

# Termination Analysis of Logic Programs

Sergio Greco and Cristian Molinaro

DIMES, University of Calabria, Italy

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# Logic Program Termination Analysis

## 1. What kind of **Logic Programs**?

- 1 Rules with function symbols.
- 2 Existential rules.

**Many applications** in knowledge representation, logic programming, and databases: answer set programming, ontological query answering, data exchange, etc.

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- Establishing termination is undecidable.

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## 2. **Termination Analysis**

- The evaluation of such programs might not terminate.
- Establishing termination is undecidable.
- **Termination Criteria**: sufficient conditions guaranteeing termination.

# Outline

- Part I: Logic Programs with Function Symbols
  - ▶ Syntax and Semantics
  - ▶ Termination Criteria
- Part II: Existential Rules
  - ▶ The Chase and the Termination Problem
  - ▶ Termination Criteria
  - ▶ Adding EGDs

# Part I

## Logic Programs with Function Symbols

# Context and Motivations

## ● **Function Symbols**

- ▶ Make modeling easier and the resulting encodings more readable and concise.
- ▶ Increase the expressive power.
- ▶ Allow us to overcome the inability of handling infinite domains.

# Context and Motivations

- **Function Symbols**

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- ▶ Increase the expressive power.
- ▶ Allow us to overcome the inability of handling infinite domains.

- **Problem:** Program evaluation might not terminate and it is undecidable whether the evaluation terminates.



# Top-down Evaluation

- Apt, Bezem. Acyclic programs. ICLP (1990).
- Bol, Apt, Klop. An analysis of loop checking mechanism for logic programs. TCS (1991).
- Sagiv. A termination test for logic programs. ICLP (1991).
- Apt, Pedreschi. Reasoning about termination of pure Prolog programs. I&C (1993).
- De Schreye, Decorte. Termination of logic programs: The never-ending story. JLP (1994).
- Lindenstrauss, Sagiv. Automatic termination analysis of logic programs. ICLP (1997).
- Codish, Taboch. A semantic basis for the termination analysis of logic programs. JLP (1999).
- Ohlebusch. Termination of logic programs: Transformational methods revisited. AAEECC (2001).
- Pedreschi, Ruggieri, Smaus. Classes of terminating logic programs. TPLP (2002).
- Bonatti. Reasoning with infinite stable models. AIJ (2004).
- Serebrenik, De Schreye. On termination of meta-programs. TPLP (2005).

# Top-down Evaluation

- Bruynooghe, Codish, Gallagher, Genaim, Vanhoof. Termination analysis of logic programs through combination of type-based norms. ACM TOCL (2007).
- Nguyen, Giesl, Schneider-Kamp, De Schreye. Termination analysis of logic programs based on dependency graphs. LOPSTR (2007).
- Baselice, Bonatti, Crisculo. On finitely recursive programs. TPLP (2009).
- Schneider-Kamp, Giesl, Nguyen. The dependency triple framework for termination of logic programs. LOPSTR (2009).
- Schneider-Kamp, Giesl, Serebrenik, Thiemann. Automated termination proofs for logic programs by term rewriting. ACM TOCL (2009).
- Nishida, Vidal. Termination of narrowing via termination of rewriting. Appl. Algebra Eng. Commun. Comput. (2010).
- Schneider-Kamp, Giesl, Stroder, Serebrenik, Thiemann. Automated termination analysis for logic programs with cut. TPLP (2010).
- Eiter, Simkus. FDNC: Decidable nonmonotonic disjunctive logic programs with function symbols. ACM TOCL (2010).
- Voets, De Schreye. Non-termination analysis of logic programs with integer arithmetics. TPLP (2011).

# Bottom-up Evaluation

- Shmueli. Decidability and Expressiveness of Logic Queries. PODS (1987).
- Ramakrishnan, Bancilhon, Silberschatz. Safety of Recursive Horn Clauses With Infinite Relations. PODS (1987).
- Kifer, Ramakrishnan, Silberschatz. An Axiomatic Approach to Deciding Query Safety in Deductive Databases. PODS (1988).
- Krishnamurthy, Ramakrishnan, Shmueli. A framework for testing safety and effective computability. SIGMOD (1988), JCSS (1996).
- Chomicki. A decidable class of logic programs with function symbols. TR 1990.
- Chomicki, Imielinski. Finite representation of infinite query answers. TODS (1993).

# Bottom-up Evaluation

- Syrjanen. Omega-restricted logic programs. LPNMR (2001).
- Gebser, Schaub, Thiele. Gringo : A new grounder for answer set programming. LPNMR (2007).
- Calimeri, Cozza, Ianni, Leone. Computable Functions in ASP: Theory and Implementation. ICLP (2008).
- Lierler, Lifschitz. One more decidable class of finitely ground programs. ICLP (2009).
- Greco, Spezzano, Trubitsyna. On the Termination of Logic Programs with Function Symbols. ICLP (2012)
- Calautti, Greco, Trubitsyna. Detecting decidable classes of finitely ground logic programs with function symbols. PPDP (2013).
- Greco, Molinaro, Trubitsyna. Logic programming with function symbols: Checking termination of bottom-up evaluation through program adornments. TPLP (2013).
- Greco, Molinaro, Trubitsyna. Bounded Programs: A New Decidable Class of Logic Programs with Function Symbols. IJCAI (2013).
- Calautti, Greco, Molinaro, Trubitsyna. Checking Termination of Logic Programs with Function Symbols through Linear Constraints. RuleML (2014).
- Calautti, Greco, Spezzano, Trubitsyna. Checking Termination of Bottom-Up Evaluation of Logic Programs with Function Symbols. TPLP (2014).

# Top-down vs. Bottom-up Evaluation

## Example

$$p(X) \leftarrow p(X).$$

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We consider **bottom-up** evaluation.

# Bottom-up Evaluation

## Example

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len([a,b,c], 0).  
len(Tail, s(N)) ← len(list(Head, Tail), N).
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Bottom-up evaluation:

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len([b,c], s(0)) ← len([a,b,c], 0) yields len([b,c], s(0))
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len([c], s(s(0))) ← len([b,c], s(0))    yields len([c], s(s(0)))
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**Fixpoint, the evaluation TERMINATES.**

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		⋮		

**The evaluation does NOT terminate.**

# Termination Criteria

- (Decidable) Sufficient conditions guaranteeing the bottom-up evaluation termination.
- The use of function symbols is restricted.

## “Terminating” Programs

We say that a program  $P$  is ***terminating*** iff **the evaluation of  $P \cup D$  terminates for every finite set of facts  $D$ .**

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## Termination Criteria

Define a decidable condition  $C$  such that for every program  $P$

**$P$  satisfies  $C \Rightarrow P$  is terminating.**



# Termination Criteria

- $\omega$ -restricted programs [Syr01]
- $\lambda$ -restricted programs [GST07]
- *Finite domain* programs [CCIL08]
- *Argument-restricted* programs [LL09]
- *Safe* and  $\Gamma$ -*acyclic* programs [CGST14]
- *Mapping-restricted* programs [CGT13]
- *Bounded* programs [GMT13b]
- *Rule-* and *cycle-bounded* programs [CGMT14]
- *Program Adornment* technique [GMT13a]

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- *Program Adornment* technique [GMT13a]
- *Size-restricted* programs, IJCAI 2015, talk on Wed 29<sup>th</sup> afternoon!

# Syntax: Datalog with Function Symbols

## Definition

We are given (pairwise disjoint) sets of **constants**, **variables**, **function symbols** (with arity  $> 0$ ), and **predicates** (with arity  $\geq 0$ ).

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- A **term** is either a constant, a variable, or of the form  $f(t_1, \dots, t_m)$ , where  $f$  is a function symbol of arity  $m$  and the  $t_i$ 's are terms.

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- A (*Datalog*) **rule** is of the form

$$\underbrace{A_0}_{\text{head}} \leftarrow \underbrace{A_1, \dots, A_n}_{\text{body}}$$

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- A (*Datalog*) **program** is a finite set of Datalog rules.

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## Example (**Safe** program)

$$p(f(X), Y) \leftarrow q(X), r(Y).$$

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Function symbols are **uninterpreted** (they are not evaluated).

# Syntax: Datalog with Function Symbols

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The **arguments** of a program  $P$  are expressions of the form  $p[i]$  where  $p$  is a predicate appearing in  $P$  and  $1 \leq i \leq \text{arity}(p)$ .

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## Example

$$\begin{aligned} p(X, Y) &\leftarrow b(X, Y). \\ q(f(X)) &\leftarrow p(X, Y). \end{aligned}$$

The **arguments** of this program are  $\mathbf{b[1]}$ ,  $\mathbf{b[2]}$ ,  $\mathbf{p[1]}$ ,  $\mathbf{p[2]}$ , and  $\mathbf{q[1]}$ .

# Termination Criteria

## $\lambda$ -Restricted Programs [GST07]

**Basic Idea:** Assign a level (i.e., an integer)  $\lambda(p)$  to each predicate  $p$  so that all head variables in rules defining  $p$  are bound by predicates  $p'$  with strictly lower level.

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**The program is  $\lambda$ -restricted.**

### Example

$$p(X) \leftarrow p(X).$$

No function symbols  $\Rightarrow$  **The evaluation always terminates.**

$\lambda(p) > \lambda(p) \Rightarrow$  **The program is not  $\lambda$ -restricted.**

# Finite Domain Programs [CCIL08]

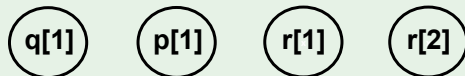
## Argument Graph

It describes the propagation of values among arguments.

- the nodes are the arguments of the program, and
- there is an edge from  $p[i]$  to  $q[j]$  if there is a rule where a term is propagated from  $p[i]$  to  $q[j]$ .

## Example (Argument Graph)

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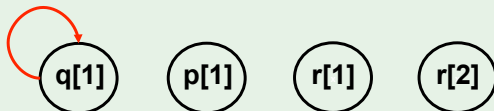
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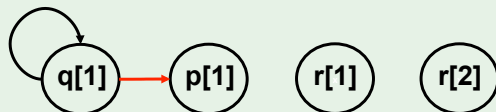
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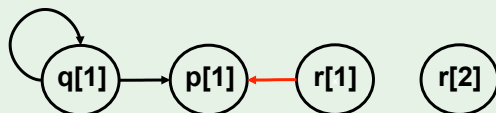
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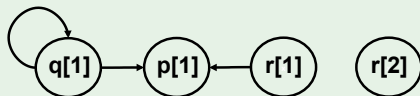
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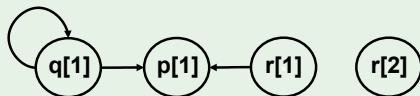


Finite domain arguments:

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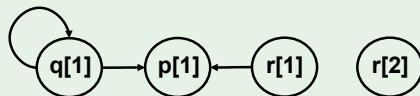
Finite domain arguments:

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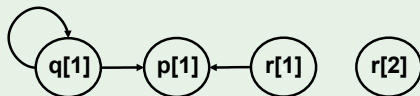
- $r[1]$  and  $r[2]$ , as they appear in no head.
- $q[1]$ , as the term in the head of the 1st rule is a subterm of that in the body.



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## Example

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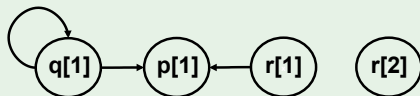
Finite domain arguments:

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# Finite Domain Programs [CCIL08]

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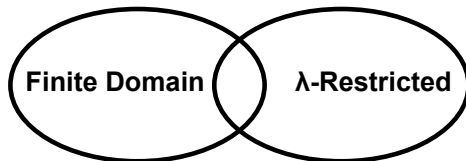


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**All arguments are finite domain  $\Rightarrow$  The program is finite domain.**

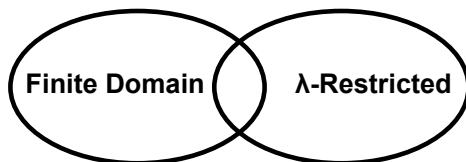
# Relative Expressivity



## Theorem

*$\lambda$ -Restricted*  $\not\equiv$  *Finite Domain*.

# Relative Expressivity



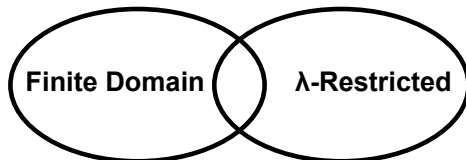
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## Argument Restriction [LL09]

**Basic idea:** assign to each argument an upper bound of the depth of terms that may occur in that argument.

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$$d(X, X) = 0$$

$$d(X, f(t_1, \dots, t_m)) = 1 + \max_{1 \leq i \leq m : t_i \text{ contains } X} d(X, t_i).$$

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## Example

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## Example

$$d(X, \mathbf{f}(X, \mathbf{g}(\mathbf{X}), Y)) = 2$$

$$d(Y, \mathbf{f}(X, \mathbf{g}(X), \mathbf{Y})) = 1$$

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### Example

$$p(f(X)) \leftarrow q(X)$$

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$$\begin{array}{l} p(\overset{1}{f}(x)) \leftarrow \overset{0}{q}(x) \\ \overset{0}{q}(x) \leftarrow p(\overset{1}{f}(x)) \end{array}$$

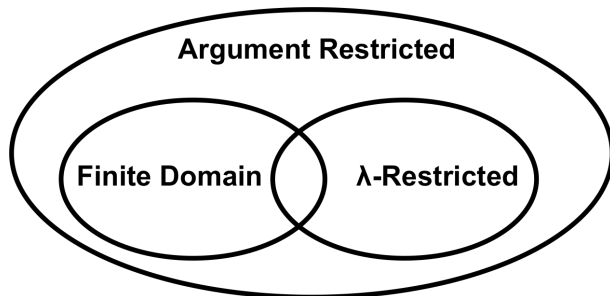
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**The program is argument-restricted**

# Relative Expressivity



## Theorem

- *Finite Domain*  $\subsetneq$  *Argument Restricted*.
- *$\lambda$ -Restricted*  $\subsetneq$  *Argument Restricted*.

# Argument Restriction [LL09]

Simple and easy (polynomial time) to compute.



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Simple and easy (polynomial time) to compute.

**Limitation:** No distinction between different function symbols.

## Example

$$p(\mathbf{f}(f(X))) \leftarrow p(\mathbf{g}(X))$$

We need to find a function  $\phi$  such that

$$\phi(p[1]) \geq \phi(p[1]) + 1$$

**No such  $\phi$  exists.** The program is **not argument-restricted**...  
... but the program **evaluation always terminates**.

Argument restriction can be used as a starting point for more complex analysis.

# Bounded Programs [GMT13b]

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## Basic Idea:

- Start with a set  $A$  of “limited” arguments.
- Iteratively apply a (monotone) operator  $\Psi(A)$  which derives more arguments as “limited”.
- If, eventually, all arguments are derived as limited, then the program is *bounded*.

The operator relies on two tools:

- the **activation graph**, and
- the **labeled argument graph**.

# Termination Analysis Tools — Activation Graph

## Activation Graph

It describes “activation” of rules.

- the nodes are the rules of the program, and
- there is an edge from  $r_i$  to  $r_j$  iff the head of  $r_i$  unifies with some body atom of  $r_j$ .

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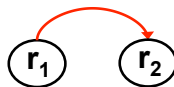
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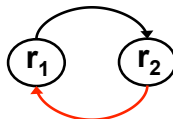
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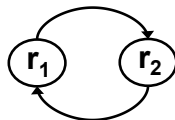
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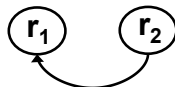
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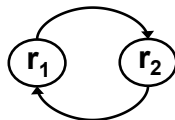
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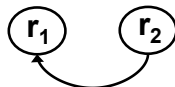
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$$r_1 : q(\mathbf{f}(X)) \leftarrow p(X)$$
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**A rule might be applied an infinite number of times only if it depends on a cycle.**

# Termination Analysis Tools — Labeled Argument Graph

## Labeled Argument Graph

It describes the propagation of values among arguments.

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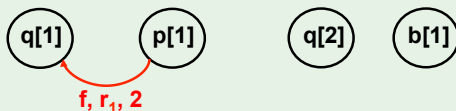
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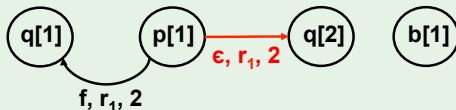
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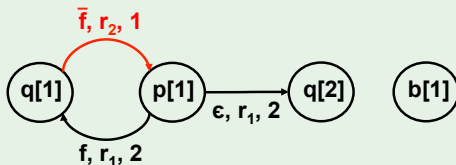
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The activation graph is also used

# Balanced, Growing, and Failing Cycles

Classification on the basis of the **first component** of the edge labels.

## Balanced / Growing / Failing Cycles

- **Balanced cycle:** a term propagated through the whole cycle remains the same.
- **Growing cycle:** a term propagated through the whole cycle grows.
- **Failing cycle:** a term propagated through the whole cycle decreases or cannot really go through the entire cycle.

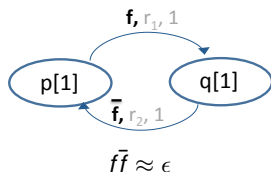
**Terms in an argument might grow infinitely only if this argument depends on a growing cycle.**

# Balanced, Growing, and Failing Cycles

## Balanced cycle

$$r_1 : q(f(X)) \leftarrow p(X)$$

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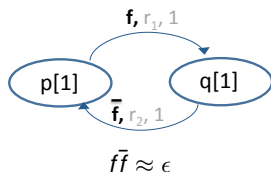
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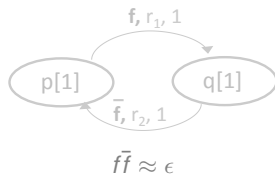
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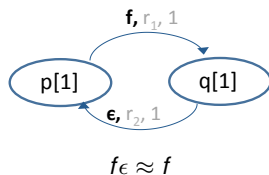
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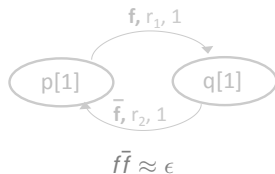
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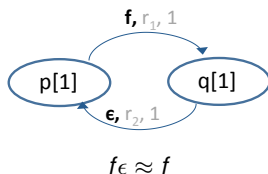
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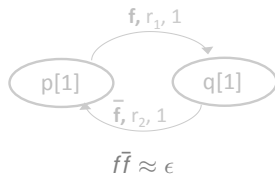
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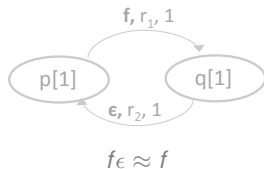
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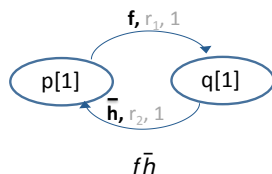
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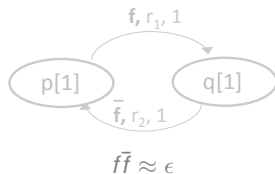


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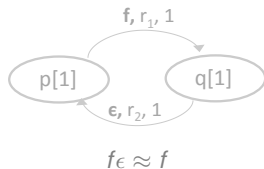
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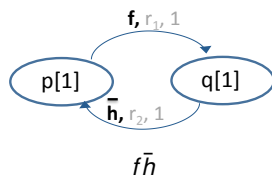
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# Active and Inactive Cycles

- Classification on the basis of the **second component** of the edge labels.
- The activation graph is also used.

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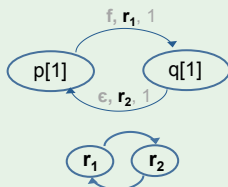
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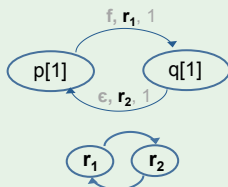
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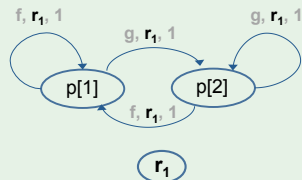
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### Example (Inactive Cycles)

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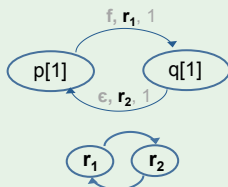
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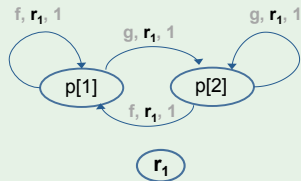
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### Example (Inactive Cycles)

$$r_1 : p(f(X), g(X)) \leftarrow p(X, X).$$



Only active cycles may be “dangerous”.

# Argument-bounded Cycles

The depth of terms in an argument might grow only if this argument depends on an **active growing cycle**.

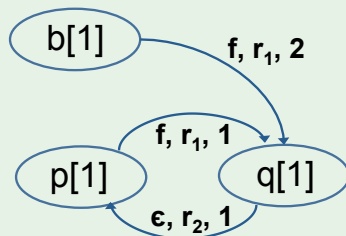
# Argument-bounded Cycles

The depth of terms in an argument might grow only if this argument depends on an **active growing cycle**.

However ...

## Example

$$\begin{aligned} r_1 : q(f(X)) &\leftarrow p(X), \mathbf{b(X)} \\ r_2 : p(X) &\leftarrow q(X). \end{aligned}$$



Since  $b[1]$  is limited, the number of values propagated in  $q[1]$  is finite.



# Twin Cycles

## Example (List length)

$r_0$ :  $\text{count}([a, b, c], 0)$ .

$r_1$ :  $\text{count}(L, s(I)) \leftarrow \text{count}([X|L], I)$ .

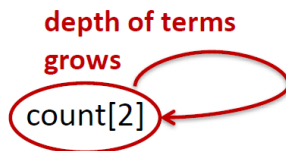
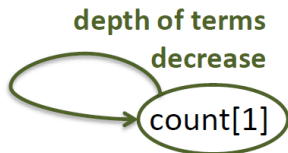
*Query goal*:  $\text{count}([], N)$ .

$\text{count}([a, b, c], 0)$

$\text{count}([b, c], s(0))$

$\text{count}([c], s(s(0)))$

$\text{count}([], s(s(s(0))))$



# Twin Cycles

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		$\text{count}([c], s(s(0)))$
<i>Query goal</i> :	$\text{count}([], N).$	$\text{count}([], s(s(s(0))))$



**The arguments may influence each other even if they do not exchange values**

The growth of  $\text{count}[2]$  is bounded by the reduction of  $\text{count}[1]$ .

# Twin Cycles

## Example (Append)

```
magic_append([a,b],[c,d]).
  magic_append(L1,L2) ← magic_append([X|L1],L2).
    append([],L,L) ← magic_append([],L).
  append([X|L1],L2,[X|L3]) ← magic_append([X|L1],L2),
    append(L1,L2,L3).
```

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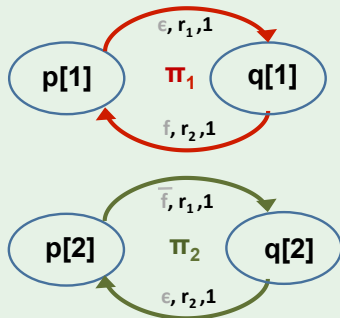
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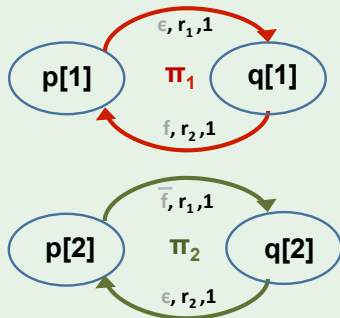
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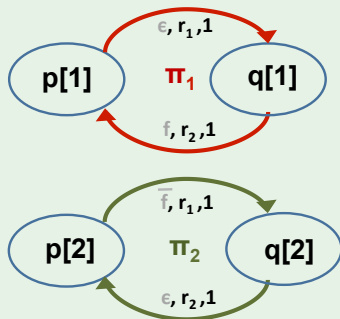
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- $\pi_1$  and  $\pi_2$  are **twin cycles**.
- The **growth** of values in  $\pi_1$  is **bounded by the reduction** in  $\pi_2$ .



# Bounded Programs

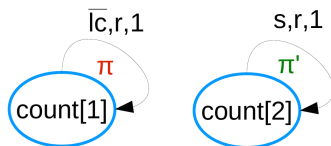
- Start with a set  $A$  of limited arguments.
- Then, add an argument  $p[i]$  if, for every cycle  $\pi$  on which  $p[i]$  depends:
  - 1  $\pi$  is not active or not growing;
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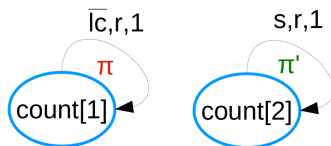
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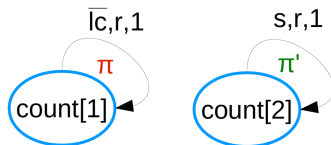
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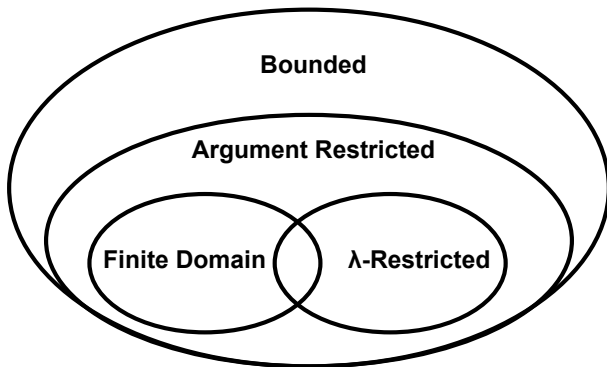
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- Both  $\pi$  and  $\pi'$  are active cycles;
- $\pi$  is failing, then  $\text{count}[1]$  is limited (Condition 1);
- $\pi'$  is a twin of  $\pi$ . Since  $\pi$  is not balanced and its arguments are limited,  $\text{count}[2]$  is also limited (Condition 2).

# Relative Expressivity



## Theorem

*Argument Restricted*  $\subsetneq$  *Bounded*.

# Rule-bounded programs [CGMT14]

Many practical programs contain rules where the “size” of the head atom **does not increase** w.r.t. the “size” of a body atom.

## Example (Bubble Sort)

```
bub(L, [], []) ← input(L).  
bub([Y|T], [X|Cur], Sol) ← bub([X|[Y|T]], Cur, Sol), X ≤ Y.  
bub([X|T], [Y|Cur], Sol) ← bub([X|[Y|T]], Cur, Sol), Y < X.  
bub(Cur, [], [X|Sol]) ← bub([X|[]], Cur, Sol).
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```

```
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## Example (Tree Visit)

```
visit(Tree, [], []) ← input(Tree).  
visit(Left, [Root|Visited], [Right|ToVisit]) ←  
    visit(tree(Root, Left, Right), Visited, ToVisit).  
visit(Next, Visited, ToVisit) ← visit(null, Visited, [Next|ToVisit]).
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## Example (List Concatenation)

```
reverse(L1, []) ← input1(L1).
reverse(L1, [X|L2]) ← reverse([X|L1], L2).
append(L1, L2) ← reverse([], L1), input2(L2).
append(L1, [X|L2]) ← append([X|L1], L2).
```

# Rule-bounded programs [CGMT14]

- **Basic idea:** check if the size of the head is bounded by the size of a body atom.
- **Linear constraints** are used to check this condition.
- **Question:** How do we measure the size of an atom?

# Rule-bounded programs - Notions of Size

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Integer coefficients  $\alpha_{p_1}, \dots, \alpha_{p_n}$  will be chosen depending on the program structure.

# Rule-bounded program - Example

## Example (List Length)

$r_1 : \text{len}([a, b, c, d], 0).$

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$$2 \cdot \alpha_1 \geq \alpha_2$$

We can choose  $\alpha_1 = \alpha_2 = 1 \Rightarrow$  **the program is rule-bounded.**

# Rule-bounded programs - Another Example

## Example (Bubble sort)

`sort([b, a, d, h, e], [], []).`

`sort([Y|T], [X|Temp], Sorted)`  $\leftarrow$  `sort([X|[Y|T]], Temp, Sorted)`,  $X \leq Y$ .

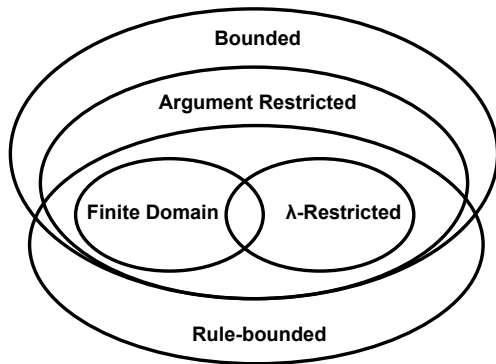
`sort([X|T], [Y|Temp], Sorted)`  $\leftarrow$  `sort([X|[Y|T]], Temp, Sorted)`,  $Y < X$ .

`sort(Temp, [], [X|Sorted])`  $\leftarrow$  `sort([X], Temp, Sorted)`.

$$\left\{ \begin{array}{l} \alpha_1 \cdot (4 + x + y + t) + \alpha_2 \cdot \mathit{temp} + \alpha_3 \cdot \mathit{sorted} \geq \\ \qquad \qquad \qquad \alpha_1 \cdot (2 + y + t) + \alpha_2 \cdot (2 + x + \mathit{temp}) + \alpha_3 \cdot \mathit{sorted} \\ \alpha_1 \cdot (4 + x + y + t) + \alpha_2 \cdot \mathit{temp} + \alpha_3 \cdot \mathit{sorted} \geq \\ \qquad \qquad \qquad \alpha_1 \cdot (2 + x + t) + \alpha_2 \cdot (2 + y + \mathit{temp}) + \alpha_3 \cdot \mathit{sorted} \\ \alpha_1 \cdot (2 + x) + \alpha_2 \cdot \mathit{temp} + \alpha_3 \cdot \mathit{sorted} \geq \\ \qquad \qquad \qquad \alpha_1 \cdot \mathit{temp} + \alpha_3 \cdot (2 + x + \mathit{sorted}) \end{array} \right.$$

A possible solution is  $\alpha_1 = 2$ ,  $\alpha_2 = 2$ ,  $\alpha_3 = 1$

# Relative Expressivity



## Theorem

- *Finite Domain*  $\subsetneq$  *Rule-bounded*.
- *$\lambda$ -Restricted*  $\subsetneq$  *Rule-bounded*.
- *Argument Restricted*  $\not\parallel$  *Rule-bounded*.
- *Bounded*  $\not\parallel$  *Rule-bounded*.

# Program Adornment [GMT13a]



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- The technique can be used in conjunction with current termination criteria allowing them to detect more programs having a terminating evaluation.
- The technique transforms a program  $P$  into an (adorned) “equivalent” program  $P^\mu$ .
- The aim is to apply termination criteria to the adorned program  $P^\mu$  rather than the original program  $P$ .

# Program Adornment

- Suppose we want to check if the evaluation of a program  $P$  terminates by applying a criterion  $C$ .
- We first transform  $P$  into an adorned program  $P^\mu$ .
- Then, we apply criterion  $C$  to  $P^\mu$  (rather than the original program  $P$ ).
- (Soundness) If  $P^\mu$  satisfies criterion  $C$ , then the evaluation of the original program  $P$  terminates.
- This approach strictly enlarges the class of programs identified by criterion  $C$ .

# Example

## **Original program**

$p(X, X) \leftarrow \text{base}(X)$

$q(X, Y) \leftarrow p(X, Y)$

$p(f(X), g(X)) \leftarrow q(X, X)$

# Example

## Original program

$$p(X, X) \leftarrow \text{base}(X)$$

$$q(X, Y) \leftarrow p(X, Y)$$

$$p(f(X), g(X)) \leftarrow q(X, X)$$

## Adorned program

$$p^{\varepsilon\varepsilon}(X, X) \leftarrow \text{base}^\varepsilon(X)$$

$$q^{\varepsilon\varepsilon}(X, Y) \leftarrow p^{\varepsilon\varepsilon}(X, Y)$$

$$p^{f_1g_1}(f(X), g(X)) \leftarrow q^{\varepsilon\varepsilon}(X, X)$$

$$q^{f_1g_1}(X, Y) \leftarrow p^{f_1g_1}(X, Y)$$

Each adorned rule is obtained from a rule in the original program by adding adornments which keep track of the structure of the terms that can be propagated during the bottom-up evaluation.

# Example

## Original program

$$p(X, X) \leftarrow \text{base}(X)$$

$$q(X, Y) \leftarrow p(X, Y)$$

$$p(f(X), g(X)) \leftarrow q(X, X)$$

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$$q^{\varepsilon\varepsilon}(X, Y) \leftarrow p^{\varepsilon\varepsilon}(X, Y)$$

$$p^{f_1g_1}(f(X), g(X)) \leftarrow q^{\varepsilon\varepsilon}(X, X)$$

$$q^{f_1g_1}(X, Y) \leftarrow p^{f_1g_1}(X, Y)$$

The adorned program is “equivalent” to the original one in the following sense: the minimal model of the original program can be obtained from the minimal model of the adorned program by dropping adornments.

# Adornment Algorithm

## **Original program**

$$p(X, f(X)) \leftarrow \text{base}(X)$$
$$p(X, f(X)) \leftarrow p(Y, X), \text{base}(Y)$$
$$p(X, Y) \leftarrow p(f(X), f(Y))$$

# Adornment Algorithm

## **Original program**

$p(X, f(X)) \leftarrow \text{base}(X)$

$p(X, f(X)) \leftarrow p(Y, X), \text{base}(Y)$

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## **Adorned program**

**Adorned  
predicate symbols**

$\text{base}^\varepsilon$

**Adornment  
definitions**

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# Adornment Algorithm

## Original program

$p(X, f(X)) \leftarrow base(X)$

$p(X, f(X)) \leftarrow p(Y, X), base(Y)$

$p(X, Y) \leftarrow p(f(X), f(Y))$

## Adorned program

$\leftarrow base^\varepsilon(X)$

**Adorned  
predicate symbols**

$base^\varepsilon$

**Adornment  
definitions**

# Adornment Algorithm

## Original program

$p(X, f(X)) \leftarrow \text{base}(X)$

$p(X, f(X)) \leftarrow p(Y, X), \text{base}(Y)$

$p(X, Y) \leftarrow p(f(X), f(Y))$

## Adorned program

$p^{\varepsilon f_1}(X, f(X)) \leftarrow \text{base}^\varepsilon(X)$

**Adorned  
predicate symbols**

$\text{base}^\varepsilon$

$p^{\varepsilon f_1}$

**Adornment  
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$f_1 = f(\varepsilon)$

# Adornment Algorithm

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$p(X, f(X)) \leftarrow \text{base}(X)$

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## Adorned program

$$p^{\varepsilon f_1}(X, f(X)) \leftarrow \text{base}^{\varepsilon}(X)$$
$$p^{f_1 f_2}(X, f(X)) \leftarrow p^{\varepsilon f_1}(Y, X), \text{base}^{\varepsilon}(Y)$$

## Adorned predicate symbols

$$\text{base}^{\varepsilon}$$
$$p^{\varepsilon f_1}$$
$$p^{f_1 f_2}$$

## Adornment definitions

$$f_1 = f(\varepsilon)$$
$$f_2 = f(f_1)$$

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The adornment algorithm terminates because no new *coherently adorned* body conjunction can be generated.

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**Adornment  
definitions**

$$f_1 = f(\varepsilon)$$

$$f_2 = f(f_1)$$

$p^{f_1 f_2}(Y, X), \text{base}^\varepsilon(Y)$  is not coherently adorned because  $Y$  is associated with the two different adornment symbols  $f_1$  and  $\varepsilon$

# Properties

## Theorem

*The adornment algorithm always terminates.*

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## Theorem

*Let  $P$  be a program and  $P^\mu$  the adorned version of  $P$ . If  $P^\mu$  satisfies a termination criterion  $C$ , then the evaluation of  $P \cup D$  terminates for any finite set of (flat) database facts  $D$ .*

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## Theorem

*By applying a termination criterion to adorned programs we are able to identify more programs whose evaluation terminates.*



# Dealing with Negation and Disjunction

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## Definition

A *Datalog*<sup>∨,¬</sup> rule is of the form

$$A_1 \vee \cdots \vee A_m \leftarrow B_1, \dots, B_k, \neg C_1, \dots, \neg C_n$$

where  $m > 0$ ,  $k \geq 0$ ,  $n \geq 0$ , and the  $A_i$ 's,  $B_i$ 's,  $C_i$ 's are atoms.

A *Datalog*<sup>∨,¬</sup> program is a finite set of *Datalog*<sup>∨,¬</sup> rules.

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Semantics: **Stable Model Semantics.**

We want to check if a *Datalog<sup>∨,¬</sup>* program has a finite number of stable models, each of finite size and that can be computed.

## Dealing with Negation and Disjunction

We want to check if a Datalog <sup>$\vee, \neg$</sup>  program  $P$  has a finite number of stable models, each of finite size and that can be computed.

We derive a Datalog program  $st(P)$  from  $P$  as follows.

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Each rule

$$A_1 \vee \dots \vee A_m \leftarrow B_1, \dots, B_k, \neg C_1, \dots, \neg C_n$$

in  $P$  is replaced with  $m$  Datalog rules

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### Proposition

*If  $st(P)$  satisfies a termination criterion, then  $P$  has a finite number of stable models, each of them is of finite size and can be computed.*

# Conclusions

- The evaluation of logic programs with function symbols might not terminate, and establishing termination is not decidable.
- One solution: (Sufficient) Termination Conditions.
- Related lines of research:
  - ▶ Ensure decidability of some reasoning tasks even if there might be infinite and infinitely many stable models (e.g.,  $\text{FDNC}$  programs [ES10, Bon11], Finitary Programs [Bon04], Finitely Recursive Programs [BBC09]).
  - ▶ Finite well-founded model [RS14].

# Directions for Future Work

- ① Combining termination criteria.  
One approach: identify arguments that are “limited” even when the program is not entirely recognized as terminating.
  - ▶ support the user in the problem formulation;
  - ▶ provide limited arguments to other techniques which can leverage this kind of information.



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- 2 Exploiting negation and disjunction.

## Example

$$p(f(X)) \leftarrow p(X), \neg p(X)$$

will be analyzed like

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- 3 Interpreted function symbols.
- 4 Testing Local Stratification.

Thanks!

Questions?

## Part II

# Existential Rules

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Special rules whose head atoms:

- may have existentially quantified variables,
- may be equality conditions (between two variables).

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Special rules whose head atoms:

- may have existentially quantified variables,
- may be equality conditions (between two variables).

Used in a variety of contexts:

- in databases to define integrity constraints;
- in data integration and data exchange to define schema mappings;
- for knowledge representation and ontological reasoning (Datalog<sup>±</sup>).

# Integrity constraints in databases

## Example

*emp*(*Emp#*, *Name*, *Address*)    *worksFor*(*Emp#*, *Prj#*)



# Integrity constraints in databases

## Example

$emp(Emp\#, Name, Address)$      $worksFor(Emp\#, Prj\#)$

- Inclusion dependencies and foreign keys:

$$worksFor(E, P) \rightarrow \exists N \exists A emp(E, N, A)$$

- Functional Dependencies and internal keys

$$emp(E, N_1, Pr_1) \wedge emp(E, N_2, Pr_2) \rightarrow N_1 = N_2$$

# Schema Mappings in Data Exchange

Data Exchange: Transform data structured under a source schema into data structured under a different target schema.

## Example

Company A

*empA(Emp#, Name, Address, Salary)*

Company B

*empB(Emp#, Name, Phone, Salary)*

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$empA(Emp\#, Name, Address, Salary)$

Company B

$empB(Emp\#, Name, Phone, Salary)$

Company A is acquired by Company B

$empA(E, N, A, S) \rightarrow \exists P empB(E, N, P, S)$

# Encoding Ontologies

- Plain Datalog allows for encoding some ontological axioms:
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- Plain Datalog allows for encoding some ontological axioms:
- TGDs can also express other important ontological axioms:

- Concept Inclusions:

$$\forall X \text{ emp}(X) \rightarrow \text{person}(X)$$

- (Inverse) Relation Inclusion:

$$\forall X \forall Y \text{ manages}(X, Y) \rightarrow \text{isManaged}(Y, X)$$

- Relation Transitivity:

$$\forall X \forall Y \forall Z \text{ mgs}(X, Y), \text{ mgs}(Y, Z) \rightarrow \text{mgs}(X, Z)$$

- Participation:

$$\forall X \text{ emp}(X) \rightarrow \exists Y \text{ report}(X, Y)$$

- Disjointness:

$$\forall X \text{ emp}(X), \text{ customer}(X) \rightarrow \text{false}$$

- Functionality:

$$\forall X \forall Y \forall Z \text{ reports}(X, Y), \text{ reports}(X, Z) \rightarrow Y = Z$$

# The Problem

# Answering queries under constraints

## The Problem

*Input:*

- *A database  $D$  (set of ground facts),*
- *A set of data dependencies (integrity constraints)  $\Sigma$ ,*
- *A (boolean) conjunctive query  $Q$*

*Question:*

- $D \cup \Sigma \models Q$

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Very old problem: CQ answering over incomplete databases



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Very old problem: CQ answering over incomplete databases

Undecidable in the general case

# Answering queries under constraints

## The Problem

*Data dependencies:*

- *Tuple generating dependencies (TGDs):*

$$\forall \bar{X} \forall \bar{Y} \varphi(\bar{X}, \bar{Y}) \rightarrow \exists \bar{Z} \psi(\bar{X}, \bar{Z})$$

- *Equality generating dependencies (EGDs):*

$$\forall \bar{X} \varphi(\bar{X}) \rightarrow X_1 = X_2$$

$\varphi(\bar{X}, \bar{Y})$ ,  $\varphi(\bar{X})$  and  $\psi(\bar{Z}, \bar{X})$  are conjunctions of atoms,  $X_1, X_2 \in \bar{X}$ .

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$K = D \cup \Sigma$  is called *knowledge base*.

# Answering queries under constraints

## Example (Models and answers)

- Database:  $D = \{person(john)\}$
- Data dependencies  $\Sigma$ :

$$\forall X \text{ person}(X) \rightarrow \exists Z \text{ fatherOf}(Z, X)$$

$$\forall X \forall Y \text{ fatherOf}(X, Y), \text{ person}(Y) \rightarrow \text{ person}(X)$$

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- Queries:

$$Q_1 = \exists X \text{ fatherOf}(X, john)$$

$$Q_2 = \exists X \text{ fatherOf}(john, X)$$

- Answers:

$$D \cup \Sigma \models Q_1$$

$$\text{certain}(Q_1, (D, \Sigma)) = \text{"yes"}$$

$$D \cup \Sigma \not\models Q_2$$

$$\text{certain}(Q_2, (D, \Sigma)) = \text{"no"}$$

All models of  $D \cup \Sigma$  contain an atom  $\text{fatherOf}(x, john)$ ,

# Datalog<sup>±</sup> (Syntax)

Datalog variant for ontological reasoning allowing in the head:

- existential variables (TGDs)
- Equality atoms (EGDs)
- Constant *false* (Denial constraints)

Also denoted as  $Datalog[\exists, =, F]$

More expressive than several ontological reasoning languages (e.g. UML Class Diagrams, DL-Lite,  $\mathcal{ELHI}^{\neg}$ , F-Logic Lite).

Query answering under Datalog<sup>±</sup> is undecidable

Query answering is undecidable



Determine decidable classes of queries



# Answering queries over incomplete databases

## Definition (Incomplete databases/Naive tables)

Databases may be *incomplete*, that is may contain (labeled) **nulls** (of the form  $\perp_j$ ), representing the presence of unknown values.

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## Definition (Possible worlds (under CWA))

Given a possibly incomplete database  $D$ , **POSS(D)** denotes the set of ground databases obtained from  $D$  by replacing nulls with constants.

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## Example (POSS(D))

- $D = \{person(john), person(frank), fatherOf(\perp_1, john)\}$

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Given a possibly incomplete database  $D$ , **POSS(D)** denotes the set of ground databases obtained from  $D$  by replacing nulls with constants.

## Example (POSS(D))

- $D = \{person(john), person(frak), fatherOf(\perp_1, john)\}$
- POSS(D) (under CWA) contains:
  - ▶  $\{person(john), person(frak), fatherOf(john, john)\}$
  - ▶  $\{person(john), person(frak), fatherOf(frak, john)\}$

# Answering queries over incomplete databases

## Definition (certain answer)

- **certain(D)** = database derived from  $D$  by deleting tuples with nulls.
- **certain(Q, D)** =  $\bigcap \{ \mathbf{Q(R)} \mid \mathbf{R} \in \mathbf{POSS(D)} \}$

# Answering queries over incomplete databases

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- **certain(Q, D)** =  $\bigcap \{ Q(R) \mid R \in \text{POSS}(D) \}$

## Theorem (weak representation systems)

*For union of conjunctive queries*

$$\text{certain}(Q(D)) = \text{certain}(Q, D)$$

Certain answers can be computed by

- 1 Evaluating (naively)  $Q(D)$
- 2 Removing tuples with nulls

# Answering queries under constraints

## Definition (Model)

Given a knowledge base  $K = D \cup \Sigma$ ,  $M$  is a **model** of  $K$  if  $M \models K$ .

## Definition (Homomorphism)

Mapping  $h : \text{Nulls} \rightarrow \text{Nulls} \cup \text{Constants}$ .

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**POSS(M) = { R | h(M)  $\subseteq$  R  $\wedge$  R is ground }.**



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**POSS(M) = { R | h(M)  $\subseteq$  R  $\wedge$  R is ground }.**

## Definition (certain answer)

**certain(Q, (D,  $\Sigma$ )) =  $\bigcap$ { Q(R) | R  $\in$  POSS(M)  $\wedge$  M is a model of D  $\cup$   $\Sigma$  }**

# Universal models

## Definition (Models comparison)

Given two models  $M_1$  and  $M_2$  we say that  $M_1$  is **at least as general** as  $M_2$  ( $M_1 \sqsupseteq M_2$ ) if  $\exists h$  such that  $h(M_1) \subseteq M_2$ .

$M_1$  is **more general** than  $M_2$  ( $M_1 \sqsupset M_2$ ) if  $M_1 \sqsupseteq M_2$  and  $M_2 \not\sqsupseteq M_1$ .

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## Theorem

- $M_1 \sqsupseteq M_2$  iff  $POSS(M_1) \supseteq POSS(M_2)$ ,
- $M_1 \subseteq M_2 \Rightarrow M_1 \sqsupseteq M_2$ .

# Universal models

## Definition (Universal model)

$M$  is an **universal model** (or **universal solution**) if for every model  $N$ ,  $M \sqsubseteq N$  (i.e.  $\exists h$  s.t.  $h(M) \subseteq N$ ).

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## Theorem (Main Th.)

For every UCQ  $Q$  and for every *arbitrary universal model*  $\mathbf{M}$  of  $D \cup \Sigma$

$$\mathbf{certain}(Q, (D, \Sigma)) = \mathbf{certain}(Q, \mathbf{M}) = \mathbf{certain}(Q(\mathbf{M}))$$

Recall that:

$$\mathbf{certain}(Q, (D, \Sigma)) = \bigcap \{ Q(R) \mid R \in \text{POSS}(M) \wedge M \text{ is a model of } D \cup \Sigma \}$$

# Universal models

## Example (Models and answers)

- Database:  $D = \{person(john)\}$
- Data dependencies  $\Sigma$ :

$$\forall X \text{ person}(X) \rightarrow \exists Z \text{ fatherOf}(Z, X)$$

$$\forall X \forall Y \text{ fatherOf}(X, Y), \text{ person}(Y) \rightarrow \text{ person}(X)$$

- Models (under OWA):

$$M_1 = \{\text{person}(john), \text{fatherOf}(john, john)\}$$

$$M_2 = \{\text{person}(john), \text{fatherOf}(\perp_1, john), \text{person}(\perp_1)\}$$

$$M_3 = \{\text{person}(john), \text{fatherOf}(\perp_2, john), \text{person}(\perp_2)\}$$

$$M_4 = \{\text{person}(john), \text{fatherOf}(\perp_1, john), \text{person}(frank)\}$$

...

# Universal models

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$$M_3 = \{person(john), fatherOf(\perp_2, john), person(\perp_2)\}$$

$$M_4 = \{person(john), fatherOf(\perp_1, john), person(frunk)\}$$

...

$$M_2 \sqsupseteq M_1, M_2 \sqsupseteq M_4, M_2 \sqsupseteq M_3, \quad M_3 \sqsupseteq M_1, M_3 \sqsupseteq M_4, M_3 \sqsupseteq M_2$$

**$M_2$  and  $M_3$  are universal models.**

# The Chase

Fixpoint algorithm designed to enforce satisfaction of dependencies.

The execution of the chase involves

- adding new facts (possibly with null values) to satisfy TGDs,
- replacing nulls (with constants or other null values) to satisfy EGDs.



# The Chase

Several problems can be solved using the chase algorithm:

- Checking query containment under dependencies
- Checking implication of dependencies
- Checking lossless decomposition of database schema
- Computing universal solutions in data exchange
- Computing certain answers in data integration
- Ontology Querying
- Database repair
- ...

# The Chase

## Chase algorithm $chase(D, \Sigma)$

Iteratively, let  $K$  be the current instance ( $K = D$  at step 0),

- select nondeterministically a constraint  $r \in \Sigma$  and an homomorphism  $h$  such that  $K \not\models h(r)$  (i.e.  $K \models \text{body}(h(r))$  and there is no extension  $h'$  of  $h$  such that  $K \models \text{head}(h'(r))$ ).
- enforce the satisfaction of  $h(r)$  by either i) adding a tuple (if  $r$  is a TGD), or ii) replacing a null value (if  $r$  is an EGD), or "fail" (if  $r$  is an EGD which cannot be enforced).

A chase step from  $K_1$  and  $r_1$  with homomorphism  $h$  to  $K_2$  is denoted as  $K_1 \xrightarrow{r_1, h} K_2$ .

The result of  $chase(D, \Sigma)$  is nondeterministic and is either

- a (possibly infinite) universal model;
- fail, if  $D \cup \Sigma$  does not have universal models.

# Chase - Enforcing data dependencies: a first example

## Example

$$\begin{array}{ll} D : & \Sigma : \\ N(a) & N(X) \rightarrow \exists Y E(X, Y) \\ S(a) & S(X) \wedge E(X, Y) \rightarrow N(Y) \end{array}$$

# Chase - Enforcing data dependencies: a first example

## Example

$D :$

$N(a)$

$S(a)$

$\Sigma :$

$N(X) \rightarrow \exists Y E(X, Y)$

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$chase(D, \Sigma) = \{N(a), S(a)\}$

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$$\text{chase}(D, \Sigma) = \{ N(a), S(a), E(a, \perp_1), N(\perp_1), E(\perp_1, \perp_2) \}$$

All dependencies are satisfied: STOP.

This and every other chase sequence terminates.

# Chase - Enforcing data dependencies: a first example

## Example

$D :$

$N(a)$

$S(a)$

$\Sigma :$

$N(X) \rightarrow \exists Y E(X, Y)$

~~$S(X) \wedge E(X, Y) \rightarrow N(Y)$~~

$\text{chase}(D, \Sigma) = \{ N(a), S(a), E(a, \perp_1), N(\perp_1), E(\perp_1, \perp_2) \}$

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$$\text{chase}(D, \Sigma) = \{ N(a), S(a), E(a, \perp_1), N(\perp_1), E(\perp_1, \perp_2), N(\perp_2), \dots \}$$

There is **no** terminating chase sequence.

# Chase – Terminating Sequence

## Example ( $\exists$ a finite sequence)

$D :$

$airport(a)$

$\Sigma :$

$r_1 : airport(X) \rightarrow \exists Y flight(X, Y)$

$r_2 : flight(X, Y) \rightarrow airport(X) \wedge airport(Y)$

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The following facts are added to  $D$  :

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**No further rule is applicable: STOP.**

# Chase – Non-terminating Sequence

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$flight(\perp_1, \perp_2)$

# Chase – Non-terminating Sequence

Example ( $\exists$  an infinite sequence)

$D$  :

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$\Sigma$  :

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$r_2 : flight(X, Y) \rightarrow airport(X) \wedge airport(Y)$

$r_3 : flight(X, Y) \rightarrow flight(Y, X)$

The following facts are added to  $D$  :

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$flight(\perp_1, \perp_2)$

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$\vdots$

By iteratively applying  $r_1$  and  $r_2$  **the chase never terminates.**

# Chase Termination

- Checking whether there is **at least one** terminating chase sequence vs. **all** chase sequences are terminating;
- for **a given instance  $D$**  vs. **for every instance**.

# Chase Termination

## Theorem

Consider a set  $\Sigma$  of TGDs:

- It is undecidable whether, **for every instance**  $D$ , **some** chase sequence of  $D$  with  $\Sigma$  terminates [GO13].
- It is undecidable whether, **for every instance**  $D$ , **all** chase sequences of  $D$  with  $\Sigma$  terminate [GM14].

# Chase Termination

## Theorem ([DNR08])

Given a set  $\Sigma$  of TGDs and a (fixed) instance  $D$ :

- *It is undecidable whether some chase sequence of  $D$  with  $\Sigma$  terminates.*
- *It is undecidable whether all chase sequences of  $D$  with  $\Sigma$  terminate.*

# Sufficient Conditions

**One Solution:** Identify sufficient conditions guaranteeing chase termination.

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Many have been proposed:

- Weak Acyclicity [FKMP05]
- Stratification [DNR08] and C-Stratification [MSL09]
- Safety and Inductive Restriction [MSL09]
- Super-weak Acyclicity [Mar09]
- Local Stratification [GST11, GST15]
- Adornment Techniques [GS10, GST15]
- Model-Faithful Acyclicity [GHK<sup>+</sup>13]
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- Acyclic Graph Rule Dependencies [BLMS11]

From now on we consider only TGDs



# Chase Variants

## Oblivious and Semi-oblivious

- The set of dependencies is *skolemized*.
- The resulting logic program is evaluated.
- The oblivious and semi-oblivious chases adopt two different skolemizations.

### Example

$$r : N(X, Y) \rightarrow \exists K, Z E(X, K, Z)$$

- **Oblivious** Chase. Skolemization:

$$N(X, Y) \rightarrow E(X, f_r^K(X, Y), f_r^Z(X, Y))$$

- **Semi-oblivious** Chase. Skolemization:

$$N(X, Y) \rightarrow E(X, f_r^K(X), f_r^Z(X))$$

# Chase Variants

Example (Complex terms represent nulls)

$$D : E(a, b) \quad \Sigma : E(X, Y) \rightarrow \exists Z E(X, Z)$$

**Standard**

**Semi-oblivious**

**Oblivious**

# Chase Variants

Example (Complex terms represent nulls)

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No chase step  
( $D \models \Sigma$ )

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⋮

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STOP (fixpoint)

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$$E(a, b) \\ E(a, \perp_2) \\ E(a, \perp_3) \\ E(a, \perp_4)$$

⋮

NO Termination  
(no fixpoint)

# Chase Variants

## Core Chase [DNR08]

Minimal universal models.

Core chase step:

- 1 Enforce all dependencies “in parallel”.
- 2 “Retract” the result (homomorphism  $h : M \rightarrow M$ ).

## Theorem (Completeness of the Core Chase [DNR08])

*If  $D$  is an instance and  $\Sigma$  is a set of TGDs. then there exists a universal model for  $\Sigma$  and  $I$  iff the core chase of  $I$  with  $\Sigma$  terminates and yields such a model.*

# Chase Variants

- $CT_{\forall}^C$ : class of sets of TGDs  $\Sigma$  s.t., for every instance, **all** c-chase sequences terminate.
- $CT_{\exists}^C$ : class of sets of TGDs  $\Sigma$  s.t., for every instance, **at least one** c-chase sequence terminates.

Theorem ([Mei10, One13] For TGDs only)

$$CT_{\forall}^{obl} = CT_{\exists}^{obl} \subsetneq CT_{\forall}^{sobl} = CT_{\exists}^{sobl} \subsetneq CT_{\forall}^{std} \subsetneq CT_{\exists}^{std} \subsetneq CT_{\forall}^{core} = CT_{\exists}^{core}$$



# Function Symbols vs. TGDs

Termination Criteria for programs with function symbols can be applied to TGDs:

## Step 1. Skolemize TGDs.

### Example

$$\begin{aligned} r &: p(X, Y) \rightarrow \exists K, Z q(X, K, Z) \\ sk(r) &: p(X, Y) \rightarrow q(X, f_r^K(X), f_r^Z(X)) \end{aligned}$$

## Step 2. Apply termination criteria to skolemized TGDs.

Given a set  $\Sigma$  of TGDs, let  $sk(\Sigma) = \{sk(r) \mid r \in \Sigma\}$ .

Termination of the bottom-up evaluation of  $sk(\Sigma)$  (i.e., the semi-oblivious chase)  $\Rightarrow$  Termination of the chase of  $\Sigma$  [One13].

# Function Symbols vs. TGDs

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$$r: p(X, Y) \rightarrow \exists Z p(X, Z)$$

# Function Symbols vs. TGDs

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**Step 1. Skolemize  $r$ :**

$$sk(r) : p(X, Y) \rightarrow p(X, f_r^Z(X))$$

We get a logic program with function symbols.

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- E.g.,  $sk(r)$  is argument-restricted.

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- E.g.,  $sk(r)$  is argument-restricted.
- Thus, the bottom-up evaluation of  $sk(r)$  always terminates.
- That is, the semi-oblivious chase of  $r$  always terminates.
- Thus, the standard chase of  $r$  always terminates.

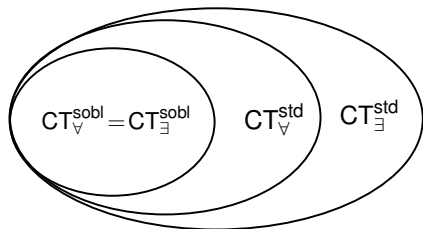


# Function Symbols vs. TGDs

**Limitations:** Recall that:

Theorem ([Mei10, One13])

$$CT_{\forall}^{\text{sobl}} = CT_{\exists}^{\text{sobl}} \not\subseteq CT_{\forall}^{\text{std}} \not\subseteq CT_{\exists}^{\text{std}}$$

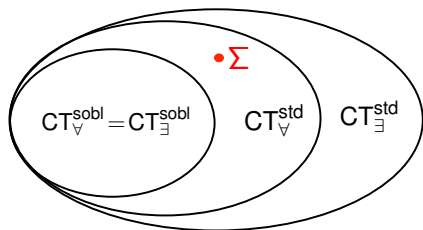


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# Function Symbols vs. TGDs

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# Function Symbols vs. TGDs

What about applying criteria for TGDs to programs with function symbols?

The latter are more general than skolemized TGDs.

Each function symbol occurs:

<b>Skolemized TGDs</b>	<b>Programs with function symbols</b>
once	arbitrary number of times
only in the head	in the body and/or head
no nesting	arbitrary nesting

# Termination Criteria

## Dependency Graph

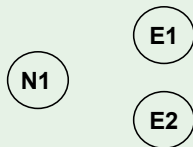
- Nodes are predicate arguments.
- Two kinds of edges:
  - 1 normal edges represent the propagation of values between arguments;
  - 2 special edges  $\xrightarrow{*}$  represent the generation of nulls.

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$$\Sigma = \begin{array}{l} N(X) \rightarrow \exists Y E(X, Y) \\ E(X, Y) \rightarrow N(Y) \end{array}$$

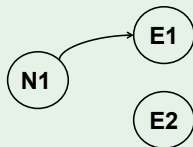


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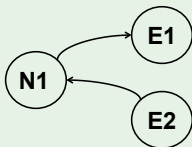


Dependency Graph  $dep(\Sigma)$ 

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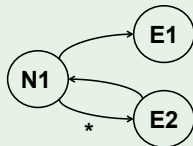


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## Example

$$\Sigma = \begin{array}{l} \mathbf{N(X)} \rightarrow \exists Y \mathbf{E(X, Y)} \\ \mathbf{E(X, Y)} \rightarrow \mathbf{N(Y)} \end{array}$$



## Definition

A set of dependencies is **weakly acyclic** if **there is no cycle going through a special edge** in the dependency graph.

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## Theorem

If  $\Sigma$  is weakly acyclic, then **for every instance  $I$ , every chase sequence terminates** (and has a polynomial length in the size of  $I$ ).

Affected Positions  $aff(\Sigma)$  [CGK13]

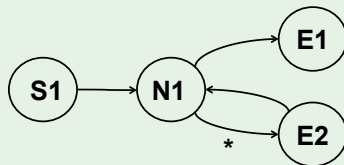
Overestimation of positions that may contain null values.

Propagation Graph  $prop(\Sigma)$ 

Restriction of the dependency graph containing only affected positions.

## Example

$$\Sigma = \begin{array}{l} r_1 : N(X) \rightarrow \exists Y E(X, Y) \\ r_2 : S(Y) \wedge E(X, Y) \rightarrow N(Y) \end{array}$$



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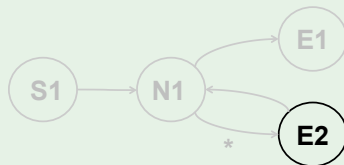
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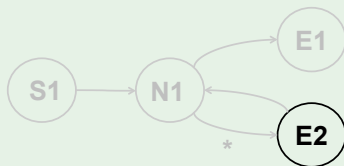
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- $aff(\Sigma) = \{E_2\}$
- $prop(\Sigma) = (\{E_2\}, \emptyset)$

### Affected Positions $aff(\Sigma)$

Overestimation of positions that may contain null values.

### Propagation Graph $prop(\Sigma)$

Restriction of dependency graph containing only affected positions.

### Safety

A set of dependencies is **safe** if **the propagation graph does not contain cycles with special edges.**



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Overestimation of positions that may contain null values.

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Restriction of dependency graph containing only affected positions.

### Safety

A set of dependencies is **safe** if **the propagation graph does not contain cycles with special edges.**

### Theorem

If  $\Sigma$  is safe, then **for every instance  $I$ , every chase sequence terminates** (and has a polynomial length in the size of  $I$ ).

Chase Graph  $G(\Sigma)$ 

- It represents how dependencies fire each other.
- Nodes: the dependencies in  $\Sigma$ .
- Edges: there is an edge from  $r_1$  to  $r_2$  ( $r_1 \prec r_2$ ) if  $r_1$  may “fire”  $r_2$ .

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Definition (Chase Graph  $G(\Sigma)$ )

$r_1 \prec r_2$  if  $\exists$  instance  $K_1$  and homomorphisms  $h_1$  and  $h_2$  such that

- 1)  $K_1 \xrightarrow{r_1, h_1} K_2$  (chase step -  $K_1 \not\models h_1(r_1)$ ),
- 2)  $K_2 \not\models h_2(r_2)$ ,
- 3)  $K_1 \models h_2(r_2)$ .

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## Example

$$\Sigma = \begin{array}{l} r_1 : N(X) \rightarrow \exists Y E(X, Y) \\ r_2 : S(Y) \wedge E(X, Y) \rightarrow N(Y) \end{array}$$

- *there exists a scenario where firing  $r_2$  causes  $r_1$  to fire ( $r_2 \prec r_1$ ).*
- $r_1 \not\prec r_2$ ,  $r_1 \not\prec r_1$  and  $r_2 \not\prec r_2$ .
- *The chase graph is acyclic and  $\Sigma$  is stratified.*

## Stratification

A set of dependencies is *stratified* if **every cycle in the chase graph  $G(\Sigma)$  is weakly acyclic.**

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A set of dependencies is *stratified* if **every cycle in the chase graph  $G(\Sigma)$  is weakly acyclic.**

## Theorem

*If  $\Sigma$  is stratified then, for every instance  $I$ , **there exists at least one chase sequence that terminates** (and whose length is polynomial in the size of  $I$ ).*

# C-Stratification [MSL09] vs Stratification

- A variation called *c-stratification* guarantees the termination of **every** chase sequence.
- Same approach of stratification, but the oblivious chase is used.
- C-Stratification
- Stratification

$r_1 \prec_c r_2$  if:

- 1)  $K_1 \xrightarrow{*, r_1/h_1} K_2$  (oblivious step),
- 2)  $K_2 \not\models h_2(r_2)$ ,
- 3)  $K_1 \models h_2(r_2)$ .

$r_1 \prec r_2$  if:

- 1)  $K_1 \xrightarrow{r_1/h_1} K_2$  (standard step),
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- 3)  $K_1 \models h_2(r_2)$ .

## Theorem

If  $\Sigma$  is *c-stratified* then, for every instance  $I$ , **all chase sequences terminate** and their length is polynomial in the size of  $I$ .

For any  $\Sigma$ ,  $G(\Sigma) \subseteq G_c(\Sigma) \Rightarrow Str \supseteq CStr$



- It improves the firing relation by considering possible propagation of null values.
- It tests safety on the (nontrivial) strongly connected components of the graph.
- It generalizes both safety and c-stratification.

## Theorem

If  $\Sigma$  is inductively restricted, then **for every instance  $I$ , every chase sequence terminates** (and has a polynomial length in the size of  $I$ ).

# Super-weak Acyclicity [Mar09] [semi-obliv. chase]

- Builds a *trigger graph* whose edges define relations among dependencies. An edge  $r_i \rightsquigarrow r_j$  means that a null value introduced by a dependency  $r_i$  is propagated (directly or indirectly) into the body of  $r_j$ .
- Different nulls in positions for the same variable  $\Rightarrow$  dependencies are not fired

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## Example

$$r_1 : N(X) \rightarrow \exists Y, Z E(X, Y, Z)$$

$$r_2 : E(X, Y, Z) \rightarrow G(X, Y, Z)$$

$$r_3 : G(X, Y, Y) \rightarrow N(Y)$$

$\Sigma$  neither safe nor stratified.

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$\Sigma$  neither safe not stratified.

$$P(\Sigma) = \begin{cases} r'_1 : N(X) \rightarrow E(X, f_Y^f(X), f_Z^f(X)) \\ r'_2 : E(X, Y, Z) \rightarrow \exists Y, Z G(X, Y, Z) \\ r'_3 : G(X, Y, Y) \rightarrow N(Y) \end{cases}$$

# Super-weak Acyclicity [Mar09]

## Super-weak Acyclicity

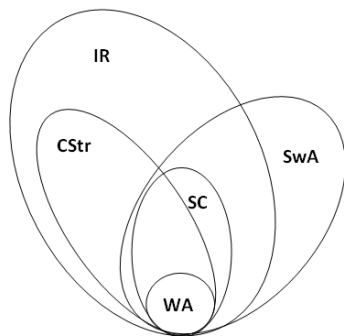
A set of dependencies is *super-weak acyclic* if **the trigger relation is acyclic**.

## Theorem

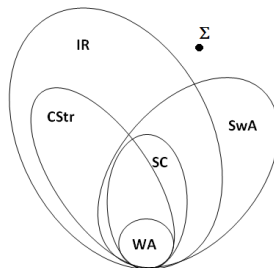
If  $\Sigma$  is super-weak acyclic, then **for every instance  $I$ , every chase sequence terminates** (and has a polynomial length in the size of  $I$ ).

# Relative Expressivity

- $WA$ : Weak Acyclicity
- $SC$ : Safety
- $CStr$ : C-stratification
- $IR$ : Inductive Restriction
- $SwA$ : Super-weak Acyclicity



# Limitations



## Example

$$r_1 : N(X) \rightarrow \exists Y \exists Z E(X, Y) \wedge S(Z, Y)$$

$$r_2 : E(X, Y) \wedge S(X, Y) \rightarrow N(Y)$$

$$r_3 : E(X, Y) \rightarrow E(Y, X)$$

# Improvements of (C-)Stratification

- Builds a *firing graph*  $\Gamma(\Sigma) = (\Sigma, E)$  representing how constraints fire each other.
- $(r_1, r_2) \in E$  if  $r_1 < r_2$  (firing  $r_1$  can cause  $r_2$  to fire)
- $r_1 < r_2$  if:
  - 1)  $K_1 \xrightarrow{r_1, h_1} K_2$ ,
  - 2)  $K_2 \cup S \not\models h_2(r_2)$ ,
  - 3)  $K_1 \cup S \models h_2(r_2)$  and
  - 4)  $\text{Null}(S) \cap (\text{Null}(K_2) - \text{Null}(K_1)) = \emptyset$ .
- $r_1 < r_2$  if:
  - 1)  $K_1 \xrightarrow{r_1, h_1} K_2$ ,
  - 2)  $K_2 \not\models h_2(r_2)$ ,
  - 3)  $K_1 \models h_2(r_2)$ .

As  $r_1$  could cause the firing of  $r_2$  not immediately,  $S$  is a set of atoms which could have been inferred after the firing of  $r_1$ .

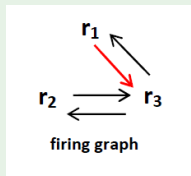


# Improvements of (C-)Stratification

## Example

$$\begin{aligned} \Sigma = & \quad r_1 : R(x) \rightarrow \exists y T(x, y) \\ & \quad r_2 : R(x) \rightarrow T(x, x) \\ & \quad r_3 : T(x, y) \wedge T(x, x) \rightarrow R(y) \end{aligned}$$

- $K_1 = \{R(a)\}$  and  $K_2 = \{R(a), T(a, \perp_1)\}$
- $S = \{T(a, a)\}$
- $r_3 : T(a, \perp_1) \wedge T(a, a) \rightarrow R(\perp_1)$
- $r_3$  is fired by  $r_1$ , then we have  $r_1 < r_3$

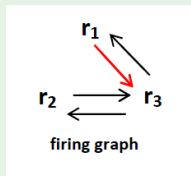


# Improvements of (C-)Stratification

## Example

$$\begin{aligned} r_1 &: R(x) \rightarrow \exists y T(x, y) \\ \Sigma = r_2 &: R(x) \rightarrow T(x, x) \\ r_3 &: T(x, y) \wedge T(x, x) \rightarrow R(y) \end{aligned}$$

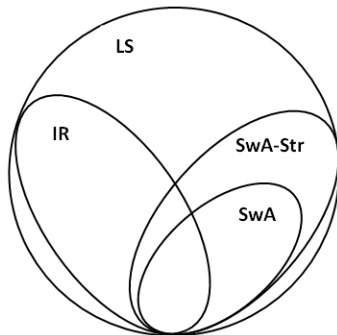
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## Local Stratification

- *WA-Str* (resp. *SC-Str*, *SwA-Str*) tests *WA* (resp. *SC*, *SwA*) over components of  $\Gamma(\Sigma)$
- *Local Stratification (LC)* combines *SwA* with  $\Gamma(\Sigma)$ : in analyzing how nulls may be propagated from a rule  $r_i$  to a rule  $r_j$ , also checks whether  $r_i < r_j$  transitively.

# Criteria Relationships



# Rewriting Techniques

# Constraints Rewriting Technique [GST15]

## Idea

- Rewrite  $\Sigma$  into an ‘equivalent’ adorned set  $\Sigma^\alpha$  and verify the structural properties for chase termination on  $\Sigma^\alpha$  (similarly to LPs)
- Rewrite  $\Sigma$  into a set of dependencies useful to analyze the structure of terms during the execution.

# Rewriting Algorithm [GST15]

## Example

$\Sigma$  :

$$r_1 : N(X) \rightarrow \exists Y E(X, Y)$$

$$r_2 : S(X) \wedge E(X, Y) \rightarrow N(Y)$$

$Adn(\Sigma)$ :

$$s_1 : N(X) \rightarrow N^b(X)$$

$$s_2 : S(X) \rightarrow S^b(X)$$

$$s_3 : E(X, Y) \rightarrow E^{bb}(X, Y)$$

$$r'_1 : N^b(X) \rightarrow \exists Y E^{bf_1}(X, Y) \quad \mathbf{f_1} = f_{r_1}^Y(b)$$

$$r'_2 : S^b(X) \wedge E^{bb}(X, Y) \rightarrow N^b(Y)$$

$$r''_2 : S^b(X) \wedge E^{bf_1}(X, Y) \rightarrow N^{f_1}(Y)$$

$$r''_1 : N^{f_1}(X) \rightarrow \exists Y E^{f_1 f_2}(X, Y) \quad \mathbf{f_2} = f_{r_1}^Y(f_1)$$

# Model-Faithful Acyclicity (MFA) [GHK<sup>+</sup>13]

## Example

$\Sigma :$

$$r : A(X) \rightarrow \exists Y, Z R(X, Y) \wedge B(Z)$$

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$$S(X_1, X_2) \rightarrow D(X_1, X_2)$$

$$D(X_1, X_2) \wedge S(X_2, X_3) \rightarrow D(X_1, X_3)$$

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$$F_r^Y(X_1) \wedge D(X_1, X_2) \wedge F_r^Y(X_2) \rightarrow C$$

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If  $I \cup MFA(\Sigma) \models C$  then a cyclic term is derived during the semi-oblivious chase execution of  $I$  and  $\Sigma$ .

# Model-Faithful Acyclicity (MFA) [GHK<sup>+</sup>13]

## Definition

$\Sigma$  is MFA w.r.t. an instance  $I$  if  $I \cup \text{MFA}(\Sigma) \not\models C$ .

## Definition

The *critical instance*  $I_\Sigma$  for  $\Sigma$  is the instance containing all facts that can be built using:

- all predicates in  $\Sigma$ ,
- all constants in the body of a dependency in  $\Sigma$ , and
- one special fresh constant  $*$ .

## Theorem ([Mar09])

*The semi-oblivious chase of  $\Sigma$  and  $I$  terminates for every  $I$  iff the semi-oblivious chase of  $\Sigma$  and  $I_\Sigma$  terminates.*

## Theorem

If  $\Sigma$  is MFA w.r.t.  $I_\Sigma$ , then **for every instance  $I$ , every (semi-oblivious) chase sequence terminates.**

# Related Approaches

So far we have discussed **sufficient conditions** ensuring chase termination.

Other lines of research:

- Identify restricted classes of dependencies for which the termination problem is decidable [CGP15].
- Identify restricted classes of dependencies guaranteeing decidability of query answering (even if the chase does not terminate).
  - ▶ Guarded and Weakly Guarded Datalog<sup>±</sup> [CGK13]
  - ▶ Sticky Datalog<sup>±</sup> [CGP10]
  - ▶ Forward and Backward chaining [BLMS11]

# Adding EGDs

# EGDs – Syntax

An Equality-Generating Dependency is of the form:

$$\forall \bar{X} \varphi(\bar{X}) \rightarrow X_1 = X_2$$

where  $\varphi(\bar{X})$  is a conjunction of atoms and  $X_1, X_2 \in \bar{X}$ .

## Example

$$\forall M_1, M_2, P \text{ directs}(M_1, P) \wedge \text{directs}(M_2, P) \rightarrow M_1 = M_2$$



# EGDs and Chase Termination

- 1 In some cases the presence of EGDs allows us to have a terminating c-chase sequence when the set consisting only of the TGDs does not have one;
- 2 In some cases in the presence of EGDs there is no terminating c-chase sequence, but the set consisting only of the TGDs does have one.

# Chase and EGDs

Adding EGDs leads to termination

## Example (No EGDs)

$D :$

$N(a)$

$\Sigma :$

$N(X) \rightarrow \exists Y E(X, Y)$   
 $E(X, Y) \rightarrow N(Y)$

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$chase(D, \Sigma) = \{N(a), E(a, \perp_1), N(\perp_1), E(\perp_1, \perp_2), \dots\}$

There is no terminating chase sequence.

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Adding an EGD to  $\Sigma$  ...

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$E(X, Y, Z) \rightarrow Y = Z$

$chase(D, \Sigma) = \{N(a), E(a, \perp_1, \perp_1)\},$

# Chase and EGDs

Adding EGDs  $\rightarrow$  No termination

Adding an EGD to  $\Sigma$  ...

## Example (TGDs + EGDs)

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No termination

# Relationship between $CT_{\forall}^c$ and $CT_{\exists}^c$

For TGDs only:

$$CT_{\forall}^{obl} = CT_{\exists}^{obl} \subset CT_{\forall}^{sobl} = CT_{\exists}^{sobl} \subset CT_{\forall}^{std} \subset CT_{\exists}^{std} \subset CT_{\forall}^{core} = CT_{\exists}^{core}$$

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$$CT_{\forall}^{obl} \subset CT_{\exists}^{obl} \not\subset CT_{\forall}^{sobl} \subset CT_{\exists}^{sobl} \not\subset CT_{\forall}^{std} \subset CT_{\exists}^{std} \subset CT_{\forall}^{core} = CT_{\exists}^{core}$$

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$$\begin{aligned} CT_{\forall}^{obl} \subset CT_{\exists}^{obl} \not\subset CT_{\forall}^{sobl} \subset CT_{\exists}^{sobl} \not\subset CT_{\forall}^{std} \subset CT_{\exists}^{std} \subset CT_{\forall}^{core} = CT_{\exists}^{core} \\ CT_{\forall}^{obl} \subset CT_{\forall}^{sobl} \subset CT_{\forall}^{std} \subset CT_{\forall}^{core} \\ CT_{\exists}^{obl} \subset CT_{\exists}^{sobl} \subset CT_{\exists}^{std} \subset CT_{\exists}^{core} \end{aligned}$$

# Termination Criteria and EGDs

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  - ▶ Natural Simulation [Gottlob et al., PODS06];
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- But they can be applied by simulating EGDs with TGDs:
  - ▶ Natural Simulation [Gottlob et al., PODS06];
  - ▶ Substitution-free simulation [Marnette, PODS09].
- However, the behaviour of EGDs cannot be fully simulated via TGDs...

# EGDs Simulation

## Example

$\Sigma :$

$$r_1 : A(x) \wedge B(x) \rightarrow C(x)$$

$$r_2 : C(x) \rightarrow \exists y A(x) \wedge B(y)$$

$$r_3 : C(x) \rightarrow \exists y A(y) \wedge B(x)$$

$$r_4 : A(x) \wedge A(y) \rightarrow x = y$$

$$r_5 : B(x) \wedge B(y) \rightarrow x = y$$

Every chase sequence is terminating, for any variation of the chase.

## Example

$\Sigma$  :

$$r_1 : A(x) \wedge B(x) \rightarrow C(x)$$

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Every chase sequence is terminating, for any variation of the chase.

However, both the natural and the substitution-free simulations of  $\Sigma$  have no terminating chase sequence.



# EGDs Simulation

*Substitution-free simulation* [Mar09]

## Example

$$A(X) \wedge B(X) \rightarrow C(X)$$

$$C(X) \rightarrow \exists Y A(X) \wedge B(Y)$$

$$C(X) \rightarrow \exists Y A(Y) \wedge B(X)$$

$$A(X) \wedge A(Y) \rightarrow X = Y$$

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# EGDs Simulation

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$$B(X) \wedge B(Y) \rightarrow X = Y$$

$$Eq(X, Y) \rightarrow Eq(Y, X)$$

$$Eq(X, Y) \wedge Eq(Y, Z) \rightarrow Eq(X, Z)$$

$$A(X) \rightarrow Eq(X, X)$$

$$B(X) \rightarrow Eq(X, X)$$

$$C(X) \rightarrow Eq(X, X)$$

# EGDs Simulation

*Substitution-free simulation* [Mar09]

## Example

$$\cancel{A(X) \wedge B(X) \rightarrow C(X)} \quad A(X) \wedge B(X_2) \wedge Eq(X, X_2) \rightarrow C(X)$$

$$C(X) \rightarrow \exists Y A(X) \wedge B(Y)$$

$$C(X) \rightarrow \exists Y A(Y) \wedge B(X)$$

$$A(X) \wedge A(Y) \rightarrow X = Y$$

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# EGDs Simulation

*Substitution-free simulation* [Mar09]

## Example

$$\cancel{A(X) \wedge B(X) \rightarrow C(X)} \quad A(X) \wedge B(X_2) \wedge \text{Eq}(X, X_2) \rightarrow C(X)$$

$$C(X) \rightarrow \exists Y A(X) \wedge B(Y)$$

$$C(X) \rightarrow \exists Y A(Y) \wedge B(X)$$

$$A(X) \wedge A(Y) \rightarrow \cancel{X=Y} \text{Eq}(X, Y)$$

$$B(X) \wedge B(Y) \rightarrow \cancel{X=Y} \text{Eq}(X, Y)$$

$$\text{Eq}(X, Y) \rightarrow \text{Eq}(Y, X)$$

$$\text{Eq}(X, Y) \wedge \text{Eq}(Y, Z) \rightarrow \text{Eq}(X, Z)$$

$$A(X) \rightarrow \text{Eq}(X, X)$$

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*Substitution-free simulation* [Mar09]

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$$B(X) \wedge B(Y) \rightarrow \cancel{X=Y} \text{Eq}(X, Y)$$

$$\text{Eq}(X, Y) \rightarrow \text{Eq}(Y, X)$$

$$\text{Eq}(X, Y) \wedge \text{Eq}(Y, Z) \rightarrow \text{Eq}(X, Z)$$

$$A(X) \rightarrow \text{Eq}(X, X)$$

$$B(X) \rightarrow \text{Eq}(X, X)$$

$$C(X) \rightarrow \text{Eq}(X, X)$$

Every chase sequence is terminating.

No terminating chase sequence for the substitution-free simulations.

# Function Symbols vs. EGDs

**Step 1.** Replace EGDs with TGDs via Substitution-free simulation [Mar09].

**Step 2.** Proceed as with TGDs.

# Function Symbols vs. EGDs

**Step 1.** Replace EGDs with TGDs via Substitution-free simulation [Mar09].

**Step 2.** Proceed as with TGDs.

Recall that:

## Example

### Terminating

$$\begin{aligned}p(X) \wedge q(X) &\rightarrow r(X) \\r(X) &\rightarrow \exists Y p(X) \wedge q(Y) \\r(X) &\rightarrow \exists Y p(Y) \wedge q(X) \\p(X) \wedge p(Y) &\rightarrow X = Y \\q(X) \wedge q(Y) &\rightarrow X = Y\end{aligned}$$

### Non – Terminating

$$\begin{aligned}p(X) \wedge q(X_2) \wedge eq(X, X_2) &\rightarrow r(X) \\r(X) &\rightarrow \exists Y p(X) \wedge q(Y) \\r(X) &\rightarrow \exists Y p(Y) \wedge q(X) \\p(X) \wedge p(Y) &\rightarrow eq(X, Y) \\q(X) \wedge q(Y) &\rightarrow eq(X, Y) \\eq(X, Y) &\rightarrow eq(Y, X) \\eq(X, Y) \wedge eq(Y, Z) &\rightarrow eq(X, Z) \\p(X) &\rightarrow eq(X, X) \\q(X) &\rightarrow eq(X, X) \\r(X) &\rightarrow eq(X, X)\end{aligned}$$

# Dealing with EGDs

## Example

$D = \{N(a)\}, \Sigma :$

$$N(x) \quad \rightarrow \quad \exists y E(x, y)$$

$$E(x, y) \quad \rightarrow \quad N(y)$$

$$E(x, y) \quad \rightarrow \quad x = y$$



# Dealing with EGDs

## Example

$D = \{N(a)\}, \Sigma :$

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$N(a)$

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$$N(a) \rightarrow E(a, \perp_1)$$

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$$N(a) \rightarrow E(a, a) \rightarrow a = \perp_1$$

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## Example

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$$E(x, y) \quad \rightarrow \quad x = y$$

$N(a) \rightarrow E(a, a) \rightarrow a = \perp_1 \rightarrow$  all constraints satisfied!

# Rewriting TGDs and EGDs

## Example

$$r_1 : N(x) \rightarrow \exists y E(x, y)$$

$$r_2 : E(x, y) \rightarrow N(y)$$

$$r_3 : E(x, y) \rightarrow x = y$$

# Rewriting TGDs and EGDs

## Example

$$r_1 : N(x) \rightarrow \exists y E(x, y)$$

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$$r'_3 : E^{bb}(x, y)$$



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$$r'_2 : E^{bb}(x, y)$$

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$$r'_3 : E^{bb}(x, y) \quad \rightarrow \quad x = y$$

$$r'_2 : E^{bb}(x, y) \quad \rightarrow \quad N^b(y)$$

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$$r'_1 : N^b(x)$$

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$$r'_2 : E^{bb}(x, y) \rightarrow N^b(y)$$

$$r'_1 : N^b(x) \rightarrow \exists y E^{bf_1}(x, y) \quad f_1 = f_{r'_1}^y(b)$$

# Rewriting TGDs and EGDs

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$$r'_1 : N^b(x) \quad \rightarrow \quad \exists y E^{bf_1}(x, y) \quad f_1 = f_{r'_1}^y(b)$$

$$r''_3 : E^{bf_1}(x, y)$$

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$$r'_2 : E^{bb}(x, y) \quad \rightarrow \quad N^b(y)$$

$$r'_1 : N^b(x) \quad \rightarrow \quad \exists y E^{bb}(x, y) \quad \cancel{f_1 = f_{r_1}^y(b)}$$

$$r''_3 : E^{bb}(x, y) \quad \rightarrow \quad x = y \quad b = f_1$$

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No cyclic symbol  $f_i$  occurs in the constraints above.

# Rewriting TGDs and EGDs

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$$r'_1 : N^b(x) \rightarrow \exists y E^{bb}(x, y)$$

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No cyclic symbol  $f_i$  occurs in the constraints above.

Thus, there exists a terminating standard chase sequence.

# Rewriting TGDs and EGDs

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$$r_1 : N(x) \quad \rightarrow \quad \exists y E(x, y)$$

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$$r_3 : E(x, y) \quad \rightarrow \quad x = y$$

$$r'_3 : E^{bb}(x, y) \quad \rightarrow \quad x = y$$

$$r'_2 : E^{bb}(x, y) \quad \rightarrow \quad N^b(y)$$

$$r'_1 : N^b(x) \quad \rightarrow \quad \exists y E^{bb}(x, y)$$

$$r''_3 : E^{bb}(x, y) \quad \rightarrow \quad x = y$$

No cyclic symbol  $f_i$  occurs in the constraints above.

Thus, there exists a terminating standard chase sequence.

In this sequence, EGDs are applied as soon as possible.

# Results

- The rewriting algorithm always terminates;
- $\Sigma^\alpha \in \text{CT}_{\exists}^{std}$  implies  $\Sigma \in \text{CT}_{\exists}^{std}$ ;
- Furthermore, if  $\nexists$  cyclic  $f_i$  in  $\Sigma^\alpha$ , then  $\Sigma \in \text{CT}_{\exists}^{std}$ ;

# Current and Future directions

- Determine decidable classes of data dependencies,
- Consider an extended framework ( $Datalog[\exists, =, F, \neg]$ ),
- Define criteria guaranteeing termination of one chase sequence,
- Determine how to compute one of the terminating sequences,
- Further exploiting of EGDs
- Complexity (not discussed here)
- Support for design tools.

Thanks!

Questions?



- [BBC09] S. Baselice, P. A. Bonatti, and G. Crisculo. On finitely recursive programs. *TPLP*, 9(2):213–238, 2009.
- [BLMS11] Jean-François Baget, Michel Leclère, Marie-Laure Mugnier, and Eric Salvat. On rules with existential variables: Walking the decidability line. *Artif. Intell.*, 175(9-10):1620–1654, 2011.
- [Bon04] P. A. Bonatti. Reasoning with infinite stable models. *Artificial Intelligence*, 156(1):75–111, 2004.
- [Bon11] P. A. Bonatti. On the decidability of fdnc programs. *Intelligenza Artificiale*, 5(1):89–93, 2011.
- [CCIL08] F. Calimeri, S. Cozza, G. Ianni, and N. Leone. Computable functions in ASP: Theory and implementation. In *ICLP*, pages 407–424, 2008.
- [CGK13] Andrea Cali, Georg Gottlob, and Michael Kifer. Taming the infinite chase: Query answering under expressive relational constraints. *JAIR*, 48:115–174, 2013.
- [CGMT14] M. Calautti, S. Greco, C. Molinaro, and I. Trubitsyna. Checking termination of logic programs with function

symbols through linear constraints. In *RuleML*, pages 97–111, 2014.

- [CGP10] Andrea Cali, Georg Gottlob, and Andreas Pieris. Advanced processing for ontological queries. *PVLDB*, 3(1):554–565, 2010.
- [CGP15] Marco Calautti, Georg Gottlob, and Andreas Pieris. Chase termination for guarded existential rules. In *PODS*, pages 91–103, 2015.
- [CGST14] M. Calautti, S. Greco, F. Spezzano, and I. Trubitsyna. Checking termination of bottom-up evaluation of logic programs with function symbols. *TPLP*, 2014.
- [CGT13] M. Calautti, S. Greco, and I. Trubitsyna. Detecting decidable classes of finitely ground logic programs with function symbols. In *PPDP*, 2013.
- [DNR08] A. Deutsch, A. Nash, and J. B. Remmel. The chase revisited. In *PODS*, pages 149–158, 2008.
- [ES10] Thomas Eiter and Mantas Simkus. FDNC: decidable nonmonotonic disjunctive logic programs with function symbols. *TOCL*, 11(2), 2010.

- [FKMP05] R. Fagin, P. G. Kolaitis, R. J. Miller, and L. Popa. Data exchange: semantics and query answering. *TCS*, 336(1):89–124, 2005.
- [GHK<sup>+</sup>13] Bernardo Cuenca Grau, Ian Horrocks, Markus Krotzsch, Clemens Kupke, Despoina Magka, Boris Motik, and Zhe Wang. Acyclicity notions for existential rules and their application to query answering in ontologies. *JAIR*, 47:741–808, 2013.
- [GM14] T. Gogacz and J. Marcinkowski. All-instances termination of chase is undecidable. In *ICALP*, pages 293–304, 2014.
- [GMT13a] S. Greco, C. Molinaro, and I. Trubitsyna. Logic programming with function symbols: Checking termination of bottom-up evaluation through program adornments. *TPLP*, 13(4-5):737–752, 2013.
- [GMT13b] Sergio Greco, Cristian Molinaro, and Irina Trubitsyna. Bounded programs: A new decidable class of logic programs with function symbols. In *IJCAI*, 2013.
- [GO13] Gösta Grahne and Adrian Onet. Anatomy of the chase. *CoRR*, abs/1303.6682, 2013.

- [GS10] S. Greco and F. Spezzano. Chase termination: A constraints rewriting approach. *PVLDB*, 3(1):93–104, 2010.
- [GST07] M. Gebser, T. Schaub, and S. Thiele. Gringo: A new grounder for answer set programming. In *LPNMR*, pages 266–271, 2007.
- [GST11] Sergio Greco, Francesca Spezzano, and Irina Trubitsyna. Stratification criteria and rewriting techniques for checking chase termination. *PVLDB*, 4(11):1158–1168, 2011.
- [GST15] Sergio Greco, Francesca Spezzano, and Irina Trubitsyna. Checking chase termination: Cyclicity analysis and rewriting techniques. *TKDE*, 27(3):621–635, 2015.
- [LL09] Y. Lierler and V. Lifschitz. One more decidable class of finitely ground programs. In *ICLP*, pages 489–493, 2009.
- [Mar09] B. Marnette. Generalized schema-mappings: from termination to tractability. In *PODS*, pages 13–22, 2009.
- [Mei10] Michael Meier. *On the Termination of the Chase Algorithm*. Albert-Ludwigs-Universität Freiburg (Germany), 2010.

- [MSL09] M. Meier, M. Schmidt, and G. Lausen. On chase termination beyond stratification. *CoRR*, abs/0906.4228, 2009.
- [One13] Adrian Onet. The chase procedure and its applications in data exchange. In *Data Exchange, Integration, and Streams*, pages 1–37. 2013.
- [RS14] F. Riguzzi and T. Swift. Terminating evaluation of logic programs with finite three-valued models. *ACM TOCL*, 2014.
- [Syr01] T. Syrjanen. Omega-restricted logic programs. In *LPNMR*, pages 267–279, 2001.