All–instances termination of chase is undecidable

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Abstract. We show that all–instances termination of chase is undecidable. More precisely, there is no algorithm deciding, for a given set \( T \) consisting of Tuple Generating Dependencies (a.k.a. Datalog\(^3\) program), whether the \( T \)-chase on \( D \) will terminate for every finite database instance \( D \). Our method applies to Oblivious Chase, Semi-Oblivious Chase and – after a slight modification – also for Standard Chase. This means that we give a (negative) solution to the all–instances termination problem for all version of chase that are usually considered.

The arity we need for our undecidability proof is three. We also show that the problem is EXPSPACE-hard for binary signatures, but decidability for this case is left open.

Both the proofs – for ternary and binary signatures – are easy. Once you know them.

1 Introduction

The chase procedure was defined in late 1970s and has been considered one of the most fundamental database theory algorithms since then. It has been applied to a wide spectrum of problems, for example for checking containment of queries under constraints [ASU79] or for testing implication between sets of database dependencies ([MMS79], [BV84]). A new wave of interest in this notion began when the theory of data integration was founded ([FKPP05]), where chase is used to compute solutions to data exchange problems. This interest was further strengthened recently by the Datalog\(^\pm\) program [CGL09], [CGL12].

The basic idea of a \( T \)-chase is as follows. We consider a set \( T \) of Tuple Generating Dependencies\(^1\), which means rules (constraints) of the form:

\[
\Phi(\bar{x}, \bar{y}) \Rightarrow \exists \bar{z} \Psi(\bar{x}, \bar{z})
\]

where \( \Phi \) and \( \Psi \) are conjunctive queries\(^2\), and where \( \bar{x}, \bar{y} \) and \( \bar{z} \) are tuples of variables. Then, for a database instance \( D \) we try – step by step – to extend \( D \),

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\(^1\) Such sets are also known as Datalog\(^3\) programs, and we will use the word “program” in this sense. While chase is sometimes also defined for other types of dependencies, we only consider Tuple Generating Dependencies in this paper.

\(^2\) \( \Phi \) and \( \Psi \) are positive, without equality. Our negative results hold for single head TGDs, which means that \( \Psi \) is a single atom.
by adding new elements and atoms, so that the new database satisfies the constraints from $T$: whenever there are some elements $\bar{a}, \bar{b}$ in the current structure, such that $\Phi(\bar{a}, \bar{b})$ is true, a tuple $\bar{c}$ of new elements is created and new relational atoms added, to make $\Psi(\bar{a}, \bar{c})$ also true. Notice that the tuple $\bar{z}$ can be empty. In such case the TGD under consideration degenerates to a plain Datalog rule.

As it turns out, there are several possible semantics of the whenever above, leading to several versions of the chase procedure. The Standard Chase is a lazy version – it only adds new elements if $\Phi(\bar{a}, \bar{b})$ is true in the current structure, but $\exists \bar{z} \Psi(\bar{a}, \bar{z})$ is (at this point of execution) false. Oblivious and Semi-Oblivious Chase ([M09]) are eager versions. Oblivious Chase always adds one tuple $\bar{c}$ for each tuple $\bar{a}, \bar{b}$ such that $\Phi(\bar{a}, \bar{b})$ is true. Semi-Oblivious Chase always adds one tuple $\bar{c}$ for each tuple $\bar{a}$ such that $\exists \bar{y} \Phi(\bar{a}, \bar{y})$ is true.

It is not hard to notice that the order of execution does not matter for Oblivious and Semi-Oblivious Chase. Whatever order the candidate tuples are picked in, we will eventually get the same structure. But Standard Chase is non-deterministic – different orders in which tuples are picked can eventually lead to different structures.

One more version of the procedure is Core Chase (see [DNR08]). It is again a lazy version, but a parallel one: all the rules applicable at some point are triggered at the same time. In this way the non-determinism of Standard Chase is got rid of. For reasons that we will not discuss here Core Chase is slightly more complicated than that (and not really practical – the cost of each step is DP-complete).

As we said before, the chase procedure is almost ubiquitous in database theory. This phenomenon is discussed in [DNR08]: "the applicability of the same tool to (...) seemingly different problems is not accidental, and it is due to a deeper, tool-independent reason: to solve these problems, it suffices to exhibit a representative (database) instance $U$ with two key properties, and the chase is an algorithm for finding such an instance." The two key properties of the instance $U$, being the result of $T$-chase on an a database instance $D$, for given set $T$ of tuple generating dependencies and for given database instance $D$ are that:

- $U$ is a model of $T$ and $D$;
- $U$ is universal - there is a homomorphism from $U$ into every model of $D$ and $T$.

But $U$, or $\text{Chase}(D, T)$, as we prefer to call the structure resulting from running a $T$-chase on $D$, is in many cases only useful when it is finite, which only happens if (and only if) the chase procedure terminates. One of the applications where finiteness of $\text{Chase}(D, T)$ is a key issue is considered in [FKPP05], and a sufficient condition on $T$, implying finiteness of $\text{Chase}(D, T)$ was studied in this paper, called Weak Acyclicity. Weak Acyclicity is a property of $T$ alone, so it implies termination regardless of $D$. This reflects the fact that the typical context in which database constraints are analyzed is the static analysis context – we want to optimize $T$ before knowing $D$. So, in particular, it is natural to want

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3 If chase does not terminate the claim is true provided the order is fair – each tuple will be eventually picked.
to be sure that \( T \)-chase on \( D \) will terminate on \( D \) before knowing the \( D \) itself. Many other conditions like that were studied. For example the Stratified-Witness property ([DT03]), which is historically earlier, and stronger (i.e. narrower), than Weak Acyclicity. Then it was the Rich Acyclicity criterion, introduced in [HS07], and proved in [GO11] to imply termination of Oblivious Chase for all instances \( D \). A condition based on stratification of rules was introduced in [DNR08]. As it turned out to only guarantee termination of the Standard Chase, another class of sets of rules – Corrected Stratified Class (CSC) was defined in [MSL09], with Oblivious Chase terminating for all instances \( D \). Then, in [MSL09a], CSC was extended to Inductively Restricted (IR) class, and further to a whole hierarchy of classes \( T[k] \), where \( T[2] = IR \).

This list is by no means exhaustive – see Adrian Onet’s thesis [O12] for a 35-pages long survey chapter about sufficient conditions for chase termination. What is however worth mentioning is that all the known conditions imply all-instances termination and thus none of them depends on \( D \).

With so much effort spent on finding the sufficient conditions it is natural to ask about decidability of the all-instances termination problem itself. But surprisingly, this fundamental problem has so far remained open. Some work was done, but mostly on a related problem of chase termination for given program \( T \) and also given database instance \( D \). It was shown to be undecidable in [DNR08] for Core Chase and Standard Chase (\( ♣ \)). In [M09] it was noticed that the proof of \( ♣ \) works also for Semi-oblivious and Oblivious chase. The only previous results concerning decidability of the all-instances chase termination problem can be found in [G013], where the problem is shown to be undecidable for Core Chase (\( ▽1 \)) and the Standard\( _3 \) sub-version (\( ▽2 \)), where we ask, for given \( T \), whether for each database instance \( D \) there exists a terminating execution path of \( T \)-Standard Chase on \( D \) (let us remind here that Standard Chase is a non-deterministic procedure). And this is again not really the most natural question as – having some \( T \) on mind – we want to be sure that whenever and however we run a \( T \)-chase, it will always terminate\(^4\).

Another result in [GO13] is undecidability of all-instances chase termination problem for sets of constraints where, apart from TGDs, a denial constraint is allowed, which is a conjunctive query \( Q \) such that when \( Q \) is proved somewhere in Chase(\( T, D \)) then the chase procedure terminates and “fails” (\( ♠ \)).

One more result from [M09], which can be slightly confusing, is undecidability of what is there – misleadingly – called “all-instances termination” (\( ♦ \)). The signature \( Σ \) of the TGDs there is a disjoint union of two sub-signatures \( Σ_1 \) and \( Σ_2 \) but only instances where the relations in \( Σ_2 \) are initially empty are allowed.

1.1 Our contribution
The main result of this paper is:

**Theorem 1** All-instance termination of Oblivious Chase is undecidable (and r.e.-hard) for programs consisting of single-head TGDs over ternary signatures.

\(^4\) The termination problem for the Standard\(_3 \) version is shown to be \( Π_2^0 \) complete in [GO13]. But the result statement there is not correct: co-r.e. completeness is claimed.
Proof of Theorem 1 is presented in Section 3. It can also be read, without any changes, as a proof of undecidability of all-instances termination of Semi-Oblivious Chase. In short Subsection 3.5 we modify the proof to show that also all-instances–all-paths Standard Chase termination is undecidable.

It is common knowledge that whatever can be said about TGDs over high arity signature usually remains true for binary signature, as long as multi-head TGDs are allowed. And also the other way round – one who is prepared to pay the arity cost can usually translate everything into the language of single-head TGDs. This fails however in the context of chase termination: one can easily modify our proof of Theorem 1 to get undecidability of all-instance chase termination for multi-head binary TGDs, but only for Semi-Oblivious Chase, not for Oblivious. See Appendix D. for details. In Section 4 we show:

**Theorem 2** All-instance termination of Oblivious Chase is EXPSPACE-hard for programs consisting of single head TGDs over binary signatures.

**Upper bounds.** It follows easily from Lemma 8 that the all-instances termination problem of Oblivious and Semi-Oblivious Chase is recursively enumerable, and so Theorem 1 provides matching lower bounds. But Lemma 8 is not true for all-instances–all-paths Standard Chase termination, and thus the only upper bound known for this problem is the $\Pi^0_2$ level of the Arithmetical Hierarchy. Our conjecture is that the problem is in fact also r.e., but much more insight into the structure of Standard Chase is needed in order to prove this claim.

The lower bound given by Theorem 1 is not matched by any upper bound, and we believe that the problem is undecidable. A similarity that is maybe worth being mentioned here (see also next subsection) – is that Datalog programs uniform boundedness is also known to be undecidable for ternary arities but decidability was left open for the binary case [M99].

## 2 Techniques

It will not be too unfair to say that the proof of ♦, in [DNR08], is not complicated. The possibility of having our favorite instance $D$ fixed gives a lot of control, and having this control it is not hard to encode a computation of a machine of one’s choice as $T$-chase for some program $T$. The same can be said about $\Diamond$, whose proof, in [M09], is an adaptation of the proof of ♦ – the input instance over signature $\Sigma_1$ is neglected, a new instance, over $\Sigma_2$, hardwired in dedicated TGDs, is created, and then the proof from [DNR08] is applied.

**Flooding rule.** The schema from [DNR08] is repeated, in a sense, in the proof of $\forall 2$ in [GO13]. The instance $D$ is treated as an input of some machine, and chase simulates the computation of this machine on given input. Chase terminates when the computation does. The problem are the instances $D$ which contain too much positive information to be understood by a Datalog $\exists$ program as a finite

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5 This is for sake of completeness, as it was earlier shown in [GO13] that undecidability of all-instances termination of Oblivious Chase implies undecidability of all-instances–all-paths Standard Chase termination.
input – for example instances that contain a loop, which is unavoidably seen by a program as an infinite path.

The trick used in [GO13] to make sure that chase will terminate on such unwelcome instances is the flooding rule – a technique earlier used in 1990s in the numerous papers dealing with the Datalog boundedness problem [GMSV93]. Let us illustrate it by an example:

**Example.** Consider the program $T$:

(i) $U(x, y, z), E(z, w) \Rightarrow \exists u U(y, u, w)$;

(ii) $E(x, y) \rightarrow E^+(x, y)$;

(iii) $E^+(x, y), E(y, z) \Rightarrow E^+(x, z)$;

(iv) $E^+(x, x) \Rightarrow U(y, u, w)$ (flooding rule)

To see what is going on here, notice that $E$-atoms are never produced. Rules (ii) and (iii) compute $E^+$, being the (non-reflexive) transitive closure of $E$. Rule (i) unfolds graph $E$: if $\langle x, y \rangle$ is an edge in the unfolding, $y$ is “over” an element $z$ in $E$ and if there is an edge $\langle z, w \rangle$ in $E$ then a new element $u$ must exist in the unfolding, being “over” $w$.

It is easy to see that whatever $D$ we begin with, $T$-Standard Chase on $D$ has a terminating path. If $E$ is acyclic, then rule (i) terminates for all chase variants. If $E$ has a cycle, then rules (ii) and (iii) can prove $E^+(a, a)$ for some $a$, and then rule (iv) can be used to “flood” the predicate $U$, so that in consequence, the head of (i) will be always satisfied and (i) will never be triggered again.

But there is no hope for this trick to work for the all–instances–all–paths Standard Chase: flooding rule only terminates a Standard Chase if we can make sure it is always used early enough to prevent new elements to be born, which means that it must be us who decides what the execution order is.

Clearly, this technique also fails for the eager chase variants. $T$-Oblivious Chase on $D$ does not terminate whenever $D$ is an instance containing an atom $U(a, b, c)$, for some $c$ belonging to a cycle in $E$.

Notice also that adding a denial constraint to the constraints (♣) is just another way of using a flooding rule – instead of flooding the database we make the chase fail.

**Drinking from the well of positivity.** The trick we invented in this paper to replace the flooding rule is as follows. We treat the instance $D$ as the only source of some positive facts: there are predicates which are never proved, they can only come with $D$.

Then the idea is that each new element $a$ of Chase uses the path leading from $D$ to $a$ to run its private computation of some Turing-complete computational model. Only Datalog rules are used in this computation so we do not need to bother about termination. In order to be able to give birth to a successor $a$ must first reach, by means of atoms created during its private computation, some atom that can only be found in $D$. Elements of Chase which are already too far away from this source of positivity cannot drink from it any more, dessicate, and do not produce offspring, thus causing the chase to terminate.

As we are going to see in the next Section, once one knows the above idea, the proof of Theorem 1 is easy.
3 Proof of Theorem 1

3.1 The well of positivity

From now on, whenever we say “chase” we mean Oblivious Chase.

Informally we say that Oblivious Chase creates one witness for each tuple satisfying the body of an existential TGD, regardless whether such a witness is already present in the current database instance or not. One of the ways how this informal statement can be formalized is to construct, for a given Datalog\footnote{Proof of Theorem 1} program $\mathcal{T}$ a new program $\mathcal{T}'$, by replacing each TGD in $\mathcal{T}$, of the form:

(i) $\Phi(\bar{x}) \Rightarrow \exists y \, \Psi(y, \bar{x})$

where (i) is the number of the rule in $\mathcal{T}$, and $\Phi$ and $\Psi$ are conjunctive queries, by a rule:

(i') $\Phi(\bar{x}) \Rightarrow \Psi(h_i(\bar{x}), \bar{x})$

where $h_i$ is a Skolem function. In this way $\text{Chase}(D, \mathcal{T})$ is the structure whose active domain is a subset of Herbrand universe, where the elements of $D$ are treated as constants, and terms are built out of constants using the Skolem functions $h_i$, and which is a minimal model for all the rules of the program $\mathcal{T}'$. Since $\mathcal{T}'$ is a Prolog program it always has such a minimal model.

Now the question whether the $\mathcal{T}$-Oblivious Chase on $D$ terminates is equivalent to the question whether $\text{Chase}(D, \mathcal{T})$, seen as a substructure of the Herbrand universe, contains, for each $k \in \mathbb{N}$, a term of depth at least $k$.

For a given signature $\Sigma$, an element $a_\Sigma$ of a database instance $D$ over $\Sigma$ will be called a well of positivity if for each relation $R \in \Sigma$ the atom $R(a_\Sigma, a_\Sigma, \ldots, a_\Sigma)$ is true in $D$. By $D_\Sigma$ we will denote the database instance consisting of a single element, being a well of positivity.

Lemma 3. The following conditions are equivalent for Datalog\footnote{Proof of Theorem 1} program $\mathcal{T}$:

(i) for each database instance $D$, $\mathcal{T}$-Oblivious Chase on $D$ terminates;
(ii) $\mathcal{T}$-Oblivious Chase terminates on $D_\Sigma$.

This is (rephrased) Theorem 2 in [M09]. We sketch its proof for completeness.

Proof. Only the (ii)$\Rightarrow$ (i) implication needs a proof. Let us assume that there exists $D$ such that $\text{Chase}(D, \mathcal{T})$, seen as a substructure of the Herbrand universe, contains, for each $k \in \mathbb{N}$, a term of depth at least $k$. What we need to prove is that also $\text{Chase}(D_\Sigma, \mathcal{T})$ does contain such a term.

So let $t$ be a term of depth at least $k$ in $\text{Chase}(D, \mathcal{T})$. This means that there is a derivation, in program $\mathcal{T}'$, having atoms of $D$ in its leaves and some atom containing $t$ in its root. When we replace all the elements of $D$, occurring in atoms of this derivation, by the well of positivity $a_\Sigma$, then we will get another valid derivation in program $\mathcal{T}'$, leading, instead of $t$, to some new term $t'$ in $\text{Chase}(D_\Sigma, \mathcal{T})$. And the depth of $t'$ is equal to the depth of $t$ – the two terms only differ at the level of constants, but are equal otherwise. $\square$
3.2 The problem to be reduced

The undecidable problem we are going to encode is the halting problem for finite automata with three counters (3CM). More precisely, the instance of the problem Halt3CM is a triple consisting of finite set $Q$ of states, of some initial state $q_1 \in Q$ and of a finite set $\Pi$ of instructions, each of them of the following format:

- if the current state is $q \in Q$,
- the value of the first counter (is/is not) zero
- and the value of the second counter (is/is not) zero
then:
- change the state to $q' \in Q$;
- (increment|decrement|keep unchanged) the value of the first counter,
- (increment|decrement|keep unchanged) the value of the second counter,
- increment the value of the third counter.

We assume here that the automaton is deterministic, which means that the part of the instruction which is after then is a function of the part occurring before then. This function is partial – if a configuration is reached with no instruction applicable then the automaton halts.

The problem, called Halt3CM, is whether, for a given 3CM $M$, executing the instructions of $M$ will ever halt when started from the state $q_1$ and three empty counters. Of course Halt3CM is undecidable. From now on each time we say “$M$ halts” we mean that it halts after started from $q_1$ and three empty counters.

Notice that the value of the third counter is never read by the automaton, and the counter is incremented in each step. This leads to the following:

**Lemma 4.** A 3CM halts if and only if the set of values of its third counter is bounded.

From now on a 3CM $M = (Q, q_1, \Pi)$ is fixed and we will construct a Datalog³ program $T_M$, over some signature $\Sigma_M$ such that $T_M$-Oblivious Chase on $D_{\Sigma_M}$ terminates if and only if $M$ halts.

3.3 Encoding the automaton as a Conway function

Now we will encode the computation of $M$ as a sequence of iterations of a Conway function. This technique is by no means new, but maybe not as widely known as some other undecidable problems, so we include this subsection for completeness.

Suppose $|Q| = m$. Let $p_1 = 2$, $p_2 = 3$, $\ldots$, $p_{m+3}$ be the first $m + 3$ primes and let $p = p_1 p_2 \ldots p_{m+3}$. Let $c$ be a configuration of $M$ with the state being $q_i$, and $c_1$, $c_2$ and $c_3$ being respectively values of the first, second and third counter. Then by $e(c)$ (or encoding of $c$) we will mean the number:

$$p p_1^{c_1} p_2^{c_2} p_{m+3}^{c_3}$$

Notice that if $c$ is the initial configuration of $M$ then $e(c) = 2$. 
For two configurations \(c, c'\) of \(M\) we will say that they are consecutive when \(c'\) is a result of executing a single step of \(M\) in \(c\) or when there is no instruction that can be executed in \(c\) and \(c = c'\). Now it is easy to see that:

**Theorem 5** There exist natural numbers \(q_0, q_1, \ldots q_{p-1}, r_0, r_1, \ldots r_{p-1}\), such that for each two consecutive configurations \(c, c'\) of \(M\), such that \(e(c) = i \mod p\) it holds that \(e(c') = \frac{q_i(c)}{r_i} \mod p\).

For the proof of this theorem notice that the reminder \(i\) of \(e(c)\) modulo \(p\) carries all the information needed for \(M\) to decide which instruction should be applied: the state is \(q_j\) if and only if \(i\) is divisible by \(p_j\) and the value of the (for example) second counter is non-zero if and only if \(i\) is divisible by \(p_{m+2}\).

It is equally easy to see that executing an instruction boils down to division (removing the old state, decrementing a counter) and multiplication (moving to a new state, incrementing a counter).

From now on the numbers \(q_0, q_1, \ldots q_{p-1}, r_0, r_1, \ldots r_{p-1}\) provided for \(M\) by Theorem 5 are fixed. Denote by \(g\) a function that maps a natural number \(n\) to \(nq_i/r_i\), where \(n = i \mod p\). Let \(G = \{g^n(2) : n \in \mathbb{N}\}\) be the smallest subset of \(\mathbb{N}\) which contains 2 and is closed under \(g\). Clearly, \(M\) halts if and only if \(G\) is bounded. So, what remains for us to do is to construct such a Datalog\(^2\) program \(T_M\) that \(T_M\)-Oblivious Chase on \(D_{\Sigma_M}\) terminates if and only if \(G\) is bounded. Notice that it is here where the third counter is important.

### 3.4 The program \(T_M\)

Denote by \(QR\) the set \(\{q_0, q_1, \ldots q_{p-1}, r_0, r_1, \ldots r_{p-1}\}\). The signature \(\Sigma_M\) will consist of the following relations:

- a binary relation \(E\), which will pretend to be the successor relation on the natural numbers;
- for each \(j \in QR\) a binary relation \(E^j\) – only needed to keep rule (d3) short;
- a unary relation \(H\), which will never occur in the head of any rule, so its only atom will be \(H(a_\Sigma)\);
- for each \(0 \leq i \leq p-1\) a ternary relation \(T^i\), with \(T^i_x(y, z)\) meaning something like “\(x\) thinks that \(\frac{x}{y} = \frac{z}{r_i}\). Normally we should of course write \(T(x, y, z)\) rather than \(T_x(y, z)\). But we like \(T_x(y, z)\) more, and it is still ternary;
- for each \(0 \leq i \leq p-1\) a binary relation \(R^i\), with \(R^i_x(y)\) meaning something like “\(x\) thinks that \(i = y \mod p\);
- a binary relation \(G\), with \(G_x(y)\) meaning “\(x\) thinks that \(y \in G\);
- a unary relation \(N\), with \(N(x)\) meaning that \(x\) is a natural number. \(N\) is not really needed, we only have it because otherwise the bodies of rules (d2) and (d4) would be empty, and we do not like rules with empty bodies.

Now we are ready to write the program \(T_M\). There is one existential rule:

\((e)\ G_x(y), H(y) \Rightarrow \exists z E(z, x)\).

Read this rule as “Once \(x\) has drunk from the well of positivity, it is allowed to give birth to a new element \(z\).”
There will be also several Datalog rules:

\[
\begin{align*}
(d0) & \quad E(y, y_1), E(y_1, y_2), \ldots, E(y_{j-1}, y_j) \Rightarrow E^j(y, y_j) \quad \text{one rule for each } j \in \mathcal{Q}R; \\
(d1) & \quad E(z, x) \Rightarrow N(z)
\end{align*}
\]

Rules of the form (d2) and (d3) form a recursive definition of multiplication by addition (remember \(x\) always thinks it equals zero):

\[
\begin{align*}
(d2) & \quad N(x) \Rightarrow T_i^0(x, x) \quad \text{one rule for each } 0 \leq i \leq p - 1; \\
(d3) & \quad T_i^j(y, z), E^h(y, y'), E^r(z, z') \Rightarrow T_i^j(y', z') \quad \text{one rule for each } 0 \leq i \leq p - 1;
\end{align*}
\]

The next two rules count modulo \(p\):

\[
\begin{align*}
(d4) & \quad N(x) \Rightarrow R^0_i(x); \\
(d5) & \quad R^i_j(y), E(y, y') \Rightarrow R^i_j(y') \quad \text{whenever } j = i + 1 \mod p;
\end{align*}
\]

Now, once we have all the predicates we need for the multiplications, and for remainders modulo \(p\), we can easily write rules which will compute the set \(G\).

First of them says – as long as \(x\) keeps assuming that it equals zero – that \(2 \in G\):

\[
\begin{align*}
(d6) & \quad E(x, y), E(y, z) \Rightarrow G_x(z)
\end{align*}
\]

Second rule for \(G\) says that \(G\) is closed with respect to the function \(g\):

\[
\begin{align*}
(d7) & \quad R^i_j(y), G_x(y), T_i^j(y, z) \Rightarrow G_x(z) \quad \text{one rule for each } 0 \leq i \leq p - 1;
\end{align*}
\]

Notice that the rules (d2)–(d7) form a sort of a private Datalog program for each \(x\), and the atoms proved by such programs for different \(x, x'\) never see each other (this is reflected in our notation, which suggests that \(x\) is more than merely an argument of the predicates, but part of their names). Rule (e) creates a new element \(z\), such that \(E(z, x)\), when the program for \(x\) can prove that \(G_x(y)\) for some \(y\) such that \(H(y)\). But, as we said, there is no rule saying that something is in \(H\) and the only element \(a\) such that \(\text{Chase}(D_{\Sigma_M}, T_M) \models \neg H(a)\) is the well of positivity \(a_{\Sigma_M}\). So (e) creates a new element \(z\), such that \(E(z, x)\), when the program for \(x\) can prove that \(G_x(a_{\Sigma_M})\).

Now we have a lemma that Theorem 1 follows from:

**Lemma 6.** \(T_M\)-Oblivious Chase on \(D_{\Sigma_M}\) terminates if and only if \(G\) is bounded.

We think that the lemma follows directly from the construction of \(T_M\). But the readers who like it more formal, are invited to read Appendix A.

### 3.5 The case of all-instances-all-paths Standard Chase termination

For any \(T\) and \(D\) any structure being a result of running a \(T\)-Standard Chase on \(D\) is a subset of (oblivious) \(\text{Chase}(D, T)\). This means that if \(G\) is bounded, then \(T_M\)-Standard Chase terminates on each instance and each path. What remains to be seen is that if \(G\) is not bounded, then there exists \(D\) such that \(T_M\)-Standard Chase does not terminate on some path. It is easy to see that a structure \(D\), consisting of the well of positivity \(a_{\Sigma_M}\) and of some \(a\) such that \(D \models E(a, a_{\Sigma_M})\), has this property.
4 Proof of Theorem 2

It is harder to prove any nontrivial lower bound for all-instances Oblivious Chase termination problem for single-head TGDs over binary signatures, then to prove undecidability in the general case. In the proof of Theorem 2 we try to repeat the idea of proof of Theorem 1, creating a new element of some $E$-path, for a binary $E$, only when some private computation, run by the last element $a$ of the current path, terminates. But, while having arity three at our disposal, we could run many mutually non-interfering computations using the same arena, now we must construct a separate arena for each element of the $E$-path being built.

This arena needs to be huge enough to contain a complex computation, but on the other hand the process of the construction of the arena should never lead to an infinite chase. In other words we need to – and we think it is not immediately clear how to do it – find a binary Datalog $\exists \Sigma$ program which builds a huge (i.e. greater than exponential, with respect to the size of the program) (Oblivious) Chase, when run on $a_\Sigma$, but finally terminates.

4.1 Constructing the arena: Chase of exponential depth

Let $m$ be a fixed natural number and let $M = 2^m$. Consider the program $T_0^b(m)$ consisting of the following rules:

- (d0) $H(x) \Rightarrow K(x)$
- (d0') $H(x) \Rightarrow C_i(x)$ (one rule for each $i \in \{0, 1, \ldots m\}$)
- (e) $K(x) \Rightarrow \exists y R(x, y)$
- (d1) $R(x, y) \Rightarrow T(y, y)$
- (d2) $T(x, y), R(x', z), R(z, x), R(y', y) \Rightarrow T(x', y')$
- (d3) $T(x, y), C_i(x) \Rightarrow C_{i+1}(y)$ (one rule for each $i \in \{0, 1, \ldots m - 1\}$)
- (d4) $R(x, y), C_m(x) \Rightarrow K(y)$

Let now $a$ be any element such that $H(a)$ (which means that $a$ may be, but may not be, a well of positivity), and let $D_a$ be a database instance containing $a$ as a single element.

**Exercise 7** Chase($D_a, T_0^b(m)$), seen as a graph over predicate $R$, is a path of length $M + 1$, having $a$ as its first element

Solution to this exercise can be found in Appendix B. **Hint:** like in Section 3 there is no rule saying that something is in $H$, and the only element satisfying $H$ plays the role of the well of positivity. Also like in Section 3 Oblivious Chase produces a path (this time it is an $R$-path) – if an element is in $K$ then it is “close enough” to $H$ to be able to produce $R$-offspring. The predicates $C_i$ are resources – the further we are from $H$ the more we are running out them.
4.2 Constructing the arena: Chase of double exponential size

For fixed natural numbers \(m\) and \(p\) consider now the program \(T^1_b(m,p)\) consisting of all the rules that can be obtained from the rules of \(T^0_b(m)\) by replacing each occurrence of the predicate \(R\) with one of the predicates \(R_1, \ldots, R_p\). For example rule (e) will be replaced by \(p\) new rules while rule (d2) will be replaced by \(p^3\) new rules. Let \(a\) and \(D_a\) be as in the previous subsection. Then the analysis of Chase\((D_a, T^1_b(m,p) \cup T)\) is analogous to the analysis of Chase\((D_a, T^0_b(m))\), except that the structure we now get is a \(p\)-ary tree of depth \(M + 1\) rather than a path of length \(M + 1\). Notice that the same elements are created regardless if \(a\) is a well of positivity, or any element just satisfying \(H(a)\).

4.3 The encoding Lemma and how it implies Theorem

Now Chase\((D_a, T^1_b(m,p))\) can be used as an arena, where we can run some computation. Let \(a\) and \(D_a\) be as before.

**Lemma 8 (The encoding Lemma).** The problem:

Given \(m, p \in \mathbb{N}\) and a Datalog program \(T\), with EDB relations \(H, R_1, R_2, \ldots, R_p\) and IDB relations \(P, (binary)\) and \(G_1, G_2, C\) (unary). Is it the case that:

\[
\text{Chase}(D_a, T^1_b(m,p) \cup T) \models C(a)\ ?
\]

is EXSPACE-hard.

The size of the instance is here the size of the program \(T^1_b(m,p) \cup T\).

For the proof of the Lemma see Appendix C. Notice that Chase\((D_a, T^0_b(m))\) has the same set of elements as Chase\((D_a, T^1_b(m,p))\) – this is because the Datalog rules of \(T\) do not prove any atoms that could be used by \(T^1_b(m,p)\).

Let now \(T^2_b(m,p)\) be \(T^1_b(m,p)\) with the following additional rules:

- \((d')\) \(E(x,y) \Rightarrow H(y)\)
- \((e')\) \(C(x) \Rightarrow \exists z \ E(x,z)\).

Proof of Theorem will be finished when we show:

**Lemma 9.** For a Datalog program \(T\), as in Lemma the following two conditions are equivalent:

- \(\text{Chase}(D_a, T^1_b(m,p) \cup T) \models C(a)\)
- \(\text{Chase}(D_\Sigma, T^2_b(m,p) \cup T)\) does not terminate.

For the proof of the Lemma first suppose that \(\text{Chase}(D_a, T^1_b(m,p) \cup T) \models C(a)\). Let us run \(T^2_b(m,p)\) on \(D_\Sigma\). Since \(C(a_\Sigma)\) is true in \(D_\Sigma\), rules \((e')\), and then \((d')\) will be triggered, creating a new element \(c\) satisfying \(H(c)\). Then, rules of \(T^1_b(m,p)\) will build the \(p\)-ary tree of depth \(M + 1\) rooted in \(c\) and \(T\) will be run on this tree, proving \(C(c)\). But this means that \((e')\) will trigger again, creating element \(c'\) such that \(E(c,c')\) and \(H(c')\), and so on.
Now suppose that $\text{Chase}(D_a, T_b^f(m, p) \cup T) \not= C(a)$. Then again, an element $c$ like above will be created, and the $p$-ary tree of depth $M + 1$ rooted in $c$ will be built, $T$ will be run on this tree, but $C(c)$ will never be proved, no new elements will be added, and chase will terminate. □

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6 Appendix A. Proof of Lemma 6

As there is no risk of confusion we will denote the structure, being the result of \( T_M \)-Oblivious Chase on \( D_{\Sigma_M} \) simply as Chase.

We are going to prove two lemmas describing the structure of Chase. First, let us think of Chase as of a graph with respect to the relation \( E \). Since we do not want to be distracted by the loop \( E(a_{\Sigma_M}, a_{\Sigma_M}) \) all the time, let \( E_0 \) be the relation \( E \) in Chase minus the edge \( E(a_{\Sigma_M}, a_{\Sigma_M}) \).

**Lemma 10.** Chase is a descending \( E_0 \)-path, finite or infinite, without self-loops. The first element of this path is the well of positivity \( a_{\Sigma_M} \).

In order to prove Lemma 10 it is enough to show that:

(i) \( E_0 \) is a connected graph;
(ii) there are no \( E_0 \)-cycles in Chase,
(iii) the \( E_0 \)-out-degree of each element of Chase is 1, except from \( a_{\Sigma_M} \), whose out-degree is 0;
(iv) the \( E_0 \)-in-degree of each element of Chase is at most 1.

To see that (i) holds true notice that for each element \( a \) of Chase, there exists a (descending) \( E_0 \) path from \( a_{\Sigma_M} \) to \( a \). It can be easily proved by induction on the structure of Chase that this path already exists at the moment when \( a \) is created.

For the proof of (ii) notice that whenever Chase \( |E_0(a, b) \) then \( b \) was created by the chase procedure earlier than \( a \).

Concerning (iii), notice that the only way for an element \( b \neq a_{\Sigma_M} \) to be in an atom \( E_0(a, b) \) in Chase, for some \( a \), is to be created by rule (e) from elements \( x = a, y = a_{\Sigma_M} \).

For claim (iv) it is enough to see that the only rule of \( T_M \) that creates atoms of \( E \) is rule (e). Since there is only one element satisfying \( H \), each element of Chase is involved, as the variable \( x \), in at most one tuple satisfying the body of rule (e). By the rules of oblivious chase this means that rule (e) is triggered at most once for each element of Chase being \( x \). This ends the proof of Lemma 10.

Let now \( a = a_0 \) be any element of Chase and for each \( i \in \mathbb{N} \) let \( a_{i+1} \) be the unique element of Chase such that Chase \( |E_0(a_i, a_{i+1}) \). Of course there exists \( k \in \mathbb{N} \) such that for each \( i \geq k \) we have \( a_i = a_{\Sigma_M} \). Let \( k_0 \) be the smallest such \( k \). Then the following lemma follows easily from Lemma 10 and from the construction of rules (d2)–(d7):

**Lemma 11.** (i) If there is an \( E \)-path in Chase from \( a \) to some \( b = a_j \) for some \( j \in \mathbb{N} \);
(ii) if \( T_\alpha^* (b, c) \) holds in Chase for some elements \( b, c \) then \( b = a_j, c = a_j' \) for some \( j, j' \in \mathbb{N} \) such that \( j \equiv j' \mod p \);
(iii) if \( R_\alpha^* (b) \) holds in Chase for some element \( b \), then \( b = a_j \) for some \( j \in \mathbb{N} \) such that \( j = i \mod p \);
(iv) if \( G \) is bounded, and \( k_0 \) is greater than all the elements of \( G \) then \( G = \{ j : \text{Chase} \models G_\alpha(a_j) \} \);
(v) if \( G \) is unbounded then \( G \subseteq \{ j : \text{Chase} \models G_\alpha(a_j) \} \); in particular in such case Chase \( \models G_\alpha(a_{\Sigma_M}) \).
Notice that if \( G \) is unbounded then it may be that \( \text{Chase} \models G_a(a_j) \) for some \( j \notin G \), even if \( j < k_0 \). We imagine the predicate \( G_a \) as \( a \) moving a pebble to the values of subsequent iterations of \( g \). But once the pebble falls to the well of positivity all the control is lost, and different things can happen.

Now, we are ready to prove Lemma 6.

Suppose first that \( G \) is unbounded. We will show that in such case \( \text{Chase} \) is an infinite descending \( E \)-path, which means that for each \( a \) in \( \text{Chase} \) there is a \( b \neq a \) such that \( \text{Chase} \models E(b, a) \). But by Lemma 11 (v) if \( G \) is unbounded then \( \text{Chase} \models G_a(a_{\Sigma M}) \), so the body of rule (e) is satisfied in \( \text{Chase} \) for \( x = a \) and its head must also be satisfied.

Now suppose that \( G \) is bounded, that \( k \) is a natural number greater than all the elements of \( G \), and that \( \text{Chase} \) is an infinite path. Let \( a \) be any element such that the \( E \)-distance between \( a \) and \( a_{\Sigma M} \) is greater than \( k \). Then, by Lemma 11 (iv) \( G_a(a_{\Sigma M}) \) is never proved, and the rule (e) could never have been triggered for \( a_0 \) as \( x \). But this contradicts the assumption that \( \text{Chase} \) was infinite. \( \square \)

7 Appendix B. Solution to Exercise 7

It is clear that \( \text{Chase}(D_a, T_0^0(m)) \), seen as an \( R \)-graph is a path: this is because rule (e) can only create one \( R \)-successor for each node.

Now, suppose that in the process of building \( \text{Chase}(D_a, T_0^0(m)) \) we always trigger Datalog rules as early as possible, and that rule (e) is only used when there are no more Datalog rules applicable (this can be assumed since – as we already noticed in the Introduction – the order in which rules are used by Oblivi ous Chase does not matter). Let \( C_N \) be the (partial) \( \text{Chase}(D_a, T_0^0(m)) \) after rule (e) was used for the \( N \)th time, which means that \( C_N \) consists of \( N + 1 \) elements, call them \( a_0 = a, a_1, a_2, \ldots a_N \) (in the order that they were created in), and after all the Datalog rules were saturated.

Notice that we do not claim that \( C_N \) exists for each \( N \).

Lemma 12. \( C_N \models T(a_i, a_j) \) if and only if \( i + N \leq 2j \)

Proof. Easy induction – on the depth of the derivation. \( \square \)

Lemma 13. Suppose \( 1 \leq i \leq m \). Then \( C_N \models C_i(a_j) \) if and only if \( \frac{j}{N} \leq \frac{2^{m-1}}{2^m} \).

Proof. Induction on \( i \). Use Lemma 12. \( \square \)

Lemma 14. \( C_N \models K(a_j) \) if and only if \( \frac{j-1}{N} \leq \frac{2^m-1}{2^{m+1}} \).

Proof. Notice that (unless \( j = 0 \)) the only rule that can prove \( K(a_j) \) is rule (d4). This means that \( C_N \models K(a_j) \) if and only if \( C_N \models C_m(a_{j-1}) \). But – due to Lemma 13 this holds if and only if \( \frac{j-1}{N} \leq \frac{2^m-1}{2^{m+1}} \). \( \square \)

Lemma 15. \( C_N \models K(a_N) \) if and only if \( N \leq M \)

Proof. Lemma 14 says that \( C_N \models K(a_N) \) if and only if \( \frac{N-1}{N} \leq \frac{2^{m-1}}{2^{m+1}} \). \( \square \)

But this means that rule (e) will be triggered in \( C_N \) if and only if \( N \leq M \). Which ends the solution of Exercise 7. \( \square \)
8 Appendix C. Proof of Lemma 8 (the encoding Lemma)

An instance of the problem we are going to encode, call it Thue2, consists of:

- a finite \( A = \{1, 2, \ldots, p\} \);
- a natural number \( m \);
- a set of productions \( \pi \subseteq A^2 \times A^2 \).

Notice that each production of our Thue process replaces an infix of length 2 by another infix of length 2. So obviously, the word problem is decidable for Thue2. It is however straightforward to prove, by a standard encoding of a Turing machine, that the problem:

\[
\text{Given an instance of Thue2. Does this instance have a solution, which means that there exists a number } k < M \text{ such that } 1p^k \xrightarrow{a} 2p^k ?
\]

is EXPSPACE-complete. Notice that, as always, \( M = 2^m \).

Let us remind the reader that Chase \( \Pi \) in exactly one step. Rule (p1) says that whenever there are two words of the form \( 1p^k \), one can reach the word \( 2p^k \) in some number of steps, in each step replacing some infix \( w \) of a current word by an infix \( w' \) in such a way that \( \langle w, w' \rangle \in \pi \). Notice that if \( 1p^k \xrightarrow{a} 2p^k \) is true for some \( k < M \) then also \( 1p^{k-1} \xrightarrow{a} 2p^{k-1} \) is true, so the statement of the problem may seem to be unnecessarily complicated. But this is how we need it.

From now on we assume that an instance \( \Pi = \langle p, m, \pi \rangle \) of Thue2 is fixed. Our goal is to build a Datalog program \( T \), over the signature as required by Lemma 8, and such that \( \text{Chase}(D_a, T^b_a(m, p)) \cup T \models C(a) \) if and only if \( \Pi \) has a solution.

Let us remind the reader that Chase \( (D_a, T^b_a(m, p)) \) is a tree, and elements of this tree can be in a natural way seen as words from \( A^{\leq M} \). The program \( T \) will first of all contain the following rules, defining some new binary relation \( P \) on \( A^{\leq M} \):

\[(p1) \quad R_i(x, y), R_i(y, y'), R_j(x, z), R_j(z, z') \Rightarrow P(y', z') \quad \text{one rule for each pair } \langle i', j' \rangle \in \pi.\]

\[(p2) \quad P(x, y), R_i(x, x'), R_i(y, y') \Rightarrow P(x', y') \quad \text{one rule for each } i \in A.\]

It is easy to see that the predicate \( P \) computes pairs of words \( \langle w, w' \rangle \in A^{\leq M} \times A^{\leq M} \) such that \( w \rightarrow_w w' \), which means that \( w \) rewrite to \( w' \) by the Thue process \( \Pi \) in exactly one step. Rule (p1) says that whenever there are two \( a \) and \( b \) elements of Chase \( (D_a, T^b_a(m, p)) \), which represent words of the form \( wii' \) and \( wjj' \), such that \( \langle ii', jj' \rangle \in \pi \) then \( P(a, b) \) holds true in \( \text{Chase}(D_a, T^b_a(m, p)) \cup T \). Rule (p2) provides a mechanism able to add any (but the same) suffix, both to \( wii' \) and \( wjj' \).

Now we are going to play a pebble game, like in Appendix A. Next three rules of \( T \) allow us to place a pebble on any element of Chase \( (D_a, T^b_a(m, p)) \) that represents a word of the form \( 1p^k \) for some \( k \leq M \):

\[(g1) \quad H(x), E_1(x, y) \Rightarrow G^1(y) ; \]
\[(g2) \quad G^1(y), E_p(y, y') \Rightarrow G^1(y') ; \]
\[(g3) \quad G_1(x) \Rightarrow G(x). \]

The next rule is the main mechanism of \( T \). It lets us follow the derivation in \( \Pi \):

\[(h1) \quad H(x), E_2(x, y) \Rightarrow G^2(y) . \]
(g4) $G(y), P(y, y') \Rightarrow G(y').$

Finally, there are three rules in $\mathcal{T}$ that make it possible to check whether the element that we placed the pebble on does indeed represent a word of the form $2^p k$:

(g5) $G(y) \Rightarrow G_2(y)$;
(g6) $G_2(y), E_p(y', y) \Rightarrow G_2(y')$
(g7) $H(x), E_2(x, y), G_2(y) \Rightarrow C(x)$.

It now follows from the construction that Chase$(D_a, T^1_b(m, p) \cup \mathcal{T} \models C(a))$ if and only if $H$ has a solution.

9 Appendix D. Low arity vs. single head TGDs. Discussion.

In the program $T_M$ we constructed in Section 3 there are several ternary rules of the form:

\[ (*) \quad T(x, y, z), E_q(y, y'), E_r(z, z') \Rightarrow T(x, y', z') \]

One could think that, if we allowed multi-head TGDs, the ternary relation in the head of rule $(*)$ could be easily split into three binary relations:

\[ (** \quad T_1(v, x), T_2(v, y), T_3(v, z), E_q(y, y'), E_r(z, z') \Rightarrow \exists w T_1(w, x), T_2(w, y'), T_3(w, z') \]

where $v$ and $w$ are “names” for atoms $T(x, y, z)$ and $T(x, y', z')$. Of course all other rules involving $T$ would also need to be changed accordingly.

This is however not so simple. Notice how careful we were, in Section 3, about existential TGDs. There was only one of them, and not easy to trigger. And, while replacing $(*)$ with $(**)$, we replace a safe Datalog rule with a potentially prolific existential one. Indeed, it is not hard to see that, if we consider the Oblivious Chase, the $w$ in the head of $(**) \text{ depends as a Skolem term, on the } v \text{ in the body, which results with recursive calls and infinite chase. However, in the case of the Semi-Oblivious Chase, } v \text{ only depends on } x, y' \text{ and } z', \text{ and the only candidates for } y' \text{ and } z' \text{ are all the elements on the } E\text{-path from the current } x \text{ to } a_\Sigma, \text{ which are finitely many. This means that using the above splitting we can really rewrite proof from Section 3 to a proof of:}

Theorem 16 All-instance termination of Semi-Oblivious Chase is undecidable (and r.e.-hard) for programs consisting of multi-head TGDs over binary signatures.

As we said in the Introduction, we do not know whether all-instances-all-paths Standard Chase termination is recursively enumerable. Actually, it could very well be the case that it is r.e. for single-head TGDs but not for multi-head TGDs. In particular, the standard translation of multi-head TGDs into single head TGDs, where a new predicate is added for the head of each rule, from which the atoms of the head are then produced using projections, does not preserve all-instances-all-paths Standard Chase termination. As an example consider a program $T$ consisting of a single rule:

\[ E(x, y) \Rightarrow \exists z E(y, z), E(z, y) \]

$T$-Standard Chase on $D$ terminates for each $D$. Let however the following $T'$ be the natural translation of $T$:

\[ E(x, y) \Rightarrow \exists z E(y, z), E(z, y) \]
\[ E(x, y) \Rightarrow \exists z \ R(y, z) \]
\[ R(y, z) \Rightarrow E(y, z) \]
\[ R(y, z) \Rightarrow E(z, y) \]

Then it is easy to see that \( T' \)-Standard Chase does not always terminate.