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# Conjunctive query containment over trees

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#### 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  $\Pi_2^P$ -complete. For the satisfiability problem most problems are classified as either in PTIME or NP-complete. © 2010 Elsevier Inc. All rights reserved.

#### 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 [12]. 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-from-where queries from SQL that only use "and" as a Boolean connective.

Recently, conjunctive queries have also been studied over tree structures [10]. 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 [10]. Moreover, unary and binary conjunctive queries over trees form a very natural fragment of XPath 2.0 [2], and therefore also of XQuery [5]. Indeed, unary and binary conjunctive queries over trees are closely related to Core XPath without *negation* and *union* (see, e.g., [9]), but with *path intersection*, as introduced in XPath 2.0 (see, e.g., [11,16]). 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 [10], 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 nonempty 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 frag-

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**Table 1** Complexities of conjunctive query containment. All coNP and  $\Pi_2^P$  results are completeness results.

	Child	Child <sup>+</sup>	Child*	NextSibling	NextSibling <sup>+</sup>	NextSibling*	Following
Child	in P	$\Pi_2^P$	$\Pi_2^P$	coNP	coNP	coNP	$\Pi_2^P$
Child <sup>+</sup>		coNP	coNP	$\Pi_2^P$	$\Pi_2^P$	$\Pi_2^P$	$\Pi_2^P$
Child*			coNP	$\Pi_2^{P}$	$\Pi_2^P$	$\Pi_2^P$	$\Pi_2^P$
NextSibling				in P	coNP	coNP	$\Pi_2^P$
NextSibling <sup>+</sup>					coNP	coNP	$\Pi_2^P$
NextSibling*						coNP	$\Pi_2^P$
Following							coNP

**Table 2**Complexities of conjunctive query satisfiability. All NP-results are completeness results.

	Child	Child <sup>+</sup>	Child*	NextSibling	NextSibling <sup>+</sup>	NextSibling*	Following
Child	in P	NP [11]	NP	in P	in P	in P	NP
Child <sup>+</sup>		in P	in P	?	?	?	?
Child*			in P	?	?	?	?
NextSibling				in P	NP	NP	NP
NextSibling <sup>+</sup>					in P	in P	in P
NextSibling*						in P	in P
Following							in P

ments [17]. 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 [4]. We give a brief overview of our results.

**Containment.** We obtain a similar classification as Gottlob et al. [10]. 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  $\Pi_2^P$ -completeness results for containment. The former translation is easy to obtain due to a polynomial size witness property for counter-examples (Lemma 1). For the latter translation, we build on some of the NP lower bound reductions by Gottlob et al. for our  $\Pi_2^P$  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. [10], we can also conclude that containment is in  $\Pi_2^P$  for conjunctive queries using all axes, and that it is in coNP for the fragments using the axes {Child, NextSibling, NextSibling\*, NextSibling\*}, {Child\*, Child\*}, and {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, conjunctive queries using either only the *NextSibling*-axis, or only the *Child*-axis. 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  $L(Q) \nsubseteq L(\text{false})$ . Our results on satisfiability are summarized in Table 2. Interestingly, we see here that the dichotomy from the evaluation and containment problems shifts. For the satisfiability problem, we obtain significantly more tractable fragments than for the containment problem. Some cases, however, still remain NP-hard.

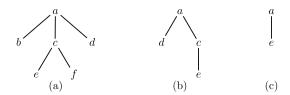
We note that the NP lower bound for satisfiability of conjunctive queries with {Child, Child+} was already obtained by Hidders [11]. We give an alternative proof (Theorem 23).

**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 [6]. In particular, it is shown that conjunctive query containment is  $\Pi_2^P$ -complete. The classes of conjunctive queries studied in [6] are, however, incomparable to ours. Also, Arenas and Libkin study data exchange for tree patterns, which form a subset of conjunctive queries, and also prove results for satisfiability w.r.t. a DTD [1]. This is an issue that we also studied in [4], but do not consider in the current paper.

#### 2. Preliminaries

#### 2.1. Trees

By  $\Sigma$  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.



**Fig. 1.** Examples of  $\mathcal{R}$ -subtrees.

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*<sup>+</sup> and *NextSibling*<sup>+</sup>) and their transitive and reflexive closures (denoted *Child*<sup>\*</sup> and *NextSibling*<sup>\*</sup>). We also use the *Following*-relation, which is inspired by XPath [7] 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*- and *NextSibling* relations in t. Furthermore, the labels of nodes are the same in t and t'. Our definition of subtree implies that, when u is a next sibling of v in a subtree t' of t, u also has to be a next sibling of v in t itself. As this is a quite severe restriction, we sometimes also want to work with a more flexible notion of 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 t in t' implies that t' is an t' in t' implies that t' in an t' subtree, we want all relations in t' to be preserved. The "standard" notion of subtree defined above corresponds with a {Child, NextSibling}-subtree. As an example of t'-subtrees, consider Fig. 1. Fig. 1(a) contains a tree t'. Fig. 1(b) contains a {Child}-subtree of t', and Fig. 1(c) a {Child}-subtree. Notice that siblings in the {Child}-subtree are also siblings in t', but their ordering can be different: the t'-labeled node can come before the t'-labeled node whereas they were ordered the other way around in t'.

#### 2.2. Conjunctive queries

Let  $X = \{x, y, z, \ldots\}$  be a set of variables. A *conjunctive query* (CQ) over alphabet  $\Sigma$  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,  $Child^+$ ,  $Child^+$ ,  $NextSibling^+$ 

**Definition 1.** Let Q be a conjunctive query, and t a tree. A *valuation* of Q on t is a total function  $\theta: Var(Q) \to Nodes(t)$ . 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 (denoted  $t \models Q$ ) 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). For readability, we often represent queries graphically. We often omit the variable names of the queries in the figures and only present their  $\Sigma$ -symbols. If a variable has no associated  $\Sigma$ -symbol, we denote this by a wildcard symbol "\*". The following example presents such an illustration of a conjunctive query.

**Example 1.** Consider the conjunctive query  $Q = Child^+(x_1, x_2) \wedge Child^+(x_2, x_4) \wedge Child^+(x_1, x_3) \wedge Child^+(x_3, x_4) \wedge Child^+(x_5, x_6) \wedge a(x_1) \wedge b(x_2) \wedge c(x_3) \wedge d(x_5) \wedge e(x_6)$ . Fig. 2(a) depicts query Q. Any tree t that models Q must have an a-labeled node u with a 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 d-labeled node with an e-labeled child somewhere. An example of such a tree is in Fig. 2(b).

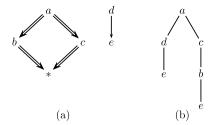


Fig. 2. Graphical representation of the query in Example 1 and a tree modeling it. Double arrows represent  $Child^+$  relations and single arrows represent Child relations.

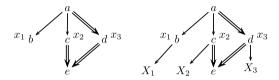


Fig. 3. How to reduce from k-ary queries to 0-ary queries.

**Definition 2.** Let Q be a conjunctive query over  $\Sigma$  with variables Var(Q). The query graph Q is the directed edge- and node-labeled multigraph  $G_Q = (N, E)$  with nodes N and edges E such that N = Var(Q), node X is labeled X if and only if X is an atom in X i

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 axis 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 3.

- 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  $L(Q) \nsubseteq L(\text{false})$ .

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*.

**Boolean versus** k-ary queries. 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 [13], 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 Fig. 3. 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  $X_i$  is a new label, the query  $P'(x_1, x_2, x_3)$ , depicted on the right of Fig. 3, is obtained. It is now easy 2 to see that, for two queries  $P(\bar{x})$  and  $Q(\bar{x})^3$  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.

 $<sup>^{2}</sup>$  The proof is analogous to the one of Proposition 1 in [13].

 $<sup>^3</sup>$  We can assume w.l.o.g. that the free variables are the same in P and Q.

## 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  $a \in \mathcal{L}$ , and a(u) holds in a(u) in a(u) in a(u) holds in a(u) hold

**Observation 1.** 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 2.** 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  $a \in \mathcal{L}$ , 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 3.** 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 6, the other direction of Observation 3 does not always hold.

#### 2.4. Chasing queries in CQ(Child) and CQ(NextSibling)

In some of our upper bound proofs using only *Child* or only *NextSibling*, will use *the chase*, which is a well-know technique in database theory. Let Q be a query in CQ(Child) or CQ(NextSibling). The idea behind the chase is that 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  $\theta$  for Q on t, we must have  $\theta(y) = \theta(x)$  for all  $y \in [x]$ . We do this as follows:

- 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 chase (Q), which is the result of applying the chase to Q. This is done by creating a new variable for each class and replacing each occurrence of a variable in Q with the variable representing its class. Notice that, in each of these cases, chase (Q) can be computed in polynomial time.

## 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 4.** 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  $L(P) \subseteq L(Q_i)$  for all  $i \in \{1, \ldots, k\}$ .

#### 3.1. PTIME upper bounds

**Theorem 5.** Containment is in PTIME for CQ(NextSibling).

**Proof.** For testing whether  $L(P) \subseteq L(Q)$ , where  $P, Q \in CQ(NextSibling)$ , we first test that both queries are satisfiable. This can be done in polynomial time by Theorem 20. If P is unsatisfiable, containment trivially holds. If Q is unsatisfiable while P is satisfiable, containment fails. As the chase can be computed in polynomial time for P and Q (see Section 2.4), we can assume that P = chase(P) and Q = chase(Q). Hence, each query is a collection of linear maximal connected subqueries, that is, there are no 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. Furthermore, by Observation 4, 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 counterexample tree  $t \in L(P) - L(Q)$  as follows. Let  $P_1, \ldots, P_k$  be the maximal connected

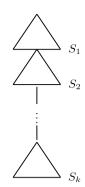
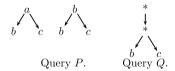


Fig. 4. The tree construction for CQ(NextSibling) containment.



**Fig. 5.** Example for which  $L(P) \subseteq L(Q)$ , but there is no homomorphism from Q to P. Each arrow denotes a *Child*-axis.

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 is depicted in Fig. 4 and has levels 0, 1, ..., 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  $i \in \{1, ..., k\}$  are children of the sole non-leaf node on level i - 1. Level i has  $n_i$  nodes, which are labeled, from left to right, with the symbols of  $S_i$ .

Clearly, t satisfies P. We claim that t does not satisfy Q. Suppose, towards a contradiction, that there would be a satisfaction of Q on t. Then this satisfaction would induce a homomorphism from Q to P, which we assumed not to be the case.  $\square$ 

#### **Theorem 6.** Containment is in PTIME for CQ(Child).

**Proof.** The proof for CQ(Child) is considerably more involved than the one for CQ(NextSibling). A naive algorithm would try to find a homomorphism of Q into P and accept if and only if it can be found. However, Fig. 5 illustrates that not finding a homomorphism from Q into P does not imply that  $L(P) \nsubseteq L(Q)$ . Indeed, there is no homomorphism from Q to P, but  $L(P) \subseteq L(Q)$ .

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 20. 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.

Any satisfiable query 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 (Section 2.4). Since P and Q have already been checked for satisfiability and the chase can be computed in polynomial time, we may assume that both P and Q are tree shaped and contain no cycles.

In the remainder of this proof, we use the notions of distance between nodes and the length of a path in a tree. When  $u_1, \ldots, u_n$  are nodes in a tree t such that, for every  $i=1,\ldots,n-1$ , either  $Child(u_i,u_{i+1})$  or  $Child(u_{i+1},u_i)$  holds, then  $p=(u_1,u_2),(u_2,u_3),\ldots,(u_{n-1},u_n)$  is a path between  $u_1$  and  $u_n$  in t. The length of p is n-1, which corresponds to the number of edges in p. The distance between two nodes u and v in t is the length of the shortest path between u and v.

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., [13]. If there is a homomorphism, we can conclude that  $L(P) \subseteq L(Q)$ . From now on, we assume that there is no homomorphism from Q to P.

If there is no homomorphism and P has only one maximal connected component, we can conclude that  $L(P) \nsubseteq L(Q)$ , 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, as in the example in Fig. 5.

We try to find a counterexample to containment, that is, a tree that satisfies P but not Q. As usual, by Observation 4, 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

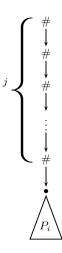


Fig. 6. P<sub>i</sub>

P and let the root variable of each  $P_i$  be  $r_{P_i}$ . 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  $P_i$ , in particular, there is no homomorphism from Q to  $P_i$  for any i. Therefore, this counterexample tree does not model Q. Thus we assume in the following that  $r_Q$  has no label.

In the following, we will reason about what criteria a counterexample tree must satisfy, and try to construct one that satisfies 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 = |Var(Q)|. For each  $j \in \{0, ..., n\}$  and  $i \in \{1, ..., m\}$ , let  $P_i^j$  be the query obtained by adding new variables  $z_1, ..., z_j$  to  $P_i$ , each of them labeled by the new symbol #, adding the atoms  $Child(z_l, z_{l+1})$  for  $1 \le l < j$ , and  $Child(z_j, r_{P_i})$ ; see Fig. 6. In particular,  $P_i^0 = P_i$ .

see Fig. 6. In particular,  $P_i^0 = P_i$ .

Now, for each  $1 \le i \le m$ , we define  $v_i$  to be the largest number smaller than n+1 such that there is no homomorphism from Q to  $P_i^{v_i}$ . Notice that  $v_i$  is well defined since there is no homomorphism from Q to P. Also notice that, if there is no homomorphism from Q to  $P_i^n$ , there is no homomorphism from Q to  $P_i^n$  for any  $l \in \mathbb{N}$ , since Q has n variables, is connected, and only uses the *Child* axis.

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  $\theta$  of P on t, the distance from the root of t to  $\theta(r_{P_i})$  cannot be larger than  $v_i$ . Indeed, since the label # does not 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  $P_i$  also satisfies Q.

For some  $P_i$ , we may have  $v_i = 0$ . This means that there is no homomorphism from Q to  $P_i^0 = P_i$ , but there is a homomorphism from Q to  $P_i^1$ . Using the above arguments, for any counterexample tree t and any satisfaction  $\theta$  for P on t, the root variable  $r_{P_i}$  of  $P_i$  must be assigned to the root of t by  $\theta$ . Let s 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  $\{s+1,\ldots,m\}$ . If there are two components,  $P_{i_1}$  and  $P_{i_2}$  such that  $i_1,i_2>s$  and the roots of  $P_{i_1}$  and  $P_{i_2}$  have different labels, they cannot both be assigned to the root of a counterexample tree. Thus no witness tree exists, and we can conclude that containment holds.

Recall that  $C_1, \ldots, C_k$  are the subqueries of Q attached to its root variable. For each  $1 \le j \le k$  and each  $1 \le i \le s$ , 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  $P_i^l$  that assigns  $r_{C_j}$  to the root variable of  $P_i^l$ . The intuition behind this definition is the following. Suppose t is a tree in L(P) and  $\theta$  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  $\theta_{C_j}$  for  $C_j$  on t such that  $\theta_{C_i}(r_{C_i})$  is a child of the root of t.

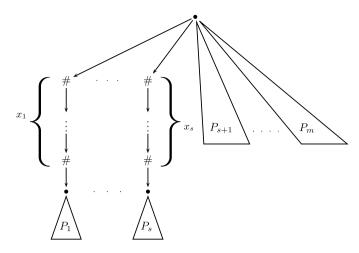
In the remainder of our procedure, we now iterate over every  $C_j$   $(1 \le j \le k)$ . In each iteration, we will try to construct a counterexample tree for containment. If, for every j, we can construct no such tree, we will prove that  $L(P) \subseteq L(Q)$ .

To this end, fix a  $C_j$ . We will try to construct a tree  $t_j$  in L(P) such that

- 1. every satisfaction for Q on  $t_i$  must assign  $r_Q$  to the root of  $t_i$ , and
- 2. there is no satisfaction for Q on  $t_i$  that assigns the root variable of  $C_i$  to a child of the root of  $t_i$ .

If we can find such a  $t_i$ , then containment fails.

For each  $1 \le i \le s$ , we pick an arbitrary number  $x_i$  from  $\{0, ..., v_i - 1\} - S_j^i$ . If, for some i, there is no such number, i.e.,  $\{0, ..., v_i - 1\} = S_j^i$ , then we set  $x_i = -1$ .



**Fig. 7.** The construction of  $t_i$ , the candidate counterexample tree for  $C_i$ .

Given these values  $x_1, \ldots, x_s$  we try to construct a counterexample tree  $t_j$  as follows. For each  $1 \le i \le s$  such that  $x_i \ge 0$ , 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  $i \in \{s+1,\ldots,m\}$  or  $x_i = -1$ , we take a minimal model of  $P_i$  and identify its root with the root of  $t_j$ . This may cause a conflict, if two or more of these minimal models already have fixed and different labels. If this is the case, our attempt to construct a counterexample tree for  $C_j$  fails and we must iterate to the next  $C_j$ . Otherwise, if at least one of them has a fixed root label a, the root of  $t_j$  gets label a. If not, then we give the root label #. This construction is depicted in Fig. 7. Clearly,  $t_j \in L(P)$ . For each  $P_i$  such that  $i \le s$ , 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  $t_j$  if it can be matched at all.

We now test whether  $t_j \models Q$ , and we have found a counterexample tree if  $t_j \not\models Q$ . If  $t_j \models Q$ , we have failed to construct a counterexample tree, and we must iterate to the next  $C_j$ .

**Claim 1.** Let  $t_j$  be the tree we have constructed for  $C_j$ . If  $t_j \models Q$  then, for every tree  $t \in L(P)$ ,

- either  $t \models Q$ , or
- there is a satisfaction for  $C_i$  on t that assigns  $r_{C_i}$  to a child of the root of t.

**Proof.** Let  $\theta_Q$  be a satisfaction for Q on  $t_j$ . By construction of  $t_j$ , we know that  $\theta_Q(r_Q)$  must be the root of  $t_j$ . Also,  $\theta_Q$  must assign all variables of  $C_j$  completely within a subtree of  $t_j$  corresponding to a minimal model of some  $P_i$  with i > s or such that  $x_i = -1$ . Otherwise,  $\theta_Q(r_{C_j})$  would have to be a child of the root of  $t_j$  that corresponds to a minimal model of  $P_i^{x_i}$ , for some  $i \le s$  and  $x_i \ge 0$ . But we know that  $x_i \notin S_j^i$  for any  $i \le s$ . Thus there is no  $P_i^{x_i}$  such that  $C_j$  can be matched in a corresponding minimal model. We conclude that there is a  $P_i$  with i > s or  $x_i = -1$  such that there is a homomorphism from  $C_j$  to  $P_i$  that maps  $P_i$  to a child of the root of  $P_i$ .

Now consider a tree  $t \in L(P)$  and a satisfaction  $\theta_P$  for P on t. If  $\theta_P(r_{P_i})$  is the root of t, then there is a satisfaction for  $C_j$  on t that assigns  $r_{C_j}$  to a child of the root of t. If, on the other hand, the distance from the root of t to  $\theta_P(r_{P_i})$  is d > 0, then we have two cases:

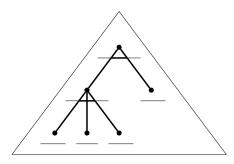
- 1. If  $d > v_i$ , then there is a satisfaction for Q on t.
- 2. If  $d \le v_i$ , then, since we have  $x_i = -1$ , we know that  $d 1 \in S_j^i$ , and again there must be a satisfaction for  $C_j$  on t that assigns  $r_{C_j}$  to a child of the root of t.

This concludes the proof of Claim 1.  $\Box$ 

If we go through all subqueries  $C_j$  of Q without being able to successfully construct a witness tree, we argue that containment holds.

Since  $t_j \models Q$  for every candidate counterexample  $t_j$ , Claim 1 tells us that for every tree  $t \in P(t)$  and every  $j \in \{1, ..., k\}$ , there is a satisfaction  $\theta_{C_j}$  for  $C_j$  on t that assigns the root of  $C_j$  to a child of the root of t.

Since we have assumed that  $r_Q$  has no label in Q, we can assign  $r_Q$  to the root of t. By combining this assignment with the assignments  $\theta_{C_i}$  we obtain a satisfaction for Q on t.  $\square$ 



**Fig. 8.** The construction of t' from t. The dots represent nodes in the set S. A horizontal line under a node represents the children of that node. The bold lines represent the nodes that are kept in t', while everything else is deleted.

## 3.2. coNP and $\Pi_2^P$ upper bounds

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

**Lemma 1.** Let P and Q be conjunctive queries. If  $L(P) \nsubseteq L(Q)$  then there exists a tree t such that  $t \models P$ ,  $t \not\models Q$ , and  $|t| \leqslant 2 \cdot |Var(P)| \cdot 2(|Var(Q)| + 5)$ .

**Proof.** Let t be a tree such that  $t \models P$  and  $t \not\models Q$ . Let  $\theta$  be a satisfaction of P on t, and let  $T = \{\theta(x) \mid x \in Var(P)\}$ . Let S be the set of nodes v of t such that v is the least common ancestor of some nonempty subset of T. Further, let R be the set of nodes that belong to a path between two nodes in S.

We will now remove nodes from t to obtain a new tree t'. The construction of t' from t is depicted in Fig. 8 and formally described as follows. We remove all nodes u from t such that

- $u \notin R$ , and
- u does not have both an earlier (left) sibling and a later (right) sibling in R, i.e., there do not exist x and y in R with  $NextSibling^+(x, u)$  and  $NextSibling^+(u, y)$ .

Thus we obtain a new tree t'. Clearly,  $t' \models P$  and  $\theta$  is a satisfaction of P on t'. Notice that there is no node u in t such that u is deleted while both its parent and one of its children belong to t'. Also, there is no u in t that is deleted while both its left sibling and its right sibling belong to t'. This means that t' is a {Child, NextSibling}-subtree of t. Since all other axes are transitive, t' is also an  $\mathcal{R}$ -subtree of t, where  $\mathcal{R}$  is the set of all axes. Hence, by Observation 2,  $t' \not\models Q$ . Notice that  $|S| < 2 \cdot |Var(P)|$ , and that t' only branches at nodes in S.

Now suppose that  $|t'| > 2 \cdot |Var(P)| \cdot 2(|Var(Q)| + 5)$ . We argue that one of the following two cases must occur:

#### **Case 1.** There is a pair u, v of nodes in t' from S such that

- 1. u is an ancestor of v;
- 2. the path  $\rho$  from u to v has length at least |Var(Q)| + 4; and
- 3. no internal node on  $\rho$  belongs to S.

#### **Case 2.** There is a pair u, v of nodes in t' such that

- 1. u and v are siblings, and children from a node in S;
- 2. there are at least |Var(Q)| + 2 nodes between u and v; and
- 3. no node between u and v has a descendant in S.

Indeed, towards a contradiction, suppose that none of the two cases applies. We will now count the nodes in t' and show that t' cannot contain more than  $2 \cdot |\text{Var}(P)| \cdot 2(|\text{Var}(Q)| + 5)$  nodes, which is a contradiction. First of all, we already noted that S contains at most 2|Var(P)| nodes. Now, to every node u in t' that belongs to S, we will associate two sets  $N_u^1$  and  $N_u^2$ , of at most |Var(Q)| + 5 nodes in t'. For each  $u \in S$ , the sum of the sizes of  $N_u^1$ ,  $N_u^2$  is at most 2(|Var(Q)| + 5). Furthermore, if we take the union of all  $N_u^1$ ,  $N_u^2$  over all nodes u in S, we have all the nodes in t'.

Consider, for each node  $u \in S$ ,

- the lowest ancestor  $u_{anc}$  of u in S, if it exists. If it does not exist, then u is the root of t'. In this case  $u_{anc}$  is undefined,  $N_u^1 = \{u\}$ , and  $N_u^2 = \emptyset$ ;
- the child  $u_{\text{anc-child}}$  of  $u_{\text{anc}}$  on the path to u;

• the leftmost child  $u_{\text{anc-child-right}}$  of  $u_{\text{anc}}$  to the right of  $u_{\text{anc-child}}$  that has a descendant in S, if it exists. Otherwise,  $u_{\text{anc-child-right}}$  is undefined and  $N_u^2 = \emptyset$ .

Let  $N_u^1$  be the set containing u and all nodes between u and  $u_{anc}$ . Let  $N_u^2$  be the set of all children of  $u_{anc}$  between  $u_{anc-child}$  and  $u_{anc-child-right}$ . Observe that

$$Nodes(t') = \bigcup_{u \in S} N_u^1 \cup N_u^2.$$

If Case 1 does not apply, then  $N_u^1$  contains at most |Var(Q)| + 4 nodes. If Case 2 does not apply, then  $N_u^2$  contains at most |Var(Q)| + 3 nodes. Hence, this implies that t' can contain not more than  $2|Var(P)| \cdot 2(|Var(Q)| + 4)$  nodes, which is a contradiction.

Hence, for trees with more than  $2 \cdot |Var(P)| \cdot 2(|Var(Q)| + 5)$  nodes, one of the two cases must apply. We now show that, in each of the above cases, we can construct a tree  $t_{small}$  in L(P) - L(Q) which is smaller than t'.

**Case 1.** Consider the path from u to v as in the description of Case 1. We now show how we obtain  $t_{\rm small}$  from t'. First, we change the label of every node in Nodes(t')-T to a new  $\Sigma$ -symbol # that does not appear in Q. Call the obtained tree  $t'_{\#}$ . Notice that such a  $\Sigma$ -symbol always exists because Q only 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  $t_{\rm small}$ . We next show that  $t_{\rm small}$  is indeed the tree we are looking for.

First of all, notice that  $t_{small}$  still models P, as  $\theta$  is still a satisfaction of P on  $t_{small}$ . Furthermore, towards a contradiction, suppose there is a satisfaction  $\theta_Q$  for Q on  $t_{small}$ . As the length of the path  $\rho'$  from u to v in  $t_{small}$  is at least |Var(Q)|+3, there is at least one interior node w of  $\rho'$  such that w does not have any siblings in  $t_{small}$  and such that no variable of Q is assigned to w by  $\theta_Q$ . Partition Var(Q) into the set Y of variables assigned by  $\theta_Q$  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  $\theta_Q$  is a satisfaction for Q on this new tree, and thus also on  $t'_\#$ . This is a contradiction.

Case 2. Consider the nodes u, v as in the description of Case 2 and consider the nodes between u and v. Without loss of generality, assume that u is to the left of v. We obtain  $t_{\rm small}$  from t' similarly as in Case 1. First, just as in Case 1, we change the label of every node in Nodes(t') - T to a new  $\Sigma$ -symbol # that does not appear in Q and we call the obtained tree  $t'_{\#}$ . Then, we remove the next sibling of u in  $t'_{\#}$ , and we adapt the NextSibling relation to reflect this change (that is, we fill in the new next sibling of u). Doing so, we obtain the tree  $t_{\rm small}$ . The argument why  $t_{\rm small}$  is the tree we are looking for is completely analogous to the argument in Case 1, except that we now need to consider NextSibling-axes instead of Child-axes. Notice that all nodes between u and v are leaves in t'.

The above two cases can be repeated until we have a witness tree of size at most  $2 \cdot |Var(P)| \cdot 2(|Var(Q)| + 5)$ .  $\Box$ 

The above lemma puts conjunctive query containment in  $\Pi_2^P$ . Indeed, for testing whether  $L(P) \nsubseteq L(Q)$ , the algorithm would guess a tree  $t_{small}$  of size at most  $2 \cdot |Var(P)| \cdot 2(|Var(Q)| + 5)$ , test in NP whether  $t_{small} \models P$  and test in coNP whether  $t_{small} \not\models Q$ . As Gottlob et al. showed that conjunctive query evaluation is in PTIME for  $CQ(Child, NextSibling, NextSibling^*, NextSibling^+)$ ,  $CQ(Child^*, Child^+)$ , and CQ(Following) [10], the above algorithm gives us a coNP upper bound for containment for these fragments. We can therefore state the following theorem.

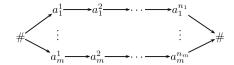
#### Theorem 7.

- 1. Containment is in  $\Pi_2^P$  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 reduce from the complement of either the Shortest Common Supersequence problem or the Shortest Common Superstring problem, both of which are known to be NP-complete [14,8]. 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 S is a superstring of  $S_0$  if  $S_0$  can be obtained by deleting a prefix and a postfix of S.

**Theorem 8.** Containment is coNP-hard for CQ(NextSibling<sup>+</sup>), CQ(NextSibling<sup>\*</sup>), CQ(Child<sup>+</sup>), CQ(Child<sup>\*</sup>), and CQ(Following).



**Fig. 9.** Structure of query P in the proof of Theorem 8.

**Proof.** All cases are proved by a reduction from the complement of Shortest Common Supersequence. To this end, let (S,k) be an instance of Shortest Common Supersequence. We now define conjunctive queries P and Q such that  $P \nsubseteq Q$  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 Fig. 9, where each arrow represents a  $NextSibling^+$ -axis and # and each  $a_i^j$  is a  $\Sigma$ -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

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

It is not difficult to see that  $P \nsubseteq Q$  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.  $\square$ 

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 Fig. 9, the arrows in Fig. 9 now denote NextSibling-axes, and that all the  $a_j^i$ -labeled nodes are connected to a common parent by Child-axes.

**Theorem 9.** Containment is coNP-hard for CQ (Child, NextSibling).

3.4.  $\Pi_2^P$  lower bounds

The  $\Pi_2^P$  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  $\{x_1, \ldots, x_{n_x}\} \cup \{y_1, \ldots, y_{n_y}\}$ . 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 [15].

**Lemma 2.**  $\forall \exists$  positive 1-in-3 SAT is  $\Pi_2^P$ -complete.

**Proof.** Membership of the problem in  $\Pi_2^P$  is trivial. For  $\Pi_2^P$ -hardness, we reduce from  $\forall\exists$  3SAT. First, we convert a  $\forall\exists$  3SAT formula  $\phi$  into a  $\forall\exists$  1-in-3 SAT formula  $\phi'$ . Second, we show how to get rid of negative literals.

Let  $C = (x \lor y \lor z)$  be a clause of  $\phi$  (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.

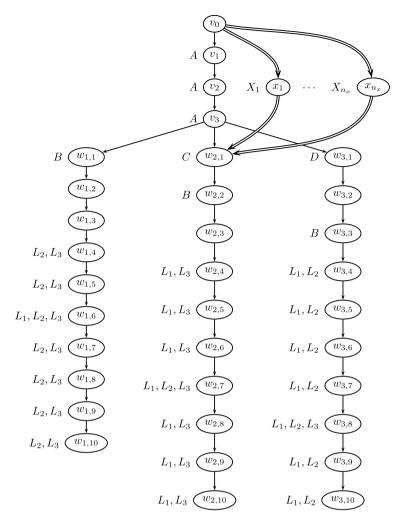
Finally, we show how to ensure that each clause contains exactly three literals. Suppose that we have a clause  $(x \lor y)$ . We introduce four new existentially quantified variables f, a, b, c and rewrite  $(x \lor y)$  as  $(x \lor y \lor f)$ ,  $(f \lor a \lor b)$ ,  $(f \lor b \lor c)$ , and  $(a \lor b \lor c)$ . 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.  $\Box$ 

**Theorem 10.** Containment is  $\Pi_2^P$ -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 [10]. We reduce from  $\forall\exists$  positive 1-in-3 SAT, which is  $\Pi_2^P$ -complete (see Lemma 2).

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.



**Fig. 10.** Illustration of the definition of query P in the proof of Theorem 10.

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

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

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  $\forall \exists$  1-in-3 SAT solution for the original formula if and only if there is one for  $\Phi$ . We need the extension  $\phi$  of our original formula for handling the x-variables in our proof.

Let query P be defined as in Fig. 10, where single lines represent the Child axis, double lines represent the Child<sup>+</sup> axis, the symbols inside the nodes are variables of *P* and the symbols to the left of nodes are their labels.

For the query Q, we introduce variables  $a_i$ ,  $b_i$  for each  $i = 1, ..., m + n_x$  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 Fig. 10 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 2) and the fact that each  $t \in L(P)$  has a {Child}-subtree in  $T_P$ .<sup>4</sup>

**Observation 11.** 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 from  $\forall \exists$  positive 1-in-3 SAT to containment for  $CQ(Child, Child^+)$  is correct, that is,

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

(⇒) Assume that, for every truth assignment  $\sigma_x$ :  $\{x_1,\ldots,x_{n_x}\}$  → {true, false}, there exists a truth assignment  $\sigma_y$ :  $\{y_1,\ldots,y_{n_y}\}$  → {true, false} such that each clause  $C_i$ ,  $1 \le i \le m$ , contains precisely one true literal under  $\sigma_x$  and  $\sigma_y$ . We show that  $t_P \models Q$  for every  $t_P \in T_P$ . According to Observation 11, 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  $\theta_P$  of P on  $t_P$ . From  $\theta_P$ , we define a truth assignment  $\sigma_{t_P}: \{x_1, \ldots, x_{n_x}\} \to \{\text{true}, \text{false}\}$  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,  $\sigma_{t_P}$  assigns a truth value to every  $x_i$ ,  $1 \le i \le n_x$ . Hence, there exists a  $\sigma_y : \{y_1, \ldots, y_{n_y}\} \to \{\text{true}, \text{false}\}$  such that each clause  $C_i$ ,  $1 \le i \le m$ , contains precisely one true literal under  $\sigma_{t_P}$  and  $\sigma_y$ . From  $\sigma_y$ , we now construct a truth assignment  $\sigma_y' : \{y_1, \ldots, y_{n_y}, y_1', \ldots, y_{n_x}', y_1'', \ldots, y_{n_x}'', y_{n_x}''\} \to \{\text{true}, \text{false}\}$  as follows:

- for each  $i = 1, ..., n_y$ ,  $\sigma'_v(y_i) = \sigma_v(y_i)$ ;
- if  $\theta_P(v_2)$  is labeled  $X_i$ , then we set  $\sigma'_v(y'_i) = \sigma'_v(y''_i) = \text{false}$ ;
- otherwise, if  $\theta_P(v_1)$  is labeled  $X_i$ , then we set  $\sigma'_v(y'_i) = \text{true}$  and  $\sigma'_v(y''_i) = \text{false}$ ;
- otherwise, we set  $\sigma'_{\nu}(y''_i) = \text{true}$  and  $\sigma'_{\nu}(y'_i) = \text{false}$ .

It is easy to see that each clause  $C_1, \ldots, C_m$  contains precisely one true literal under  $\sigma_{t_P}$  and  $\sigma_y$  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  $\sigma_{t_P}$  and  $\sigma'_y$ .

We will show how  $\sigma_{t_P}$  and  $\sigma'_y$  induce a satisfaction  $\theta_Q$  of Q on  $t_P$ . Let  $\sigma:\{1,\ldots,m+n_x\}\to\{1,2,3\}$  be defined as  $\sigma(i)=k'$  if and only if the k'-th literal in  $C_i$  is true under  $\sigma_{t_P}$  and  $\sigma'_y$ . Notice that  $\sigma$  is total and well defined. We first define a valuation  $\theta_Q$  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  $i = 1, ..., m + n_x$ ;
- $\theta_Q(b_i) = \theta_P(w_{\sigma(i),\sigma(i)})$  for each  $i = 1, ..., m + n_x$ ; and
- $\theta_O(c_{k,l,i,j}) = \theta_P(w_{\sigma(i),5+k-l+\sigma(j)})$  for each variable  $c_{k,l,i,j}$ .

We now prove that  $\theta_Q$  is a satisfaction of Q on  $t_P$ . Our choice of  $\theta_Q$  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  $X_{i-m}$ . Therefore,  $\theta_Q$  is indeed a satisfaction of Q on  $t_P$  and  $t_P \models Q$ .

( $\Leftarrow$ ) Assume that  $t_P \models Q$  for every  $t_P \in T_P$ . We show that, for each truth assignment  $\sigma_X : \{x_1, \dots, x_{n_X}\} \to \{\text{true}, \text{false}\}$ , there exists a truth assignment  $\sigma_Y : \{y_1, \dots, y_{n_Y}\} \to \{\text{true}, \text{false}\}$  such that each clause  $C_i$ ,  $1 \leqslant i \leqslant m$ , contains precisely one true literal under  $\sigma_X$  and  $\sigma_Y$ .

Let  $\sigma_x : \{x_1, \dots, x_{n_x}\} \to \{\text{true}, \text{ false}\}\$  be a truth assignment. We define the tree  $t_x$  as the tree implied by the variables and *Child*-axes in Fig. 10 with the additions that, for each  $i = 1, \dots, n_x$ ,

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

 $<sup>^4\,</sup>$  Recall the definition of  $\mathcal{R}\text{-subtrees}$  from Section 2.

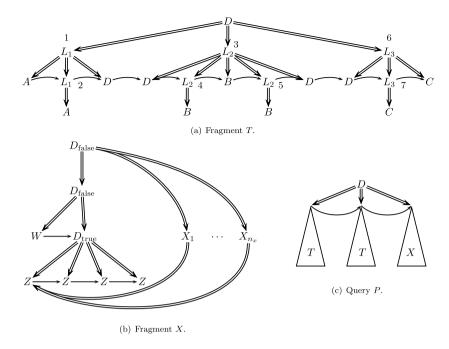


Fig. 11. Definition of query P in the proof of Theorem 13.

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

Let  $\theta$  be a satisfaction of Q on  $t_x$ . We show that  $\theta$  induces a truth assignment  $\sigma_y : \{y_1, \ldots, y_{n_y}\} \to \{\text{true}, \text{false}\}$  such that each clause  $C_i$ ,  $1 \le i \le m$  contains precisely one true literal under  $\sigma_x$  and  $\sigma_y$ . We first show that  $\theta$  induces a truth assignment  $\sigma_y' : \{y_1, \ldots, y_{n_y}, y_1', \ldots, y_{n_x}', y_1'', \ldots, y_{n_x}''\} \to \{\text{true}, \text{false}\}$  such that each clause  $C_i$ ,  $1 \le i \le m + n_x$  contains precisely one true literal under  $\sigma_x$  and  $\sigma_y'$ .

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  $\sigma_X(x_i) = \text{true}$ . 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  $\sigma_v$  we are looking for is  $\sigma'_v$  restricted to  $\{y_1, \ldots, y_{n_v}\}$ .

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 Fig. 10,  $z_0$  takes the place of z, while each  $z_i$ ,  $1 \le i \le k$  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 12.** Containment is  $\Pi_2^P$ -hard for CQ(Child, Following).

**Proof.** We adapt the proof of Theorem 10 by simulating  $Child^+$  with Child and Following. To this end, we begin by equipping each of the variables u in query P defined in Fig. 10 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  $\Sigma$ -label # that doesn't appear in the queries P and Q. Now, whenever  $Child^+(u,v)$  is used in one of the queries, we can replace it by

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

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 #.  $\square$ 

**Theorem 13.** Containment is  $\Pi_2^P$ -hard for CQ(Child<sup>+</sup>, Following) and for CQ(Child<sup>\*</sup>, Following).

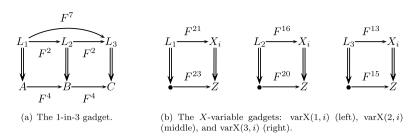


Fig. 12. Gadgets for the definition of query Q in the proof of Theorem 13.

**Table 3** The function NAND(k, l) [10].

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

**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  $\forall \bar{x} \exists \bar{y} C_1, \ldots, C_m$  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 [10]).

The construction of query P is illustrated in Fig. 11. Here, every double-lined edge represents a  $Child^+$ -axis and every directed edge represents a Following-axis. Fig. 12 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 I-in-3 gadget from Fig. 12(a) by  $v_1$ ,  $v_2$ , and  $v_3$ , respectively. Moreover, we annotate the query fragment T in Fig. 11(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 Fig. 11(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 14.** (See [10].) Every satisfaction  $\theta$  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  $\theta$  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 Fig. 11(c). That is, P contains two copies of the fragment T, followed by a copy of the X-fragment from Fig. 11(b). The ordering between the subqueries of P is enforced by F sollowing-axes: the root of T's left copy has a F sollowing-axis to the root of T's right copy, and the root of T's right copy has a T sollowing-axis to the root of the T-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  $\bar{x}$  and  $\bar{y}$  actually make one literal per clause of  $\forall \bar{x} \exists \bar{y} C_1, \ldots, C_m$  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  $\bar{x}$ -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" otherwise (see Fig. 11(b)). For example, all  $X_i$ -labeled descendants of the  $D_{\text{true}}$  node are interpreted as "true", and all  $X_i$ -labeled ancestors of the lower  $D_{\text{false}}$  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 T in P. This will be taken care of by the X-assignment gadgets in Q, which are illustrated in Fig. 12(b).

Formally, query Q is defined as follows. Each clause  $C_i$  is represented by two copies of the 1-in-3 gadget of Fig. 12(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 X0, labeled X1 in their respective subqueries in X1 cannot both match the topmost node labeled X2, respectively X3, in the left, respective right, copy of X4 min by adding an atom of the form X6 form X7 by to the query X7. To see this, we exemplify the case where X2 is defined as follows.

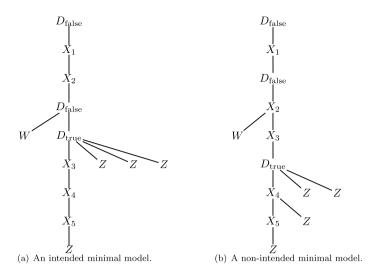


Fig. 13. Minimal models of the X-fragment of P.

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  $Q_i'$  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  $Following^{NAND(k,l)}(x, y)$  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  $\bar{x}$  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  $\bar{x}$ , 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  $D_{\text{true}}$ -labeled node is a child of the lower  $D_{\text{false}}$ -labeled node;
- 2. the W-labeled node is a child of the lower  $D_{\text{false}}$ -labeled node; and
- 3. the three rightmost Z-labeled nodes are children of the  $D_{true}$ -labeled node.

Fig. 13 contains an intended (left) and a non-intended minimal model (right) 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  $D_{\text{true}}$ ), and we say that  $t_P$  picks  $x_l$  to be false otherwise (i.e., if it is an ancestor of the lower  $D_{\text{false}}$ -labeled node). We can now make the following observation:

**Observation 15.** Let  $t_P$  be an intended minimal model of P. Then, for every satisfaction  $\theta$  of Q on  $t_P$ , the following holds. If the k-th literal of  $C_i$  is a universally quantified variable  $x_l$ , then  $\theta$  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 15 can be easily verified by testing the possible homomorphisms from the X-variable gadgets of Q (Fig. 12(b)) to the query P (Fig. 11). 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  $\theta$  chooses the first literal  $x_l$  of  $C_i$  to be true on  $t_P$ . Then  $\theta$  also matches the  $L_1$ -labeled node of the leftmost X-variable gadget in Fig. 12(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  $\theta$ 

must match the  $X_l$ -labeled node as a descendant of  $D_{\text{true}}$ . 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  $\theta$  chooses the first literal  $x_l$  of  $C_l$  to be false on  $t_P$ . Then  $\theta$  also matches the  $L_1$ -labeled node of the leftmost X-variable gadget in Fig. 12(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  $X_I$  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  $X_I$  with 23 Following steps. This is only possible if the  $X_I$ -labeled node occurs as an ancestor of  $D_{\text{true}}$ , which means that  $t_P$  chooses  $x_I$  to be false. This concludes the proof of Observation 15. □

This concludes the reduction for Theorem 13. We proceed to proving that the reduction is also correct. That is, we show that

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\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  $\sigma_X : \{x_1, \dots, x_{n_X}\} \to \{\text{true}, \text{false}\}$ , there exists a truth assignment  $\sigma_{v}: \{y_{1}, \dots, y_{n_{v}}\} \to \{\text{true}, \text{false}\}\$ such that each clause  $C_{i}, 1 \leq i \leq m$  contains precisely one true literal

Let  $\sigma_x: \{x_1, \dots, x_{n_v}\} \to \{\text{true}, \text{false}\}\$  be an arbitrary but fixed truth assignment. 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  $D_{\text{true}}$ , two  $D_{\text{false}}$ , the W, the four Z, and the  $n_X$   $X_i$ -labeled nodes in Fig. 11(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  $D_{\rm false}$ -labeled nodes (we call this area  $X^{\rm false}$ ), or it can occur as a descendant of the  $D_{\rm true}$ -labeled node (we call this area  $X^{\text{true}}$ ).

The correspondence between  $\sigma_x$  and the third subtree of  $t_x$  is now encoded as follows:

- if σ<sub>X</sub>(X<sub>i</sub>) = true, then the label X<sub>i</sub> occurs in X<sup>true</sup> and not in X<sup>false</sup>; and
   if σ<sub>X</sub>(X<sub>i</sub>) = false, then the label X<sub>i</sub> occurs in X<sup>false</sup> and not in X<sup>true</sup>.

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

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  $\sigma_v$ . We argue that  $\sigma_v$  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  $C_i$  and we select L in  $C_i$ , then we also select L in  $C_i$ . Finally, the varX(k,l) gadgets ensure that  $\theta$  picks the same values for the  $x_i \in \bar{x}$  as  $t_x$  (Observation 15), and therefore also  $\sigma_x$ . Hence, the existence of  $\theta$  implies the existence of a valuation  $\sigma_y$ such that each clause  $C_i$ ,  $1 \le i \le m$  contains precisely one true literal under  $\sigma_x$  and  $\sigma_y$ .

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

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 10, we make the following observation.

#### **Observation 16.** 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 2, as each tree in L(P) has a {Child<sup>+</sup>, Following}-subtree in  $T_P$ . We show that  $t_P \models Q$  for every  $t_P \in T_P$ . According to Observation 16, this implies that  $L(P) \subseteq L(Q)$ . Given  $t_P \in T_P$ , we define a truth assignment  $\sigma_{t_P} : \{x_1, \dots, x_{n_x}\} \to \{\text{true}, \text{false}\}$  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,  $\sigma_{t_P}$  assigns a truth value to every  $x_i$ ,  $1 \le i \le n_X$ . Hence, there exists a  $\sigma_y : \{y_1, \dots, y_{n_v}\} \to \{\text{true, false}\}$ such that each clause  $C_i$ ,  $1 \le i \le m$ , contains precisely one true literal under  $\sigma_{t_P}$  and  $\sigma_y$ .

We will show how  $\sigma_{t_P}$  and  $\sigma_y$  induce a satisfaction  $\theta$  of Q on  $t_P$ . Let  $\tau:\{1,\ldots,m\} \to \{1,2,3\}$  be defined as  $\tau(i)=k$  if and only if the k-th literal in  $C_i$  is true under  $\sigma_{t_P}$  and  $\sigma_y$ . Notice that  $\tau$  is total and well defined. We first define a valuation  $\theta$  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 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 if \tau(i) = 3.
```

By definition of  $Q_i$ ,  $\theta$  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*<sup>NAND(k,l)</sup> 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  $\theta$  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 Fig. 13). If  $X_l$  is not reachable from the W-node with a *Following* step a descendant of  $D_{\text{true}}$ , then we can always map the Z-labeled node in the gadgets to  $z_4$  (see also Fig. 13).

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<sup>+</sup>, we immediately have the following corollary.

**Corollary 1.** Containment is  $\Pi_2^P$ -hard for  $CQ(Child^*, NextSibling^+)$ .

**Theorem 17.** Containment is  $\Pi_2^P$ -hard for

```
(1) CQ(Child*, NextSibling),
(2) CQ(Child*, NextSibling*),
(3) CQ(Child+, NextSibling),
(4) CQ(Child+, NextSibling+), and
(5) CQ(Child+, NextSibling*).
```

**Proof.** For each of these fragments, the proof of Theorem 13 can be adapted by the same methods as in the article by Gottlob et al. [10]. 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.  $\Box$ 

**Theorem 18.** Containment is  $\Pi_2^P$ -hard for CQ(Following, NextSibling).

**Proof.** Unfortunately, the arguments we use in Theorem 17 do not work seamlessly for *Following* and *NextSibling* $^{\alpha}$ , 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)
```

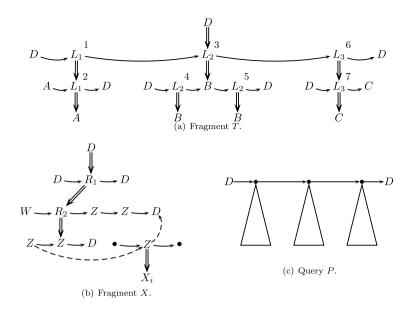
and by giving x and y different labels, the extra introduced nodes  $x_1$  and  $x_2$  for this encoding introduce difficulties for the  $X_i$ -labeled nodes of the P-query in the proof of Theorem 17. We therefore need to take a slightly different approach.

Figs. 14 and 15 illustrate how to change P and Q in the proof of Theorem 13. 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

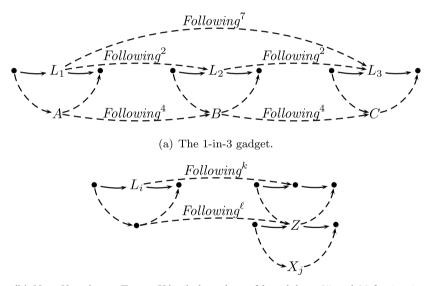
```
Descendant(x, y) = NextSibling(x_1, x) \land NextSibling(x, x_2) \land Following(x_1, y) \land Following(y, x_2),
```

where  $x_1$  and  $x_2$  are the variables left and right from x in Fig. 14, 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 13. 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.



**Fig. 14.** Definition of query P in the proof of Theorem 18.



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

Fig. 15. How to adapt O for Theorem 18.

Given that a double line denotes the above *Descendant* gadget, the 1-in-3 gadget of Q are almost the same as in Fig. 12(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 Fig. 15(a). The gadgets  $Q_i$  and  $Q_i'$  are then wired in precisely the same manner as in the proof of Theorem 13.

The most significant change in the Q-query is in the varX(i, j)-gadgets for the X-variables. How to adapt these gadgets is illustrated in Fig. 15(b). The remainder of the proof is analogous to the proof of Theorem 13.  $\Box$ 

 $\textbf{Theorem 19.} \ \textit{Containment is} \ \Pi_2^{P} - \textit{hard for CQ}(\textit{Following}, \textit{NextSibling}^+) \ \textit{and CQ}(\textit{Following}, \textit{NextSibling}^*).$ 

**Proof.** The reduction for  $CQ(Following, NextSibling^+)$  is analogous to the one in Theorem 18, i.e., we can replace every NextSibling in the proof of Theorem 18 with a  $NextSibling^+$ . The reduction of Theorem 18 can be adapted to a reduction for  $CQ(Following, NextSibling^*)$  by replacing every NextSibling(x, y) in P with  $NextSibling^*(x, xy) \land H(xy) \land NextSibling^*(xy, y)$ ,

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 18, depicted in Fig. 15(b). 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+i times. 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. This means, that in the new varX(i, j)-gadgets (see Fig. 15(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 1.  $\Box$ 

### 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 1.** 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  $\theta$ . Now let t' be a tree that

- contains the set T of nodes of t onto which variables are matched by  $\theta$ ;
- 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.  $\square$ 

#### 4.1. PTIME upper bounds

**Theorem 20.** Satisfiability is in PTIME for CQ(Child) and CQ(NextSibling).

**Proof.** First, we apply the chase to Q and obtain query chase(Q) (cf. Section 2.4). It should be clear that chase(Q) is satisfiable if and only if Q is satisfiable.

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

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

Each of these conditions can be tested in polynomial time.  $\Box$ 

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

**Definition 4.** 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 21.** Satisfiability is in PTIME for CQ (NextSibling<sup>+</sup>, NextSibling<sup>\*</sup>, Following) and CQ (Child<sup>+</sup>, Child<sup>\*</sup>).

**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 20, 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  $\theta$  for Q on t such that  $\theta$  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  $\theta$  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  $\theta$  by assigning x to the new node. After doing this for all such pairs x, y, the modified  $\theta$  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 22.** Satisfiability is in PTIME for CQ(Child, NextSibling) and CQ(Child, NextSibling<sup>+</sup>, 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  $NextSibling^{\alpha}$ -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  $a \ne b$ . 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 20. When this procedure merges classes of variables, carry these merges over to  $P_{\rm ns}$ ,  $P_{\rm c}$ , and  $P_{\rm c}$ .
- 3. For each maximal connected component of  $P_c$ , check satisfiability as in the proof of Theorem 20. When this procedure merges classes of variables, carry these merges over to  $P_{ns}$ ,  $P_c$ , and P.

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  $\theta$  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 every 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.

- (a) 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.
- (b) Since step (2) cannot merge any more classes, we know that each connected component of  $P_{ns}$  is string-shaped, that is, forms a non-branching sequence of variable classes, connected by *NextSibling*-axes.
- (c) Since step (3) cannot merge any more classes, we know that  $P_c$  is forest-shaped.
- (d) We also know that no two variable classes that belong to the same connected component of  $P_{ns}$  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  $C_i \prec C_k$ .

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 (a) above, and thus a contradiction.



Fig. 16. Gadget for the proof of Theorem 23.

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

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 Child- and NextSibling-atoms are satisfied. Thus  $\theta_P$  is a satisfaction of P on t. It is straightforward to see that  $\theta_Q$ , 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  $P_{ns}$  is done as in the proof of Theorem 21. This means that after the iteration terminates, it is not necessarily the case that each connected component of  $P_{ns}$  is string-shaped. Each such component is, however, satisfiable, and we can, as argued in the proof of Theorem 21, find a string model for it by considering a topological sorting.  $\Box$ 

#### 4.2. NP lower bounds

Theorem 23. Satisfiability is NP-hard for

- (1) CQ(Child, Child<sup>+</sup>),
- (2) CQ(Child, Child\*),
- (3) CQ(NextSibling, NextSibling<sup>+</sup>),
- (4) CQ(NextSibling, NextSibling\*),
- (5) CQ(NextSibling, Following), and
- (6) CQ(Child, Following).

**Proof.** All reductions are from Shortest Common Supersequence. For cases (1)–(5), the reductions are very similar. 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  $i = 1, \ldots, m$ ,  $s_i = a_i^1 \cdots a_i^{n_i}$ . Let # be a symbol not occurring in any string in S.

The construction of P is depicted in Fig. 16. 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 (" $\bullet$ ") 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 12, 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 [4] 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.

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