignment to the universally quantified variables from \overline{x} defined by a minimal model of P is respected. If Q_i contains the unary atom $L_k(x)$ and the k-th literal of C_i is a (universally quantified) variable x_l from \overline{x} , then we add a copy of the gadget varX(k, l) to the query, in which we identify the L_k -labeled node with the query variable x.

Intuitively, the gadget varX(k, l) ensures that if x_l is the k-th literal of Q_i , then Q_i picks the value for x_l that is generated by the tree. We explain this more formally. First, observe that, if t_P is a minimal model of P, then the label X_l occurs precisely once in t_P . (Because, if X_l occurs multiple times, t_P is not minimal.) Next, we need to define our *intended minimal models*. A minimal model t_P of P is an intended minimal model if

- (1) the D_{true} -labeled node is a child of the lower D_{false} -labeled node;
- (2) the W-labeled node is a child of the lower D_{false} -labeled node; and
- (3) the three rightmost Z-labeled nodes are children of the D_{true} -labeled node.

Figure 11 contains an intended (left) and a non-intended minimal model (right) of the X-fragment of P.



Fig. 11. Minimal models of the X-fragment of P.

Let t_P be an intended minimal model of P. We say that t_P picks x_l to be true if the X_l -labeled node can be reached with a Following-step from the W-labeled node in t_P (i.e., if it is a descendant of D_{true}), and we say that t_P picks x_l to be false otherwise (i.e., if it is an ancestor of the lower D_{false} -labeled node). We can now make the following observation: **Observation 20** Let t_P be an intended minimal model of P. Then, for every satisfaction θ of Q on t_P , the following holds. If the k-th literal of C_i is a universally quantified variable x_l , then θ selects the k-th literal x_l of C_i to be true on t_P if and only if t_P picks x_l to be true.

PROOF. Observation 20 can be easily verified by testing the possible homomorphisms from the X-variable gadgets of Q (Figure 10(b)) to the query P (Figure 9). We provide a proof for one of the cases, the arguments for all the other cases are analogous. For the direction from left to right, say that θ chooses the first literal x_l of C_i to be true on t_P . Then θ also matches the L_1 -labeled node of the leftmost X-variable gadget in Figure 10(b) to the upper L_1 -labeled node in the first subtree of t_P . From here, the W-labeled node in t_P can be reached by 20 Following steps, but not by 21. This means that θ must match the X_l -labeled node as a descendant of D_{true} . Also note that the upper L_1 -labeled node in the first subtree of t_P has a descendant (the left A-labeled child) from which the Z-descendant of X_l can be reached with 23 Following steps.

For the direction from right to left, say that θ chooses the first literal x_l of C_i to be false on t_P . Then θ also matches the L_1 -labeled node of the leftmost X-variable gadget in Figure 10(b) to the lower L_1 -labeled node in the first subtree of t_P . From here, the W-labeled node in t_P can be reached by 21 *Following* steps, so in principle we can still match X_l everywhere. However, the X-variable gadget also requires that the L_1 -labeled node has a descendant (which can only be its A-labeled child in t_P), from which we can reach a Z-labeled descendant of X_l with 23 *Following* steps. This is only possible if the X_l -labeled node occurs as an ancestor of D_{true} , which means that t_P chooses x_l to be false. This concludes the proof of Observation 20.

This concludes the reduction for Theorem 18. We proceed to proving that the reduction is also correct. That is, we show that

$$\forall \overline{x} \exists \overline{y} C_1, \dots, C_m \Leftrightarrow L(P) \subseteq L(Q).$$

(\Leftarrow) Suppose that $L(P) \subseteq L(Q)$. We show that, for each truth assignment σ_x : $\{x_1, \ldots, x_{n_x}\} \rightarrow \{\text{true, false}\}, \text{there exists a truth assignment } \sigma_y : \{y_1, \ldots, y_{n_y}\} \rightarrow \{\text{true, false}\} \text{ such that each clause } C_i, 1 \leq i \leq m \text{ contains precisely one true literal under } \sigma_x \text{ and } \sigma_y.$

Let $\sigma_x : \{x_1, \ldots, x_{n_x}\} \to \{\text{true, false}\}\ \text{be an arbitrary but fixed truth assignment.}$ ment. We define an intended minimal model t_x of P as follows. The root of t_x has three children. The first and second child correspond to the left and right T-subquery of P, respectively. For the two copies of subquery T in P, t_x has a child relation for every occurrence of a descendant relation in T, and the ordering of the nodes in t_x is given by the *Following*-relations in P. For the fragment X, it is slightly more complicated. The third subtree of t_x has $8 + n_x$ nodes, corresponding to the D_{true} , two D_{false} , the W, the four Z, and the n_x X_i -labeled nodes in Figure 9(b). The structure of the subtree is given by conditions (1)–(3) of an intended minimal model. Because of condition (1), this leaves two possibilities for the X_i -labeled nodes to occur: an X_i -labeled node either occurs between the two D_{false} -labeled nodes (we call this area X^{false}), or it can occur as a descendant of the D_{true} -labeled node (we call this area X^{false}).

The correspondence between σ_x and the third subtree of t_x is now encoded as follows:

- if $\sigma_x(x_i) = \text{true}$, then the label X_i occurs in X^{true} and not in X^{false} ; and
- if $\sigma_x(x_i) = \text{false}$, then the label X_i occurs in X^{false} and not in X^{true} .

Obviously, t_x is an intended minimal model of P. As we assumed that $L(P) \subseteq L(Q)$, we have that $t_x \models Q$.

Let θ be a satisfaction of Q on t_x . We show that θ induces a truth assignment $\sigma_y : \{y_1, \ldots, y_{n_y}\} \to \{\text{true, false}\}\$ such that each clause $C_i, 1 \leq i \leq m$ contains precisely one true literal under σ_x and σ_y .

To this end, if x is an L_k -labeled variable of Q_i , the k-th literal of C_i is existentially quantified, and $\theta(x)$ is the topmost position of L_k in t_x , we interpret this as the k-th literal of clause C_i being chosen to be true by σ_y . We argue that σ_y is indeed the truth assignment we are looking for.

As argued in the construction of Q, in every valuation of Q on t_x , the 1-in-3 gadgets select precisely one literal from each clause C_i in this way. Furthermore, the *Following*^{NAND(k,l)} atoms ensure that if a literal L occurs in two clauses C_i and C_j and we select L in C_i , then we also select L in C_j . Finally, the varX(k, l) gadgets ensure that θ picks the same values for the $x_i \in \overline{x}$ as t_x (Observation 20), and therefore also σ_x . Hence, the existence of θ implies the existence of a valuation σ_y such that each clause C_i , $1 \leq i \leq m$ contains precisely one true literal under σ_x and σ_y .

 (\Rightarrow) Assume that, for every truth assignment $\sigma_x : \{x_1, \ldots, x_{n_x}\} \to \{\text{true, false}\},$ there exists a truth assignment $\sigma_y : \{y_1, \ldots, y_{n_y}\} \to \{\text{true, false}\}$ such that each clause $C_i, 1 \leq i \leq m$ contains precisely one true literal under σ_x and σ_y .

We show that $L(P) \subseteq L(Q)$. To this end, let T_P be the set of minimal models of P, including non-intended minimal models. Analogously as in the proof of Theorem 15, we make the following observation.

Observation 21 The following are equivalent:

- $\forall t_P \in T_P : t_P \models Q$
- $\forall t \in L(P) : t \models Q.$

The observation follows from Observation 6, as each tree in L(P) has a $\{Child^+, Following\}$ -subtree in T_P .

We show that $t_P \models Q$ for every $t_P \in T_P$. According to Observation 21, this implies that $L(P) \subseteq L(Q)$.

Given $t_P \in T_P$, we define a truth assignment $\sigma_{t_P} : \{x_1, \ldots, x_{n_x}\} \to \{\text{true}, \text{false}\}$ as follows:

- if the X_i -labeled node can be reached from the W-labeled node by a Following-step in t_P , then we set $\sigma_{t_P}(x_i) = \text{true};$
- otherwise, we set $\sigma_{t_P}(x_i) = \text{false.}$

By definition of P, σ_{t_P} assigns a truth value to every x_i , $1 \leq i \leq n_x$. Hence, there exists a $\sigma_y : \{y_1, \ldots, y_{n_y}\} \to \{\text{true, false}\}$ such that each clause C_i , $1 \leq i \leq m$, contains precisely one true literal under σ_{t_P} and σ_y .

We will show how σ_{t_P} and σ_y induce a satisfaction θ of Q on t_P . Let τ : {1,...,m} \rightarrow {1,2,3} be defined as $\tau(i) = k$ if and only if the k-th literal in C_i is true under σ_{t_P} and σ_y . Notice that τ is total and well-defined. We first define a valuation θ of Q on t_P and then show that all query atoms are satisfied. Let Q_i be a 1-in-3 gadget of Q and let v_1, v_2 , and v_3 be the nodes labeled L_1, L_2 , and L_3 in Q_i , respectively. By 1–7 we denote the nodes in t_P that correspond to the nodes 1–7 in the left copy of T in P. We set

- $\theta(v_1) = 1, \ \theta(v_2) = 4, \ \theta(v_3) = 7 \text{ if } \tau(i) = 1;$
- $\theta(v_1) = 2, \ \theta(v_2) = 3, \ \theta(v_3) = 7$ if $\tau(i) = 2$; and
- $\theta(v_1) = 2, \ \theta(v_2) = 5, \ \theta(v_3) = 6 \text{ if } \tau(i) = 3.$

By definition of Q_i , θ can be extended to a valuation of Q_i on t_P for each *i*. We define the valuation of Q'_i on the second subtree of t_P completely analogously. By definition, the *Following*^{NAND(k,l)} atoms connecting Q_i and Q'_i are also satisfied. It only remains to show that the gadgets varX(k, l) can be satisfied.

We argue that these gadgets can be satisfied by matching each X_l -labeled node in varX(k, l) onto the unique occurrence of X_l in t_P . Thereto, let z_1, z_2, z_3, z_4 be the nodes in t_P that correspond to the four Z-labeled nodes in fragment X, from left to right. If X_l is reachable from the W-node with a Following-step, then $\sigma_{t_P}(x_l) =$ true. This means that θ also selects x_l to be true. To satisfy the X-variable gadgets, we can now always map the Z-labeled node in the gadgets to z_1 which is always a descendant of X_l (see also Figure 11). If X_l is not reachable from the W-node with a Following step a descendant of D_{true} , then we can always map the Z-labeled node in the gadgets to z_4 (see also Figure 11).

This concludes the proof for $CQ(Child^+, Following)$. The proof for $CQ(Child^*, Following)$ is completely analogous. The reason is that, for each occurrence of $Child^+(x, y)$ in P, either x and y bear different alphabet labels, or x has a descendant z with a different alphabet label, from which y can be reached with a *Following*-axis. Hence, y can never be matched to the same node as x. \Box

As *Following* can be defined in terms of $Child^*$ and $NextSibling^+$, we immediately have the following corollary.

Corollary 22 Containment is Π_2^P -hard for $CQ(Child^*, NextSibling^+)$.

Theorem 23 Containment is Π_2^P -hard for

- (1) $CQ(Child^*, NextSibling),$ (4) $CQ(Child^+, NextSibling^+),$ and
- (2) $CQ(Child^*, NextSibling^*),$ (5) $CQ(Child^+, NextSibling^*).$
- (3) CQ(Child⁺,NextSibling),

PROOF. For each of these fragments, the proof of Theorem 18 can be adapted by the same methods as in the article by Gottlob et al. [9]. For the fragments (2)-(5), we also need to adapt the query P, such that P accepts trees in which the T-fragments have the shape from the proof by Gottlob et al. This is, however, straightforward for each of the fragments.

Theorem 24 Containment is Π_2^P -hard for CQ(Following, NextSibling).

PROOF. Unfortunately, the arguments we use in Theorem 23 do not work seamlessly for *Following* and *NextSibling*^{α}, where $\alpha \in \{1, +, *\}$. Even though we can express that, e.g., y must be a descendant of x by the formula

 $NextSibling(x_1, x) \land NextSibling(x, x_2) \land Following(x_1, y) \land Following(y, x_2)$

and by giving x and y different labels, the extra introduced nodes x_1 and x_2 for this encoding introduce difficulties for the X_i -labeled nodes of the P-query in the proof of Theorem 23. We therefore need to take a slightly different approach.

Figures 12 and 13 illustrate how to change P and Q in the proof of Theorem 18. Here, every solid arrow denotes a *NextSibling* axis, every dotted arrow denotes a Following axis, and every double line from x to y (where x is above and y is below) denotes the gadget Descendant(x, y) =

 $NextSibling(x_1, x) \land NextSibling(x, x_2) \land Following(x_1, y) \land Following(y, x_2),$

where x_1 and x_2 are the variables left and right from x in Figure 12, respectively. It is easy to see that Descendant(x, y) expresses that $Child^+(x, y)$ must hold in all cases: either x and y are labeled differently, or one of their siblings is labeled differently.

Furthermore, the placement of the X_i -labeled nodes is different from their placement for Theorem 18. Here, the idea is that the X_i labeled nodes are either descendants of the Z-labeled descendant of the R_2 -labeled node, or descendants of one of the two rightmost Z-labeled right siblings of the R_2 labeled node.

Given that a double line denotes the above *Descendant* gadget, the 1-in-3 gadget of Q are almost the same as in Figure 10(a). The only difference is that the L_1 , L_2 , and L_3 labeled nodes need extra left and right siblings in the gadget to express the descendant relation. This is illustrated in Figure 13(a). The gadgets Q_i and Q'_i are then wired in precisely the same manner as in the proof of Theorem 18.

The most significant change in the Q-query is in the gadgets for the X-variables. How to adapt these gadgets is illustrated in Figure 13(b). The remainder of the proof is analogous to the proof of Theorem 18.

Theorem 25 Containment is Π_2^P -hard for $CQ(Following, NextSibling^+)$ and $CQ(Following, NextSibling^*)$.

PROOF. The reduction for CQ(*Following*, *NextSibling*⁺) is analogous to the one in Theorem 24, i.e., we can replace every *NextSibling* in the proof of Theorem 24 with a *NextSibling*⁺. The reduction of Theorem 24 can be adapted to a reduction for CQ(*Following*, *NextSibling*^{*}) by replacing every *NextSibling*(x, y) in P with *NextSibling*^{*}(x, xy) $\wedge H(xy) \wedge NextSibling^*(xy, y)$, where H is a new label; replacing every *NextSibling*(x, y) in Q with *NextSibling*^{*}(x, y); and by changing the number of *Following*-steps in the X-gadgets of Q.

Consider a varX(i, j)-gadget from the proof of Theorem 24. We observe that in an intended minimal model of P, if we take k minimal following steps from a node labeled L_i , we will actually be following a *NextSibling*-axes k - 4 + itimes. Thus, after the modification of P, where each *NextSibling*-axes has been doubled, we will need 2(k - 6 + i) + 6 - i = 2k - 6 + i Following steps to reach the corresponding node. For l, the corresponding new number is 2k - 5 + i.



Fig. 12. Definition of query P in the proof of Theorem 24.



(b) New X-gadgets. For varX(i, j) the values of k and ℓ are 25 and 28 for i = 1, 20 and 25 for i = 2, and 17 and 20 for i = 3, respectively.

Fig. 13. How to adapt Q for Theorem 24.

This means, that in the new varX(i, j)-gadgets (see Figure 13(b)), when i = 1 we get k = 45 and l = 52. For i = 2 we get k = 26 and l = 47, while the numbers for i = 3 are k = 31 and l = 38.

The correctness proofs for both cases are obtained through Observation 5. \Box

4 Satisfiability

We first note that a conjunctive query Q is satisfiable if and only if all its maximal connected components are satisfiable. We therefore assume in our proofs that Q has only one maximal connected component.

Proposition 26 Satisfiability for CQs is in NP.

PROOF. It is easy to see that if a CQ is satisfiable, then it is satisfiable in a linear size tree. Indeed, let Q be a CQ and let t be a tree satisfying Q under valuation θ . Now let t' be a tree that

- contains the set T of nodes of t onto which variables are matched by θ ;
- contains, for each nonempty $S \subseteq T$, the least common ancestor of the nodes in S;
- contains no other nodes; and
- preserves the descendant relation and document order (i.e., depth-first-left-to-right order) from t.

It is easy to see that t' contains less than $2 \cdot |Var(Q)|$ nodes and that t' models Q. Thus we can guess this tree, guess a valuation for Q on t', and verify in polynomial time that the valuation is actually a satisfaction, i.e., that all atoms of Q are satisfied. \Box

4.1 PTIME Upper Bounds

Theorem 27 Satisfiability is in PTIME for CQ(Child) and CQ(NextSibling).

PROOF. First, we apply the chase on the relations in Q, i.e., we compute equivalence classes [x] of variables such that [x] is the maximal set of variables such that for any $t \in L(Q)$ and any satisfaction θ for Q on t, we must have $\theta(y) = \theta(x)$ for all $y \in [x]$.

For $Q \in CQ(Child)$, we start with one class for each variable in Var(Q), and iteratively merge classes [x] and [y] if there are $x' \in [x]$, $y' \in [y]$, and a variable z such that both Child(x', z) and Child(y', z) are atoms of Q. For $Q \in CQ(NextSibling)$ we do the same, with the addition that we also merge classes [x] and [y] if there are $x' \in [x]$, $y' \in [y]$, and z such NextSibling(z, x')and NextSibling(z, y') are both atoms of Q.

Once we have computed the equivalence classes, i.e., when no more classes can be merged using the rules above, we rewrite Q, obtaining a new query Q'. This is done by creating a new variable for each class and replacing each occurance of a variable in Q with the variable representing it's class. It should be clear that Q' is satisfiable if and only if Q is satisfiable.

Our new query Q' is satisfiable if and only if the following two conditions are met.

- (1) There is no variable x in Q' that has two labels, i.e., there is no x such that both a(x) and b(x) are atoms of Q', with $a \neq b$.
- (2) There are no cycles in Q', i.e., the query graph of Q' is acyclic.

Each of these conditions can be tested in polynomial time.

Before we state the next theorem, we introduce the concept of a siblinghood, which will be useful in our next two proofs.

Definition 28 In a tree t, a siblinghood is a subset S of Nodes(t) such that all nodes in S have the same parent, i.e., there is a node $u \in Nodes(t)$ such that Child(u, v) holds for all $v \in S$.

Theorem 29 Satisfiability is in PTIME for $CQ(NextSibling^+, NextSibling^*, Following)$ and $CQ(Child^+, Child^*)$.

PROOF. We start by checking for cycles. If the query graph of Q has a cycle on which at least one edge is labeled by $NextSibling^+$, Following, or $Child^+$, then Q is unsatisfiable. Unlike in the proof of Theorem 27, however, a query may have cycles of $Child^*(resp., NextSibling^*)$ axes and still be satisfiable. On such cycles, there can be no variables x, y such that a(x) and b(y) are atoms, for $a \neq b$. If there is, Q is unsatisfiable. Allowed cycles, i.e., those consisting of only $Child^*(resp. NextSibling^*)$ axes and without multiple labels, can be removed by identifying all variables on the cycle. In the remainder of the proof, we assume that the query is cycle free.

For CQ(NextSibling⁺,NextSibling^{*},Following), we argue that if Q is satisfiable, then there is a tree t and a satisfaction θ for Q on t such that θ assigns all variables of Q to nodes of t that belong to the same siblinghood. As a first step, we note that if Q is satisfiable, then Q', obtained by replacing all NextSibling^{*}-atoms of Q by NextSibling⁺-atoms is also satisfiable. Indeed, if θ is a satisfaction of Q on tree t, NextSibling^{*}(x, y) is an atom of Q, and $\theta(x) = \theta(y)$, we can modify t by inserting a new node between $\theta(x)$ and its left sibling (or at the beginning of the siblinghood if there is no left sibling), and modify θ by assigning x to the new node. After doing this for all such pairs x, y, the modified θ is a satisfaction of both Q and Q'. Next, we note that any acyclic query Q in $CQ(NextSibling^+, Following)$ induces a strict partial ordering on the variables. A topological sorting according to this partial ordering gives us a string of variables such that if $NextSibling^+(x, y)$ or Following(x, y) is an atom of Q, then x appears before y in the string. From such a string it is easy to construct a tree with a siblinghood that satisfies Q. This shows that any $Q \in CQ(NextSibling^+, NextSibling^*, Following)$ that passes the acyclicity tests at the beginning of this proof is satisfiable.

For CQ(*Child*⁺, *Child*^{*}) we use the same arguments as for CQ(*NextSibling*⁺, *NextSibling*^{*}, *Following*), except that instead of a siblinghood we use a unary tree, i.e., a tree that does not branch. \Box

Theorem 30 Satisfiability is in PTIME for CQ(Child, NextSibling) and $CQ(Child, NextSibling^+, NextSibling^*)$.

PROOF. For a conjunctive query Q in either of the classes above, we let Q_{ns} be the subquery obtained by removing all *Child*-atoms from Q. Similarly, let Q_c be the subquery obtained by removing all *NextSibling*^{α}-atoms, for $\alpha \in \{1, +, *\}$. We note that if variables x and y belong to the same maximal connected component of Q_{ns} , then, for any tree $t \in L(Q)$, any satisfaction for Q on t has to assign x and y to nodes that belong to the same siblinghood of t.

We first present an algorithm for checking satisfiability of queries Q in CQ(*Child*, *NextSibling*). If the query graph of Q has cycles, it is always unsatisfiable. Thus we assume that Q is acyclic.

In the description of the algorithm, we make use of a copy P of Q, which will be modified by the algorithm. Actually we can see P as being defined on equivalence classes [x] of variables, which the algorithm sometimes merges. At the beginning, P thus has one singleton class [x], for each variable $x \in Var(Q)$.

The algorithm first iterates over the following three steps, and stops when no merges occurred in the last iteration.

- (1) For each pair [x], [y], check whether there exist [z], [z'] that belong to the same connected component of P_{ns} and such that Child([x], [z]) and Child([y], [z']) are atoms of P. If this is the case, try to merge [x] and [y]. This try fails if there are $x \in [x]$ and $y \in [y]$ such that a(x) and b(y) are atoms of Q, for $a \neq b$. If the check fails, P is unsatisfiable.
- (2) For each maximal connected component of $P_{\rm ns}$, check satisfiability as in the proof of Theorem 27. When this procedure merges classes of variables, carry these merges over to $P_{\rm ns}$, $P_{\rm c}$, and P.
- (3) For each maximal connected component of P_c , check satisfiability as in the proof of Theorem 27. When this procedure merges classes of variables,

carry these merges over to $P_{\rm ns}$, $P_{\rm c}$, and $P_{\rm c}$.

If the iteration stops without reporting unsatisfiability, the algorithm performs one extra test. This is an acyclicity test on an extended query graph G_P^+ of P, namely the graph where *Child*-edges are, as usual, considered directed, while *NextSibling*-edges are considered undirected (or, equivalently, can be traversed in both directions). In G_P^+ , we test whether there is a cycle that uses at least one *Child*-edge. If this is the case, P is unsatisfiable. Indeed, if t is a tree and θ a valuation for P on t such that $\theta([y])$ is a child of $\theta([x])$, then $\theta([x])$ can never be reached from $\theta([y])$ by taking any number of *Child*- or *NextSibling*-steps in t.

Notice that the steps of the iteration above only try to merge variables that always have to be assigned to the same tree node by any satisfaction for P. They report unsatisfiability if such a merge fails or if a merge has introduced cycles. This immediately implies that Q is unsatisfiable as well.

If all tests above succeed, we claim that P is satisfiable.

- i Since step (1) of the iteration cannot merge any more classes, we know that for each connected component C of P_{ns} , there is *at most one* variable [x] of P that can have child axes to variables in C.
- ii Since step (2) cannot merge any more classes, we know that each connected component of $P_{\rm ns}$ is string-shaped, that is, forms a non-branching sequence of variable classes, connected by *NextSibling*-axes.
- iii Since step (3) cannot merge any more classes, we know that $P_{\rm c}$ is forest-shaped.
- iv We also know that no two variable classes that belong to the same connected component of $P_{\rm ns}$ are connected via *Child*-axes.

Let $C = \{C_1, \ldots, C_k\}$ be the maximal connected components of P_{ns} . We define the relation \prec on $C \times C$ by $C_i \prec C_j$ if there are variables $[x] \in Var(C_i)$ and $[y] \in Var(C_j)$ such that Child([x], [y]) is an atom of P. We argue that the directed graph $G_{\prec} = (C, \prec)$ of \prec is a forest. To do this, we must show that G_{\prec} has no cycles, and that there are no $i \neq j \neq k$ such that $C_i \prec C_k$ and $C_j \prec C_k$.

A cycle in G_{\prec} would immediately imply a cycle in G_P^+ containing at least one child axis, which the algorithm has already tested for. Thus G_{\prec} is acyclic.

Suppose there are $i \neq j \neq k$ such that $C_i \prec C_k$ and $C_j \prec C_k$. Then there must be $[x] \in Var(C_i), [y] \in Var(C_j)$, and $[z], [z'] \in Var(C_k)$ such that both Child([x], [z]) and Child([y], [z']) are atoms of P. This is ruled out by (i) above, and thus a contradiction.

Given this knowledge, we can construct a witness tree t and an accompanying satisfaction θ_P as follows. For each maximal connected component C_i of $P_{\rm ns}$, we construct a siblinghood S_i modeling the component, and let θ_P assign variables to nodes in the straightforward way. This is always possible by (ii).

For each pair of variables [x], [y] such that Child([x], [y]) is an atom of P, we add a child edge from $\theta_P([x])$ to each node in the siblinghood $\theta_P([y])$ belongs to. Since we know that G_{\prec} is a forest, the resulting structure is a forest. To complete the construction, we add a new root node, and connect it to the root of each tree in the forest. It immediately follows that all Childand NextSibling-atoms are satisfied. Thus θ_P is a satisfaction of P on t. It is straightforward to see that θ_Q , defined by $\theta_Q(x) = \theta_P([x])$ is a satisfaction of Q on t.

For $CQ(Child, NextSibling^+, NextSibling^*)$, the process is similar. The differences lie in steps (1) and (2) of the iteration. In (1), we allow merging variables that are connected with the NextSibling^{*}-axes, but not those connected by the NextSibling⁺-axes. In (2), satisfiability checking for $P_{\rm ns}$ is done as in the proof of Theorem 29. This means that after the iteration terminates, it is not necessarily the case that each connected component of $P_{\rm ns}$ is string-shaped. Each such component is, however, satisfiable, and we can, as argued in the proof of Theorem 29, find a string model for it by considering a topological sorting. \Box

4.2 NP lower bounds



Fig. 14. Gadget for the proof of Theorem 31.

Theorem 31 Satisfiability is NP-hard for

- (1) $CQ(Child, Child^+)$,
- (2) $CQ(Child, Child^*)$,

- (4) $CQ(NextSibling, NextSibling^*),$
- (5) CQ(NextSibling,Following), and
- (3) $CQ(NextSibling,NextSibling^+)$, (6) CQ(Child, Following).

PROOF. All reductions are from Shortest Common Supersequence. For cases (1)-(5), the reductions are very similar. Let let S and k be an instance of SHORTEST COMMON SUPERSEQUENCE. For each of the fragments, we define a conjunctive query P such that P is satisfiable if and only if there exists a shortest common supersequence for S of length at most k. Let $S = \{s_1, \ldots, s_m\}$

where, for each i = 1, ..., m, $s_i = a_i^1 \cdots a_i^{n_i}$. Let # be a symbol not occurring in any string in S.

The construction of P is depicted in Figure 14. The dotted arrows denote $Child^+$, $Child^*$, $NextSibling^+$, $NextSibling^*$, or Following-axes and the solid arrows denote Child or NextSibling axes, whichever are relevant for the fragment under consideration. The bulleted ("•") nodes represent unlabeled variables. The idea is that the path with the solid arrows contains 2k - 1 bulleted nodes. Hence, there exists a tree model for the query if and only if there exists a shortest common supersequence for S of length at most k.

For fragment (6), the above reduction does not work. It can be fixed, however, by using the same trick as in Theorem 17, i.e., replacing all occurrences of $Child^+(u, v)$ in the proof for CQ($Child, Child^+$) by

 $Child(u, z_1) \wedge Child(u, z_2) \wedge Following(z_1, v) \wedge Following(v, z_2).$

5 Conclusions

We have determined the complexity of the containment problem for all sets of axes built from *Child*, *NextSibling*, their transitive, respectively reflexive and transitive, closures, and *Following*. The complexity of the satisfiability problem was pinpointed for most sets, but the cases involving transitive closures of *Child* and *NextSibling* (which we believe will be quite similar) are still open.

All these results were obtained in a schema-less setting. Since XML processing is mostly done with respect to a schema, this is far from the complete picture. In a recent paper [3] we studied the containment, satisfiability, and validity problems for conjunctive queries with respect to schemas. It turns out that the presence of a schema dramatically increases the complexity. In particular containment of CQs with respect to DTDs is shown to be 2EXPTIME-complete.

References

- A. Berglund, S. Boag, D. Chamberlin, M. F. Fernández, M. Kay, J. Robie, and J. Siméon. XML Path Language (XPath) 2.0. Technical report, World Wide Web Consortium, January 2007. http://www.w3.org/TR/xpath20/.
- [2] H. Björklund, W. Martens, and T. Schwentick. Conjunctive query containment over trees. In 11th International Symposium on Database Programming Languages (DBPL), pages 66–80, 2007.

- [3] H. Björklund, W. Martens, and T. Schwentick. Optimizing conjunctive queries over trees using schema information. Manuscript, submitted for publication, 2008.
- [4] S. Boag, D. Chamberlin, M. F. Fernández, D. Florescu, J. Robie, and J. Siméon. Xquery 1.0: An XML query language. Technical report, World Wide Web Consortium, January 2007. http://www.w3.org/TR/xquery/.
- [5] E. P. F. Chan and R. van der Meyden. Containment and optimization of objectpreserving conjunctive queries. *Siam Journal on Computing*, 29(4):1371–1400, 2000.
- [6] J. Clark and S. DeRose. XML Path Language (XPath) version 1.0. Technical report, World Wide Web Consortium, 1999. http://www.w3.org/TR/xpath/.
- [7] J. Gallant, D. Maier, and J. A. Storer. On finding minimal length superstrings. Journal of Computer and System Sciences, 20(1):50–58, 1980.
- [8] G. Gottlob, C. Koch, R. Pichler, and L. Segoufin. The complexity of XPath query evaluation and XML typing. *Journal of the ACM*, 52(2):284–335, 2005.
- [9] G. Gottlob, C. Koch, and K. U. Schulz. Conjunctive queries over trees. Journal of the ACM, 53(2):238–272, 2006.
- [10] J. Hidders. Satisfiability of XPath expressions. In 9th International Workshop on Database Programming Languages (DBPL), pages 21–36, 2003.
- [11] P. G. Kolaitis and M. Y. Vardi. Conjunctive-query containment and constraint satisfaction. Journal of Computer and System Sciences, 61(2):302–332, 2000.
- [12] G. Miklau and D. Suciu. Containment and equivalence for a fragment of XPath. Journal of the ACM, 51(1):2–45, 2004.
- [13] K.J. Räihä and E. Ukkonen. The shortest common supersequence problem over binary alphabet is NP-complete. *Theoretical Computer Science*, 16(2):187–198, 1981.
- [14] B. ten Cate and C. Lutz. The complexity of query containment in expressive fragments of XPath 2.0. In 26th International Symposium on Principles of Database Systems (PODS), pages 73–82, 2007.
- [15] S. Vansummeren. On deciding well-definedness for query languages on trees. Journal of the ACM, 54(4), 2007.