

# THE POWER OF THE TERMINATING CHASE

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Fig. 1: The Chase

Part 1:

# Tuple-Generating Dependencies

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# Existential Rules

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Datalog<sup>+</sup>

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$$\forall x, y. \varphi[x, y] \rightarrow \exists z. \psi[x, z]$$

# Tuple-generating dependencies a.k.a. Existential rules a.k.a. Datalog<sup>+</sup>

**Definition:** A **rule** is a formula of the form:

$$\forall x, y. \varphi[x, y] \rightarrow \exists z. \psi[x, z]$$

**Rule body:** conjunction of atoms

- using variables from  $x \cup y$
- possibly using constants

**Rule head:** conjunction of atoms

- using variables from  $x \cup z$
- possibly using constants

**Frontier:** variables  $x$  used on both sides

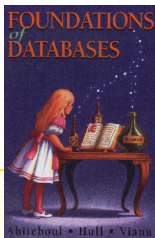
**Facts** can be encoded as variable-free rules with empty body

**Universal quantifiers** are usually omitted

## Example 1: Inclusion dependencies

The following inclusion dependency from the Alice Book:

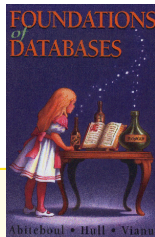
$\text{Showings}[\text{Title}] \subseteq \text{Movies}[\text{Title}]$



relates tables  $\text{Showings}[\text{Theatre, Screen, Title, Snack}]$  and  $\text{Movies}[\text{Title, Director, Actor}]$



# Example 1: Inclusion dependencies



The following **inclusion dependency** from the Alice Book:

$$\text{Showings}[\text{Title}] \subseteq \text{Movies}[\text{Title}]$$

relates tables  $\text{Showings}[\text{Theatre}, \text{Screen}, \text{Title}, \text{Snack}]$  and  $\text{Movies}[\text{Title}, \text{Director}, \text{Actor}]$

This can be expressed by the rule:

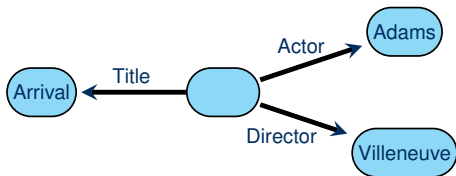
$$\text{Showings}(y_{\text{Theatre}}, y_{\text{Screen}}, x_{\text{Title}}, y_{\text{Snack}}) \rightarrow \exists z_{\text{Director}}, z_{\text{Actor}}. \text{Movies}(x_{\text{Title}}, z_{\text{Director}}, z_{\text{Actor}})$$

## Example 2: Data exchange and data integration

Different databases often require different structures.

**Example:** The [W3C RDB2RDF](#) standard specifies how to translate relational databases into graph databases in RDF format

The tuple `Movies(Arrival, Villeneuve, Adams)`, e.g., is translated to a graph of the form

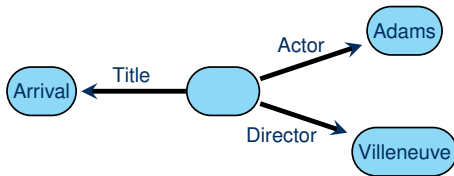


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This can be expressed by the rule:

$$\text{Movies}(x_{\text{Title}}, x_{\text{Director}}, x_{\text{Actor}}) \rightarrow \exists z. \text{Title}(z, x_{\text{Title}}) \wedge \text{Director}(z, x_{\text{Director}}) \wedge \text{Actor}(z, x_{\text{Actor}})$$

## Example 3: Ontology-based Query Answering

**Ontologies** have been proposed as means to represent and exchange descriptive schema-level knowledge

**Example:** The [W3C OWL Web Ontology Language](#) is based on [description logics](#) (DLs). Many popular OWL/DL fragments can be translated into rules. The following ontology specifies some facts about parts of compound objects (corresponding DL syntax axiom in parenthesis):

$\text{Bicycle}(x) \rightarrow \exists v.\text{hasPart}(x, v) \wedge \text{Wheel}(v)$  (Bicycle  $\sqsubseteq$   $\exists\text{hasPart}.\text{Wheel}$ )

$\text{Wheel}(x) \rightarrow \exists w.\text{properPartOf}(x, w) \wedge \text{Bicycle}(w)$  (Wheel  $\sqsubseteq$   $\exists\text{properPartOf}.\text{Bicycle}$ )

$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$  (properPartOf  $\sqsubseteq$  partOf)

$\text{hasPart}(x, y) \rightarrow \text{partOf}(y, x)$  (hasPart  $\sqsubseteq$  partOf<sup>-</sup>)

$\text{partOf}(x, y) \rightarrow \text{hasPart}(y, x)$  (partOf  $\sqsubseteq$  hasPart<sup>-</sup>)

# Problems for Existential Rules

One of the main computational problems in these applications is the following:

## QUERY ANSWERING UNDER CONSTRAINTS

**Input:** A concrete database  $\mathcal{D}$ , a set of rules  $\Sigma$ , and a conjunctive query  $q$

**Problem:** What are the certain answers of  $q$  over  $\mathcal{D}$  and  $\Sigma$ ? More formally:

- which substitutions  $\sigma$  from free variables in  $q$  to constants of  $\mathcal{D}$
- satisfy the first-order entailment  $\Sigma, \mathcal{D} \models q\sigma$ ?

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The corresponding decision problem is as follows:

## QUERY ENTAILMENT UNDER CONSTRAINTS

**Input:** A concrete database  $\mathcal{D}$ , a set of rules  $\Sigma$ , and a Boolean CQ  $q$

**Problem:** Does  $\Sigma, \mathcal{D} \models q$  hold?

# Two perspectives on the use of rules

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## **Rules as “Ontologies”**

- Logical theories encode knowledge
- Rules are exchanged and re-combined
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- Standard exchange syntax
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## Rules as “Programs”

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## Requirements

- Appeal to human engineers
- Expressive power as query language (w.r.t. data)
- Fast reasoners, robust to database changes

# Existential rules vs. logic programming

Note that query entailment under existential rules is inter-reducible to query (or fact) entailment for definite logic programs (Horn rules without  $\exists$  but with function symbols).



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“**Existential rules**  $\rightarrow$  **definite LP rules**” **Skolemisation**: replace existentially quantified variables by function terms that apply fresh skolem functions to the frontier variables

**Example:** Skolemising the rule  $\text{Wheel}(x) \rightarrow \exists w. \text{partOf}(x, w) \wedge \text{Bicycle}(w)$  yields  $\text{Wheel}(x) \rightarrow \text{partOf}(x, f(x)) \wedge \text{Bicycle}(f(x))$ , with  $f$  a skolem function.

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“**Definite LP rules**  $\rightarrow$  **existential rules**” Flatten function terms: for each  $n$ -ary function  $f$ , we introduce an  $(n + 1)$ -ary predicate  $p_f$ , used to encode “ $x = f(\mathbf{t})$ ” as  $p_f(x, \mathbf{t})$

**Example:** The rule  $R(x, y, f(x, y)) \rightarrow S(g(f(y, x)))$  is translated to  $R(x, y, z) \wedge p_f(z, x, y) \rightarrow \exists v, w.p_f(w, y, x) \wedge p_g(v, w) \wedge S(w)$ .

## Reasoning for existential rules is difficult

**Theorem:** Query entailment under constraints is undecidable (but recursively enumerable). There is a fixed rule set  $\Sigma$  and BCQ  $q$ , such that  $\{\mathcal{D} \mid \Sigma, \mathcal{D} \models q\}$  is undecidable.

**Proof (sketch):** Use a standard encoding of a Turing machine in logical rules, and apply it to a universal Turing machine. Existential quantifiers are used to create new memory cells and time points. □

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This also implies that we cannot restrict to finite models.

**Example:** Consider a database  $r(a, b)$  with constraints

$$r(x, y) \rightarrow \exists z. r(y, z)$$

$$r(x, y) \rightarrow t(x, y)$$

$$t(x, y) \wedge r(y, z) \rightarrow t(x, z)$$

The BCQ  $\exists x. t(x, x)$  is not entailed by this theory, but it holds in all finite models.



# Universal models

**Certain answer semantics:** What is true in **all** models?

But it is often enough to consider “most general models”:

**Definition:** A model  $\mathcal{I}$  of a set of rules  $\Sigma$  is **universal** if it admits a homomorphism  $h : \mathcal{I} \rightarrow \mathcal{J}$  to every model  $\mathcal{J}$  of  $\Sigma$ .

**Fact:** The BCQs entailed by rule set  $\Sigma$  are exactly the BCQs that hold true on any of its universal models.

(The same works for all query languages whose models are closed under homomorphisms)

# Decidable fragments

In the search for decidable fragments, several main principles have been explored:

- **Finite models:** there is a finite universal model
  - full dependencies (no  $\exists$ )
  - many acyclicity notions (more on this later)
- **Tree-like models:** there is universal model of bounded treewidth
  - Guarded rules
  - Frontier-guarded rules
- **Rewritability:** entailment can be reduced to first-order model checking
  - Linear tgds
  - Sticky rules

None of the general criteria are decidable, but the concrete conditions are.

Part 2:  
The Chase

# Applying a rule

Database  $\mathcal{D}$

Rule  $\rho = \varphi[\mathbf{x}, \mathbf{y}] \rightarrow \exists \mathbf{z}. \psi[\mathbf{x}, \mathbf{z}]$

**Definition:** Rule  $\rho$  is **applicable** to  $\mathcal{D}$  if:

1. there is a function  $h : \mathbf{x} \cup \mathbf{y} \rightarrow \text{adom}(\mathcal{D})$  such that  $h(\varphi) \subseteq \mathcal{D}$  (a **match**)
2. there is no function  $h' : \mathbf{x} \cup \mathbf{z} \rightarrow \text{adom}(\mathcal{D})$  with  $h'(x) = h(x)$  for all  $x \in \mathbf{x}$  and  $h'(\psi) \subseteq \mathcal{D}$

The  $\mathcal{D}'$  is the result of **applying  $\rho$  to  $\mathcal{D}$  under  $h$**  if  $\mathcal{D}' = \mathcal{D} \cup \hat{h}(\psi)$  and:

- $\hat{h}(x) = h(x)$  for all  $x \in \mathbf{x}$
- $\hat{h}(z)$  is a fresh null for all  $z \in \mathbf{z}$

# The Chase(s)

A chase constructs a sequence of databases  $\mathcal{D}_0 = \mathcal{D}, \mathcal{D}_1, \mathcal{D}_2, \dots$  by applying rules.

## The Standard Chase (a.k.a. restricted chase)

- Apply rules to matches in some order (strategy)

## The Skolem Chase (a.k.a. semi-oblivious chase)

- Apply skolemised rules (in any order)

## The Datalog-first Chase

- Apply rules to matches in some order that prioritises the application of rules without existential quantifiers

Other prominent chases: **oblivious chase** and **core chase**

# Will it terminate?

$\mathcal{D} = \{\text{Bicycle}(c)\}$

$\text{Bicycle}(x) \rightarrow \exists v.\text{hasPart}(x, v) \wedge \text{Wheel}(v)$

$\text{Wheel}(x) \rightarrow \exists w.\text{properPartOf}(x, w) \wedge \text{Bicycle}(w)$

$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$

$\text{hasPart}(x, y) \rightarrow \text{partOf}(y, x)$

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$$\mathcal{D}_1 = \mathcal{D} \cup \{\text{hasPart}(c, n_1), \text{Wheel}(n_1)\}$$



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$$\mathcal{D}_5 = \mathcal{D}_4 \cup \{\text{partOf}(n_1, n_2)\}$$

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The chase can continue forever ...



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**Applying the Datalog-first chase yields:**

# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

$$\text{Bicycle}(x) \rightarrow \exists v. \text{hasPart}(x, v) \wedge \text{Wheel}(v)$$

$$\text{Wheel}(x) \rightarrow \exists w. \text{properPartOf}(x, w) \wedge \text{Bicycle}(w)$$

$$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$$

$$\text{hasPart}(x, y) \rightarrow \text{partOf}(y, x)$$

$$\text{partOf}(x, y) \rightarrow \text{hasPart}(y, x)$$

## Applying the Datalog-first chase yields:

$$\mathcal{D}_1 = \mathcal{D} \cup \{\text{hasPart}(c, n_1), \text{Wheel}(n_1)\}$$

# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

$$\text{Bicycle}(x) \rightarrow \exists v. \text{hasPart}(x, v) \wedge \text{Wheel}(v)$$

$$\text{Wheel}(x) \rightarrow \exists w. \text{properPartOf}(x, w) \wedge \text{Bicycle}(w)$$

$$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$$

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## Applying the Datalog-first chase yields:

$$\mathcal{D}_1 = \mathcal{D} \cup \{\text{hasPart}(c, n_1), \text{Wheel}(n_1)\}$$

$$\mathcal{D}_2 = \mathcal{D}_1 \cup \{\text{partOf}(n_1, c)\}$$

# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

$$\text{Bicycle}(x) \rightarrow \exists v. \text{hasPart}(x, v) \wedge \text{Wheel}(v)$$

$$\text{Wheel}(x) \rightarrow \exists w. \text{properPartOf}(x, w) \wedge \text{Bicycle}(w)$$

$$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$$

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$$\mathcal{D}_1 = \mathcal{D} \cup \{\text{hasPart}(c, n_1), \text{Wheel}(n_1)\}$$

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$$\mathcal{D}_3 = \mathcal{D}_2 \cup \{\text{properPartOf}(n_1, n_2), \text{Bicycle}(n_2)\}$$

# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

$$\text{Bicycle}(x) \rightarrow \exists v.\text{hasPart}(x, v) \wedge \text{Wheel}(v)$$

$$\text{Wheel}(x) \rightarrow \exists w.\text{properPartOf}(x, w) \wedge \text{Bicycle}(w)$$

$$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$$

$$\text{hasPart}(x, y) \rightarrow \text{partOf}(y, x)$$

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$$\mathcal{D}_3 = \mathcal{D}_2 \cup \{\text{properPartOf}(n_1, n_2), \text{Bicycle}(n_2)\}$$

$$\mathcal{D}_4 = \mathcal{D}_3 \cup \{\text{partOf}(n_1, n_2)\}$$

# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

$$\text{Bicycle}(x) \rightarrow \exists v. \text{hasPart}(x, v) \wedge \text{Wheel}(v)$$

$$\text{Wheel}(x) \rightarrow \exists w. \text{properPartOf}(x, w) \wedge \text{Bicycle}(w)$$

$$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$$

$$\text{hasPart}(x, y) \rightarrow \text{partOf}(y, x)$$

$$\text{partOf}(x, y) \rightarrow \text{hasPart}(y, x)$$

## Applying the Datalog-first chase yields:

$$\mathcal{D}_1 = \mathcal{D} \cup \{\text{hasPart}(c, n_1), \text{Wheel}(n_1)\}$$

$$\mathcal{D}_5 = \mathcal{D}_4 \cup \{\text{hasPart}(n_2, n_1)\}$$

$$\mathcal{D}_2 = \mathcal{D}_1 \cup \{\text{partOf}(n_1, c)\}$$

$$\mathcal{D}_3 = \mathcal{D}_2 \cup \{\text{properPartOf}(n_1, n_2), \text{Bicycle}(n_2)\}$$

$$\mathcal{D}_4 = \mathcal{D}_3 \cup \{\text{partOf}(n_1, n_2)\}$$

# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

$$\text{Bicycle}(x) \rightarrow \exists v. \text{hasPart}(x, v) \wedge \text{Wheel}(v)$$

$$\text{Wheel}(x) \rightarrow \exists w. \text{properPartOf}(x, w) \wedge \text{Bicycle}(w)$$

$$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$$

$$\text{hasPart}(x, y) \rightarrow \text{partOf}(y, x)$$

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## Applying the Datalog-first chase yields:

$$\mathcal{D}_1 = \mathcal{D} \cup \{\text{hasPart}(c, n_1), \text{Wheel}(n_1)\}$$

$$\mathcal{D}_5 = \mathcal{D}_4 \cup \{\text{hasPart}(n_2, n_1)\}$$

$$\mathcal{D}_2 = \mathcal{D}_1 \cup \{\text{partOf}(n_1, c)\}$$

$$\mathcal{D}_3 = \mathcal{D}_2 \cup \{\text{properPartOf}(n_1, n_2), \text{Bicycle}(n_2)\}$$

$$\mathcal{D}_4 = \mathcal{D}_3 \cup \{\text{partOf}(n_1, n_2)\}$$

No further rules are applicable. The chase terminates.



# Will it terminate?

$\mathcal{D} = \{\text{Bicycle}(c)\}$

$\text{Bicycle}(x) \rightarrow \text{hasPart}(x, w(x)) \wedge \text{Wheel}(w(x))$

$\text{Wheel}(x) \rightarrow \text{properPartOf}(x, b(w)) \wedge \text{Bicycle}(b(w))$

$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$

$\text{hasPart}(x, y) \rightarrow \text{partOf}(y, x)$

$\text{partOf}(x, y) \rightarrow \text{hasPart}(y, x)$

# Will it terminate?

$\mathcal{D} = \{\text{Bicycle}(c)\}$

$\text{Bicycle}(x) \rightarrow \text{hasPart}(x, w(x)) \wedge \text{Wheel}(w(x))$

$\text{Wheel}(x) \rightarrow \text{properPartOf}(x, b(w)) \wedge \text{Bicycle}(b(w))$

$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$

$\text{hasPart}(x, y) \rightarrow \text{partOf}(y, x)$

$\text{partOf}(x, y) \rightarrow \text{hasPart}(y, x)$

**Applying the skolem chase yields:**

# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

$$\text{Bicycle}(x) \rightarrow \text{hasPart}(x, w(x)) \wedge \text{Wheel}(w(x))$$

$$\text{Wheel}(x) \rightarrow \text{properPartOf}(x, b(w)) \wedge \text{Bicycle}(b(w))$$

$$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$$

$$\text{hasPart}(x, y) \rightarrow \text{partOf}(y, x)$$

$$\text{partOf}(x, y) \rightarrow \text{hasPart}(y, x)$$

## Applying the skolem chase yields:

$$\mathcal{D}_1 = \mathcal{D} \cup \{\text{hasPart}(c, w(c)), \text{Wheel}(w(c))\}$$

# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

$$\text{Bicycle}(x) \rightarrow \text{hasPart}(x, w(x)) \wedge \text{Wheel}(w(x))$$

$$\text{Wheel}(x) \rightarrow \text{properPartOf}(x, b(w)) \wedge \text{Bicycle}(b(w))$$

$$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$$

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## Applying the skolem chase yields:

$$\mathcal{D}_1 = \mathcal{D} \cup \{\text{hasPart}(c, w(c)), \text{Wheel}(w(c))\}$$

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# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

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$$\text{properPartOf}(x, y) \rightarrow \text{partOf}(x, y)$$

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# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

$$\text{Bicycle}(x) \rightarrow \text{hasPart}(x, w(x)) \wedge \text{Wheel}(w(x))$$

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## Applying the skolem chase yields:

$$\mathcal{D}_1 = \mathcal{D} \cup \{\text{hasPart}(c, w(c)), \text{Wheel}(w(c))\} \quad \mathcal{D}_4 = \mathcal{D}_3 \cup \{\text{partOf}(w(c), b(w(c)))\}$$

$$\mathcal{D}_2 = \mathcal{D}_1 \cup \{\text{partOf}(w(c), c)\}$$

$$\mathcal{D}_3 = \mathcal{D}_2 \cup \{\text{properPartOf}(w(c), b(w(c))), \\ \text{Bicycle}(b(w(c)))\}$$

# Will it terminate?

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$$\mathcal{D}_6 = \mathcal{D}_5 \cup \{\text{hasPart}(b(w(c)), w(b(w(c)))) \\ \text{Wheel}(w(b(w(c))))\}$$



# Will it terminate?

$$\mathcal{D} = \{\text{Bicycle}(c)\}$$

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The chase will certainly continue forever ...

# Chase termination

## **Some observations:**

- Termination is strategy-dependent for standard and Datalog-first chase, but not for skolem chase
- Whenever skolem chase terminates, standard chase terminates for all strategies
- Whenever standard chase terminates (for some/all strategies), Datalog-first chase terminates (for all/some strategies)
- Termination always depends on the concrete database instance

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- Termination always depends on the concrete database instance

## We can define rule classes based on their termination behaviour:

Termination on . . .	instance $\mathcal{D}$	all instances
Skolem chase	$CT_{\mathcal{D}}^{\text{sk}}$	$CT_{\forall}^{\text{sk}}$
Standard chase (all strategies)	$CT_{\mathcal{D}\forall}^{\text{std}}$	$CT_{\forall\forall}^{\text{std}}$
Datalog-first chase (all strategies)	$CT_{\mathcal{D}\forall}^{\text{dlf}}$	$CT_{\forall\forall}^{\text{dlf}}$

# The chase termination problem

**Theorem (Gogacz & Marcinkowski, ICALP'14; Grahne & Onet, Fund.Inf.'18):**  
The classes  $CT_{\mathcal{D}(\mathcal{V})}^x$  and  $CT_{\mathcal{V}(\mathcal{V})}^x$  are undecidable for all  $x \in \{\text{sk}, \text{std}, \text{dlf}\}$ .

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The cases  $CT_{\mathcal{D}(\mathcal{V})}^x$  are simple:

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These cases are recursively enumerable (r.e.).

Membership of  $CT_{\mathcal{V}}^{\text{sk}}$  in r.e. is also simple, due to the following result [Marnette, PODS'09]:

**Proposition:**  $\Sigma \in CT_{\mathcal{V}}^{\text{sk}}$  if and only if  $\Sigma \in CT_{\mathcal{D}^*}^{\text{sk}}$ , where  $\mathcal{D}^*$  is the **critical instance** consisting of all atoms that can be stated over the signature using constants from  $\Sigma$  and an additional constant  $*$ .

# Universal chase termination

Hardness of  $CT_{\forall}^{\text{sk}}$  is more tricky: **how to simulate a Turing machine starting from  $\mathcal{D}^*$ ?**

- Every conjunctive query already matches
- It is difficult to apply rules in any orderly fashion

Solved by [Gogacz & Marcinkowski, ICALP'14] (showing r.e.-completeness)

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The case of  $CT_{\forall\forall}^{\text{std}}$  (and with it  $CT_{\forall\forall}^{\text{dif}}$ ) is more difficult.

The critical instance is no longer relevant for all-instances termination:

**Observation:** Every rule set is in  $CT_{\mathcal{D}^*}^{\text{std}}$ .

Indeed,  $CT_{\forall\forall}^{\text{std}}$  and  $CT_{\forall\forall}^{\text{dif}}$  are no longer r.e., although the exact degree of their undecidability remains open.



# Decidable cases

# Decidable cases

The (supposed) undecidability of chase termination has motivated significant research activities for finding sufficient termination criteria:

- omega-restrictedness [Syrjänen, LPNMR 2001]
- weak-acyclicity [Fagin et al., Theo. Comp. Sci. 2005]
- lambda restrictedness [Gebser, Schaub, Thiele, LPNMR 2007]
- finite domain [Calimeri et al. ICLP 2008]
- super-weak acyclicity [Marnette, PODS 2009]
- safety [Meier, Schmidt, & Lausen, Proc. VLDB 2009]
- argument restrictedness [Lierler & Lifschitz, ICLP 2009]
- joint acyclicity [MK & Rudolph, IJCAI 2011]
- acyclic graph of rule dependencies [Baget et al., Artif. Intell. 2011]
- $\Omega$ -acyclicity [Greco, Spezzano, & Trubitsyna, ICLP 2012]
- model faithful & model summarising acyclicity [Cuenca Grau et al., J. Artif. Intell. Res. 2013]

All of these criteria apply to  $CT_{\forall}^{\text{sk}}$ .

# Chase variants in practice

Standard chase rule applications are harder than skolem chase rule applications:

- Skolem chase: guess match and verify absence of conclusions – NP
- Standard chase: guess match and verify non-entailment of conclusion –  $\text{NP}^{\text{NP}}$  ( $= \Sigma_p^2$ )

# Chase variants in practice

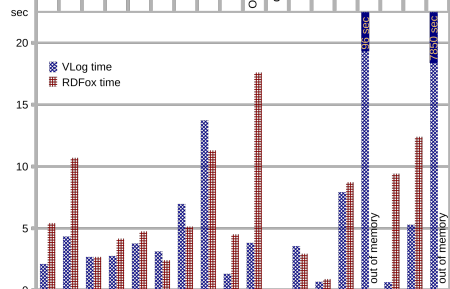
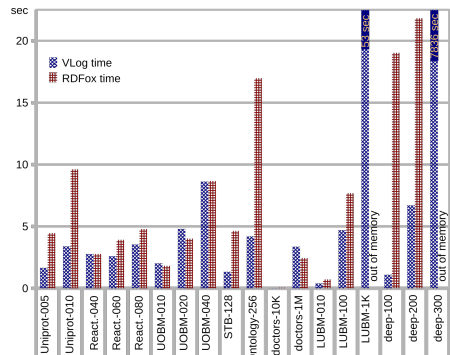
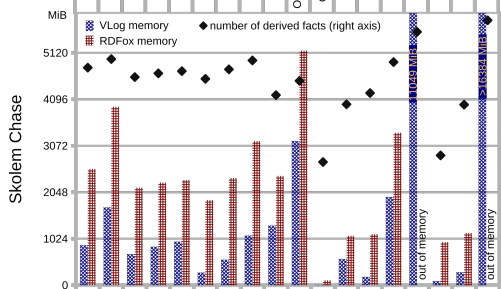
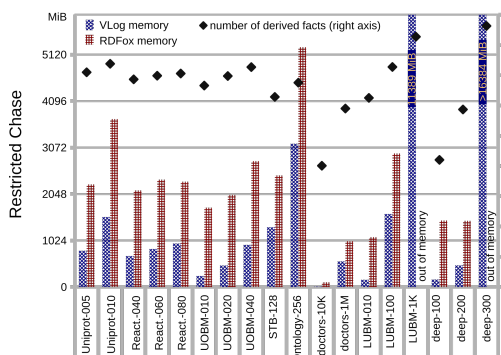
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Nevertheless, the standard chase is implemented by many existential rule engines:

- DEMo [Pichler & Savenkov, VLDB'09]
- RDFox [Motik et al., AAAI'14]
- Llunatic [Geerts et al., VLDB'14]
- Pegasus [Meier, VLDB'14]
- PDQ [Benedikt, Leblay, & Tsamoura, VLDB'14; VLDB'15]
- Graal [Baget et al., RuleML'15]
- VLog [Urbani, Jacobs, & MK, AAAI'16; Urbani et al., IJCAR'18]

See [Benedikt et al., PODS'17] and [Urbani et al., IJCAR'18] for recent benchmarks.



Part 3:  
Expressivity

# Expressive power

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Follow-up question: what is “expressive power”?

~ descriptive, not computational complexity

**Definition:** Consider a finite signature  $\mathbf{R}^{\text{EDB}}$  of (extensional) database relations.

An **abstract query** over  $\mathbf{R}^{\text{EDB}}$  is a set  $\mathcal{D}$  of concrete databases over  $\mathbf{R}^{\text{EDB}}$ .

A set of rules  $\Sigma$  and BCQ  $q$  **realise**  $\mathcal{D}$  if, for every database  $\mathcal{D}$  over  $\mathbf{R}^{\text{EDB}}$ ,

$$\mathcal{D}, \Sigma \models q \text{ exactly if } \mathcal{D} \in \mathcal{D}.$$

where  $\Sigma$  and  $q$  may use additional relations beyond  $\mathbf{R}^{\text{EDB}}$ .

~ Expressivity = abstract queries that can be realised (by a rule fragment)

**Note:** This is closer to the program view than to the ontology view.

# A note on Datalog

Distinguishing **extensional (EDB)** and **intensional (IDB)** predicates is common for Datalog.

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## Datalog as Second-Order Language

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**We only use EDB predicates to define expressivity. Everything here is first order.**

# Data complexity for $CT_{\forall}^{sk}$

Marnette [PODS 2009] showed the following general result:

**Theorem:** For every  $\Sigma \in CT_{\forall}^{sk}$  and concrete database  $\mathcal{D}$ , the skolem chase over  $\Sigma$  and  $\mathcal{D}$  is polynomial in the size of  $\mathcal{D}$ .

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**Proof:** There is a tuple-preserving mapping  $h$  from any database  $\mathcal{D}$  to the critical instance  $\mathcal{D}^*$ :

- $h(c) = c$  for all constants in  $\Sigma$
- $h(c) = *$  for all other constants

$h$  extends to function terms by setting  $h(f(c)) = f(h(c))$ .

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$h$  extends to function terms by setting  $h(f(c)) = f(h(c))$ .

This extended mapping satisfies:  $r(\mathbf{t}) \in \text{chase}_{\text{sk}}(\Sigma, \mathcal{D})$  implies  $r(h(\mathbf{t})) \in \text{chase}_{\text{sk}}(\Sigma, \mathcal{D}^*)$ .

# Data complexity for $CT_{\forall}^{\text{sk}}$

Marnette [PODS 2009] showed the following general result:

**Theorem:** For every  $\Sigma \in CT_{\forall}^{\text{sk}}$  and concrete database  $\mathcal{D}$ , the skolem chase over  $\Sigma$  and  $\mathcal{D}$  is polynomial in the size of  $\mathcal{D}$ .

The data complexity of BCQ entailment over  $CT_{\forall}^{\text{sk}}$  is PTime-complete.

**Proof:** There is a tuple-preserving mapping  $h$  from any database  $\mathcal{D}$  to the critical instance  $\mathcal{D}^*$ :

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The only data-dependent part are the additional constants in  $\mathcal{D}$ : the number of distinct terms and tuples is polynomial in this respect. □

# From $\text{CT}_{\forall}^{\text{sk}}$ to Datalog

The previous insight can be taken further

[MK & Rudolph IJCAI'11; Zhang, Zhang & You AAI'15]

**Theorem:** For every  $\Sigma \in \text{CT}_{\forall}^{\text{sk}}$  and BCQ  $q$ , there is a set of Datalog rules  $\Sigma'$  and BCQ  $q'$  such that  $\{\mathcal{D} \mid \mathcal{D}, \Sigma \models q\} = \{\mathcal{D} \mid \mathcal{D}, \Sigma' \models q'\}$ .

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**Proof (idea):** The terms in any skolem chase over  $\Sigma$  are bounded in size.

One can “flatten” such terms by increasing the arity of predicates, e.g.,

$$p(f(a, b)) \mapsto \hat{p}(f, a, b)$$

Arities must be large enough to accommodate all possible terms, but unused positions can be filled by a special constant  $\square$ , e.g.,

$$q(f(s(a, b), t(c, d))) \mapsto \hat{q}(f, s, a, b, t, c, d)$$

$$q(f(a, g(b))) \mapsto \hat{q}(f, a, \square, \square, g, b, \square)$$

It is easy to apply these replacements to rules and queries. □

# Discussion

**Summary:** Essentially all known chase termination criteria recognise fragments of existential rules that are basically syntactic simplifications of Datalog.

- Existential rules are usually more concise (flattening may incur exponential predicate arity)
- Combined complexity is accordingly higher (typically 2ExpTime-complete)
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## Thesis

Previous research on chase termination is best motivated from an ontological view, while not leading to significant advances for using rules as declarative programs/queries.

# Beyond P

Surprisingly, this severe restriction in expressivity is specific to the skolem chase:

**Theorem:** There is a rule set  $\Sigma \in \text{CT}_{\forall\forall}^{\text{dif}}$  and a BCQ  $q$  that express a non-elementary Boolean query.

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**Proof:** We reduce from the following non-elementary decision problem:

Input: A Turing machine  $\mathcal{M}$  and a number  $k$

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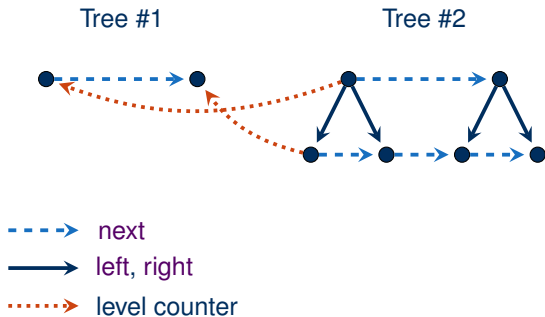
Three ingredients are needed:

1. Rules that receive an input chain  $\text{first}(e_0)$ ,  $\text{next}(e_0, e_1)$ ,  $\dots$ ,  $\text{next}(e_{k-1}, e_k)$ ,  $\text{last}(e_k)$  and construct a  $k$ -exponentially long chain (interesting)
2. Rules that simulate a Turing machine in time and space bounded by this chain (boring)
3. Rules that ensure termination even on malformed inputs (interesting)



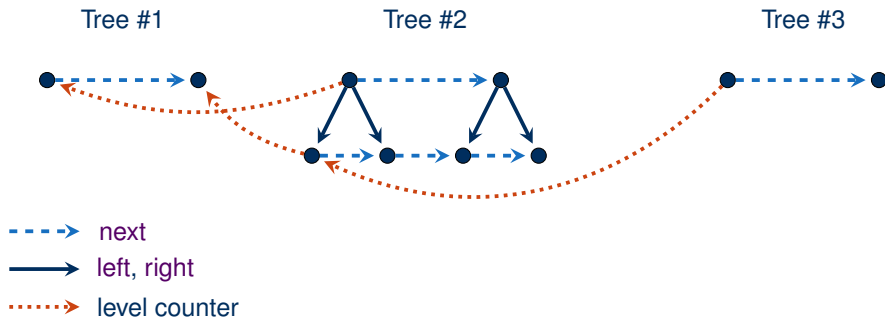
# Building a long chain

We construct a series of  $k$  full binary trees of depth  $2, 2^2, 2^{2^2}, \dots$  (we omit the roots)



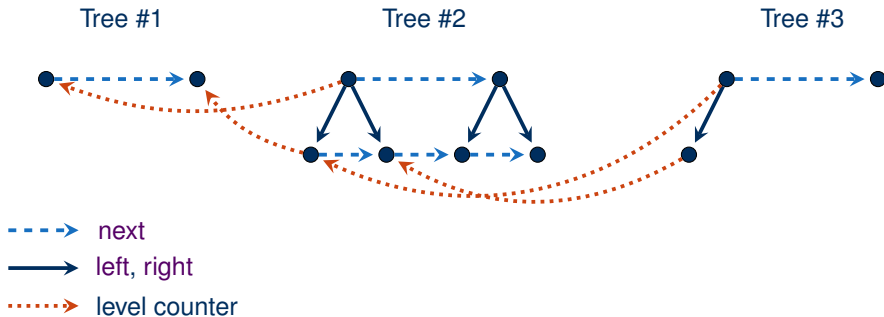
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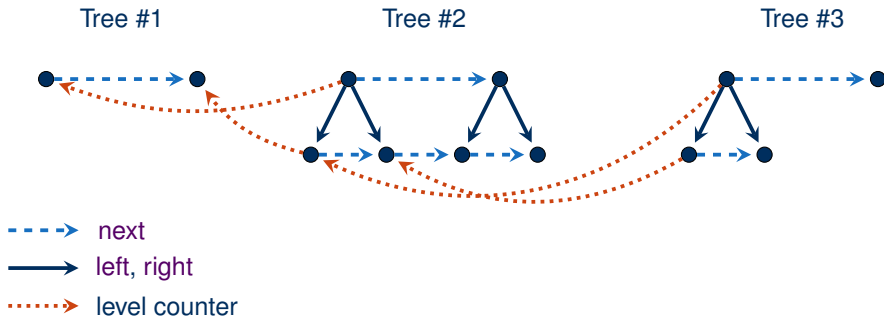
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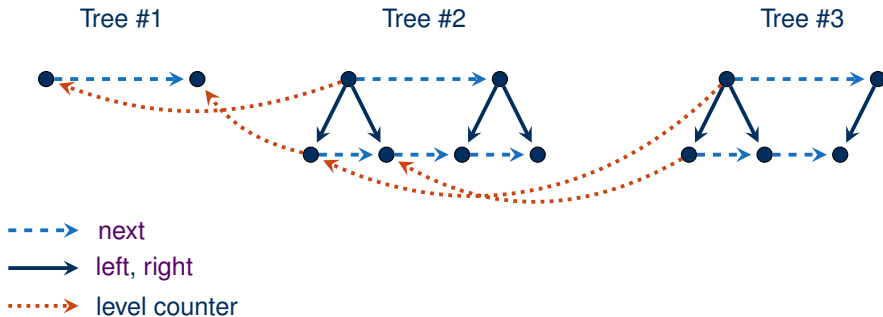
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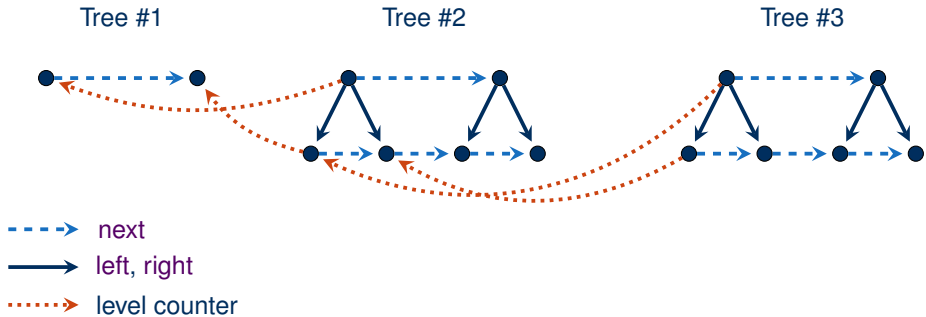
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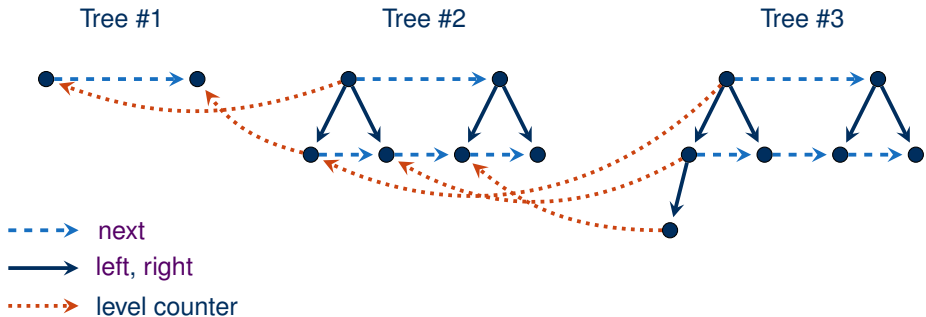
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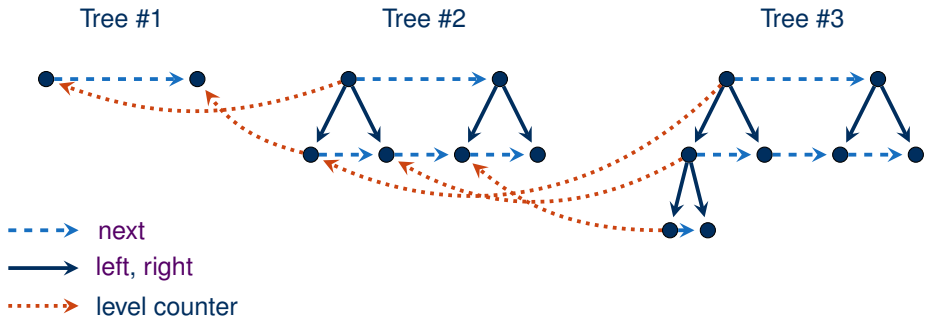
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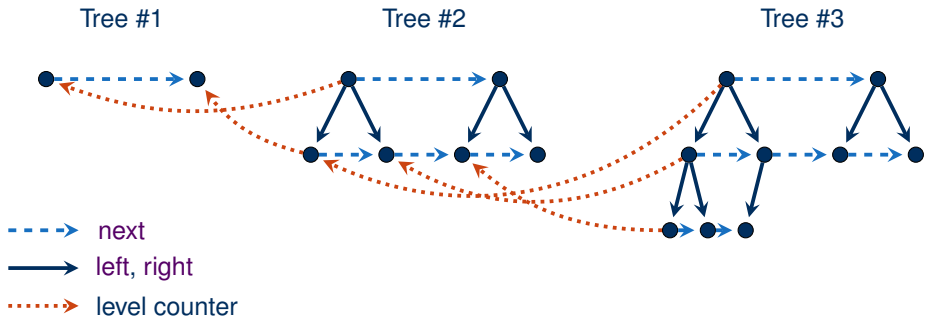
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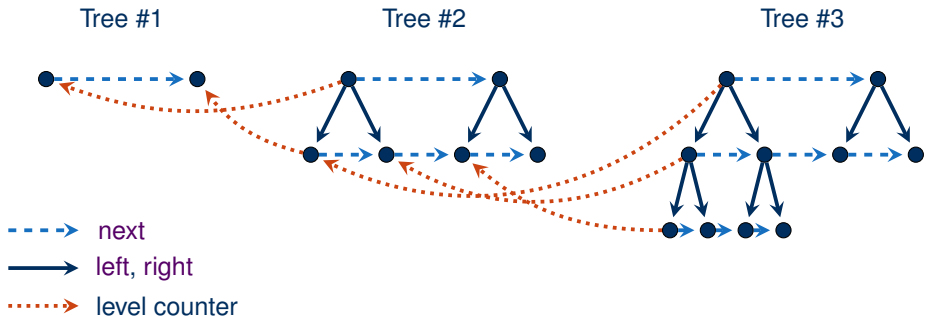
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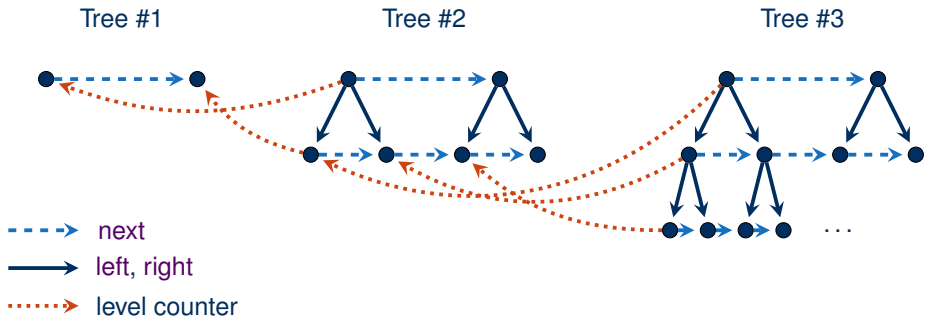
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And here are the rules:

$$\text{first}(v) \rightarrow \exists x. \text{start}(x, x, v) \wedge \text{end}(x)$$

$$\text{start}(x, u, v) \wedge \text{end}(u) \wedge \text{next}(v, v') \rightarrow \exists y_1, y_2. \text{start}(y_1, x, v') \wedge \text{succ}(y_1, y_2) \wedge \text{end}(y_2)$$

$$\text{start}(x, u, v) \wedge \text{succ}(u, u') \rightarrow \exists y. \text{left}(x, y) \wedge \text{start}(y, u', v)$$

$$\text{left}(x, y) \rightarrow \exists y'. \text{right}(x, y') \wedge \text{succ}(y, y')$$

$$\text{right}(x, y) \wedge \text{succ}(x, x') \rightarrow \exists y'. \text{left}(x', y') \wedge \text{succ}(y, y')$$

$$\text{end}(x) \wedge \text{right}(x, y) \rightarrow \text{end}(y)$$

# Ensuring termination

The chain construction works even in the skolem chase for inputs

**first**( $e_0$ ), **next**( $e_0, e_1$ ),  $\dots$ , **next**( $e_{k-1}, e_k$ ), **last**( $e_k$ ).

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But every chase fails to terminate if the next-graph is cyclic.

We can ensure termination for Datalog-first chase using rules as follows:

$$\begin{array}{ll} \text{next}(x, y) \rightarrow \text{tnext}(x, y) & \text{cycle detection} \\ \text{tnext}(x, y) \wedge \text{next}(y, z) \rightarrow \text{tnext}(x, z) & \\ \text{tnext}(x, x) \wedge \text{adom}(u) \wedge \text{adom}(v) \wedge \text{adom}(w) \rightarrow \text{start}(u, v, w) & \text{flooding rule(s)} \\ \dots & \end{array}$$

- The rules are applied before applying existential rules (Datalog first)
- The flooding rules stop the introduction of new nulls
- Further rules are needed:  
also flood other atoms; extract **adom** from other atoms

## Beyond P in standard chase

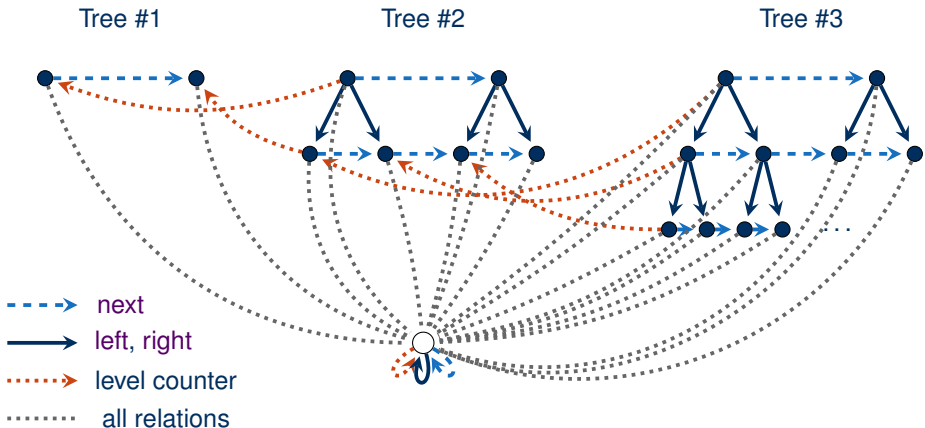
**Problem:** We rely on the Datalog-first chase. With standard chase, cycles will eventually be found (fairness), but flooding rules might be applied too late to prevent new nulls



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## Beyond P in standard chase

**Problem:** We rely on the Datalog-first chase. With standard chase, cycles will eventually be found (fairness), but flooding rules might be applied too late to prevent new nulls

**Solution:** Create an **emergency break** that can be activated with just one rule application.

- Use a new unary predicate **real**, expected for all domain elements used so far, and require it for all elements in all rules
- Introduce a unique “break” element that is almost a critical instance, but not **real**:

$$\rightarrow \exists x. \text{break}(x) \wedge \underbrace{\text{start}(x, x, x) \wedge \text{succ}(x, x) \wedge \text{end}(x) \wedge \dots}_{\text{all possible facts over } x, \text{ except } \text{real}(x)}$$

- Connect new nulls to this “break” element upon creation  
(doing this properly requires a rewriting to split **start** into several binary relations)
- To activate the emergency break, use a rule like

$$\text{tnext}(x, x) \wedge \text{break}(u) \rightarrow \text{real}(u)$$

# Beyond P in standard chase

## Summary

- The emergency-break technique ensures termination by relying on fairness
- Hence, there are no runtime guarantees, but the rules are in  $CT_{\forall\exists}^{\text{std}}$

**Theorem:** There is a rule set  $\Sigma \in CT_{\forall\exists}^{\text{std}}$  and a BCQ  $q$  that express a non-elementary Boolean query.

**Note:** Very recent research found that fairness is never needed for termination when rules have a single head atom only [Gogacz et al. CoRR abs/1901.03897, 2019]

Part 3:  
Expressivity in P

# Back to P

High expressivity can be a plus, but many practical problems are in P anyway.

Can the terminating skolem chase solve every problem in P?

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Can the terminating skolem chase express every homomorphism-closed query in P?

No, but this is less obvious:

- Dawar & Kreutzer discovered a homomorphism closed, polynomial time query that cannot be expressed in Datalog [ICALP'08]



# The Dawar & Kreutzer query

**Input:** A directed graph  $G$  with two distinguished vertices  $s$  and  $t$

**Question:**

Is  $G$  cyclic, or is there a simple path from  $s$  to  $t$  of length  $2^{2^{n^2}}$  for some  $n \in \mathbb{N}$ ?

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**Definition:** The **DK query**  $\mathcal{Q}_{\text{DK}}$  is the abstract Boolean query containing exactly those concrete databases that encode an instance of this decision problem using a binary relation **edge** and constant symbols **s** and **t**.

This query is:

- closed under homomorphisms
- solvable in polynomial time
- not expressible in Datalog, hence not expressible in  $\text{CT}_{\forall}^{\text{sk}}$

# DK in Datalog-first chase

We find that Datalog-first is more powerful than skolem even for polynomial problems:

**Theorem:** There is a rule set  $\Sigma \in \text{CT}_{\forall\forall}^{\text{dlf}}$  and BCQ  $q$  that realise the DK query.  
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**Proof:** The rule set consists of three parts:

- (A) Rules to create a “yardstick” to measure the length of single paths starting in  $s$
- (B) Rules to derive simple arithmetic relations on this yardstick
- (C) Rules to stop the chase if there is a cycle

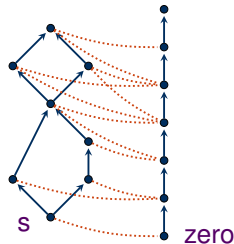
Our construction is inspired by a previous encoding by Rudolph & Thomazo [IJCAI'15].

## (A) The Yardstick

$\rightarrow \exists x.\text{zero}(x)$	start chain
$\text{zero}(x) \rightarrow \text{dist}(s, x)$	distance of $s$ to $s$
$\text{dist}(v, x) \rightarrow \exists x'.\text{succ}(x, x')$	extend chain
$\text{dist}(v_1, x_1) \wedge \text{edge}(v_1, v_2) \wedge \text{succ}(x_1, x_2) \rightarrow \text{dist}(v_2, x_2)$	distance of $v_2$ to $s$

This creates a chain of **succ** relations, starting from a **zero** element

- The chain is only extended if its last elements are needed to measure some distance
- There will only be one chain for the whole graph



## (B) The Arithmetic

$$\text{zero}(x) \wedge \text{dist}(v, y) \rightarrow \text{add}(x, y, y) \wedge \text{mul}(x, y, x)$$

$$\text{add}(x, y, z) \wedge \text{succ}(x, x') \wedge \text{succ}(z, z') \rightarrow \text{add}(x', y, z')$$

$$\text{mul}(x, y, z) \wedge \text{succ}(x, x') \wedge \text{add}(z, y, z') \rightarrow \text{mul}(x', y, z')$$

$$\text{zero}(x) \wedge \text{succ}(x, x') \rightarrow \text{exp}(x, x')$$

$$\text{exp}(x, y) \wedge \text{succ}(x, x') \wedge \text{add}(y, y, y') \rightarrow \text{exp}(x', y')$$

$$\text{mul}(x, x, y) \wedge \text{exp}(y, y') \wedge \text{exp}(y', z) \wedge \text{dist}(t, z) \rightarrow \text{goal}$$

The arithmetic is easy to implement:

- $\text{add}(x, y, z)$  means “ $x + y = z$ ”;  $\text{mul}(x, y, z)$  means “ $x * y = z$ ”;  $\text{exp}(x, y)$  means “ $2^x = y$ ”
- The last rule recognises the query condition

## (C) The Cycle Stopper

$$\text{edge}(v_1, v_2) \rightarrow \text{path}(v_1, v_2)$$

$$\text{edge}(v_1, v_2) \wedge \text{path}(v_2, v_3) \rightarrow \text{path}(v_1, v_3)$$

$$\text{path}(v, v) \rightarrow \text{goal}$$

$$\text{dist}(v, x) \wedge \text{goal} \rightarrow \text{succ}(x, x)$$

Rules for cycle detection and ensuring termination

- The first three rules derive the goal on cycles
- The last rule will prevent new nulls to be created for extending the Yardstick

This completes the construction. It is easy to verify the claim. □

## How about the standard chase?

The rules we used to realise the DK query are in  $CT_{\forall\forall}^{\text{dif}}$  but not in  $CT_{\forall\forall}^{\text{std}}$ :

- The flooding rule  $\text{dist}(v, x) \wedge \text{goal} \rightarrow \text{succ}(x, x)$  might be applied too late



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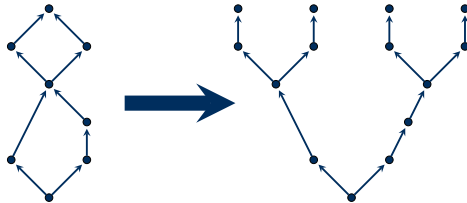
But we can do better:

**Theorem:** There is a rule set  $\Sigma \in CT_{VV}^{std}$  and BCQ  $q$  that realise the DK query. The standard chase on  $\Sigma$  is at most exponential in the size of the input database.

# DK in exponential standard chase

## Basic idea:

- Replace the Yardstick by a **tree** that is an unravelled version of the graph:

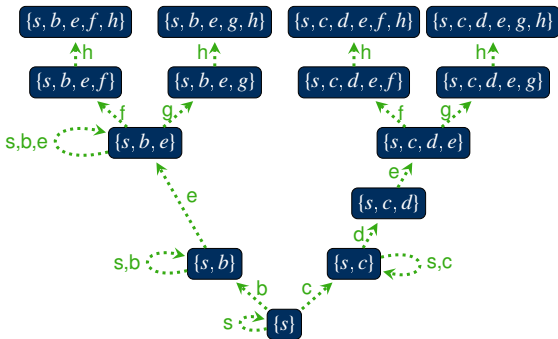
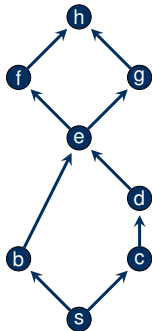


- It is then easy to use tree nodes as distances and to do arithmetic on the tree.

# DK in exponential standard chase

Elements of the tree correspond to simple paths, considered as sets of vertices

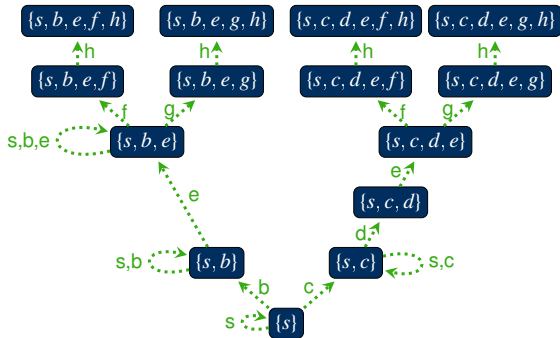
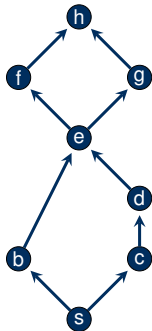
- **ins**( $v, x, y$ ) means " $\{v\} \cup x = y$ "; especially **ins**( $v, x, x$ ) means " $v \in x$ "
- **subset**( $v, w$ ) means " $x \subset y$ "



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- **subset**( $v, w$ ) means " $x \subset y$ "



Termination trick:

- new sets are only created by adding elements that are not included yet
- relevant facts **ins**( $v, x, x$ ) derived in Datalog, before adding more elements

# DK in exponential standard chase

And here are the rules for building the tree:

$$\rightarrow \exists x. \text{zero}(x) \quad (1)$$

$$\text{zero}(x) \rightarrow \text{dist}(s, x) \wedge \text{ins}(s, x, x) \wedge \text{done}(x) \quad (2)$$

$$\text{dist}(v_1, x_1) \wedge \text{edge}(v_1, v_2) \wedge \text{done}(x_1) \rightarrow \exists x_2. \text{ins}(v_2, x_1, x_2) \wedge \text{subset}(x_2, x_2) \quad (3)$$

$$\text{subset}(x_1, x_2) \wedge \text{ins}(v, x_0, x_1) \rightarrow \text{ins}(v, x_2, x_2) \wedge \text{subset}(x_0, x_2) \quad (4)$$

$$\text{subset}(x_1, x_2) \wedge \text{zero}(x_1) \rightarrow \text{ins}(s, x_2, x_2) \wedge \text{done}(x_2) \quad (5)$$

$$\text{dist}(v_1, x_1) \wedge \text{edge}(v_1, v_2) \wedge \text{ins}(v_2, x_1, x_2) \rightarrow \text{dist}(v_2, x_2) \wedge \text{succ}(x_1, x_2) \quad (6)$$

When a new set is created (3), we recursively copy elements from all subsets (4), and only consider the new set “done” when the empty set is reached (5).

(the rest of the proof is as before)

# Discussion and Questions



# What we have shown

## Notation:

- We refine termination classes by adding a function class, e.g.,

$CT_{\forall\forall}^{\text{std}}(\text{poly})$  means “rules on which the standard chase terminates in polynomial time (for all strategies and instances)”

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- Any restricted chase implementation is sound and complete for  $CT_{VV}^{\text{std}}$ , not just for decidable fragments thereof
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**Lesson 3:** A Datalog-first chase strategy might have worst-case runtime benefits

- Other ways of slightly regulating chase strategies might also work
- Even (truly) random strategies might lead to almost certain termination

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- **Decidable termination criteria** Which sufficient criteria can detect termination beyond skolem?
- **Termination on restricted database classes** Should we consider “not-quite-universal termination” that imposes requirements on the database? Which?
- **Practical applications** How to exploit beyond-skolem expressive power in practice?





One can still learn new things about the chase . . .

## Lessons learnt

- Skolem chase is just Datalog in disguise – standard chase is more
- Existing rule reasoners could be used ways not considered yet
- Rule reasoners should offer some chase strategy control

. . . but there are many open questions

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