Querying Data Sources That Export Infinite Sets of Views

Bogdan Cautis Telecom ParisTech cautis@telecom-paristech.fr Alin Deutsch* UC San Diego deutsch@cs.ucsd.edu Nicola Onose* UC San Diego nicola@cs.ucsd.edu

ABSTRACT

We study the problem of querying data sources that accept only a limited set of queries, such as sources accessible by Web services which can implement very large (potentially infinite) families of queries. We revisit a classical setting in which the application queries are conjunctive queries and the source accepts families of conjunctive queries specified as the expansions of a (potentially recursive) Datalog program.

We say that query Q is *expressible* by the program \mathcal{P} if it is equivalent to some expansion of \mathcal{P} . Q is *supported* by \mathcal{P} if it has an equivalent rewriting using some finite set of \mathcal{P} 's expansions. We present the first study of expressibility and support for sources that satisfy integrity constraints, which is generally the case in practice.

1. INTRODUCTION

The recent proliferation of data sources accessible via Web services has renewed interest in the problem of querying sources with restricted querying capabilities [21, 15, 26, 27]. One reason is that, due to commercial, load-control or privacy considerations, Web sources do not typically accept arbitrary application queries against their schema. Instead, they allow only a (potentially infinite) family of parameterized queries implemented by the Web services. For instance, Amazon provides a service that takes an author name as parameter and returns the corresponding books, but will not allow queries that list all the available books. We refer to the queries accepted by a source as *views*.

In this setting, an application query issued against the source schema can experience two levels of service. It can be fully answerable at the source when the query is equivalent to some view exported by the source (provided the right view can be identified). In many cases, the set of answerable queries is extended by a *source wrapper* [21], which intercepts client queries and answers them by automatically identifying a series of relevant views, issuing the corresponding Web service calls and post-processing their results locally.

In this paper, we revisit the setting of [15, 27], in which the application queries are conjunctive queries and the source accepts families of possibly parameterized conjunctive queries specified as the expansions of a (potentially recursive) Datalog program. The program is said to *generate* these views. As argued in [15, 27] and illustrated below, the choice of Datalog as the view specification formalism enables concise yet expressive descriptions of large (even infinite) sets of views

over a given schema.

We say that query Q is *expressible* by the program \mathcal{P} if it is equivalent to some view generated by \mathcal{P} . Expressible queries can therefore be evaluated at the source, requiring no post-processing at the wrapper. Q is *supported* by \mathcal{P} if it has an equivalent rewriting R using some finite set \mathcal{V} of views generated by \mathcal{P} . Note that finding such R and \mathcal{V} witnessing support enables the following execution plan at the wrapper: call the Web services implementing the queries in \mathcal{V} , materialize their results locally and run query R over the materialized database.

The challenge in deciding expressibility and support lies in the fact that the family of views to pick from can be very large or even infinite. This renders infeasible any systematic enumeration of views. Remarkably, the two problems were previously shown to be decidable [15], however only when ignoring any knowledge of constraints satisfied by the source. In this work, we investigate the effect of source constraints.

The following example shows that source constraints generate new opportunities for detecting support, calling for algorithms which exploit them. (Example 1.1 illustrates a limited-query-capability setting and will be our running example in this paper.)

EXAMPLE 1.1. Consider a travel information source conforming to the following schema:

flight(origin, destination) shuttle(origin, destination) train(origin, destination) bus(origin, destination).

The source admits only views concerning arbitrary-length itineraries by plane, such that Paris is reachable by train or bus from the destination airport. This family of views is described as the set of all expansions of the distinguished IDB predicate ans in program \mathcal{P} below:

 $\begin{array}{l} ans(A,B):= \ f(A,C), ind(C,B) \\ ind(C,B):= \ f(C,B), b(B, ``Paris") \\ ind(C,B):= \ f(C,C'), ind(C',B) \\ ind(C,B):= \ f(C,B), t(B, ``Paris") \end{array}$

Consider a query that asks for 2-leg itineraries ending in an airport from which Paris is reachable by train, bus and shuttle.

$$\begin{array}{ll} Q: & q(A,B):=f(A,C), f(C,B), t(B, ``Paris"), \\ & b(B, ``Paris"), s(B, ``Paris") \end{array}$$

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Clearly, Q is neither expressible nor supported by \mathcal{P} because the views generated by \mathcal{P} do not even mention shuttle information. However, suppose we knew the following constraint to hold on the source (stating that any city pair connected by train and bus is also connected by shuttle):

$$\forall A, S \ t(A, S) \land b(A, S) \longrightarrow s(A, S). \tag{1}$$

Then we would like the wrapper to find the rewriting

(R)
$$r(A,B) := V_1^b(A,B), V_1^t(A,B)$$

where $\{V_i^b\}_{i\geq 1}$ (resp. $\{V_i^t\}_{i\geq 1}$) are families of views generated by \mathcal{P} , returning endpoints of itineraries of *i* flight legs where the destination has a bus link (resp. a train link) to Paris. Indeed, it can be checked that *R* is equivalent to *Q* on all databases satisfying (1). Therefore *Q* is supported by \mathcal{P} when (1) holds.

The problem of deciding support is also of interest for implementing security policies. For security reasons, a source would only allow data access via a set of *authorized views*, which are meant to enforce security policies and check user credentials [18, 24]. This type of access control is provided in particular by the so-called "non-Truman" access control model [24], in which the only allowed queries are those that are equivalent to authorized views or a combination thereof. The difference with respect to the previous scenario is that the system does not actually need to build a rewriting, as it will run the original query, provided that support holds.

Authorized views may be parameterized. For example, a security policy may require that a physician access a patient record only after providing the corresponding record identifier (see Example 1.2). In the so-called non-Truman access control model [24], a user query is considered legal only if it has an equivalent rewriting based on authorized views, or, in our terminology, if it is supported. Illegal queries are rejected by the source.

EXAMPLE 1.2. Consider a source for medical data, which grants access to patient records only under some conditions. The source conforms to the following schema (where the recordNumber attribute refers to patient visit record number):

mrecord(patientId, recordNumber)
visit(symptoms, diagnosis, recordNumber)
nextVisit(vID, vID')

and assume that recordNumber is a key for the visit relation. A physician may have access to a limited amount of information concerning patients whose medical records belong to other colleagues, as described by the policy:

"A physician can access the diagnosis for patients only as follows. (1) He can obtain the diagnosis provided he knows the patient identifier and the visit record number. (2) He can also access the diagnosis of visits for patients with symptoms similar to those of a patient whose id and visit record number he knows, as well as of any other follow-up visits. (3) However, the physician can access neither the patient id, nor the visit number for the visits from (2)."

Parts (1) and (2) of the policy could be implemented by separate services, whose authorized views are represented by expansions of distinguished IDB predicate ans_1 and ans_2 respectively in program \mathcal{P}' below (the ? annotation denotes parameters):

$ans_1(S, D)$:-	mrecord(?N,?R), visit(S,D,?R),
$ans_2(D)$:-	mrecord(?N,?R), visit(S, D',?R),
		$ind_1(S, D, R')$
$ind_1(S, D, R)$:-	$visit(S, D', R), ind_2(D, R)$
$ind_1(S, D, R)$:-	visit(S, D, R)
$ind_2(D, R)$:-	$nextVisit(R, R'), ind_2(D, R')$
$ind_2(D,R)$:-	visit(S, D, R)

The physician wants to find the symptoms for the visit with record number r_1 of a patient identified by pid_1 , together with the diagnosis D_1 for any visit with similar symptoms, and the diagnosis D_2 for a subsequent visit. A conjunctive query that is supported by \mathcal{P}' and provides the information needed is q' below. The primary key constraint is needed to validate the authorization because otherwise there would be no correlation between the information about symptoms used by the views witnessing support.

$$\begin{array}{lll} q'(S,D_{1},D_{2}) &: - & \textit{mrecord}(\textit{pid}_{1},r_{1}),\textit{visit}(S,D_{0},r_{1}), \\ & \textit{visit}(S,D_{1},R_{1}),\textit{visit}(S,D_{1}',R_{1}'), \\ & \textit{nextVisit}(R_{1}',R_{2}),\textit{visit}(S_{2},D_{2},R_{2}) \end{array}$$

Note that the system rejects any query trying to retrieve patient ids or visit record numbers, conforming to part (3) of the policy.

Contributions. In this paper, we carry out the (to the best of our knowledge) first study of the problems of expressibility and support under source constraints. In particular, our contributions include:

Most permissive restrictions for decidability. We identify practically relevant restrictions on the program which ensure decidability under a mix of key and weakly acyclic foreign key constraints and beyond. The restrictions are particularly useful as they enable decidability via a reduction to the constraint-free case, which allows one to modularly "plug in" any existing algorithm to this end (such as those in [15, 26, 27] or the one we propose here for an improved upper bound). We show that these restrictions are as permissive as possible, since their slightest relaxation leads to undecidability in the presence of even a single key constraint. This result is counter-intuitive, since the existence of a rewriting of a conjunctive query using a finite set of non-parameterized conjunctive query views under key constraints (and beyond) is known to be decidable in NP.

A widely-applicable sound test. It is unsatisfactory in practice to refuse to test support and expressibility when the decidability restrictions are violated. A more useful approach consists in devising an algorithm which functions as a decision procedure under these restrictions, yielding only a best-effort "approximation" otherwise. One pragmatic articulation of what "approximation" could mean in this context is the following: the algorithm should be *sound* (i.e. no false positives) yet it may return false negatives (i.e. is not *complete*) for inputs that do not obey the decidability restrictions. We present such an algorithm for both expressibility and support, applicable to arbitrary programs under weakly acyclic sets of embedded dependencies [1], which are sufficiently expressive to capture key and foreign key constraints and beyond. The algorithm runs in deterministic exponential time in the size of the query, the size of the program and the maximum size of a constraint, which is as good as the best algorithm for rewriting queries using a finite list of views.

As a side-effect of our investigation, we *settle two open problems* left from prior work in the constraint-free setting.

Improved, practically tight upper bounds. We improve the previously best known upper bounds for deciding support in the constraint-free case: from non-deterministic exponential time in [27] and doubly-exponential time in [15], to deterministic exponential time in combined query and program size. Notice that in a practical implementation, the non-deterministic exponential time upper bound of [27] would still result in a doubly-exponential algorithm. The improvement is achieved using the sound algorithm mentioned above, which provably acts as an exponential-time decision procedure in the absence of constraints. We show our algorithm to be optimal in the program size (we give a deterministic EXPTIME lower bound for fixed query) and optimal for practical purposes in the query size (we give an NP lower bound for fixed program). The question of the tightness of this NP lower bound remains open. An interesting consequence of our new upper bound is that, in practical implementations, rewriting using an infinite set of views is no more expensive than using finitely many views listed individually (still deterministic exponential time).

The relationship between expressibility and support. We establish that expressibility and support are inter-reducible in PTIME in both the absence and the presence of constraints. This enables us to characterize the complexity of expressibility as well, and to employ the same algorithm for solving both problems. The result comes as a pleasant surprise, since prior work reports distinct upper bounds for these problems, suggesting (in line with intuition) that finding a rewriting of the query using program expansions is harder than finding a single equivalent expansion.

A one-size-fits-all solution. It is remarkable (and practically appealing) that all our upper bound results are based on the same algorithm for support, which serves simultaneously as (i) an essentially optimal decision procedure in the constraint-free case, improving prior upper bounds, (ii) a decision procedure under constraints in all known decidable cases, (iii) a sound procedure in general, and (iv) all of the above for the problem of expressibility, due to our inter-reducibility result.

Parameters. For presentation simplicity, we ignore, at first, the presence of parameters in the views generated by the program, and show how parameters are handled in Appendix C.

Paper outline. After introducing preliminary concepts, results and notation in Section 2, in Section 3 we establish the PTIME inter-reducibility of expressibility and support. Section 4 presents decidable restrictions and Section 5 contains a sound algorithm in the case of general constraints. We also show there the improved upper bounds for the constraint-free setting (Section 5.1). We map the boundaries of decidability in Section 6. We discuss related work

in Section 7 and conclude in Section 8. The proofs are given in Appendix D.

2. PRELIMINARIES

We denote with CQ the language of conjunctive queries. Constraints. We consider constraints ξ of the form

$$\forall \bar{u} \forall \bar{w} \ \phi(\bar{u}, \bar{w}) \longrightarrow \exists \bar{v} \ \psi(\bar{u}, \bar{v})$$

where ϕ (the *premise*) and ψ (the *conclusion*) are conjunctions of relational or equality atoms. Such constraints are known as embedded dependencies and are sufficiently expressive to specify all usual integrity constraints, such as keys, foreign keys, inclusion, join, multivalued dependencies, EGDs, TGDs etc. [1]. We call ϕ the premise and ψ the conclusion. If \bar{v} is empty, then ξ is a full dependency. If ψ consists only of equality atoms, then ξ is an equality-generating dependency (EGD). If ψ consists only of relational atoms, then ξ is a tuple-generating dependency (TGD). If the premise and conclusion of a TGD contain one atom each, we call it an inclusion dependency (IND). An IND in which the variables \bar{u} appear precisely in the key attributes of the relation mentioned in the conclusion is a foreign key constraint. A key constraint on relation R can be expressed by the EGD $\forall \bar{u}, \bar{v}_1, \bar{v}_2 \ R(\bar{u}, \bar{v}_1) \land R(\bar{u}, \bar{v}_2) \longrightarrow$ $\bar{v}_1 = \bar{v}_2$. We write $A \models C$ if the instance A satisfies all the constraints in C.

Containment and Equivalence. Query Q_1 is contained in query Q_2 under the set \mathcal{C} of constraints (denoted $Q_1 \sqsubseteq_{\mathcal{C}} Q_2$) iff $Q_1(D) \subseteq Q_2(D)$ for every database $D \models \mathcal{C}$, where Q(D) denotes the result of Q on D. Q_1 is equivalent to Q_2 under \mathcal{C} (denoted $Q_1 \equiv_{\mathcal{C}} Q_2$) iff $Q_1 \sqsubseteq_{\mathcal{C}} Q_2$ and $Q_2 \sqsubseteq_{\mathcal{C}} Q_1$.

Mappings. A partial mapping from CQ query Q_1 to CQ query Q_2 is a function h from the variables and constants of Q_1 to the variables and constants of Q_2 such that (i) h is the identity mapping on all constants, and (ii) for every relational atom (also called subgoal) $R(\bar{X})$ of Q_1 , if h is defined for all variables in (\bar{X}) , then $R(h(\bar{X}))$ is a subgoal of Q_2 . A homomorphism from a set of subgoals C_1 to a set of subgoals C_2 is a partial mapping from the query $Q_1() := C_1$ to the query $Q_2() := C_2$ which is defined on all variables of Q_1 . A containment mapping from CQ query Q_1 with tuple of head variables \bar{X}_1 to CQ query Q_2 with tuple of head variables X_2 is a homomorphism h from Q_1 to Q_2 such that $h(\bar{X}_1) = \bar{X}_2$. We represent mappings as sets of pairs associating variables with either variables or constants, and use the notation X : Y for the pair (X, Y). The union of two mappings is simply the union of their sets of pairs. A mapping is *consistent* if it does not map the same variable to two distinct values. A set of mappings is *compatible* if their union is consistent. Composition of mappings is the standard function composition, denoted by the operator \circ .

Expansion using views. Given a CQ query R formulated in terms of a set of view names $\mathcal{V}($ where the views are also CQs), the *expansion* of query R w.r.t. the views in \mathcal{V} (denoted $expand_{\mathcal{V}}(R)$) is the query E obtained as follows: every subgoal $V(\bar{X})$ in R is replaced by a copy of the body of V, in which the head variables of V are renamed to \bar{X} and all other variables are replaced by variables occurring in no other view bodies introduced during the expansion. It is easy to see that this variable renaming defines a homomorphism h from V into the expansion E, which we refer to as the expansion homomorphism.

Rewriting using views. We say that a conjunctive query R formulated in terms of view names \mathcal{V} is a rewriting of a query Q using \mathcal{V} under a set \mathcal{C} of dependencies iff $Q \equiv_{\mathcal{C}} expand_{\mathcal{V}}(R)$.

Equivalence under views and constraints. Given queries R_1, R_2 formulated in terms of the view names in \mathcal{V} and a set of dependencies \mathcal{C} , we say that R_1 is equivalent to R_2 under \mathcal{V} and \mathcal{C} , denoted $R_1 \equiv_{\mathcal{C}}^{\mathcal{V}} R_2$, iff

 $expand_{\mathcal{V}}(R_1) \equiv_{\mathcal{C}} expand_{\mathcal{V}}(R_2).$

The chase. We will use the classical chase procedure for rewriting conjunctive queries using a set of embedded dependencies [1]. For arbitrary sets \mathcal{C} of dependencies, the chase is not guaranteed to terminate. The least restrictive condition on \mathcal{C} known to date which is sufficient to ensure termination of the chase with \mathcal{C} regardless of the query Qis called *weak acyclicity* [10] (see also [9]). Weak acyclicity of \mathcal{C} implies termination of the chase of Q with \mathcal{C} in time polynomial in the size of Q and exponential in the size of C. Assuming termination of the chase, we denote with $chase_{\mathcal{C}}(Q)$ the query obtained by chasing conjunctive query Q with C to termination (this query is unique up to equivalence). Besides introducing new variables (for instance due to chasing with TGDs), the chase may equate the original variables of Q to constants or to each other (for instance due to chasing with key constraints) [1]. Denoting this variable renaming with r, it is a well-known fact that r is a homomorphic mapping from Q into $chase_{\mathcal{C}}(Q)$, also called the *chase* homomorphism [1].

Datalog expansions. A finite expansion (in short "expansion") of an IDB predicate p of a Datalog program \mathcal{P} is a CQ query with head $p(\bar{X})$ and body obtained as follows: initialize the body to $body := p(\bar{X})$, then apply the following expansion step a finite number of times until no more IDBs are left in the body: for every IDB goal g_i in the body, pick a rule r_i in \mathcal{P} defining g_i and collect all picked rules in a list \mathcal{V} . Treating \mathcal{V} as views, replace body with $expand_{\mathcal{V}}(body)$, where each g_i is expanded using r_i . The set of expansions of \mathcal{P} is infinite if \mathcal{P} is recursive.

Convention. In the remainder of this paper, unless explicitly stated otherwise, all queries and views are conjunctive queries, all programs are Datalog programs, and all dependencies are embedded dependencies.

3. EXPRESSIBILITY VERSUS SUPPORT

We say that a view V is generated by program \mathcal{P} if V is a CQ expansion of \mathcal{P} .

DEFINITION 3.1. Given a Datalog program \mathcal{P} , a conjunctive query Q and a set of embedded dependencies \mathcal{C} , we say that

- Q is supported by P under C (denoted SUPP^P_P(Q)), iff there is a finite set of views V generated by P and a conjunctive query rewriting of Q using V under C.
- Q is expressible by P under C (denoted EXPR^P_P(Q)), iff Q is equivalent under C to some view V generated by P.

In previous work, the problems of support and expressibility were introduced separately (in [15], respectively [27]). They were shown to be decidable, yet their reported complexity upper bounds were different even in the absence of constraints: doubly-exponential deterministic time for support [15], and EXPTIME for expressibility [27]. These results seemed to follow the intuition that finding a rewriting of the query using some expansions of the program is harder than finding a single equivalent expansion.

We establish a counter-intuitive relationship between the two problems, showing them to be inter-reducible in polynomial time even in the presence of dependencies.

THEOREM 3.1. Let C be a weakly acyclic set of embedded dependencies. Then there is a reduction from the problem of support of a query Q by a program \mathcal{P} under C to an instance of the expressibility problem, which is in PTIME in the size of Q and \mathcal{P} and in EXPTIME in the size of C.

COROLLARY 3.1. If the size of the schema (with dependencies) is bounded by a constant, then there is a PTIME reduction from support to expressibility provided the set of embedded dependencies is weakly acyclic.

COROLLARY 3.2. In the absence of dependencies, there is a PTIME reduction from support to expressibility.

The next result shows the existence of a polynomial-time reduction in the other direction, requiring no restrictions on the embedded dependencies.

THEOREM 3.2. Expressibility reduces in PTIME to support.

In particular, since dependency-free support is known to be decidable [15], Theorem 3.2 implies decidability of dependency-free expressibility, with the same complexity.

4. DECIDABLE CASES

In this section, we give restrictions under which the problems of expressibility and support are decidable under constraints. As will be seen in Section 6, the restrictions are needed because the two problems are in general undecidable, and they are fairly tight, in the sense that even slight relaxations thereof lead to undecidability.

Because it is interesting in its own right, we show a particular route to decidability based on reducing to the dependency-free setting, which is known to be decidable [15]. However, this does not yet provide the improved upper bound, which requires improving prior results for the dependencyfree case. We shall do so in Section 5, obtaining a more general result: a novel algorithm that does not rely on reduction to the dependency-free case, but serves as an optimal decision procedure when dependencies are absent or when they satisfy the restrictions presented in this section, and gracefully degenerates to a sound procedure otherwise.

We introduce properties of the program and of the views it generates that suffice for our reduction to the dependencyfree case. The idea is to pre-process the program to explicitly incorporate into it the knowledge about the dependencies, so that these can then be ignored, thus reducing the problem to dependency-free expressibility and support for the new program. The pre-processing technique relies on the *chase* procedure. This was a natural choice, as the chase tool has been traditionally employed successfully to reduce classical decision problems (such as query equivalence or implication of dependencies [1]) from the presence of dependencies to their absence. We start with expressibility.

Given a Datalog program \mathcal{P} , we denote with $chase_{\mathcal{C}}(\mathcal{P})$ the program obtained by chasing each rule of \mathcal{P} with \mathcal{C} .

DEFINITION 4.1 (C-LOCAL PROGRAM). Let C be a weakly acyclic set of dependencies. We say that a program \mathcal{P} is C-local iff for every view V generated by \mathcal{P} there is a view W generated by chase_C(\mathcal{P}), and for every view W generated by chase_C(\mathcal{P}) there is a view V generated by \mathcal{P} , such that chase_C(V) is equivalent to W even in the absence of dependencies.

The intuition behind \mathcal{C} -locality is as follows. Recall that when checking expressibility under \mathcal{C} , one needs to exhibit some view V generated by \mathcal{P} , such that $Q \equiv_{\mathcal{C}} V$. By the chase theorem [1, 17], if the chase terminates, the equivalence under \mathcal{C} reduces to the following equivalence in the absence of dependencies (i.e. under the empty set of dependencies): $chase_{\mathcal{C}}(Q) \equiv_{\emptyset} chase_{\mathcal{C}}(V)$. \mathcal{C} -locality ensures that the chase of view V can be avoided by simply searching among the views generated by $chase_{\mathcal{C}}(\mathcal{P})$. These must include some W with $W \equiv_{\emptyset} chase_{\mathcal{C}}(V)$, so the existence of V as above is equivalent to the existence of W generated by $chase_{\mathcal{C}}(\mathcal{P})$, with $chase_{\mathcal{C}}(Q) \equiv_{\emptyset} W$. This in turn is by definition dependency-free expressibility of query $chase_{\mathcal{C}}(Q)$ by program $chase_{\mathcal{C}}(\mathcal{P})$. Indeed, we can show the following.

THEOREM 4.1. Let Q be a conjunctive query, C a weakly acyclic set of dependencies, and \mathcal{P} a C-local program. Then $\operatorname{Expr}_{\mathcal{P}}^{\mathcal{C}}(Q)$ holds iff $\operatorname{Expr}_{chase_{\mathcal{C}}(\mathcal{P})}^{\emptyset}(chase_{\mathcal{C}}(Q))$ holds.

The reduction of support to the dependency-free case requires an additional restriction on the views generated by the program. In this case, we need to exhibit a set \mathcal{V} of views generated by \mathcal{P} and a rewriting R of Q in terms of \mathcal{V} . Again by the chase theorem [1, 17], this is equivalent (provided the chase terminates) to exhibiting \mathcal{V} and R such that $chase_{\mathcal{C}}(Q) \equiv_{\emptyset} chase_{\mathcal{C}}(expand_{\mathcal{V}}(R))$. The idea behind the reduction is to require the views to be such that no matter how they are used in R, chasing R's expansion gives the same result as first chasing each view individually and then expanding R with the chased views: $chase_{\mathcal{C}}(expand_{\mathcal{V}}(R)) \equiv_{\emptyset} expand_{\{chase_{\mathcal{C}}(V_1),\ldots,chase_{\mathcal{C}}(V_n)\}}(R).$ Now if \mathcal{P} is \mathcal{C} -local, then the chased views are equivalent to some views $\mathcal{W} = \{W_1, \ldots, W_n\}$ generated by $chase_{\mathcal{C}}(\mathcal{P})$, and we have $chase_{\mathcal{C}}(Q) \equiv_{\emptyset} chase_{\mathcal{C}}(expand_{\mathcal{W}}(R))$, which is the definition of dependency-free support of $chase_{\mathcal{C}}(Q)$ by $chase_{\mathcal{C}}(P)$. We formalize this intuition next.

DEFINITION 4.2 (C-INDEPENDENT VIEW SET). Let C be a weakly acyclic set of dependencies. We say that a set of views $\mathcal{V} = \{V_1, \ldots, V_n\}$ is C-independent iff, for every query R' formulated in terms of \mathcal{V} , there exists query R also formulated in terms of \mathcal{V} , such that

(i)
$$R' \equiv_{\mathcal{C}}^{\mathcal{V}} R$$
,

$$chase_{\mathcal{C}}(expand_{\{V_1,\ldots,V_n\}}(R))$$

is equivalent even in the absence of dependencies to

$$expand_{\{chase_{\mathcal{C}}(V_1),\ldots,chase_{\mathcal{C}}(V_n)\}}(R)$$

Notice that we do not require property (ii) in Definition 4.2 to hold for all queries R' over \mathcal{V} , since there are potentially many equivalent forms of R'. It sufficies if one of them satisfies (ii). In that case, we can show the following.

THEOREM 4.2. Let Q be a conjunctive query, C a weakly acyclic set of dependencies, and \mathcal{P} a C-local program. Then, if the views generated by \mathcal{P} are C-independent, then $\text{Supp}_{\mathcal{P}}^{\mathcal{C}}(Q)$ iff $\text{Supp}_{chase_{\mathcal{C}}(\mathcal{P})}^{\emptyset}(chase_{\mathcal{C}}(Q))$. We next provide various syntactic restrictions on the dependencies in C and on \mathcal{P} to guarantee C-independence and C-locality.

THEOREM 4.3. Let C be a weakly acyclic set of inclusion dependencies. Then any Datalog program \mathcal{P} is C-local and every finite subset of its generated views is C-independent.

Theorems 4.1, 4.2 and 4.3 immediately imply that for weakly acyclic sets of inclusion dependencies, expressibility and support reduce to the dependency-free versions:

COROLLARY 4.1. If C is a weakly acyclic set of inclusion dependencies, then for any program \mathcal{P} and query Q, $\operatorname{Expr}_{\mathcal{P}}^{\mathcal{C}}(Q)$ iff $\operatorname{Expr}_{chase_{\mathcal{C}}(\mathcal{P})}^{\emptyset}(chase_{\mathcal{C}}(Q))$ and $\operatorname{Supp}_{\mathcal{P}}^{\mathcal{O}}(Q)$ iff $\operatorname{Supp}_{chase_{\mathcal{C}}(\mathcal{P})}^{\emptyset}(chase_{\mathcal{C}}(Q))$.

EXAMPLE 4.1. Consider a source for travel data using the following schema:

train(origin, destination, operator)
bus(origin, destination, operator)

where each origin-destination pair is connected by a non-stop leg. It accepts queries for train itineraries with arbitrary many legs in which the same operator is used. It returns the origin, the destination, one intermediary stop and the operator. This family of queries is described by program \mathcal{P} :

$$\begin{array}{rcl} (\mathcal{P}) & ans(A,B,C,O) & :- & ind(A,B,O), ind(B,C,O) \\ & ind(B,C,O) & :- & t(B,B',O), ind(B',C,O) \\ & ind(B,C,O) & :- & t(B,C,O) \end{array}$$

Let Q be an application query searching for a one-way trip with connection in Paris, such that starting from Paris one can either continue the trip by bus, and stay with the first operator, or take another train with any available operator.

(Q) $q(A,B) := t(A,C,O_1), b(C,B,O_1), t(C,B,O_2), C = "Paris"$

Notice that Q is not supported by \mathcal{P} in the absence of constraints (the source does not even allow views mentioning the bus predicate): $\mathrm{Supp}_{\mathcal{P}}^{\emptyset}(Q)$ does not hold.

Assume that the source satisfies C which contains the inclusion dependency (2) below, stating that an operator will also cover by bus any leg important enough to be covered by train.

$$\forall X, Y, O \qquad t(X, Y, O) \longrightarrow b(X, Y, O) \tag{2}$$

Since C is (trivially) a weakly acyclic set of INDs, by Corollary 4.1 SUPP $_{\mathcal{P}}^{\mathcal{C}}(Q)$ holds if and only if so does SUPP $_{chase_{\mathcal{C}}(\mathcal{P})}^{\emptyset}(chase_{\mathcal{C}}(Q)).$

Chase steps apply on the extensional parts of the second and third rules of \mathcal{P} , yielding the new rules (we underline the newly added tuples):

$$ind(B,C,O) := t(B,B',O), \underline{b(B,B',O)}, ind(B',C,O)$$

$$ind(B,C,O) := t(B,C,O), \underline{b(B,C,O)}$$

The new program $chase_{\mathcal{C}}(\mathcal{P})$ generates the views V_{ij} denoting the expansion with *i* legs from the origin to the intermediary point and *j* legs from the intermediary point to the destination. This includes the view V_{11} , which gives the shortest itineraries:

 (V_{11}) v(A, B, C, O) := t(A, B, O), b(A, B, O), t(B, C, O), b(B, C, O)

By chasing also the query, we obtain $Q' = chase_{\mathcal{C}}(Q)$:

$$\begin{array}{lll} (Q') \ q(A,B) &: - & t(A,C,O_1), \underline{b(A,C,O_1)}, b(C,B,O_1), \\ & t(C,B,O_2), \overline{b(C,B,O_2)}, C = ``Paris'' \end{array}$$

Observe that $\operatorname{Supp}_{chase_{\mathcal{C}}(\mathcal{P})}^{\emptyset}(chase_{\mathcal{C}}(Q))$ (and $\operatorname{Supp}_{\mathcal{P}}^{\mathcal{C}}(Q)$) still does not hold because all the views V_{ij} require that only one operator be used. To enforce this requirement on Q', one would need a constraint enforcing that the subgoals $b(C, B, O_1)$ and $b(C, B, O_2)$ from Q' refer to the same operator, making the equality $O_1 = O_2$ hold.

Key safety. We next introduce the notion of a program being "key-safe", which guarantees C-locality and C-independence in the presence of key constraints.

Let R be a relation with an *n*-attribute composite key and let $\overline{P} = (p_1, \ldots, p_k)$ be an ordered sequence of k distinct values in the range 1 to n. We say that a rule of \mathcal{P} outputs the key of R, by positions \overline{P} , into the sequence of head variables $\overline{X} = (X_{i_1}, \ldots, X_{i_k})$ if \overline{X} appears in the rule body either

- in the positions p_1, \ldots, p_k of the key attribute sequence of some *R*-subgoal, with the remaining n - k positions (if any) of the key being bound to constant values, or
- in the positions j_1, \ldots, j_k of some *p*-subgoal, where p is an IDB predicate with at least one rule that in turn outputs the key of R by key positions \overline{P} into the sequence of head variables with indices j_1, \ldots, j_k .

We say that a subgoal g outputs the key of R, by positions $\overline{P} = (p_1, \ldots, p_k)$, into the sequence of variables $\overline{X} = (X_{i_1}, \ldots, X_{i_k})$ if

- g uses EDB predicate R and \bar{X} appears in positions p_1, \ldots, p_k in the key attributes of g, with the remaining n k positions (if any) of the key being bound to constant values, or
- g uses IDB predicate p and there exists some rule defining p which outputs the key of R, by the key positions *P*, into variables *X*.

We say that a rule is *safe* for the key constraint on R if whenever one of its IDB subgoals outputs the key of R by some sequence of k key positions \bar{P} into k variables $\bar{X} = (X_{i_1}, \ldots, X_{i_k})$, no other subgoal does the same (for the same key positions \bar{P}). Notice that several EDB subgoals may output the key of the same R by the same key positions and into the same sequence of variables \bar{X} , as long as no IDB goal does.

EXAMPLE 4.2. Suppose that, in Example 4.1, C contains also a key constraint on the b table, stating that bus operators cover disjoint legs:

$$\forall X, Y, O \qquad b(X, Y, O), b(X, Y, O') \longrightarrow O = O' \qquad (3)$$

Notice that $chase_{\mathcal{C}}(\mathcal{P})$ is the same as in Example 4.1 because no chase step applies with the key constraint.

The rules in $chase_{\mathcal{C}}(\mathcal{P})$ are safe. Indeed, in the second rule, b outputs the key into the sequence B, B', while ind outputs it into B', C. The two subgoals in the first rule also output the key, but into different sequences: A, B and B, C respectively.

Intuitively, safety of the rules in a program \mathcal{P} is designed to guarantee \mathcal{C} -locality. It disallows two IDB goals in a rule from outputting the key of some EDB R into the same variables because this could lead, in the expansion of the rule, to two R goals agreeing on the key attributes and thus triggering a chase step with the key constraint. Since the R goals would come from the expansion of distinct IDB goals in the rule, the effect of this chase would not be reproducible by chasing the program rules in isolation (as in the definition of $chase_{\mathcal{C}}(P)$).

We now give a condition ensuring that every set of views generated by \mathcal{P} is \mathcal{C} -independent. This requires additional restrictions on the rules of the distinguished predicates.

DEFINITION 4.3. A program \mathcal{P} is key-safe for a set of key constraints \mathcal{K} if

- 1. each rule is safe for all key constraints in \mathcal{K} , and
- 2. for all distinguished predicates and of \mathcal{P} , all defining rules r of and, and all relational symbols R in the schema, if r outputs the key attributes \overline{A} (as defined above) of some goal $R(\overline{A}, \overline{B})$, it also outputs all nonkey attributes \overline{B} (by the same definition that applied to the key attributes).

If \mathcal{I} is a set of weakly acyclic INDs, we say that \mathcal{P} is key-safe for $\mathcal{C} = \mathcal{K} \cup \mathcal{I}$ if $chase_{\mathcal{I}}(\mathcal{P})$ is key-safe for \mathcal{K} .

Note that key-safety can be checked in PTIME in the size of $\mathcal P$ and $\mathcal K.$

EXAMPLE 4.3. Continuing Example 4.2, we observe that distinguished predicate ans outputs the pairs of key attributes A, B and B, C, but it also outputs O, the only non-key attribute. Therefore, \mathcal{P} is key-safe.

Intuitively, the key safety condition on the distinguished predicates ensures that, given query R' in terms of some views \mathcal{V} generated by \mathcal{P} , there is query $R \equiv_{\mathcal{C}}^{\mathcal{V}} R'$ such that no chase step with a key constraint will apply to $expand_{\mathcal{V}}(R)$. This is because, if two view atoms in R' happen to output the key of some EDB goal G into the same variables \bar{A} , then by key-safety they each must also output all non-key attributes of G, say in variables \bar{B}_1 , respectively \bar{B}_2 . But then there is a query R, equivalent to R', obtained by adding to R' the equalities $\bar{B}_1 = \bar{B}_2$. This equality is preserved in $expand_{\mathcal{V}}(R)$, so the chase step with the key constraint does not apply on $expand_{\mathcal{V}}(R)$. More formally, we can show the following.

THEOREM 4.4. Let C consist of key constraints and an acyclic set of inclusion dependencies. Any Datalog program \mathcal{P} that is key-safe for C is also C-local and all views generated by it are C-independent.

COROLLARY 4.2. If C consists of key constraints and an acyclic set of INDs and \mathcal{P} is key-safe for C, then for any query Q, $\operatorname{Expr}_{\mathcal{P}}^{\mathcal{C}}(Q)$ iff $\operatorname{Expr}_{chase_{\mathcal{C}}(\mathcal{P})}^{\emptyset}(chase_{\mathcal{C}}(Q))$ and $\operatorname{Supp}_{\mathcal{P}}^{\mathcal{C}}(Q)$ iff $\operatorname{Supp}_{chase_{\mathcal{C}}(\mathcal{P})}^{\emptyset}(chase_{\mathcal{C}}(Q))$.

EXAMPLE 4.4. Continuing Example 4.3, a chase step with (3) applies on Q', introducing the equality atom $O_1 = O_2$. With this, $\text{SUPP}_{chase_{\mathcal{C}}(\mathcal{P})}^{\emptyset}(chase_{\mathcal{C}}(Q))$ holds, as witnessed by the rewriting

 $q(A, B) := V_{11}(A, "Paris", B, O).$

Remarks. The definition of key-safety described above is over-conservative: it considers all constants as being equatable in a chase step. This is because it only keeps track of the positions bound to constants, ignoring the actual constant values. We describe in Appendix B a refined version of key-safety that takes into account these values. This refined notion of key-safety is implied by the one presented here and detects strictly more decidable cases, but, for ease of presentation, it is omitted from the main text.

According to the results presented so far in this section, and in Section 3, under the decidability restrictions (C-independence and C-locality), we can solve expressibility under Ceven by using our favorite solver for dependency-free support (first reduce to dependency-free expressibility, then reduce to dependency-free support). Symmetrically, we can solve support under C using any solver for dependency-free expressibility. It turns out that the same cross-use of solvers can be achieved by first reducing from expressibility under C to support under C (using Theorem 3.2), and then to dependency-free support (using Theorem 4.1) (and symmetrically for support), as the reductions preserve restrictions for decidability. More details can be found in Appendix E.

5. A WIDELY APPLICABLE SOUND TEST

We next present a sound algorithm for testing support, applicable to any program and set of weakly acyclic dependencies. It is a decision procedure (no false negatives) under the decidability restrictions of Section 4, and in the dependencyfree case (where it provides an exponentially better upper bound than previous work).

Our solution is based on the following overall strategy. Since a systematic enumeration of all (potentially infinitely many) views generated by a program \mathcal{P} is infeasible, we instead "describe the behavior" (in a sense formalized shortly) of any view generated by \mathcal{P} w.r.t. a decision procedure (described below) for the existence of a rewriting under C using *finitely* many views. This description will abstract away from the view body, focusing on how the view behaves in essential tests performed by this decision procedure. As it will turn out, under our decidability restrictions, there are only *finitely* many distinct behaviors, each exhibited by a possibly infinite set of views. It suffices therefore to find one representative view from each set, thus reducing the problem of checking support by \mathcal{P} to checking the existence of a rewriting using the finitely many representatives. This problem is known to be decidable under weakly acyclic dependencies (Lemma 5.1 below). We start by describing the associated decision procedure.

Canonical Rewriting Candidate. Given a finite set of views \mathcal{V} , an acyclic set of constraints \mathcal{C} , and a query Q, call the *canonical rewriting candidate* of Q using \mathcal{V} under \mathcal{C} , denoted $CRC_{\mathcal{V}}^{\mathcal{C}}(Q)$, the query obtained as follows: (i) it has the same head variables as Q, and (ii) its body is constructed by evaluating each view $V \in \mathcal{V}$ over the body of $chase_{\mathcal{C}}(Q)$ (viewed as a symbolic database, also known as the canonical instance [1]) and adding the subgoal V(t) for every tuple tin the result of the evaluation. We show next that the canonical rewriting candidate yields a decision procedure for the existence of a rewriting. This result reformulates a theorem in [9] (see also [8])¹:

LEMMA 5.1 (COROLLARY OF [9]). Q has a rewriting using \mathcal{V} under \mathcal{C} iff $CRC_{\mathcal{V}}^{\mathcal{C}}(Q)$ is one. Moreover, this in turn holds iff (a) $CRC_{\mathcal{V}}^{\mathcal{C}}(Q)$ is safe (its head variables appear in its body), and (b) there is a containment mapping from Q into the result of chasing with \mathcal{C} the expansion of $CRC_{\mathcal{V}}^{\mathcal{C}}(Q)$: chase_c(expand_{\mathcal{V}}($CRC_{\mathcal{V}}^{\mathcal{C}}(Q)$)) $\sqsubseteq Q$.

EXAMPLE 5.1. Revisiting Example 1.1, consider the following set of views $\mathcal{V} = \{V_1, V_2\}$:

 $\begin{array}{lll} (V_1) & ans^1(Z_1,Z_2) & :- & f(Z_1,X), f(X,Z_2), t(Z_2, ``Paris") \\ (V_2) & ans^2(Z_1,Z_2) & :- & f(Z_1,Y), f(Y,Z_2), b(Z_2, ``Paris") \end{array}$

generated (among others) by \mathcal{P} . We will follow, step by step, the rewriting algorithm from [9]. The first step consists in finding mappings from the view queries into the body of Qand adding, to Q, atoms corresponding to the head of the view query. V_1 is mapped into Q by $m_1 = \{Z_1 : A; X :$ $<math>C; Z_2 : B\}$, which leads to adding ans¹(A, B). Similarly, for V_2 we discover the mapping $m_2 = \{Z_1 : A; Y : C; Z_2 : B\}$ and add ans²(A, B). We stop here, since no more mappings can be inferred. The result is an expanded query

$$\begin{split} U: \quad q(A,B) :&- f(A,C), f(C,B), t(B, ``Paris"), \\ &\quad b(B, ``Paris"), s(B, ``Paris"), \\ &\quad \underline{ans^1(A,B), ans^2(A,B)} \end{split}$$

in which the newly added atoms are underlined. U is called the universal plan in [9], and it is guaranteed that any exact rewriting of Q is a subquery of U.

 $R = CRC_{\mathcal{V}}^{\mathcal{C}}(Q)$ is then obtained from U by keeping only the atoms from the view schema:

$$R(A,B) := ans^1(A,B), ans^2(A,B).$$

R is equivalent to Q under dependency (1), as can be verified by first constructing the expansion $E = expand_{\mathcal{V}}(CRC_{\mathcal{V}}^{\mathcal{C}}(Q))$ as:

$$\begin{array}{lll} E(A,B) &: - & f(A,X'), f(X',B), t(B, ``Paris"), \\ & & f(A,Y'), f(Y',B), b(B, ``Paris") \end{array}$$

which chases with (1) to query (cE):

$$\begin{array}{lll} cE(A,B) &: - & f(A,X'), f(X',B), t(B, ``Paris"), \\ & f(A,Y'), f(Y',B), b(B, ``Paris"), \\ & s(B, ``Paris") \end{array}$$

into which there is a containment mapping from Q, $cm_q = \{A : A, B : B, C : X'\}$. The reverse containment also holds, as witnessed by the containment mapping from cE into Q, $cm_e = \{A : A, B : B, X' : C, Y' : C\}$, hence R is indeed a rewriting.

Note that both views contribute to the rewriting, since both t and b atoms are needed as images of the t and b atoms from Q. The contribution of V_1 consists in m_{v1} , a partial mapping

¹Lemma 5.1 is a corollary of [9], where it is also proven that there are only finitely many rewritings of Q using \mathcal{V} that are minimal under \mathcal{C} , and that all of them are subqueries of $CRC_{\mathcal{V}}^{\mathcal{C}}(Q)$.

of Q into cE, obtained by restricting the domain of cm_q to the first three atoms of Q:

$$m_{v1} = \{A : A, B : B, C : X'\}.$$

In this case, the image of m_{v1} , E^1 , is the entire expansion of ans^1 :

$$E^{1} = f(A, X'), f(X', B), t(B, "Paris")$$

The contribution of V_2 is enabled by a partial mapping

$$m_{v2} = \{B:B\}$$

from (the b atom of) Q into the expansion of ans^2 , with the image

$$E^2 = b(B, "Paris").$$

 m_{v1} and m_{v2} agree on the common B variable, and, since together they cover the whole of the body of Q, we obtain by combining them the containment mapping cm_q that maps the entire Q into cE.

Redundant views Let us add now to program \mathcal{P} a new rule, corresponding to the definition of the view V_3 given below:

$$V_3$$
) $ans^3(Z_1, Z_3) := f(Z_1, T), f(T, Z_2), b(Z_3, "Paris").$

(

Running the same rewriting algorithm as above on the set $\mathcal{V}' = \{V_1, V_2, V_3\}$, we discover that V_3 maps into Q by $m_3 = \{Z_1 : A, T : C, Z_2 : B, Z_3 : B\}$, which leads to a rewriting candidate $CRC_{\mathcal{V}'}^c(Q)$ of the form

$$R'(A, B) := ans^{1}(A, B), ans^{2}(A, B), ans^{3}(A, B).$$

 V_3 does not modify the way in which the expansion query (which already had t and b atoms) chases, hence the resulting chased expansion of R' is:

$$\begin{array}{lll} cE'(A,B) &: - & f(A,X'), f(X',B), t(B, ``Paris"), \\ & f(A,Y'), f(Y',B), b(B, ``Paris"), \\ & f(A,T'), f(T',T''), b(B, ``Paris"), \\ & s(B, ``Paris") \end{array}$$

We can argue here that V_2 and V_3 are mutually redundant w.r.t. finding a rewriting of Q. The partial mapping $m_{v3} = \{B : B\}$ from Q into the expansion of ans³, with the image b(B, "Paris"), is isomorphic to the partial mapping m_{v2} from Q into the expansion of ans². To this, add the fact that both mappings from the bodies of the two views into Q, v_2 and v_3 , agree on the images of the distinguished variables, mapping them into variables A and B of Q. Without going into further details, this would be enough to allow us to discard one of the two views and to obtain as a rewriting either $ans^1(A, B)$, $ans^2(A, B)$ or $ans^1(A, B)$, $ans^3(A, B)$.

According to Lemma 5.1 and the observations above, in order for a view to contribute to the rewritability of Q

(i) it must generate a subgoal g of the canonical rewriting candidate

e.g. V_1 generates $ans^1(A, B)$, introduced by the mapping m_1 from V_1 into Q;

(ii) g's expansion may participate in the chase with C of the expansion E of the canonical rewriting candidate e.g. the expansion E¹ of ans¹(A, B) contains the atom t(B, "Paris"), which, together with the expansion of V₂, E² = b(B, "Paris"), allows a chase step with dependency (1) to apply; (iii) since Q maps into the chase of E, the expansion of g must include (after the chase) the image of a partial map from Q

e.g. E^1 is the image of m_{v1} .

We shall therefore describe a view V with respect to its behavior for (i), (ii) and (iii), using the notion of *descriptor*.

Normalized program. For uniformity of treatment, we will assume from now on w.l.o.g. that the program \mathcal{P} is normalized as follows. For every k-ary IDB predicate p, every rule for p has the head variables $\overline{Z} = Z_1, \ldots, Z_k$, in that order. Furthermore, for every EDB predicate e, introduce a new IDB e', replace each occurrence of e in \mathcal{P} with e', and add the rule $e'(\overline{Z}) : -e(\overline{Z})$. The normalized program has only two kinds of rules: those whose bodies consist of a single EDB subgoal (called *EDB rules*), or solely of IDB subgoals (called *IDB rules*). For technical reasons, we additionally compute (as in [15]), the *closure* of the program, which consists in adding for every rule r in \mathcal{P} all rules obtained from r by systematically equating in all possible ways the head variables of r with each other and with the constants in Q.

DEFINITION 5.1 (DESCRIPTORS). For a query Q and a program \mathcal{P} , $E^{(p(t),fr)}$ is called a descriptor w.r.t Q and \mathcal{P} iff

- p is an IDB predicate from \mathcal{P} ,
- E is a conjunctive query body over EDBs from \mathcal{P} ,
- \mathcal{P} generates as expansion of p a query of head variables \overline{Z} , $p(\overline{Z}) := body$,
- there is a homomorphism to : body \rightarrow chase_C(Q) s.t. to(\overline{Z}) = t,
- fr is a partial variable mapping from Q into chase_C(body) such that the image of Q under fr is E.

We call E the expansion fragment described by the descriptor, and (p(t), fr) the adornment of E. We call variables $\{Z_1, \ldots, Z_k\}$ (where k is the arity of p) the distinguished variables of the descriptor, while all other variables in the range of fr are hidden.

In the following, when referring to a descriptor we will omit the program \mathcal{P} and the query Q if they are obvious from the context.

EXAMPLE 5.2. In the setting of Example 5.1, $d_1 = E_1^{(p_1(t_1),fr_1)}$ and $d_2 = E_2^{(p_2(t_2),fr_2)}$ below are descriptors for the views V_1 and V_2 , respectively:

$$\begin{aligned} d_1 &: E_1 = [f(Z_1, X), f(X, Z_2), t(Z_2, "Paris")], \\ p_1(t_1) &= ans(A, B), fr_1 = \{A : Z_1, C : X, B : Z_2\} \\ d_2 &: E_2 = [b(Z_2, "Paris")], \\ p_2(t_2) &= ans(A, B), fr_2 = \{B : Z_2\} \end{aligned}$$

Note that, though the two views contribute the same ans(A, B) goal to the canonical rewriting candidate, the two descriptors distinguish among V_1 and V_2 by the images of Q into the view bodies (E_1 includes the image of Q's t and two f goals, E_2 only the b goal).

Before explaining in detail how descriptors are found, we show how they can be used to soundly infer support. Intuitively, a descriptor represents the fragment of a chased view generated by \mathcal{P} that serves as image of the partial mapping from Q. Our goal is to put together such fragments in a consistent way to create (if it exists) the image of Q under a *containment mapping*.

Partial rewriting candidate. More formally, consider a finite set of descriptors w.r.t. to query Q, program \mathcal{P} and dependencies \mathcal{C} : $\mathcal{D} = \{E_i^{(p_i(t_i),f_{r_i})}\}_{1 \leq i \leq n}$, where all p_i are (not necessarily distinct) distinguished IDBs of \mathcal{P} . Introduce for each predicate p_i a fresh predicate p_i^i (using the rank iof the predicate in an arbitrary ordering of the descriptor set) such that $p_i^i \neq p_j^j$ for all $1 \leq i, j \leq n, i \neq j$. Assuming w.l.o.g. that Q's tuple of head variables is \bar{X} , we call the query

$$R(\bar{X}):=p_1^1(t_1),\ldots,p_n^n(t_n)$$

the partial rewriting candidate described by \mathcal{D} . The set $\mathcal{V} := \{ VF_i : p_i^i(\bar{Z}) :- E_i \}_{1 \leq i \leq n}$ is called the view fragments described by \mathcal{D} . The view fragments VF_i are not necessarily safe queries, if not all the head variables serve as image of the partial mapping fr_i .

EXAMPLE 5.3. For the set of descriptors $\mathcal{D} = \{d_1, d_2\}$ from Example 5.2, the fresh view goals are ans¹, ans² respectively. The partial rewriting candidate described by \mathcal{D} is

$$R(A,B) := ans^{1}(A,B), ans^{2}(A,B)$$

(it happens to coincide with the canonical rewriting candidate shown in Example 5.1). The view fragments are

$$(VF_1)$$
 ans¹(Z₁, Z₂) :- $f(Z_1, X), f(X, Z_2), t(Z_2, "Paris")$
 (VF_2) ans²(Z₁, Z₂) :- $b(Z_2, "Paris").$

Notice how VF_1 's, VF_2 's bodies are isomorphic to fragments of the bodies of V_1 , respectively V_2 from Example 5.1. Also, VF_2 is not safe as variable Z_1 does not appear in the body.

The following result allows us to test support, as in Lemma 5.1, but using descriptors instead of explicit views. The key idea is to use the partial rewriting candidate instead of the canonical rewriting candidate.

COROLLARY 5.1 (OF LEMMA 5.1). Let \mathcal{D} be a finite set of descriptors w.r.t. query Q, program \mathcal{P} and dependencies $\mathcal{C}: \mathcal{D} = \{E_i^{(p_i(t_i),fr_i)}\}_{1 \le i \le n}$. Denote with

- R the partial rewriting candidate described by \mathcal{D} ,
- \mathcal{V} the view fragments described by \mathcal{D} ,
- E the expansion $expand_{\mathcal{V}}(R)$.

If (a) R is safe and (b) there exists a containment mapping $cfr from Q into chase_{\mathcal{C}}(E)$, then Q is supported by \mathcal{P} under \mathcal{C} .

We say that any set \mathcal{D} as in Corollary 5.1 *witnesses support*. Notice that conditions (a) and (b) in Corollary 5.1 reformulate the corresponding conditions from Lemma 5.1 in terms of descriptors.

EXAMPLE 5.4. The set of descriptors \mathcal{D} in Example 5.3 witnesses support for the query, program and dependency in our running Example 1.1. Indeed, if we apply the test of Corollary 5.1 to the partial rewriting candidate R and the view fragments VF₁ and VF₂ described by \mathcal{D} (shown in Example 5.3), we obtain

• the expansion

$$EF(A, B) := f(A, X'), f(X', B), t(B, "Paris"), b(B, "Paris")$$

• the result (cEF) of chasing EF with dependency (1),

$$cEF(A, B) :- f(A, X'), f(X', B), t(B, "Paris"), b(B, "Paris"), s(B, "Paris")$$

Notice that EF and cEF are fragments of E, respectively cE from Example 5.1. Let cfr be the mapping $\{A : A, B : B, C : X'\}$. Observe that (a) R is safe; and (b) cfr is a containment mapping from Q into cEF, thus satisfying the conditions of Corollary 5.1.

The number of descriptors is infinite due to the unbounded set of hidden variables, but there are only finitely many isomorphism types of descriptors modulo renaming of the hidden variables, in the following sense:

DEFINITION 5.2 (SIMILARITY). Two descriptors $E_1^{(p_1(t_1),f_{r_1})}$ and $E_2^{(p_2(t_2),f_{r_2})}$ are similar iff $p_1 = p_2$ (and hence the distinguished variables of the descriptors are the same), $t_1 = t_2$, and there is an isomorphism i between the ranges of f_{r_1} and f_{r_2} which is the identity on the distinguished variables, and i witnesses the isomorphism of E_1 and E_2 .

Intuitively, the condition on fr_1 and fr_2 ensures that the partial containment mapping of Corollary 5.1, restricted to the view fragment, is the same for both descriptors. It is easy to see that similarity is an equivalence relation, and that there are only finitely many equivalence classes of descriptors under similarity. Indeed in $E^{(p(t),fr)}$, p is a predicate from \mathcal{P} ; t a tuple of variables and constants from $chase_{\mathcal{C}}(Q)$, thus the number of distinct values it can take is polynomial in the size of $chase_{\mathcal{C}}(Q)$ and exponential in the arity of p; the number of distinct (up to isomorphism) partial mappings fris exponential in the number of variables in Q.

Similarity plays a key role in our support test. Indeed we can show that any representative of a similarity equivalence class is as good as any member of the class for the purpose of witnessing support, in the following sense:

if descriptor d_1 is similar to d_2 , then for any set \mathcal{D}

(†) of descriptors, $\mathcal{D} \cup \{d_1\}$ is a support witness if and only if $\mathcal{D} \cup \{d_2\}$ is one.

Algorithm findDescriptors. We next present a bottomup algorithm for computing representatives of descriptor equivalence classes under similarity. The algorithm findDescriptors consists in initializing a set of descriptors \mathcal{D} to the empty set, then repeatedly carrying out the rule steps described below until \mathcal{D} reaches a fixpoint (under similarity), finally returning \mathcal{D} .

EDB rule step. Consider an EDB rule

$$e'(Z_1,\ldots,Z_k):=e(Z_1,\ldots,Z_k)$$

For every variable mapping to from Z_1, \ldots, Z_k into Q's variables and constants, such that the goal $e(to(Z_1), \ldots, to(Z_k))$ appears in $chase_{\mathcal{C}}(Q)$; and every partial variable mapping fr from the variables of Q to $\{Z_1, \ldots, Z_k\}$ (including the empty-domain one), add to \mathcal{D} the descriptor $E^{(e(to(\bar{Z})),fr)}$, where $E = e(\bar{Z})$. Note that descriptors with empty-domain mappings capture the situation when none of the query goals maps into the described e goal².

IDB rule step. Consider an IDB rule

$$p(\bar{X}):-p_1(\bar{X}_1),\ldots,p_n(\bar{X}_n)$$

If there exists a homomorphism h from the rule body into $chase_{\mathcal{C}}(Q)$, and a set of descriptors

$$E_1^{(p_1(h(\bar{X}_1)),fr_1)},\ldots,E_n^{(p_n(h(\bar{X}_n)),fr_n)}$$

in \mathcal{D} , then:

Construct the views $V_i : p_i(\overline{Z}_i) := E_i$. Denote with E the expansion of the rule body using these views, and with xh_i the corresponding expansion homomorphism $xh_i: E_i \to E$ (i.e. the variable renaming performed on each V_i during expansion). Chase E with C and denote with ch the corresponding chase homomorphism $ch: E \to chase_{\mathcal{C}}(E)$. If the set $\{ch \circ xh_i \circ fr_i\}_{1 \le i \le n}$ of partial mappings from Q into $chase_{\mathcal{C}}(E)$ is compatible, construct the combined mapping $cfr := \bigcup_{i=1}^{n} ch \circ xh_i \circ fr_i$, otherwise exit the rule step. For every partial mapping fr from Q into $chase_{\mathcal{C}}(E)$ which extends cfr (including the trivial extension fr = cfr) by mapping additional variables of Q into fresh variables added during the chase, compute the descriptor $d = F^{(p(h(\bar{X})),fr)}$, where F is the image under fr of all goals in Q such that fr is defined on all their variables. If d is not similar to any descriptor in \mathcal{D} , add it to \mathcal{D} .

EXAMPLE 5.5. We next illustrate the rule steps of algorithm findDescriptors for Example 1.1 showing how descriptors d_1 and d_2 from Example 5.2 are derived. First, observe that no chase step applies on Q, so $Q = chase_{\mathcal{C}}(Q)$.

For brevity, we work on the unnormalized program \mathcal{P} . Applications of EDB rule steps produce (among others) the fol*lowing descriptors:*

$$\begin{split} &d_{3} = [f(Z_{1}, Z_{2})]^{(f(A,C), \{A:Z_{1}, C:Z_{2}\})} \\ &d_{4} = [f(Z_{1}, Z_{2})]^{(f(A,C), \{\})} \\ &d_{5} = [f(Z_{1}, Z_{2})]^{(f(C,B), \{C:Z_{1}, B:Z_{2}\})} \\ &d_{6} = [f(Z_{1}, Z_{2})]^{(f(C,B), \{\})} \\ &d_{7} = [t(Z_{1}, "Paris")]^{(t(B, "Paris"), \{B:Z_{1}\})} \\ &d_{8} = [b(Z_{1}, "Paris")]^{(b(B, "Paris"), \{B:Z_{1}\})}. \end{split}$$

Notice that for the same match of EDB goal $f(Z_1, Z_2)$ into goal f(A, B) of chase c(Q), several partial mappings from the query are considered. We show only two here (in descriptors d_3 and d_4 , where the latter uses the empty mapping, meaning that no query variable is mapped into its fragment).

An IDB rule step for the fourth \mathcal{P} rule combines the descriptors d_5 and d_7 yielding a new descriptor:

 $(d_9) \quad [f(Z_1,Z_2),t(Z_2,``Paris")]^{(ind(C,B),\{C:Z_1,B:Z_2\})}$

which combines with d_3 using the first rule of \mathcal{P} , yielding d_1 . Descriptors d_6 and d_8 combine via an IDB rule step with the third rule in \mathcal{P} to

 $(d_{10}) [b(Z_2, "Paris")]^{(ind(C,B), \{B:Z_2\})}$

which combines with d_4 using the first rule of \mathcal{P} , yielding d_2 .

We next prove that the inflationary process for descriptor discovery implemented by algorithm findDescriptors always terminates for weakly acyclic sets of constraints.

LEMMA 5.2. If C is weakly acyclic, then algorithm find-**Descriptors** is quaranteed to

- (a) terminate in time exponential in the sizes of $\mathcal{P}, \mathcal{C},$ and Q.
- (b) output only descriptors, which are all pairwise dissimilar.

Algorithm testSupport. Our algorithm for testing support amounts to deciding if the descriptors computed by algorithm **findDescriptors** give a support witness (in the sense of Corollary 5.1). According to Corollary 5.1, the existence of such a witness is sufficient for support, but, due to our undecidability results, when the program is unrestricted (see Section 6), it is not always a necessary condition. That is why algorithm *testSupport* is in general only sound.

algorithm testSupport

input: query Q, program \mathcal{P} , set of dependencies \mathcal{C} ; begin $\mathcal{N} :=$ the normalization of \mathcal{P} ; $\mathcal{D} := \mathbf{findDescriptors}(Q, \mathcal{N}, \mathcal{C});$ $\mathcal{D}' :=$ all descriptors from \mathcal{D} pertaining to distinguished predicates of \mathcal{N} ; if \mathcal{D}' witnesses support (tested as in in Corollary 5.1) then return true: else return false;

end

Algorithm **testSupport** satisfies the following properties.

THEOREM 5.1. If C is weakly acyclic, the following hold: (1) algorithm testSupport is sound for testing support, and (2) it runs in time exponential in the size of \mathcal{P} , \mathcal{C} , and Q.

Algorithm testSupport produces strictly less false negatives than the approach of reducing away dependencies described in Section 4. First, it is a decision procedure whenever the reduction succeeds:

THEOREM 5.2. If C is weakly acyclic and P is a C-local program generating C-independent views, then algorithm test-**Support** is a decision procedure for support.

COROLLARY 5.2. If C is a weakly acyclic set of key and foreign key constraints, and $chase_{\mathcal{C}}(P)$ is safe for the keys in C, then **testSupport** is a decision procedure for support.

Second, the setting of Example 1.1 exhibits a case in which the restrictions required in Section 4 for reduction to the dependency-free case do not apply (they involve keys and foreign keys, while dependency (1) is neither). Indeed, it is easy to check that the chased program does not support the chased query in the absence of dependencies. We therefore need a qualitatively better approach, which is provided by algorithm **testSupport**: Example 5.5 shows that the call to findDescriptors yields (among others) the descriptors d_1, d_2 , which, according to Example 5.4, witness support.

Algorithm testExpressibility. While we could use the reduction from expressibility to support used in Theorem 3.2, the following variation on **testSupport** constitutes a direct

 $^{^2 \}mathrm{Technically},$ descriptors for EDB rule IDBs using emptydomain partial mappings do not fully conform to Definition 5.1 as the expansion fragment contains a goal that is not the image under the partial mapping. As seen in the IDB rule step, the definition holds for all other IDBs, which are the pre-normalization IDBs.

test: call **findDescriptors**, keep only the descriptors for distinguished IDB predicates, and perform the test of Corollary 5.1 only on singleton sets of descriptors.

Finding the actual views. So far, we have only provided algorithms for deciding support and expressibility. To turn them into algorithms exhibiting the actual views generated by \mathcal{P} as well as the rewriting using it requires extra bookkeeping. All we need to do is to carry along with a descriptor d the actual expansion built during its derivation, noticing that the derivation tree of d coincides with the expansion tree of the expansion described by d.

Finding minimized witnesses for support. Let us note that while the partial rewriting candidate described by \mathcal{D}' in algorithm **testSupport** may contain redundant atoms, in security applications we only need to check if a query is supported by authorized views [24], which amounts to checking the existence of a rewriting without ever using it. Instead, the original query is executed once it is authorized. For non-security applications in which the wrapper needs to find and execute the rewriting in order to service a user application, one can plug in any technique for minimization under constraints already studied in the literature. One of them is the backchase minimization [8] which starts from the rewriting candidate (corresponding to R from Corollary 5.1) and considers subsets of view atoms at a time. This technique is amenable to further optimization by reusing the information from the partial mappings stored in the descriptors: find subsets of descriptors whose partial mappings are compatible and yield a total mapping from the query into the partial rewriting candidate. The presentation of such an optimization algorithm combining the discovered descriptors more efficiently goes beyond the scope of this paper.

5.1 Revisiting the Dependency-free Case

Based on algorithm **testSupport**, we now improve the previously best-known upper bound for checking support in the dependency-free setting. [15] reported a deterministic doubly-exponential upper bound in the size of the query and program. We obtain an exponentially improved upper bound, implied by Theorem 5.2 and Theorem 5.1:

COROLLARY 5.3. In the absence of dependencies, algorithm testSupport

- (a) is a decision procedure for support of a query Q by an arbitrary program \mathcal{P} , and
- (b) runs in deterministic EXPTIME in the sizes of \mathcal{P} and Q.

We next show that this upper bound is tight in the program size, and tight for practical purposes in the query size.

THEOREM 5.3. $\text{SUPP}_{\mathcal{P}}^{\emptyset}(Q)$ is NP-hard in the size of Q and EXPTIME-complete in the size of \mathcal{P} .

6. BOUNDARIES OF DECIDABILITY

We next justify the restrictions of Section 4 by exploring the boundaries of decidability for the problems of expressibility and support. To calibrate our results, we start with the following: allowing unrestricted sets of constraints immediately leads to undecidability even if the program expresses a single view. This result is unsurprising given that unrestricted sets of embedded dependencies notoriously lead to undecidability of many fundamental database decision problems, such as equivalence of queries and implication of dependencies [1]: THEOREM 6.1. If C contains arbitrary embedded dependencies, $\text{EXPR}^{\mathcal{P}}_{\mathcal{P}}(Q)$ and $\text{SUPP}^{\mathcal{C}}_{\mathcal{P}}(Q)$ are undecidable even if \mathcal{P} expresses a single view.

Theorem 6.1 shows that decidability requires the set of constraints to conform at least to the restrictions yielding decidability in the single-view case. The most permissive restriction known to date requires C to be a weakly acyclic set of embedded dependencies [9, 10]. As we show below, weak acyclicity turns out to be too generous for sets of views described by unrestricted programs.

Indeed, it turns out that the interaction of recursion in the program and the presence of dependencies leads to undecidability even under strong restrictions on the dependencies and on the program which are known to lead to decidability in many classical decision problems as long as recursion and dependencies are mutually exclusive. For instance, query rewritability using finitely many views (listed explicitly, not described by a program) is known to be decidable under weakly acyclic dependencies [9], in particular under only functional dependencies (which include key constraints), or only full TGDs. In the absence of dependencies, expressibility and support for arbitrary recursive programs is decidable [15]. Moreover, many classical undecidable Datalog-related problems, such as containment and boundedness (undecidable by [12]) are known to become decidable for recursive *monadic* programs [6]. However when considering recursion and dependencies together, we obtain surprisingly strong undecidability results.

Recall that a program is *monadic* if all its IDB predicates have arity 1, and it is *linear* if each rule body contains at most one intentional subgoal.

THEOREM 6.2. If \mathcal{P} is recursive and not key-safe, then $\text{Expr}_{\mathcal{P}}^{\mathcal{C}}(Q)$ is undecidable even if \mathcal{C} consists of a single key constraint, and \mathcal{P} is a linear monadic program.

This justifies our key-safety restriction, showing that it is maximally permissive. Theorems 6.2 and 3.2 immediately yield:

COROLLARY 6.1. If \mathcal{P} is recursive and not key-safe, then SUPP $_{\mathcal{P}}^{\mathcal{P}}(Q)$ is undecidable even if \mathcal{C} consists of a single key constraint and \mathcal{P} is a linear monadic program.

Sets of full TGDs are trivially weakly acyclic, and yet we have:

THEOREM 6.3. If \mathcal{P} is recursive, then $\operatorname{ExpR}_{\mathcal{P}}^{\mathcal{C}}(Q)$ is undecidable even if \mathcal{C} contains only full TGDs and \mathcal{P} is a monadic program.

COROLLARY 6.2. If \mathcal{P} is recursive, then $\operatorname{Supp}_{\mathcal{P}}^{\mathcal{C}}(Q)$ is undecidable even if \mathcal{C} contains only full TGDs and \mathcal{P} is monadic.

Since INDs are a particular case of TGDs, it is interesting to contrast Theorem 6.3 and Corollary 4.3. Notice that there is no contradiction here, as weakly acyclic sets of INDs and sets of full TGDs have incomparable expressive power: weakly acyclic sets of INDs can express non-full TGDs, but INDs allow only one atom in the premise, while full TGDs allow multiple atoms.

7. RELATED WORK

The necessity of describing infinite families of views exported by the source was first argued in [21] and the problem of deciding support first solved (in the absence of constraints) in [14, 15]. [15] pioneers the idea of reducing support to rewriting the query using finitely many views. Views generated by the program are compared for *interchangeability*: V_1 and V_2 are interchangeable if in every rewriting R of Q, by replacing the V_1 goals with V_2 goals we still obtain a rewriting. [15] shows that interchangeability induces finitely many equivalence classes on the set of all views generated by the program, and gives an algorithm to find one representative of each class. This finite set of representative views is then used to check for a rewriting. The resulting algorithm runs in doubly-exponential deterministic time. We can show however that interchangeability under dependencies yields infinitely many equivalence classes, thus precluding the reduction from [15] (see Example F.1 in Appendix F). Even in the absence of dependencies, we observe that interchangeability is unnecessarily strong, leading to a refinement of the view equivalence classes that yields exponentially more representatives than truly needed. Intuitively, instead of interchangeability in *every* rewriting of Q, the descriptor similarity condition (†) from Section 5 detects interchangeability with respect to only the *canonical* rewriting. This allows us to manipulate mapping/partial mapping pairs rather than sets thereof as in [15], which yields the upper bound improvement from doubly-exponential to single-exponential time.

In the dependency-free setting, [27] improves the upper bound for support of [15] to non-deterministic exponential time in the combined query and program size. However for practical purposes this still yields implementations that run in doubly-exponential time. In addition to the extension to constraints, our solution improves on [27] even in the dependency-free case, by achieving an exponentially better upper bound, proven to be essentially tight.

The problem of support strictly extends that of rewriting queries using finitely many views. The latter was treated in depth in the literature, considering various extensions pertaining to the language of queries and views [13, 3, 2, 5], to adding limited access patterns for the views [11, 19], to adding constraints (see the references in [8]), and to mixing such extensions [7]. The problem is NP-complete in the size of the query and views, in practice leading to deterministic exponential-time implementations, which is no better than for support. Prior work on information integration [16] studied answering queries using a finite set of views with limited access patterns with a different goal, namely finding maximally contained answers.

8. CONCLUSION

In this paper, we revisit the problem of deciding support and expressibility of a conjunctive query by (possibly parameterized) views generated as the expansions of a Datalog program, investigating for the first time the effect of source constraints.

We identify practically relevant restrictions on the program which lead to decidability for the most prevalent constraints in practice (weakly acyclic sets of keys and foreign keys). Moreover, we show that even slight relaxations to our restrictions lead to undecidability. We present an algorithm which is applicable to unrestricted programs and weakly acyclic sets of embedded dependencies, yielding a decision procedure in all known decidable cases, and a sound test in general.

We also settle two problems left open by work on the constraint-free case. First, we show that in the absence of

constraints our algorithm is a decision procedure which improves the previously known upper bounds for support in the absence of constraints (from 2-EXPTIME [14] and NEXP-TIME [27] to EXPTIME in the query and program size). We also give practically tight lower bounds, showing EXPTIMEhardness for fixed query and NP-hardness for fixed program. Second, we show that expressibility and support are interreducible in PTIME (even under constraints), which allows us to use essentially the same algorithm for solving them.

Note that the support problem discussed in this paper and in prior work decides whether a user query can be handled by the source or not by testing the existence of an *exact rewriting* using the generated views. This indeed represents the fundamental functionality one may expect in a limited-query-capability setting. However, when no equivalent rewriting exists, a user may also accept a best-effort approach in which instead of the exact answer she obtains its tightest *approximations*, either from below (contained in the answer) or from above (containing the answer). These approximations are known in the literature on view-based query rewriting as the *minimally containing* and *maximally contained* rewritings of the query. We leave for future research the problem of answering approximately a query using a potentially infinite family of views.

9. **REFERENCES**

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APPENDIX

A. WEAK ACYCLICITY

We repeat for the reader's convenience the definition of weakly acyclic set of dependencies, and the associated result.

DEFINITION A.1. (Weakly Acyclic)[9, 10] A position is a pair (R, i) (which we write R^i) where R is a relation symbol of arity r and i satisfies $1 \le i \le r$. The dependency graph of a set Σ of TGDs is a directed graph where the vertices are the positions of the relation symbols in Σ and, for every TGD ξ of the form

$$\forall \bar{u}, \bar{w} \ \phi(\bar{u}, \bar{w}) \longrightarrow \exists \bar{v} \ \psi(\bar{u}, \bar{v})$$

there is an edge between R^i and S^j whenever (1) some $u \in \{\bar{u}\}$ occurs in R^i in ϕ and in S^j in ψ or (2) some $u \in \{\bar{u}\}$ appears in R^i in ϕ and some $v \in \{\bar{v}\}$ occurs in S^j in ψ . Furthermore, these latter edges are labeled with \exists and we call them existential edges. A set Σ of TGDs and EGDs is weakly acyclic if the dependency graph of its TGD set has no cycles through an existential edge.

THEOREM A.1. For every weakly acyclic set C of embedded dependencies, there are b and c such that, for any set of subgoals A, regardless of the order of the chase, $chase_{C}(A)$ is guaranteed to terminate in $O(|A|^{b})$ steps and in time $O(|A|^{c})$, where |A| denotes the size of A.

B. REFINED VERSION OF KEY-SAFETY

The notion of key-safety presented in the main text keeps only track of the positions bound to constants, ignoring the actual constant values that may appear in these positions. As a consequence, it may fail to detect decidable instances of expressibility or support, where the constraints can still be ignored after the program and the query have been chased. We give in this section a refined definition for the key-safety restriction that addresses this problem, allowing us to solve strictly more cases by reduction to the dependency-free case.

Let R be a relation with an n-attribute composite key. By a *template of constants* (in short, template) for the key of R we denote a sequence of values $T = (v_1, \ldots, v_n)$, where each v_i can be either a constant value or a special value denoted *blank*. By the variable positions of T we denote the ordered sequence of positions $\bar{P}_T = (p_1, \ldots, p_k)$ of T that are occupied by *blank*.

We say that a rule of \mathcal{P} outputs the key of R, by template T, into the sequence of head variables $\overline{X} = (X_{i_1}, \ldots, X_{i_k})$ if \overline{X} appears in the rule body either

- in the positions $\bar{P}_T = (p_1, \ldots, p_k)$ of the key attribute sequence of some *R*-subgoal, with the remaining n - kpositions (if any) of the key being bound to the constant values given in *T*.
- in the positions j_1, \ldots, j_k of some *p*-subgoal, where *p* is an IDB predicate with at least one rule that in turn outputs the key of *R* by the template *T*, into the sequence of head variables with indices j_1, \ldots, j_k .

We say that a subgoal g outputs the key of R, by template T, into the sequence of variables $\bar{X} = (X_{i_1}, \ldots, X_{i_k})$ if

• g uses EDB predicate R and \bar{X} appears in positions $\bar{P}_T = (p_1, \ldots, p_k)$ in the key attributes of g, with the remaining n-k positions (if any) of the key being bound to the constant values given in T, or

 g uses IDB predicate p and there exists some rule defining p which outputs the key of R, by the template T
, into variables X
.

We say that a *rule is safe* for the key constraint on R if whenever one of its IDB subgoals outputs the key of R by some template of constants T into k variables $\bar{X} = (X_{i_1}, \ldots, X_{i_k})$, no other subgoal does the same (for the same template T). Notice that several EDB subgoals may output the key of the same R by the same template and into the same sequence of variables \bar{X} , as long as no IDB goal does. A program \mathcal{P} is key-safe for a set of key constraints \mathcal{K} if

- each rule is safe for all key constraints in \mathcal{K} , and
- for all distinguished predicates ans of \mathcal{P} , all defining rules r of ans, and all relational symbols R in the schema, if r outputs the key attributes \overline{A} , by some template, of some goal $R(\overline{A}, \overline{B})$, it also outputs all non-key attributes \overline{B} , by some template (using the same definition that applied to the key attributes).

If \mathcal{I} is a set of weakly acyclic INDs, we say that \mathcal{P} is key-safe for $\mathcal{C} = \mathcal{K} \cup \mathcal{I}$ if $chase_{\mathcal{I}}(\mathcal{P})$ is key-safe for \mathcal{K} . Notice that this new definition of key-safety can still be checked in PTIME in the size of \mathcal{P} and \mathcal{K} .

EXAMPLE B.1. Assuming the schema and constraints of Example 4.3, consider the following modified program \mathcal{P}'

$$\begin{array}{rcl} (\mathcal{P}') & ans(A,B,C,O) & :- & ind(A,B,O), ind'(B,O) \\ & ind(B,C,O) & :- & t(B,B',O), ind(B',C,O) \\ & ind(B,C,O) & :- & t(B,C,O) \\ & ind'(B,O) & :- & ind_P(B,O), ind_{SD}(B,O) \\ & ind_P(B,O) & :- & t(B, "Paris",O) \\ & ind_{SD}(B,O) & :- & t(B, "SanDiego",O) \end{array}$$

Notice that \mathcal{P}' is not key-safe under the weaker restriction, since the rule defining ind' is not safe. But we can easily see that the two constants appearing in the second attribute of the key cannot be equated during the chase, and \mathcal{P}' is indeed key-safe under the refined definition.

More precisely, in the sixth rule, b outputs the key into the sequence of variables B, by the template

$$T_1 = (blank, "SanDiego").$$

Similarly, in the fifth rule, b outputs the key into the sequence of variables B, by the template

$$T_2 = (blank, "Paris").$$

Since T_1 and T_2 are different, the rule defining ind' is safe. Then, in the first rule, the ind' subgoal outputs the key in B, by any of these two templates. Finally, and outputs the key attributes in A, B (by the ind subgoal) and in B (by the ind' subgoal) but in both cases it also outputs O, the non-key attribute.

C. PARAMETERS

Our solutions to checking expressibility and support can be extended to the case when sources implement parameterized queries, expecting applications to provide the parameter values (recall Example 1.2).

There are two kinds of query evaluation plans one may adopt in the presence of parameters. The straightforward execution consists in the wrapper issuing in a first stage a series of service calls to the source without inspecting any intermediate results to determine how to instantiate parameters for the other calls. Once all call results come in, during the second stage the rewriting query is run over them and the result passed to the application query. This is the approach taken in [14, 15]. We shall call this approach the *two-stage* evaluation. A more sophisticated evaluation strategy is based on the idea of interleaving query execution at the wrapper with service calls to the source. The evaluation of a subquery of the rewriting can thus provide parameter values for the subsequent calls needed by the non-evaluated part of the rewriting. This approach is used in [26] and, for finite sets of parameterized views, in [11], where it is known as the *dependent-join* evaluation.

If only two-stage evaluation is considered, there is an immediate reduction to the problem of non-parameterized views, based on the following observation:

LEMMA C.1. In two-stage evaluation, for the views to be relevant to the problem of support or expressibility, their parameters must be filled in with constants appearing in the query or the source dependencies.

This result follows immediately from Lemma 5.1 and generalizes a similar observation from [14] to the presence of dependencies. It implies that it suffices to generate a new program in which the parameters are replaced in all possible ways by the (bounded) set of constants in Q and C, and test support and expressibility for the new program. In practice, an efficient implementation would extend the rewriting algorithm as suggested in [14], by mapping parameters into constants from the query.

We next present expressibility and support under the more advanced dependent-join evaluation strategy. Our solution comes with no complexity overhead, in the sense that in the dependency-free case, our decision procedures have the same complexity as in the parameter-less case. This is nontrivial since the number of parameters that can occur in an expansion is a priori unbounded.

We start by introducing some auxiliary notions.

Notation. For parameters we adopt the ?X notation of [14, 15], enabling the generation of parameterized views. We stress that by this notation, an input variable ?X will be considered different from some other variable X appearing in the same program rule.

Access patterns. An access pattern for a view $V(X_1, \ldots, X_k)$ is an expression α in $\{o, i\}^k$. We say that the X_j is an *output* (resp. *input*) variable if $\alpha(j) = o$ (resp. $\alpha(j) = i$). A view with access pattern α is denoted $V^{\alpha}(X_1, \ldots, X_k)$. Views generated by a Datalog program with parameters will be presented using this notation, by introducing an input head variable for each parameter.

Executable query. Notice that, for a (rewriting) query whose atoms have access patterns, there may not always be a way to satisfy the bindings for the input variables, i.e. the query is not executable. Following [20, 7], we say that a query R formulated in terms of view names with binding patterns \mathcal{V} is *executable* if the access patterns of R are such that every input variable appears first in an output position of some previous goal.

Expressibility / **Support.** We are now ready to extend the definitions of expressibility and support in the presence of parameterized views. We say that a query Q is *expressible* by a program \mathcal{P} iff the query is equivalent to a query obtained from an expansion of \mathcal{P} by replacing all input variables by constants. Note that this is the natural choice,

since expressibility captures the cases in which a query can be fully answered by just one "service call", without any post-processing. We say that Q is *supported* by \mathcal{P} iff there exists an executable rewriting R using some finite set \mathcal{V} of views with access patterns generated by \mathcal{P} .

Before going into the specific details, we first give a brief outline on how the solutions of the previous section can be extended to deal with parameterized programs. As before, we aim at reducing these problems to query answering using only a finite family of the specified views, defined by descriptors. First, since by the dependent-join mechanism input variables play an important role in how view goals interact in a rewriting, we need to keep track in descriptors of their query-view and view-query mappings. While this leads to descriptors of unbounded size (since the number of input variables is not bounded), we show that only a finite set of descriptors needs to be considered. Then, we extend algorithm **testSupport** to find an *executable* ordering of a rewriting in terms of descriptors. For this phase, we show that an expensive ordering search can be avoided, by relying on a canonical executable rewriting candidate. In conclusion, similar to the case without parameters, we obtain a sound, exponential-time, algorithm for expressibility and support, which becomes complete in the absence of constraints or under restrictions on the interaction between program and constraints.

Modifying Example 1.1, the running example in this section is the following:

EXAMPLE C.1. Consider the schema from Example 1.1 extended with a relation airport(name) and the set of views specified by the parameterized Datalog program \mathcal{P}'' , with 2 distinguished IDB predicates (ans₁ and ans₂):

$$\begin{array}{rcl} ans_1(A) & :- & a(A) \\ ans_2(A,B) & :- & f(A,C), ind(C,B) \\ ind(C,B) & :- & f(C,C'), ind(C',B) \\ ind(C,?B) & :- & f(C,?B), b(?B, "Paris") \\ ind(C,?B) & :- & f(C,?B), t(?B, "Paris") \end{array}$$

Note that the program differs from the one of Example 1.1 in two aspects: (a) the source admits direct access to the airport relation (by ans_1) and (b) the destination of views concerning itineraries (by ans_2) is an input variable.

Besides dependency (1)

$$\forall A, S \ t(A, S) \land b(A, S) \longrightarrow s(A, S)$$

we assume the source verifies also the dependency

$$\forall A \ b(A, "Paris"), t(A, "Paris") \longrightarrow a(A) \tag{4}$$

which guarantees that any airport with a bus and train connection to Paris can be found in the airport relation.

Consider that the user asks the same query as in Example 1.1, i.e., itineraries of length 2 ending in an airport from which Paris is reachable by all the three transportation means

$$\begin{array}{rcl} (Q) & q(A,B) & :- & f(A,C), f(C,B), t(B, ``Paris"), \\ & b(B, ``Paris"), s(B, ``Paris"). \end{array}$$

We recall that this query was supported in the setting of Example 5.1, as witnessed by the rewriting

(R)
$$r(A,B) := V_1(A,B), V_2(A,B).$$

However, under the given access patterns, R no longer witnesses support since the conjunction of $V_1^{oi}(A, B)$ and $V_2^{oi}(A, B)$ is not executable. But by adding to this rewriting $U^o(B)$ as a first subgoal, where U is the one view generated by predicate ans₁, the query becomes executable:

$$(R') r(A,B) := U^o(B), V_1^{oi}(A,B), V_2^{oi}(A,B),$$

and moreover equivalent to Q. Indeed, the values for B are now passed by the dependent-join, and it can be easily checked that R' is equivalent to Q under the two dependencies, since the airport goal of the rewriting maps in the result of chasing Q with (4).

We next discuss the decision procedure for view-based query answering using a *finite* set of parameterized views, under dependencies. This procedure will be then adapted to a finite set of view descriptors.

Answerable part. Given a query R formulated in terms of view names with access patterns \mathcal{V} , we call the *answerable part* of R (denoted ans(R)) the executable query with the same head as R and the body built one goal at a time from body(R) as follows:

- start with an empty set of bounded variables, \mathcal{B} , then repeatedly
- find the first view goal g^α(X̄) in R not added to ans(R) such that all the input variables of g are in B; add this goal to ans(R) and add its head variables X̄ to B.

Clearly, ans(R) is an executable query and this procedure runs in quadratic time.

Executable Canonical Rewriting Candidate. Given a finite set of views with access patterns \mathcal{V} , an acyclic set of constraints \mathcal{C} , and a query Q, we call the *executable canonical rewriting candidate* of Q using \mathcal{V} under \mathcal{C} , denoted $ECRC_{\mathcal{V}}^{\mathcal{C}}(Q)$, the query obtained as follows:

(i) compute $CRC^{\mathcal{C}}_{\mathcal{V}}(Q)$ (as described in Section 5),

(ii) find its answerable part, $\operatorname{ans}(CRC_{\mathcal{V}}^{\mathcal{C}}(Q))$.

Similar to Lemma 5.1, results from [7] guarantee that Q has a rewriting using \mathcal{V} under \mathcal{C} iff $ECRC_{\mathcal{V}}^{\mathcal{C}}(Q)$ is itself one. We omit further details and only illustrate the main idea by an example.

EXAMPLE C.2. Revisiting Example C.1, we know that the views V_1, V_2, U , generated (among others) by \mathcal{P}'' , give an executable rewriting for Q, under $\mathcal{C} = \{(1), (4)\}$.

Assume that an additional distinguished predicate is present in \mathcal{P}'' , defined by the rule:

$$ans_3(A,B) := f(A,?C), ind(?C,B)$$

This rule generates, among others, two views that have the same subgoals as V_1 and V_2 , but in which the intermediary stop is an input variable, too. Hence these views, denoted W_1 and W_2 , will have one output and two inputs, their access pattern being (o, i, i).

Consider the set of views $\mathcal{V} = \{V_1, V_2, U, W_1, W_2\}$, which can all be mapped into $chase_{\mathcal{C}}(Q)$. By evaluating them on the body of $chase_{\mathcal{C}}(Q)$, we obtain the intermediary $CRC_{\mathcal{V}}^{\mathcal{C}}(Q)$:

$$R(A,B) := V_1^{oi}(A,B), V_2^{oi}(A,B), U^o(B), W_1^{oii}(A,C,B), W_2^{oii}(A,C,B)$$

which is not executable since no value can be assigned to the C input variable. However, by computing $\operatorname{ans}(CRC^{\mathbb{C}}_{\mathcal{V}}(Q))$,

we eliminate the last two goals and reorder the rewriting, obtaining the $ECRC_{\mathcal{V}}^{\mathcal{C}}(Q)$:

$$R(A,B) := U^{o}(B), V_{1}^{oi}(A,B), V_{2}^{oi}(A,B)$$

which we know is indeed an executable rewriting of Q.

Testing expressibility and support. Similarly to the case without access patterns, we capture the usefulness of a view in the executable rewriting candidate by a descriptor, which takes also into account parameters and the access patterns they impose. Once the set of descriptors is obtained, checking expressibility amounts to checking if one of them denotes a view which becomes equivalent to Q after replacing input variables by constants. For testing support, we first construct the *partial rewriting candidate*, as described in Section 5. Since this candidate may not be executable, we need to compute its answerable part, which we call the *executable partial rewriting candidate*. Finally, we check as in Corollary 5.1 whether this candidate is equivalent to Q under the dependencies, starting from the corresponding view fragments.

Finding descriptors. Since now we need to describe also the role of view goals in the answerable part of the rewriting candidate, we enrich the descriptor definition by taking into account input variables. More precisely, (a) input variables are treated as head variables, and (b) we add the corresponding access patterns to each descriptor, thus discriminating among views which are similar according to Definition 5.2 if they have distinct access patterns. We omit the formal definition and illustrate these changes on the setting of Example C.2:

EXAMPLE C.3. The descriptor for the view V_1^{oi} has, besides the components given in Example 5.2, the access pattern $\alpha_1 = (o, i)$. Similarly, the descriptor for the view W_1^{oii} has the components

$$E_{1} = [f(Z_{1}, Z_{2}), f(Z_{2}, Z_{3}), t(Z_{3}, "Paris")]$$

$$p_{1}(t_{1}) = ans(A, C, B)$$

$$fr_{1} = \{A : Z_{1}, C : Z_{2}, B : Z_{3}\}$$

$$\alpha_{2} = (o, i, i)$$

One difficulty in extending descriptors in this way comes from the fact that there may be no bound on the number of input variables of generated views, leading to an unbounded number of descriptors and excluding any rewriting approach based on descriptors. However, we know from [22] that, in the absence of constraints, if a rewriting using views with binding patterns exists, then one with at most n (the number of variables of Q) distinct variables exists. This can in fact be extended to the case with constraints, showing that if a rewriting with a finite set of views exists, then there is also one in which the view atoms have at most n input variables, n being the number of variables of $chase_{\mathcal{C}}(Q)$. The intuition for this is that if a view with more than n parameters appears in a rewriting, then for sure some of those parameters will be bound to the same value. Hence it is sufficient to consider only descriptors with at most n inputs. Moreover, it was shown in previous work [9], that if the constraints C are weakly acyclic, n is upper-bounded by a polynomial in the size of Q whose largest exponent depends only on C.

Therefore, the procedure **findDescriptors** can easily be extended to take parameters into account. The bottomup step will infer descriptors in which the binding pattern component may contain up to n distinct input variables. The following theorem summarizes the results of this section:

THEOREM C.1. If C is weakly acyclic, the following hold:

- Procedure findDescriptors extended with parameters outputs all pairwise dissimilar descriptors and is guaranteed to terminate in time exponential in the sizes of P, C and Q.
- Procedure testSupport extended with parameters is a sound algorithm for checking support and runs in time exponential in the sizes of P, C and Q. It becomes a complete decision procedure if P is a C-local program generating C-independent views.

D. PROOFS

Proof: (Theorem 3.1) Given Q, \mathcal{P} and \mathcal{C} , we construct a new query Q', program \mathcal{P}' and set of dependencies \mathcal{C}' , such that Q is supported by \mathcal{P} under \mathcal{C} iff Q' is expressible by \mathcal{P}' under \mathcal{C}' .

The reduction starts from the following result, which generalizes a result of [13] to the presence of dependencies:

LEMMA D.1. Let C be a weakly acyclic set of embedded dependencies. Then $\text{SUPP}^{\mathcal{P}}_{\mathcal{P}}(Q)$ holds iff there is a rewriting R of Q under C using views generated by \mathcal{P} , where R has no more variables than chase_C(Q).

It was shown in prior work [7] that, if C is weakly acyclic, then $chase_{\mathcal{C}}(Q)$ contains v variables, where v is upper-bounded by a polynomial in the number of goals in Q and exponential in the maximum arity of a relation appearing in the conclusion of a dependency in C.

From this, we will build in PTIME in the size of $chase_{\mathcal{C}}(Q)$ and \mathcal{P} a new program \mathcal{P}' that basically enumerates all possible conjunctions of expansions of \mathcal{P} .

For this proof, it helps to consider Q and conjunctions of expansions as *rectified*. More precisely, no constants are allowed in predicate subgoals, and no variable appears twice in subgoals. Instead, joins are made explicit by subgoals equals(X, Y), and selections with a constant c by subgoals equals(X, c). Note that we can pass from any conjunctive query to its rectified version and vice-versa in linear time.

Given Q, denote with a_Q the arity of Q (the number of its distinguished variables). Assume w.l.o.g. that the distinguished predicate of \mathcal{P} is *ans*, of arity $a_{\mathcal{P}}$. We add a new IDB predicate *ans'*, as well as a new unary EDB predicate D.

We build the following program, of distinguished predicate

ans':

$$ans'(V_1, ..., V_{a_Q}) := pick(V_1, X_1, ..., X_v), pick(V_2, X_1, ..., X_v), pick(V_{a_Q}, X_1, ..., X_v), temp(X_1, ..., X_v) := D(X_1), ..., D(X_v) temp(X_1, ..., X_v) := ans(Y_1, Y_2, ..., Y_{a_P}), pick(Y_1, X_1, ..., X_v), pick(Y_2, X_1, ..., X_v), pick(Y_{a_P}, X_1, ..., X_v), pick(Y_{a_P}, X_1, ..., X_v), temp(X_1, ..., X_v) pick(V, X_1, ..., X_v) := equals(V, X_1), D(V), D(X_1), ..., D(X_v) pick(V, X_1, ..., X_v) := equals(V, X_2), D(V), D(X_1), ..., D(X_v) pick(V, X_1, ..., X_v) := equals(V, X_v), D(V), D(X_1), ..., D(X_v) pick(V, X_1, ..., X_v) := equals(V, X_v), D(V), D(X_1), ..., D(X_v)$$

modified rules of \mathcal{P}

with D atoms for all variables

The rules of $\mathcal P$ appear in $\mathcal P'$ modified as follows. Each rule of $\mathcal P$ of the form

$$head_i(\bar{X}_i) :- body_i(\bar{X}_i, \bar{Y}_i)$$

is transformed into a rule

$$head_i(X_i) := body_i(X_i, Y_i),$$
$$D(X_1), \dots, D(X_{n_i}), D(Y_1), \dots, D(Y_{m_i})$$

The rules added in addition to those of \mathcal{P} have the task of expressing all possible conjunctive queries with a_Q head variables and at most v variables in total, formulated against the distinguished goal of \mathcal{P} , ans. The ans subgoals are then expanded into views generated by \mathcal{P} (plus D subgoals), due to the inclusion of the (modified) rules of \mathcal{P} into \mathcal{P}' .

Note that the *temp* subgoal lists the pool of v variables the expansions of \mathcal{P}' will use. Each *temp* subgoal expands into arbitrarily many *ans* subgoals which will build the body of the rewriting. The variables appearing in the head *ans* ' and in the various *ans* subgoals in the body are each associated with *pick* subgoals. The *pick* subgoal has v possible expansions, each having the role of picking one of the v variables in the pool to equate with the variable in its first argument. In this way, every assignment of variables from the pool to variables of (the head and body of) the conjunctive query over *ans* subgoals is realizable by some expansion of the *pick* subgoals.

The D predicate is introduced for technical purposes, to avoid generating unsafe Datalog rules for the *pick* goal. Its effect is that each view generated by \mathcal{P}' has a D subgoal for each of its variables. This does not influence expressibility as long as we add such subgoals for all variables appearing in the query and in the dependencies. Indeed, if Q has the form

$$Q(Z_1,\ldots,Z_{a_Q})$$
 :- $body(Z_1,\ldots,Z_{v_Q})$

with body a conjunction of subgoals, we build a new boolean query

$$Q'(Z_1, \dots, Z_{a_Q}) := body(Z_1, \dots, Z_{v_Q}),$$
$$D(Z_1), \dots, D(Z_{v_Q})$$

Finally, we construct a new set of dependencies \mathcal{C}' by adding in the conclusion of each dependency σ from \mathcal{C} the predicate D(X) for every variable X appearing in σ .

Notice that \mathcal{C}' and Q' are obtained in linear time from \mathcal{C} and Q, respectively. \mathcal{P}' is obtained in PTIME from \mathcal{P} and v, where the latter is polynomial in the size of Q but exponential in the maximum arity of a relation appearing in the conclusion of some dependency from \mathcal{C} .

It is easy to show that $\text{SUPP}_{\mathcal{P}}^{\mathcal{C}}(Q)$ holds if and only if $\text{EXPR}_{\mathcal{P}'}^{\mathcal{C}'}(Q')$ does.

Proof: (Theorem 3.2) Given Q, \mathcal{P} and \mathcal{C} , we construct a boolean query Q'', boolean program \mathcal{P}'' and set of dependencies \mathcal{C}'' , such that Q is expressible by \mathcal{P} under \mathcal{C} iff Q''is supported by \mathcal{P}'' under \mathcal{C}'' .

For presentation simplicity, we first show a first-cut solution which works only if the query graph is connected, then we explain how the reduction can be adapted to arbitrary queries.

Denote with a_Q the arity of Q and assume w.l.o.g. that Q has the form

$$Q(Z_1,\ldots,Z_{a_Q})$$
 :- $body(Z_1,\ldots,Z_{v_Q})$

with body a conjunction of subgoals and $v_Q \ge a_Q$ the total number of variables appearing in Q. We build the boolean query

Q'() :- $head(Z_1,\ldots,Z_{a_Q}), body(Z_1,\ldots,Z_{v_Q})$

using a fresh EDB relation *head* of arity a_Q .

Assume w.l.o.g. that the distinguished IDB of \mathcal{P} is ans. Notice that, for Q to be expressible by \mathcal{P} , ans must have the same arity as Q. \mathcal{P}' is constructed by adding to the rules of \mathcal{P} a new rule defining a fresh, boolean IDB predicate ans':

$$ans'() :- ans(X_1, \ldots, X_{a_O}), head(X_1, \ldots, X_{a_O}).$$

The distinguished IDB predicate of \mathcal{P}' is ans'.

Note that the views generated by \mathcal{P}' are in one-to-one correspondence to those generated by \mathcal{P} : any view V' generated by \mathcal{P}' simply extends the body of some view V generated by \mathcal{P} with a *head* subgoal containing the head variables of V. Q is equivalent to V if and only if Q' is equivalent to the corresponding view V': the *head* subgoals appearing in both Q' and V' ensure the desired correspondence between the distinguished variables of Q and those of V. We have thus proven

Claim 1. $\operatorname{Expr}_{\mathcal{P}'}^{\mathcal{C}}(Q')$ iff $\operatorname{Expr}_{\mathcal{P}}^{\mathcal{C}}(Q)$.

Also note that, since each view generated by \mathcal{P}' is boolean, any rewriting using such views is really a Cartesian product thereof. We therefore make the following claim:

Claim 2. Consider a boolean query Q'' and the set of embedded dependencies \mathcal{C}'' . If

(a) Q" performs no Cartesian products (i.e. if its hypergraph [1] is connected), and (b) all constraints in \mathcal{C}'' have premises with connected hypergraph,

then Q'' is equivalent under \mathcal{C}'' to some boolean conjunctive query R iff it is equivalent under \mathcal{C}'' to a connected subquery of R.

Proof of Claim 2. The "if" direction is immediate, we prove the "only if" direction next.

Let Q'' be connected, and $R() := V_1(), V_2()$, where the hypergraphs of V_1 and V_2 are disjoint.

Assume toward a contradiction that $V_1 \not\sqsubseteq_{\mathcal{C}''} V_2$ and $V_2 \not\sqsubseteq_{\mathcal{C}''} V_1$. Then there must exist two databases, DB_1, DB_2 , with disjoint active domains, such that both DB_1, DB_2 satisfy \mathcal{C}'' , and such that $V_1(DB_1) = true, V_2(DB_1) = false, V_1(DB_2) = false$ and $V_2(DB_2) = true$. Since Q'' is equivalent to R under \mathcal{C}'' , we obtain that $Q(DB_1) = Q(DB_2) = false$.

Let DB_3 be the database obtained by unioning the two: $DB_3 := DB_1 \cup DB_2$.

We claim that DB_3 satisfies C'' as well: the components DB_1, DB_2 do so by hypothesis, and their disjoint union cannot violate any constraint in C'' because all constraint premises are connected and thus cannot match across the databases.

Note that $Q''(DB_3) = false$, as Q'' has no match into any of DB_1, DB_2 , and no match across them because it is connected. Also note that $R(DB_3) = true$, as V_1 and V_2 have a match against the sub-databases DB_1 and DB_2 , respectively.

We have thus exhibited a database $DB_3 \models \mathcal{C}''$ such that $R(DB_3) = true$, but $Q''(DB_3) = false$, contradicting the equivalence of Q to R under \mathcal{C}'' . Therefore, either of V_1, V_2 must be contained in the other under \mathcal{C}'' , so R can be minimized under \mathcal{C}'' to just one component. The reasoning extends to arbitrarily many components by induction. End of proof of Claim 2.

By Claim 2, we have that, under restrictions (a) and (b), all rewritings of Q' under C using views generated by \mathcal{P}' contain a single view goal or can be minimized to a single view goal. This implies that $\text{ExpR}_{\mathcal{P}'}^{\mathcal{C}}(Q')$ holds if and only if $\text{Supp}_{\mathcal{P}'}^{\mathcal{C}}(Q')$ does. Considering also Claim 1, we obtain $\text{Supp}_{\mathcal{P}'}^{\mathcal{C}}(Q')$ iff $\text{ExpR}_{\mathcal{P}'}^{\mathcal{C}}(Q')$.

We now refine the reduction, lifting restrictions (a) and (b). To this end, we obtain from Q', \mathcal{P}' and $\mathcal{C}, Q'', \mathcal{P}''$ and $\mathcal{C}^{"}$ such that $\text{Expr}_{\mathcal{P}}^{\mathcal{C}}(Q)$ holds iff $\text{Expr}_{\mathcal{P}''}^{\mathcal{C}''}(Q'')$ does, and such that Q'' and the premises of all constraints in \mathcal{C}'' are connected. Then Claim 2 will apply to Q'' and \mathcal{C}'' , completing the proof.

The head of Q'' is the same as that of Q'. The distinguished IDB of \mathcal{P}'' is the same as that of \mathcal{P}' . Every remaining goal and subgoal of Q' and \mathcal{P}' , say $G(\bar{X})$ of arity a, is extended to an a + 1-ary goal $G(\bar{X}, U)$, where U is a fresh variable shared across all goals.

We replace in the same way all subgoals appearing in dependencies in \mathcal{C} : for every $\sigma \in \mathcal{C}$ of form

$$\forall \bar{X} \ premise(\bar{X}) \rightarrow \exists \bar{Y} \ conclusion(\bar{X}, \bar{Y}),$$

we construct σ'' of form

$$\forall \bar{X} \forall U \ premise''(\bar{X}, U) \to \exists \bar{Y} \ conclusion''(\bar{X}, \bar{Y}, U),$$

where *premise*" and *conclusion*" are obtained from *premise* and *conclusion*, respectively, by extending the goals with the

new variable U, as done above for Q' and \mathcal{P}' .

Claim 3. $\operatorname{ExpR}_{\mathcal{P}'}^{\mathcal{C}}(Q')$ holds iff $\operatorname{ExpR}_{\mathcal{P}''}^{\mathcal{C}''}(Q'')$ does.

The theorem follows from Claims 1, 3 and 2. *Proof:* (**Theorem 4.1**) Let V be the view generated by

 \mathcal{P} , witnessing $\operatorname{Expr}_{\mathcal{P}}^{\mathcal{C}}(Q)$, i.e.

$$Q \equiv_{\mathcal{C}} V \tag{5}$$

Because C is weakly acyclic, the chase with it is guaranteed to terminate, so (5) is equivalent to

$$chase_{\mathcal{C}}(Q) \equiv chase_{\mathcal{C}}(V)$$
 (6)

Since \mathcal{P} is \mathcal{C} -local, there is W generated by $chase_{\mathcal{C}}(\mathcal{P})$ with

$$chase_{\mathcal{C}}(V) \equiv W.$$
 (7)

By (7) and (6) and by transitivity of \equiv relation, we obtain

$$chase(Q) \equiv W$$
 (8)

and thus W witnesses $\operatorname{ExpR}^{\emptyset}_{chase_{\mathcal{C}}(\mathcal{P})}(chase_{\mathcal{C}}(Q))$. The opposite direction is analogous.

Proof: (Theorem 4.2)

Assume w.l.o.g. that $\text{SUPP}_{\mathcal{P}}^{\mathcal{C}}(Q)$ is witnessed by the views $\mathcal{V} = \{V_1, \ldots, V_n\}$ generated by \mathcal{P} and the rewriting R in terms of \mathcal{V} . By \mathcal{C} -locality of \mathcal{P} , there are W_1, \ldots, W_n generated by $chase_{\mathcal{C}}(\mathcal{P})$ such that $W_i \equiv chase_{\mathcal{C}}(V_i)$ for every $1 \leq i \leq n$. We therefore have:

$$Q \equiv_{\mathcal{C}} expand_{\mathcal{V}}(R), \tag{9}$$

which is equivalent (due to weak acyclicity of
$$\mathcal{C}$$
) to

$$chase_{\mathcal{C}}(Q) \equiv chase_{\mathcal{C}}(expand_{\mathcal{V}}(R)),$$
 (10)

By \mathcal{C} -independence of \mathcal{V} there must be another query R' over the view schema such that

$$chase_{\mathcal{C}}(expand_{\mathcal{V}}(R')) \equiv expand_{\{chase_{\mathcal{C}}(V_1),\dots,chase_{\mathcal{C}}(V_n)\}}(R')$$
(11)

and

$$expand_{\mathcal{V}}(R') \equiv_{\mathcal{C}} expand_{\mathcal{V}}(R)$$
 (12)

hence again by weak acyclicity of \mathcal{C} ,

$$chase_{\mathcal{C}}(expand_{\mathcal{V}}(R')) \equiv chase_{\mathcal{C}}(expand_{\mathcal{V}}(R)) \equiv chase_{\mathcal{C}}(Q)$$
(13)

From (11) and (13), we infer:

$$chase_{\mathcal{C}}(Q) \equiv expand_{\{chase_{\mathcal{C}}(V_1),\dots,chase_{\mathcal{C}}(V_n)\}}(R'), (14)$$

which is equivalent to

$$chase_{\mathcal{C}}(Q) \equiv expand_{\{W_1,\dots,W_n\}}(R')$$
(15)

because the semantics of a query composition is preserved under replacement with equivalent queries.

But then $\{W_1, \ldots, W_n\}$ and R' witness

$$\operatorname{SUPP}_{chase_{\mathcal{C}}(\mathcal{P})}^{\emptyset}(chase_{\mathcal{C}}(Q)).$$

Proof: (Theorem 4.3)

Let α be a conjunctive query whose body contains both EDB and IDB relations of \mathcal{P} . We say that β is obtained in one expansion step with rule r, denoted

$$\alpha \stackrel{\prime}{\Longrightarrow} \beta,$$

iff β is obtained by expanding with r one of α 's subgoals that uses the IDB relation defined by r. Given a sequence of expansion steps

$$ans(\bar{X}) = \alpha_0 \stackrel{r_1}{\Longrightarrow} \alpha_1 \dots \stackrel{r_n}{\Longrightarrow} \alpha_n$$

where each r_i is a rule of \mathcal{P} and *ans* is a distinguished IDB of \mathcal{P} , we call each α_i a *partial expansion* of \mathcal{P} .

To prove the theorem, we claim more specifically that the result of chasing any partial expansion α of \mathcal{P} can be alternatively obtained by replacing the rules in the derivation of α with their chased form (these are rules of $chase_{\mathcal{C}}(\mathcal{P})$):

Claim 1. For every n, and every sequence of expansion steps

$$ans(\bar{X}) = \alpha_0 \stackrel{r_1}{\Longrightarrow} \alpha_1 \dots \stackrel{r_n}{\Longrightarrow} \alpha_n,$$

there is a sequence of expansion steps

$$ans(\bar{X}) = \beta_0 \stackrel{chase_{\mathcal{C}}(r_1)}{\Longrightarrow} \beta_1 \dots \stackrel{chase_{\mathcal{C}}(r_n)}{\Longrightarrow} \beta_n$$

such that for every $0 \le i \le n$, β_i is isomorphic to $chase_{\mathcal{C}}(\alpha_i)$.

Notice that, if the claim holds, then any partial expansion α_n of \mathcal{P} can be obtained (up to isomorphism) as an expansion β_n of $chase_{\mathcal{C}}(\mathcal{P})$. This immediately gives \mathcal{C} -locality for the particular case of full expansions (which are the generated views): to find the view W generated by $chase_{\mathcal{C}}(\mathcal{P})$ that corresponds to view V generated by \mathcal{P} , retrace the derivation of V by \mathcal{P} using the chased rules instead.

Proof of Claim 1. The claim is proven by induction on n, using for the induction step the observation that INDs have only one atom in the premise, so the EDB goals in α_i cannot cooperate with the new EDB goals introduced in α_{i+1} to enable a chase step. The chase of α_{i+1} therefore progresses in isolation on the goals appearing in α_i , and on the new goals introduced by the expansion step. Its effect is therefore alternatively achievable by expanding $chase_C(\alpha_i)$ in one step using the chased rule $chase_C(r_{i+1})$. End of proof of Claim 1.

 $\mathcal C\text{-independence}$ of the views follows from essentially the same observation about INDs, which actually gives a stronger result:

Claim 2. Any set \mathcal{V} of views (regardless of whether generated by some program or not) is \mathcal{C} -independent if \mathcal{C} consists only of INDs.

Indeed, any join of renamed copies of view bodies (corresponding to the expansion of some rewriting), when chased, gives the same result as chasing the view bodies in isolation and then joining them. This is because the single-atom premises of the INDs preclude the interaction (w.r.t. enabling chase steps) of goals from distinct view bodies.

Proof: (**Theorem 4.4**) C-locality follows from the fact that Claim 1 in the proof of Theorem 4.3 holds also for $C = \mathcal{I} \cup \mathcal{K}$, where \mathcal{I} is a set of INDs, and \mathcal{K} a set of key constraints.

The proof is similar, using for the induction step a few additional observations about the chase with key constraints and INDs.

A first observation is that the chase of any query α with C yields a result that is equivalent to that obtained by first

chasing with \mathcal{I} , then with \mathcal{K} :

$$chase_{\mathcal{C}}(\alpha) \equiv chase_{\mathcal{K}}(chase_{\mathcal{I}}(\alpha)),$$

as the key constraints from \mathcal{K} never introduce new variables or new relational atoms.

A second observation is the following. Since \mathcal{P} is key-safe for \mathcal{C} , this means by definition that $chase_{\mathcal{I}}(\mathcal{P})$ is key-safe for \mathcal{K} . This in turn implies that the result of $chase_{\mathcal{K}}(chase_{\mathcal{I}}(\alpha_i))$ can be alternatively obtained by expanding $chase_{\mathcal{K}}(chase_{\mathcal{I}}(\alpha_{i-1}))$ with $chase_{\mathcal{C}}(r_i)$. This is because the chase steps with INDs, as shown in the proof of Theorem 4.3, are triggered by single subgoals and not by the interaction of the EDB goals in α_{i-1} and the new goals in α_i . The same isolation property holds for the chase steps with key constraints. Key constraints (when expressed as dependencies) do have two goals in the premise, but due to the key safety restriction (see condition 1 in the definition of key-safety), the image of the premise can never span the EDB goals in α_{i-1} and the new EDB goals in α_i .

C-independence follows similarly. Say that the set of generated views is \mathcal{V} , and CR is a canonical rewriting of query Q using \mathcal{V} .

The way in which views are joined in CR depends on Q, and therefore CR could conceivably contain two view subgoals both of which output the key of some relation into the same variables \bar{X} . But then by construction of the canonical rewriting, Q itself joins two R-goals on the key, $R(\bar{X}, \bar{Y})$ and $R(\bar{X}, \bar{U})$. But then the key constraint applies when chasing Q (and this is a step in constructing the canonical rewriting), so \bar{U} must be the same variables as \bar{Y} in $chase_{\mathcal{C}}(Q)$, and therefore in CR. Since by key-safety of \mathcal{P} , all views that output keys must also output the non-key attributes, the equality of \bar{Y} and \bar{U} is guaranteed in the expansion of CR, and the chase step of CR with R's key constraint does not apply. In summary, we obtain that

$$chase_{\mathcal{C}}(expand_{\mathcal{V}}(CR)) =$$
$$ase_{\mathcal{K}}(chase_{\mathcal{I}}(expand_{chase_{\mathcal{C}}(\mathcal{V})}(CR))) = (16)$$

$$chase_{\mathcal{I}}(expand_{chase_{\mathcal{C}}(\mathcal{V})}(CR)) = (17)$$

$$expand_{chase_{\mathcal{T}}(chase_{\mathcal{C}}(\mathcal{V}))}(CR) = (18)$$

$$expand_{chase_{\mathcal{C}}(\mathcal{V})}(CR).$$
 (19)

Here, (18) follows from Claim 2 in the proof of Theorem 4.3, since any set of views, including $chase_{\mathcal{C}}(\mathcal{V})$ is \mathcal{I} -independent. (19) follows from the fact that $chase_{\mathcal{C}}(\mathcal{V})$ yields views on which no further chase step with any constraint in \mathcal{C} applies, in particular with the constraints in $\mathcal{I} \subseteq \mathcal{C}$.

ch

Proof: (**Theorem 5.1**) (1.) follows from Lemma 5.2(b) and Corollary 5.1.

(2.) follows from Lemma 5.2(a) and Corollary 5.1, noticing that the containment mapping cfr can be computed in EXPTIME in the size of Q and in PTIME in the size of the result of chasing the partial rewriting candidate. In turn, the size of the chase result is exponential in the maximum arity of a constraint in C and polynomial in that of the partial rewriting candidate [9]. The size of the partial rewriting candidate is given by the maximum number of distinct descriptors that can be built, which by Lemma 5.2(a) is worst-case exponential in the size of Q, C, and the maximum arity of a predicate in \mathcal{N} (which remains unchanged during normalization, so it coincides with the maximum arity of a predicate in \mathcal{P}). Notice from the proof of Lemma 5.2 that only the maximum arity a of the program \mathcal{N} , and the size s of its rules may appear in the exponent. The number of rules in \mathcal{N} does not appear in the exponent. It is easy to check that the normalization process affects only the number of rules (which blows it up exponentially from \mathcal{P} to \mathcal{N}) and preserves the values a and s from \mathcal{P} .

Proof: (Lemma 5.2) (a) Notice that the initialization stage and each individual rule step terminate, since the chase terminates when C is weakly acyclic. The set D must saturate, as there are only finitely many dissimilar descriptors. Their number is upper bounded by an exponential in the maximum arity of a predicate in P and the size of Q, which bounds the number of rule step applications. At every rule step, finding that the rule applies involves matching it against the set of descriptors, which is exponential in the rule size. By Theorem A.1, the ensuing chase terminates in time exponential in the size of C and polynomial in the size of the descriptor.

(b) An easy proof by induction on the structure of the derivation tree of each descriptor.

Proof: (Theorem 5.3) The NP lower bound follows from a reduction from the problem of checking conjunctive query equivalence (NP-complete by [4]), via the problem of checking expressibility. Given conjunctive queries Q_1, Q_2 , we have that $Q_1 \equiv Q_2$ iff Q_1 is expressible by the single-rule Datalog program Q_2 . The latter reduces in PTIME to the problem of support by Theorem 3.2.

The EXPTIME lower bound is obtained by a reduction from the problem of checking containment of a query Q in a Datalog program \mathcal{P} , known to be PTIME in the size of Q and EXPTIME-complete in the size of \mathcal{P} [23]. First, we carry out a reduction to the problem of checking expressibility, then compose it with the PTIME reduction from expressibility to support given by Theorem 3.2:

Given query $Q(\bar{X}) := body(\bar{X}, \bar{Y})$ and program \mathcal{P} of distinguished predicate ans (necessarily of same arity as the query), we construct program \mathcal{P}' which includes all rules of \mathcal{P} , the additional rule $ans'(\bar{Z}) := ans(\bar{Z}), body(\bar{Z}, \bar{Y})$ and pick ans' as the new distinguished predicate of \mathcal{P}' . Notice that \mathcal{P}' generates all intersections of Q with views generated by \mathcal{P} , whence we have that Q is contained in \mathcal{P} iff $\text{EXPR}^{\emptyset}_{\mathcal{P}'}(Q)$ holds.

Proof: (**Theorem 6.1**) The undecidability of support follows from the undecidability of expressibility and the reduction of Theorem 3.2. As for the undecidability of expressibility, it follows from a reduction from query containment under embedded dependencies (known to be undecidable [1]) to support of a query by a non-recursive Datalog program which expresses a single view.

Proof: (**Theorem 6.2**) The proof is by reduction from the Post Correspondence Problem (PCP), known to be undecidable [25, 1]. Let $\{v_i\}_{1 \le i \le n}$, $\{w_i\}_{1 \le i \le n}$ be the PCP instance, where v_i, w_i are words over alphabet $\{a, b\}$. This is a "yes" instance iff there exists a natural number l and a sequence of integers $\sigma \in \{1, \ldots, n\}^l$ such that

$$v_{\sigma(1)} \circ v_{\sigma(2)} \circ \ldots \circ v_{\sigma(l)} = w_{\sigma(1)} \circ w_{\sigma(2)} \circ \ldots \circ w_{\sigma(l)}$$

where $\sigma(i)$ denotes the i^{th} integer in the sequence, and \circ is the word concatenation operator. Any such σ is called a *solution* of the PCP problem. Any sequence σ (regardless of whether it is a solution) determines a word obtained by concatenating the corresponding *w*-words, and one obtained by concatenating the corresponding *v*-words.

We construct a monadic, linear (recursive) Datalog program \mathcal{P} , the singleton set \mathcal{C} comprising a key constraint, and a query Q such that the PCP problem has a solution iff $\operatorname{ExpR}_{\mathcal{P}}^{\mathcal{C}}(Q)$.

We use only one EDB relation e(X, l, Y), intended to denote a directed edge with source X, target Y and label l. The (boolean) query Q is the following, where all lower-case letters (e.g. l, r, a, b, c, d) are constants, and upper-case letters are variables:

$$\begin{array}{lll} Q() & : - & e(A,l,B), e(A,r,C), e(D,c,A), \\ & & e(D,a,D), e(D,b,D), e(D,d,D). \end{array}$$

The program \mathcal{P} is constructed as follows (again, lower-case letters are constants and upper-case letters are variables). \mathcal{P} consists of

- the rule V() :- C(X);
- $\bullet\,$ the rule

$$C_{r}(X) := e(X, d, Y),$$

$$e(X, c, X'), e(X', l, Z),$$

$$e(Y, c, Y'), e(Y', r, T),$$

$$e(U, a, U), e(U, b, U), e(U, d, U),$$

$$e(U, c, X');$$

• for every $1 \leq i \leq n$, assuming w.l.o.g. that $v_i = \alpha_1^i \dots \alpha_{k_i}^i$ and $w_i = \beta_1^i \dots \beta_{l_i}^i$, the rules

$$C(X) := e(X, \alpha_{1}^{i}, X_{1}), \dots, e(X_{k_{i}-1}, \alpha_{k_{i}}^{i}, X_{k_{i}}),$$

$$e(X, \beta_{1}^{i}, Y_{1}), \dots, e(Y_{l_{i}-1}, \beta_{l_{i}}^{i}, Y_{l_{i}}),$$

$$e(X_{k_{i}}, d, Y_{l_{i}}), C_{r}(X_{k_{i}});$$

$$C_{r}(X) := e(X, d, Y),$$

$$e(X, \alpha_{1}^{i}, X_{1}), \dots, e(X_{k_{i}-1}, \alpha_{k_{i}}^{i}, X_{k_{i}}),$$

$$e(Y, \beta_{1}^{i}, Y_{1}), \dots, e(Y_{l_{i}-1}, \beta_{l_{i}}^{i}, Y_{l_{i}}),$$

$$e(X_{k_{i}}, d, Y_{l_{i}}), C_{r}(X_{k_{i}});$$

 \mathcal{C} comprises just one key constraint, stating that the source and label of an edge determine its target:

$$\forall X, L, Y, Y' \ e(X, L, Y) \land e(X, L, Y') \to Y = Y'.$$

 \mathcal{P} is designed to generate, for every sequence σ of integers from $\{1, \ldots, n\}$, an expansion which encodes the two concatenations of *v*-words and *w*-words determined by σ . A word is encoded by a chain of edges, each edge label encoding a character in the word. The expansion thus contains two chains of words (one for the v_i 's, one for the w_i 's), each of them ended by a c-labeled edge followed by an 1-edge, respectively an **r**-edge. The chains start from the same node (according to the *C* rule), and continue in parallel, chaining together pairs of subchains which correspond to pairs of words (v_i, w_i) for some *i* (this is the role of the repeated expansion is ended by a subgraph given by the expansion of the first rule of C_r , whose role will be explained shortly.

To enable mappings from the arbitrarily long chains of the expansions into the query, Q contains cycles into which every pair of chains can map. Indeed, it is easy to see that any expansion of \mathcal{P} has a containment mapping into Q. Since the cycles in Q cannot map into the straight chains in the

expansions of \mathcal{P} , the *v*-chain is ended by the cycles generated by the first rule of C_r .

We therefore have that $\text{EXPR}^{\mathcal{P}}_{\mathcal{P}}(Q)$ holds iff \mathcal{P} expresses some view V such that $V \sqsubseteq_{\mathcal{C}} Q$ (since the opposite containment holds for every expansion, even in the absence of constraints). Because \mathcal{C} contains only a key constraint, the chase with it is guaranteed to terminate, and $V \sqsubseteq_{\mathcal{C}} Q$ holds iff $chase_{\mathcal{C}}(V) \sqsubseteq Q$ [1].

Observe that successive expansions of the C_r IDB chain only the v-words together; the v_i -words in the expansion of each rule start from variable X which is also the end of the previous v_j -word in the concatenation, but the w_i words start from the fresh variable Y which is not connected to the end Y_{k_j} of the previous w_j -word. Connecting the successive w-words explicitly would require IDB C_r to be binary, carrying both ends of v and w-words. To use only monadic rules, we rely instead on the key constraint: the variable beginning any w-word and the variable ending the previous w-word in the chain are both targets of d-edges emanating from the junction of the previous and current vword. The chase with the key constraint will "glue" the two chain segments corresponding to the w-words.

The intuition behind the construction is that, if we log for each one-step expansion of IDB C_r the *i* corresponding to the rule used, the obtained sequence of integers is the candidate for the solution of the PCP problem. All possible sequences of one-step expansions thus generate all possible solution candidates.

The theorem follows from the following claim, stating that a candidate solution is verified as a true solution only by finding a containment mapping from Q into the chase result of the corresponding expansion:

Claim. There is some view V generated by \mathcal{P} such that

$$chase_{\mathcal{C}}(V) \sqsubseteq Q$$

iff V encodes a solution of the PCP problem. \diamond

Proof. Notice that the chase of any expansion E with the key constraint can only start at the common origin of the v- and w-chains, and can only continue as long as the labels in the chains situated at the same distance from the origin coincide. The chains are determined by a solution to the PCP problem if and only if they match on their entire length, which is equivalent to the chase proceeding to collapse the chains all the way to their ends. This is detected by the fact that the 1- and the **r**-edges eventually share the same source, which in turn is the only way in which the query pattern can map into the chase result of E. End of proof of claim.

Proof: (Theorem 6.3) We use a reduction from PCP, adapting the construction from the proof of Theorem 6.2. The main difficulty here is to control that the two chains of v- and w-words are determined by the same sequence of integers, and that the chains match each other in length and labels. This was achieved in the proof of Theorem 6.2 by chasing with the key constraint.

We introduce fresh edge labels, $1, \ldots, n$, for n being the number of PCP words. We also use the labels *sync*, *end*, *up*, *down*.

We construct a monadic, (recursive) Datalog program \mathcal{P} , the set \mathcal{C} comprising three families of TGDs, and a query Qsuch that the PCP problem has a solution iff $\text{ExpR}_{\mathcal{P}}^{\mathcal{C}}(Q)$. The Datalog program \mathcal{P} contains:

- a rule for the distinguished IDB predicate ans: ans() : C(X);
- for every $1 \leq i \leq n$, assuming w.l.o.g. that $v_i = \alpha_1^i \dots \alpha_{k_i}^i$ and $w_i = \beta_1^i \dots \beta_{l_i}^i$, the rules

$$C(X) := e(X, sync, X), e(X, \alpha_1^i, X_1), \dots, e(X_{k_i-1}, \alpha_{k_i}^i, X_{k_i}), e(X, \beta_1^i, Y_1), \dots, e(Y_{l_i-1}, \beta_{l_i}^i, Y_{l_i}), e(X, i, X_{k_i}), e(X, i, Y_{l_i}), C_v(X_{k_i}), C_w(Y_{l_i});$$

$$C_{v}(X) := e(X, \alpha_{1}^{i}, X_{1}), \dots, e(X_{k_{i}-1}, \alpha_{k_{i}}^{i}, X_{k_{i}}), \\ e(X, i, X_{k_{i}}), C_{v}(X_{k_{i}});$$

$$C_w(X) := e(X, \beta_1^i, Y_1), \dots, e(Y_{l_i-1}^{\prime}, \beta_{l_i}^i, Y_{l_i}), \\ e(X, i, Y_{l_i}), C_w(Y_{l_i});$$

• the rules

$$C_{v}(X) := e(X, end, Y), e(Y, up, Z)$$

$$C_{w}(X) := e(X, end, Y), e(Y, down, Z)$$

$$e(X, a, X), e(X, b, X),$$

$$e(X, 1, X), \dots, e(X, n, X),$$

$$e(X, sync, X)$$

The program expands into chains that are not necessarily synchronized. We control synchronization by constraints. More precisely, we use TGDs to control that the two chains are determined by the same sequence of integers, and to control that the two chains match. C comprises:

• for each $1 \le i \le n$, the full TGD

$$\forall X, Y, Z, T$$

$$e(X, sync, Y) \land e(X, i, Z) \land e(Y, i, T) \rightarrow e(Z, sync, T)$$

$$(20)$$

• for each $l, l' \in \{a, b\}$, the full TGD

$$\forall X, Y, Z, T$$

$$e(X, l, Y), e(X, l, Z), e(Y, l', T) \rightarrow e(Z, l', T)$$

$$(21)$$

• for each $l \in \{a, b\}$, the full TGD

$$\begin{aligned} \forall X, Y, Z, T, U, V, W \qquad (22)\\ e(X, l, Y), e(X, l, Z), e(Z, end, T),\\ e(Y, end, V), e(T, up, U), e(V, down, W),\\ e(Y, sync, Z) &\rightarrow e(V, up, U) \end{aligned}$$

Intuitively, an expansion has an *end*-edge to signal the end of each chain, then an *up*-edge to signal the end of the chain of *v*-words, and a *down*-edge to signal the end of the chain of *w*-words.

The sync edges are added by the chase of the expansion with the family of TGDs (20), to mark the pairs of nodes on the two chains which represent chain prefixes determined by the same sequence of integers from $\{1, \ldots, n\}$.

Since the two chains of the expansion have a common origin (due to the expansion of IDB C), the chase with the

family of TGDs (21) can only start at the origin, and continues down the chains only as far as the labels of the chain prefixes match. The two chains match entirely if and only if the chase with (21) stops at the chain ends (marked by *end*-edges).

If the chase with both families of TGDs (20) and (21) goes all the way to the end of the two chains, then both the sequence of integers and the sequence of labels coincide, hence the chains encode a PCP solution. This is detected by the family of TGDs (22), which apply only in that case, recording this fact by copying the up-edge from the end of the v-chain to the end of the w-chain, thus creating a node with both up and down edges emanating from it.

This is precisely what the query checks for:

Similar to proof of Theorem 6.2, in order to enable mappings from the arbitrarily long chains of the expansions into the query, Q contains cycles into which every pair of chains can map. Indeed, it is easy to see that any expansion of \mathcal{P} has a containment mapping into Q. Since the cycles in Q cannot map into the straight chains in the expansions of \mathcal{P} , the w-chain is ended by the cycles generated by the second rule of C_w .

It is easy to verify that Q can be mapped into the result of chasing some expansion of \mathcal{P} with \mathcal{C} iff the expansion encodes a PCP solution.

E. INTER-REDUCIBILITY PRESERVES DE-CIDABILITY RESTRICTIONS

PROPOSITION E.1. Let Q be a conjunctive query, C a weakly acyclic set of dependencies, and \mathcal{P} a C-local Datalog program. Let Q', C' and \mathcal{P}' be obtained, in PTIME, as in the reduction used in the proof of Theorem 3.1 such that we have $\operatorname{SUPP}_{\mathcal{P}}^{\mathcal{P}}(Q)$ iff $\operatorname{ExpR}_{\mathcal{P}'}^{\mathcal{P}'}(Q')$. Then (i) \mathcal{P}' is C'-local and (ii) if any finite set of views generated by \mathcal{P} is C-independent, then the views generated by \mathcal{P}' are C'-independent.

Please also note that, as in Theorem 4.3, if C is a weakly acyclic set of inclusion dependencies, then so is C', hence \mathcal{P}' is also C'-local and the views it expresses are C'-independent.

Proof: In the following, for a query Q, we will denote by $body_Q$ the conjunction of atoms in the body of Q.

(i) Any view V' generated by \mathcal{P}' is of the form

$$ans'(Z_1, \dots, Z_m) := equals(Z_1, X_{i1}), \dots, equals(Z_m, X_{im}), \\ D(X_1), \dots, D(X_m), \\ body_{V_1}(Y_1^{(1)}, \dots, Y_{k_1}^{(1)}, U_1^{(1)}, \dots, U_{l_1}^{(1)}), \\ equals(Y_1^{(1)}, X_{j1}^{(1)}), \dots, equals(Y_{k_1}^{(1)}, X_{jk_1}^{(1)}), \\ D(Y_1^{(1)}), \dots, D(Y_{k_1}^{(1)}), D(U_1^{(1)}), \dots, D(U_{l_1}^{(1)}), \\ \dots body_{V_n}(Y_1^{(n)}, \dots, Y_{k_n}^{(n)}, U_1^{(n)}, \dots, U_{l_n}^{(n)}), \\ equals(Y_1^{(n)}, X_{j1}^{(n)}), \dots, equals(Y_{k_n}^{(n)}, X_{jk_n}^{(n)}), \\ D(Y_1^{(n)}), \dots, D(Y_{k_n}^{(n)}), D(U_1^{(n)}), \dots, D(U_{l_n}^{(n)}), \\ \end{array}$$

where $V_i(Y_1^{(i)}, \ldots, Y_{k_i}^{(i)})$ are views generated by \mathcal{P} . Let us write it shortly

$$ans'(\bar{Z}) := extra(\bar{Z}, \bar{X}, \bar{Y}, \bar{U}), body_{V_1}(\bar{Y}_1, \bar{U}_1), \dots body_{V_n}(\bar{Y}_n, \bar{U}_n)$$

where \bar{Y} and \bar{U} are the union of all \bar{Y}_i and all \bar{U}_i variables, respectively. Since there are no D atoms in C', we have that $chase_{C'}(V')$ is obtained by chasing only $body_{V_i}$, i.e.

$$extra(\bar{Z}, \bar{X}, \bar{Y}, \bar{U}), chase_{\mathcal{C}'}(body_{V_1}(\bar{Y}_1, \bar{U}_1)), \dots, chase_{\mathcal{C}'}(body_{V_n}(\bar{Y}_n, \bar{U}_n))$$

which is of the form

$$extra(\bar{Z}, \bar{X}, \bar{Y}, \bar{U}),$$

$$chase_{\mathcal{C}}(body_{V_{1}}(\bar{Y}_{1}, \bar{U}_{1})), D(F_{1}^{(1)}), \dots D(F_{p_{1}}^{(1)}), \dots$$

$$chase_{\mathcal{C}}(body_{V_{n}}(\bar{Y}_{1}, \bar{U}_{1})), D(F_{1}^{(n)}), \dots D(F_{p_{n}}^{(n)})$$

for some sets of variables $\bar{F}^{(i)} \subset \bar{U}_i \cup \bar{Y}_i$.

Since \mathcal{P} is \mathcal{C} -local, for each V_i , there is a view W_i generated by $chase_{\mathcal{C}}(\mathcal{P})$ such that $chase_{\mathcal{C}}(V_i) \equiv W_i$. If we denote

$$w'_i = body_{W_i}(\bar{Y}_i, \bar{U}_i), D(F_1^{(i)}), \dots D(F_{p_i}^{(i)})$$

we have that

$$chase_{\mathcal{C}'}(V') \equiv extra(\bar{Z}, \bar{X}, \bar{Y}, \bar{U}), w'_1(\bar{Y}_1, \bar{U}_1), \dots$$
$$w'_n(\bar{Y}_n, \bar{U}_n).$$

Hence $chase_{\mathcal{C}'}(V')$ is equivalent to a view W' generated by $chase_{\mathcal{C}'}(\mathcal{P}')$ because chase steps only apply on the rules of \mathcal{P}' obtained from the rules of \mathcal{P} (by adding D atoms) and all the w'_i can be obtained by chasing the rules inherited from \mathcal{P} .

For the converse, we consider a view W' generated by $chase_{\mathcal{C}'}(\mathcal{P}')$. Using the same observation, W' has the subgoals from the bodies of n views W_1, \ldots, W_n generated by $chase_{\mathcal{C}}(\mathcal{P})$ (for some $n \geq 1$) plus the ones in the conjunction extra defined above and some other D atoms introduced by the chase. As \mathcal{P} is \mathcal{C} -local, for each W_i there is a view V_i generated by \mathcal{P} such that $chase_{\mathcal{C}}(V_i) \equiv W_i$. Reasoning in the same manner as above, we can put all the V_i views together and obtain a view V' generated by $chase_{\mathcal{C}'}(\mathcal{P}')$ such that $chase_{\mathcal{C}'}(\mathcal{P}')$ such that $chase_{\mathcal{C}'}(\mathcal{P}')$ such that $chase_{\mathcal{C}'}(\mathcal{P}') \equiv W'$.

(ii) Consider *n* views V'_1, \ldots, V'_n expressed by \mathcal{P}' . Each V'_i is of the form

$$V'_{i}(\bar{Y}_{i}) := V_{1}^{(i)}(\bar{Y}_{1}^{(i)}), \dots, V_{k_{i}}^{(i)}(\bar{Y}_{k_{i}}^{(i)}), extra_{i}(\bar{Z}_{i}, \bar{X}_{i}, \bar{Y}_{i}, \bar{U}_{i})$$

where $\bigcup_{j=1}^{k_i} \bar{Y}_{k_i}^{(i)} = \bar{Y}_i$ and $extra_i$ is a conjunction of Dand equals predicates, similar to extra from (i). Let extra be the conjunction of all the $extra_i$ and R be a query over $\{V'_1, \ldots, V'_n\}$. Please note that the chase with \mathcal{C}' of $expand_{\{V'_1, \ldots, V'_n\}}(R)$ will only apply to the bodies of the $V_j^{(i)}$ views generated by \mathcal{P} . It follows that

$$chase_{\mathcal{C}'}(expand_{\{V'_1,\ldots,V'_n\}}(R))$$

is equivalent to a query having a body of the form

$$extra(\bar{Z}, \bar{X}, \bar{Y}, \bar{U}), chase_{\mathcal{C}'}\left(expand_{\{V_1, \dots, V_n\}}(conj_{\mathcal{P}}(\bar{Y}, \bar{U}))\right),$$

where *extra* is a conjunction of *extra*_i subqueries and $conj_{\mathcal{P}}$ is a conjunction of views V_i generated by \mathcal{P} .

Since \mathcal{C}' is obtained from \mathcal{C} by adding D atoms in the conclusions, the latter conjunction is equivalent to:

where the $D(F_{ijk})$ subgoals are added by the conclusions of dependencies from \mathcal{C}' .

But we assumed the views generated by \mathcal{P} to be \mathcal{C} -independent. Hence there is a query $T \equiv_{\mathcal{C}}^{\mathcal{V}} conj_{\mathcal{P}}$ (T outputs all the variables of $conj_{\mathcal{P}}$) such that

$$chase_{\mathcal{C}}(expand_{\{V_1,\dots,V_n\}}(T)) \equiv$$
$$expand_{\{chase_{\mathcal{C}}(V_1),\dots,chase_{\mathcal{C}}(V_n)\}}(T).$$
(23)

Let T' be the query obtained from T in the following way. We replace every view atom V_i (from \mathcal{V}) with a V'_j , (from \mathcal{V}') such that

- V'_j is constructed using only one expansion of the second rule (from program \mathcal{P}') for the *temp* IDB predicate, such that the variables \bar{X} of the *ans* predicate are the output variables of the $V_i(\bar{X})$ atom;
- notice also that the equalities between pairs of output variables of $V'_j(\bar{X})$ are already satisfied in T because those equalities are needed in order for $conj_{\mathcal{P}}$ to map into T.

From $T \equiv_{\mathcal{C}}^{\mathcal{V}} conj_{\mathcal{P}}$, it follows that

$$chase_{\mathcal{C}}(expand_{\mathcal{V}}(T)) \equiv chase_{\mathcal{C}}(expand_{\mathcal{V}}(conj_{\mathcal{P}})).$$

The mappings witnessing the latter equivalence can be extended to show that

$$chase_{\mathcal{C}'}(expand_{\mathcal{V}'}(T')) \equiv chase_{\mathcal{C}'}(expand_{\mathcal{V}'}(R)),$$

proving that $R \equiv_{\mathcal{C}'}^{\mathcal{V}} T'$. We can extend the mapping because chasing with \mathcal{C}' instead of \mathcal{C} only brings D goals that can be inferred using bodies of views from \mathcal{V} (there is no D goal in the premise of a rule from \mathcal{C}'). Hence $expand_{\mathcal{V}'}(T')$ and $expand_{\mathcal{V}'}(R)$ will behave the same way during the chase, since their subqueries based on \mathcal{V} views, $expand_{\mathcal{V}}(T)$ and $conj_{\mathcal{P}}$ respectively, are equivalent. The rest of the subgoals are D atoms coming from the expansions of the views from \mathcal{V}' (which are formed by bodies of views from \mathcal{V} plus Datoms).

From equivalence (23) we can also infer that

$$chase_{\mathcal{C}'}(expand_{\{V'_1,\dots,V'_n\}}(T')) \equiv \\ expand_{\{chase_{\mathcal{C}'}(V'_1),\dots,chase_{\mathcal{C}'}(V'_n)\}}(T')$$

because the D atoms, including those from the bodies of \mathcal{V}' views, are not involved in the chase and because chasing with \mathcal{C}' is only different from chasing with \mathcal{C} in that D atoms are added. But, by construction of \mathcal{C}' , the variables in these D atoms are variables that already existed in (23) hence the mappings witnessing (23) extend to the D atoms.

Since $R \equiv_{\mathcal{C}'}^{\mathcal{V}'} T'$, we can conclude that the views generated by \mathcal{P}' are \mathcal{C}' -independent.

PROPOSITION E.2. Let Q be a conjunctive query, C a weakly acyclic set of dependencies, and \mathcal{P} a C-local Datalog program. Let Q'', C'' and \mathcal{P}'' be obtained, in PTIME, as in the reduction used in the proof of Theorem 3.2 such that we have $\mathrm{EXPR}^{\mathcal{P}}_{\mathcal{P}}(Q)$ iff $\mathrm{SUPP}^{\mathcal{P}''}_{\mathcal{P}''}(Q'')$. Then (i) \mathcal{P}'' is \mathcal{C}'' -local and (ii) if every finite set of views generated by \mathcal{P} is C-independent, then the views generated by \mathcal{P}'' are \mathcal{C}'' -independent.

Proof: Notice that the constraints do not mention *head* predicate used in the definition of Q', hence the *head* atoms are not involved in the chase.

Let V'' be a view generated by \mathcal{P}'' of the form

$$\mathcal{V}''():=head(\bar{X}), body_{\mathcal{P}''}(\bar{X}, \bar{Y}, U).$$

One can see that if we replace all EDB predicates of $body_{\mathcal{P}''}$ with predicates that do not use the U variable, we obtain the unfolding of a view V generated by \mathcal{P} , whose body is a conjunction $body_{\mathcal{P}}(\bar{X}, \bar{Y})$.

Under the assumption that \mathcal{P} is \mathcal{C} -local, for every view V generated by \mathcal{P} , there is a view W generated by $chase_{\mathcal{C}}(\mathcal{P})$ such that $chase_{\mathcal{C}}(V) \equiv W$. We can prove by induction on the length of the chase sequence that there is a terminating chasing sequence for $body_{\mathcal{P}''}(\bar{X}, \bar{Y}, U)$ similar to the one for $body_{\mathcal{P}}(\bar{X}, \bar{Y})$ modulo the transformation of the EDB predicates. Hence there is also a view W'' produced by $chase_{\mathcal{C}''}(\mathcal{P}'')$ with $chase_{\mathcal{C}''}(V'') \equiv W''$.

To conclude (i), we can show that, conversely, for each view W'' generated by $chase_{\mathcal{C}'}(\mathcal{P}'')$ there is a corresponding W generated by $chase_{\mathcal{C}}(\mathcal{P})$ and a V generated by \mathcal{P} such that $W \equiv chase_{\mathcal{C}}(V)$ implies $W'' \equiv chase_{\mathcal{C}''}(V'')$, where body of V'' is formed by body of V and a *head* atom. For that, we can use the same argument, namely the isomorphism between the two chase sequences.

To prove (ii), let R'' be a query formulated in terms of views \mathcal{V}'' . Let R' be a query in terms of \mathcal{V} , obtained from R'' by removing the *head* predicates and the replacing the subgoals with corresponding subgoals on the original schema, by removing the U variable. Since the views produced by \mathcal{P} are \mathcal{C} -independent, there is a query $R \equiv_{\mathcal{V}}^{\mathcal{C}} R'$ such that $chase_{\mathcal{C}}(expand_{\mathcal{V}}(R)) \equiv expand_{\{chase_{\mathcal{C}}(V_1),\ldots\}}(R)$. Let then T be the query obtained from R by introducing back all the *head* atoms that were removed and by replacing the other subgoals with predicates that have one more variable, U. By extending the mappings that witness $R \equiv_{\mathcal{V}}^{\mathcal{C}} R'$ to predicates with the arity increased by one and to the *head* subgoals, we can also show that $T \equiv_{\mathcal{V}}^{\mathcal{C}} R''$. We conclude by noticing that $chase_{\mathcal{C}''}(expand_{\mathcal{V}''}(T)) \equiv expand_{\{chase_{\mathcal{C}''}(V_1''),\ldots\}}(T)$ follows from the similar property satisfied by R.

F. INTERCHANGEABILITY IS UNHELPFUL UNDER DEPENDENCIES

The following example shows that under dependencies, there are infinitely many equivalence classes of views with respect to interchangeability. This precludes the reduction described in [15] from the problem of support to that of rewriting using finitely many views, as it involves focusing on representatives of the equivalence classes.

EXAMPLE F.1. We have a program \mathcal{P} that produces unary views as follows:

$$V(X) := e(X, a, Y), C_r(Y)$$

$$C_r(X) := e(X, a, Y), C_r(Y)$$

$$C_r(X) := e(X, b, Y), e(Y', b, Y), e(Y', a, Y'),$$

$$e(Y, up, Z)$$

$$C_r(X) := e(X, b, Y), e(Y', b, Y), e(Y', a, Y'),$$

$$e(Y, down, Z)$$

Expansions are chains of a-labeled edges ending with a blabeled edge and one of up or down. Consider the query Q:

$$Q() \quad :- \quad e(D, a, D), e(D, b, A),$$
$$e(A, up, B), e(A, down, C)$$

The source obeys also one key constraint for each $l \in \{a, b\}$:

$$\forall X, Y', Y'' \ e(X, l, Y'), e(X, l, Y'') \longrightarrow Y' = Y''.$$

We write V_n for the expansion with n a-labeled edges and ending with up. We write U_n for the expansion with n alabeled edges and ending with down.

We can see that, for any n, the rewriting R_n defined as

$$R_n():-V_n(X), U_n(X)$$

is an equivalent rewriting of Q.

However, replacing in R_n the V_n goal with any other view $(V_i \text{ or } U_i)$ would not yield another equivalent rewriting. So each V_n (and each U_n) is in its own equivalence class w.r.t. interchangeability in rewritings for Q. There are therefore infinitely many such equivalence classes.