On monotonic determiancy and rewritability
for recursive queries and views

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ABSTRACT
A query \( Q \) is monotonically determined over a set of views \( V \) if \( Q \) can be expressed as a monotonic function of the view image. In the case of relational algebra views and queries, monotonic determinacy coincides with rewritability as a union of conjunctive queries, and it is decidable in important special cases, such as for CQ views and queries [9, 23]. We investigate the situation for views and queries in the recursive query language Datalog. We give both positive and negative results about the ability to decide monotonic determinacy, and also about the co-incidence of monotonic determinacy with Datalog rewritability.

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1 INTRODUCTION
View definitions allow complex queries to be represented by simple relation symbols. They have many uses, including as a means to protect access to data, as a means to raise the level of abstraction available to data users, and as a means to speed up the evaluation of queries [2]. Views represent a restricted interface to a dataset, and thus an associated question is what class of queries can be answered via accessing this interface. More formally, given a query \( Q \) expressed as a logical formula over the base relations, can the answer to be obtained via accessing the views. There are several different formulations of this computational problem, depending on what one means by “answering a query accessing the views”. One can ask whether \( Q \) is expressible as an arbitrary function of the views, or as an arbitrary monotone function of the views. Alternatively, one can choose a particular query language \( L \) and ask whether \( Q \) can be transformed to a query \( Q' \) over the views, where \( Q' \) is in \( L \). The first choice is that \( Q \) is determined over the views, the second that \( Q \) is monotonically determined over the views, and the last that \( Q \) is \( L \)-rewritable over the views. Each of these notions can be relativized to finite instances.

These questions were studied initially in the case where both queries and views are given by conjunctive queries (CQs). It is known that:

- determinacy of CQ query over a collection of CQ views is equivalent to rewritability of \( Q \) over the views in relational algebra [23]
- determinacy of a CQ over CQ views does not agree neither with determinacy over finite instances [17] nor with monotone determinacy [3]
- determinacy of a CQ query over CQ views is undecidable [17], and the same holds for determinacy over finite instances [18]
- determinacy is decidable for queries and views given as path-CQs [3]
- monotonic determinacy of a CQ query over CQ views implies rewritability of \( Q \) as a CQ [9], agrees with monotonic determinacy over finite instances and is NP-complete to decide [21]

These results have been generalized to the case of queries and views built up with more general constructs of active-domain first-order logic (or equivalently, in relational algebra). Then monotonic determinacy becomes, like determinacy, undecidable, and monotonic determinacy, like determinacy, disagrees with its variant over finite instances. But there is still a relationship between determinacy/monotonic determinacy and rewritability in a logic: determinacy is the same as rewritability in first-order logic; monotonic determinacy is the same as rewritability.

Less is known where queries and views are recursive, for example, when views and queries are in the common recursive query language Datalog. For specialized recursive queries and views over a graph schema, the regular path queries, both the determinacy and monotonic determinacy problem have been studied. For one- and two-way regular path queries and views monotonic determinacy (aka “losslessness with respect to the sound view assumption”) is decidable in ExpSpace ([10] for 1-way,[11] for 2-way), and implies Datalog rewritability [15], while plain determinacy is undecidable [16]. It follows from [14] that monotonic determinacy is undecidable for Datalog queries and CQ views and implies rewritability in Datalog over views.

The status of these questions for more general recursive queries — e.g., queries and views in Datalog over higher-arity relations — is to the best of our knowledge unknown.

EXAMPLE 1. Consider a schema with a ternary relation \( T \), and binary relation \( B \) and unary relations \( U_1, U_2 \). Consider the Boolean Datalog query \( Q \) given as:

\[
\text{GOAL}_Q \leftarrow U_1(x), W_1(x)
\]

\[
W_1(x) \leftarrow T(x, y, z), B(z, w), B(y, w), W_1(w)
\]

\[
W_1(x) \leftarrow U_2(x)
\]
Consider the following CQ views:

\[ V_0(x, w) := T(x, y, z), B(z, w), B(y, w) \]
\[ V_1(x) := U_1(x) \quad \forall x \]
\[ V_2(x) := U_2(x) \quad \forall x \]
\[ V_3(y, z) := U_3(x), T(x, y, z) \]

and the binary Datalog view \( V_4 \):

\[ \text{Goal}_{V_4}(y, z) \leftarrow T(x, y, z), B(z, w), B(y, w) \]
\[ \text{Goal}_{V_4}(q, r) \leftarrow \text{Goal}_{V_4}(q, r) \]
\[ \text{Goal}_{V_4}(y, z) \leftarrow B(y, w), B(z, w), U_2(w) \]

We can see that \( Q \) is monotonically determined over the views \( V_0-V_2 \). In fact there is a Datalog rewriting, obtained from \( Q \) by first replacing the second rule by \( W_1(x) \leftarrow V_0(x, w), W_1(w) \) and then replacing each \( U_i \) by \( V_i \) in the other rules. Further \( Q \) is monotonically determined using views \( V_3 - V_4 \), since it can be rewritten as the CQ \( \exists y z V_3(y, z) \land V_4(y, z) \).

Note that query \( Q \) is not contained in any of the classes considered in past work (e.g. regular path queries).

Our results. We give results on the complexity of deciding monotonic determinacy and on the ability to rewrite monotonically-determined queries into suitable languages, for views and queries expressed in Datalog or in sublanguages such as Monadic Datalog (MDL), or frontier-guarded Datalog (FGDL).

We provide new positive results about rewratability, showing monotonic determinacy implies \( L \)-rewritability for some natural query languages \( L \). We show that monotonic determinacy implies rewratability in Datalog for Datalog queries and FGDL views (Theorem 1), as well as for MDL queries and a collection of FGDL and CQ views (Theorem 2). We observe that for CQ \( Q \) and Datalog \( V \), monotonic determinacy implies rewratability as a CQ, and the same holds if CQ is replaced with UCQ. Note that an analysis of the “inverse rules” algorithm [14] implies that FGDL queries monotonically determined over CQ views have Datalog rewritings. On the negative side, we show that MDL queries monotonically determined over CQ views are not necessarily rewratable in MDL (Theorem 7). This contrasts with the observation from [14] mentioned above. In contrast to Theorem 2, we give an example of an MDL query monotonically determined over UCQ views without a Datalog rewriting (Theorem 8). Our results on rewratability are summarized in Figure 1 where “nn” stands for “not necessarily”.

We now turn to results about deciding monotonic determinacy. We show that monotonic determinacy is

- decidable in \( 2\text{ExpTime} \) for CQ queries and Datalog views (Theorem 5),
- decidable in \( 2\text{ExpTime} \) for queries and views in frontier-guarded Datalog (Theorem 3),
- decidable in \( 3\text{ExpTime} \) for MDL queries and a collection of MDL and CQ views (Theorem 4),
- \( 2\text{ExpTime} \)-hard for CQ queries and MDL queries and for MDL queries and CQ views (Proposition 9)
- undecidable for MDL queries and UCQ views (Theorem 6)

Known and new results on decidability of monotonic determinacy are presented in Figure 2 where we use [upper bound]/[lower bound] notation for sources.

Alongside with \( L \)-rewritability we can ask whether there are computable functions lying within a certain complexity class which separate the images of instances where \( Q \) is true from images of those where \( Q \) is false. We call such a function a separator for \( Q \) over \( V \). Note that Datalog rewritings give rise to \( \text{PTIME} \) separators, while UCQ-rewritings produce \( \text{AC}^0 \) separators. Our additional observations on separators, outside of those that follow from rewratability results, are: (1) for Datalog queries and UCQ views there is always a separator in \( \text{NP} \) as well as one in \( \text{co-NP} \); (2) for any primitive recursive function \( f \) there are Datalog queries monotonically determined over Datalog views without a separator in \( \text{TIME} (f(x)) \) (Theorem 9).

Techniques. A contribution of the paper is to show how techniques arising from earlier work can be adapted for the analysis of monotone determinacy. For our positive results, a key tool is an automata-theoretic technique, involving bounds on the treewidth of view images and the forward-backward method developed for analysis of guarded logics [8, 19]. For our negative results, we show how to adapt some of the coding ideas used in showing undecidability of determinacy [16–18] to the setting of monotonic determinacy, and we also show how tools from constraint satisfaction [4] can be used to provide monotonically-determined queries that have no Datalog rewriting.

Organization. Section 2 contains preliminaries about Datalog and monotonic determinacy, while Section 3 presents key tools that we make use of in our positive results. Section 4 presents our rewratability results, while Section 5 gives results on deciding monotonic determinacy. Section 6 contains lower bounds on detecting monotonic determinacy, while Section 7 provides non-rewritability results. The paper ends with conclusions and some open questions in Section 8. The details of many proofs are deferred to the appendix.

## 2 PRELIMINARIES

We will work with relational schemas, consisting of a finite set of relations, with each relation \( R \) associated with a number the arity of \( R \). For \( R \) of arity \( n \), an \( R \)-fact is an expression \( R(c_1, \ldots, c_n) \), where \( c_1, \ldots, c_n \) are elements. A fact over schema \( S \) is an \( R \) fact for some relation \( R \) of \( S \).

A database instance (or simply instance) is a set of facts over a schema. A query is a set of facts over a schema. The active domain of an instance \( I \), denoted \( \text{dom}(I) \), is the set of elements that occur as \( c_i \) in some fact \( R(c_1, \ldots, c_n) \) of \( I \). A query of arity \( n \) over schema \( S \) is a function from instances of \( S \) to relations of arity \( n \). A Boolean query is a query of arity 0. The output of a query \( Q \) on instance \( I \) is denoted \( \text{Output}(Q, I) \). We will also write \( I \models Q(e) \) or \( I, e \models Q \) to indicate that \( e \) is in the output of \( Q \) on input \( I \).

A homomorphism from instance \( I \) to instance \( I' \) is a mapping \( h \) such that \( R(c_1, \ldots, c_n) \in I \) implies \( R(h(c_1), \ldots, h(c_n)) \in I' \). If there is a homomorphism from \( I \) to \( I' \) then we write \( I \rightarrow I' \).

The Gaifman graph of an instance \( I \) is the graph whose nodes are the elements of \( \text{dom}(I) \) and whose edges connect any \( c_i \) and \( c_j \) in \( e \) such that \( R(e) \) holds. The radius of a graph \( G \) is defined as \( \min_{u \in \text{vertices}(G)} \max_{v \in \text{vertices}(G)} \text{dist}_G(u, v) \) where \( \text{dist}_G(u, v) \) is the distance between \( u \) and \( v \) in \( G \).
Conjunctive queries and Datalog. A conjunctive query (CQ) is a logical formula of the form \( q(x) = \exists y \phi(x, y) \), where \( \phi(x, y) \) is a conjunction of atoms. Given any CQ \( Q \), its canonical database, denoted \( \text{CanonDB}(Q) \), is the instance formed by turning each atom \( R(x_1 \ldots x_n) \) into a fact \( R(c_{x_1} \ldots c_{x_n}) \), where for each variable or constant \( x \) in \( Q \) we have a constant \( c_x \). Each CQ \( Q \) with free variables ordered as \( x_1 \ldots x_n \) defines a query of arity \( n \) in the obvious way: a tuple \( t_1 \ldots t_n \) is in the output of \( Q \) on \( I \) if there is a homomorphism of \( \text{CanonDB}(Q) \) into \( I \) mapping each \( x_i \) to \( t_i \). The radius of a CQ is the radius of the Gaifman graph of its canonical database.

Datalog is a language for defining queries over a relational schema \( S \). Datalog rules are of the form:

\[
P(x) \leftarrow \phi(x)
\]

where \( P(x) \) is an atom over a relation \( P \) that is not in \( S \), \( \phi(x) \) is a conjunctive query and every variable in \( P(x) \) occurs in \( \phi \). The left side of the rule is the head, while the right side is the body of the rule. In a set of rules, the relation symbols that occur in the head of a rule are called the intensional relations of the rule. A Datalog program is a finite collection of rules. For a database instance \( I \) and a set of Datalog rules \( \Pi \) by FPEval(\( \Pi, I \)) we denote the minimal IDB-extension of \( I \) satisfying \( \Pi \). A Datalog query \( Q = (\Pi, \text{Goal}) \) is a Datalog program \( \Pi \) together with a distinguished intensional goal relation \( \text{Goal} \) of arity \( k \geq 0 \). The output of Datalog query \( Q \) on an instance \( I \) (denoted as \( \text{Output}(Q, I) \) or simply \( Q(I) \)) consists of all tuples \( e \) such that \( \text{Goal}(e) \in \text{FPEval}(\Pi, I) \).

For example, consider a signature where there is a binary relation \( R \) and unary relation \( U \). The formula expressing that \( x \) has a path consisting of \( R \) edges to an element in \( U \) would be written in Datalog as the following query \( \text{Conn}(x) = (\Pi, \text{Goal}(x)) \) where \( \Pi \) consists of the following rules:

\[
P(x) \leftarrow U(x)
\]
\[
P(x) \leftarrow R(x, y), P(x)
\]
\[
\text{Goal}(x) \leftarrow P(x)
\]

Above, \( P(x) \) and \( \text{Goal}(x) \) are intensional relations while \( R(x, y) \) and \( U(x) \) are extensional. We follow conventions concerning Datalog rules and omit the existential quantifiers on the variables in the body that do not appear in the head; we also use \( ", " \) for conjunction.

A Datalog query \( Q_1 \) is contained in a Datalog query \( Q_2 \) if \( \text{Output}(Q_1, I) \subseteq \text{Output}(Q_2, I) \) for every instance \( I \). Datalog containment is known to be undecidable in general [25].

Table 1: Rewritability of Queries Monotonically Determined by the Views

<table>
<thead>
<tr>
<th>Query \ Views</th>
<th>CQ</th>
<th>MDL, FGDL</th>
<th>FGDL + CQ</th>
<th>UCQ</th>
<th>Datalog</th>
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| CQ            | CQ [Prop. 8, (a)] | MDL [Th. 1] | Datalog, not necessarily MDL; for example, in an MDL program we can have a rule \( I_1(x) \leftarrow I_2(x) \), where \( I_1 \) and \( I_2 \) are both intensional. However every MDL program can be rewritten to be in FGDL, and thus we declare, as a convention, that any MDL program is Frontier-guarded. Frontier-Guarded Datalog containment is known to be decidable (e.g. [6]).

Table 2: Decidability and Complexity of Monotonic Determination

<table>
<thead>
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<td>-</td>
<td>2ExpTime-c</td>
</tr>
<tr>
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<td>-</td>
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where \( q(z) \) is in \( \text{CQAppr}(\Pi, V(z), k) \) for \( k \leq i \) and \( \sigma \) unifies \( V(z) \) with \( V(y) \) by sending \( z \) to \( y \).

A CQ approximation of a Datalog query \( (\Pi, \text{GOAL}(x)) \) is any element of \( \text{CQAppr}(\Pi, \text{GOAL}(x), i) \) for some \( i \).

**Proposition 1.** For any Datalog query \( Q \), if \( I, c \models Q \) then there is a CQ approximation \( Q_0 \) of \( Q \) such that \( I \models Q_0(c) \).

We often identify an approximation \( Q_0 \) of a Datalog query \( Q \) with its canonical database; for example, for another Datalog query \( Q' \), we can write \( \text{Output}(Q', Q_0) \) to indicate the output of \( Q' \) on \( \text{CANDIR}(Q_0) \). We can also talk about the approximation of an atom \( A \) in a Datalog program, which is defined by considering the program with \( A \) as the goal predicate.

Views, determinacy, and rewritability. A view over some relational schema \( S \) is a tuple \((V, Q_V)\) where \( V \) is a view relation and \( Q_V \) is an associated query over \( S \) whose arity matches that of \( V \). \( Q_V \) is referred to as the definition of view \( V \). By \( V \) we denote a collection of views over a schema \( S \). We sometimes refer to the vocabulary of the definitions \( Q_V \) as the base schema for \( V \), denoting it as \( 2^V \), while the predicates components \( V \) are referred to as the view schema, denoted \( \Sigma_V \). For an instance \( I \) and set of views \( V = \{ (V, Q_V) \mid V \in \Sigma_V \} \), the view image of \( I \), denoted by \( \text{Output}(Q, I) \), is the instance where each view predicate \( V \in \Sigma_V \) is interpreted by \( \text{Output}(Q, I) \). A query \( Q \) over schema \( S \) is determined over \( V \) if

\[
\text{for any two instances } I_1, I_2 \text{ such that } V(I_1) = V(I_2) \text{ we have } \text{Output}(Q, I_1) = \text{Output}(Q, I_2).
\]

A query \( Q \) over schema \( S \) is monotonically determined over \( V \) if

\[
\text{for any two instances } I_1, I_2 \text{ such that } V(I_1) \subseteq V(I_2) \text{ we have } \text{Output}(Q, I_1) \subseteq \text{Output}(Q, I_2).
\]

Given views \( V \) and a query \( Q \), a query \( R \) over the view schema \( \Sigma_V \) is a separator of \( Q \) with respect to \( V \) if: for each \( I \) over \( S \), the output of \( R \) on \( \text{Output}(Q, I) \) is the same as the output of \( Q \) on \( I \). A separator that can be specified in a particular language \( L \) (e.g. Datalog, CQs) is an L-rewriting of \( Q \) w.r.t. \( V \), and if this exists we say \( Q \) is L-rewritable over \( V \).

It is clear that if \( Q \) has a rewriting in a language that defines only monotone queries, like Datalog, then \( Q \) must be monotonically determined. We will be concerned with the converse to this question. The main questions we will consider, fixing languages \( L_Q \) and \( L_V \) for the queries and views (e.g. Datalog, fragments of Datalog) are:

- can we decide whether a \( Q \) in \( L_Q \) is monotonically determined over \( V \)?
- fixing another language \( L \) for rewritings, if \( Q \) is monotonically determined over \( V \), does it necessarily have a rewriting in \( L \)?

In this paper, for simplicity we will always consider the determinacy and rewritability problems restricting to the case when the query \( Q \) is Boolean. But all of our results extend to the non-Boolean case. In addition, we allow our instances to be finite or infinite, but all of the results extend when the instances are assumed to be finite. See the appendix for details.

### 3 Forward and Backward Between Datalog and Automata

We overview an automata-theoretic technique that will prove useful in rewriting results. It involves tree width bounds, along with the idea of combining forward mappings from Datalog to automata, projection of an automaton onto a subvocabulary, and backward mappings from an automaton to Datalog. The approach derives from work on guarded logics \([8, 19]\).

Treewidth and tree codes. For a number \( k \) a tree decomposition of width \( k \) for an instance \( I \) is a pair \( \mathcal{D} = (\tau, \lambda) \) consisting of a rooted directed tree \( \tau = (V, E) \) and a map \( \lambda \) associating a tuple of distinct elements \( \lambda(v) \) of length at most \( k \) (called a bag) to each vertex \( v \) in \( V \) such that the following conditions hold:

- for any atom \( R(c) \) in \( I \), there is a vertex \( v \in V \) with \( c \subseteq \lambda(v) \);
- for any element \( c \) in \( I \), the set \( \{ v \in V \mid \varepsilon \in \lambda(v) \} \) is connected in \( \tau \).

Above we abuse notation slightly by using \( \lambda(v) \) also to refer to the underlying set of elements as well as the tuple. Also in the literature the width associated to such a decomposition is \( k - 1 \), but this distinction will not be important for any of our results. Will also talk about a tree decomposition of width \( k \) for a pair \((I, a)\) consisting of an instance and a tuple. In this case we add to the requirements above that \( a \) is an initial segment of \( \lambda(r) \) for \( r \) the root of the tree.

The treewidth of an instance \( I \), \( \text{tw}(I) \), is the minimum width of a tree decomposition of \( I \). For a tree decomposition \( \mathcal{D} \) of data instance \( I \) let \( l(\mathcal{D}) \) be the maximum over elements \( e \) of \( I \) of the number of bags containing \( e \).

We will now discuss how to represent tree decompositions by labeled trees called codes. In this context, we will always assume that in tree decompositions, all vertices \( v \in V \) have outdegree at most 2. It is easy to show that if an instance has any tree decomposition of width \( k \), it has one with this property.

We represent such tree decompositions as instances in a signature \( \text{CODE}(S, k) \) which contains the following relations:

- for every relation \( R \in S \) of arity \( m \) and every sequence \( n = n_1, \ldots, n_m \) of numbers of size at most \( k \) there is a unary relation \( T^R_k \) in \( \text{CODE}(S, k) \) to mark the nodes \( v \) in \( \tau \) such that the atom \( R(b_{n_1}, \ldots, b_{n_m}) \) is in \( I \), where \( \lambda(v) = (b_{n_1}, \ldots, b_{n_k}) \).
- for every partial 1-1 map \( s \) from \( \{1, \ldots, k\} \) to \( \{1, \ldots, k\} \), there is a binary relation \( T_s \) to indicate the “same as” relation between positions in neighboring bags. For example, if \( (u, v) \in T_s \) and \( s(3) = 1 \), then the position 3 in \( u \) and the position 1 in \( v \) stand for the same element.

All relations \( T_s \) are directed from a parent to a child.

We use \( \text{UniPred}(S, k) \) and \( \text{BinPred}(S, k) \) to denote the sets of all unary and binary predicates in \( \text{CODE}(S, k) \) respectively. A tree over this signature will be referred to as a tree code of width \( k \) for \( S \).

It should be clear how each tree decomposition of \((I, a)\) of width \( k \) gives rise to a tree code of width \( k \) for \( S \); if there are bags with less than \( k \) elements, we fill them up with dummy elements to the length \( k \). We now show how to decode an instance from such a code \( \mathcal{T} \). For nodes \( u, v \) in a code \( \mathcal{T} \), we
write \((u, i) \equiv (v, j)\) if \((u, v) \in T_i\) holds in \(T\) and \(s(i) = j\). For a node \(u\) and position \(i\) we let \([u, i]\) be the equivalence class of \((u, i)\) in the equivalence relation generated by \(\equiv\). In words, the position \(i\) in the node \(u\) corresponds to the position \(j\) in the node \(v\) if there is an undirected path leading from \(u\) to \(v\) with the edge labels that in a step-by-step manner establish a match between \(i\) in \(u\) and \(j\) in \(v\). The decoding of \(T\), denoted \(I = \Sigma(T)\), is the database instance \(I\) consisting of atoms \(R[(v_1, i_1), \ldots, (v_r, i_r)]\) where each \(R\) from \(S\) is applied to exactly those tuples \([(v_1, i_1), \ldots, (v_r, i_r)]\) for which there is some node \(w \in \text{dom}(T)\) such that \(w \in T_{j_1 \ldots j_r}\) and \([w, j_m] = [v_m, i_m]\) for all \(m \in \{1, \ldots, r\}\). In this case we also say the \(T\) is a code of \(I\).

Monadic Datalog Normalisation. A Monadic Datalog query is said to be normalised if the body of any recursive rule does not contain IDB atoms with the head variable. A well-known simple fact is that any MDL query can be transformed into a normalized one.

**Proposition 2 ([12]).** For each MDL query \(Q\) there exists a normalized MDL query \(Q'\) which is equivalent to \(Q\).

Normalization is useful in connection with tree codes, since it is easy to see that the CQ approximations of normalized queries have decompositions with small "treespan":

**Lemma 1.** Let \(Q\) be a normalized Monadic Datalog query. Then there is a number \(k = O(|Q|)\) such that all CQ-approximations of \(Q\) have tree decomposition \(D\) of width \(k\) with \(l(D) \leq 2\).

Bounding the treewidth of view images. We will present results showing that, for certain classes of sets of views \(V\) and Datalog queries \(Q\), we can find a uniform bound on the treewidth of the \(V\)-image of the approximations of \(Q\).

It is easy to see that expanding an instance with the evaluation of all intensional predicates of a frontier-guarded program does not blow-up treewidth:

**Lemma 2.** If \(\Pi \in \text{FGDL}\) and \(I\) is an instance of treewidth \(k\), then \(\text{FPEval}(\Pi, I)\) is of treewidth \(k\).

A locality argument shows that applying connected CQ views preserves bounded treewidth:

**Lemma 3.** Let \(D\) be a tree decomposition of a data instance \(I\) of width \(k\) with \(l(D) \leq 2\). Let \(V\) be a set of connected CQ views, and \(V(I)\) the view image of \(I\) under \(V\). Let \(r\) be the greatest radius of a \(CQ\) in \(V\). Then the treewidth of \(V(I)\) is at most \(\frac{k(k+r-1)}{k-1}\).

Tree automata. We describe our variant of tree automata that accept binary trees \(T\) with edges labelled by binary relations from the set \(\text{BinPred}(S, k)\) and nodes labelled with unary predicates from a set \(\text{UnPred}(S, k)\). We consolidate node- and edge-labels by considering a tree alphabet \(\text{TreeAlph}\). \(\text{TreeAlph}\) contains labels for internal nodes \(\sigma_{L}^{s_1, s_2}\) indexed by sets of unary predicates \(L \subseteq \text{UnPred}(S, k)\) and pairs of binary predicates \(s_1, s_2 \in \text{BinPred}(S, k)\). It also contains leaf labels \(\sigma_L\) indexed by \(L \subseteq \text{UnPred}(S, k)\). We sometimes treat trees as terms over this alphabet: a tree with root labeled \(\sigma_{L}^{s_1, s_2}\) with children \(t_1\) and \(t_2\) would be written as \(\sigma_{L}^{s_1, s_2}(t_1, t_2)\).

A non-deterministic finite tree automaton (NTA) over \(\text{TreeAlph}\) is a tuple \(\mathcal{A} = (Q, Q_f, \Delta_0, \Delta_2)\), where

- \(Q\) is a finite set of states
- \(Q_f \subseteq Q\) is a set of final states,
- \(\Delta_0\) is a set of initial transitions of the form \(\sigma_L \rightarrow q\), and
- \(\Delta_2\) is a set of transitions of the form \(q_1, q_2, \sigma_{L}^{s_1, s_2} \rightarrow q\).

A run of \(\mathcal{A}\) on a tree \(T\) is a labelling function \(f : \text{Nodes}(T) \rightarrow Q\) satisfying the following: if \(t_0 = \sigma_{L}^{s_1, s_2}(t_1, t_2)\) for \(\sigma_{L}^{s_1, s_2} \in \text{TreeAlph}\), then \((f(t_1), f(t_2), \sigma_{L}^{s_1, s_2}) \rightarrow (f(t))\) \(\in \Delta_2\) and if \(f(v) = q\) for a leaf \(v\) of \(T\) with \(f(v) = \sigma_L\), then \(\sigma_L \rightarrow q \in \Delta_0\).

We say that \(T\) is accepted by \(\mathcal{A}\) if there is a run of \(\mathcal{A}\) on \(T\) that labels the root of \(T\) with a final state.

Forward from Datalog to NTA. We now show how to create a tree automaton accepting the views images of approximations of a given Datalog query. We say that a class \(\mathcal{C}\) of instances is \(k\) regular if the treewidth of instances in \(\mathcal{C}\) is at most \(k\), and there is an automaton \(\mathcal{A}\) such that

- for codes \(T\) of width \(k\), \(\mathcal{A}\) accepts \(T\) implies \(\Sigma(T) \in \mathcal{C}\).
- for each instance \(\mathcal{F} \in \mathcal{C}\) there is a code \(T\) such that \(\Sigma(T) = \mathcal{F}\) and \(\mathcal{A}\) accepts \(T\).

In this case we say \(\mathcal{A}\) captures \(\mathcal{C}\). If the stronger condition "for all codes \(T\), \(\mathcal{A}\) accepts \(T\) if \(\Sigma(T) \in \mathcal{C}\) holds, we say \(\mathcal{A}\) recognizes \(\mathcal{C}\).

The following simple "forward mapping" proposition shows that we can capture the approximations of Datalog queries with an automaton:

**Proposition 3.** For any Datalog query \(Q = (\Pi, \text{Goal})\), there is an \(\text{ExpTime}\) function that outputs an NTA \(\mathcal{A}_Q\) that captures the set of canonical databases of CQ approximations of \(Q\).

If we restrict to instances of a fixed treewidth, we can do better, obtaining an NTA that recognizes all trees that satisfy the Datalog program considered as a set of Horn clauses:

**Proposition 4.** For any Datalog program \(\Pi\), the class \(\{\mathcal{F} \mid \mathcal{F} \models \Pi, \text{tw} (\mathcal{F}) \leq k\}\) (here \(\mathcal{F}\) are finite instances which contain both EDBs and IDBs of \(\Pi\)) is \(k\) regular and is recognized by an NTA at most doubly-exponential sized in \(k\) and singly-exponential in \(|\Pi|\).

We also note that if we have captured a class of codes of instances with an automaton, we can project away some of the signature and still capture:

**Proposition 5.** If \(\mathcal{C}\) is a \(k\) regular class in \(\Sigma\) captured by NTA \(\mathcal{A}\) and \(\Sigma' \subseteq \Sigma\), then the class \(\mathcal{C} \cap \Sigma' = \{\mathcal{F} \mid \Sigma' \models \mathcal{F} \in \mathcal{C}\}\) is also \(k\) regular, captured by an automaton of size at most \(|\mathcal{A}|\). The same holds with "captured" replaced by "recognized".

Our next "forward mapping" result shows that we can recognize the set of codes of small treewidth which fail to satisfy clauses of a frontier-guarded program:

**Proposition 6.** For a FGDL query \(Q = (\Pi, \text{Goal})\) the set \(\{(I, a) \mid I \not\models Q(a), \text{tw}(I) \leq k\}\) is \(k\) regular and recognized by an NTA of size at most doubly-exponential in \(k\).

**Proof.** Follows from Propositions 4 and 5 since for a frontier-guarded \(Q = (\Pi, \text{Goal})\) we have, using Lemma 2,

\[\{(I, a) \mid I \not\models Q(a), \text{tw}(I) \leq k\} = \{(\mathcal{F} \mid \Sigma, a) \mid \mathcal{F} \models \Pi, \mathcal{F} \not\models \text{Goal}(a), \text{tw}(\mathcal{F}) \leq k\}\]
where $\Sigma$ is the signature of the EBDs in $\Pi$. □

In applying these results, we will sometimes use implicitly that if $C_1$ is captured by $\mathcal{A}$ and $C_2$ is recognized by $\mathcal{A}^*$, then $C_1 \cap C_2$ is captured by the product of $\mathcal{A}$ and $\mathcal{A}^*$. Note that, in contrast, classes of instances that are captured are not closed under intersection.

Homomorphic determinacy. A query $Q$ is said to be homomorphically determined by views $V$ if:

Whenever we have two instances $I_1$ and $I_2$ and a homomorphism $h$ from $V(I_1)$ to $V(I_2)$, then for each tuple $(c_1, \ldots, c_k) \in Q(I_1)$ we also have $(h(c_1), \ldots, h(c_k)) \in Q(I_2)$.

Note that if $Q$ is rewritable over $V$ in Datalog, or any other homomorphism-invariant query language, then $Q$ must be homomorphically determined by $V$.

Homomorphic determinacy of $Q$ over $V$ always implies monotonic determinacy of $Q$ over $V$; monotonic determinacy is simply the case where $h$ is the identity. Surprisingly, for Datalog queries and views the converse also holds:

**Lemma 4.** For any Datalog query $Q$ and Datalog views $V$, if $Q$ is monotonically determined over $V$ then it is homomorphically determined over $V$.

Backwards from NTAs to Datalog. Consider arbitrary NTA $\mathcal{A}$ that works on tree codes of width $k$. From $\mathcal{A}$ we construct a Datalog program. For every transition of the form $q_1, q_2, \sigma_L^{n_1, n_2} \rightarrow q$ with $L = \{T^{R_1}, \ldots, T^{R_m}\}$ we create a rule

$P_q(x_1, \ldots, x_k) \leftarrow \bigwedge_{i=1}^k \text{Adom}(x_i) \wedge P_{q_1}(x_1^{(1)}, \ldots, x_k^{(1)}) \wedge P_{q_2}(x_1^{(2)}, \ldots, x_k^{(2)})$ \wedge \bigwedge_{i \in \text{dom}(s_1)} x_i = x_1^{(1)} \wedge \bigwedge_{i \in \text{dom}(s_2)} x_i = x_2^{(2)} \wedge \bigwedge_{l=1}^m R^l(x_{n_l})$

where $j$ ranges over 1 and 2, $x_i^{(j)}$ are fresh variables for indices $j \in \{1, 2\}$ and $i \in \{1, \ldots, k\}$, and for $n_l = (n_1^{(1)}, \ldots, n_2^{(1)})$ we have $x_{n_l} = (x_{n_l}^{(1)}, \ldots, x_{n_l}^{(2)})$. For initial transitions of the form $\sigma_L \rightarrow q$ with $L = \{T^{R_1}, \ldots, T^{R_m}\}$ we have rules

$P_q(x_1, \ldots, x_k) \leftarrow \bigwedge_{i=1}^k \text{Adom}(x_i) \wedge \bigwedge_{l=1}^m R^l(x_{n_l})$.

For accepting states $q$ we add the rules $\text{GoalAdom}_q \leftarrow P_q(x_1, \ldots, x_k)$ for the goal predicate $\text{GoalAdom}_q$; recall that we are assuming here that the original query $Q$ is Boolean, so we are looking for a Boolean Datalog rewriting. We also add a standard set of rules which, when evaluated on any data instance $I$ under fixed-point semantics, guarantees that the interpretation of the IDB $\text{Adom}(x)$ is the active domain of $I$. Denote the resulting backward map Datalog query by $Q_{\Pi}$.

We now get to the main result of this section, which states that if we assume homomorphic determinacy and begin with an automaton representing view images of approximations of $Q$, then applying the backward mapping produces a Datalog rewriting of $Q$ over $V$. The proof is mostly a matter of working with the definitions. Homomorphic determinacy is used in the direction from right to left.

**Proposition 7.** Let $Q$ be homomorphically determined over $V$ and $\mathcal{A}$ be any automaton working on k-codes such that $V(Q_1) \cap \omega \subseteq \Sigma(L(\mathcal{A})) \subseteq \Sigma(D) \cap V(Q_1)$ maps into $D$ for some $i \in \omega$. More precisely, we require that:

1. For each CQ approximation $Q_1$ of $Q$, there is a Datalog query $Q$ such that $Q(V_1) \subseteq Q(V_2)$ and $Q$ is homomorphically determined by $Q$.

2. For each $Q$ accepted by $\mathcal{A}$, there is a CQ approximation $Q_1$ of $Q$ and a homomorphism from $V(Q_1)$ into $V(\mathcal{A})$.

Then for each data instance $I$ we have $I \models Q$ if $V(I) \models Q_q(a)$ for some $a \in \text{Adom}(I)^k$.

## 4 Rewritability

We are now ready to present our main results about rewritings of queries that are monotonically determined over views. The following result exhibits how the forward and backward mappings help us obtain Datalog rewritings.

**Theorem 1.** Suppose $Q$ is a Datalog query and $V$ is a collection of FGDL views. If $Q$ is monotonically determined over $V$, then $Q$ is rewritable over $V$ in Datalog. The size of the rewriting is at most double-exponential in $|Q|$ and exponential in $|V|$. If $Q$ is MDL such a rewriting exists in MDL as well.

**Proof.** Consider the class $\mathcal{C}$ of canonical databases of CQ approximations of $Q$. By Proposition 3, $\mathcal{C}$ is $k$-regular for some $k = O(|Q|)$ and is captured by an NTA $\mathcal{A}'$ of at most exponential size in $|Q|$. By Lemma 2, the treewidth of the class of view images of $C$ is also bounded by $k$. We claim that there is an automaton $\mathcal{A}$ that captures $V = \{F | tw(C) \leq k, V(C) \subseteq F \}$ for some $i \in \omega$ of size at most double-exponential in $|Q|$ and single-exponential in $|V|$ ("of required size" below) and argue that it satisfies the conditions of Proposition 7.

Without loss of generality we assume that the sets of IDBs of programs for different views are disjoint, and that their goal predicates are identical with the view predicates. Denote by $V$ the union of all rules in Datalog queries in $V$. Note that by definition, for any instance $I$, the restriction ofFP$\Pi_\mathcal{F}(\Pi_\mathcal{V}, I)$ on the view signature is exactly $V(I)$.

By Proposition 4, there is an NTA $\mathcal{A'}$ of required size which recognizes all codes of $(F | tw(F) \leq k)$. Therefore the class $\mathcal{F} = \{F \mid \mathcal{F} \cap \Sigma = \mathcal{F}, \mathcal{F} = \Pi_\mathcal{V}, tw(F) \leq k\}$ is captured by the intersection of $\mathcal{A}'$ and $\mathcal{A}$, which is also of required size. Observe that $V$ is the projection of $\mathcal{F}$ on the signature of view predicates and so $V$ is captured by some NTA $\mathcal{A}$ of required size by Proposition 5.

By Lemma 4, $Q$ is homomorphically determined over $V$. Since $V(Q_1) \cap \mathcal{A}$ is a CQ approximation of $Q$, the restriction of $FP\Pi_\mathcal{F}(\Pi_\mathcal{V}, I) \cap \Sigma = \mathcal{F}, FP\Pi_\mathcal{F} = \Pi_\mathcal{V}, tw(F) \leq k$ is captured by the intersection of $\mathcal{A}'$ and $\mathcal{A}$, which is also of required size. Therefore $V$ is the projection of $\mathcal{F}$ on the signature of view predicates and so $V$ is captured by some NTA $\mathcal{A}$ of required size by Proposition 5.
the condition of Proposition 7 holds for \( \forall \). Now applying Proposition 7, we conclude that \( Q \) is Datalog rewritable, and that the rewriting is of required size.

If \( Q \) is MDL the construction can be refined to produce an MDL rewriting; see the appendix for details. \( \square \)

We can use the same technique in the setting where the views are combinations of Monadic Datalog and CQs, while the query is Monadic Datalog, using normalization (Lemma 1) and the bound of Lemma 3. Normalization is used to enforce the bound on \( I \) required in Lemma 3. Although Lemma 3 requires connectivity, we can show that disconnected views can be replaced by connected ones.

**Theorem 2.** Suppose \( Q \) is a normalized Monadic Datalog query and \( V \) is a collection of Monadic Datalog and CQ views. If \( Q \) is monotonically determined by \( V \), then \( Q \) is rewritable over \( V \) in Datalog. The size of the rewriting is at most double-exponential in \( K = O(|Q|(|V|)) \).

The previous rewriting results involved restricting the views. We now note that if we restrict the query to be a UCQ, monotonic determinacy implies not only Datalog rewritability, but even UCQ rewritability, for arbitrary Datalog views:

**Proposition 8.** For views \( V \) in arbitrary Datalog we have:
1. If a CQ \( Q \) is monotonically determined by \( V \), then there is a CQ-rewriting of \( Q \) in terms of \( V \).
2. If a UCQ \( Q \) is monotonically determined by \( V \), then there is a UCQ-rewriting of \( Q \) in terms of \( V \).

In both cases the rewritings are polynomial size in \( |Q| \) and \( |V| \).

**Proof.** This can be seen as a "degenerate" variant of the forward-backward technique, which is well-known in the DB and KR literature [3, 22, 23]. Let \( Q \) be the disjunction of \( Q_1 : i \in S \). Let \( Q' = \bigvee_{i \in S} V(Q_i) \) denote the query that holds on an instance \( I' \) of the view schema exactly when for some \( i \in S \), there is a homomorphism of \( V(Q_i) \) into \( I' \). Equivalently, this is the query obtained by applying the views to each canonical database of a disjunct of \( Q \), and then interpreting the resulting facts as a query.

We claim that if \( Q \) is monotonically determined by \( V \), then \( Q' \) is a rewriting of \( Q \). In particular, if \( Q \) is a CQ, then \( Q' \) is just a CQ. We need to show that for each instance \( I, J \models Q \) iff \( V(I) \models Q' \).

(\( \Rightarrow \)) If some \( Q_k \) maps into \( I \), then \( V(Q_k) \) maps into \( V(J) \).

(\( \Leftarrow \)) Suppose some \( V(Q_k) \) maps into \( V(J) \). Monotonic determinacy implies homomorphic determinacy by Lemma 4 in the definition of homomorphic determinacy, take \( I_1 = Q_k \) and \( I_2 = I \). It is easy to check that \( V(I_1) \) maps into \( V(I_2) \) and \( I_1 \models Q \). It follows that \( I_2 \models Q \), in other words, that \( I \models Q \). \( \square \)

**5 DECIDABILITY**

We move from rewritability results to decision procedures for monotonic determinacy.

Monotonic determinacy testing procedure. Our decidability results will depend upon an characterization of monotonic determinacy, which we review here. Given a Datalog query \( Q \) and Datalog views \( V \), a canonical test for Monotonic Determinacy is a tuple \( (Q_i, D') \) that consists of:

- A CQ \( Q_i \) that is a CQ-approximation of \( Q \)
- An instance \( D' \) of the input schema formed by taking each fact \( F = V(c) \) in \( V(Q_i) \), choosing a CQ approximation \( Q' \) of \( Q_i \), and replacing \( F \) with fresh elements and facts from \( Q' \) that witness \( V(c) \). That is, firing the rule \( \forall x. V(x) \rightarrow Q'(x) \). In this case we say that \( D' \) is obtained from \( V(Q_i) \) by applying inverses of view definitions.

Such a test succeeds if \( D' \) satisfies \( Q \). It is easy to see that monotonic determinacy is characterized using tests:

**Lemma 5.** \( Q \) is monotonically determined over \( V \) if and only if every test succeeds.

We show that monotonic determinacy is decidable for some classes of views by bounding the treewidth of all instances \( D' \) that are the second component of some test.

**Theorem 3.** Suppose \( Q \) and \( V \) are Frontier-guarded Datalog queries. Then there is an algorithm that decides if \( Q \) is monotonically determined by \( V \) in \( 2ExpTime \).

**Proof.** In this proof the words "of required size" mean "doubly-exponential in \( Q \) and single-exponential in \( V \), \( C \) stands for the class of all CQ-approximations of \( Q \), \( \Sigma_V \) is the view signature and \( \Sigma_B \) is the initial signature.

We must check whether \( Q \) holds on all tests. As observed in the proof of Theorem 1, there is an integer \( k = O(|Q|, |V|) \) bounding the treewidth of all CQ-approximations of \( Q \) and views in \( V \). Let \( V = \{ F \mid \Sigma_V \mid F \mid \Sigma_B \in C, tw(F) \leq k, F \models \Pi_V \} \). As argued in the proof of Theorem 1, \( V \) is \( k \)-regular and captured by an NTA \( \mathcal{A}_V \) of required size.

Since \( Q \) is a monotone query, instead of checking whether all tests succeed, we will check an equivalent condition that \( Q \) holds on the class \( ETTEST(Q, V) \) which consists of all instances \( D' \) which can be obtained from an instance in \( V \) by applying inverses of view definitions while keeping the atoms of the view signature. Note that the treewidth of all instances in \( ETTEST(Q, V) \) is also bounded by \( k \). By Proposition 3, for each view \( V \) with definition \( Q_V \) there exists an automaton \( \mathcal{A}_{ETTEST} \) running on codes \( T \) which for each atom \( V(c) \) at a node \( n \) checks whether \( n \) has a descendant \( n' \) such that \( n' \) contains \( c \) and the subtree of \( T \) rooted at \( n' \) is a code of some CQ approximation of \( Q_V \). It should be clear that the automaton \( \mathcal{A} \) obtained as the product of \( \mathcal{A}_V \) and \( \mathcal{A}'_V \) for all \( V \in \mathcal{V} \) captures \( ETTEST(Q, V) \).

By Proposition 6, there is an NTA \( \mathcal{A}' \) of required size which recognizes those codes which do not satisfy \( Q \). So to check if \( Q \) is monotonically determined by \( V \) we construct the intersection of \( \mathcal{A}_V \) and \( \mathcal{A}' \) (which is of required size) and check if it is empty. The latter check is linear in the size of the automaton. \( \square \)

Using MDL normalization and the treewidth bounds of Lemma 3 we can use the same proof technique to extend this to a mix of CQ and Frontier-guarded Datalog views, provided that \( Q \) is in Monadic Datalog.
We now begin our negative results, starting with lower bounds
for testing monotonic determinacy. We first note some lower
bounds on monotonic determinacy that can be obtained through
straightforward reductions from containment or equivalence:

**Theorem 4.** Suppose Q is in Monadic Datalog, and V is a
collection of CQ and Frontier-guarded Datalog views. Then there
is an algorithm that decides if Q is monotonically determined
by V in 3ExpTime.

The previous cases of decidability required restricting the
views. We now observe that if we only restrict Q to be a CQ,
then we can reduce monotonic determinacy to checking equiva-
ience between a recursive and a non-recursive query, the one
created by the "simple forward backward method" of Proposition
8.

**Theorem 5.** If Q is a CQ and V is a collection of Datalog
views, then the problem of monotonic determinacy of Q over V
is decidable in 2ExpTime.

6 LOWER BOUNDS ON TESTING
MONOTONIC DETERMINACY

We now begin our negative results, starting with lower bounds
for testing monotonic determinacy. We first note some lower
bounds on monotonic determinacy that can be obtained through
straightforward reductions from containment or equivalence:

**Proposition 9.** Monotonic determinacy is

- NP-hard for CQ queries and views [9, 21]
- \( \Pi_2^P \)-hard for UCQ queries and UCQ views
- 2ExpTime-hard for CQ queries and MDL views
- 2ExpTime-hard for MDL queries and a fixed atomic view
- undecidable for Datalog queries and a fixed atomic view
  (cf. [14])

It is more challenging to get undecidability results in settings
where the equivalence problem for the views and queries is
decidable, as is the case for UCQs and Monadic Datalog [13].
The remainder of this section will be devoted to developing
techniques for this case.

A tiling problem is a tuple \( TP = (\text{Tiles}, HC, VC, IT, FT) \)
where \( \text{Tiles} = \{T_1, \ldots, T_k\} \), HC and VC are binary relations
("horizontal and vertical compatibility"), and IT and FT are
subsets of tiles that must be placed at the bottom left and top
right corner respectively.

A solution to a tiling problem consists of numbers \( n \) and \( m \),
and map \( \tau : \{1, \ldots, n\} \times \{1, \ldots, m\} \rightarrow \text{Tiles} \) such that

\[
\begin{align*}
(T1) & \quad (\tau(i, j), \tau(i + 1, j)) \in HC \text{ for } 1 \leq j \leq m \text{ and } 1 \leq i < n; \\
(T2) & \quad (\tau(i, j), \tau(i, j + 1)) \in VC \text{ for } 1 \leq j < m \text{ and } 1 \leq i \leq n. \\
(T3) & \quad \tau(1, 1) \in IT \text{ and } (T4) \quad \tau(n, m) \in FT. 
\end{align*}
\]

By a standard reduction from the halting problem for Turing
machines, it is easy to show that the problem "given a tiling
problem TP, tell if it has a solution" is undecidable. By reducing
this tiling problem to the problem of monotonic determinacy
for MDL queries and UCQ views we obtain

**Theorem 6.** The problem of monotonic determinacy for MDL
queries and UCQ views is undecidable.

The idea of the reduction is, given TP, to construct \( Q_{TP} \)
and \( V_{TP} \) which generate tests for monotonic determinacy that
look like \( (n, m) \)-grids with assignments of tiles. The query
\( Q_{TP} \) will have disjuncts that return "true" when they detect
violations of conditions (T1)–(T4). Thus \( Q_{TP} \) and \( V_{TP} \) will

have a failing test for monotonic determinacy iff the tiling
problem \( TP \) has a solution.

Figure 1, (a) shows such a test. We code the grid using
four binary relations \( YSucc, XSucc, YProj, XProj \) and unary
markers \( XEnd, YEnd \). Vertical and horizontal axes are
represented as chains of \( YSucc \)- and \( XSucc \)-atoms respectively.
The "grid points" are linked via \( XProj \) and \( YProj \)-edges to
their projections on the axes. The unary predicates \( XEnd \) and
\( YEnd \) mark the ends of the axes.

Note how CQs \( HA(z_1, z_2, x_1, x_2, y) = YProj(y, z_1) \wedge YProj(y, z_2) \wedge
YProj(x_1, z_1) \wedge YProj(x_2, z_2) \wedge XSucc(x_1, x_2) \text{ and } VA(z_1, z_2, x, y_1, y_2) =
YProj(y_1, z_1) \wedge YProj(y_2, z_2) \wedge XProj(x, z_1) \wedge XProj(x, z_2) \wedge
XSucc(y_1, y_2) \) (see Figure 1, (b)) can be used to check vertical
and horizontal adjacency between grid points. For example,
\( HA(z_1, z_2, x_1, x_2, y) \) says \( z_1 \) and \( z_2 \) have the same \( y \)-projection,
while the \( x \)-projection of \( z_2 \) is next to the \( x \)-projection of \( z_1 \).
Query \( H(z_1, z_2) = \exists y_1 \exists x_2 \ HA(z_1, z_2, x_1, x_2, y) \) holds for
grid points \( z_{1,1}^0 \) and \( z_{2,2}^0 \) iff \( z_{1,1}^0 \) is the right neighbour of \( z_{2,2}^0 \).

Given a tiling problem \( TP \), we define the query \( Q_{TP} \) as a
disjunction \( Q_{\text{START}} \lor Q_{\text{HELPER}} \lor Q_{\text{VERIFY}} \) where Monadic Datalog
query \( Q_{\text{START}} \) and UCQs \( Q_{\text{HELPER}} \) and \( Q_{\text{VERIFY}} \) are defined by
the following programs:

1. \( Q_{\text{START}} \leftarrow A(x), B(x) \)
2. \( A(x) \leftarrow XSucc(x, x'), A(x'), C(x') \)
3. \( A(x) \leftarrow XEnd(x) \)
4. \( B(y) \leftarrow YSucc(y, y'), B(y'), D(y') \)
5. \( B(y) \leftarrow YEnd(y) \)
6. \( Q_{\text{HELPER}} \leftarrow C(u), YProj(y, z), XProj(x, z) \)
7. \( Q_{\text{HELPER}} \leftarrow D(u), YProj(y, z), X Proj(x, z) \)
8. \( Q_{\text{VERIFY}} \leftarrow HA(z_1, z_2, y_1, y_2) \land T_1(z_1), T_2(z_2) \)
   for all pairs \( (T_1, T_2) \notin HC \)
We can verify that a solution of our tiling problem corresponds to monotonic determinacy, which will prove useful in both our undecidability and non-rewritability results.

A typical CQ-approximation of \( Q_{\text{Start}} \) is shown in Figure 2 (a), and it generates the axes of the grid which are marked with unary predicates \( C \) and \( D \). The view-image of such \( C \) is shown in Figure 2 (b). This view image for each grid-point contains an \( S \)-atom, and so a grid-like test as in Figure 1 (a) can be constructed out of this view image by replacing each of these \( S \)-atoms with any of the disjunctions other than the first disjunct in the definition of the grid-generating view.

When we run \( Q \) on the tests, \( Q_{\text{Verify}} \) comes into play. Note the correspondence between rules \( 8 \) – \( 11 \) for \( Q_{\text{Verify}} \) and the negations of conditions (1) – (4) in the definition of a solution of a tiling problem. Thus, when executed on a grid-test from Figure 1 (a), \( Q_{\text{Verify}} \) returns false iff a grid test is a solution to \( TP \). The query \( Q_{\text{Helper}} \) ensures that we are not harmed in the case where the grid-generating views are applied with the first rule.

We can verify that a solution of our tiling problem corresponds to monotonic determinacy, which will prove useful in both our undecidability and non-rewritability results:

Proposition 10. \( Q_{TP} \) is not monotonically determined by \( V_{TP} \) iff \( TP \) has a solution.

7. NON-REWITABILITY

We now turn to negative results concerning rewritability.

Pebble games. A partial homomorphism from \( I \) to \( I' \) is a mapping \( h \) from a subset \( D \triangleq \text{ADOM}(I) \) to \( \text{ADOM}(I') \) such that \( R(c_1 \ldots c_n) \in I \) implies \( R(h(c_1) \ldots h(c_n)) \in I' \), provided each \( c_i \in D \). Let \( k \geq 2 \). In the existential \( k \)-pebble game we have two players, the Spoiler and the Duplicator, each having a set of pebbles \( \{p_1, \ldots, p_k\} \) and \( \{q_1, \ldots, q_k\} \), respectively. The game is played on two instances \( I \) and \( I' \) over the same schema. In each round, the Spoiler either places a pebble \( p_i \) on some element of \( I \) or removes \( p_i \) from \( I \), to which the Duplicator responds by placing its corresponding pebble \( q_i \) on some element of \( I' \) or by removing \( q_i \) from \( I' \), respectively. The Duplicator wins the game if he has a winning strategy, i.e., if he can indefinitely continue playing the game in such a way that after each round, if \( a_1, \ldots, a_k \) are the elements in \( I \) marked by the Spoiler’s pebbles \( \{p_1, \ldots, p_k\} \), and \( a'_1, \ldots, a'_k \) are the elements in \( I' \) marked by the Duplicator’s pebbles \( \{q_1, \ldots, q_k\} \), then the relation \( \{(a_1, a'_1), \ldots, (a_k, a'_k)\} \) is a partial homomorphism from \( I \) to \( I' \).

Recall that if there is a homomorphism from \( I \) to \( I' \), we write \( I \rightarrow I' \). Similarly, if Duplicator wins the game on \( I \) and \( I' \), then we write \( I \rightarrow_k I' \). Observe that \( I \rightarrow I' \) implies \( I \rightarrow_k I' \), for every \( k \geq 2 \).

The following property relates the game to homomorphisms from structures of bounded treewidth:

Fact 1. [5] Let \( k \geq 2 \). Let \( I \) and \( I' \) be two instances over the same schema. Then the following are equivalent:

(1) \( I \rightarrow_k I' \).

(2) For every instance \( I'' \) of treewidth \( \leq k - 1 \), if \( I'' \rightarrow I \), then \( I'' \rightarrow I' \).

Existential pebble games with \( k \) pebbles preserve truth of Boolean Datalog queries with rule bodies of size at most \( k \). Thus games can be used to show non-rewritability in Datalog:

Fact 2. [20] Let \( Q \) be a Boolean query. Suppose there exists two instances \( I_k \) and \( I_k' \) such that \( Q(I_k) = \text{TRUE} \) and \( I_k \rightarrow_k I_k' \), for infinitely many \( k \)’s. Then \( Q \) is not definable in Datalog.

Let \( I \) be an instance and \( k \geq 2 \) be an integer. An instance \( I' \) is a \( k \)-unravelling of \( I \) if there is a homomorphism \( \Phi \) from \( I' \) to \( I \) and a tree decomposition \( (\tau, (\lambda(u))_{u \in \text{VERTICES}(\tau)}) \) of \( I' \) of width at most \( k \), such that:

(1) For each \( u \in \text{VERTICES}(\tau) \), the mapping \( \Phi \mid_{\lambda(u)} \) is a partial isomorphism from \( I' \) to \( I \).

(2) For \( u \in \text{VERTICES}(\tau) \) with children \( u_1, \ldots, u_r \), the set \( \{\Phi(\lambda(u_1)), \ldots, \Phi(\lambda(u_r))\} \) contains the collection of all non-empty subsets of \( I \) of size \( \leq k \).

If, in addition we have \( |\lambda(u) \cap \lambda(v)| \leq 1 \) for all non-equal \( u \) and \( v \) in \( \text{VERTICES}(\tau) \), then we say that \( I' \) is a \( (1, k) \)-unravelling of \( I \). Duplicator has a winning strategy between an instance and its \( (1, k) \)-unravelling in a variation of the \( k \)-pebble games.
in which at most one pebble can remain in place in each move. Such games preserve Boolean Monadic Datalog queries with bodies of size $k$, and hence each Boolean Monadic Datalog query is preserved under $(1,k)$-unravellings for sufficiently large $k$. So we have the following variant of Fact 2:

**Fact 3.** Let $Q$ be a Boolean query. Suppose there exist two instances $I_k$ and $I'$ such that $Q(I_k) = TRUE$, $Q(I'_k) = FALSE$ and $I'_k$ is a $(1,k)$-unravelling of $I_k$, for infinitely many $k$’s. Then $Q$ is not definable in Monadic Datalog.

Note that the treewidth of any $k$-unravelling is at most $k - 1$. Observe also that all $k$-unravellings of an instance $I$ are homomorphically equivalent. The following facts about unravellings will be useful (see the appendix):

**Fact 4.** Let $k \geq 2$. Let $I$ be an instance and $U$ be any $k$-unravelling of $I$. Then the following hold:
1. $U \rightarrow I$ and $I \rightarrow U$.
2. For every instance $I'$, we have $I \rightarrow_{k} I'$ iff $U \rightarrow I'$.

Non-rewrriwritability in Monadic Datalog. We recall that Monadic Datalog queries monotonically determined over CQ views always have FGDL rewritings (e.g. [14], or Thm 2). We show that they may not be rewritable in MDL:

**Theorem 7.** There exists a Monadic Datalog query $Q$ and a set of CQ views $V$ such that $Q$ is rewritable with respect to $V$ in Datalog, but not in Monadic Datalog.

**Proof.** Consider the following Monadic Datalog query $Q$

- $W(x) \leftarrow A(x, y), B(y, v), C(x, z), D(z, v), U(v)$
- $W(x) \leftarrow A(x, y), B(y, v), C(x, z), D(z, v), W(v)$
- $S(x, y, z) \leftarrow A(x, y), B(y, v), C(x, z)$
- $R(y, z, x', z') \leftarrow B(y, v), D(z, v), A(v, y'), C(v, z')$
- $T(y, z, v) \leftarrow U(v), B(y, v), D(z, v)$

Set $Q$ checks whether the instance $I$ contains the points $s \in M^I$ and $t \in U^I$ which are connected by a sequence of “diamonds” (see Figure 3, (a)).

We claim that there is no Monadic Datalog rewriting of $Q$ in terms of these views. To prove this, given an integer $k$, we construct two instances $I_k$ and $I'_k$ such that $I_k \models Q$, $I'_k \not\models Q$, but Duplicator wins in the $(1,k)$-game for the view images $V(I_k)$ and $V(I'_k)$.

Let $I_k$ be a sequence of $k+1$ diamonds from Figure 3, (a) and $I'_k$ be the view image of $I_k$ from Figure 3, (b). Let $J'_k$ be the (infinite) $(1,k)$-unravelling of $J_k$. Let $I'_k$ be the result of applying “inverse rules”:

- $S(x, y, z) \rightarrow M(x) \land A(x, y) \land C(x, z)$
- $R(y, z, x', z') \rightarrow \exists v B(y, v) \land D(z, v) \land A(v, y') \land C(v, z')$
- $T(y, z, v) \rightarrow U(v) \land B(y, v) \land D(z, v)$

$J'_k$ and removing the view predicates. There are two types of elements in $I'_k$, those that were present in $I'_k$ and those introduced by the existential quantifier over $v$ in the second rule which are called anonymous.

We first claim that the view image of $I'_k$ is $J'_k$. To see this, consider a homomorphism $h$ from the body of $q(y, z, y', z') = B(y, v), D(z, v), A(v, y'), C(v, z')$ into $I'_k$. Note that $h$ must map $v$ into an anonymous point, and so $y, z, y'$ and $z'$ must be mapped to the points $y_0, z_0, y'_0, z'_0$ for some $R(y_0, z_0, y'_0, z'_0) \in J'_k$. If follows that any $R$-atom in $V(I'_k)$ is in $J'_k$. This also holds for $S$-atoms thanks to the unary predicate $M$, and also for $R$-atoms, thanks to the unary predicate $U$.

Our next claim is that $I'_k \models Q$ iff $J'_k \models Q$. Indeed, $I_n$ maps into $I'_k$ iff $J_n$ maps into $J'_k$.

Finally, we claim that $J'_k \not\models Q$. We show that for any $n$ there is no homomorphism from $J_k$ to $J'_k$. Indeed, as $J'_k$ maps homomorphically onto $J_k$, for any points $s, t$ in $J'_k$ with $s \in \pi_1(S^I), t \in \pi_3(T^I)$, the distance between $s$ and $t$ (measured in the Gaifman graph for $J'_k$) cannot be less than $k + 1$. This implies the claim for $1 \leq n \leq k$. For $n \geq k + 1$ the claim follows from the observation that there is no homomorphism from the query describing the pattern in Figure 4 into $J'_k$. Indeed, it can be easily shown by induction that if such a homomorphism existed, then there would be a single bag in $J'_k$ containing its whole image. This is a contradiction, as all bags are of size $k$, and the query in question has $2k + 2$ variables.

It follows that $I_k$ and $I'_k$ are as required. □

Non-rewritability in Datalog. Now we show that monotonic determinacy does not imply Datalog rewritability, even for Monadic Datalog queries and UCQ views:

**Theorem 8.** There exists a Monadic Datalog query $Q$ and a set of UCQ views $V$ such that $Q$ is monotonically determined by $V$ but there is no Datalog rewriting of $Q$ over $V$.

The query $Q$ and views $V$ we will use have the form $Q_{TP}$ and $V_{TP}$ for a particular tiling problem $TP$ as defined in Section 6. We define a schema $\delta := (h, V, I, F)$ where $h, V$ are binary and $I, F$ are unary relations. Let $TP = (Tiles, HC, VC, IT, FT)$ be a tiling problem. Give a database instance $I$ for schema
δ, we say that I can be tiled by TP if there is an assignment of each element of I to a tile where V, I, F satisfy the horizontal, vertical, initial and final constraints. We denote by ITP the tiling problem TP viewed as a relational structure for δ, with Tiles the domain. Then an instance for δ can be tiled by TP exactly when it has a homomorphism into ITP.

For n, m ≥ 1, we denote by I_n,m the database instance with domain \{((i, j) : 1 ≤ i ≤ n and 1 ≤ j ≤ m) and facts I((1, 1)), F((n, m)), H((i, j), (i + 1, j)), \} for every 1 ≤ i ≤ n and 1 ≤ j ≤ m, and V((i, j), (i, j + 1)), for every 1 ≤ i ≤ n and 1 ≤ j < m. Then TP has a solution in the usual sense if I_n,m can be tiled with TP.

We can adapt techniques of [4] to show that there is a tiling problem for which no n × m rectangular grid can be tiled, but where for each k large enough grids can be “k-approximately tiled”, in the sense of having k-unravellings that can be tiled.

**Lemma 6.** There is a tiling instance TP* such that I_n,m can not be tiled with TP* for each n, m ≥ 1 but for each n, m ≥ 3 and each k with 2 ≤ k < \min(n, m) any k-unravelling of I_n,m can be tiled with TP*.

**Proof of Theorem 8.** Let TP* be the tiling instance from Lemma 6 and let QTP* and VTP* be the ML query and UCQ views from Theorem 6. Recall that QTP* and VTP* are defined over the schema

\[ \sigma := \{X Succ, Y Succ, \text{C, D, XEND, YEND, XPROJ, YPROJ, T}_1, \ldots, T_p\} \]

where \(T_1, \ldots, T_p\) is the tile set of TP*. Since I_n,m can not be tiled with TP*, for each n, m ≥ 1, the tiling instance TP* has no solution, and hence QTP* is monotonically determined by VTP*.

Fix \(\ell ≥ 10\). We shall define instances I_\ell and I_\ell' over \(\sigma\) such that VTP*(I_\ell) \rightarrow [\sqrt{\ell - 1}] VTP*(I_\ell'), \quad QTP*(I_\ell) = \text{TRUE} and \quad QTP*(I_\ell') = \text{FALSE} By applying Fact 2, this implies that QTP* has no Datlog rewriting over VTP*, as required. The instance I_\ell has domain \(X_0 \cup X \cup Y\), where \(X := \{x_1, \ldots, x_{\ell}\}\) and \(Y := \{y_1, \ldots, y_{\ell}\}\), and facts \(D(x_i), C(y_i)\), for all 1 ≤ i ≤ \ell, along with

\[ \text{X Succ}(x_i, x_{i+1}), Y Succ(y_i, y_{i+1}) \text{ for all } 1 ≤ i < \ell \]

along with XEND(x_\ell), YEND(y_\ell), YSucc(z_0, y_1), and X Succ(z_0, x_1). Figure 2 (a) depicts I_\ell. Informally, I_\ell is the expansion of QTP* (more precisely of QSTART) representing the \((1, \ell)\)-grid. In particular, QTP*(I_\ell) = \text{TRUE}.

Intuitively, we would now like to define I_\ell' so that its view image contains a \([\sqrt{\ell - 1}]\) unravelling of the view image of I_\ell. By Fact 4 we would have VTP*(I_\ell) \rightarrow [\sqrt{\ell - 1}] VTP*(I_\ell') as required. But using Lemma 6 and the definition of QTP* we hope to show QTP*(I_\ell') = \text{FALSE}. We will follow this intuition, but to define the appropriate I_\ell' we will need to construct several auxiliary instances. Let E_\ell := VTP*(I_\ell). Figure 2 (b) depicts E_\ell. Recall that view images are defined over schema \(\tau:\)

\[ \{V_X Succ, V_Y Succ, V_X END, V_Y END, V_{T_1}, \ldots, V_{T_p}, V_{\text{C HELPER}}, V_{\text{D HELPER}}, V_{\text{HH}}, V_{\text{VA}}, V_{\text{V}}, V_{\text{S}}\} \]

Intuitively, E_\ell copies the X Succ, Y Succ, X END and Y END-facts from I_\ell, while the S-facts correspond to the product Y × X.

Let U_\ell be a \([\sqrt{\ell - 1}]\)-unravelling of E_\ell, which is witnessed by a homomorphism \(\Phi : U_\ell \rightarrow E_\ell\) and a tree decomposition \((\tau, (\lambda(u)_{u \in \text{vertices}(\tau)})\) of U_\ell.

In order to exploit Lemma 6, we need to interpret U_\ell as an unravelling of the grid I_\ell\ell'. The idea is to define a new instance W_\ell over schema \(\delta = \{H, V, I, F\}\) (recall that \(\delta\) is the schema of I_\ell\ell') whose domain consists of all the \(S\)-facts of U_\ell and the horizontal and vertical successor relations are interpreted in the natural way. Thus we can think of W_\ell as an unravelling of the S-facts of E_\ell, which in turn correspond to grid points of I_\ell\ell' (the fact \(S(y_j, x_i)\) corresponds to the point \((i, j)\)). Formally, W_\ell is defined as follows:

1. The domain of W_\ell contains all pairs \((z, w)\) such that \(S(w, z)\) is a fact in U_\ell.  
2. I((w, z)) is a fact iff \(\Phi(w) = y_1\) and \(\Phi(z) = x_1\). Similarly, \(F((w, z))\) is a fact iff \(\Phi(w) = u_\ell\) and \(\Phi(z) = x_1\).
3. \(H((w, z))\) is a fact iff \(w = w'\) and \(V_{\text{S Succ}}(z, z')\) is a fact in U_\ell. Similarly, \(V((w, z))\) is a fact iff \(z = z'\) and \(V_{\text{S Succ}}(w, w')\) is a fact in U_\ell.

**Claim 1.** W_\ell can be tiled by TP*.

**Proof.** We use the characterization for tilings of W_\ell as homomorphisms into ITP*. Then by Lemma 6 and Fact 1, it suffices to show (a) W_\ell → I_\ell\ell' and (b) \(\text{tw}(W_\ell) ≤ \ell - 2\).

For (a), we can take the homomorphism \(\psi\) such that for every \((w, z)\) in W_\ell, we have \(\psi((w, z)) = (i, j)\) iff \(\Phi(w) = y_j\) and \(\Phi(z) = x_i\), for 1 ≤ i ≤ \ell, j ≤ \ell. Let us argue that \(\psi\) is a homomorphism. If \(I((w, z))\) is a fact in W_\ell, by definition we have \(\Phi(w) = y_1\) and \(\Phi(z) = x_1\), and then \(\psi(w) = y_1\) and \(\psi(z) = x_1\), which is a fact in I_\ell\ell'. If \(H((w, z))\) is a fact in W_\ell, then \(w = w'\) and \(S(w, z), S(w', z')\) and \(V_{\text{S Succ}}(z, z')\) are facts in U_\ell. It follows that \(\Phi(w) = \Phi(w') = y_j\), \(\Phi(z) = x_i\) and \(\Phi(z') = x_{i+1}\), for some 1 ≤ i ≤ \ell and 1 ≤ j ≤ \ell. Hence \(\psi((w, z))\) and \(\psi((w', z'))\) are facts in I_\ell\ell'. The argument is analogous for F and V-facts.

For condition (b), recall that \((\tau, (\lambda(u)_{u \in \text{vertices}(\tau)})\) is a decomposition of U_\ell with \(|\lambda(u)| ≤ \sqrt{\ell - 1}\), for all \(u \in \text{vertices}(\tau)\). We define a decomposition \((\tau', (\lambda'(u)_{u \in \text{vertices}(\tau')}))\) for W_\ell with \(\tau' := \tau\) and, for each \(u \in \text{vertices}(\tau')\), we have

\[ \lambda'(u) = \{(w, z) : (w, z) \subseteq \lambda(u)\} \]

along with all the \(u \in \text{vertices}(\tau')\), we have

\[ \lambda'(u) = \{(w, z) : (w, z) \subseteq \lambda(u)\} \]

The connectedness condition is inherited from \(\tau\). Suppose that we have a fact \(H((w, z), (w', z'))\) in W_\ell. Then \(w = w'\), and \(S(w, z), S(w', z), V_{\text{S Succ}}(z, z')\) are facts in U_\ell. There must exist \(u \in \text{vertices}(\tau)\) such that \(w = w', z, z' \subseteq \lambda(u)\) (as every clique is always contained in a bag). It follows that \(\lambda(u) \subseteq \lambda'(u)\). The argument for V-facts is analogous. Finally, note that \(|\lambda'(u)| ≤ |\lambda(u)|^2 ≤ \ell - 1\), for all \(u \in \text{vertices}(\tau')\), we conclude that the treewidth of W_\ell is ≤ \ell - 2 as required.

Using the tiling solution \(\chi\) for W_\ell given by Claim 1 and “chasing with the inverse rules of the view definitions” we can move to the desired instance I_\ell' for the base schema \(\sigma\). The instance I_\ell' is obtained from I_\ell by replacing each fact \(V_{\text{S Succ}}(w, z), V_{\text{S Succ}}(w, z), V_{\text{XEND}}(w)\) by
What about separators, which are like rewritings, but not views, there may be no separator in computation is halting and return the input. Determinism of an input and a valid computation of a high-complexity de-

by considering a query verifying that the base data represent

tained Datalog views and queries where the certain answers

over computable time bound. We show that when we turn to general Datalog queries and may be no Datalog rewriting even in the case of UCQ views. This is true because every view image is the view image

for

or

D

monotonically determined over UCQ views, it may not be Datalog rewritable, we do not know whether a rewriting can be obtained by expanding the language, e.g. to stratified Dat-

datolog; it is true of the particular constructions constructed in

8 CONCLUSION

We have taken some basic steps in understanding monotonic determinacy for recursive queries. We leave quite a number of gaps in both the understanding of rewritability and decidabil-

complexity of separators. Thus far we have seen that there


determinacy for recursive queries. We leave quite a number

of gaps in both the understanding of rewritability and decidabil-

ity of testing monotonic determinacy, as one can see from Figures 1 and 2. To highlight just one, while we have shown that monotonic determinacy of a Datalog query over Datalog views does not imply a rewriting in any reasonable complexity class, we do not know what can be said when the query is restricted; e.g. to be in Frontier-guarded Datalog. While we have shown that when a Monadic Datalog query is monotonically determined over UCQ views, it may not be Datalog rewritable, we do not know whether a rewriting can be obtained by expanding the language, e.g. to stratified Data-

datalog; it is true of the particular constructions constructed in

Theorem 8. See the appendix.

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APPENDIX

FINITE VARIANTS

In the body of the paper we used a semantics in terms of arbitrary instances, but we claimed that all the results hold when instances are restricted to be finite. One can relativize all of the definitions to finite instances. We say \( Q \) is monotonically determined over \( V \) with respect to finite instances if whenever two finite instances agree on \( V \) they must agree on \( Q \). Similarly we can talk about \( Q \) being rewritable in logic \( L \) with respect to views \( V \) over finite instances if there is a query \( R \in L \) such that for every finite instance \( I \) evaluating \( R \) on the \( V \)-image of \( I \) gives the same result as evaluating \( Q \) on \( I \).

For the results about deciding monotonic determinacy and the positive results about rewriting monotonically determined queries in Datalog, the equivalence of the finite and unrestricted variants follows from the following well-known fact:

PROPOSITION 11. If Datalog query \( Q \) is monotonically determined over Datalog views \( V \) for finite instances, then \( Q \) is monotonically determined over \( V \) (over all instances).

For all the languages \( L \) we consider here (Datalog, MDL, etc.) if \( Q \) is \( L \)-rewritable over \( V \) with respect to finite instances, it is \( L \)-rewritable with respect to all instances.

PROOF. We prove the first statement. Assume \( Q \) is monotonically determined over Datalog views \( V \) for finite instances, and suppose we have two instances \( I \) and \( I' \), perhaps infinite with \( V(I) \subseteq V(I') \) and \( Q(I) \) not contained in \( Q(I') \). Fix \( t \in Q(I) - Q(I') \). There is a finite subinstance \( I_0 \) of \( I \) with \( t \in Q(I_0) \), \( V(I_0) \subseteq V(I') \) and \( V(I_0) \) is finite, so there is a finite subinstance \( I_0' \) of \( I' \) with \( V(I_0) \subseteq V(I_0') \). Now \( I_0 \) and \( I_0' \) contradict the hypothesis on \( Q \) and \( V \).

For the theorem, we use the fact that equivalence of Datalog queries over finite instances implies equivalence over all instances.

It remains to consider our negative results about rewritings and separators. An easy case is Theorem 9, showing that no computable function bounds the time of a separator for Datalog queries monotonically-determined over Datalog views. The query and views in the example are monotonically determined over all instances, hence over finite instances; the argument that the query does not have separators growing at a given time bound does make no use of infinitary methods, and hence holds to show that there is no such separator over finite instances.

The proofs of Theorems 7 and 8 both make use of unravellings, which can be infinite. For Theorem 7, we can argue just by looking at the statement: if the query \( Q \) can be rewritten to a Monadic Datalog query \( R \) over the views \( V \) with respect to finite instances, then consider the Datalog query \( R_Y \) formed by composing the rules for \( V \) and \( R \), treating each view predicate as an intensional predicate of \( R_Y \). Then \( Q \) is equivalent to \( R' \) over all finite instances. But then, using the fact that a witness to non-containment of Datalog must be finite, we see that \( Q \) is equivalent to \( R' \) over all instances, a contradiction of the theorem. The same argument holds for Theorem 8.
which are view images of an instance of the base schema. Given a query \( Q \) where \( \Gamma \) is any instance of the base schema, \( \Gamma \) is a skolem function.

Consider a collection of CQ views \( V = \{ (V, Q_V) \mid V \in \Sigma_V \} \) over a base schema \( S \). We associate a set of TGDs: \( \Gamma_V = \{ V(x) \rightarrow \exists y \, Q_V(x, y) \} \) consisting of inverse rules. By replacing the existential quantifiers with skolem functions, we can consider \( \Gamma_V \) as a logic program with input signature the view schema and the base schema as intensional relations.

**Example 2.** Suppose we have just one view, \( V(x, y, z) = \exists u \, S(x, y, u) \land S(u, y, z) \). Then the corresponding inverse rules are:

\[
S(x, y, f(x, y, z)) \leftarrow V(x, y, z) \\
S(f(x, y, z), y, z) \leftarrow V(x, y, z)
\]

where \( f \) is a skolem function.

Note that these rules have a single atom in each rule body. We call such rules atomic. We refer to the intensional predicates of the rules as extensionally-based IDBs.

Let \( Q \) be a Boolean Datalog query over the base signature, with goal predicate \( \text{Goal}_Q \). We write \( Q \cup \Gamma_V \), the inverse rules logic program to indicate the query that unions the rules of \( Q \) and those of \( \Gamma_V \), using the goal predicate \( \text{Goal}_Q \). This is a Boolean logic program query over the base signature.

**Example 3.** Consider the frontier-guarded Datalog query \( Q \) with goal predicate \( \text{Goal} \) and rules:

\[
\text{Goal} \leftarrow \text{Conn}(x, x) \\
\text{Conn}(x, y) \leftarrow S(x, y, z), \text{Conn}(x, z), \text{Conn}(z, y) \\
\text{Conn}(x, y) \leftarrow S(x, y, z)
\]

Then the inverse rules logic program \( Q \cup \Gamma_V \) for query \( Q \) and \( V \) the single view in Example 2 would have all the rules for \( Q \) along with the two inverse rules for \( V \) coming from Example 2. Observe that \( S \) is now an intensional predicate along with \( \text{Goal} \) and \( \text{Conn} \).

When \( Q \) is frontier-guarded, the inverse-rules logic program is not necessarily frontier-guarded. Indeed, the rules in \( \Gamma_V \) contain only intensional predicates. However, it is immediate that the head variables of each rule are contained in some atom with an extensionally-based intensional predicate.

One can characterize the output of the inverse rules logic program on an arbitrary instance of the view schema, not just those which are view images of an instance of the base schema. Given a query \( Q \) over the input schema and an instance \( \mathcal{J} \) of the view schema, the certain answers of \( Q \) with respect to \( V \) over \( \mathcal{J} \) is the intersection of \( Q(\mathcal{I}) \) over all \( \mathcal{I} \) such that \( V(\mathcal{I}) \subseteq \mathcal{J} \). It is easy to see that:

**Theorem 10.** [14] For any such \( \mathcal{J} \), \( Q \cup \Gamma_V \) evaluated on \( \mathcal{J} \) gives the certain answers of \( Q \) with respect to \( V \) over \( \mathcal{J} \).

It follows from the theorem that if \( Q \) is monotonically determined by \( V \) and \( \mathcal{I} \) is any instance of the base schema, \( Q \cup \Gamma_V \) evaluated over \( V(\mathcal{I}) \) is the same as \( Q(\mathcal{I}) \). Using the terminology from the body of the paper this can be restated as: \( Q \cup \Gamma_V \) is a separator for \( Q \) with respect to \( V \).

We now turn to the de-functionalization step of the inverse rules algorithm. This replaces the inverse rules logic program with an ordinary Datalog program over a different set of intensional predicates. These predicates are obtained by annotating the intensional predicates of the logic program. The idea is that an IDB atom \( R(x, f(x, y)) \) containing skolem term \( f \) created during the construction of the fixpoint of the logic program will be mimicked by an atom \( R^{1/1,2}(x, y) \) created during the fixpoint of the de-functionalized program. We refer to the original paper [14] for the details, but illustrate the idea with an example.

**Example 4.** We give part of the de-functionalization of the inverse rules logic program from Example 3. The inverse rules themselves will translate to the rules:

\[
S^{1,2, f(1,2,3)}(x, y, z) \leftarrow V(x, y, z) \\
S^{f(1,2,3), 2, 3}(x, y, z) \leftarrow V(x, y, z)
\]

The following rule in the inverse rules logic program:

\[
\text{Conn}(x, y) \leftarrow S(x, y, z), \text{Conn}(x, z), \text{Conn}(z, y)
\]
will generate many annotated rules. One can consider the substitution \( x = f(x_1, y_1, z_1), y = y_1 z = f(x_2, y_2, z_2) \) into the rule above, which gives the rule:

\[
\text{CONN}(f(x_1, y_1, z_1), y_1) \leftarrow S(f(x_1, y_1, z_1), y_1, z_1), \text{CONN}(f(x_1, y_1, z_1), f(x_2, y_2, z_2)), \text{CONN}(f(x_2, y_2, z_2), y_1)
\]

The corresponding annotated rule would be:

\[
\text{CONN}^{f(1,2,3),2}(x_1, y_1, z_1) \leftarrow S^{f(1,2,3),2}(x_1, y_1, z_1), \text{CONN}^{f(1,2,3),f(4,5,6)}(x_1, y_1, z_1, x_2, y_2, z_2), \text{CONN}^{f(1,2,3),4}(x_2, y_2, z_2, y_1)
\]

We can see that annotated rules as produced by the standard inverse-rules algorithm are not frontier-guarded. However, we note that:

- each IDB that is an annotation of an extensionally-based IDB appears in exactly one rule, and that rule is atomic. Hence in particular the unique such rule has a view atom as a frontier-guard.
- the head variables of each rule co-occur in an atom that is an annotation of an extensionally-based relation.

From this it follows that we can conjoin to each rule a view atom that makes the rule frontier-guarded: namely, we can conjoin the view atom corresponding to the annotated extensionally-based relation in the second item.

**Example 5.** Continuing the example above, the annotated rule:

\[
\text{CONN}^{f(1,2,3),2}(x_1, y_1, z_1) \leftarrow S^{f(1,2,3),2}(x_1, y_1, z_1), \text{CONN}^{f(1,2,3),f(4,5,6)}(x_1, y_1, z_1, x_2, y_2, z_2), \text{CONN}^{f(1,2,3),4}(x_2, y_2, z_2, y_1)
\]

is converted to the frontier-guarded rule:

\[
\text{CONN}^{f(1,2,3),2}(x_1, y_1, z_1) \leftarrow V(x_1, y_1, z_1), S^{f(1,2,3),2}(x_1, y_1, z_1), \text{CONN}^{f(1,2,3),f(4,5,6)}(x_1, y_1, z_1, x_2, y_2, z_2), \text{CONN}^{f(1,2,3),4}(x_2, y_2, z_2, y_1)
\]
PROOFS FOR SECTION 3: TREENWIDTH BOUNDS AND THE FORWARD-BACKWARD METHOD

Proof of Lemma 1
Recall the statement:

Let $Q$ be a normalized Monadic Datalog query. Then there is a number $k = O(|Q|)$ such that all CQ-approximations of $Q$ have tree decomposition $\mathcal{D}$ of width $k$ with $l(\mathcal{D}) \leq 2$.

Proof. In fact, $k$ is the maximal number of variables in a body of $Q$. Then the definition of a CQ-approximation gives rise to a tree decomposition $\mathcal{D}$ of width $k$. The property $l(\mathcal{D}) \leq 2$ follows from the fact that $Q$ is normalized. \qed

Proof of Lemma 2
Recall the statement:

If $\Pi$ is a Datalog program such that all its rules are frontier-guarded, and $I$ is an instance of treewidth $k$, then $\text{FPEval}(\Pi, I)$ is of treewidth $k$.

Proof. Monadic rules introduce only monadic predicates which do not increase the treewidth of the instance. Guarded rules introduce atoms which are wholly inside the EDB guards. So treewidth does not increase when the rules are fired. \qed

PROOF OF LEMMA 3
Recall the statement:

Let $\mathcal{D}$ be a tree decomposition of a data instance $I$ of treewidth $k$ with $l(\mathcal{D}) \leq 2$. Let $V$ be a set of connected CQ views, and $V(I)$ the view image of $I$ under $V$. Let $r$ be the greatest radius of a CQ in $V$. Then the treewidth of $V(I)$ is at most $k' = \frac{k(k+1)}{k-1}$.

Proof. For a bag $b$ of $\mathcal{D}$ and an integer $n$ define recursively its $n$-extension by setting $\text{ext}(b, 0) = b$ and $\text{ext}(b, n) = \{u \mid \exists v \in \text{ext}(b, n-1) \text{ such that } u$ and $v$ belong to a same bag of $\mathcal{D}\}$. Since $l(\mathcal{D}) \leq 2$, it is easy to see by induction that $|\text{ext}(b, n)| \leq k + k^2 + \ldots + k^{n+1} = \frac{k(k^{n+1}-1)}{k-1}$. Let $\mathcal{D}'$ be a tree of bags whose set of nodes is $\text{ext}(b, r)$, with an edge between $\text{ext}(b, r)$ and $\text{ext}(b', r)$ exactly when there is an edge from $b$ to $b'$ in $T$. We claim that $\mathcal{D}'$ is a tree decomposition of $V(I)$.

First we show that for any element $v$ the set of all bags in $\mathcal{D}'$ containing $v$ is connected. Suppose that two nodes $n_1$ and $n_2$ of $\mathcal{D}'$ contain $v$. Then there are bags $b_1$ and $b_2$ in $\mathcal{D}$ such that $n_1 = \text{ext}(b_1, r)$ and $n_2 = \text{ext}(b_2, r)$. Thus $v$ belongs to some bags $b'_1$ and $b'_2$ which are at most $r$ steps away from $b_1$ and $b_2$ respectively. Let $\pi$ be a unique simple path connecting $b_1$ and $b_2$ in $\mathcal{D}$. Now we have a number cases depending on the length of $\pi$ and relative positions of $b_1$ and $b_2$ with respect to $\pi$ (see Figure 5).

In each of the cases we use the fact that $v$ must belong to all bags on a unique simple path between $b'_1$ and $b'_2$ (highlighted by bold lines) to conclude that $v$ must also belong to all $r$-extensions of bags on $\pi$. For example, in Case 4, $v$ belongs to all bags between $b'_1$ and $b'_2$ and so to their extensions. But also $v$ belongs to all $r$-extensions of bags between $b_1$ and $b'_2$ because all such bags are within distance $r$ from $b'_2$ and $v$ belongs to $b'_2$. Similarly $v$ belongs to all $r$-extensions of all bags between $b'_1$ and $b_2$. In Case 14 since $v$ belongs to $b'_2$, by the same argument it follows that it belongs to all $r$-extension of all bags between $b_1$ and $b_2$.

Other cases are similar.

Secondly we show that for each atom $S(e)$ from $V(I)$ there is a node in $\mathcal{D}'$ containing $e$. Suppose that $S(e)$ was generated by the view definition $S(x) \leftarrow \phi(x, y)$ for a connected CQ $\phi(x, y)$ with free variables $x$ and quantified variables $y$ under some assignment $\eta$ defined on both $x$ and $y$. As $\phi$ is of radius at most $r$, it should have a variable $z \in x \cup y$ such that all other variables are at distance at most $r$ from $z$ in the Gaifman graph of $\phi(x, y)$. Therefore the range of $\eta$ lies within distance $r$ from $\eta(z)$. Let $b$ be any bag of $\mathcal{D}$ containing $\eta(z)$. If follows that $e$ is contained in the $r$-extension of $b$.

Finally, it is easy to see that the sizes of bags of $\mathcal{D}'$ are as required. \qed

Proof of Proposition 3
Recall the statement:

For any Datalog query $Q = (\Pi, \text{GOAL})$, there is an ExpTime function that outputs an NTA $\mathcal{A}_Q$ that captures the set of canonical databases of CQ approximations of $Q$.

Proof. Without any loss of generality we assume that all rules of $\Pi$ have either 0 or 2 IDB atoms.

The states of $\mathcal{A}_Q$ will be rule heads of $\Pi$ paired with an injective mapping $m$ from the head variables to $\{1, \ldots, k\}$. For example, if our state in node $v$ is $\{P(x, y) \mid \{x \mapsto 1, y \mapsto 3\}\}$ this means that we are looking for witnesses to the fact $P([v, 1], [v, 3])$.

In a state $(U(x), m)$ we non-deterministically choose a rule body with the head $U(x)$ and a consistent extension $m'$ of $m$ to all of the variables in the body. Consistent here means that for every EDB atom $R(y)$ in the body the unary predicate $T_{R_{m(x)}}^y$ is in the label of the current node.
Now let’s turn to intensional predicates. If there are no intensional atoms in the body, we accept. If $F_1(y_1)$ and $F_2(y_2)$ are the intensional atoms in the rule body in some canonical order, then we have a transition which for $i = 1, 2$ goes to the $i$-th child of the current node and switches the state into $(F_i(y_i), m_i)$, where $m_i$ is the restriction of $m$ to $y_i$. We also check that the edge label leading to the $i$-th child is the restriction of the identity map on $\{1, \ldots, k\}$ to the image of $m_i$.

To see that the conditions for capturing hold, note that all CQ approximations of $Q$ have a standard tree decomposition of width $k$, where there is one-to-one correspondence between bags and rule bodies. Therefore we have a standard $k$-code, where variables in each rule body are ordered in such a way that common variables in two adjacent bags occur in exactly same positions. This code is accepted by $A_Q$, which gives the second requirement for capturing. And all codes that are accepted by $A_Q$ are extensions of this standard code, which gives the first required property of capturing.

$\square$

Proof of Proposition 4

Recall the statement
For any Datalog program $\Pi$, the class $\{ \mathcal{F} : \mathcal{F} \models \Pi, \text{tw}(\mathcal{F}) \leq k \}$ (here $\mathcal{F}$ are finite instances which contain both EDBs and IDBs of $\Pi$) is $k$-regular and is recognized by an NTA of at most double-exponential size in $k$ and single-exponential size in $|\Pi|$.

**Proof.** First we construct a two-way alternating tree automaton which, for each of the rules of $\Pi$ of the form $R(x) \leftarrow \phi(x, y)$, guesses non-deterministically moving in both directions the valuations $a$ and $b$ of $x$ and $y$, respectively, and then checks whether $\neg\phi(a, b) \vee R(a)$ holds. Note that its size is linear in $\Pi$ and single-exponential in $k$. Then using Theorem A.2 of [13] we convert it into an NTA with an exponential blow-up. 

**Proof of Proposition 5**

Recall the statement:

If $C$ is a $k$-regular class in $\Sigma$ captured by NTA $\mathcal{A}$ and $\Sigma' \subseteq \Sigma$, then the class $C \mid \Sigma' = \{ \mathcal{F} \mid \Sigma' \mid \mathcal{F} \in C \}$ is also $k$-regular and is captured by an automaton of size at most $|\mathcal{A}|$. The same holds with “capture” replaced by “recognize”.

**Proof.** We prove only the first part, with the second part being similar.

Consider an automaton $\mathcal{A}$ for $C$. It has transitions of the form $q_1, q_2, \sigma_L^{x,y} \rightarrow q$. Let $\mathcal{A}'$ have the same states and accepting $\mathcal{A}$, the alphabet $\sigma_L^{x,y}$ with $L \subseteq \Sigma'$, and the transition table $\{ q_1, q_2, \sigma_L^{x,y} \rightarrow q \mid q_1, q_2, \sigma_L^{x,y} \rightarrow q $ is a transition of $\mathcal{A} \}$. We claim that $\mathcal{A}'$ is of required size and captures $C \mid \Sigma'$. Indeed, take $\mathcal{F}' \in C \mid \Sigma'$. Then there is $\mathcal{F} \in C$ such that $\mathcal{F}' = \mathcal{F} \mid \Sigma'$. As $\mathcal{A}$ captures $C$, there is a code $\mathcal{T}$ of $\mathcal{F}$ such that $\mathcal{A}$ accepts $\mathcal{T}$. From the definition of $\mathcal{A}'$ it follows that $\mathcal{A}'$ accepts $\mathcal{T} \mid \Sigma'$ which is a code of $\mathcal{T}'$. And the other way round, if $\mathcal{A}'$ accepts $\mathcal{T}' \mid \Sigma'$ via a run $f'$, then there is a run $f$ of $\mathcal{A}$ which accepts some extension $\mathcal{T}$ of $\mathcal{T}'$. As $\mathcal{A}$ captures $C$, it follows that there is a database instance $\mathcal{F} \in C$ such that $\mathcal{T}$ is a code of $\mathcal{F}$, and so $\mathcal{T}'$ is a code of $\mathcal{F} \mid \Sigma'$. Thus $\mathcal{A}(\mathcal{T}') \in C \mid \Sigma'$.

**Proof of Lemma 4**

Recall the statement:

For any Datalog query $Q$ and Datalog views $V$, if $Q$ is monotonically determined over $V$ then it is homomorphically determined over $V$.

**Proof.** Assume monotonic determinacy and consider instances $I_1, I_2$ with $I_1 \models Q$, and a homomorphism $h$ from $V(I_1)$ into $V(I_2)$.

It follows that there is a CQ $Q_1$ that is an approximation of $Q$, and a homomorphism $\alpha$ from $Q_1$ into $I_1$. Note that $\alpha$ is also a homomorphism from $V(Q_1)$ into $V(I_1)$. Thus $\alpha$ followed by $h$, denoted $\alpha(h)$, is a homomorphism from $V(Q_1)$ into $V(I_2)$.

We create an instance $I'$ such that $V(Q_1) \subseteq V(I')$, along with a homomorphism $h'$ taking $I'$ into $I_2$. We will construct $I'$ as the union of a set of facts $S_F$ obtained by chasing each fact in $V(Q_1)$ with the inverse of the view definitions (see also the proof of Lemma 5). More precisely, consider a fact $V(a_1 \ldots a_n) \in V(Q_1)$. Then $V(h(a_1)\ldots h(a_n))$ is in $V(I')$, and thus there is some CQ-approximation $\rho$ of $V$ such that $\rho(h(a_1)\ldots h(a_n))$ holds in $I'$. We let $S_F$ be obtained from $\rho(h(a_1)\ldots h(a_n))$ by replacing each $h(a_i)$ with $a_i$ and each existentially quantified variable with a fresh null. One can easily check that $I'$ is as required. Thus by monotonic determinacy we have $I' \models Q$. But since there is a homomorphism of $I'$ into $I_2$, we conclude that $I_2 \models Q$ as required. 

**Jointly-annotated terms**

We recall the definition of the backward mapping query $Q_{\text{opt}} = (\Pi_{\text{opt}}, \text{GOAL}_{\text{opt}})$:

For every transition of the form $q_1, q_2, \sigma_L^{x,y} \rightarrow q$ with $L = (T_{n_1}^R, \ldots, T_{n_m}^R)$ we create a rule

$$P_q(x_1, \ldots, x_k) \leftarrow \bigwedge_{i=1}^k \text{Adom}(x_i) \land P_{q_1}(x_1^i, \ldots, x_k^i) \land P_{q_2}(x_1^2, \ldots, x_k^2) \land \bigwedge_{i \in \text{dom}(s_1)} x_1 = x_1^{s_1(i)} \land \bigwedge_{i \in \text{dom}(s_2)} x_1 = x_1^{s_2(i)} \land \bigwedge_{l=1}^m R_l^R(x_{n_l})$$  (1)

where $j$ ranges over $1$ and $2$, $x_j^i$ are fresh variables for indices $j \in \{1, 2\}$ and $i \in \{1, \ldots, k\}$, and for $n_l = (n_{1,l}, \ldots, n_{k,l})$ we have $x_{n_l} = (x_{n_{1,l}}, \ldots, x_{n_{k,l}})$. For initial transitions of the form $\sigma_L \rightarrow q$ with $L = (T_{n_1}^R, \ldots, T_{n_m}^R)$ we have rules

$$P_q(x_1, \ldots, x_k) \leftarrow \bigwedge_{i=1}^k \text{Adom}(x_i) \land \bigwedge_{l=1}^m R_l^R(x_{n_l}).$$  (2)

For accepting states $q$ we add the rules $\text{GOAL}_{\text{opt}}(x_1, \ldots, x_k) \leftarrow P_q(x_1, \ldots, x_k)$ for the goal predicate $\text{GOAL}_{\text{opt}}$. We also add a standard set of rules which, when evaluated on any data instance $I$ under fixed-point semantics, guarantee that the interpretation of the IDB $\text{Adom}(x)$ is the active domain of $I$. 

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Proof terms and annotated proof terms. To show correctness of the backward mapping construction (Proposition 7) we will need the notion of a "proof certificate" for backward mappings of an automaton.

When a Datalog query $Q = (\Pi, \text{GOAL})$ holds for a tuple $d$ in an instance $I$, there is a derivation that witnesses this, which has a tree-like structure. A proof term for $I \models Q(d)$ is a labelled finite tree in which every node $v$ is labelled with a ground fact $\text{FactOf}(v)$ over the predicates mentioned in $\Pi$, and every non-leaf node $v$ is additionally labelled with a rule $\text{RuleOf}(v)$ of $\Pi$ such that:

- If $v$ is the root, $\text{FactOf}(v) = \text{GOAL}(d)$
- If $v$ is a leaf then $\text{FactOf}(v)$ is a fact over the extensional predicates of $Q$, and this fact holds in $I$
- If $v$ is not a leaf, let $I_v$ be the instance consisting of $\text{FactOf}(v)$ and all facts $\text{FactOf}(c)$ for $c$ a child of $v$. Then there is

  a map $h_v$ from the variables in the body of $\text{RuleOf}(v)$ into the active domain of $I_v$ that maps the facts in the body of $\text{RuleOf}(v)$ onto the facts of $I_v$, and maps the head of $\text{RuleOf}(v)$ to $\text{FactOf}(v)$.

It is well-known [1] and easy to see that proof terms represent a semantics for Datalog: $I \models Q(d)$ exactly when there is a proof term that witnesses this.

We now give a notion of a witness for acceptance of an automaton running over codes. A jointly-annotated term for automaton $\mathcal{A}$, instance $I$, and $k$-tuple $a$ is a pair $(\mathcal{T}, b)$ where

- $\mathcal{T}$ is a tree code accepted by $\mathcal{A}$;
- the map $b$ assigns each vertex of $\mathcal{T}$ to a $k$-tuple of elements from $I$, with the root of $\mathcal{T}$ mapped to $a$;

which satisfy the following condition: if $t_v = \sigma^U_{L}(t_{v_1}, t_{v_2})$ with $L = (R^L_{a_1}, \ldots, R^L_{a_m})$, $b(v) = (b_1, \ldots, b_k)$, $b(v_j) = (b^j_1, \ldots, b^j_k)$ for $j = 1, 2$ then

$$I \models \bigwedge_{i = 1}^{m} b_i = b^j_i(s(i)) \wedge \bigwedge_{j = 1}^{2} R^j(L(b_{n_i})).$$

(3)

We also require that

$$t = \sigma_L \text{ is a leaf symbol in } \mathcal{T} \text{ with } L = (R^L_{a_1}, \ldots, R^L_{a_m}) \text{ then the atoms } R^l(L(b_{n_i})) \text{ are in } I \text{ for } l = 1, \ldots, m.$$ (4)

In other words, $b$ can be considered as a homomorphism from $\mathcal{T}(T)$ into $I$.

We now verify the key property of a jointly-annotated term:

**Proposition 12.** For each data instance $I$, $I, \Pi_{\mathcal{A}} \models \text{GOAL}_{\mathcal{A}}(a)$ if and only if there is a jointly-annotated term for $\mathcal{A}, I$, and $a$.

Proof. We prove the two directions of the if and only if separately.

$(\Rightarrow)$ Take a proof term $t$ that witnesses $I, \Pi_{\mathcal{A}} \models \text{GOAL}_{\mathcal{A}}(a)$. We transform $t$ into a jointly-annotated term $(\mathcal{T}, b)$ on the set of all vertices of $t$ with $\text{FactOf}(v)$ being an IDB. Note that this gives us a binary tree since all rule bodies in $\Pi_{\mathcal{A}}$ have either 0 or 2 IDBs by assumption. For each vertex $v$ we take some ordering $(u^v_1, \ldots, u^v_k)$ of elements in $I_v$ without duplicates; we use fresh dummy elements to fill up the tuple if $I_v$ has less than $k$ elements. Now we define unary labels of $\mathcal{T}$ by setting $TR^R_{a_1, \ldots, a_m}(v) \in \mathcal{T}$ iff $R(u^v_1, \ldots, u^v_k) \in I_v$. We define edge labels $s$ between a parent $v$ and its child $v$ by setting $s(n) = m$ if $u^v_n$ is the same element as $u^v_m$; it should be clear that $s$ is a partial bijection. This constitutes the definition of $\mathcal{T}$. It remains to define $b$ by setting $b(v)$ to be $(u^v_1, \ldots, u^v_k)$. We can create an accepting run $f$ by setting $f(v)$ to be the state of the automaton $q$ such that $\text{FactOf}(v)$ is labelled by and $b$.

$(\Leftarrow)$ It is easy to show by induction on the length of the if and only if separately.

If $v$ is a leaf, then then $I, \Pi_{\mathcal{A}} \models \text{GOAL}_{\mathcal{A}}(a)$ by the rule (2) because its body holds due to condition (4) and the fact that for all $u$ in $b(v)$ we have $I_v, \Pi_{\mathcal{A}} \models \text{Adom}(u)$.

Indeed, if $u$ is a leaf, then then $I_v, \Pi_{\mathcal{A}} \models \text{GOAL}_{\mathcal{A}}(a)$ by the rule (2) because its body holds due to condition (4) and the fact that for all $u$ in $b(v)$ we have $I_v, \Pi_{\mathcal{A}} \models \text{Adom}(u)$.

If $v$ has children $v_1$ and $v_2$, then there must be $q_1$ and $q_2$ such that $f(v_1) = q_1, f(v_2) = q_2$, production $(q_1, q_2, \sigma_L \rightarrow g)$ is a transition of $\mathcal{A}$, and the vertex label of $v$ is $L$ while edge labels between $v, v_1$ and $v_2$ are $s_1$ and $s_2$.

We claim that $I_v, \Pi_{\mathcal{A}} \models \text{GOAL}(a)$ can be inferred by the rule (1) for this production under assignment $\{(x_1, \ldots, x_k) : b(v), (x^v_1, \ldots, x^v_k) : b(v_1), (x^v_1, \ldots, x^v_k) : b(v_2)\}$. Indeed, we have $I_v, \Pi_{\mathcal{A}} \models \text{Adom}(u)$ for all elements in the body of the rule, we have $I_v, \Pi_{\mathcal{A}} \models \text{GOAL}(a)$ by the induction hypothesis, and the rest of the rule by (3). 

$\square$

**Proof of Proposition 7**

Recall the statement:

Let $Q$ be homomorphically determined over $V$ and $\mathcal{A}$ be any automaton working on $k$-codes such that $\{V(Q_i) \mid i \in \omega\} \subseteq \mathcal{T}(L(\mathcal{A})) \subseteq \{\mathcal{D} \mid V(Q_i) \text{ maps into } \mathcal{D} \text{ for some } i \in \omega\}$. More precisely, we require that

1. for each CQ approximation $Q_i$ of $Q$ there is a code $\mathcal{T}$ such that $\mathcal{T}(\mathcal{T}) = V(Q_i)$ and $\mathcal{T}$ is accepted by $\mathcal{A}$;
2. for each tree code $\mathcal{T}$ accepted by $\mathcal{A}$ there is a CQ approximation $Q_i$ of $Q$ and a homomorphism from $V(Q_i)$ into $\mathcal{T}(\mathcal{T})$.
Then for each data instance \( I \) we have \( I \models Q \) iff \( V(I) \models Q_\mathbb{R}(a) \) for some \( a \in \text{ADOM}(I)^k \).

**Proof.** Suppose \( \Pi \) is a Datalog program containing intensional predicate \( A \), \( I \) is an instance for the extensional (input) signature of \( \Pi \), and \( a \) is a tuple of elements from \( I \). Below we write \( I, \Pi \models A(a) \) to indicate that the least fixpoint of \( \Pi \) on \( I \) contains \( A(a) \).

\( \Rightarrow \) Suppose that \( I \models Q \). Then there is an approximation \( Q_i \) of \( Q \) and a homomorphism \( h \) from \( Q_i \) into \( I \), which is also a homomorphism from \( V(Q_i) \) to \( V(I) \). As \( Q_i \) is an approximation of \( Q \), by the first inclusion for \( L(\mathbb{R}) \), \( \mathbb{R} \) must accept some code \( T \) of \( V(Q_i) \). Choose an arbitrary element \( e_0 \) from \( \text{ADOM}(V(I)) \).

For a vertex \( v \) of \( T \) we define \( b(v) \) to be the tuple \((e_1, \ldots, e_k)\) of elements of \( V(I) \) where each \( e_i \) is defined as follows:

\[
e_i = \begin{cases} h([v, i]), & \text{if } [v, i] \in \text{ADOM}(I) \\ e_0, & \text{otherwise}. \end{cases}
\]

We claim that \((T, b)\) is a jointly-annotated term for \( \mathbb{R} \), \( V(I) \) and the \( b \)-image of the root of \( T \). Indeed, if equation (3) contains an equality \([v, i] = [u, j]\), it follows that \([v, i]\) and \([u, j]\) are indeed equivalent. The R-atoms of equations (3) and (4) hold because \( h \) is a homomorphism, and also because they are never applied to dummies. It follows (by Proposition 12) that \( V(I) \models Q_\mathbb{R}(a) \) for some \( a \).

\( \Leftarrow \) Suppose that \( V(I), \Pi_\mathbb{R} \models \text{GOAL}_\mathbb{R}(a) \). Let \((T, b)\) be a jointly-annotated term for the inference of \( \text{GOAL}_\mathbb{R}(a) \) for \( V(I), \Pi_\mathbb{R} \) and \( a \) (which exists by Proposition 12), and \( f \) be an accepting run of \( \mathbb{R} \) on \( T \). Thus, by the second inclusion for \( L(\mathbb{R}) \), there must be a homomorphism \( h \) from \( V(Q_i) \) for some \( i \) into \( \Sigma(T) \). Note that by Proposition 12, we know that \( b \) can be considered as a homomorphism from \( \Sigma(T) \) into \( V(I) \). By composing \( h \) with \( b \), we obtain a homomorphism \( g \) from \( V(Q_i) \) into \( V(I) \). Now we have a data instance \( I' = Q_i \) such that \( I' \models Q \) and a homomorphism \( g \) from \( V(I') \) into \( V(I) \). Therefore, as \( Q \) is homomorphically determined by \( V \), we have \( I \models Q \). \( \square \)
PROOFS FOR SECTION 4: REWRITABILITY RESULTS

Proof of the last part of Theorem 1

Recall that Theorem 1 stated that if $Q$ is in MDL, $V$ are a collection of FGDV views, and $Q$ monotonically determined by $V$ then $Q$ has a rewriting in MDL. We sketch how to modify the prior argument for this claim.

A tree decomposition is frontier-one if the intersection of any two neighboring bags has at most one element. It is clear that approximations of MDL queries have such decompositions, provide that we now allow decompositions that have arbitrary outdegree, not necessarily binary. We can further normalize so that in each bag other than the root, the element that is shared with its parent (if such exists) has the first local name in the code.

When we apply frontier-guarded views, we annotate the bags of the tree decomposition with view predicates, but we do not change the intersection of neighboring bags. And when we project such a decomposition onto the view predicates, we do not change this intersection either. Thus in the proof of Theorem 1, we can consider an automaton $\mathfrak{A}$ that enforces the frontier-one restriction.

We can modify the backward mapping for frontier-one decompositions so that it produces an MDL query; our modification will have only unary intensional predicates $P_q$ for each state $q$ of the automaton, corresponding only to the element coded in the frontier.

More formally, for every transition of the form $q_1 \ldots q_r \sigma_{L_1}^{s_1} \ldots \sigma_{L_m}^{s_m} \rightarrow q$ with $L = [\tau_1^{R_1}, \ldots, \tau_m^{R_m}]$, we know that for the $i^{th}$ child node, the label $L$ contains at most one equality of a local name $n_i$ with the first local name of the child.

We create a rule of the form:

$$P_q(x_1) = \bigwedge_{i=1}^{k} \text{Adom}(x_i) \land \bigwedge_{i=1}^{r} P_q(x_{n_i}) \land \bigwedge_{i=1}^{m} R^i(x_{n_i})$$

Similar modifications are applied to the leaf rules.

Proof of Theorem 2

Recall the statement:

Suppose $Q$ is a normalized Monadic Datalog query and $V$ is a collection of Monadic Datalog and CQ views. If $Q$ is monotonically determined by $V$, then $Q$ is rewritable over $V$ in Datalog. The size of the rewriting is at most double-exponential in $K = O(|Q||V|)$ ("of required size" below).

Proof. We first argue that without any loss of generality we can assume that all CQ views are connected. If $V$ is a CQ view which is not connected then it can be replaced by a few connected CQs. For example, the disconnected view $V(x, y) = Q_1(x) \land Q_2(y)$ can be replaced by the free-variable-connected views $V_1(x) = Q_1(x) \land \exists y. Q_2(y)$ and $V_2(y) = (\exists x. Q_1(x)) \land Q_2(y)$. Indeed, given $V$, we can restore $V_1$ and $V_2$ as its projections on $x$ and $y$ respectively. And the other way round, given $V_1$ and $V_2$, we can restore $V$ as their product since $V_1(x) \land V_2(y) = Q_1(x) \land (\exists y. Q_2(y)) \land (\exists x. Q_1(x)) \land Q_2(y)$ is equivalent in first-order logic to $Q_1(x) \land Q_2(y) = V(x, y)$.

We need to show that there is an automaton $\mathfrak{A}$ such that

$$\{V(Q_i) \mid i \in \omega\} \subseteq \mathfrak{T}_I(L(\mathfrak{A})) \subseteq \{\mathcal{F} \mid V(Q_i) \text{ maps into } \mathcal{F} \text{ for some } i \in \omega\}$$

Consider the class $\mathcal{C}$ of canonical databases of CQ approximations of $Q$. By Lemma 3 (applied with the maximal radius $r$ of the CQ views in $V$ where $r = O(|V|)$), the treewidth of the class $\mathcal{C} = \{V(F) \mid F \in \mathcal{C}\}$ of view images of $C$ is also bounded by some $K = O(|Q||V|)$. We can strengthen Proposition 3 to show that for any treewidth $K$ greater than or equal to the maximal number of variables in the rules of $Q$, the class $\mathcal{C}$ of approximations is $K$-regular and there is an NTA $\mathfrak{A}_{\text{base}}$ of at most exponential size in $K$ that captures $\mathcal{C}$.

Without any loss of generality we assume that the sets of IDBs of programs for different views are disjoint, and that their goal predicates are identical with the view predicates. Denote by $\Pi_V$ the union of all rules in Datalog queries in $V$, including the rules for the CQ views. By Proposition 4, there is an NTA $\mathfrak{A}_{\Pi_V}$ of required size which recognizes all codes of $\{F \mid F \models \Pi_V, \text{tw}(F) \leq K\}$.

We claim that the automaton $\mathfrak{A}' = \mathfrak{A}_{\text{base}} \cap \Pi_{\Pi_V}$ satisfies

$$\{\text{FPEval}(\Pi_V, Q_i) \mid i \in \omega\} \subseteq \mathfrak{T}_I(L(\mathfrak{A}')) \subseteq \{\mathcal{F} \mid F \models \Pi_V\}$$

Now the automaton $\mathfrak{A}$ that is the projection of $\mathfrak{A}'$ on the signature of view predicates (which exists by Proposition 5) captures $V = \{F \mid \Sigma_V \land F \land \Sigma_B \in \mathcal{C}, \text{tw}(F) \leq K, F \models \Pi_V\}$ and so satisfies two conditions of Proposition 7. Now applying Proposition 7, we conclude that $Q$ is Datalog rewritable over views, and that the rewriting is of required size.

Another observation will be useful later (see proof of Theorem 4) is that $\mathfrak{A}$ captures $V = \{F \mid \Sigma_V \land F \land \Sigma_B \in \mathcal{C}, \text{tw}(F) \leq K, F \models \Pi_V\}$.

Now applying Proposition 7, we conclude that $Q$ is Datalog rewritable over views, and that the rewriting is of required size. \[\square\]
PROOFS FOR SECTION 5: DECIDABILITY RESULTS ON MONOTONIC DETERMINACY

Proof of Lemma 5

Recall the statement:

\( Q \) is monotonically determined over \( V \) if and only if every test succeeds.

We first need a bit of infrastructure. When a Datalog query \( Q = (I, \text{Goal}) \) holds for a tuple \( d \) in an instance \( I \), there is a derivation that witnesses this, which has a tree-like structure. A proof term for \( I \models Q(d) \) is a labelled finite tree in which every node \( v \) is labelled with a ground fact \( \text{FactOf}(v) \) over the predicates mentioned in \( I \), and every non-leaf node \( v \) is additionally labelled with a rule \( \text{RuleOf}(v) \) of \( I \) such that:

- If \( v \) is the root, \( \text{FactOf}(v) = \text{Goal}(d) \)
- If \( v \) is a leaf then \( \text{FactOf}(v) \) is a fact over the extensional predicates of \( Q \), and this fact holds in \( I \)
- If \( v \) is not a leaf, let \( I_v \) be the instance consisting of \( \text{FactOf}(v) \) and all facts \( \text{FactOf}(c) \) for \( c \) a child of \( v \). Then there is a map \( h_v \) from the variables in the body of \( \text{RuleOf}(v) \) into the active domain of \( I_v \) that maps the facts in the body of \( \text{RuleOf}(v) \) onto the facts of \( I_v \), and maps the head of \( \text{RuleOf}(v) \) to \( \text{FactOf}(v) \).

It is well-known [1] and easy to see that proof terms represent a semantics for Datalog: \( I \models Q(d) \) exactly when there is a proof term that witnesses this.

We are now ready for the proof of the lemma.

Proof. We assume \( Q \) is Boolean for simplicity. In one direction, assume \( Q \) is monotonically determined over \( V \), and consider a test \( (Q_1, D') \). By virtue of \( (Q_1, D') \) being a test, we have \( V(D) \subseteq V(D') \). Monotonic determinacy and \( Q_1 \models Q \) thus imply that \( D' \models Q \).

In the other direction, assume every test succeeds, and consider instance \( I_1 \) and \( I_2 \) with \( I_1 \) satisfying \( Q \) and \( V(I_1) \subseteq V(I_2) \). As \( I_1 \models Q \), there is a homomorphism \( \alpha \) from some \( Q_1 \) into \( I_1 \). Since the views are preserved under homomorphism, \( \alpha \) is also a homomorphism from \( V(Q_1) \) into \( V(I_1) \).

We will now create a \( D' \) such that \( (Q_1, D') \) forms a test, along with an extension of \( \alpha \) that is a homomorphism taking \( D' \) into \( I_2 \). \( D' \) will be the union of a set of facts \( S_F \) (defined below) for every fact \( F \) from \( V(Q_1) \). For a fact \( F = V(c) \) from \( V(Q_1) \) let \( F' = \alpha(F) \). Note that \( F' \) is in \( V(I_1) \). By assumption, \( F' \) is also in \( V(I_2) \). Thus there is a proof term \( \tau_F \) witnessing that \( I_2 \models F' \). Moving top-down on \( \tau_F \), we form a proof term for \( F \). The root of the term \( \tau_F \) is labelled with the fact \( \text{Goal}_V(\alpha(c_1) \ldots \alpha(c_n)) \) for the goal predicate \( \text{Goal}_V \) of the Datalog program \( Q_V \). Since \( \alpha \) is not injective, \( c_i \) may not be unique, but we choose one such tuple \( c_1 \ldots c_n \) and fix it for the transformation of \( \tau_F \). This choice will impact the proof term that we create, but will not impact the homomorphism extending \( \alpha \). We first transform \( \tau_F \) by replacing any element \( \alpha(c_i) \) occurring in \( \tau_F \) by \( c_i \). We then continue our transformation by proceeding top-down on the partially-transformed term. At the root of the term we do nothing more. In the inductive step, we consider an intensional fact \( U(d) \) in \( \tau_F \) witnessed by a set of facts \( J \) that are a substitution instance of some rule body \( B \). In \( J \), we uniformly replace any witness \( w \) to an existentially quantified variable \( x \) of \( B \) by a fresh element \( d_w \), and extend the homomorphism to take \( d_w \) to \( w \). We set \( S_F \) to be the union of all EDB facts occurring in the proof term we have constructed for \( F \).

It is easy to see that the union of the facts \( S_F \) forms an appropriate \( D' \) giving a test. By assumption this test succeeds, so \( D' \models Q \). But since \( D' \) is homomorphically embedded into \( I_2 \), this means that \( I_2 \models Q \) as required. \( \square \)

Proof of Theorem 5

Recall the statement:

If \( Q \) is a CQ and \( V \) is a collection of Datalog views, then the problem of monotonic determinacy of \( Q \) over \( V \) is decidable in \( 2\expTime \).

Proof. Let \( Q' = V(Q) \) and let \( Q' \) inherit all answer variables \( x \) from \( Q \). Let \( Q'' = (I, \text{Goal}) \) where \( I \) is obtained by taking all rules defining \( V \) and adding the rule \( \text{Goal}(x) \leftrightarrow Q' \).

It is easy to see that the following statements are equivalent:

1. \( Q \) is monotonically determined by \( V \);
2. \( Q' \) is a CQ rewriting of \( Q \) in terms of \( V \);
3. for all \( I, I \models Q \) iff \( V(I) \models Q' \);
4. \( Q'' \) is equivalent to \( Q \);
5. \( Q'' \) is contained in \( Q \).

Indeed, (1) implies (2) by the proof of Proposition 8, and all other implications between adjacent statements are trivial. It remains to note that the containment (5) can be decided in \( 2\expTime \) by Theorem 5.12 of [12]. \( \square \)
Proof of Theorem 4
Recall the statement:

Suppose $Q$ is in Monadic Datalog, and $V$ is a collection of CQ and Frontier-guarded Datalog views. Then there is an algorithm that decides if $Q$ is monotonically determined by $V$ in $3\text{ExpTime}$.

Proof. In this proof the words “of required size” mean “doubly-exponential in $K$” where $K$ is some integer defined below, $\mathcal{C}$ stands for the class of all CQ-approximations of $Q$, $\Sigma_V$ is the view signature and $\Sigma_B$ is the initial signature.

As in the proof of Theorem 2, we we can assume that all CQ views are connected.

We have to check whether $Q$ holds on all tests. As observed in the proof of Theorem 2, there is an integer $K = O(|Q|^{|V|})$ such that both the treewidth of the view images of CQ-approximations of $Q$ and the treewidth of the CQ-approximations of the views in $V$ are at most $K$. Let $\mathcal{V} = \{F \mid \Sigma_V \mid F \mid \Sigma_B \in \mathcal{C}, \text{tw}(F) \leq K, F \models \Pi_V\}$. As argued in the proof of Theorem 2, $\mathcal{V}$ is $K$-regular and captured by an NTA $A_V$ of required size.

We follow the template of Theorem 2 We will check the equivalent condition that $Q$ holds on each element of the class $\text{ETEST}(Q, V)$, which consists of all instances $D'$ which can be obtained from an instance in $V$ by applying inverses of view definitions while keeping the atoms of the view signature. Note that the treewidth of all database instances in $\text{ETEST}(Q, V)$ is also bounded by $K$. By Proposition 3, for each view $(V, Q_V)$ there exists an automaton $A'_V$ which for each atom $V(c)$ at a node $n$ in $T$ checks whether $n$ has a descendant $n'$ such that $n'$ contains $c$ and the subcode of $T$ rooted at $n'$ is a code of some CQ approximation of $Q_V$. The automaton $A'_{\text{ETEST}}$ defined as the product of $A_V$ and $A'_V$ for all views $V$ in $V$ (thus accepting the intersection of these languages) captures $\text{ETEST}(Q, V)$.

By Proposition 6, there is an NTA $A''$ of required size which recognizes those codes which do not satisfy $Q$. So to check if $Q$ is monotonically determined by $V$ we construct the intersection of $A'_{\text{ETEST}}$ and $A''$ (which is of required size) and check if it is empty. The latter check is linear in the size of the automaton. It should be clear that the time complexity of this procedure is doubly exponential in $K$, and so triply exponential in the size of the input. □
PROOFS FOR SECTION 6: LOWER BOUNDS ON MONOTONIC DETERMINACY

Proof of Proposition 9

Recall the statement:

Monotonic determinacy is

- NP-hard for CQ queries and views
- $\Pi_2^p$-hard for UCQ queries and UCQ views
- 2ExpTime-hard for CQ queries and MDL views
- 2ExpTime-hard for MDL queries and a fixed atomic view
- undecidable for Datalog queries and a fixed atomic view

The first three bullet items will follow from a reduction from Datalog equivalence:

**Lemma 7.** Let $Q$ and $Q_V$ be arbitrary Datalog queries. Then $Q$ is monotonically determined by $V = \{ (V, Q_V) \}$ iff $Q$ and $Q_V$ are equivalent.

**Proof.** Let $Q = \bigvee_{i=0}^n Q_i^Q$ and $Q_V = \bigvee_{j=0}^\beta Q_j^Q$ with non-empty $Q_i^Q$ and $Q_j^Q$.

First we show that each $Q_i^Q$ satisfies $Q_V$. Indeed, if $Q_V$ is not true on some $Q_i$, then there is a test built on $Q_i$ with no atoms. Clearly this test does not satisfy $Q$.

Then we show that each $Q_j^Q$ satisfies $Q$. Fix some CQ-approximation $Q_0^Q$ of $Q$. We claim that $(Q_0^Q, Q_j^Q)$ is a test for $Q$ and $V$ for any $j \in \alpha$. Indeed, $Q_V$ evaluates to true on $Q_0^Q$, and then $V = 1$ during the inverse step can be replaced by any $Q_j^Q$. Thus $Q_j^Q$ must satisfy $Q$.

The first bullet item now follows from the NP-hardness of equivalence of CQs; the second item follows from the $\Pi_2^p$ hardness of equivalence for UCQs [24], while the third follows from the 2ExpTime-hardness of a CQ and an MDL query [7].

The results for fixed views follow from a reduction found in [14]:

**Lemma 8.** Let $Q_1$ and $Q_2$ be arbitrary Datalog queries. Consider the query $Q = Q_1 \land e \lor Q_2$ where $e$ is a fresh extensional predicate of arity 0 and a set of views $V$ which has views $P'$ for all extensional relations $P$ occurring in $Q$ except $e$. Then $Q_1$ is contained in $Q_2$ iff $Q$ is monotonically determined by $V$.

**Proof.** ($\Rightarrow$) Note that the tests for $Q$ and $V$ consist of all CQ-approximations of $Q_1$ and $Q_2$. It follows that if $Q_1$ is contained in $Q_2$, then all tests pass.

($\Leftarrow$) We assume monotonic determinacy and show that $Q_1$ is contained in $Q_2$. Pick some CQ-approximation $Q_1^Q$ of $Q_1$. Then it’s easy to see that $(Q_1^Q, e, Q_1^Q)$ is a test for $Q$ and $V$. By monotonic determinacy it follows that $Q_1^Q \models Q$, and so either $Q_1^Q \models Q_1 \land e$ or $Q_1^Q \models Q_2$. The first option is impossible because $Q_1^Q$ contains no $e$-atoms. Therefore, $Q_1^Q \models Q_2$. Thus $Q_1$ is contained in $Q_2$.

The second to the last item now follows from [13] and the last item from [25], noting that the lower bounds only require a single extensional predicate.

Proof of Proposition 10

Recall the statement:

$Q_{TP}$ is not monotonically determined by $V_{TP}$ iff TP has a solution.

We recall the definition of the query and views, giving names to the special views.

1. $Q_{\text{START}} \leftarrow A(x), B(x)$
2. $Q_{\text{Succ}} \leftarrow XSucc(x, x'), A(x'), C(x')$
3. $Q_{\text{End}} \leftarrow XEnd(x)$
4. $Q_{\text{Succ}} \leftarrow YSucc(y, y'), B(y'), D(y')$
5. $Q_{\text{End}} \leftarrow YEnd(y)$
6. $Q_{\text{Helper}} \leftarrow C(u), YProj(y, z), XProj(x, z)$
7. $Q_{\text{Helper}} \leftarrow D(u), YProj(y, z), XProj(x, z)$
8. $Q_{\text{Verify}} \leftarrow HA(z_1, z_2, y, x_1, x_2), T_i(z_1), T_j(z_2)$ for all pairs $(T_i, T_j) \notin HC$
9. $Q_{\text{Verify}} \leftarrow VA(z_1, z_2, y_1, y_2, x), T_i(z_1), T_j(z_2)$ for all pairs $(T_i, T_j) \notin VC$
10. $Q_{\text{Verify}} \leftarrow YSucc(o, y), YSucc(y, z), XSucc(o, x), XProj(x, z), T_i(z)$ for all $T_i \notin IT$

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(11) \( Q_{\text{VERIFY}} \leftarrow \text{YEND}(y), \text{YProj}(y, z), T_i(z), \text{XProj}(x, z), \text{XEND}(x) \)
for all \( T_i \notin FT \)

The set of views \( V_{FP} \) consists of
- the grid-generating view
  \[
  S(x, y) \leftarrow C(x), D(y)
  \]
  \[
  S(x, y) \leftarrow \text{XProj}(x, z), T_i(z), \text{YProj}(y, z) \text{ for all } T_i \text{ in } Tiles;
  \]
- the atomic views \( V_{\text{Succ}}, V_x, V_{\text{YEND}}, V_{\text{XEND}} \) and \( V_T \) for EDBs \( \text{YSucc}, \text{XSucc}, \text{YEND}, \text{XEND} \) and each \( T_i \text{ in } Tiles; \)
- the following special views
  \[
  (SP1) \quad V_{\text{Helper}}(u, x, y, z) \leftarrow C(u), \text{XProj}(x, z), \text{YProj}(y, z)
  \]
  \[
  (SP2) \quad V_{\text{Helper}}(u, x, y, z) \leftarrow D(u), \text{XProj}(x, z), \text{YProj}(y, z)
  \]
  \[
  (SP3) \quad V_{\text{HA}}(x_1, x_2, y_1, y_2) \leftarrow \text{HA}(x_1, x_2, y_1, y_2)
  \]
  \[
  (SP4) \quad V_{\text{VA}}(x_1, x_2, y_1, y_2) \leftarrow \text{VA}(x_1, x_2, y_1, y_2)
  \]
  \[
  (SP5) \quad V_{\text{F}}(x, y, z) \leftarrow \text{XProj}(x, z), \text{YProj}(y, z)
  \]
  \[
  (SP6) \quad V_{\text{F}}(x, y, z) \leftarrow \text{XProj}(x, z), \text{XEND}(x), \text{YEND}(y), \text{YProj}(y, z).
  \]

We are now ready to begin the proof of Proposition 10.

**Proof.** Suppose that \( T = (Q_i, I') \) is a test for \( Q_{FP} \) and \( V_{FP} \). Following Gogacz and Marcinkowski [17], we call \( Q_i \) the Green instance and \( I' \) the Red instance of the test. We say that \( T = (Q_i, I') \) is a main test if its Green instance is generated from the \( Q_{\text{START}} \)-atom. Otherwise \( T \) is said to be a side test. Note that due to the choice of special and atomic views, all side tests always pass. Also note that all special views are empty when applied to an approximation of a \( Q_{\text{START}} \)-atom (see Figure 2, (a)).

(⇒) Suppose that \( Q_{FP} \) is not monotonically determined by \( V_{FP} \). Then there exists a test \( T = (Q_i, I') \) for \( Q_{FP} \) and \( V_{FP} \) that fails \( Q_{FP} \). Note that \( T \) can’t be a side test. Therefore \( T \) must be a main test. Note that there are three kinds of main tests (see Figure 2; all tests are obtained from (b) by non-deterministic replacement of the \( S \)-atoms by their definitions):

1) a test in which the second rule of the \( S \)-view never fires. In this case the Red instance contains the same \( C \) and \( D \)-atoms as in the Green instance, and hence \( Q_{\text{START}} \) must hold.

2) a test in which both rules of the \( S \)-view fire at least once. In this case, the Red instance will contain both \( C \)-facts, \( D \)-facts, and also some \( \text{XProj} \)-facts that joins with some \( \text{YProj} \)-fact, and thus using SP1-SP2 and \( Q_{\text{Helper}} \) we see that \( Q \) will hold on the Red instance

3) a test in the second rule of the \( S \)-view which fires at least once, but the first rule never fires. In this case the Red instance is isomorphic to a grid from the picture with some \( T_i \)-predicate at each point of the grid.

We claim that these \( T_i \)-predicates give rise to a correct tiling \( \tau \). Indeed, as 8) and 9) do not set \( Q_{\text{VERIFY}} \) to True on \( I' \), \( \tau \) must respect horizontal and vertical compatibility constraints. Similarly, due to rules 10) and 11), \( \tau \) should have a tile from \( IT \) at (1, 1) and from \( FT \) at \( (n, m) \).

(⇐) Suppose that there are integers \( m \) and \( n \) and a tiling of the \( n \times m \) grid with a tile from \( IT \) at (1, 1) and from \( FT \) at \( (n, m) \). Then this tiling (when placed on the \( n \times m \) grid in Figure 1) is \( I' \) for some grid test of monotonically determinacy. Thus \( Q_{FP} \) is not monotonically determined by \( V_{FP} \).

\( \square \)
PROOFS FOR SECTION 7: NON-REWRITABILITY RESULTS

Proof of Fact 4
Recall the statement:

Let $k \geq 2$. Let $I$ be an instance and $U$ be any $k$-unravelling of $I$. Then the following hold:
1. $U \rightarrow I$ and $I \rightarrow_k U$.
2. For every instance $I'$, we have $I \rightarrow_k I'$ iff $U \rightarrow I'$.

For the first part, $U \rightarrow I$ by definition. To see $I \rightarrow_k U$, we form a strategy for the duplicator inductively, preserving the invariant that the pebbles of the duplicator are contained in a single bag of the tree decomposition. The induction step is accomplished using the second property of an unravelling.

We turn to the second part, fixing $I'$. If $U \rightarrow I'$ via some homomorphism $h$, we can apply $h$ to the strategy witnessing $I \rightarrow_k U$ to see $I \rightarrow_k I'$. Conversely, suppose $I \rightarrow_k I'$. Given $u \in U$ we know there is some some bag of the tree decomposition containing $u$ with at most $k$ elements, and $\Theta$ is a partial isomorphism on this bag. Consider a play for Spoiler in the pebble game from $I'$ to $I$ going down the branch of the tree decomposition to $u$. In this play, once Spoiler moves a pebble off of an element, he will never move back on to the element. Let $h(u)$ be the element in $I'$ corresponding to $u$ in the response of the duplicator playing according to his winning strategy witnessing $I \rightarrow_k I'$. One can verify that $h(u)$ is a homomorphism.

Proof of Lemma 6
Recall the statement:

There is a tiling instance $TP^*$ such that $I_{n,m}^{grid}$ can not be tiled with $TP^*$ for each $n, m \geq 1$ but for each $n, m \geq 3$ and each $k$ with $2 \leq k < \min(n, m)$ any $k$-unravelling of $I_{n,m}^{grid}$ can be tiled with $TP^*$.

We can rephrase a tiling problem as a homomorphism problem. For a tiling problem $TP = (Tiles, HC, VC, IT, FT)$, we denote by $I_{TP}$ the database instance over $\delta = [H, V, I, F]$ with domain $Tiles$ and facts $h(T(T'))$ (resp. $v(T(T'))$) for every $(T, T') \in HC$ (resp. $(T, T') \in VC)$, and $I(T)$ (resp. $F(T)$) for every $T \in IT$ (resp. $T \in FT$). Then an instance can be tiled according to $TP$ exactly when it has a homomorphism to $I_{TP}$. We can thus rephrase the lemma as:

There is a tiling problem $TP^*$ such that $I_{n,m}^{grid} \not\rightarrow I_{TP^*}$ for each $n, m \geq 1$, but $I_{n,m}^{grid} \rightarrow_k I_{TP^*}$ for each $n, m \geq 3$ and each $k$ with $2 \leq k \leq \min(n, m)$.

Before going into the proof, we state a well-known characterization of winning strategies for the Duplicator in the existential pebble game:

FACT 5. Let $k \geq 2$ and let $I, I'$ be two instances over the same schema. The Duplicator has a winning strategy in the existential $k$-pebble game on $I$ and $I'$ if and only if there is a non-empty collection $\mathcal{H}$ of partial homomorphisms from $I$ to $I'$ with domain size $\leq k$ such that: (1) if $f \in \mathcal{H}$ and $g \subseteq f$, then $g \in \mathcal{H}$, and (2) for each $f \in \mathcal{H}$ with domain size $\leq k$ and each $a \in \text{dom}(I)$, there is $g \in \mathcal{H}$ with $f \subseteq g$ whose domain contains $a$.

Proof. Our proof is an adaptation of a construction from [4]. It was shown in [4] that if an instance $I$ has a core of treewidth strictly bigger than $k$ with $k \geq 2$, then there exists an instance $I^*$ such that $I \rightarrow I^*$ and $I \rightarrow_k I^*$. We could apply this result to each $I_{n,m}^{grid}$, where $n, m \geq 3$, and obtain $I_{n,m}^{grid}$ such that $I_{n,m}^{grid} \not\rightarrow I_{TP^*}$ and $I_{n,m}^{grid} \rightarrow_k I_{TP^*}$, for $2 \leq k \leq \min(n, m)$. By adapting the arguments in [4], we show that the family $(I_{n,m}^{grid})_{n,m \geq 3}$ can actually be collapsed into a single instance $I_{TP^*}$ with the desired properties.

For $n, m \geq 1$, let $G_{n,m}$ be the $(n \times m)$-grid graph. That is, $\text{vertices}(G_{n,m}) := \{(i, j) : 1 \leq i \leq n, 1 \leq j \leq m\}$ and $\text{edges}(G_{n,m}) := \{(i, j), (i', j') : |i - i'| + |j - j'| = 1\}$. Observe that $G_{n,m}$ is precisely the Gaifman graph of the database instance $I_{n,m}^{grid}$. Intuitively, a solution for our tiling problem on $G_{n,m}$ will describe a 0/1 assignment to the edges of the grid $G_{n,m}$. In order to define $TP^*$, we consider the grid $G_{3,3}$. Intuitively, we want to think of grid points within $G_{3,3}$ as "grid point types" that can be assigned to a grid point in some larger grid $G_{n,m}$. For example, the tile $(2, 1)$ that lies in the center of the lower border represents the type of all elements that lie on the lower border of $G_{n,m}$, excluding the corner points. Our tiles will enhance each abstract grid point with a 0/1 assignment to its incident edges.

For each vertex $u \in \text{vertices}(G_{3,3})$, we denote by $d_u$ the degree of $u$ (note that $d_u \leq 4$) and fix an enumeration $e_{d_u}^1, \ldots, e_{d_u}^d$ of all the edges in $G_{3,3}$ that are incident to $u$. The set of tiles $\text{Tiles}^*$ of $TP^*$ contains all the tuples $(u, b_1, \ldots, b_{d_u})$ such that:
1. $u \in \text{vertices}(G_{3,3})$ and $b_1, \ldots, b_{d_u} \in \{0, 1\}$.
2. $b_1 + \cdots + b_{d_u} \equiv 0 \pmod{2}$ if $u \neq (1, 1)$.
3. $b_1 + \cdots + b_{d_u} \equiv 1 \pmod{2}$ if $u = (1, 1)$.

That is, we consider assignments where the number of edges set to 1 is odd for the left-lower point but the number of edges set to 1 is even elsewhere.

Let us denote $\pi_1 : \text{Tiles}^* \rightarrow \text{vertices}(G_{3,3})$ the first-coordinate projection. We define the set of initial and final tiles to be $IT^* := \{t \in \text{Tiles}^* : \pi_1(t) = (1, 1)\}$ and $FT^* := \{t \in \text{Tiles}^* : \pi_1(t) = (3, 3)\}$, respectively.
Our compatibility relation will ensure that the 0/1 assignment to incident edges is consistent among adjacent nodes: if a grid point \( n \) has the outgoing edge to its right set to \( b \in \{0, 1\} \) and \( n' \) is the neighbor of \( n \) to the right, then \( n' \) has the incoming edge to its left set to \( b \).

We first give the constraints for pairs of grid points that are assigned to distinct abstract grid points in \( G_{3,3} \). For each edge \( e = [u, v] = \{(i, j), (i + 1, j)\} \in \text{edges}(G_{3,3}) \) with \( 1 \leq i < 3 \) and \( 1 \leq j < 3 \), we add to the horizontal compatibility relation \( HC^e \) the pair \((e, b_1, \ldots, b_{du}), (v, b'_1, \ldots, b'_{du})\) iff \( e = e^u = e^v, \) for some \( \ell, m \) and \( b_\ell = b'_m \). Similarly, for each edge \( e = [u, v] = \{(i, j), (i, j + 1)\} \in \text{edges}(G_{3,3}) \) with \( 1 \leq i \leq 3 \) and \( 1 \leq j < 3 \), we add to the vertical compatibility relation \( VC^e \) the pair \((e, b_1, \ldots, b_{du}), (v, b'_1, \ldots, b'_{du})\) iff \( e = e^u = e^v, \) for some \( \ell, m, \) and \( b_\ell = b'_m \).

We now give the consistency restrictions for pairs of grid points that are assigned the same abstract grid point. We add the following pairs to \( HC^e \) and \( VC^e \):

- For \( u = (2, j) \) with \( j \in \{1, 3\} \), the pair \((e, b_1, \ldots, b_{du}), (u, b'_1, \ldots, b'_{du})\) \( \in HC^e \) iff \( e = \{(2, j), (3, j)\}, e' = \{(1, j), (2, j)\}, e = e^u, e' = e^v, \) for some \( \ell, m, \) and \( b_\ell = b'_m \).
- For \( u = (i, 2) \) with \( i \in \{1, 3\} \), the pair \((e, b_1, \ldots, b_{du}), (u, b'_1, \ldots, b'_{du})\) \( \in VC^e \) iff \( e = \{(i, 2), (i, 3)\}, e' = \{(i, 1), (i, 2)\}, e = e^u, e' = e^v, \) for some \( \ell, m, \) and \( b_\ell = b'_m \).
- For \( u = (2, 2) \), the pair \((e, b_1, \ldots, b_{du}), (u, b'_1, \ldots, b'_{du})\) \( \in HC^e \) iff \( e = \{(2, 2), (3, 2)\}, e' = \{(1, 2), (2, 2)\}, e = e^u, e' = e^v, \) for some \( \ell, m, \) and \( b_\ell = b'_m \), and the pair \((e, b_1, \ldots, b_{du}), (u, b'_1, \ldots, b'_{du})\) \( \in VC^e \) iff \( e = \{(2, 2), (3, 2)\}, e' = \{(1, 2), (2, 2)\}, e = e^u, e' = e^v, \) for some \( \ell, m, \) and \( b_\ell = b'_m \).

Let \( n, m \geq 3 \). We define a function \( \Psi \) from \( \text{vertices}(G_{n,m}) \) to \( \text{vertices}(G_{3,3}) \) as follows. We let \( \Psi((1, 1)) = (1, 1), \Psi((n, 1)) = (3, 1), \Psi((1, m)) = (1, 3) \) and \( \Psi((n, m)) = (3, 3) \). For \( 1 < i < n \) and \( 1 < j < m \), we define \( \Psi((i, j)) = (2, 2), \Psi((i, j)) = (1, 2), \Psi((n, i)) = (3, 2), \Psi((i, 1)) = (2, 1) \) and \( \Psi((m, i)) = (2, 3) \). We can now enumerate incident edges of \( a \) in \( G_{n,m} \) according to the already-defined enumeration for \( \Psi(a) \) in \( G_{3,3} \). For each \( a \in \text{vertices}(G_{n,m}) \), we define a bijection \( \Delta_a \) from its incident edges in \( G_{n,m} \) to the incident edges of \( \Psi(a) \) in \( G_{3,3} \) in the natural way: if \( e \) corresponds to the incident edge of \( a \) to the "up" direction in the grid \( G_{n,m} \) then \( \Delta_a(e) \) is also the incident edge of \( \Psi(a) \) in the grid \( G_{3,3} \), to the "up" direction; similarly for the "right", "down" and "left" directions. Then for each \( a \in \text{vertices}(G_{n,m}) \), we enumerate its incident edges as \( e_1^{a}, \ldots, e_{du}^{a} = \Delta_a^{-1}(e_1^{\Psi(a)}, \ldots, e_{du}^{\Psi(a)}) \), where \( e_1^{\Psi(a)}, \ldots, e_{du}^{\Psi(a)} \) is the enumeration for \( \Psi(a) \) already fixed in the construction of \( TP^a \).

We now formalize the intuition that the parity and consistency conditions ensure that a rectangular grid cannot be tiled:

**Claim 2.** \( I_{n,m}^{grid} \not\twoheadrightarrow I_{TP}^p \), for every \( n, m \geq 1 \).

**Proof.** Note that \( I_{n,m}^{grid} \not\twoheadrightarrow I_{TP}^p \) if \( \min(n, m) \leq 2 \). Towards a contradiction, suppose \( I_{n,m}^{grid} \twoheadrightarrow I_{TP}^p \) for some \( n, m \geq 3 \), via a homomorphism \( h \). By construction, we must have \( \pi_1(h(a)) = \Psi(a) \), for every \( a \) in \( I_{n,m}^{grid} \) and hence \( h \) corresponds to a 0/1 assignment of the edges of the Gaifman graph \( G_{n,m} \) of \( I_{n,m}^{grid} \). In particular, there exists a 0/1 vector \( (x_e)_{e \in \text{E}(G_{n,m})} \) such that for each \( a \in \text{vertices}(G_{n,m}) \), we have \( \hat{h}(a) = (\Psi(a), x_{e_1^{a}}, \ldots, x_{e_1^{a}}) \). Now we have

\[
\sum_{a \in \text{vertices}(G_{n,m})} (x_{e_1^{a}} + \cdots + x_{e_{du}^{a}}) = \sum_{a \in \text{vertices}(G_{n,m}))} (x_{e_1^{a}} + \cdots + x_{e_{du}^{a}}) = 1 \quad \text{(mod 2)}
\]

But this is impossible as each edge \( e \in \text{edges}(G_{n,m}) \) is counted exactly twice in \( \sum_{a \in \text{vertices}(G_{n,m})} (x_{e_1^{a}} + \cdots + x_{e_{du}^{a}}) \); a contradiction.

While there is no total mapping from \( I_{n,m}^{grid} \) to \( I_{TP}^p \) that is a homomorphism, by considering partial mappings with domains that are not too large, we can easily satisfy the correct parity conditions, and hence we can define partial homomorphisms from \( I_{n,m}^{grid} \) to \( I_{TP}^p \). The next claim tells us that these partial homomorphisms can be chosen to be consistent.

**Claim 3.** \( I_{n,m}^{grid} \rightarrow_k I_{TP}^p \), for every \( n, m \geq 3 \) and \( 2 \leq k < \min(n, m) \).

**Proof.** Let \( P = (a_0, a_1, \ldots, a_k) \) be a walk in \( G_{n,m} \). For every edge \( e \in \text{edges}(G_{n,m}) \), we define:

1. \( x_{e}^P = 1 \) if \( P \) visits \( e \) an odd number of times.
2. \( x_{e}^P = 0 \) if \( P \) visits \( e \) an even number of times.

We also define \( h^P(a) = (\Psi(a), x_{e_1^{a}}^P, \ldots, x_{e_{du}^{a}}^P) \), for each \( a \in \text{vertices}(G_{n,m}) \) (i.e., in the domain of \( I_{n,m}^{grid} \)).
Let \( W \) be the collection of all walks \( P = (a_0, a_1, \ldots, a_t) \) in \( G_{n,m} \) with \( a_0 = (1,1) \) and \( a_t \neq a_0 \). We claim that for each \( P = (a_0, a_1, \ldots, a_t) \in W \) and each \( a \neq a_t \) in \( \text{vertices}(G_{n,m}) \), the tuple \( h^P(a) \) always belongs to the domain of \( I^{TP}_P \). Note that \( x_{e_i}^{a} + \cdots + x_{e_{a_t}}^{a} = |\{ e \in P : e \text{ is incident to } a \}| \) (mod 2). For \( a \neq a_0 \), we have \( |\{ e \in P : e \text{ is incident to } a \}| = 2 \cdot |\{ i : 0 < i < \ell \text{ and } a_i = a \}| = 0 \) (mod 2), and hence \( h^P(a) = (\Psi(a), x_{e_0}^{a}, x_{e_1}^{a}, \ldots, x_{e_{a_t}}^{a}) \) belongs to \( I^{TP}_P \) (as \( \Psi(a) \neq (1,1) \)). On the other hand, for \( a = a_0 \), we have \( |\{ e \in P : e \text{ is incident to } a \}| = 1 + 2 \cdot |\{ i : 0 < i < \ell \text{ and } a_i = a \}| = 1 \) (mod 2), and hence \( h^P(a) = (\Psi(a), x_{e_0}^{a}, x_{e_1}^{a}, \ldots, x_{e_{a_t}}^{a}) \) belongs to \( I^{TP}_P \) (as \( \Psi(a) = (1,1) \)).

Thus we can define for each walk \( P = (a_0, a_1, \ldots, a_t) \in W \) a partial mapping \( h^P \) from \( I^{grid}_{n,m} \) to \( I^{TP}_P \) with domain \( \text{vertices}(G_{n,m}) \setminus \{ a_t \} \). By definition of \( TP^P \) and since \( h^P \) is defined from a 0/1 vector \( (x_{e}^{P})_{e \in \text{EDGES}(G_{n,m})} \), we have that \( h^P \) is actually a partial homomorphism.

We define a non-empty collection \( \mathcal{H} \) of partial homomorphisms from \( I^{grid}_{n,m} \) to \( I^{TP}_P \) as follows. For \( 1 \leq p \leq n \) and \( 1 \leq q \leq m \), we denote by \( C_{p,q} \) the \( (p,q) \)-cross of \( G_{n,m} \) defined as \( C_{p,q} = \{ (p,j) : 1 \leq j \leq m \} \cup \{ (i,q) : 1 \leq i \leq n \} \). For every non-empty subset \( S \subseteq \text{vertices}(G_{n,m}) \) with \( |S| \leq k \) (recall that \( 2 \leq k < \min(n,m) \)), and every walk \( P = (a_0, \ldots, a_t) \in W \) such that there are \( p,q \) with \( a_e \in C_{p,q} \) and \( C_{p,q} \cap S = \emptyset \), we add to \( \mathcal{H} \) the restriction \( h^P|_S \). We prove that \( \mathcal{H} \) is a winning strategy for the Duplicator and then \( I^{grid}_{n,m} \rightarrow_k I^{TP}_P \) as required. Condition (1) of Fact 5 holds by definition, so we focus on condition (2). Let \( h^P|_S \in \mathcal{H} \) for some \( S \) with \( |S| < k \) and walk \( P = (a_0, \ldots, a_t) \in W \) such that \( a_e \in C_{p,q} \) and \( C_{p,q} \cap S = \emptyset \) for some \( p,q \). Let \( a \in \text{vertices}(G_{n,m}) \setminus S \) and \( S' = S \cup \{ a \} \). Since \( k < \min(n,m) \), there exist \( p',q' \) such that \( C_{p',q'} \cap S' = \emptyset \). Moreover, since \( C_{p,q} \) is connected and \( |C_{p,q} \cap C_{p',q'}| \geq 2 \), there is a walk \( P'' = (a_e, a_{\ell+i}, \ldots, a_{\ell+t}) \) such that \( a_{\ell+i} \in C_{p,q} \) for all \( 0 \leq i \leq r \), \( a_{\ell+r} \in C_{p',q'} \) and \( a_{\ell+r} \neq a_0 \). Let \( P' = (a_0, \ldots, a_{\ell+r}) \) be the concatenation of \( P \) and \( P'' \). Then \( h^P|_S \in \mathcal{H} \). Finally, observe that \( h^P(b) = h^P(b) \) for every \( b \in S \), since \( x_{e}^{P} \) can only differ for edges \( e = \{ b', b'' \} \subseteq C_{p,q} \) and \( C_{p,q} \cap S = \emptyset \). It follows that \( h^P|_S \subseteq h^P|_S \), and hence condition (2) holds.

\[ \square \]

Additional comments on non-Datalog-rewritable examples

We mentioned in the conclusion of the paper that for the example query \( Q_{TP} \) in views produced in the proof of Theorem 8 there is a rewriting in a slightly larger language, stratified Datalog. The details of stratified Datalog will not concern us here, except that it includes positive Boolean combinations of Datalog queries and relational algebra queries. We will show that the example has a rewriting that is such a Boolean combination. We now explain this. In fact, what we show is that for every tiling problem TP for which rectangular grids can not be tiled, the query \( Q_{TP} \) from Theorem 6 has a rewriting that is a positive Boolean combination of Datalog queries and relational algebra queries. In particular, this shows that \( Q_{TP} \) always has a separator in PTIME.

Denote by \( Q^{\text{START}}_S \) the query obtained from \( Q^{\text{START}} \) by replacing \( C \) and \( D \) by the first and second projections of \( S \), respectively. Let \( Q^{\text{VERIFY}}_S \) be obtained from \( Q^{\text{VERIFY}} \) by using the views. That is, by replacing:

- \( \text{CQ HA} \) by view \( V_{\text{HA}} \) and similarly for \( \text{VA} \),
- \( \text{relations } T_j \) by the corresponding atomic views,
- rewriting rules corresponding to the second to last bullet item as \( V_j(o,x,y,z), V_{T_j}(z) \), and similarly rewriting rules corresponding to the final bullet item using \( V_F \).

Let \( \text{PRODUCTTest} \) be a query that tests whether \( S \) is the product of its projections. \( \text{PRODUCTTest} \) can be expressed in relational algebra, hence in stratified Datalog.

Consider the query \( R \) formed by existentially quantifying

\[ V_{C}^{\text{HELP}} \lor V_{D}^{\text{HELP}} \lor Q_{\text{VERIFY}}^{S} \lor (Q_{\text{START}}^{S} \land \text{PRODUCTTest}) \]

Clearly \( R \) is a positive Boolean combination of Datalog queries and the relational algebra query \( \text{PRODUCTTest} \). We claim that \( R \) is a rewriting of \( q \).

In one direction, suppose \( Q \) returns true on \( I \) and let \( J \) be the view image. We do a case analysis depending on which of the top-level disjuncts holds. If \( Q^{\text{HELP}} \) holds on \( I \) then \( V_{C}^{\text{HELP}} \lor V_{D}^{\text{HELP}} \) is non-empty, and thus \( R \) holds in \( J \). If \( Q^{\text{VERIFY}} \) holds on \( I \) then \( Q_{\text{VERIFY}}^{S} \) holds on \( J \) and hence we conclude again that \( R \) holds on \( J \). Finally, suppose \( Q_{\text{START}}^{S} \) holds on \( I \). If \( Q^{\text{VERIFY}} \) fails, we know one of \( C \) or \( D \) is empty. But then \( Q_{\text{START}} \) cannot hold, a contradiction to our assumption. Thus \( \text{PRODUCTTest} \) must hold. From this, it is easy to see that \( Q_{\text{START}}^{S} \) holds. This completes the proof of this direction.

Conversely suppose that \( R \) holds on the view image \( J \). Again we do a case analysis on the top-level disjuncts. If \( V_{C}^{\text{HELP}} \) or \( V_{D}^{\text{HELP}} \) is nonempty on \( J \), then \( Q^{\text{HELP}} \) holds on \( I \) and hence \( Q \) holds on \( I \). If \( Q_{\text{VERIFY}}^{S} \) holds on \( I \), then \( Q^{\text{VERIFY}} \) holds on \( I \), and again we conclude that \( Q \) holds on \( I \). Finally, suppose \( Q_{\text{START}}^{S} \land \text{PRODUCTTest} \) holds on \( J \), and suppose that none of the disjuncts of \( Q \) hold. Note that since \( Q_{\text{HELP}} \) fails, \( V_{C}^{\text{HELP}} \) and \( V_{D}^{\text{HELP}} \) must be empty. Thus we have two possibilities for \( S \). There is the “projection case”, where either one of \( C \) or \( D \) is empty, and all the \( S \) atoms are generated by the second rule. The alternative is the “product case”, where both \( C \) and \( D \) are both nonempty and all the atoms of \( S \) are generated by the first rule.
We claim that we must be in the “product case” for $S$ above. If we are in the projection case, then every pair must be associated with a tile. Further, since $Q_{\text{VERIFY}}$ and $Q_{\text{HELPER}}$ fail, we have a tiling of a rectangular grid, contradicting the hypothesis that there is no tiling. Since we have argued that we are in the product case, it follows that $Q_{\text{START}}$ holds on $I$ and thus $Q$ holds in $I$ as required.

**Proof of Theorem 9**

Recall the statement:

There is no integer-valued function $F$ such that for all $Q, V$ such that $V$ and $Q$ are in Datalog and $Q$ is monotonically determined over $V$, there is a separator of $Q$ over $V$ that runs in time $F(V(l))$.

We now give the proof of Theorem 9. We assume the opposite, aiming for a contradiction. We use the following fact, which is a consequence of the time hierarchy theorem:

For any computable function $F$ there is a deterministic Turing machine $M_F$ which halts on all of its inputs, and such that no Turing machine running in time $F$ can decide the same language as $M_F$.

Fix such a machine $M$ for $F$.

Let $V_{\text{INPUT}}$ be the input alphabet of $M$, and $\Sigma_M$ be a suitable alphabet for encoding configurations of $M$.

We consider a base signature with relations $\Sigma_{\text{INPUT}}$ for the input signature of $M$ along with symbols $\text{Succ}(x, y), U_a(x) : a \in \Sigma_{\text{INPUT}}$ for the configuration signature of $M$.

A *pre-run-string* is a string in the regular language formed by intersecting

$$\sigma_{\text{INPBEGIN}} \Sigma_{\text{INPUT}}^* \sigma_{\text{INPEND}} \Sigma_{\text{M}}^+ \sigma_{\text{RUNEND}}$$

with a regular expression enforcing that the last maximal segment of $\Sigma_M$ strings that does not contain $\varepsilon$; encodes a halting state. Above:

- $\sigma_{\text{INPBEGIN}}$ is a marker designating the beginning of the input while
- $\sigma_{\text{INPEND}}$ designates the end of the input,
- $\varepsilon$ is a marker indicating the separator between configurations, while
- $\sigma_{\text{RUNEND}}$ marks the end of the run.

A *well-shaped string* will consist of an initial letter with a special symbol $\sigma_{\text{INPBEGIN}}$ and ending with $\sigma_{\text{INPEND}}$, followed by a code for a run of $M$, ending with a special symbol $\sigma_{\text{RUNEND}}$. A string is *badly-shaped* if it is not well-shaped. It is easy to see that a badly-shaped string $w$ has at least one of the following *bad properties*: $w$ is not a pre-run string, $w$ contains a sub-string $\sigma_i \sigma_{i+1}$, where $\sigma_{i+1}$ does not encode a next configuration after $\sigma_i$, $w$ contains a string $\sigma_{\text{INPBEGIN}} \varepsilon \sigma_{\text{INPEND}} \sigma_{\text{RUNEND}} c_1$ such that $c_1$ does not encode initial configuration of $M$ with input $w_in$.

A *pre-run instance* will be a relational encoding of a homomorphic image of a pre-run string using the relations $\text{Succ}(x, y), U_a(x) : a \in \Sigma_{\text{INPUT}}$ for the coding of the initial segment, symbols $\text{Succ}'(x, y), U_a'(x) : a \in \Sigma_{\text{M}}$ for the remaining part of the run, and additional symbols for the separators. That is, in the relational encoding we allow the same element to represent different places in the string. A *well-shaped string instance* will be a relational encoding of a homomorphic image of a well-shaped string, again using the relations $\text{Succ}(x, y), U_a : a \in \Sigma_{\text{INPUT}}$ for the initial segment and the primed copies for the remaining segments. We define a badly-shaped string instance analogously.

A standard argument shows

**Proposition 13.** There is a Datalog query whose approximations are (up to isomorphism) exactly the badly-shaped string instances.

Note that if we had enforced that codings were *alternating*, with every other configuration reversed, then we could use a PDA to detect bad properties on a string and a context-free path query to detect it on the encoding. With the power of general Datalog, no alternation is needed.

Our views $V$ will include:

- the input views, with one binary view returning exactly $\text{Succ}(x, y)$, and for each $a \in \Sigma_{\text{INPUT}}$ a unary view returning $U_a(x)$.
- a nullary view $\text{BADLY-SHAPED}$ which returns $\text{TRUE}$ whenever the instance contains a badly-shaped string instance. That is, $\text{BADLY-SHAPED}$ returns $\text{TRUE}$ when the input contains the homomorphic image of a relational encoding of a string starting with the symbol $\sigma_{\text{INPBEGIN}}$ and ending with the symbol $\sigma_{\text{INPEND}}$ which has one of the bad properties. By Proposition 13, a Datalog view with this property exists.
- a unary view $\text{PRE-RUN}(x)$ which holds for $x$ if there is a sub-instance that is a pre-run instance in which the occurrence of $\sigma_{\text{INPEND}}$ corresponds to $x$.

Our query $Q$ will be the sentence obtained from $\text{BADLY-SHAPED}$ disjoined with $Q_{\text{ACCEPT}}$, where $Q_{\text{ACCEPT}}$ returns true exactly when we detect a relational encoding of a pre-run string that ends in an accept state.

We now argue that $Q$ is monotonically determined over $V$.

Consider instances $I_1$ and $I_2$ with $V(I_1) \subseteq V(I_2)$ and $Q(I_1)$ being true.

$Q(I_2)$ could be true because $\text{BADLY-SHAPED}$ holds, in this case, $Q(I_2)$ also holds since $\text{BADLY-SHAPED}$ is one of the views. So we can assume that $\text{BADLY-SHAPED}$ does not hold in $I_1$ or $I_2$, since if it does hold then $I_2$ satisfies $Q$. 

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$Q(I_1)$ could also be true because $V^{\text{badly-shaped}}$ fails but $Q^{\text{accept}}$ holds. We know there is a relational encoding of some string

$$\sigma_{\text{InpBegin}} w_0 \sigma_{\text{InpEnd}} w_1 \cdots w_n \sigma_{\text{RunEnd}}$$

witnessing that $Q^{\text{accept}}$ holds in $I_1$. Let $x$ be the element corresponding to the label $\sigma_{\text{InpEnd}}$ in this encoding. Note that $V^{\text{pre-run}}$ must hold of $x$ in $I_1$, hence in $I_2$. The latter must be witnessed via a relational encoding of some string of the form

$$\sigma_{\text{InpBegin}} w'_0 \sigma_{\text{InpEnd}} w'_1 \cdots w'_k$$

with $\sigma_{\text{InpBegin}} w'_0 \sigma_{\text{InpEnd}}$ relationally encoded in the unprimed signature, the $w'_i$ encoded in the primed signature, with the element labelled by $\sigma_{\text{InpEnd}}$ corresponding to $x$. Note that by the definition of pre-run, $w'_i$ must include a halting state.

Since we have views for all of the input signature elements, and $V(I_1) \subseteq V(I_2)$, we know that we also have an encoding of a string $\sigma_{\text{InpBegin}} w_0 \sigma_{\text{InpEnd}}$ in $I_2$, with the encoding done in the unprimed signature, with the

We now consider the string

$$s = \sigma_{\text{InpBegin}} w_0 \sigma_{\text{InpEnd}} w'_1 \cdots w'_k \sigma_{\text{RunEnd}}$$

$s$ begins with the input string, and ends with a halting state. Note that since a relational encoding of $\sigma_{\text{InpBegin}} w_0 \sigma_{\text{InpEnd}}$ lies in $I_1$, the encoding of $s$ must lie in $I_2$, due to the input views. Since $V^{\text{badly-shaped}}$ is false in $I_2$, we know that in $I_2$:

- For every relational encoding of a string of the form:

  $$\sigma_{\text{InpBegin}} w_0 \sigma_{\text{InpEnd}} w'_1$$

  with $\sigma_{\text{InpBegin}} w_0 \sigma_{\text{InpEnd}}$ encoded in the unprimed signature and $w'_1$ encodes a state with tape configuration $w_0$ and state the initial state of $M$, under the transition relation of $M$.

- For every relational encoding a string of the form:

  $$w'_1; w'_2$$

  with the encoding being in the primed signature, $w'_1$ must encode a state that is a successor in the transition relation of $M$ of the state encoded by $w'_2$.

From this we infer that $s$ is an encoding of a run of $M$ on $w_0$, ending at a halting state. But since $M$ is deterministic, $s$ must be the same as

$$\sigma_{\text{InpBegin}} w_0 \sigma_{\text{InpEnd}} w'_1 \cdots w_n \sigma_{\text{RunEnd}}$$

which ends in an acceptance state.

Since a relational encoding of $s$ lies in $I_2$, we can conclude that $Q$ holds in $I_2$. This completes the argument for monotonic determinacy of $Q$ with respect to $V$.

Now, suppose $Q$ has a separator $R$ that runs in time $F$. Then $R$ will allow us to check in time $F$ whether $M$ accepts or rejects on its input, a contradiction. Thus we have completed the proof of Theorem 9.