

# Parallel Tractability of Ontology Materialization: Technique and Practice

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## Abstract

Materialization is an important reasoning service for many ontology-based applications, but the rapid growth of semantic data poses the challenge to efficiently perform materialization on large-scale ontologies. Parallel materialization algorithms work well for some ontologies, which is not always explained by the theoretical results. For example, for the YAGO ontology parallelization works well, although the reasoning problem for the used ontology language is not in NC, i.e., the theoretical complexity class for *parallel tractability*. This motivates us to study the problem of *parallel tractability* of ontology materialization from a theoretical perspective. We focus on the datalog rewritable ontology languages DL-Lite and Description Horn Logic (DHL) and propose algorithms, called NC algorithms, to identify classes of ontologies for which materialization is tractable in parallel. To verify the practical usability of the above results, we analyze real-world datasets and show that for many ontologies expressed in DHL materialization is tractable in parallel. The implementation of our optimized parallel algorithm shows performance improvements over the highly optimized state-of-the-art reasoner RDFS on ontologies for which materialization is tractable in parallel.

**Keywords:** ontology, materialization, datalog, parallel tractability, NC complexity

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## 1. Introduction

The Web Ontology Language (OWL)<sup>1</sup> is an important standard for ontology languages in the Semantic Web and knowledge-based applications. In many of these applications, *materialization* plays an important role, which is the reasoning task of computing all implicit *facts* that follow from a given ontology [1]. Ontology developers and users employ materialization to optimize tasks such as query answering, ontology diagnosis or debugging. Since large amounts of semantic data are being generated at an increasing pace by sensor networks, government authorities and social media [2, 3], it is challenging to conduct materialization on such large-scale ontologies efficiently.

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<sup>1</sup>The latest version is OWL 2: <http://www.w3.org/TR/owl2-overview/>

For the ontology language RDFS and its extended fragments, approaches for parallel reasoning and for employing parallel computing platforms exist [4, 5, 6]. Several optimization strategies, e.g., dictionary encoding, balancing workload and data partitioning, are further studied to enhance parallel RDFS reasoning. There are also parallel implementations for scalable reasoning over ontologies that use highly expressive ontology languages [7, 8]. The different approaches utilize different kinds of computing platforms to make reasoning tasks more efficient in parallel, e.g., supercomputers [9, 10], MapReduce [11] and GPU servers [12].

The effectiveness of parallel reasoning for the above mentioned techniques is empirically verified on different test datasets. However, materialization is not tractable in parallel for most of the popular ontology languages with PTime-complete or higher complexity [13]. In particular, this is the case for RDFS and datalog rewritable ontology languages and, therefore, the efficiency of reasoning may not be improved using a parallel implementation. A possible reason for the apparent high performance of parallel reasoning is that the utilized test datasets do not fall into the worst cases in terms of computational complexity. Also some well-known, large-scale ontologies such as YAGO show good performance for parallel reasoning [14], although they are expressed in ontology languages that are, in theory, not tractable in parallel. On the other hand, there are ontologies for which materialization performance does not improve by parallelism and our experiments in Section 6 confirm this. The current theoretical work of parallel ontology reasoning can hardly explain this. While one can try out different parallel implementations to see whether an ontology can be handled by (one of) them efficiently, we study the problem of parallel tractability in theory and identify properties that make an ontology tractable in parallel. These properties can further be used to optimize parallel algorithms and to guide ontology engineers in creating ontologies for which parallel tractability can be guaranteed theoretically.

According to Motik et al. [4], many real large-scale ontologies are essentially expressed in ontology languages that can be rewritten into datalog rules. We follow this argumentation and focus on such datalog rewritable ontology languages in this paper. The main target of this paper is to identify classes of datalog rewritable ontologies such that materialization over these ontologies is tractable in parallel, i.e., falls into the parallel complexity class NC [13]. This complexity class consists of problems that can be solved efficiently in parallel. To show that a problem is in NC, one can give an *NC algorithm* that handles this problem using parallel computation [13]. An NC algorithm is required to terminate in parallel poly-logarithmic time. However, current materialization algorithms of datalog rewritable ontology languages (e.g., the core algorithm used in RDFox [4]) are not NC algorithms since they are designed for general datalog programs and have PTime-complete complexity. Thus, we first give NC algorithms that perform materialization and then identify the corresponding classes of datalog rewritable ontologies (called *parallel tractability classes*) that can be handled by these NC algorithms. To make the proposed NC algorithms practical, we also discuss how to optimize and implement them.

In the practical part of this work, we study specific datalog rewritable ontology languages. We first focus on the ontology language DL-Lite [15] to clarify how to apply the above NC algorithms to study parallel tractability of ontology materialization. We show that DL-Lite<sub>core</sub> and DL-Lite<sub>R</sub> are tractable in parallel. We then study the ontology language Description Horn Logic (DHL) [16], which is the intersection of datalog and OWL in terms of expressivity. We give a case of a DHL ontology where materialization can hardly be parallelized. Based on the analysis of this case, we propose to restrict the usage of DHL such that materialization over the restricted ontologies can be handled by the proposed NC algorithms. We further extend the results to an extension of DHL that also allows complex role inclusion axioms. Finally, we analyze well-known benchmarks and real-world datasets and show that many real ontologies following the

proposed restrictions belong to the parallel tractability classes. We implement a system based on an optimized NC algorithm for DHL materialization, which we compare to the state-of-the-art reasoner RDFS. The experimental results show that the optimizations proposed in this paper result in a better performance on ontologies that are tractable in parallel compared with RDFS.

The remainder of the paper is organized as follows. In Section 2, we introduce some basic notions. We then give two NC algorithms for ontology materialization in Section 3. We study parallel tractability of materialization in DL-Lite, DHL and an extension of DHL in Section 4, respectively. In Section 5, we discuss how to optimize and implement the given NC algorithms. We analyze real-world datasets and evaluate our implementation in Section 6. Finally, we discuss related work in Section 7 and conclude in Section 8. Detailed proofs of the theorems and lemmas in this paper are given in the appendix.

## 2. Background Knowledge

In this section, we introduce some basic notions that are needed to introduce our approach.

### 2.1. Datalog

We discuss the main issues in this paper using standard datalog notions. In datalog [17], a *term* is a variable or a constant. An *atom*  $A$  is defined by  $A \equiv p(t_1, \dots, t_n)$  where  $p$  is a *predicate* (or *relational*) name,  $t_1, \dots, t_n$  are terms, and  $n$  is the arity of  $p$ . If all the terms in an atom  $A$  are constants, then  $A$  is called a *ground atom*. A datalog *rule* is of the form: ' $B_1, \dots, B_n \rightarrow H$ ',<sup>2</sup> where  $H$  is referred to as the *head atom* and  $B_1, \dots, B_n$  the *body atoms*. Each variable in the head atom of a rule must occur in at least one body atom of the same rule. A *fact* is a rule of the form ' $\rightarrow H$ ', i.e., a rule with an empty body and the head  $H$  being a ground atom. A datalog program  $P$  consists of rules and facts. A *substitution*  $\theta$  is a partial mapping of variables to constants. For an atom  $A$ ,  $A\theta$  is the result of replacing each variable  $x$  in  $A$  with  $\theta(x)$  if the latter is defined. We call  $\theta$  a *ground substitution* if each defined  $A\theta$  is a ground atom. A *ground instantiation* of a rule is obtained by applying a ground substitution on all the terms in this rule with respect to a finite set of constants occurring in  $P$ . Furthermore the ground instantiation of  $P$ , denoted by  $P^*$ , consists of all ground instantiations of rules in  $P$ . The predicates occurring only in the body of some rules are called *EDB predicates*, while the predicates that may occur as head atoms are called *IDB predicates*.

### 2.2. DHL and DL-Lite

In what follows, we use **CN**, **RN** and **IN** to denote three disjoint countably infinite sets of *concept names*, *role names*, and *individual names* respectively. The set of roles is defined as  $\mathbf{R} := \mathbf{RN} \cup \{R^- \mid R \in \mathbf{RN}\}$  where  $R^-$  is the *inverse role* of  $R$ . For ease of discussion, we focus on the *simple forms* of axioms shown in the left column of Table 1. These simple forms can be obtained by using well-known *structural transformation* techniques [18, 19].

DHL (short for *description horn logic*) [16] is introduced as an intersection of description logic (DL) and datalog in terms of expressivity. We define a DHL ontology  $\mathcal{O}$  as a triple:  $\mathcal{O} = \langle \mathcal{T}, \mathcal{R}, \mathcal{A} \rangle$ , where  $\mathcal{T}$  denotes the TBox containing axioms of the forms (T1-T3);  $\mathcal{R}$  is the RBox that is a set of axioms of the forms (R1-R3);  $\mathcal{A}$  is the ABox containing *assertions* of the forms

<sup>2</sup>In datalog rules, a comma represents a Boolean conjunction ' $\wedge$ '.

Table 1: Axioms and corresponding datalog rules

	Axioms	Datalog Rules
(T1)	$A \sqsubseteq B$	$A(x) \rightarrow B(x)$
(T2)	$A_1 \sqcap A_2 \sqsubseteq B$	$A_1(x), A_2(x) \rightarrow B(x)$
(T3)	$\exists R.A \sqsubseteq B$	$R(x, y), A(y) \rightarrow B(x)$
(T4)	$A \sqsubseteq \exists R$	$A(x) \rightarrow R(x, o_R^A)$
(R1)	$S \sqsubseteq R$	$S(x, y) \rightarrow R(x, y)$
(R2)	$S \sqsubseteq R^-$	$S(x, y) \rightarrow R(y, x)$
(R3)	$R \circ R \sqsubseteq R$	$R(x, y), R(y, z) \rightarrow R(x, z)$
(R4)	$R_1 \circ R_2 \sqsubseteq R$	$R_1(x, y), R_2(y, z) \rightarrow R(x, z)$
(A1)	$A(a)$	$A(a)$
(A2)	$R(a, b)$	$R(a, b)$

(A1) and (A2). In an axiom of either of the forms (T1-T3 and R1-R3), concepts  $A_{(i)}$  and  $B$  are either concept names, the *top concept* ( $\top$ ) or the *bottom concept* ( $\perp$ );  $R$  and  $S_{(i)}$  are roles in  $\mathbf{R}$ . For an axiom of the form  $A \sqsubseteq \forall R.B$  that is also allowed in DHL, we only consider its equivalent form  $\exists R^-.A \sqsubseteq B$ .

DHL is related with other ontology languages. First, DHL is essentially a fragment of the description logic Horn-*SHOIQ* with disallowing *nominals*, *number restrictions* and right-hand side *existential restrictions* ( $A \sqsubseteq \exists R.B$ ). Second, the expressivity of DHL covers that of RDFS to some extent [16]. Reasoning with RDFS ontologies is NP-complete [20] and, thus, is not tractable in parallel. However, by applying some simplifications and restrictions, RDFS ontologies can be expressed in DHL [16], which has PTime-complete complexity for materialization.

In the initial work of DHL [16], *complex role inclusion axioms* (complex RIAs) of the form  $R_1 \circ \dots \circ R_n \sqsubseteq R$  are not considered, although they can be naturally transformed into datalog rules. In this paper, we also consider an extension of DHL (denoted by DHL( $\circ$ )) that allows complex RIAs. Since a complex RIA can be transformed to several axioms of form (R4), we then require that an RBox  $\mathcal{R}$  of a DHL( $\circ$ ) ontology can contain axioms of the forms (R1-R4). Note that (R3) is actually a special case of (R4).

DL-Lite is a group of ontology languages designed for highly-efficient query answering over knowledge bases and underpins the OWL profile OWL QL [15]. DL-Lite<sub>core</sub> and DL-Lite<sub>R</sub> are the two basic fragments of DL-Lite. DL-Lite<sub>core</sub> requires that a TBox only contains axioms of the forms (T1)  $A \sqsubseteq B$ , (T2)  $A_1 \sqcap A_2 \sqsubseteq B$  where  $B \equiv \perp$ , (T3)  $\exists R.A \sqsubseteq B$  where  $A \equiv \top$  or  $B \equiv \perp$ , and (T4)  $A \sqsubseteq \exists R$ , an ABox contains *assertions* of the forms (A1) and (A2), and RBoxes are not allowed. DL-Lite<sub>R</sub> is an extension of DL-Lite<sub>core</sub>, which allows involving RBoxes that contain axioms of the forms (R1-R2). If not specially specified, a *DL-Lite ontology* denotes a *DL-Lite<sub>R</sub> ontology* in the following paragraphs.

### 2.3. Ontology Materialization via Datalog Programs

An ontology expressed in DHL, DHL( $\circ$ ), DL-Lite<sub>core</sub> or DL-Lite<sub>R</sub> can be transformed into a datalog program (see the corresponding rules in the right column of Table 1). Note that an axiom of form (T4)  $A \sqsubseteq \exists R$  should be transformed into a first-order logic rule of the form  $A(x) \rightarrow \exists y(R(x, y))$ , where  $y$  is called a *free variable* (it is not occurring in the body atom  $A(x)$ ). In this work, such a free variable  $y$  is eliminated via Skolemization into a fresh individual  $o_R^A$  that

corresponds to the concept  $A$  and the role  $R$ . The elimination of free variables does not influence the result of materialization [15].

In what follows, for an ontology  $O = \langle \mathcal{T}, \mathcal{R}, \mathcal{A} \rangle$ , we use  $P = \langle R, \mathbf{I} \rangle$  to represent the corresponding datalog program where  $R$  is the set of rules obtained by transforming the axioms in  $\mathcal{T}$  and  $\mathcal{R}$  and  $\mathbf{I}$  is the set of facts that are directly copied from the assertions in  $\mathcal{A}$ . Further, we use  $R_1 \sqsubseteq_* R_2$  to denote the smallest transitive reflexive relation between roles such that  $R_1 \sqsubseteq R_2 \in \mathcal{R}$  implies  $R_1 \sqsubseteq_* R_2$  and  $R_1^- \sqsubseteq_* R_2^-$ . In this paper, we also use the notion of *simple role*, which was initially proposed to restrict the usage of highly expressive ontology languages [21]. Specifically, a role  $S \in \mathbf{R}$  is *simple* if, (1) it has no subrole (including  $S$ ) occurring on the right-hand side of axioms of the forms (R3) and (R4); (2)  $S^-$  is simple.

Based on the above representations, ontology materialization corresponds to the evaluation of datalog programs. Specifically, given a datalog program  $P = \langle R, \mathbf{I} \rangle$ , let  $T_R(\mathbf{I}) = \{H\theta \mid \forall B_1, \dots, B_n \rightarrow H \in R, B_i\theta \in \mathbf{I}, 1 \leq i \leq n\}$ , where  $\theta$  is some substitution; further let  $T_R^0(\mathbf{I}) = \mathbf{I}$  and  $T_R^i(\mathbf{I}) = T_R^{i-1}(\mathbf{I}) \cup T_R(T_R^{i-1}(\mathbf{I}))$  for each  $i > 0$ . The smallest integer  $n$  such that  $T_R^n(\mathbf{I}) = T_R^{n+1}(\mathbf{I})$  is called *stage*, and *materialization* refers to the computation of  $T_R^n(\mathbf{I})$  with respect to  $R$  and  $\mathbf{I}$ .  $T_R^n(\mathbf{I})$  is also called the *fixpoint* and denoted by  $T_R^\omega(\mathbf{I})$ . We say that an atom  $A$  is *derivable* or can be *derived* with respect to the datalog program  $P = \langle R, \mathbf{I} \rangle$  if  $A \in T_R^\omega(\mathbf{I})$ . In this paper, we consider the data complexity of materialization, i.e., we assume that the rule set  $R$  is fixed for any class of datalog programs.

#### 2.4. The Complexity Class NC

The parallel complexity class NC, known as Nick's Class [13], is studied by theorists as a parallel complexity class where each decision problem can be efficiently solved in parallel. Specifically, a decision problem in the NC class can be solved in poly-logarithmic time on a PRAM (parallel, random-access machine) with a polynomial number of processors. We also say that an NC problem can be solved in *parallel poly-logarithmic time*. Although the NC complexity class is a theoretical analysis tool, it has been shown that many NC problems can be solved efficiently in practice [13].

From the perspective of implementations, NC problems are also highly feasible in parallel for other parallel models like BSP [22] and MapReduce [23]. The NC complexity class was originally defined as a class of decision problems. Since we study the problem of materialization, we do not require in this work that a problem is a decision problem in NC. In addition, since many parallel reasoning systems (see related work in Section 7) are implemented on shared-memory platforms, we study all the issues in this work by assuming that the running machines are in shared-memory configurations.

### 3. Parallel Tractability of Datalog Programs

Our target is to find for which kinds of ontologies (not ontology languages) materialization is tractable in parallel. We consider data complexity, i.e., we consider classes of datalog programs which share a fixed the rule set. The data complexity of the materialization problem in this case is PTime-complete, which is considered to be inherently sequential in the worst case [13]. In other words, the materialization problem of datalog programs cannot be solved in parallel poly-logarithmic time unless  $P=NC$ . Thus, we say that *materialization for a class of datalog programs is tractable in parallel if there exists an algorithm that handles this class of datalog programs and runs in parallel poly-logarithmic time. Such an algorithm is also called an NC algorithm*. In this section, we identify such classes of datalog programs by studying different NC algorithms.

### 3.1. Parallel Tractability Classes

We first give the following definition for a class of datalog programs that is tractable in parallel, i.e., an NC algorithm exists for handling each datalog program in this class.

**Definition 1. (Parallel Tractability Class)** Given a class  $\mathcal{D}$  of datalog programs sharing the same rule set, we say that  $\mathcal{D}$  is a class of datalog program with parallel tractability, short a PTD class, if there exists an NC algorithm that performs sound and complete materialization for each datalog program in  $\mathcal{D}$ . The corresponding class of ontologies of  $\mathcal{D}$  is called a class of ontologies with parallel tractability, short a PTO class.

According to the above definition, if we find an NC algorithm  $A$  for datalog materialization, then we can identify a PTD class  $\mathcal{D}_A$ , which is the class of all datalog programs that can be handled by  $A$ . However, current materialization algorithms of datalog rewritable ontology languages (e.g., the core algorithm used in RDFS [4]) are not NC algorithms due to their PTime-complete complexity, since they are designed for handling general datalog programs. Thus, we proceed to devise specific NC algorithms. In the following, we first give a parallel materialization algorithm that works for general datalog programs. We then restrict this algorithm to an NC version and identify the target PTD class.

### 3.2. Materialization Graph

In order to give a parallel materialization algorithm, we introduce the notion of *materialization graphs*, which facilitates the analysis of the given algorithm.

**Definition 2. (Materialization Graph)** A materialization graph, with respect to a datalog program  $P = \langle R, I \rangle$ , is a directed acyclic graph denoted by  $\mathcal{G} = \langle V, E \rangle$  where,

- $V$  is the node set and  $V \subseteq T_R^\omega(I)$ ;
- $E$  is the edge set and  $E \subseteq T_R^\omega(I) \times T_R^\omega(I)$ .

For each edge of the form  $e(v_1, v_2)$  where  $v_1$  and  $v_2$  are two nodes, we say that the node  $v_1$  is the parent (node) of  $v_2$  and the node  $v_2$  is the child (node) of  $v_1$ . Further,  $\forall H, B_1, \dots, B_n \in V$ ,  $\mathcal{G}$  satisfies the following conditions:

- $H$  has a parent node or a child node;
- if  $H$  has the in-degree of 0, then  $H$  is an original fact of  $P$ ;
- $B_1, \dots, B_n \rightarrow H \in P^*$  is satisfied if  $e(B_1, H), \dots, e(B_n, H) \in E$  and  $B_1, \dots, B_n$  are all the parents of  $H$ .

For some derived atom  $H$ , there may exist several rule instantiations where  $H$  occurs as a head atom. This also means that  $H$  can be derived in different ways. The condition in the definition above results in only one way of deriving  $H$  being described by a materialization graph. Suppose  $\mathcal{G}$  is a materialization graph such that the nodes with in-degree 0 are the original facts in  $I$ . The size of  $\mathcal{G}$ , denoted by  $|\mathcal{G}|$ , is the number of nodes in  $\mathcal{G}$ . The depth of  $\mathcal{G}$ , denoted by  $\text{depth}(\mathcal{G})$ , is the maximal length of a path in  $\mathcal{G}$ . We next give an example of a materialization graph.

**Example 1.** Consider a  $DHL(\circ)$  ontology  $O_{ex_1}$  where the  $TBox$  is  $\{\exists R.A \sqsubseteq A\}$ , the  $RBox$  is  $\{S \circ R \sqsubseteq R\}$  and the  $ABox$  is  $\{A(b), R(a_1, b), S(a_i, a_{i-1})\}$  for  $2 \leq i \leq k$  and  $k$  an integer greater than 2. The corresponding datalog program of this ontology is  $P_{ex_1} = \langle R, I \rangle$  where  $I$  contains all the assertions in the  $ABox$  and  $R$  contains the two rules ' $R(x, y), A(y) \rightarrow A(x)$ ' and ' $S(x, y), R(y, z) \rightarrow R(x, z)$ '. The graph in Figure 1 is a materialization graph with respect to  $P_{ex_1}$ , denoted by  $\mathcal{G}_{ex_1}$ . The nodes with in-degree 0 are the original facts in  $I$ ; each of the other nodes corresponds to a ground instantiation of some rule. For example, the node  $A(a_k)$  corresponds to the ground rule instantiation ' $R(a_k, b), A(b) \rightarrow A(a_k)$ '. The size of this materialization graph is the number of nodes, that is  $3k$ . The depth of  $\mathcal{G}_{ex_1}$  is  $k$ .

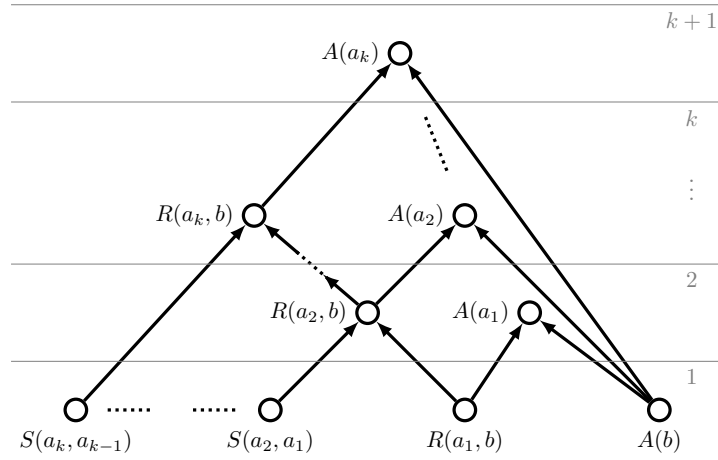


Figure 1: An example materialization graph

We say that a materialization graph  $\mathcal{G}$  is a *complete materialization graph* when  $\mathcal{G}$  contains all ground atoms in  $T_R^\omega(I)$ . The set of nodes in a complete materialization graph is actually the result of materialization. Thus, the procedure of materialization can be transformed into the construction of a complete materialization graph. In the remainder, we only consider complete materialization graphs and do not distinguish the term from the notion of ‘materialization graphs’. It should also be noted that there may exist several materialization graphs for a datalog program.

### 3.3. A Basic Parallel Algorithm

In this part, we propose a parallel algorithm (denoted by Algorithm  $A_{bsc}$ ) that constructs a materialization graph for a given datalog program. Our computational model is a PRAM (parallel, random-access machine) that is mostly used to analyze parallel complexity. This model allows us to introduce *the parallel assumption*: for any datalog program  $P = \langle R, I \rangle$ , any substitution of some atom and any rule instance in  $P^*$  can be mapped to a unique memory location; further, a one-to-one relation can be established between processors and rule instances. Under this assumption, a processor can check the applicability of its corresponding rule instance and access the state of an atom occurring in this rule instance in constant time. Algorithm  $A_{bsc}$  is then given as follows.

**Algorithm  $A_{\text{bsc}}$ .** Given a datalog program  $P = \langle R, \mathbf{I} \rangle$ , the algorithm returns a materialization graph  $\mathcal{G}$  of  $P$ . Suppose we have  $|P^*|$  processors, and each rule instantiation in  $P^*$  is assigned to one processor. Initially  $\mathcal{G}$  is empty. The following three steps are then performed:

(Step 1) Add all facts in  $\mathbf{I}$  to  $\mathcal{G}$ .

(Step 2) For each rule instantiation  $B_1, \dots, B_n \rightarrow H$ , if the body atoms are all in  $\mathcal{G}$  while  $H$  is not in  $\mathcal{G}$ , the corresponding processor adds  $H$  to  $\mathcal{G}$  and creates edges pointing from  $B_1, \dots, B_n$  to  $H$ .

(Step 3) If no processor can add more nodes and edges to  $\mathcal{G}$ , terminate, otherwise continue with Step 2.  $\square$

**Example 2.** We consider the datalog program  $P_{\text{ex}_1}$  in Example 1 again, and perform Algorithm  $A_{\text{bsc}}$  on it. Initially, all the facts  $(A(b), R(a_1, b), S(a_2, a_1), \dots, S(a_k, a_{k-1}))$  are added to the result  $\mathcal{G}_{\text{ex}_1}$  (Step 1). Then in different iterations of Step 2, the remaining nodes are added to  $\mathcal{G}_{\text{ex}_1}$  by different processors. For example a processor  $p$  is allocated a rule instantiation ' $R(a_2, b), A(b) \rightarrow A(a_2)$ '. Then, processor  $p$  adds  $A(a_2)$  to  $\mathcal{G}_{\text{ex}_1}$  after it checks that  $A(b)$  and  $R(a_2, b)$  are in  $\mathcal{G}_{\text{ex}_1}$ . Algorithm  $A_{\text{bsc}}$  halts when  $A(a_k)$  has been added to  $\mathcal{G}_{\text{ex}_1}$  (Step 3).

Lemma 1 shows the correctness of Algorithm  $A_{\text{bsc}}$  and that, for any datalog program  $P$ , Algorithm  $A_{\text{bsc}}$  always constructs a materialization graph with the minimum depth among all the materialization graphs of  $P$ . The detailed proofs of Lemma 1 and other lemmas and theorems can be found in the appendix.

**Lemma 1.** Given a datalog program  $P = \langle R, \mathbf{I} \rangle$ , we have

1. Algorithm  $A_{\text{bsc}}$  halts and returns a materialization graph  $\mathcal{G}$  of  $P$ ;
2.  $\mathcal{G}$  has the minimum depth among all the materialization graphs of  $P$ .

*Proof sketch.* This lemma can be proved by performing an induction on  $T_R^\omega(\mathbf{I})$ . The stage (see the related contents in Section 2) of  $P$  is the lower-bound of the depth of the materialization graphs. Based on the previous induction, one can further check that, for the materialization graph  $\mathcal{G}$  constructed by Algorithm  $A_{\text{bsc}}$ , its depth equals the depth of the stage.  $\square$

We now discuss the parallel complexity of Algorithm  $A_{\text{bsc}}$ . Given a class of datalog programs  $\mathbb{P}$  where a rule set is shared for each datalog program  $P = \langle R, \mathbf{I} \rangle$  in  $\mathbb{P}$ . Let  $e$ ,  $v$  and  $r$  represent the maximum arity of any predicate in  $\mathbf{I}$ , the maximum number of variables in any datalog rule, and the number of datalog rules respectively. We then have that the number of constants is at most  $|\mathbf{I}|e$ , and the number of all possible rule instances in  $P^*$  is at most  $r(|\mathbf{I}|e)^v$ . Note that  $e$ ,  $v$  and  $r$  depend only on the rule set  $R$  and not on the fact set  $\mathbf{I}$ . Thus, the memory space for storing the atoms and the rule instances is polynomial in the size of  $\mathbf{I}$ . This also means that the number of processors is polynomially bounded. The computing time of Step 1 and Step 3 occupy constant time (denoted by  $c_1$ ) because of parallelism. Since Algorithm  $A_{\text{bsc}}$  works under the parallel assumption, one iteration of Step 2 also costs constant time (denoted by  $c_2$ ). Thus, the whole computing time of Algorithm  $A_{\text{bsc}}$  turns out to be  $c_1 + \ell \cdot c_2$  where  $\ell$  denotes the number of iterations of Step 2.

An NC algorithm should meet two requirements: first, it works on a polynomial number of processors; second, it halts in poly-logarithmic time. As discussed above, Algorithm  $A_{\text{bsc}}$  meets the first requirement. If we want to restrict Algorithm  $A_{\text{bsc}}$  to be an NC algorithm, we can make



the number of iterations of Step 2 to be poly-logarithmically bounded. We use the symbol  $\psi$  to denote a poly-logarithmically bounded function. For any datalog program, if we restrict the number of iterations of Step 2 to be bounded by  $\psi(|\mathbf{I}|)$ , the computing time of Algorithm  $\mathbf{A}_{\text{bsc}}$  is  $c_1 + \psi(|\mathbf{I}|) \cdot c_2$ . With this restriction, Algorithm  $\mathbf{A}_{\text{bsc}}$  is an NC algorithm denoted by  $\mathbf{A}_{\text{bsc}}^\psi$ .

Based on  $\mathbf{A}_{\text{bsc}}^\psi$ , we can identify a class of datalog programs  $\mathcal{D}_{\mathbf{A}_{\text{bsc}}^\psi}$  such that all the datalog programs in it can be handled by  $\mathbf{A}_{\text{bsc}}^\psi$ . It is obvious that  $\mathcal{D}_{\mathbf{A}_{\text{bsc}}^\psi}$  is a PTD class. We use the following theorem to further show that this class can be captured based on the materialization graphs of the datalog programs in  $\mathcal{D}_{\mathbf{A}_{\text{bsc}}^\psi}$ .

**Theorem 1.** *For any datalog program  $P$ ,  $P \in \mathcal{D}_{\mathbf{A}_{\text{bsc}}^\psi}$  iff  $P$  has a materialization graph whose depth is upper-bounded by  $\psi(|P|)$ .*

*Proof sketch.* We can first prove that the number of iterations of Step 2 is actually the depth of the constructed materialization graph. This theorem then follows by considering Lemma 1.  $\square$

Consider Example 1 again. Let the integer  $k$  be a variable. We can get a class of datalog programs, denoted by  $\mathbb{P}_{\text{ex}_1}$ , where the rule set is  $\{R(x, y), A(y) \rightarrow A(x), S(x, y), R(y, z) \rightarrow R(x, z)\}$  and the fact set varies according to  $k$ . The algorithm  $\mathbf{A}_{\text{bsc}}^\psi$  is restricted in the sense that it cannot even work on the rather simple datalog program class  $\mathbb{P}_{\text{ex}_1}$ . Let  $\mathcal{G}_{\text{ex}_1}$  be some materialization graph corresponding to some datalog program in  $\mathbb{P}_{\text{ex}_1}$ . It can be checked that  $\text{depth}(\mathcal{G}_{\text{ex}_1}) = k$  for some  $k$ . This means that the depths of the materialization graphs are linearly bounded by  $k$ . On the other hand, the sizes of the datalog programs in  $\mathbb{P}_{\text{ex}_1}$  are polynomial in  $k$ . Thus, for any  $\psi$  that is poly-logarithmically bounded, we can always find a  $k$  large enough such that  $\mathbf{A}_{\text{bsc}}^\psi$  terminates without constructing a materialization graph for each datalog program in  $\mathbb{P}_{\text{ex}_1}$ . However, there indeed exists an NC algorithm that can handle  $\mathbb{P}_{\text{ex}_1}$ . We discuss this in the next part.

### 3.4. Optimizing Algorithm $\mathbf{A}_{\text{bsc}}$ via Single-Way Derivability

In this part, we optimize Algorithm  $\mathbf{A}_{\text{bsc}}$  such that  $P_{\text{ex}_1}$  can be handled. Based on the optimized variant of Algorithm  $\mathbf{A}_{\text{bsc}}$ , we can identify another PTD class.

We discuss our optimization based on a specific case in Example 3. We find that, in this kind of case, the construction of a materialization graph can be accelerated.

**Example 3.** *Consider the snapshot of Algorithm  $\mathbf{A}_{\text{bsc}}$  in Figure 2. A materialization graph  $\mathcal{G}$  is being constructed for some datalog program  $\langle R, \mathbf{I} \rangle$ . The nodes in the dashed box denote the ones that have been added to  $\mathcal{G}$ . In this snapshot,  $v_0$  has been newly added to  $\mathcal{G}$  in the  $\ell^{\text{th}}$  ( $\ell \geq 1$ ) iteration. Each of the nodes  $v_i$  ( $1 \leq i \leq k$ ) has at most one parent node not in  $\mathcal{G}$ , while  $v'$  has two parent nodes not in  $\mathcal{G}$ . All of the nodes  $v_i$  ( $1 \leq i \leq k$ ) and  $v'$  would be added to  $\mathcal{G}$  afterwards.*

In Example 3,  $v_k$  would be added to  $\mathcal{G}$  after at least  $k$  iterations by performing Algorithm  $\mathbf{A}_{\text{bsc}}$ . We can check that each edge  $(v_{i-1}, v_i)$  ( $1 \leq i \leq k$ ) in this stage adheres to the condition: *if the parent node  $v_{i-1}$  is derivable, the child node  $v_i$  is derivable* (this is because for a node  $v_i$ , the node  $v_{i-1}$  is the only parent node that has not been added to  $\mathcal{G}$ ). We call this condition the *single-way derivability condition*, which intuitively says that the derivability of a node only depends on one of its parent nodes. Observe that node  $v_k$  is reachable from node  $v_0$  through the path  $\tau = (v_0, v_1, \dots, v_k)$  where each edge  $(v_{i-1}, v_i)$  ( $1 \leq i \leq k$ ) satisfies the single-way derivability condition. We call such a path a *single-way derivable (SWD) path*. Further, the starting node  $v_0$  in  $\tau$  has been added to  $\mathcal{G}$ ; in other words,  $v_0$  is derivable. Thus, all of the nodes  $v_i$  ( $1 \leq i \leq k$ ) can be added to  $\mathcal{G}$  immediately. Based on the above idea, we optimize Algorithm  $\mathbf{A}_{\text{bsc}}$  by checking

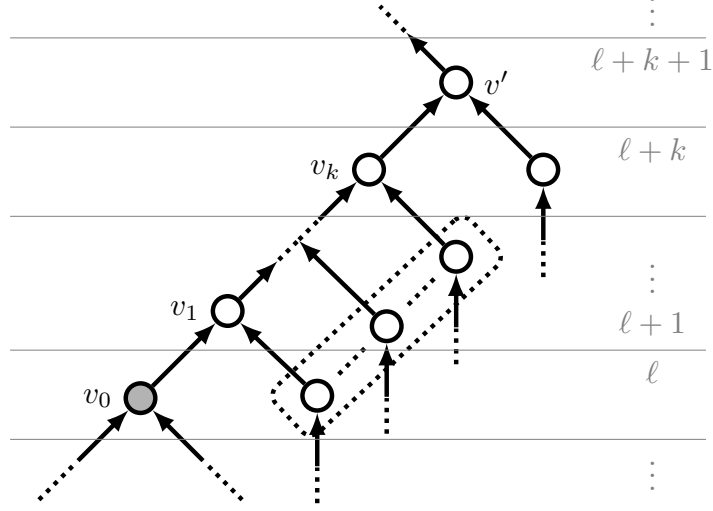


Figure 2: A partial materialization graph

whether such an SWD path exists for some node  $v$ . If there is an SWD path for node  $v$ , this node can be added to  $\mathcal{G}$  due to the single-way derivability condition.

For Example 3, there exists an SWD path  $(v_0, \dots, v_i)$  for each node  $v_i$  ( $1 \leq i \leq k$ ), hence, all of these nodes can be added to  $\mathcal{G}$  right after  $v_0$ . On the other hand, node  $v'$  has no SWD path since it has two parent nodes not in  $\mathcal{G}$ .

We next discuss how to determine the existence of SWD paths. Note that, SWD paths require us to describe the reachability between two nodes. To this extent, we use a binary transitive relation  $\text{rch} \subseteq T_R^\omega(\mathbf{I}) \times T_R^\omega(\mathbf{I})$ , e.g.,  $\text{rch}(v_1, v_2)$  means that  $v_2$  is reachable from  $v_1$ . In each iteration of Step 2 in Algorithm  $\mathbf{A}_{\text{bsc}}$ , we further compute a  $\text{rch}$  relation (denoted by  $S_{\text{rch}}$ ) by performing the following process:

(†) For each rule instantiation of the form  $B_1, \dots, B_i, \dots, B_n \rightarrow H$  such that  $H$  has not been added to  $\mathcal{G}$ :

1. if the all body atoms  $B_1, \dots, B_n$  have been added to  $\mathcal{G}$ , put  $\text{rch}(B_1, H), \dots, \text{rch}(B_n, H)$  in  $S_{\text{rch}}$ ;
2. if  $B_i$  is the only node in the body that has not been added to  $\mathcal{G}$ , put  $\text{rch}(B_i, H)$  in  $S_{\text{rch}}$ .  $\square$

We then compute the transitive closure (denoted by  $S_{\text{rch}}^*$ ) with respect to  $S_{\text{rch}}$ . Based on the transitive closure, we can perform the following optimization: for a node  $v$ , if there is a relation  $\text{rch}(v', v) \in S_{\text{rch}}^*$  such that  $v'$  has been added to  $\mathcal{G}$  and  $v$  has an SWD path, then  $v$  can be added to  $\mathcal{G}$ . The following algorithm applies this optimization strategy.

**Algorithm OPT.** The algorithm requires two inputs: a datalog program  $P = \langle R, \mathbf{I} \rangle$  and a (partial) materialization graph  $\mathcal{G}$  that is being constructed from  $P$ . The following steps are performed:

- (i) Compute the  $\text{rch}$  relation  $S_{\text{rch}}$  by following the above process (see (†)).

- (ii) Compute the transitive closure  $S_{rch}^*$  of  $S_{rch}$ .
- (iii) Update  $\mathcal{G}$  as follows: for any  $\text{rch}(B_i, H) \in S_{rch}$  that corresponds to ' $B_1, \dots, B_i, \dots, B_n \rightarrow H$ ' and there exists a node  $B'$  such that  $\text{rch}(B', B_i) \in S_{rch}^*$  and  $B'$  is in  $\mathcal{G}$ ; if  $H$  is not in  $\mathcal{G}$  or  $H$  is in  $\mathcal{G}$  but has no parent pointing to it, add  $H$  and  $B_i$  (if  $B_i$  is not in  $\mathcal{G}$ ) to  $\mathcal{G}$ , and create the edges  $e(B_1, H), \dots, e(B_n, H)$  in  $\mathcal{G}$ . Do nothing for other statements  $\text{rch}(B_j, H) \in S_{rch}$ .  $\square$

It is well known that there is an NC algorithm for computing the transitive closure [24]. Based on this result and Algorithm OPT, we propose an optimized variant of Algorithm  $A_{\text{bsc}}$ :

**Algorithm  $A_{\text{opt}}$ .** Given a datalog program  $P = \langle R, \mathbf{I} \rangle$ , the algorithm returns a materialization graph  $\mathcal{G}$  of  $P$ . Suppose we have  $|P^*|$  processors, and each rule instantiation in  $P^*$  is assigned to one processor. Initially  $\mathcal{G}$  is empty. The following steps are then performed:

(Step 1) Add all facts in  $\mathbf{I}$  to  $\mathcal{G}$ .

(Step 2) Compute  $S_{rch}$  by performing (i) in Algorithm OPT; use an NC algorithm to compute the transitive closure  $S_{rch}^*$  (see (ii) in Algorithm OPT); update  $\mathcal{G}$  by performing (iii) in Algorithm OPT.

(Step 3) If no node has been added to  $\mathcal{G}$  (in Step 2), terminate, otherwise iterate Step 2.  $\square$

It should be noted that there has to be an SWD path for any derivable node in some iteration when performing Algorithm  $A_{\text{opt}}$ . The following lemma shows the correctness of Algorithm  $A_{\text{opt}}$ .

**Lemma 2.** *Given a datalog program  $P = \langle R, \mathbf{I} \rangle$ ,  $A_{\text{opt}}$  halts and outputs a materialization graph  $\mathcal{G}$  of  $P$ .*

**Example 4.** *We perform Algorithm  $A_{\text{opt}}$  on the datalog program  $P_{\text{ex1}}$  in Example 1. Initially,  $R(a_1, b)$  is in the materialization graph  $\mathcal{G}_{\text{ex1}}$ . In the first iteration of Step 2, all the rule instantiations are in two kinds of forms: ' $R(a_i, b), A(b) \rightarrow A(a_i)$ ' and ' $S(a_i, a_{i-1}), R(a_{i-1}, b) \rightarrow R(a_i, b)$ ',  $2 \leq i \leq k$ ,  $S_{rch}$  is the set  $\{\text{rch}(R(a_{i-1}, b), R(a_i, b)) \mid 2 \leq i \leq k\} \cup \{\text{rch}(R(a_i, b), A(a_i)) \mid 1 \leq i \leq k\}$ . In the transitive closure of  $S_{rch}$ , one can check that  $\text{rch}(R(a_1, b), R(a_i, b)), \text{rch}(R(a_1, b), A(a_i)) \in S_{rch}^*$ ,  $2 \leq i \leq k$ . Thus,  $R(a_i, b)$  and  $A(a_i)$ ,  $2 \leq i \leq k$ , can all be added to  $\mathcal{G}_{\text{ex1}}$  in the first iteration of Step 2.*

We now analyse the parallel complexity of Algorithm  $A_{\text{opt}}$ . We use the same symbols  $e$ ,  $v$  and  $r$  that are used for analyzing the complexity of Algorithm  $A_{\text{bsc}}$  (see Section 3.3). Similarly to the analysis for Algorithm  $A_{\text{bsc}}$ , it can be checked that the number of processors is polynomial in  $|\mathbf{I}|$ , i.e.,  $r(|\mathbf{I}|e)^v$ . We now consider the computing time of Algorithm  $A_{\text{opt}}$ . It is obvious that Step 1 and Step 3 occupy constant time (denoted by  $c_1$ ). In Step 2, the phases of computing  $S_{rch}$  and updating  $\mathcal{G}$  also take constant time (denoted by  $c_2$ ) under the parallel assumption. Recall that the number of constants is at most  $|\mathbf{I}|e$  (see Section 3.3). Let  $w$  and  $p$  represent the maximum arity of any predicate and the number of predicates, respectively. We have that the number of all possible substitutions of atoms is at most  $p(|\mathbf{I}|e)^w$ . The relation  $S_{rch}$  consists of pairs of the atoms. Thus, the size of  $S_{rch}$  is bounded by  $p^2(|\mathbf{I}|e)^{2w}$ . It can be checked that the computing time of the NC algorithm for computing the transitive closure of  $S_{rch}$  is poly-logarithmic in the size of  $\mathbf{I}$ , more precisely, it is upper bounded by  $\log^2(p^2(|\mathbf{I}|e)^{2w}) (= 4\log^2(p) + 4w^2\log^2(|\mathbf{I}|e) + 8w\log(p)\log(|\mathbf{I}|e))$ . We now have that the total computing time of Algorithm  $A_{\text{opt}}$  is  $c_1 + \ell c_3 + \ell c_4 \log(|\mathbf{I}|e) + \ell c_5 \log^2(|\mathbf{I}|e)$  where  $c_3 = c_2 + 4\log^2(p)$ ,  $c_4 = 8w\log(p)$ ,  $c_5 = 4w^2$  and  $\ell$  denotes the number of iterations of Step 2.

Algorithm  $A_{\text{opt}}$  can be restricted to an NC algorithm analogously to the process for Algorithm  $A_{\text{bsc}}$ . That is the number of iterations of Step 2 is poly-logarithmically bounded, i.e., it is bounded by  $\psi(|I|)$  where  $\psi$  is a poly-logarithmic function. In this way, the computing time of Algorithm  $A_{\text{bsc}}$  turns out to be  $c_1 + c_3\psi(|I|) + c_4\psi(|I|)\log(|I|e) + c_5\psi(|I|)\log^2(|I|e)$ , which is still poly-logarithmical in the size of  $I$ . We denote the NC variant of Algorithm  $A_{\text{opt}}$  by  $A_{\text{opt}}^\psi$ .

Based on  $A_{\text{opt}}^\psi$ , we can identify a PTD class  $\mathcal{D}_{A_{\text{opt}}^\psi}$ . Further, we have the following corollary, which implies that  $A_{\text{opt}}^\psi$  performs better than  $A_{\text{bsc}}^\psi$  in terms of computing time.

**Corollary 1.** *For any poly-logarithmically bounded function  $\psi$ , we have that  $\mathcal{D}_{A_{\text{bsc}}^\psi} \subseteq \mathcal{D}_{A_{\text{opt}}^\psi}$ .*

*Proof sketch.* Suppose  $P \in \mathcal{D}_{A_{\text{bsc}}^\psi}$ . According to Theorem 1, the depth of the materialization graph  $\mathcal{G}$  constructed by  $A_{\text{bsc}}^\psi$  is upper-bounded by  $\psi(|P|)$ . It is obvious that the number of nodes in each path of  $\mathcal{G}$  is also upper-bounded by  $\psi(|P|)$ . According to the optimization strategy applied in  $A_{\text{opt}}$ , if  $\mathcal{G}$  can be constructed by  $A_{\text{opt}}$ , the number of iterations of  $A_{\text{opt}}$  has to be upper-bounded by  $\psi(|P|)$ ; if  $\mathcal{G}$  is not the materialization graph constructed by  $A_{\text{opt}}$ , then there has to exist another materialization graph  $\mathcal{G}'$  constructed by  $A_{\text{opt}}$  and  $\mathcal{G}'$  has a smaller depth compared to  $\mathcal{G}$ .  $\square$

#### 4. Parallel Tractability of Ontology Materialization in OWL

In this section, we study the issue of parallel tractability for materialization of DL-Lite and DHL (DHL( $\circ$ )) ontologies based on Algorithm  $A_{\text{opt}}$ . We show that, for any DL-Lite<sub>core</sub> or DL-Lite<sub>R</sub> ontology  $\mathcal{O}$ , there exists a poly-logarithmically bounded function  $\psi$  such that Algorithm  $A_{\text{opt}}^\psi$  can handle materialization of  $\mathcal{O}$ . However, for DHL and DHL( $\circ$ ), there exist ontology classes such that Algorithm  $A_{\text{opt}}^\psi$  does not work. We illustrate the reason why Algorithm  $A_{\text{opt}}^\psi$  cannot always work by studying specific cases. Further, we propose to restrict the usage of DHL and DHL( $\circ$ ) in order to achieve parallel tractability of materialization.

##### 4.1. Materialization of DL-Lite Ontologies via Algorithm $A_{\text{opt}}$

In this part, we show how to use Algorithm  $A_{\text{opt}}$  to handle DL-Lite materialization and analyze its parallel tractability. Based on the analysis, we have that, for any DL-Lite<sub>core</sub> or DL-Lite<sub>R</sub> ontology there always exists an SWD path for each atom of the form  $A(a)$  or  $R(a, b)$ . In other words, all atoms of the forms  $A(a)$  and  $R(a, b)$  can be added to the constructed materialization graph in the first iteration of Step 2 by performing Algorithm  $A_{\text{opt}}$ .

In order to show how Algorithm  $A_{\text{opt}}$  handles DL-Lite materialization, we use the following example.

**Example 5.** *Given a DL-Lite ontology  $\mathcal{O}_{\text{ex}_5}$  where the TBox and RBox contain the following axioms:  $A \sqsubseteq B_1$ ,  $A \sqsubseteq B_2$ ,  $B_1 \sqcap B_2 \sqsubseteq \perp$ ,  $\exists R.B_2 \sqsubseteq \perp$ ,  $Q \sqsubseteq S^-$ ,  $S \sqsubseteq R$ ; its ABox contains an assertion  $Q(a, b)$ . We denote the corresponding datalog program of  $\mathcal{O}_{\text{ex}_5}$  by  $P_{\text{ex}_5} = \langle R, I \rangle$ , where  $R$  contains the rules that are transformed from the above axioms;  $I$  contains  $Q(a, b)$  as the only fact. The unique materialization graph of  $P_{\text{ex}_5}$  is denoted by  $\mathcal{G}_{\text{ex}_5}$  (see Figure 3).*

Consider performing Algorithm  $A_{\text{opt}}$  on the ontology  $\mathcal{O}_{\text{ex}_5}$  in Example 5. (I) First, Algorithm  $A_{\text{opt}}$  adds all ABox assertions (only  $Q(a, b)$  here) to the initially empty graph  $\mathcal{G}_{\text{ex}_5}$  (Step 1 of Algorithm  $A_{\text{opt}}$ ). (II) In the first iteration of Step 2, Algorithm  $A_{\text{opt}}$  checks that all of the nodes  $A(a)$ ,  $S(b, a)$ ,  $B_1(a)$ ,  $B_2(a)$  and  $R(b, a)$  have corresponding SWD paths starting from  $Q(a, b)$ .

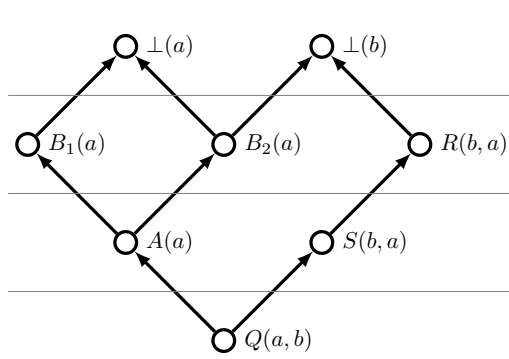


Figure 3: The materialization graph of  $\mathcal{O}_{\text{ex}_5}$

Thus, Algorithm  $\mathbf{A}_{\text{opt}}$  adds these nodes to  $\mathcal{G}_{\text{ex}_5}$  immediately. We take an example of node  $B_1(a)$ , for which an SWD path  $(Q(a, b), A(a), B_1(a))$  exists. When updating  $\mathcal{G}_{\text{ex}_5}$  by adding  $B_1(a)$  to it, Algorithm  $\mathbf{A}_{\text{opt}}$  first checks whether the parent node  $A(a)$  of  $B_1(a)$  has been in  $\mathcal{G}_{\text{ex}_5}$ ; if  $A(a)$  is already in  $\mathcal{G}_{\text{ex}_5}$ ,  $B_1(a)$  is added to  $\mathcal{G}_{\text{ex}_5}$  by creating an edge pointing from  $A(a)$  to  $B_1(a)$ ; if  $A(a)$  has not been added to  $\mathcal{G}_{\text{ex}_5}$ , Algorithm  $\mathbf{A}_{\text{opt}}$  adds  $A(a)$  ( $B_1(a)$ , respectively) to  $\mathcal{G}_{\text{ex}_5}$  and creates an edge pointing from  $Q(a, b)$  to  $A(a)$  (an edge pointing from  $A(a)$  to  $B_1(a)$ , respectively).<sup>3</sup> The other nodes  $A(a)$ ,  $S(b, a)$ ,  $B_2(a)$  and  $R(b, a)$  are processed similarly. (III) The two nodes  $\perp(a)$  and  $\perp(b)$  have no SWD path in the first iteration of Step 2. They are left to be processed in the second iteration. (IV) Finally, Algorithm  $\mathbf{A}_{\text{opt}}$  finishes constructing  $\mathcal{G}_{\text{ex}_5}$  after two iterations (Step 3).

From the above example, we can observe that SWD paths exist for all nodes except  $\perp(a)$  and  $\perp(b)$  in the first iteration of Algorithm  $\mathbf{A}_{\text{opt}}$ . Further, we give the following lemma that holds for any  $\text{DL-Lite}_{\text{core}}$  or  $\text{DL-Lite}_{\mathcal{R}}$  ontology.

**Lemma 3.** *For any  $\text{DL-Lite}_{\text{core}}$  or  $\text{DL-Lite}_{\mathcal{R}}$  ontology  $\mathcal{O}$ , there exists a materialization graph  $\mathcal{G}$  such that each atom of the form  $A(x)$  ( $A \neq \perp$ ) or  $R(x, y)$  in  $\mathcal{G}$  has an SWD path.*

The above lemma guarantees that all atoms of the form  $A(x)$  ( $A \neq \perp$ ) or  $R(x, y)$  have to be added to the constructed materialization graph in the first iteration of Step 2 by applying Algorithm  $\mathbf{A}_{\text{opt}}$ . For each atom of the form  $\perp(x)$ , there may not exist an SWD path in the first iteration of Step 2, since its derivability depends on its two parent nodes (see nodes  $\perp(a)$  and  $\perp(b)$  in  $\mathcal{G}_{\text{ex}_5}$  of Example 5). On the other hand, according to the syntax of DL-Lite,  $\perp$  does not occur on the left-hand side of any axiom. In other words, an atom of the form  $\perp(x)$  cannot be the parent of any other node. This allows for adding all atoms of the form  $\perp(x)$  to the constructed materialization graph in at most two iterations of Step 2 by applying Algorithm  $\mathbf{A}_{\text{opt}}$ . Based on the above discussion, for any DL-Lite ontology  $\mathcal{O}$ , there always exists a poly-logarithmically bounded function  $\psi$  such that Algorithm  $\mathbf{A}_{\text{opt}}^\psi$  can handle the materialization of  $\mathcal{O}$ ; more precisely, we can set that  $\psi = 2$ . This result is consistent with that by Calvanese et al. [15]. Formally, we use  $\mathcal{D}_{\text{dl-lite}}$  to denote the set of all  $\text{DL-Lite}_{\text{core}}$  and  $\text{DL-Lite}_{\mathcal{R}}$  ontologies and give the following theorem.

**Theorem 2.** *There exists a poly-logarithmically bounded function  $\psi$  such that  $\mathcal{D}_{\text{dl-lite}} \subseteq \mathcal{D}_{\mathbf{A}_{\text{opt}}^\psi}$ .*

<sup>3</sup>These two cases may happen simultaneously, since node  $A(a)$  and node  $B_1(a)$  are being processed in parallel.

#### 4.2. Parallel Tractability of DHL Materialization

In this part, we study whether Algorithm  $A_{\text{opt}}^\psi$  can handle DHL ontologies. Unfortunately there exist DHL ontology classes such that Algorithm  $A_{\text{opt}}^\psi$  does not work for any poly-logarithmically bounded function  $\psi$ . In the following, we first give such a case to illustrate the reason why Algorithm  $A_{\text{opt}}^\psi$  cannot work. Based on the analysis of this case, we propose to restrict the usage of DHL in order to achieve parallel tractability of materialization.

We find that, an unlimited usage of axioms of the form  $B_1 \sqcap B_2 \sqsubseteq A$  makes it impossible for Algorithm  $A_{\text{opt}}$  to construct a materialization graph in a poly-logarithmic number of iterations of Step 2. We use the following example to illustrate this.

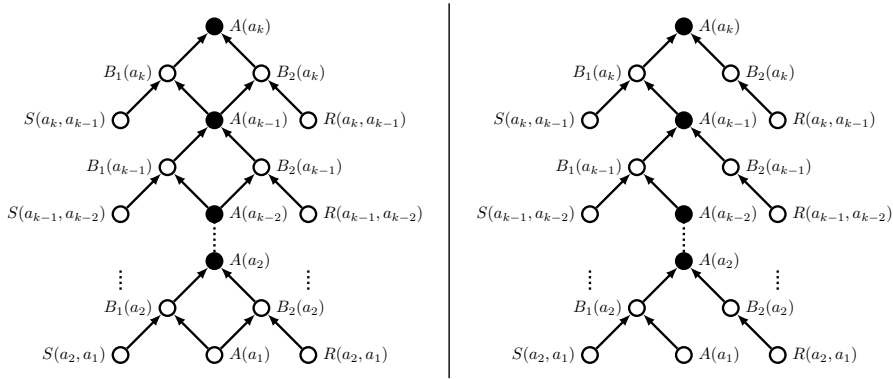


Figure 4: The materialization graph of  $O_{\text{ex}_6}$  (left) and of  $O_{\text{ex}_7}$  (right)

**Example 6.** Given a DHL ontology  $O_{\text{ex}_6}$  where the TBox contains three axioms:  $B_1 \sqcap B_2 \sqsubseteq A$ ,  $\exists S.A \sqsubseteq B_1$  and  $\exists R.A \sqsubseteq B_2$ ; the ABox is  $\{S(a_i, a_{i-1}), R(a_i, a_{i-1}), A(a_1)\}$  for  $2 \leq i \leq k$  and  $k$  an integer greater than 2. We denote the corresponding datalog program of  $O_{\text{ex}_6}$  by  $P_{\text{ex}_6} = \langle R, I \rangle$ , where  $R$  contains three rules: ' $B_1(x), B_2(x) \rightarrow A(x)$ ', ' $S(x, y), A(y) \rightarrow B_1(x)$ ' and ' $R(x, y), A(y) \rightarrow B_2(x)$ '. The materialization graph of  $P_{\text{ex}_6}$  constructed by  $A_{\text{opt}}$  is denoted by  $\mathcal{G}_{\text{ex}_6}$  (see Figure 4 (left)).

One can check that  $\mathcal{G}_{\text{ex}_6}$  is the unique materialization graph of  $P_{\text{ex}_6}$ . Observe that there exists a path (e.g.,  $A(a_1), B_1(a_2), A(a_2), \dots, A(a_k)$ ) between  $A(a_1)$  and  $A(a_k)$ . When performing Algorithm  $A_{\text{opt}}$  on  $P_{\text{ex}_6}$ , it can be checked that each node of the form  $A(a_i)$  (filled with black color, and  $2 \leq i \leq k$ ) has no SWD path until the  $i^{\text{th}}$  iteration; in the  $i^{\text{th}}$  iteration, the parent nodes of  $A(a_i)$  have been added to  $\mathcal{G}_{\text{ex}_6}$ , and,  $A(a_i)$  can also be added to  $\mathcal{G}_{\text{ex}_6}$ . Similar to the datalog program class  $\mathbb{P}_{\text{ex}_1}$ , we can also get a datalog program class  $\mathbb{P}_{\text{ex}_6}$  for the ontology  $O_{\text{ex}_6}$  when  $k$  is a variable. Based on the above analysis, Algorithm  $A_{\text{opt}}$  cannot complete the materialization for the datalog programs of  $\mathbb{P}_{\text{ex}_6}$  in a poly-logarithmic number of iterations. The intuitive reason is that at least two paths exist from  $A(a_1)$  to  $A(a_k)$ . These paths *twist* mutually and share the same joint nodes (see the black nodes). This invalidates the optimization used in Algorithm  $A_{\text{opt}}$ . That is, for each node  $A(a_i)$ ,  $2 \leq i \leq k$ , until its parents ( $B_1(a_i)$  and  $B_2(a_i)$ ) are added to  $\mathcal{G}_{\text{ex}_6}$ , there would not exist an available SWD path for  $A(a_i)$ . We use the term '*path twisting*' to refer to such cases.

In order to make Algorithm  $A_{\text{opt}}$  terminate in a poly-logarithmic number of iterations, we consider restricting the usage of axioms of the form  $B_1 \sqcap B_2 \sqsubseteq A$  to avoid ‘path twisting’. An intuitive idea is to ensure that *there is only one path between each two atoms of the form  $A(x)$  generated from the rules corresponding to (T1)*. We explain it by using the following example where the ontology is modified from that in Example 6.

**Example 7.** Consider an ontology  $O_{\text{ex7}}$  where the TBox contains three axioms:  $B_1 \sqcap B_2 \sqsubseteq A$ ,  $\exists S.A \sqsubseteq B_1$  and  $B_3 \sqsubseteq B_2$ ; the ABox is  $\{S(a_i, a_{i-1}), B_3(a_i), A(a_1)\}$  for  $2 \leq i \leq k$  and  $k$  an integer greater than 2. We denote the corresponding datalog program by  $P_{\text{ex7}}$  where the rule set contains: ‘ $B_1(x), B_2(x) \rightarrow A(x)$ ’, ‘ $S(x, y), A(y) \rightarrow B_1(x)$ ’ and ‘ $B_3(x) \rightarrow B_2(x)$ ’.  $P_{\text{ex7}}$  has a unique materialization graph denoted by  $\mathcal{G}_{\text{ex7}}$  (see Figure 4 (right)).

In the above example, for the axiom  $B_1 \sqcap B_2 \sqsubseteq A$ , all derived atoms of the form  $B_2(x)$  must not be child nodes of an atom  $A(y)$  for some  $y$ . This ensures that only one path exists between each two nodes among  $A(a_2), \dots, A(a_k)$ . Further, when constructing  $\mathcal{G}_{\text{ex7}}$ , Algorithm  $A_{\text{opt}}$  can terminate after two iterations of Step 2. Specifically, in the first iteration, Algorithm  $A_{\text{opt}}$  adds all of the nodes  $B_3(a_i)$  and  $B_2(a_i)$ ,  $2 \leq i \leq k$ , to  $\mathcal{G}_{\text{ex7}}$  since they have corresponding SWD paths; after that, all other nodes can be added to  $\mathcal{G}_{\text{ex7}}$  in the second iteration (because each node has an SWD path). Motivated by this example, we consider restricting the usage of the axioms  $B_1 \sqcap B_2 \sqsubseteq A$  such that all atoms of the form  $B_1(x)$  or  $B_2(x)$  cannot be generated by an atom  $A(y)$  for some  $y$ . To this end, we first define *simple concepts* as follows:

**Definition 3.** Given an ontology  $O = \langle \mathcal{T}, \mathcal{R}, \mathcal{A} \rangle$ , a concept  $A \in \mathcal{CN}$  is *simple*, if (1)  $A$  does not occur on the right-hand side of some axiom; or (2)  $A$  satisfies the following conditions:

1. for each  $B \sqsubseteq A \in \mathcal{T}$ ,  $B$  is simple;
2. for each  $\exists R.B \sqsubseteq A \in \mathcal{T}$ ,  $B$  is simple;
3. there is no axiom of the form  $B_1 \sqcap B_2 \sqsubseteq A$  in  $\mathcal{T}$ .

Based on simple concepts, we restrict DHL ontologies such that, in all axioms of the form  $B_1 \sqcap B_2 \sqsubseteq A$ , at least one concept of  $B_1$  and  $B_2$  should be a simple concept (we call this the *simple-concept restriction*). Intuitively, for such restricted DHL ontologies, the situation of ‘path twisting’ does not happen. This is because, for each axiom of the form  $B_1 \sqcap B_2 \sqsubseteq A$  such that, w.l.o.g.,  $B_1$  is a simple concept, none of the ancestors of  $B_1(x)$  for some  $x$  is generated from the rules corresponding to (T1).

**Example 8.** In the ontology of Example 6, all of  $A$ ,  $B_1$  and  $B_2$  are non-simple concepts. In the ontology of Example 7,  $A$  and  $B_1$  are non-simple concepts, while  $B_3$  and  $B_2$  are simple concepts. Further, it can be checked that the ontology of Example 7 follows the simple-concept restriction and can be handled by Algorithm  $A_{\text{opt}}^\psi$  for some poly-logarithmic function  $\psi$ .

We define the following class of DHL ontologies based on the above restriction and give Theorem 3 to show that any DHL ontology that satisfies the simple-concept restriction can be handled by Algorithm  $A_{\text{opt}}^\psi$  for some poly-logarithmic function  $\psi$ .

**Definition 4.** Let  $\mathcal{D}_{\text{dhl}}$  be a class of datalog programs where each program is rewritten from a DHL ontology that follows the condition that, for all axioms of the form  $A_1 \sqcap A_2 \sqsubseteq B$ , at least one concept of  $A_1$  and  $A_2$  should be a simple concept.

**Theorem 3.** There exists a poly-logarithmically bounded function  $\psi$  s.t.  $\mathcal{D}_{\text{dhl}} \subseteq \mathcal{D}_{A_{\text{opt}}^\psi}$ .

### 4.3. Parallel Tractability of DHL( $\circ$ ) Materialization

In this part, we study materialization tractable in parallel for DHL( $\circ$ ) ontologies. In addition to the rules in DHL, we also have to consider complex RIAs (R4). We next show that complex RIAs may also cause the situation of ‘path twisting’. Consider the following example:

**Example 9.** Given a DHL( $\circ$ ) ontology  $\mathcal{O}_{ex_9}$  where the TBox is empty; the RBox  $\mathcal{R}$  contains three axioms:  $R_1 \circ R_2 \sqsubseteq R$ ,  $R_3 \circ R \sqsubseteq R_1$  and  $R \circ R_4 \sqsubseteq R_2$ ; the ABox  $\mathcal{A}$  is  $\{R(a_1, a_1), R_3(a_i, a_{i-1}), R_4(a_{i-1}, a_i)\}$  for  $2 \leq i \leq k$  and  $k$  an integer greater than 2. The corresponding datalog program  $P_{ex_5}$  contains three rules: ‘ $R_1(x, y), R_2(y, z) \rightarrow R(x, z)$ ’, ‘ $R_3(x, y), R(y, z) \rightarrow R_1(x, z)$ ’ and ‘ $R(x, y), R_4(y, z) \rightarrow R_2(x, z)$ ’. The materialization graph of  $P_{ex_5}$  constructed by Algorithm  $A_{opt}$  is denoted by  $\mathcal{G}_{ex_9}$ .

One can check that the materialization graph  $\mathcal{G}_{ex_9}$  has the same shape as that of  $\mathcal{G}_{ex_6}$  in Figure 4. A twisted path exists in  $\mathcal{G}_{ex_9}$  involving  $R(a_i, a_i)$ ,  $2 \leq i \leq k$ , as the joint nodes. Further, all the roles  $R_1, R_2, R_3, R_4$  and  $R$  in this example are non-transitive roles.

Inspired by what we do for axioms  $B_1 \sqcap B_2 \sqsubseteq A$ , we require that, for all axioms of the form  $R_1 \circ R_2 \sqsubseteq R$ , if  $R$  is not a transitive role and no transitive role  $S$  exists such that  $R \sqsubseteq_* S$ , then, at least one of  $R_1$  and  $R_2$  is a *simple role*.<sup>4</sup> We now consider such an axiom  $R_1 \circ R_2 \sqsubseteq R$  (denoted by  $\alpha_1$ ) where  $R$  is a transitive role. That is, we also have  $R \circ R \sqsubseteq R$  (denoted by  $\alpha_2$ ). By replacing  $R$  on the left-hand side of  $\alpha_2$  using  $R_1$  and  $R_2$ , we can get a complex RIA in the form of  $R_1 \circ R_2 \circ R_1 \circ R_2 \sqsubseteq R$  (denoted by  $\alpha_3$ ). If one of  $R_1$  and  $R_2$  is not a simple role, the corresponding rule of  $\alpha_3$  may also lead to ‘path twisting’.<sup>5</sup> This can be explained as follows: Without loss of the generality,  $R_2$  is a simple role, while  $R_1$  is not. For some atom  $R(x, y)$ , it may depend on two different nodes of the predicate  $R_1$  through the corresponding rule of  $\alpha_3$ . A similar analysis applies to the cases of  $\alpha_1$  where  $R$  is not a transitive role, while another transitive role  $S$  exists such that  $R \sqsubseteq_* S$ . That is, we can obtain a complex RIA of the form  $R_1 \circ R_2 \circ R_1 \circ R_2 \sqsubseteq S$ . Further, the situation of path twisting also exists. To tackle the above issue, we require both of  $R_1$  and  $R_2$  in  $\alpha_1$  to be simple roles (we call the above restriction for transitive and non-transitive roles the *simple-role restriction*). Combined with the simple-concept restriction, we define a class of DHL( $\circ$ ) ontologies as follows:

**Definition 5.**  $\mathcal{D}_{dhl(\circ)}$  is a class of datalog programs where each program is rewritten from a DHL( $\circ$ ) ontology and the following conditions are satisfied:

1. for all axioms of the form  $A_1 \sqcap A_2 \sqsubseteq B$ , at least one concept of  $A_1$  and  $A_2$  is a simple concept;
2. for all axioms of the form  $R_1 \circ R_2 \sqsubseteq R$ , if there does exists a transitive role  $S$  such that  $R \sqsubseteq_* S$ , then both  $R_1$  and  $R_2$  are simple roles; otherwise at least one of  $R_1$  and  $R_2$  is a simple role.

**Example 10.** For the ontology  $\mathcal{O}_{ex_7}$  in Example 9, all of the roles  $R_1, R_2$  and  $R$  are non-simple roles. Thus,  $\mathcal{O}_{ex_7}$  does not follow the simple-role restriction because of  $R_1 \circ R_2 \sqsubseteq R$ . Consider the ontology  $\mathcal{O}_{ex_1}$  in Example 1 again. The role  $R$  is a non-simple role, while  $S$  is a simple role. Thus,  $\mathcal{O}_{ex_1}$  follows the simple-role restriction. All the implicit nodes in  $\mathcal{G}_{ex_1}$  have corresponding SWD paths in the first iteration. Thus, ‘path twisting’ cannot occur when materializing  $\mathcal{O}_{ex_1}$  by Algorithm  $A_{opt}$ .

We further give Theorem 4 to show that Algorithm  $A_{opt}^\psi$  can handle all datalog programs in  $\mathcal{D}_{dhl(\circ)}$  for some poly-logarithmic function  $\psi$ .

**Theorem 4.** There exists a poly-logarithmically bounded function  $\psi$  s.t.  $\mathcal{D}_{dhl(\circ)} \subseteq \mathcal{D}_{A_{opt}^\psi}$ .

<sup>4</sup>See the definition of a simple role in Section 2.

<sup>5</sup>Obviously, applying the rules of  $\alpha_1$  and  $\alpha_2$  separately has the same effect to that of only applying the rule of  $\alpha_3$ .



Table 2: RDFS statements and the corresponding DL axioms

	RDFS statements	Axiom
(1)	$P \text{ rdfs:domain } C$	$\top \sqsubseteq \forall P^-.C$
(2)	$P \text{ rdfs:range } C$	$\top \sqsubseteq \forall P.C$
(3)	$B \text{ rdfs:subClassOf } C$	$B \sqsubseteq C$
(4)	$R \text{ rdfs:subPropertyOf } S$	$R \sqsubseteq S$
(5)	$a \text{ rdf:type } C$	$C(a)$
(6)	$a R b$	$R(a, b)$

#### 4.4. Parallel Tractability of Reasoning over RDFS Ontologies

In this part, we discuss parallel tractability of reasoning over RDFS ontologies. Although RDFS is not directly based on a description logic, it can (partly) be described by a set of DL axioms [16]. The correspondence between RDFS statements and DL axioms is listed in Table 2. We call RDFS statements of the form (1–4) *the schema data* (which correspond to TBox axioms), statements of the form (5) and (6) *the instance data* (which correspond to ABox axioms).

The original rule set for RDFS reasoning is given by Hayes [25]. By applying the rules in this rule set, new schema data could be derived. Thus, the reasoning task of RDFS is different from the task of materialization. Further, an RDFS ontology allows statements about blank nodes which act like variables. This possibly leads to infinite entailments. Additionally, the original rule set is incomplete with the RDF restriction that blank nodes cannot occur in predicate positions [20], however, by allowing blank nodes in predicate position the rule set turns out to be complete. The computational complexity of RDFS reasoning is NC-complete [20], which is not considered to be tractable in parallel.

The situation of path twisting may also happen when conducting reasoning on RDFS ontologies. Suppose  $R_{\text{subp}}$  denotes the RDFS built-in property `rdfs:subPropertyOf`. We define three further properties  $R_1$ ,  $R_2$  and  $R_3$  by the following axioms:  $R_1 \sqsubseteq R_{\text{subp}}$ ,  $R_2 \sqsubseteq R_{\text{subp}}$  and  $R_{\text{subp}} \sqsubseteq R_3$ . It can be checked that these three axioms entail the complex RIA  $R_1 \circ R_2 \sqsubseteq R_3$  (denoted by  $\beta$ ). If axiom  $\beta$  does not meet the simple-role restriction (see Section 4.3), the situation of path twisting may also happen as shown in Example 9.

In this work, we study the task of materialization, hence, we can only focus on newly entailed instance data for the RDFS ontologies. To this end, we assume that any class of RDFS ontologies has the same schema data (I). In this way, axioms such as  $\beta$  cannot contribute to the computational complexity, since they only apply to schema data. Further, we assume blank nodes not to occur in any class of RDFS ontologies (II). This allows for expressing RDFS ontology in DHL (see Table 2). Based on the two restrictions (I) and (II), the materialization of RDFS ontologies is tractable in parallel (see Section 4.2).

## 5. Further Optimizing DHL( $\circ$ ) Materialization

In this section, we first discuss why Algorithm  $A_{\text{opt}}$  can hardly work in practice. In order to make Algorithm  $A_{\text{opt}}$  more practical, we propose to modify Algorithm  $A_{\text{opt}}$  and give an algorithm variant Algorithm  $A_{\text{prc}}$ . We show that Algorithm  $A_{\text{prc}}$  can also be restricted to an NC version when materializing DHL( $\circ$ ) ontologies that satisfy both the simple-concept and the simple-role restriction.

### 5.1. Reducing Computing Space

In previous sections, Algorithm  $A_{opt}$  is mainly used for theoretical analysis. However, this algorithm can hardly work in practice due to its inherently high requirement of computing space. Specifically, Algorithm  $A_{opt}$  constructs a materialization graph of the given datalog program  $P$  by checking all possible rule instantiations in  $P^*$ . One can check that  $|P^*|$  could be the quadratic or cubic in the number of constants occurring in  $P$ . Recall the analysis of the parallel complexity of Algorithm  $A_{bsc}$  and Algorithm  $A_{opt}$ . The number of constants is at most  $|I|e$  where  $e$  denotes the maximum arity of any predicate in  $I$ . Consider a datalog rule of the form ' $R(x, y), A(y) \rightarrow B(x)$ '. Since there are two variables in this rule, the number of all possible rule instantiations is  $(|I|e)^2$ . Similarly, for a datalog rule rewritten from an axiom of the form (R3) or (R4), the number of all possible rule instantiations is  $(|I|e)^3$ . Undoubtedly, a plain implementation of Algorithm  $A_{opt}$  would be slow when the target datalog program or ontology tends to be large in size. On the other hand, from Examples 1, 5 and 6, we can observe that the rule instantiations used for the construction of materialization graphs always cover a small part of  $P^*$  with respect to the target datalog program  $P$ . Thus, we consider reducing the computing space of Algorithm  $A_{opt}$  by narrowing down the scope of rule instantiations to be checked. Our strategy is to restrict that, *in each iteration of Algorithm  $A_{opt}$ , each of the checked rule instantiations should involve at least one body atom that has been added to the constructed materialization graph*. Since  $DHL(\circ)$  is our focus, we explain how to apply the above strategy in  $DHL(\circ)$  materialization as follows.

We first consider a datalog rule of the form ' $A(x) \rightarrow B(x)$ ' that corresponds to (T1). The above strategy requires that, in each iteration, Algorithm  $A_{opt}$  can only check the rule instantiations of the form ' $A(a) \rightarrow B(a)$ ' where  $A(a)$  has been added to the constructed materialization graph  $\mathcal{G}$ . If there are  $n$  atoms of the form  $A(x)$  that have been added to  $\mathcal{G}$ , the number of all checked rule instantiations of the above rule is also  $n$ , instead of  $(|I|e)$ . This is because, for each assertion  $A(a)$ , the datalog rule ' $A(x) \rightarrow B(x)$ ' has only one rule instantiation ' $A(a) \rightarrow B(a)$ ' where the variable  $x$  is substituted by the constant  $a$ . The cases of (R1) and (R2) can be analyzed similarly. We now consider datalog rules that have two body atoms, i.e., datalog rules of the form ' $R(x, y), A(y) \rightarrow B(x)$ ' (see (T3)), ' $A_1(x), A_2(x) \rightarrow B(x)$ ' (see (T2)) and ' $R_1(x, y), R_2(y, z) \rightarrow R(x, z)$ ' (see (R3) and (R4)). We require that, for each rule instantiation of these rules checked by Algorithm  $A_{opt}$ , at least one body atom has been added to the constructed materialization graph  $\mathcal{G}$ ; thus, it can be checked that, in each iteration of Algorithm  $A_{opt}$ , the number of checked rule instantiations is at most  $k(|I|e)$  where  $k$  is the number of atoms that have been added to  $\mathcal{G}$ . Note that, for rule instantiations of two body atoms, we do not require that both of these two body atoms have been added to the constructed materialization graph. Otherwise, the algorithm would perform as Algorithm  $A_{bsc}$  and, thus, the optimizations used in Algorithm  $A_{opt}$  cannot further work.

### 5.2. Further Optimizing Algorithm $A_{opt}$

We use the above method of narrowing down the scope of rule instantiations to modify Algorithm  $A_{opt}$  and obtain an algorithm variant Algorithm  $A_{prc}$ . Recall that, in each iteration of Step 2, Algorithm  $A_{opt}$  checks all possible rule instantiations and computes an  $rch$  relation and its transitive closure to determine the existence of SWD paths. We let Algorithm  $A_{prc}$  conduct the same work with narrowing down the scope of rule instantiations to be checked as well. This can be described by the following algorithm.

**Algorithm PRC.** This algorithm has as inputs (1) a DHL( $\circ$ ) ontology  $O = \langle \mathcal{T}, \mathcal{R}, \mathcal{A} \rangle$  and its datalog program  $P = \langle R, \mathbf{I} \rangle$ ; (2) a (partial) materialization graph  $\mathcal{G} = \langle V, E \rangle$  that is constructed from  $P$ . This algorithm outputs an  $\text{rch}$  relation  $S_{rch}$  that is computed as follows:

- add  $\text{rch}(A(a), B(a))$  to  $S_{rch}$  where  $A(a) \rightarrow B(a) \in P^*$ ,  $O \models A \sqsubseteq_* B$  and  $A(a) \in V$ ;
- add  $\text{rch}(A(b), B(a))$  to  $S_{rch}$  where  $R(a, b), A(b) \rightarrow B(a) \in P^*$ ,  $\exists R.A \sqsubseteq B \in \mathcal{T}$  and  $R(a, b) \in V$ ;
- add  $\text{rch}(A_2(a), B(a))$  to  $S_{rch}$  where  $A_1(a), A_2(a) \rightarrow B(a) \in P^*$ ,  $A_1 \sqcap A_2 \sqsubseteq B \in \mathcal{T}$  and  $A_1(a) \in V$ ;
- add  $\text{rch}(A_1(a), B(a))$  to  $S_{rch}$  where  $A_1(a), A_2(a) \rightarrow B(a) \in P^*$ ,  $A_1 \sqcap A_2 \sqsubseteq B \in \mathcal{T}$  and  $A_2(a) \in V$ ;
- add  $\text{rch}(R(a, b), S(a, b))$  to  $S_{rch}$  where  $R(a, b) \rightarrow S(a, b) \in P^*$ ,  $O \models R \sqsubseteq_* S$  and  $R(a, b) \in V$ ;
- add  $\text{rch}(R(a, b), S(b, a))$  to  $S_{rch}$  where  $R(a, b) \rightarrow S(b, a) \in P^*$ ,  $O \models R \sqsubseteq_* S^-$  and  $R(a, b) \in V$ ;
- add  $\text{rch}(R_2(b, c), R_3(a, c))$  to  $S_{rch}$  where  $R_1(a, b), R_2(b, c) \rightarrow R_3(a, c) \in P^*$ ,  $R_1 \circ R_2 \sqsubseteq R_3 \in \mathcal{R}$  (the case where  $R_1 \equiv R_2 \equiv R_3$  is also included) and  $R_1(a, b) \in V$ ;
- add  $\text{rch}(R_1(a, b), R_3(a, c))$  to  $S_{rch}$  where  $R_1(a, b), R_2(b, c) \rightarrow R_3(a, c) \in P^*$ ,  $R_1 \circ R_2 \sqsubseteq R_3 \in \mathcal{R}$  and  $R_2(b, c) \in V$ .  $\square$

Based on Algorithm PRC, we modify Algorithm  $A_{\text{opt}}$  by replacing Step (i) of Algorithm OPT with Algorithm PRC. Algorithm  $A_{\text{prc}}$  is, thus, given as follows:

**Algorithm  $A_{\text{prc}}$ .** Given a DHL( $\circ$ ) ontology  $O = \langle \mathcal{T}, \mathcal{R}, \mathcal{A} \rangle$  and its datalog program  $P = \langle R, \mathbf{I} \rangle$ , this algorithm returns a materialization graph  $\mathcal{G}$  of  $P$ . Initially  $\mathcal{G}$  is empty. The following steps are then performed:

(Step 1) Add all facts in  $\mathbf{I}$  to  $\mathcal{G}$ .

(Step 2) Compute  $S_{rch}$  by performing Algorithm PRC; use an NC algorithm to compute the transitive closure  $S_{rch}^*$  (see (ii) in Algorithm OPT); update  $\mathcal{G}$  by performing (iii) in Algorithm OPT.

(Step 3) If no node has been added to  $\mathcal{G}$  (in Step 2), terminate, otherwise iterate Step 2.  $\square$

**Theorem 5.** For any DHL( $\circ$ ) ontology  $O$ , Algorithm  $A_{\text{prc}}$  halts and outputs a materialization graph of  $O$ .

The above theorem is given to show the correctness of Algorithm  $A_{\text{prc}}$ . Since Algorithm  $A_{\text{prc}}$  is almost identical with Algorithm  $A_{\text{opt}}$ , it has the same upper bound of the computing time, i.e.,  $c_1 + \ell' c_3 + \ell' c_4 \log(|\mathbf{I}|e) + \ell' c_5 \log^2(|\mathbf{I}|e)$ , where  $\ell'$  denotes the number of iterations of Step 2, and the symbols  $c_1, c_3, c_4, c_5$  have the same meanings as those in the analysis of Algorithm  $A_{\text{opt}}$ . The difference between the two algorithms is that Algorithm  $A_{\text{prc}}$  considers a smaller scope of rule instantiations. This gives rise to smaller sizes of the relation  $S_{rch}$ . Based on the above discussion and Algorithm PRC, the size of relation  $S_{rch}$  is bounded by  $h(|\mathbf{I}|e)$  (where  $h$  denotes the number of nodes that have been added to the graph), while the size of the relation  $S_{rch}$  as computed by Algorithm  $A_{\text{opt}}$  is at most  $p^2(|\mathbf{I}|e)^{2w}$ .

We next use an example to show how Algorithm  $A_{\text{prc}}$  handles DHL( $\circ$ ) materialization.

**Example 11.** Consider performing Algorithm  $A_{prc}$  on the ontology  $O_{ex_1}$  of Example 1. Note that the individual set  $IN$  is  $\{a_1, \dots, a_k, b\}$ , i.e.,  $k + 1$  individuals are involved. Initially, Algorithm  $A_{prc}$  adds all the facts  $(A(b), R(a_1, b), S(a_2, a_1), \dots, S(a_k, a_{k-1}))$  to the result  $\mathcal{G}_{ex_1}$  (Step 1). In the first iteration of Step 2, Algorithm  $A_{prc}$  computes  $S_{rch}$  first. According to Algorithm PRC, for each atom of the form  $S(a_i, a_{i-1})$ ,  $2 \leq i \leq k$ , and  $\forall o \in IN$ , an  $rch$  relation of the form  $rch(R(a_{i-1}, o), R(a_i, o))$  is added to  $S_{rch}$ . Since the atom  $R(a_1, b)$  has been added to  $\mathcal{G}_{ex_1}$ , Algorithm  $A_{prc}$  checks that all atoms of the form  $R(a_i, b)$ ,  $2 \leq i \leq k$ , have SWD paths and are added to  $\mathcal{G}_{ex_1}$  in the first iteration. In the second iteration of Step 2, since all atoms of the form  $R(a_i, b)$ ,  $1 \leq i \leq k$ , have been in  $\mathcal{G}_{ex_1}$ , the  $rch$  relations of the form  $rch(A(b), A(a_i))$  are checked with respect to the axiom  $\exists R.A \sqsubseteq A$ ; further all atoms of the form  $A(a_i)$ ,  $1 \leq i \leq k$ , are finally added to  $\mathcal{G}_{ex_1}$  by Algorithm  $A_{prc}$ .

From the above example, one can find that Algorithm  $A_{prc}$  terminates after two iterations of Step 2. This is the same as Algorithm  $A_{opt}$  (see Example 4). We use Theorem 6 to show that Algorithm  $A_{prc}$  also has an NC version when handling ontologies that satisfy the simple-concept and the simple-role restrictions. The correctness of this theorem is based on the fact that the method of narrowing down the scope of rule instantiations in Algorithm  $A_{prc}$  does not influence the determination of the existence of SWD paths. A detailed analysis can be found in the proof in the appendix.

**Theorem 6.** For any  $DHL(\circ)$  ontology  $O$  that satisfies the simple-concept and the simple-role restriction, there exists a poly-logarithmically bounded function  $\psi$ , such that Algorithm  $A_{prc}^\psi$  outputs a materialization graph of  $O$ .

## 6. Evaluation

In the first part of this section, we analyze different kinds of ontologies and datasets and investigate the usability of the results proposed above. In the second part, we evaluate the implementation of Algorithm  $A_{prc}$  and compare it to the reasoning system RDFox.

### 6.1. The Parallel Tractability of the LUBM Datasets and YAGO Ontologies

In this part, we analyze two popular datasets, LUBM and YAGO, and show the relations between these two datasets and the theoretical results regarding parallel tractability.

**LUBM.** In the Semantic Web community, LUBM (The Lehigh University Benchmark) is proposed to facilitate the evaluation of ontology-based systems in a standard and systematic way. In the latest version of LUBM,<sup>6</sup> the core ontology contains 48 classes and 32 properties, used to describe the departments and the staff of universities. By setting a different number of universities for an ontology-generator, users can get datasets of any size based on the core ontology. The statements about properties in the LUBM core ontology, such as inverse property statements, can be rewritten into the datalog rules of the form (R1), (R2) and (R3) in Table 1. Most of the statements about classes can be rewritten into the datalog rules of the form (T1) and (T3) in Table 1. Five axioms have, however, the form  $A \sqsubseteq \exists R.B$ , which requires existentially quantified variables in the rule head when rewriting the axiom into a logic rule:  $A(x) \rightarrow \exists y(R(x, y) \wedge B(y))$  ( $\tau$ ), where a free variable  $y$  is introduced. This kind of axiom is a general case of (T4) in Table 1

<sup>6</sup><http://swat.cse.lehigh.edu/projects/lubm/>

for which  $A$  is actually replaced by the top concept  $\top$ . Similarly to how we handle (T4), we can also eliminate the free variable  $y$  in rule  $\tau$  via Skolemization, i.e., by replacing the variable  $y$  with a new constant  $o_R^{A,B}$ . In this way, rule  $\tau$  can be rewritten into  $A(x) \rightarrow R(x, o_R^{A,B}) \wedge B(o_R^{A,B})$ . If we only focus on the materialization task, the rewriting approach via Skolemization guarantees the completeness and correctness [27]. On the other hand, rule  $\tau$  is not considered when using OWL RL reasoners to handle LUBM [11, 26]. In summary, if the above kind of rule is not considered or rewriting is used, the materialization of a LUBM dataset can be handled by Algorithm  $A_{opt}^\psi$ . (Did you use rewriting or did you drop the axioms?)

**YAGO.** The knowledge base YAGO<sup>7</sup> is constructed from Wikipedia and WordNet. The latest version YAGO3 [28] has more than 10 million entities (e.g., persons, organizations, cities, etc.) and contains more than 120 million facts about these entities.

In order to balance the expressiveness and computing efficiency, a YAGO-style language, called the *YAGO model*, is proposed based on a slight extension of RDFS [29]. The YAGO model defines a set of properties: `domain`, `range`, `subClassOf`, `subRelationOf` and `type`, and a set of classes: `entity`, `class`, `relation` and `acyclicTransitiveRelation`. The facts in the YAGO model are stated by triples, e.g.,  $(r_1, \text{subRelationOf}, r_2)$ , which are similar to the RDFS statements. A group of rules for reasoning over YAGO ontologies is specified as follows [29]:

- (1)  $(r, \text{domain}, c), (x, r, y) \rightarrow (x, \text{type}, c)$
- (2)  $(r, \text{range}, c), (x, r, y) \rightarrow (y, \text{type}, c)$
- (3)  $(c_1, \text{subClassOf}, c_2), (x, \text{type}, c_1) \rightarrow (x, \text{type}, c_2)$
- (4)  $(r_1, \text{subRelationOf}, r_2), (x, r_1, y) \rightarrow (x, r_2, y)$
- (5)  $(r, \text{type}, \text{acyclicTransitiveRelation}), (x, r, y), (y, r, z) \rightarrow (x, r, z)$

According to the semantics given to YAGO [29], the built-in properties in YAGO, i.e., `domain`, `range`, `subClassOf`, `subRelationOf` and `type` act in the same way as the terms in RDFS statements of the form (1–5) in Table 2, respectively. In contrast to RDFS, YAGO also allows for defining acyclic transitive properties using the class `acyclicTransitiveRelation`; further, any fact in some YAGO ontology cannot be described using blank nodes. By carefully checking the rules (1–5) for the reasoning over YAGO ontologies, one can see that these rules can be rewritten into the datalog rules of the form (T3),<sup>8</sup> (T1), (R1) and (R3) in Table 1, respectively. I am not sure why in YAGO it is called *acyclic transitive relation*. The rule (in YAGO) is just for transitivity and their argument the subclassof is acyclic and transitive does not make much sense as I can have equivalent classes:  $C \text{ subclassof } D, D \text{ subclassof } C$ . That is allowed and cyclic. I would tend to just omit the next remark. Further, DHL does not restrict any role to be acyclic. Based on the above analysis, any YAGO ontology can be expressed in DHL and satisfies the simple-concept and the simple-role restrictions. We then have that, for any well-constructed class of YAGO ontologies, Algorithm  $A_{opt}^\psi$   $A_{pre}$ ? both is correct, but as we are evaluating  $A_{pre}$  this might be better? can handle all of the ontologies in the class.

<sup>7</sup><http://www.mpi-inf.mpg.de/home/>

<sup>8</sup>Both of Rule (1) and Rule (2) can be rewritten into (T3).

In addition to LUBM and YAGO, we further investigate different kinds of ontologies and datasets including benchmarks, real-world ontologies and datasets that can be expressed in ontology languages. These ontologies and datasets are collected from the Protégé ontology library,<sup>9</sup> Swoogle<sup>10</sup> and the Oxford ontology library.<sup>11</sup> Based on the analysis of these ontologies, we found that, ignoring imports, many of them belong to  $\mathcal{D}_{dhl}$  or  $\mathcal{D}_{dhl(\circ)}$ . All of these investigated ontologies and the analysis results are available online.<sup>12</sup>

## 6.2. Evaluating the Implementation of Algorithm $A_{pre}$

In this part, we evaluate the implemented prototype system ParallelDHL for DHL( $\circ$ ) materialization based on Algorithm  $A_{pre}$ . We also provide evaluation results for the state-of-the-art reasoning system RDFS [4], which can also handle parallel ontology materialization. Since RDFS is already a mature system, developed by several people over several years, we only provide the results for reference, as a direct comparison with our initial, not yet optimized prototype is not possible.

**Datasets.** We select eight ontologies from the data sources given in the previous subsection; four of the eight ontologies belong to  $\mathcal{D}_{dhl(\circ)}$  and the rest of them does not satisfy the simple-concept or the simple-role restriction. The chosen eight ontologies are real ontologies and are applied in different fields. The basic information for these ontologies is summarized in Table 3.

Table 3: Information about the test ontologies, where  $|\mathcal{T}| + |\mathcal{R}|$  denotes the number of axioms occurring in the ontology,  $|\mathcal{A}|$  denotes the number of assertions in the ABox and the last column states whether the ontology belongs to  $\mathcal{D}_{dhl(\circ)}$

Ontology Name	Field	$ \mathcal{T}  +  \mathcal{R} $	$ \mathcal{A} $	$\mathcal{D}_{dhl(\circ)}$
Finance	finance	1,934	6,152	no
Molecule	chemistry	43	0	no
GrossAnatomy	anatomy	2,276	13	no
Skeleton	medicine	815	0	no
ChemistryPrimitive	chemistry	167	0	yes
FacebookOnto	social networking	185	28	yes
Mahabharata	literature	69	2,036	yes
Transportation	traffic	925	511	yes

In order to compare ABoxes of a similar scale, we generate a limited number of ABox assertions with respect to the TBoxes and RBoxes for the test ontologies using the generation method proposed by Elhaik et al. [30]. For each test ontology, e.g., the Finance ontology, we generate 5 new ontologies with different ABoxes, denoted Finance- $i$  ( $i \in \{1, 2, 3, 4, 5\}$ ). The index  $i$  indicates that the number of ABox assertions of Finance- $i$  is  $i \times 100,000$ . We also use the term “Finance series” to denote the five generated ontologies for the Finance ontology. The other test ontologies are processed similarly.

**The Experimental Results.** We evaluated ParallelDHL and RDFS over the above eight ontology series. The running environment is a DELL server with a memory of 16 GiB and 4 cores. For fairness, we set the same number of threads (i.e., 1, 2, 4, 6 and 8 threads, respectively)

<sup>9</sup>[http://protegewiki.stanford.edu/wiki/Protege\\_Ontology\\_Library](http://protegewiki.stanford.edu/wiki/Protege_Ontology_Library)

<sup>10</sup><http://swoogle.umbc.edu/>

<sup>11</sup><http://www.cs.ox.ac.uk/isg/ontologies/lib/>

<sup>12</sup><https://github.com/quanzz/PT>

Table 4: Analysis of the reasoning results, where the *minimal time ratio* is the minimal reasoning time of ParallelDHL divided by the minimal reasoning time of RDFox and the *average speedup* is  $1/4(T_1/T_2 + T_2/T_4 + T_4/T_6 + T_6/T_8)$ , where  $T_i$  is the reasoning time with  $i$  threads allocated

series	1	2	3	4	5
minimal time ratio					
Finance	0.54	1.36	0.94	0.99	1.70
Molecule	0.94	0.79	0.96	0.98	1.07
GrossAnatomy	0.65	0.48	0.84	0.77	0.83
Skeleton	1.11	<b>3.08</b>	<b>3.03</b>	<b>3.04</b>	<b>2.75</b>
ChemistryPrimitive	1.29	1.54	1.68	1.14	0.93
FacebookOnto	<b>0.23</b>	<b>0.22</b>	<b>0.19</b>	<b>0.24</b>	<b>0.22</b>
Mahabharata	1.19	1.19	0.99	0.84	0.89
Transportation	0.24	0.52	0.51	0.65	0.49
average speedup of ParallelDHL with increasing number of threads					
Finance	3.06	1.88	2.29	2.95	2.73
Molecule	1.04	1.24	1.13	1.07	1.05
GrossAnatomy	1.41	1.69	1.33	1.32	1.22
Skeleton	3.28	1.84	1.65	1.74	1.96
ChemistryPrimitive	<b>1.46</b>	<b>1.43</b>	<b>1.44</b>	<b>1.51</b>	<b>1.65</b>
FacebookOnto	<b>1.81</b>	<b>1.83</b>	<b>1.82</b>	<b>1.61</b>	<b>1.78</b>
Mahabharata	<b>1.41</b>	<b>1.46</b>	<b>1.51</b>	<b>1.47</b>	<b>1.56</b>
Transportation	<b>1.98</b>	<b>1.57</b>	<b>1.61</b>	<b>1.55</b>	<b>1.59</b>
average speedup of RDFox with increasing number of threads					
Finance	1.01	1.04	1.13	1.11	1.19
Molecule	0.90	1.03	0.95	1.01	0.99
GrossAnatomy	0.87	0.99	0.94	0.96	0.98
Skeleton	1.13	1.01	1.06	1.08	1.08
ChemistryPrimitive	1.28	1.16	1.14	1.11	1.15
FacebookOnto	1.22	1.08	1.07	1.09	1.09
Mahabharata	1.24	1.07	1.13	1.10	1.12
Transportation	1.14	1.04	1.03	1.08	1.11

for ParallelDHL and RDFox in each experiment. The resulting reasoning times<sup>13</sup> are presented in the 16 line graphs (denoted by  $lg1, \dots, lg16$ ) of Figure 5, where the abscissa of each line graph records the numbers of ABox assertions, the ordinate records the reasoning times (in milliseconds), and each of the five curves in different colors denotes the trend of reasoning time with the corresponding number of threads allocated (we use line- $k$  to denote the curve corresponding to  $k$  threads).

We further process the collected data and fill the results in Table 4, where each cell corresponds to a test ontology (see the row label) and a newly generated ABox (distinguished by the series column labels). The minimal time ratio describes the performance of ParallelDHL by using RDFox as the baseline. The indicator ‘speedup’ and its derived indicators are the most common tools used for measuring the capacity of parallelism [4, 31, 11]. Here, we use the aver-

<sup>13</sup>The experimental results can be found at <https://github.com/quanzz/PT>.

age speedup [32] to describe the average capacity of parallelism with different threads allocated. In the following, we give a detailed analysis based on the contents in Figure 5 and Table 4.

**Analysis of the Experimental Results.** According to the theoretical results in Section 4, materialization of the ontologies not belonging to  $\mathcal{D}_{dhl(\odot)}$  (Finance, Molecule, GrossAnatomy and Skeleton) may not be tractable in parallel. In other words, parallel techniques may not work for improving the efficiency of materialization. This is shown in the line graphs of Figure 5. We can see that, in the line graphs (*lg1-lg8*) of the ontology series that do not belong to  $\mathcal{D}_{dhl(\odot)}$ , the four lines, line-2, line-4, line-6 and line-8, intersect to some degree. This is more obvious in the line graphs for the Finance and GrossAnatomy series. For example, in line graph *lg1*, line-2 stays higher than line-6, but lower than line-4. The intersection of lines indicates that reasoning time cannot obviously be reduced with more threads allocated. This is also supported by the results of the average speedups. From Table 4, we can see that the average speedups of ParallelDHL and RDFox for the ChemistryPrimitive, FacebookOnto, Mahabharata and Transportation series are more stable compared with the other ones, i.e., with an average of 1.4 for ParallelDHL. This means that the reasoning time can be reduced on average by 1.4 times with two more threads allocated for ParallelDHL, and, 1.1 for RDFox. For the four ontology series that do not belong to  $\mathcal{D}_{dhl(\odot)}$ , the average speedups have a higher volatility. For example, the Finance and Skeleton series lead to high average speedups that reach up to 3, while, for the Molecule and GrossAnatomy series, the average speedups are even below 1. In summary, from the experiments on the test ontology series, parallelism leads to a more effective improvement for materializing the ontology series that belong to  $\mathcal{D}_{dhl(\odot)}$  compared with the ones outside of this fragment.

By analyzing the experimental results of ParallelDHL and RDFox, we have that ParallelDHL is competitive system despite not yet being highly optimized. From Table 4, we can find that most of the minimal time ratios are close to 1. This means that the reasoning times of ParallelDHL are close to that of RDFox in parallel. In particular, for the GrossAnatomy, FacebookOnto and Transportation series, ParallelDHL performs much better than RDFox; therein, for materializing FacebookOnto-3, ParallelDHL only requires one fifth of the minimal reasoning time of RDFox. On the other hand, for most ontology series, the average speedups of ParallelDHL are overall higher than that of RDFox. The main difference between ParallelDHL and RDFox lies in that ParallelDHL applies the optimizations designed for Algorithm  $A_{opt}$  (see Section 3). The higher average speedups of ParallelDHL also verifies the validity of the optimizations used in Algorithm  $A_{opt}$ . One may note that ParallelDHL is on average 3 times slower than RDFox when handling the Skeleton series. The main reason is that the computation of the *rch* relations requires a large amount of time and the situation of path twisting occurs in Skeleton series. This makes the the optimization based on *rch* relations invalid as discussed in Section 4. **What does the bold font in the table mean?**

## 7. Related Work

The parallel reasoner RDFox [4] is used to evaluate our implementation. RDFox is a state-of-the-art system that handles reasoning on datalog rewritable ontology languages. Algorithm  $A_{bsc}$  proposed in Section 3 is similar to the main algorithm of RDFox (see [4], Sections 3 and 4). The difference lies in that, a group of rule instantiations are handled by one processor (namely a thread) in RDFox, while in Algorithm  $A_{bsc}$ , each rule instantiation is assigned to a unique processor. Thus, in theory, the materialization of the datalog program in Example 1 is serial on RDFox. It is also shown from the experiments that parallelism does not lead to a remarkable improvement when RDFox handles the ontology series in  $\mathcal{D}_{dhl(\odot)}^-$ . We use Algorithm  $A_{opt}$  to show



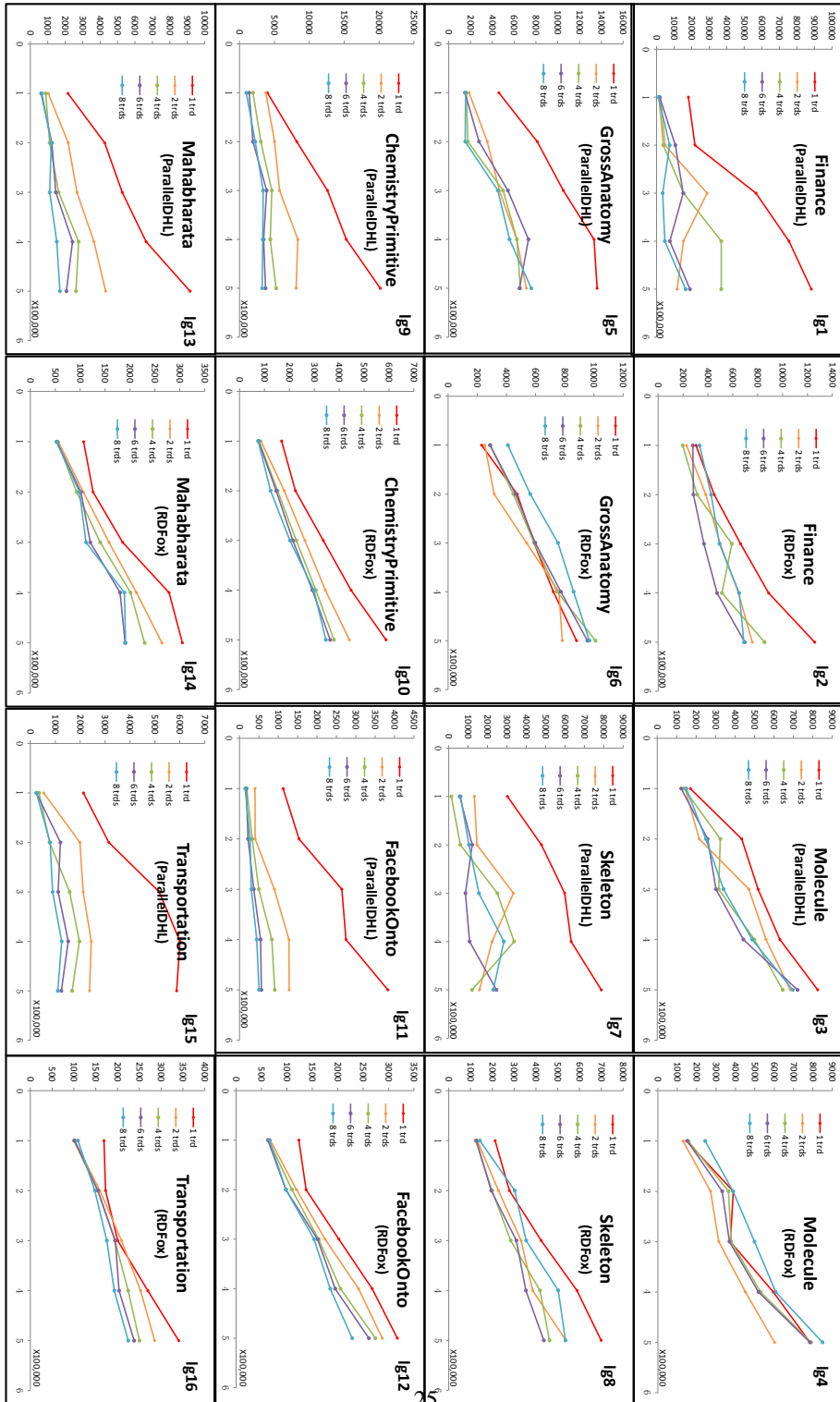


Figure 5: The line graphs of all experiments. The 16 line graphs are labeled by  $lg1, \dots, lg16$  respectively.

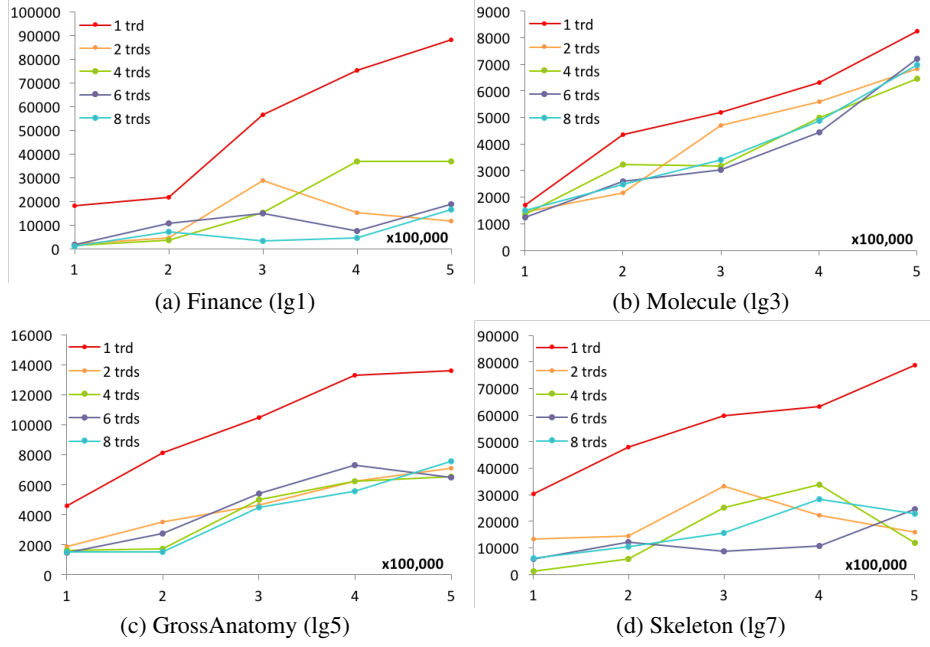


Figure 6: Materialization time in ms for ParallelDHL for the ontologies that do not belonging to  $\mathcal{D}_{dhl(\odot)}$ .

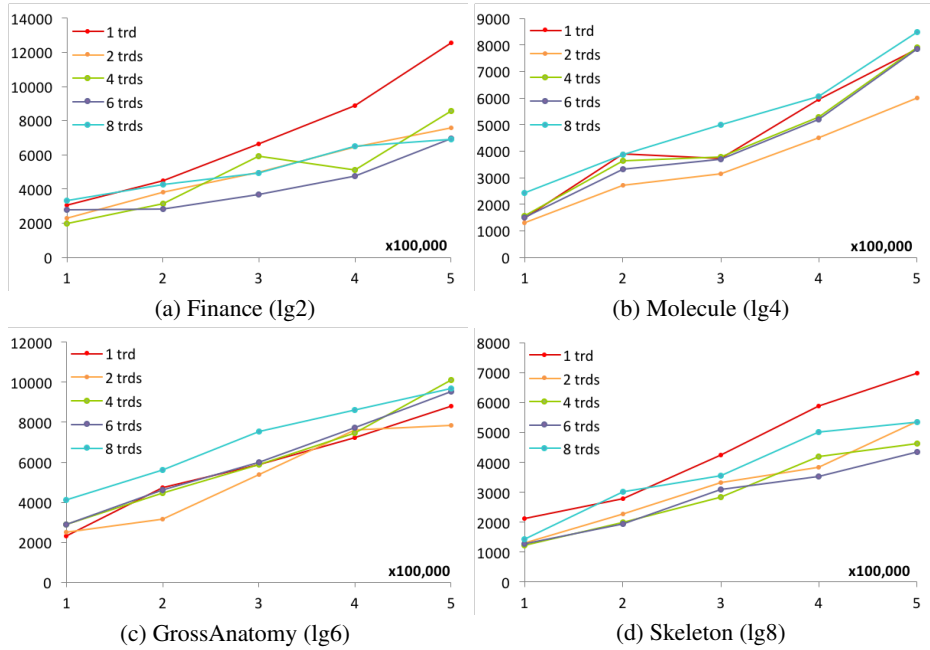


Figure 7: Materialization time in ms for RDFox for the ontologies that do not belonging to  $\mathcal{D}_{dhl(\odot)}$ .

that the ontology in Example 1 is also parallelly tractable. Our implementation ParallelDHL, which is based on Algorithm  $A_{pre}$ , has also been shown to have a better performance than RDFS when handling the test ontologies that belong to  $\mathcal{D}_{dhl(\odot)}$ .

There is work that studies parallel reasoning in RDFS and OWL. The current methods mainly focus on optimizing reasoning algorithms from different aspects. The authors of [10] propose a new kind of encoding method for RDF triples to achieve a high performance. This method can significantly yield a throughput improvement and optimize the parallel RDF reasoning and query answering. In the work of [5], the RETE algorithm is used to accelerate rule matching for RDFS reasoning. The authors of [6] propose a more efficient storage technique and optimize the join operations in parallel reasoning. The issue of balance distribution of parallel tasks is also studied [33, 26]. Two approaches are explored, i.e., *rule partitioning* (allocating parallel tasks to different processors based on rules) and *data partitioning* (allocating parallel tasks based on data). The evaluation results indicate that the efficiency of balance distribution varies with respect to different datasets. On the other hand, parallel reasoning is also implemented for OWL fragments, e.g., OWL RL [14], OWL EL [31], OWL QL [34], and even highly expressive languages [7, 35, 36, 8]. In current work, several techniques are proposed to adapt parallel computation to OWL reasoning tasks. A kind of graph-based method is discussed in [34] to enhance OWL QL classification. The authors of [35, 36, 8] propose pruning techniques to optimize the Tableaux algorithm. The *lock-free technique* is applied in the work [31, 7].

Another line of optimizing parallel reasoning is to utilize high-performance computing platforms. For in-memory platforms, different supercomputers, like Cray XMT, Yahoo S4, have early been used in parallel RDF reasoning [9, 10]. The authors of [12] report their work on RDFS reasoning based on massively parallel GPU hardware. The distributed parallel platforms, like MapReduce and Peer-to-Peer networks, are also used for RDFS reasoning. The representative systems are WebPIE [11], Marvin [37] and SAOR [38]. Different techniques are discussed in this work to tackle the special problems in distributed computing. However, to study the issue of parallel tractability on distributed platforms, we have to discuss more issues, e.g., *network structures* and *communications*. This is not the focus in this paper.

Different from the above work, the purpose of this paper is to study the issue of the parallel tractability of materialization from the perspective of data. The results given in this paper guarantee the parallel tractability theoretically, regardless of what optimization techniques and platforms as discussed above are used.

## 8. Conclusions

In this paper, we studied the problem of parallel tractability of materialization on the datalog rewritable ontologies. To identify the parallelly tractable classes, we proposed two NC algorithms, Algorithm  $A_{bsc}^\psi$  and Algorithm  $A_{opt}^\psi$ , that perform materialization on datalog rewritable ontology languages. Based on these algorithms, we identified the corresponding parallelly tractable datalog program (PTD) classes such that materialization on the datalog programs in these classes is in the complexity class NC. We further studied two specific ontology languages, DL-Lite and DHL (including one of its extension). We showed that any ontology expressed in DL-Lite<sub>core</sub> or DL-Lite<sub>R</sub> is parallelly tractable. For DHL and DHL( $\odot$ ), we proposed two restrictions such that materialization is parallelly tractable.

We verified the usefulness of our theoretical techniques in two ways. On the one hand, we analyzed different kinds of datasets, including well-known benchmarks, real-world ontologies

and a famous dataset YAGO. Our analysis showed that many real ontologies belong to the parallelly tractable class  $\mathcal{D}_{dhl}$  or  $\mathcal{D}_{dhl(\odot)}$ . The developers and users can also refer to  $\mathcal{D}_{dhl}$  and  $\mathcal{D}_{dhl(\odot)}$  to create large-scale ontologies for which parallel tractability is theoretically guaranteed. On the other hand, we used an optimization strategy based on SWD paths to give a practical algorithm variant Algorithm  $A_{prc}$ , which can also be restricted to an NC version, and can be implemented for practice. We implemented a system based on Algorithm  $A_{prc}$  and compared it to the state-of-the-art reasoner RDFox. The experimental results showed that the optimization proposed in this paper results in a better performance on parallelly tractable ontologies compared with RDFox.

In the future work, we will extend our work in two lines. One line is a further study of the parallel tractability of other expressive ontology languages, in addition to datalog rewritable ones. Since different expressive OWL languages have different syntaxes and higher reasoning complexities than PTime-complete, we need to explore more restrictions that are practical and make materialization parallelly tractable. Another line of the future work is to study how to further apply the results in practice. One idea is to apply the technique of SWD paths to enhance other reasoning-based tasks, like ontology classification, ontology debugging and query answering.

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## Appendix

### Appendix A. Proof of Lemma 1

**Lemma 1** *Given a datalog program  $P = \langle R, I \rangle$ , we have (1)  $A_{bsc}$  halts and returns a materialization graph  $\mathcal{G}$  of  $P$ ; (2)  $\mathcal{G}$  has the minimum depth among all the materialization graphs of  $P$ .*

*Proof:* (1) First, whenever a processor  $p$  adds a new node  $v$  to  $\mathcal{G}$ , it has to first check whether  $v$  has already been in  $\mathcal{G}$  and does nothing if  $v$  is in  $\mathcal{G}$ . Thus  $\mathcal{G}$  turns out to be an acyclic graph. Second, to show  $\mathcal{G}$  is a complete materialization graph, we perform an induction on  $T_R^\omega(I)$ . Specifically, all the facts in  $T_R^0(I)$  have to be in  $\mathcal{G}$  by Step 1 of  $A_{bsc}$ . For  $i > 0$ , suppose that the ground atoms in  $T_R^i(I)$  are in  $\mathcal{G}$ . It can be checked that the atoms in  $T_R^{i+1}(I)$  have to be added to  $\mathcal{G}$ , since whenever a new node is derived from  $T_R^i(I)$ , there has to be a processor that would add it to  $\mathcal{G}$ .

(2) The *stage* (see the related contents in Section 2.3) of  $P$  is the lower bound of the depth of all materialization graphs. One can further check that, for the materialization graph  $\mathcal{G}$  constructed by  $A_{bsc}$ , its depth equals to the stage based on the induction above. Thus,  $\mathcal{G}$  has the minimal depth among all the materialization graphs of  $P$ .  $\square$

### Appendix B. Proof of Theorem 1

**Theorem 1** *For any datalog program  $P$ ,  $P \in \mathcal{D}_{A_{bsc}^\psi}$  iff  $P$  has a materialization graph whose depth is upper-bounded by  $\psi(|P|)$ .*

*Proof:* We first prove that the number of iterations of Step 2 is actually the depth of the constructed materialization graph. We define the depths of nodes iteratively as follows: for each explicit node  $v$ ,  $\text{depth}(v)=0$ ; for each implicit node  $v'$  whose parents are  $v_1, \dots, v_i$ ,  $\text{depth}(v')=\max\{\text{depth}(v_1), \dots, \text{depth}(v_i)\} + 1$ . By performing an induction on the number of iterations of Step 2, one can check that an implicit node  $v'$  has to be added to  $\mathcal{G}$  in the  $n^{\text{th}}$  iteration where  $n=\text{depth}(v')$ . Further,  $\text{depth}(\mathcal{G})=\max\{\text{depth}(v_i)|v_i \text{ is in } \mathcal{G}\}$ . We then have that the number of iterations of Step 2 is  $\text{depth}(\mathcal{G})$ .

( $\Rightarrow$ )  $P \in \mathcal{D}_{A_{bsc}^\psi}$  means that  $A_{bsc}^\psi$  can return a materialization graph  $\mathcal{G}$  of  $P$ . Recall that the number of iterations of Step 2 is bounded by  $\psi(|P|)$ . Thus  $\text{depth}(\mathcal{G})$  is also bounded by  $\psi(|P|)$ .

( $\Leftarrow$ ) Suppose  $P$  has a materialization graph  $\mathcal{G}$  whose depth is upper-bounded by  $\psi(|P|)$ . If  $\mathcal{G}$  has the minimal depth among other materialization graphs of  $P$ , the number of iterations of Step 2 is also  $\text{depth}(\mathcal{G})$  (Lemma 1) and, thus, upper-bounded by  $\psi(|P|)$ . If  $A_{bsc}$  does not return  $\mathcal{G}$ , then the returned graph  $\mathcal{G}'$  should have a smaller depth compared with  $\mathcal{G}$ . In this case, this conclusion still holds.  $\square$

### Appendix C. Proof of Lemma 2

**Lemma 2** *Given a datalog program  $P = \langle R, I \rangle$ ,  $A_{opt}$  halts and the output  $\mathcal{G}$  is a materialization graph of  $P$ .*

*Proof:* this lemma can be proved by two stages: (1) the graph  $\mathcal{G}$  returned by  $A_{opt}$  is a materialization graph; (2)  $\mathcal{G}$  is a complete materialization graph. We first show that (1) holds by an induction on the iterations of Step 2 of  $A_{opt}$ .

*Base case.* Initially,  $\mathcal{G}$  only contains all the explicit nodes. In this case,  $\mathcal{G}$  is obviously a materialization graph.

*Inductive case.* According to the induction hypothesis, the partial graph constructed after the  $i^{th}$  ( $i > 1$ ) iteration is a materialization graph, denoted by  $\mathcal{G}^i$ . We have to show that the partial graph constructed after the  $i + 1^{th}$  iteration is also a materialization graph, denoted by  $\mathcal{G}^{i+1}$ . The partial graph  $\mathcal{G}^{i+1}$  is updated in  $\mathbf{A}_{opt}$  by performing the step (iii) of Algorithm Opt. Suppose that  $\text{rch}(B_k, H) \in S_{rch}$  is checked. It corresponds to the rule instantiation  $B_1, \dots, B_k, \dots, B_n \rightarrow H$ . Algorithm Opt next checks that there exists a node  $B'$  such that  $\text{rch}(B', B_k) \in S_{rch}^*$  and  $B'$  is in  $\mathcal{G}$ . This means that  $B_k$  and  $H$  are derivable. The algorithm then adds new nodes  $B_k$  and  $H$  to  $\mathcal{G}^i$  in three cases. (1) if  $H$  is not in  $\mathcal{G}^i$  while  $B_k$  is in  $\mathcal{G}^i$ , then  $H$  is added to  $\mathcal{G}^{i+1}$ , and the edges  $e(B_1, H), \dots, e(B_n, H)$  are created. (2) when neither of  $H$  and  $B_k$  is in  $\mathcal{G}^i$ , then  $H$  and  $B_k$  are added to  $\mathcal{G}^{i+1}$ , and the edges  $e(B_1, H), \dots, e(B_n, H)$  are also created. In the above two cases,  $H$  is a new node. Thus,  $\mathcal{G}^{i+1}$  is acyclic. (3) if  $H$  is in  $\mathcal{G}^i$  and  $H$  has no parent,  $B_k$  is added to  $\mathcal{G}^{i+1}$ , and the edges  $e(B_1, H), \dots, e(B_n, H)$  are also created. For the case that  $H$  is in  $\mathcal{G}^i$  and  $H$  has parents, the algorithm does nothing. Thus,  $\mathcal{G}^{i+1}$  is acyclic and satisfies the definition of materialization graph.

To show that (2) holds, we use the same method in the proof for Lemma 1. We want to show that all the ground atoms in  $T_R^{i+1}(\mathbf{I})$  have to be added to  $\mathcal{G}$ , with the induction hypothesis that the ground atoms in  $T_R^i(\mathbf{I})$  are in  $\mathcal{G}$ . Suppose the ground atoms in  $T_R^i(\mathbf{I})$  have been added to  $\mathcal{G}$  by performing  $\mathbf{A}_{opt}$ . For each atom  $\alpha$  in  $T_R^{i+1}(\mathbf{I})$ ,  $\alpha$  actually has a special SWD path of the length 1. This is because that all parents of  $\alpha$  have been in  $\mathcal{G}$ . According to the optimization strategy,  $\alpha$  has to be added to  $\mathcal{G}$  by applying Step 2 of  $\mathbf{A}_{opt}$ .  $\square$

#### Appendix D. Proof of Lemma 3

**Lemma 3** *Given a DL-Lite ontology  $\mathcal{O}$ , for any materialization graph  $\mathcal{G}$  of  $\mathcal{O}$ , each atom of the form  $A(x)$  ( $A \neq \perp$ ) or  $R(x, y)$  in  $\mathcal{G}$  has an SWD path.*

*Proof:* this lemma can be proved by an induction on applications of the datalog rules corresponding to DL-Lite axioms.

*Base case.* For each explicit node  $v$  of the form  $A(x)$  ( $A \neq \perp$ ) or  $R(x, y)$ ,  $v$  has a special SWD path with  $v$  as the unique node. This conclusion holds for any materialization graph.

*Inductive cases.* For each datalog rule of the form ' $B_1, \dots, B_n \rightarrow H$ ' that is rewritten from some DL-Lite axiom, we have the induction hypothesis that  $B_1, \dots, B_n$  have SWD paths. We are left to prove that  $H$  has an SWD path as well in all materialization graphs of  $\mathcal{O}$ .

If  $H$  is in the form of  $A(x)$  ( $A \neq \perp$ ), it may be derived by applying (T1) and (T3). Thus, we conduct the induction by distinguish these two cases as follows.

Case 1.1  $B \sqsubseteq A$ . According to the induction hypothesis, node  $B(x)$  has an SWD path, denoted by  $(v_1, \dots, v_n, B(x))$ . We have that, in some materialization graph, node  $A(x)$  has an SWD path of the form  $(v_1, \dots, v_n, B(x), A(x))$ .

Case 1.2  $\exists R \sqsubseteq A$ . Node  $R(x, y)$  has an SWD path, denoted by  $(v_1, \dots, v_n, R(x, y))$ , according to the induction hypothesis. It is obvious that node  $A(x)$  has an SWD path of the form  $(v_1, \dots, v_n, B(x), R(x, y))$  in some materialization graph.

Since node  $A(x)$  can only be derived in either Case 1.1 or Case 1.2, we have that node  $A(x)$  has an SWD path in all materialization graphs of  $\mathcal{O}$ .

It is similar to prove the case where  $H$  is in the form of  $R(x, y)$ . Since node  $R(x, y)$  may be derived by applying (R1) and (R2), we discuss these two cases.

Case 2.1  $S \sqsubseteq R$ . According to the induction hypothesis, node  $S(x, y)$  has an SWD path, denoted by  $(v_1, \dots, v_n, S(x, y))$ . Obviously, node  $R(x, y)$  has an SWD path of the form  $(v_1, \dots, v_n, S(x, y), R(x, y))$  in some materialization graph.

Case 2.2  $S \sqsubseteq R^-$ . This case is similar to the above case.

In both of Case 2.1 and Case 2.2, node  $R(x, y)$  has an SWD path. Thus, node  $R(x, y)$  has an SWD path in all materialization graphs of  $\mathcal{O}$ .  $\square$

## Appendix E. Proof of Theorem 2

**Theorem 2** *There exists a poly-logarithmically bounded function  $\psi$  s.t.  $\mathcal{D}_{dl\text{-}lite} \subseteq \mathcal{D}_{\mathcal{A}_{opt}^\psi}$ .*

*Proof:* this theorem can be easily proved based on Lemma 3. Specifically, for any DL-lite ontology  $\mathcal{O}$ , there always exists a poly-logarithmically bounded function  $\psi$  such that  $\mathcal{A}_{opt}^\psi$  can handle the materialization of  $\mathcal{O}$ , since in any materialization graph, each node has an SWD path. More precisely, we can set that  $\psi = 2$ .  $\square$

## Appendix F. Proof of Theorem 3

**Theorem 3** *There exists a poly-logarithmically bounded function  $\psi$  s.t.  $\mathcal{D}_{dhl} \subseteq \mathcal{D}_{\mathcal{A}_{opt}^\psi}$ .*

*Proof:* observe that, for a datalog program that is transformed from a DHL ontology, a rule with a binary atom as its head can only be either of the form (R1), (R2) or (R3). This also indicates that, the materialization of a DHL ontology can be separated into two stages: in the first stage (Stage 1), all the rules of the forms (R1-R3) are exhaustively applied; the consequences of the first stage also serve as the facts in the second stage (Stage 2). The rules of the forms (T1-T3) are then applied in Stage 2. It is obvious that, if both of Stage 1 and Stage 2 can be handled by performing  $\mathcal{A}_{opt}^\psi$  for some poly-logarithmical function  $\psi$ , then the whole materialization can be handled by  $\mathcal{A}_{opt}^\psi$ . In what follows, we investigate the above two stages respectively, and show that, for any datalog program in  $\mathcal{D}_{dhl}$ , such a poly-logarithmical function  $\psi$  exists.

In Stage 1, the rules of the forms (R1-R3) are applied to add new nodes to the constructed materialization graph. We can observe that, rule (R3) is used for computation of transitive roles. As mentioned before, there exists an NC algorithm for transitivity computation. Inspired by this, we can prove that  $\mathcal{A}_{opt}^\psi$  handles Stage 1 for some poly-logarithmical function  $\psi$ . The proof of this result can be shown by separately considering non-transitive roles and the roles that are transitive or influenced by transitive roles. Specifically, we say that role  $R$  is *transitively influenced* (TI) if (1)  $R \circ R \sqsubseteq R \in \mathcal{R}$ ; or (2) there exists a TI role  $R'$  such that  $R' \sqsubseteq_* R$  or  $R' \sqsubseteq_* R^-$ . We say that a role is an NTI (non-transitively influenced) role if it is not a TI role. We further define a set  $\delta_R$  for each role  $R \in \mathbf{R}$  as follows:

**Definition 6.** *For each  $R \in \mathbf{R}$ , let  $\delta_R$  be the set of all assertions as follows:*

1. *for each  $R(a, b) \in \mathcal{A}$ ;*



2.  $R(a, b)$ , for each  $R'(a, b) \in \mathcal{A}$  and  $R' \sqsubseteq_* R$ ;
3.  $R(a, b)$ , for each  $R'(b, a) \in \mathcal{A}$  and  $R' \sqsubseteq_* R^-$ .

Let  $\delta_R^*$  be the transitive closure of  $\delta_R$  where  $R$  is a transitive role. We then have the following lemma.

**Lemma 4.**  $P \models R(a, b)$  implies: (1) if  $R$  is an NTI role,  $R(a, b) \in \delta_R$ ; (2) if  $R$  is a transitive role,  $R(a, b) \in \delta_R^*$ ; (3) if  $R$  is a TI role, then  $R(a, b) \in \delta_R$ , or there exists a transitive role  $R'$  such that  $R' \sqsubseteq_* R$  and  $R'(a, b) \in \delta_{R'}^*$ .

Note that, for all roles  $R$ ,  $\delta_R$  can be computed by only applying (R1) and (R2). In this sense, Lemma 4 also indicates that: (1) for each implicit node  $R(a, b)$  where  $R$  is an NTI role, it can be added to a materialization graph by only applying (R1) and (R2); (2) for each transitive role  $R$ , all implicit nodes of the form  $R(a, b)$  are in the transitive closure  $\delta_R^*$ , which can be computed by an NC algorithm on  $\delta_R$ ; (3) for each role  $R$  that is a TI role but not a transitive role, one can further perform (R1) and (R2) iteratively based on all transitive closures  $\delta_{R'}^*$  where  $R'$  is a transitive role. Since all nodes generated by only applying (R1) and (R2) have SWD paths, the computations of (1) and (3) can be handled by  $\mathcal{A}_{opt}^\psi$ . Further, transitive computation in part (2) can also be handled by  $\mathcal{A}_{opt}^\psi$ . Thus, there exists a poly-logarithmical function  $\psi$  such that  $\mathcal{A}_{opt}^\psi$  handles Stage 1.

The results of Stage 1 serve as the facts of Stage 2. In other words, all binary atoms are explicit nodes in Stage 2. This also means that the rules of the forms (T1) and (T3) generate nodes with SWD paths. Due to the simple-concept restriction, for each rule instantiation  $A_1(a), A_2(a) \rightarrow B(a)$ , one of  $A_1(a)$  and  $A_2(a)$  always has an SWD path. Thus, Stage 2 can be handled by  $\mathcal{A}_{opt}$  in at most two iterations. Based on the above analysis, this theorem holds.

We are now left to prove Lemma 4. We conduct the proof by an induction on the derivation of  $P = \langle R, \mathbf{I} \rangle$ . We distinguish inductive cases by different rules (R1-R3) that are possibly applied to derive  $R(a, b)$ .

*Basic case.* If  $R(a, b) \in \mathbf{I}$ ,  $R(a, b) \in \delta_R$ . In this case, regardless of that  $R$  is a TI or an NTI role, all of (1), (2) and (3) hold.

*Inductive case.* We first study the case where  $R$  is an NTI role.

Case 1.1  $R(a, b)$  is derived by applying the rule (R1), w.l.o.g.,  $R'(a, b) \rightarrow R(a, b)$ , which also means that  $R' \sqsubseteq R \in \mathcal{R}$ . Since  $R$  is an NTI role,  $R'$  has to be an NTI role according to Definition 6. According to the induction hypothesis,  $R'(a, b) \in \delta_{R'}$ . (Case 1.1.1) If  $R'(a, b) \in \mathcal{A}$ ,  $R(a, b) \in \delta_R$  holds according to Definition 6; (Case 1.1.2) If there exists some role  $R''$  such that  $R''(a, b) \in \mathcal{A}$  and  $R'' \sqsubseteq R'$ , we also have  $R(a, b) \in \delta_R$  since  $R'' \sqsubseteq_* R$  holds; (Case 1.1.3) If there exists some role  $R''$  such that  $R''(b, a) \in \mathcal{A}$  and  $R'' \sqsubseteq R'^-$ , this case is similar to (Case 1.1.2).

Case 1.2  $R(a, b)$  is derived by applying the rule (R2), w.l.o.g.,  $R'(a, b) \rightarrow R(b, a)$ , which also means that  $R' \sqsubseteq R^- \in \mathcal{R}$ . This case is similar to Case 1.1.

Case 1.3  $R(a, b)$  is derived by applying the rule (R3). This case is impossible, since  $R$  is an NTI role.

We next study the case where  $R$  is a transitive role.

Case 2.1  $R(a, b)$  is derived by applying the rule (R1), w.l.o.g.,  $R'(a, b) \rightarrow R(a, b)$ , which also means that  $R' \sqsubseteq R \in \mathcal{R}$ .

Case 2.1.1  $R'$  is an NTI role. By the induction hypothesis,  $R'(a, b) \in \delta_{R'}$ . Similar to (Case 1.1.1),  $R(a, b) \in \delta_R$  holds. Obviously,  $R(a, b) \in \delta_R^*$  also holds.

Case 2.1.2  $R'$  is a TI role. By the induction hypothesis, if  $R'(a, b) \in \delta_{R'}$ , then  $R(a, b) \in \delta_R$  also holds. On the other hand, if there exists a transitive role  $R''$  (if  $R'$  is a transitive role, then  $R'' \equiv R'$ ) such that  $R'' \sqsubseteq_* R'$  and  $R''(a, b) \in \delta_{R''}^*$ . Since  $\delta_{R''}^*$  is the transitive closure of  $\delta_{R''}$ , we then have that there must exist such atoms  $(R''(a, c_1), R''(c_1, c_2), \dots, R''(c_n, b))$  in  $\delta_{R''}$ . Further, due to  $R'' \sqsubseteq_* R$  and Definition 6, we have  $R(a, c_1), R(c_1, c_2), \dots, R(c_n, b) \in \delta_R$ . Thus,  $R(a, b) \in \delta_R^*$  also holds.

Case 2.2  $R(a, b)$  is derived by applying the rule (R2), w.l.o.g.,  $R'(b, a) \rightarrow R(a, b)$ , which also means  $R' \sqsubseteq R^- \in \mathcal{R}$ .

Case 2.2.1  $R'$  is an NTI role. By induction hypothesis,  $R'(b, a) \in \delta_{R'}$ . Similar to Case 1.2,  $R(a, b) \in \delta_R$  holds. Obviously,  $R(a, b) \in \delta_R^*$  also holds.

Case 2.2.2  $R'$  is a TI role. By the induction hypothesis, if  $R'(b, a) \in \delta_{R'}$ , then  $R(a, b) \in \delta_R$  also holds. On the other hand, if there exists a transitive role  $R''$  (if  $R'$  is a transitive role, then  $R'' \equiv R'$ ) such that  $R'' \sqsubseteq_* R'$  and  $R''(b, a) \in \delta_{R''}^*$ . Since  $\delta_{R''}^*$  is the transitive closure of  $\delta_{R''}$ , we then have that there must exist such atoms  $(R''(b, c_1), R''(c_1, c_2), \dots, R''(c_n, a))$  in  $\delta_{R''}$ . Further, due to  $R'' \sqsubseteq_* R^-$  and Definition 6, we have  $R(a, c_n), R(c_n, c_{n-1}), \dots, R(c_1, b) \in \delta_R$ . Thus,  $R(a, b) \in \delta_R^*$  also holds.

Case 2.3  $R(a, b)$  is derived by applying the rule (R3), w.l.o.g.,  $R(a, c), R(c, b) \rightarrow R(a, b)$ , which also means that  $R \circ R \sqsubseteq R \in \mathcal{R}$ . By the induction hypothesis,  $R(a, c), R(c, b) \in \delta_R^*$  holds. Obviously,  $R(a, b) \in \delta_R^*$  also holds.

We finally study the case where  $R$  is a TI but not a transitive role.

Case 3.1  $R(a, b)$  is derived by applying the rule (R1), w.l.o.g.,  $R'(a, b) \rightarrow R(a, b)$ , which also means that  $R' \sqsubseteq R \in \mathcal{R}$ .

Case 3.1.1  $R'$  is an NTI role. By the induction hypothesis,  $R'(a, b) \in \delta_{R'}$  holds. Similar to (Case 1.1.1),  $R(a, b) \in \delta_R$  holds. Obviously,  $R(a, b) \in \delta_R^*$  also holds.

Case 3.1.2  $R'$  is a TI role. By the induction hypothesis, if  $R'(a, b) \in \delta_{R'}$ , then  $R(a, b) \in \delta_R$  also holds. On the other hand, if there exists a transitive role  $R''$  (if  $R'$  is a transitive role, then  $R'' \equiv R'$ ) such that  $R'' \sqsubseteq_* R'$  and  $R''(a, b) \in \delta_{R''}^*$ . Since  $R'' \sqsubseteq_* R$  and  $R''$  is the transitive role, third consequence in this Lemma is satisfied.

Case 3.2  $R(a, b)$  is derived by applying the rule (R2), w.l.o.g.,  $R'(a, b) \rightarrow R(b, a)$ , which also means that  $R' \sqsubseteq R^- \in \mathcal{R}$ . This case is similar to Case 3.1.

Case 3.3  $R(a, b)$  is derived by applying the rule (R3). This is impossible, since  $R$  is not a transitive role.  $\square$

## Appendix G. Proof of Theorem 4

**Theorem 4** *There exists a poly-logarithmically bounded function  $\psi$  s.t.  $\mathcal{D}_{dhl(\circ)} \subseteq \mathcal{D}_{\mathcal{A}_{opt}^\psi}$ .*

*Proof:* the proof idea of this theorem is similar to that of Theorem 3. That is, we separate the materialization of  $\text{DHL}(\circ)$  ontologies into two stages: in Stage 1, all the rules of the forms (R1-R4) are exhaustively applied; in Stage 2, the rules of the forms (T1-T2) are then applied while the results of Stage 1 serve as facts. Stage 2 is as same as that of DHL. Thus we only consider Stage 1 here.

Our target is to show that  $\mathcal{A}_{opt}^\psi$  handles Stage 1. To this end, we also distinguish all roles by whether they are transitively influenced. Since we have to consider complex RIAs, we re-define TI and NTI roles as follows: a role  $R$  is *transitively influenced* (TI) if (1)  $R \circ R \sqsubseteq R \in \mathcal{R}$ ; or (2) there exists a TI role  $R'$  such that  $R' \sqsubseteq_* R$  or  $R' \sqsubseteq_* R^-$ ; or (3) there exist a TI role  $R'$  and an axiom of either of the form  $R' \circ R'' \sqsubseteq R$  or  $R'' \circ R' \sqsubseteq R$ . We say that a role is an NTI role if it is not transitively influenced. The set  $\delta_R$  for each role  $R \in \mathbf{R}$  is re-defined based on *role sequence set* that is defined as follows:

**Definition 7.** *Let  $\mathcal{L}(R)$  be the set of role sequences with respect to  $R$  as follows:*

1.  $R'$ , for each  $R' \sqsubseteq_* R$ ;
2.  $R'^-$ , for each  $R' \sqsubseteq_* R^-$ ;
3.  $L_1 L_2$ , for each axiom of the form (except R3)  $R_1 \circ R_2 \sqsubseteq R' \in \mathcal{R}$ ,  $L_i \in \mathcal{L}(R_i) (1 \leq i \leq 2)$  and  $R' \in \mathcal{L}(R)$ ;
4.  $L_2^- L_1^-$ , for each axiom of the form (except R3)  $R_1 \circ R_2 \sqsubseteq R' \in \mathcal{R}$ ,  $L_i \in \mathcal{L}(R_i) (1 \leq i \leq 2)$  and  $R'^- \in \mathcal{L}(R)$ .

In the above definition, for a role sequence  $L = R_1 R_2, \dots, R_n$ , let  $L^- = R_n^-, \dots, R_2^- R_1^-$ . We then give the following definition for  $\delta_R$ .

**Definition 8.** *For each  $R \in \mathbf{RN}$ , let  $\delta_R$  be the set of all assertions as follows:*

1. for each  $R(a, b) \in \mathcal{A}$ ;
2.  $R(a, b)$ , for each  $R'(a, b) \in \mathcal{A}$  and  $R' \sqsubseteq_* R$ ;
3.  $R(a, b)$ , for each  $R'(b, a) \in \mathcal{A}$  and  $R' \sqsubseteq_* R^-$ ;
4.  $R(a, b)$ , for each  $R_0 R_2, \dots, R_n \in \mathcal{L}(R)$ , where  $R_i(x_i, x_{i+1}) \in \mathcal{A}$  and  $x_0 = a, x_{n+1} = b$  for  $0 \leq i \leq n$ .

**Lemma 5.**  $P \models R(a, b)$  implies: (1) if  $R$  is an NTI role,  $R(a, b) \in \delta_R$ ; (2) if  $R$  is a transitive role, then  $R(a, b) \in \delta_R^*$ ; (3) if  $R$  is a TI but not transitive role, then there exists a role sequence  $R_1 R_2, \dots, R_n \in \mathcal{L}(R)$  such that  $R_i(x_i, x_{i+1}) \in \delta_{R_i}$ , or  $R_i(x_i, x_{i+1}) \in \delta_{R_i}^*$  if  $R_i$  is a transitive role, where  $1 \leq i \leq n$  and  $x_0 = a, x_{n+1} = b$ .

The set  $\delta_R^*$  also denotes the transitive closure of  $\delta_R$  where  $R$  is a transitive role. Note that, for all roles  $R$ ,  $\delta_R$  can be computed by applying (R1), (R2) and (R4). The above lemma says: (1) for each implicit node  $R(a, b)$  where  $R$  is an NTI role, it can be added to a materialization graph by only applying (R1), (R2) and (R4); (2) for transitive roles  $R$ , all implicit nodes are in

the transitive closure  $\delta_R^*$ , which can be computed by an NC algorithm on  $\delta_R$ ; (3) for each role  $R$  that is a TI but not a transitive role, one can further perform (R1), (R2) and (R4) iteratively based on all transitive closures  $\delta_{R'}^*$ , where  $R'$  is a transitive role. All nodes generated by applying (R1) and (R2) have SWD paths in the first iteration; at least one role in the left hand side of (R4) are restricted to be a simple role. Thus, the computations of (1) and (3) can be handled by  $\mathcal{A}_{opt}^\psi$ . Further, transitive computation in part (2) can also be handled by  $\mathcal{A}_{opt}^\psi$ . In summary, there exists a poly-logarithmical function  $\psi$  such that  $\mathcal{A}_{opt}^\psi$  handles Stage 1.

We now prove Lemma 5. We conduct the proof by an induction on the derivation of  $P = \langle R, \mathbf{I} \rangle$ . We distinguish inductive cases by different rules (R1-R4) that are possibly applied to derive  $R(a, b)$ .

*Basic case.* If  $R(a, b) \in \mathbf{I}$ ,  $R(a, b) \in \delta_R$ . In this case, regardless of the case that  $R$  is a TI or an NTI role, all of (1), (2) and (3) hold.

*Inductive case.* We first study the case where  $R$  is an NTI role.

Case 1.1  $R(a, b)$  is derived by applying the rule (R1), w.l.o.g.,  $R'(a, b) \rightarrow R(a, b)$ , which also means that  $R' \sqsubseteq R \in \mathcal{R}$ . Since  $R$  is an NTI role,  $R'$  has to be an NTI role according to Definition 8. By the induction hypothesis,  $R'(a, b) \in \delta_{R'}$  holds. (Case 1.1.1) If  $R'(a, b) \in \mathcal{A}$ ,  $R(a, b) \in \delta_R$  holds according to Definition 8; (Case 1.1.2) If there exists some role  $R''$  such that  $R''(a, b) \in \mathcal{A}$  and  $R'' \sqsubseteq R'$ , we also have that  $R(a, b) \in \delta_R$  because  $R'' \sqsubseteq_* R$  holds; (Case 1.1.3) If there exists some role  $R''$  such that  $R''(b, a) \in \mathcal{A}$  and  $R'' \sqsubseteq R'^-$ , this case is similar to (Case 1.1.2); (Case 1.1.4) There exists  $R_0 R_2, \dots, R_n \in \mathcal{L}(R')$ , where  $R_i(x_i, x_{i+1}) \in \mathcal{A}$  and  $x_0 = a, x_{n+1} = b$  for  $0 \leq i \leq n$ . According to Definition 7,  $R_0 R_2, \dots, R_n \in \mathcal{L}(R)$  holds since  $R' \sqsubseteq R \in \mathcal{R}$ .

Case 1.2  $R(a, b)$  is derived by applying the rule (R2), w.l.o.g.,  $R'(a, b) \rightarrow R(b, a)$ , which also means that  $R' \sqsubseteq R^- \in \mathcal{R}$ . Since  $R$  is an NTI role,  $R'$  has to be an NTI role according to Definition 8. By the induction hypothesis,  $R'(b, a) \in \delta_{R'}$ . (Case 1.2.1) If  $R'(b, a) \in \mathcal{A}$ ,  $R(a, b) \in \delta_R$  holds according to Definition 8; (Case 1.2.2) If there exists some role  $R''$  such that  $R''(b, a) \in \mathcal{A}$  and  $R'' \sqsubseteq R'$ , we also have that  $R(a, b) \in \delta_R$  because  $R'' \sqsubseteq_* R^-$  holds; (Case 1.2.3) If there exists some role  $R''$  such that  $R''(a, b) \in \mathcal{A}$  and  $R'' \sqsubseteq R'^-$ , this case is similar to (Case 1.2.2); (Case 1.2.4) there exists  $R_0 R_2, \dots, R_n \in \mathcal{L}(R')$ , where  $R_i(x_i, x_{i+1}) \in \mathcal{A}$  and  $x_0 = b, x_{n+1} = a$  for  $0 \leq i \leq n$ . According to Definition 7,  $R_n^-, \dots, R_1^- \in \mathcal{L}(R)$  holds since  $R' \sqsubseteq R^- \in \mathcal{R}$ .

Case 1.3  $R(a, b)$  is derived by applying the rule (R4), w.l.o.g.,  $R_1(a, c_2), R_2(c_2, b) \rightarrow R(a, b)$ , which also means that  $R_1 \circ R_2 \sqsubseteq R \in \mathcal{R}$ . Since  $R$  is an NTI role, both of  $R_1$  and  $R_2$  are NTI roles. By the induction hypothesis, for each  $R_i(c_i, c_{i+1}) (1 \leq i \leq 2)$  where  $c_1 = a, c_3 = b$ ,  $R_i(c_i, c_{i+1}) \in \delta_{R_i}$  hold. We have a role sequence  $L_i \in \mathcal{L}(R_i)$  that may be constructed in following different cases:

Case 1.3.1 If there exists a role  $R'_i$  such that  $R'_i(c_i, c_{i+1}) \in \mathcal{A}$  and  $R'_i \sqsubseteq_* R_i$ , then we have that  $L_i = R'_i$ .

Case 1.3.2 If there exists a role  $R'_i$  such that  $R'_i(c_{i+1}, c_i) \in \mathcal{A}$  and  $R'_i \sqsubseteq_* R_i^-$ , then we have that  $L_i = R'_i^-$ .

Case 1.3.3 If there exists a role sequence  $R_0 R_2, \dots, R_m \in \mathcal{L}(R')$ , where  $R_j(x_j, x_{j+1}) \in \mathcal{A}$  and  $x_0 = c_i, x_{n+1} = c_{i+1}$  for  $0 \leq i \leq m$  and  $R' \sqsubseteq_* R_i$ , then  $L_i = R_0 R_2, \dots, R_m$  holds.

Based on  $L_i(1 \leq i \leq 2)$ , we have that  $L_1 L_2 \in \mathcal{L}(R)$ . Further  $R(a, b) \in \delta_R$  holds.

Case 1.4  $R(a, b)$  is derived by applying the rule (R3). This case is impossible, since  $R$  is an NTI but not transitive role.

We next study the case where  $R$  is a transitive role.

Case 2.1  $R(a, b)$  is derived by applying the rule (R1), w.l.o.g.,  $R'(a, b) \rightarrow R(a, b)$ , which also means that  $R' \sqsubseteq R \in \mathcal{R}$ .

Case 2.1.1  $R'$  is an NTI role. By the induction hypothesis,  $R'(a, b) \in \delta_{R'}$ . Similar to Case 1.1,  $R(a, b) \in \delta_R$  holds and obviously,  $R \in \mathcal{L}(R)$  holds.

Case 2.1.2  $R'$  is a TI but not a transitive role. By the induction hypothesis, there exists  $R_1, \dots, R_n \in \mathcal{L}(R')$  such that  $R_i(x_i, x_{i+1}) \in \delta_{R_i}$  or  $R_i(x_i, x_{i+1}) \in \delta_{R_i}^*$ , where  $1 \leq i \leq n$  and  $x_0 = a, x_{n+1} = b$ . Since  $R' \sqsubseteq R$  holds,  $R_1, \dots, R_n \in \mathcal{L}(R)$  also holds.

Case 2.1.3  $R'$  is a transitive role. By the induction hypothesis, there exists  $R'(a, b) \in \delta_{R'}^*$ . Further, let  $R'(a, c_1), R'(c_1, c_2), \dots, R'(c_n, b) \in \delta_{R'}$ . Since  $R' \sqsubseteq R \in \mathcal{R}$  holds, we also have that  $R(a, c_1), R(c_1, c_2), \dots, R(c_n, b) \in \delta_R$ . Obviously  $R(a, b) \in \delta_R^*$  holds.

Case 2.2  $R(a, b)$  is derived by applying the rule (R2), w.l.o.g.,  $R'(b, a) \rightarrow R(a, b)$ , which also means  $R' \sqsubseteq R^- \in \mathcal{R}$ .

Case 2.2.1  $R'$  is an NTI role. By the induction hypothesis,  $R'(b, a) \in \delta_{R'}$ . Similar to Case 1.2,  $R(a, b) \in \delta_R$  holds and obviously,  $R \in \mathcal{L}(R)$  holds.

Case 2.2.2  $R'$  is a TI but not transitive role. By the induction hypothesis, there exists  $R_1, \dots, R_n \in \mathcal{L}(R')$  such that  $R_i(x_i, x_{i+1}) \in \delta_{R_i}$  or  $R_i(x_i, x_{i+1}) \in \delta_{R_i}^*$ , where  $1 \leq i \leq n$  and  $x_0 = b, x_{n+1} = a$ . Since  $R' \sqsubseteq R^-$  holds,  $R_n, \dots, R_1 \in \mathcal{L}(R)$  also holds.

Case 2.2.3  $R'$  is a transitive role. By the induction hypothesis, there exists  $R'(b, a) \in \delta_{R'}^*$ . Further, let  $R'(b, c_1), R'(c_1, c_2), \dots, R'(c_n, a) \in \delta_{R'}$ . Since  $R' \sqsubseteq R^- \in \mathcal{R}$ , we also have that  $R(a, c_1), R(c_1, c_2), \dots, R(c_n, b) \in \delta_R$ . Obviously  $R(a, b) \in \delta_R^*$  holds.

Case 2.3  $R(a, b)$  is derived by applying the rule (R4), w.l.o.g.,  $R_1(a, c_2), R_2(c_2, b) \rightarrow R(a, b)$ , which also means that  $R_1 \circ R_2 \sqsubseteq R \in \mathcal{R}$ . Further,  $R$  is a transitive role. According to the simple-role restriction, both of  $R_1$  and  $R_2$  are simple roles. This also means that  $R_1, R_2$  are NTI roles. Further,  $R_1(a, c_2) \in \delta_{R_1}$  and  $R_2(c_2, b) \in \delta_{R_2}$ . This case is similar to Case 1.3.

Case 2.4  $R(a, b)$  is derived by applying the rule (R3), w.l.o.g.,  $R(a, c), R(c, b) \rightarrow R(a, b)$ , which also means that  $R \circ R \sqsubseteq R \in \mathcal{R}$ . Since  $R$  is a TI role, by the induction hypothesis,  $R(a, c), R(c, b) \in \delta_R^*$  holds. Obviously,  $R(a, b) \in \delta_R^*$  holds.

We finally study the case where  $R$  is a TI but not a transitive role.

Case 3.1  $R(a, b)$  is derived by applying the rule (R1), w.l.o.g.,  $R'(a, b) \rightarrow R(a, b)$ , which also means that  $R' \sqsubseteq R \in \mathcal{R}$ .

- Case 3.1.1  $R'$  is an NTI role. By the induction hypothesis,  $R'(a, b) \in \delta_{R'}$  holds. Similar to Case 1.1,  $R(a, b) \in \delta_R$  holds and obviously,  $R \in \mathcal{L}(R)$  holds.
- Case 3.1.2  $R'$  is a TI but not a transitive role. By the induction hypothesis, there exists  $R_1, \dots, R_n \in \mathcal{L}(R')$  such that  $R_i(x_i, x_{i+1}) \in \delta_{R_i}$  or  $R_i(x_i, x_{i+1}) \in \delta_{R_i}^*$ , where  $1 \leq i \leq n$  and  $x_0 = a, x_{n+1} = b$ . Since  $R' \sqsubseteq R$ ,  $R_1, \dots, R_n \in \mathcal{L}(R)$  also holds.
- Case 3.1.3  $R'$  is a transitive role. By the induction hypothesis, there exists  $R'(a, b) \in \delta_{R'}^*$ . According to Definition 7,  $R' \in \mathcal{L}(R)$  holds. Then,  $R'$  is actually the role sequence that satisfies the condition in this lemma.

Case 3.2  $R(a, b)$  is derived by applying the rule (R2), w.l.o.g.,  $R'(b, a) \rightarrow R(a, b)$ , which also means that  $R' \sqsubseteq R^- \in \mathcal{R}$ .

- Case 3.2.1  $R'$  is an NTI role. By the induction hypothesis,  $R'(b, a) \in \delta_{R'}$ . Similar to Case 1.2,  $R(a, b) \in \delta_R$  holds and obviously,  $R \in \mathcal{L}(R)$  holds.
- Case 3.2.2  $R'$  is a TI but not a transitive role. By the induction hypothesis, there exists  $R_1, \dots, R_n \in \mathcal{L}(R')$  such that  $R_i(x_i, x_{i+1}) \in \delta_{R_i}$  or  $R_i(x_i, x_{i+1}) \in \delta_{R_i}^*$ , where  $1 \leq i \leq n$  and  $x_0 = b, x_{n+1} = a$ . Since  $R' \sqsubseteq R^-$ ,  $R_n^-, \dots, R_1^- \in \mathcal{L}(R)$  also holds.
- Case 3.2.3  $R'$  is a transitive role. By the induction hypothesis, there exists  $R'(b, a) \in \delta_{R'}^*$ . Similarly to Case 3.1 (Case 1.1.3), we have that  $R'^-$  is actually the role sequence that satisfies the condition in this lemma.

Case 3.3  $R(a, b)$  is derived by applying the rule (R4), w.l.o.g.,  $R_1(a, c_2), R_1(c_2, b) \rightarrow R(a, b)$ , which also means that  $R_1 \circ R_2 \sqsubseteq R \in \mathcal{R}$  holds. Further,  $R$  is a TI but not transitive role. According to the simple-role restriction, at least one of  $R_1$  and  $R_2$  is a simple role. W.l.o.g., let  $R_2$  be a simple role, which also means that  $R_2$  is an NTI role. Further,  $R_2(c_2, b) \in \delta_{R_2}$  holds. For  $R_1$ , by the induction hypothesis,  $R_1(a, c_2) \in \delta_{R_1}^*$  holds. On the other hand, a role sequence  $L_{R_2}$  exists in  $\mathcal{L}(R_2)$  such that  $c_2$  and  $b$  are the starting and ending individual respectively. Further, we have that  $R_1 L_{R_2} \in \mathcal{L}(R)$  which satisfies the second condition in Lemma 5.

Case 3.4  $R(a, b)$  is derived by applying the rule (R3), w.l.o.g.,  $R(a, c), R(c, b) \rightarrow R(a, b)$ , which also means that  $R \circ R \sqsubseteq R \in \mathcal{R}$ . Since  $R$  is not a transitive role, this case is impossible.  $\square$

## Appendix H. Proof of Theorem 5

**Theorem 5** For any  $DHL(\circ)$  ontology  $\mathcal{O}$ , Algorithm  $A_{prc}$  halts and outputs a materialization graph  $\mathcal{O}$ .

*Proof:* similar to the proof of Lemma 2, this theorem is proved in two stages: (1) the graph  $\mathcal{G}$  returned by  $A_{prc}$  is a materialization graph; (2)  $\mathcal{G}$  is a complete materialization graph. One can show that (1) holds by performing the same induction on the iterations of Step 2 of  $A_{opt}$  (see the proof of Lemma 2). This is because that such an induction is performed regardless of how the relation  $S_{rch}$  is computed.

To prove that (2) holds, we conduct the induction used in the proof for Lemma 1. We aim to show that all the ground atoms in  $T_R^{i+1}(\mathbf{I})$  have to be added to  $\mathcal{G}$ , with the induction hypothesis

that the ground atoms in  $T_R^i(\mathbf{I})$  are in  $\mathcal{G}$ . Suppose the ground atoms in  $T_R^i(\mathbf{I})$  have been added to  $\mathcal{G}$  by performing  $A_{prc}$ . For each atom  $\alpha$  in  $T_R^{i+1}(\mathbf{I})$ , there has to be a relation  $(\beta, \alpha)$  in  $S_{rch}$  obtained in some iteration; further  $\beta$  is already in  $\mathcal{G}$  according to the inductive hypothesis. Thus,  $\alpha$  has to be added to  $\mathcal{G}$  by applying Step 2 of  $A_{prc}$ .  $\square$

## Appendix I. Proof of Theorem 6

**Theorem 6** *For any DHL( $\circ$ ) ontology  $\mathcal{O}$  that follows the simple-concept and the simple-role restrictions, there exists a poly-logarithmically bounded function  $\psi$ , such that  $A_{prc}^\psi$  outputs a materialization graph of  $\mathcal{O}$ .*

*Proof:* similar to what we do when proving Theorem 3 and Theorem 4, we separate the materialization of a DHL( $\circ$ ) ontology into two stages: the role materialization (all the rules of the forms (R1-R4) are exhaustively applied) and the concept materialization (the rules of the forms (T1-T3) are then applied). The difference is that, we further separate the role materialization into successively two stages: *the simple role materialization* (Stage SRM) and *the non-simple role materialization* (Stage NSRM). Similarly, the concept materialization is also separated into successively two stages: *the simple concept materialization* (Stage SCM) and *the non-simple concept materialization* (Stage NSCM). Specifically, in Stage SRM, all atoms of the form  $R(a, b)$  are derived where  $R$  is a simple role, while all atoms of the form  $R(a, b)$  for the non-simple role  $R$  are derived in Stage NSRM. Stage SCM and Stage NSCM can be explained similarly.

We aim to prove that, there always exists a poly-logarithmically bounded function  $\psi$  such that  $A_{prc}^\psi$  can handle each of the four stages, Stage SRM, Stage NSRM, Stage SCM and Stage NSCM. Based on the previous result, we have that there exists a poly-logarithmically bounded function  $\psi'$  such that  $A_{prc}^{\psi'}$  handles the whole materialization.

We first consider Stage SCM. Stage SCM (the simple concept materialization) is conducted on axioms of two forms  $A \sqsubseteq B$  and  $\exists R.A \sqsubseteq B$ , which correspond to datalog rules of the forms  $A(x) \rightarrow B(x)$  and  $R(x, y), A(y) \rightarrow B(x)$  respectively. Since role assertions are supposed to be fixed during in Stage SCM, role  $R$  in each rule of the form  $R(x, y), A(y) \rightarrow B(x)$  can be viewed as an EDB predicate. This further means that, for each atom of the form  $A(a)$  where  $A$  is a simple concept,  $A(a)$  has an SWD path in the first iteration of  $A_{prc}$ . Recall Algorithm Prc. We use an induction to show that such an SWD path of  $A(a)$  can be determined by  $S_{rch}$ . The atom  $A(a)$  can be derived through either of the following rule instantiations:

- Case 1  $B(a) \rightarrow A(a)$ , where  $\mathcal{O} \models B \sqsubseteq_* A$ . We have that  $\text{rch}(B(a), A(a))$  is in  $S_{rch}$  according to Algorithm Prc. By the induction hypothesis, the existence of the SWD path for  $B(a)$  can be determined. Thus, the existence of the SWD path of  $A(a)$  can also be determined by  $S_{rch}$ .
- Case 2  $R(a, b), B(b) \rightarrow A(a)$  where  $\exists R.B \sqsubseteq A \in \mathcal{T}$ . According to Algorithm Prc and the fact that  $R(a, b)$  has to be derived, we have that  $\text{rch}(B(b), A(a))$  is in  $S_{rch}$ . By the induction hypothesis, the existence of the SWD path for  $B(b)$  can be determined. Thus, the existence of the SWD path of  $A(a)$  can also be determined.

We next discuss the non-simple concept materialization (Stage NSCM), which is conducted on axioms of forms (T1-T3). According to the simple-concept restriction, for each axiom of the form  $A_1 \sqcap A_2 \sqsubseteq B$ , one of  $A_1$  and  $A_2$  should be simple concept and can be viewed as an EDB

predicate, since Stage SCM is completed. In Stage NSCM, for each atom of the form  $A(a)$  where  $A$  is not a simple concept,  $A(a)$  has an SWD path in the first iteration of  $\mathbf{A}_{prc}$  due to the simple concept restriction. We also use an induction to show that such an SWD path of  $A(a)$  can be determined by distinguishing different rule instantiations. The case of the rule instantiation of the form  $B(a) \rightarrow A(a)$  (resp.,  $R(a, b), B(b) \rightarrow A(a)$ ) is referred to Case 1 (resp., Case 2). We only consider the case of the rule instantiation of the form  $A_1(a), A_2(a) \rightarrow A(a)$  here.

Case 3  $A_1(a), A_2(a) \rightarrow A(a)$  where  $A_1 \sqcap A_2 \sqsubseteq A \in \mathcal{T}$ . If  $A_1(a)$  is derived where  $A_1$  is a simple concept, we have that  $\text{rch}(A_2(a), A(a))$  is in  $S_{rch}$  according to Algorithm Prc. By the induction hypothesis, the existence of the SWD path for  $A_2(a)$  can be determined. Thus, the existence of the SWD path of  $A(a)$  can also be determined. It is similar for the case where  $A_2(a)$  is derived and  $A_2$  is a simple concept.

The simple-role materialization (Stage SRM) can be simply conducted by checking the transitive closure of role inclusions, i.e., the axioms of the forms (R1-R2). It is easy to check that the simple-role materialization can be finished after the first iteration of  $\mathbf{A}_{prc}$ .

We now discuss Stage NSRM. For each atom of the form  $R(a, b)$  where  $R$  is not a simple role and not a transitive role,  $R(a, b)$  has an SWD path in the first iteration of  $\mathbf{A}_{prc}$  due to the simple role restriction. We use an induction to show that such an SWD path of  $R(a, b)$  can be determined by distinguishing different rule instantiations as follows.

Case 4  $S(a, b) \rightarrow R(a, b)$  where  $\mathcal{O} \models S \sqsubseteq^* R$ . We have that  $\text{rch}(S(a, b), R(a, b))$  is in  $S_{rch}$  according to Algorithm Prc. By the induction hypothesis, the existence of the SWD path for  $S(a, b)$  can be determined. Thus, the existence of the SWD path of  $R(a, b)$  can also be determined.

Case 5  $S(b, a) \rightarrow R(a, b)$  where  $\mathcal{O} \models S \sqsubseteq^* R^-$ . We have that  $\text{rch}(S(b, a), R(a, b))$  is in  $S_{rch}$  according to Algorithm Prc. By the induction hypothesis, the existence of the SWD path for  $S(b, a)$  can be determined. Thus, the existence of the SWD path of  $R(a, b)$  can also be determined.

Case 6  $R_1(a, c), R_2(c, b) \rightarrow R(a, b)$  where  $R_1 \circ R_2 \sqsubseteq R$ . If  $R_1(a, c)$  is derived where  $R_1$  is a simple role, we have that  $\text{rch}(R_2(c, b), R(a, b))$  is in  $S_{rch}$  according to Algorithm Prc. By the induction hypothesis, the existence of the SWD path for  $R_2(c, b)$  can be determined. Thus, the existence of the SWD path of  $R(a, b)$  can also be determined. It is similar for the case where  $R_2(c, b)$  is derived and  $R_2$  is a simple role.

We finally consider the derivation of the atoms of the form  $R(a, b)$ , where  $R$  is a transitive role. Recall Lemma 5. We have that,  $R(a, b) \in \delta_R$  or  $R(a, b) \in \delta_R^*$  holds. According to the simple role restriction, for each sub-role  $S$  of  $R$ ,  $S$  is a simple role, or there exists such axioms of the form  $S_1 \circ S_2 \sqsubseteq S$  where  $S_1$  and  $S_2$  are simple roles. One can check the set  $\delta_R$  can be computed by  $\mathbf{A}_{prc}$  in at most two iterations (see Case 4, Case 5 and Case 6). We should prove that the transitive closure  $\delta_R^*$  can also be computed by  $\mathbf{A}_{prc}^\psi$  for some poly-logarithmically bounded function  $\psi$ .

Case 7  $R(a, c), R(c, b) \rightarrow R(a, b)$  where  $R$  is a transitive role. One can construct a binary tree  $t$  where the root node is  $R(a, b)$ , the leaves are in  $\delta_R$  and the height of  $t$  is upper-bounded by  $\log(|\delta_R|)$ . It can be checked that,  $\mathbf{A}_{prc}$  can generate all nodes in each level of  $t$  in one iteration from the bottom. Thus,  $R(a, b)$  can be derived in at most  $\log(|\delta_R|)$  iterations.  $\square$