

# PRACTICAL USES OF EXISTENTIAL RULES IN KNOWLEDGE REPRESENTATION

## Part 3: Implementing a Calculus for Horn-*ALC* using Existential Rules

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# The Description Logic Horn- $\mathcal{ALC}$ : Syntax

**Definition.** A Horn- $\mathcal{ALC}$  ontology is a set of Horn- $\mathcal{ALC}$  axioms:

$$A \sqsubseteq \perp \quad T \sqsubseteq B \quad A \sqsubseteq B \quad A \sqcap E \sqsubseteq B \quad \exists R.A \sqsubseteq B \quad A \sqsubseteq \forall R.B \quad A \sqsubseteq \exists R.B$$

In the above;  $A$ ,  $B$ , and  $E$  are concept names; and  $R$  is a role name.

**Remark.** Note the axioms of the form  $A \sqsubseteq \forall R.B$ , which are not  $\mathcal{EL}$ , such as:

$$\text{CheesePizza} \sqsubseteq \forall \text{HasTopping}.\text{Cheese}$$

The axiom states that “all toppings in a cheese pizza are cheese toppings”.

Even though Horn- $\mathcal{ALC}$  is not much more expressive than  $\mathcal{EL}$ , (Krötzsch, Rudolph, and Hitzler 2013) have showed that:

**Theorem.** Solving classification over Horn- $\mathcal{ALC}$  is ExpTime-complete.

# The Description Logic Horn- $\mathcal{ALC}$ : Semantics

**Definition.** We define the semantics of Horn- $\mathcal{ALC}$  axioms via translation into equivalent first-order logic formulas:

$A \sqsubseteq \perp$	$\mapsto$	$\forall x.(A(x) \rightarrow \perp)$
$\top \sqsubseteq B$	$\mapsto$	$\forall x.B(x)$
$A \sqsubseteq B$	$\mapsto$	$\forall x.(A(x) \rightarrow B(x))$
$A \sqcap E \sqsubseteq B$	$\mapsto$	$\forall x.(A(x) \wedge E(x) \rightarrow B(x))$
$\exists R.A \sqsubseteq B$	$\mapsto$	$\forall x.(R(x, y) \wedge A(y) \rightarrow B(x))$
$A \sqsubseteq \forall R.B$	$\mapsto$	$\forall x.(A(x) \wedge R(x, y) \rightarrow B(y))$
$A \sqsubseteq \exists R.B$	$\mapsto$	$\forall x.(A(x) \rightarrow \exists y.R(x, y) \wedge B(y))$

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In the above;  $A$ ,  $B$ , and  $E$  are concept names, and  $R$  is a role name.

Often, we remove universal quantifiers from first-order logic formulas.

# A Consequence-Based Calculus to Solve Classification

$$\begin{array}{ll}
 R_A^C) \frac{}{A \sqsubseteq A} : A \in \text{Concepts}(O) & R_{\exists}^+) \frac{C \sqsubseteq A}{C \sqsubseteq \exists R.B} : A \sqsubseteq \exists R.B \in O \\
 R_A^{\exists}) \frac{C \sqsubseteq \exists R.D}{D \sqsubseteq D} : D \in \mathcal{D} & R_{\exists}^-) \frac{C \sqsubseteq \exists R.D \quad D \sqsubseteq A}{C \sqsubseteq B} : \exists R.A \sqsubseteq B \in O \\
 R_{\sqcap}^1) \frac{C \sqsubseteq A}{C \sqsubseteq B} : \top \sqsubseteq B \in O & R_{\exists}^{\perp}) \frac{C \sqsubseteq \exists R.D \quad D \sqsubseteq \perp}{C \sqsubseteq \perp} \\
 R_{\sqcap}^1) \frac{C \sqsubseteq A}{C \sqsubseteq B} : A \sqsubseteq B \in O & R_{\forall}^-) \frac{C \sqsubseteq \exists R.D \quad C \sqsubseteq A}{C \sqsubseteq \exists R.(D \sqcap B)} : A \sqsubseteq \forall R.B \in O \\
 R_{\sqcap}^2) \frac{C \sqsubseteq A \quad C \sqsubseteq E}{C \sqsubseteq B} : A \sqcap E \sqsubseteq B \in O &
 \end{array}$$

Figure: Classification Calculus for Horn- $\mathcal{ALC}$ . Where  $A$ ,  $B$ , and  $E$  are concept names;  $R$  is a role name; and  $C$  and  $D$  are conjunctions of concept names

**Remark.** Original calculus by (Kazakov 2009).

# Consequence-Based Calculus: Soundness

**Soundness.** Show via induction that each rule only produces sound inferences.

For instance, let us show that the following production rule is indeed sound:

$$(R_{\forall}) \frac{C \sqsubseteq \exists R.D \quad C \sqsubseteq A}{C \sqsubseteq \exists R.(D \sqcap B)} : A \sqsubseteq \forall R.B \in \mathcal{O}$$

**Proof:**

1. By IH:  $\mathcal{O} \models \bigwedge_{C \in \mathcal{C}} C(x) \rightarrow \exists y.(R(x, y) \wedge \bigwedge_{D \in \mathcal{D}} D(y))$
2. By IH:  $\mathcal{O} \models \bigwedge_{C \in \mathcal{C}} C(x) \rightarrow A(x)$
3. By the precondition of the rule:  $\mathcal{O} \models A(x) \wedge R(x, y) \rightarrow B(y)$
4. By (1–3) and the semantics of first-order logic:  
 $\mathcal{O} \models \bigwedge_{C \in \mathcal{C}} C(x) \rightarrow \exists y.(R(x, y) \wedge \bigwedge_{D \in \mathcal{D}} D(y) \wedge B(y))$

# Consequence-Based Calculus: Completeness

To show **completeness**, we verify the following theorem:

**Theorem.** If an axiom of the form  $A \sqsubseteq B$  is not derived by the previously proposed calculus on input  $O$ , then  $O \not\models A \sqsubseteq B$ .

**Proof Sketch:** Using the output of the calculus on input  $O$ , we can construct a model for this ontology that contains an element that is in the domain of  $A$  but not in the domain of  $B$ . Therefore,  $O \not\models A \sqsubseteq B$ .

**Remark.** For a complete proof, check the following references:

- (Kazakov 2009)
- (Simancik, Kazakov, and Horrocks 2011)

# Consequence-Based Calculus: Complexity

**Theorem.** The Horn- $\mathcal{ALC}$  classification calculus runs in exponential time in the size of the input ontology  $\mathcal{O}$ .

**Remark.** Note that this calculus produces inferences of the form

$$(1) \mathbb{C} \sqsubseteq B \quad \text{and} \quad (2) \mathbb{C} \sqsubseteq \exists R.D$$

where  $B$  is a concept name,  $R$  is a role name, and  $\mathbb{C}$  and  $\mathbb{D}$  are conjunctions of concept names. Therefore, the calculus may produce at most

$$2^{|\text{Concepts}(\mathcal{O})|} \times |\text{Concepts}(\mathcal{O})| \quad \text{and} \quad 2 \times 2^{|\text{Concepts}(\mathcal{O})|} \times |\text{Roles}(\mathcal{O})|$$

inferences of type (1) and (2), respectively.



# Implementing the Consequence-Based Calculus: Datalog

Because of the following result, we can not implement the Horn- $\mathcal{ALC}$  classification calculus using a fixed Datalog rule set:

**Theorem.** The data complexity of fact entailment over Datalog is in P.

## Proof:

1. Consider a Datalog rule set  $\mathcal{R}$ , a fact set  $\mathcal{F}$ , and a fact  $\varphi$ .
2. Let  $\mathcal{R}'$  be the grounding of  $\mathcal{R}$  using the constants in  $\mathcal{F}$ .
3. We have that  $\mathcal{R}' \cup \mathcal{F} \models \varphi$  if and only if  $\mathcal{R} \cup \mathcal{F} \models \varphi$ .
4. Checking if  $\mathcal{R}' \cup \mathcal{F} \models \varphi$  can be reduced to fact entailment over propositional logic, which can be solved in polynomial time.
5. If  $\mathcal{R}$  is fixed, then  $\mathcal{R}'$  is polynomial in the number of constants in  $\mathcal{F}$ .
6. By (3) and (4): if  $\mathcal{R}$  is fixed, we can decide if  $\mathcal{R} \cup \mathcal{F} \models \varphi$  in polynomial time.

# Implementing the Consequence-Based Calculus: Datalog

Because of the following result, we can not implement the Horn- $\mathcal{ALC}$  classification calculus using a fixed Datalog rule set:

**Theorem.** The data complexity of fact entailment over Datalog is in P.

Assume that we can implement the Horn- $\mathcal{ALC}$  classification calculus with a fixed Datalog rule set (as done with the  $\mathcal{EL}$  classification calculus). Then:

1. By Theorem 3.4: we could solve Horn- $\mathcal{ALC}$  classification in polynomial time.
2. By (1): we could solve an ExpTime-hard problem in polynomial time.
3. By (2):  $P = \text{ExpTime}$  ( $\zeta$ )

**Remark.** To implement the Horn- $\mathcal{ALC}$  classification calculus (or any other procedure that solves Horn- $\mathcal{ALC}$  classification), we need a rule-based language with ExpTime-hard data complexity!

# A Horn- $\mathcal{ALC}$ Classification Calculus with Datalog(S)

We study Datalog(S), an extension of Datalog that can model exponential computations.

**Example.** Consider the following Datalog(S) rule set:

$$\text{Person}(x) \rightarrow \text{LikesAll}(x, \emptyset)$$

$$\text{LikesAll}(x, X) \wedge \text{Likes}(x, y) \rightarrow \text{LikesAll}(x, X \cup \{y\})$$

$$\text{LikesAll}(x, X) \rightarrow \text{AllLikeAll}(\{x\}, X)$$

$$\text{AllLikeAll}(X, Y) \wedge \text{LikesAll}(x, Y) \rightarrow \text{AllLikeAll}(X \cup \{x\}, Y)$$

$$\text{AllLikeAll}(X, X) \wedge \text{alice} \in X \rightarrow \text{CliqueOfAlice}(X)$$

**Theorem.** Checking fact entailment for Datalog(S) is ExpTime-complete for both data and combined complexity.

See (Carral et al. 2019) for a complete proof of the above result.

# A Horn- $\mathcal{ALC}$ Classification Calculus with Datalog(S)

Using a function to encode the axioms and entities in an input ontology as facts and a fixed Datalog(S) rule set, we can implement the Horn- $\mathcal{ALC}$  classification calculus.

**Example.** For an ontology  $\mathcal{O}$ , let  $\text{Facts}(\mathcal{O})$  be the fact set such that:

$$A \sqsubseteq \perp \in \mathcal{O} \mapsto \text{ax}_{\sqsubseteq}(c_A, c_{\perp})$$

$$\exists R.A \sqsubseteq B \in \mathcal{O} \mapsto \text{ax}_{\exists\sqsubseteq}(c_A, c_R, c_B)$$

$$\top \sqsubseteq B \in \mathcal{O} \mapsto \text{ax}_{\sqsubseteq}(c_{\top}, c_B)$$

$$A \sqsubseteq \forall R.B \in \mathcal{O} \mapsto \text{ax}_{\sqsubseteq\forall}(c_A, c_R, c_B)$$

$$A \sqsubseteq B \in \mathcal{O} \mapsto \text{ax}_{\sqsubseteq}(c_A, c_B)$$

$$A \sqsubseteq \exists R.B \in \mathcal{O} \mapsto \text{ax}_{\sqsubseteq\exists}(c_A, c_R, c_B)$$

$$A \sqcap E \sqsubseteq B \in \mathcal{O} \mapsto \text{ax}_{\sqcap\sqsubseteq}(c_A, c_E, c_B)$$

$$A \in \text{Concepts}(\mathcal{O}) \mapsto \text{Concept}(c_E)$$

In the above;  $c_A$ ,  $c_B$ ,  $c_E$ ,  $c_{\top}$ , and  $c_{\perp}$  are fresh constants unique for  $A$ ,  $B$ ,  $E$ ,  $\top$ , and  $\perp$ , respectively; and  $c_R$  is a fresh constant unique  $R$ .

# A Horn- $\mathcal{ALC}$ Classification Calculus with Datalog(S)

We translate the production rules in the Horn- $\mathcal{ALC}$  classification calculus (left) into analogous Datalog(S) rules (right):

$R_A^C$	$\frac{}{A \sqsubseteq A} : A \in \text{Concepts}(O)$	Concept( $x$ ) $\rightarrow \text{SC}(\{x\}, x)$
$R_A^{\exists}$	$\frac{C \sqsubseteq \exists R.D}{D \sqsubseteq D} : D \in \mathcal{D}$	$\text{Ex}(C, r, D) \wedge d \in D$ $\rightarrow \text{SC}(D, d)$
$R_{\sqcap}^0$	$\frac{C \sqsubseteq A}{C \sqsubseteq B} : \top \sqsubseteq B \in O$	$\text{SC}(C, a) \wedge \text{ax}_{\sqsubseteq}(c_{\top}, b)$ $\rightarrow \text{SC}(C, b)$
$R_{\sqcap}^1$	$\frac{C \sqsubseteq A}{C \sqsubseteq B} : A \sqsubseteq B \in O$	$\text{SC}(C, a) \wedge \text{ax}_{\sqsubseteq}(a, b)$ $\rightarrow \text{SC}(C, b)$
$R_{\sqcap}^2$	$\frac{C \sqsubseteq A \quad C \sqsubseteq E}{C \sqsubseteq B} : A \sqcap E \sqsubseteq B \in O$	$\text{SC}(C, a) \wedge \text{SC}(C, e) \wedge \text{ax}_{\sqcap}(a, e, b)$ $\rightarrow \text{SC}(C, b)$

# A Horn- $\mathcal{ALC}$ Classification Calculus with Datalog(S)

We translate the production rules in the Horn- $\mathcal{ALC}$  classification calculus (left) into analogous Datalog(S) rules (right):

$R_{\exists}^+$	$\frac{C \sqsubseteq A}{C \sqsubseteq \exists R.B} : A \sqsubseteq \exists R.B \in \mathcal{O}$	$SC(C, a) \wedge ax_{\sqsubseteq \exists}(a, r, b) \rightarrow Ex(C, r, \{B\})$
$R_{\exists}^-$	$\frac{C \sqsubseteq \exists R.D \quad D \sqsubseteq A}{C \sqsubseteq B} : \exists R.A \sqsubseteq B \in \mathcal{O}$	$Ex(C, r, D) \wedge SC(D, a) \wedge ax_{\sqsubseteq \exists}(r, a, b) \rightarrow SC(C, b)$
$R_{\exists}^{\perp}$	$\frac{C \sqsubseteq \exists R.D \quad D \sqsubseteq \perp}{C \sqsubseteq \perp}$	$Ex(C, r, D) \wedge SC(D, c_{\perp}) \rightarrow SC(C, c_{\perp})$
$R_{\forall}^{\perp}$	$\frac{C \sqsubseteq \exists R.D \quad C \sqsubseteq A}{C \sqsubseteq \exists R.(D \sqcap B)} : A \sqsubseteq \forall R.B$	$Ex(C, r, D) \wedge SC(C, a) \wedge ax_{\sqsubseteq \forall}(a, r, b) \rightarrow Ex(C, r, D \cup \{b\})$

# A Horn- $\mathcal{ALC}$ Classification Calculus with Datalog(S)

**Definition.** Let  $\mathcal{R}_{\text{HALC}}$  be the following Datalog(S) rule set:

$$\begin{aligned} \text{Concept}(x) &\rightarrow \text{SC}(\{x\}, x) & \text{SC}(C, a) \wedge \text{ax}_{\sqsubseteq}(c_{\top}, b) &\rightarrow \text{SC}(C, b) \\ \text{Ex}(C, r, D) \wedge d \in D &\rightarrow \text{SC}(D, d) & \text{SC}(C, a) \wedge \text{ax}_{\sqsubseteq}(a, b) &\rightarrow \text{SC}(C, b) \\ \text{SC}(C, a) \wedge \text{SC}(C, e) \wedge \text{ax}_{\sqcap\sqsubseteq}(a, e, b) &\rightarrow \text{SC}(C, b) \\ \text{SC}(C, a) \wedge \text{ax}_{\sqsubseteq\exists}(a, r, b) &\rightarrow \text{Ex}(C, r, \{B\}) \\ \text{Ex}(C, r, D) \wedge \text{SC}(D, a) \wedge \text{ax}_{\exists\sqsubseteq}(r, a, b) &\rightarrow \text{SC}(C, b) \\ \text{Ex}(C, r, D) \wedge \text{SC}(D, c_{\perp}) &\rightarrow \text{SC}(C, c_{\perp}) \\ \text{Ex}(C, r, D) \wedge \text{SC}(C, a) \wedge \text{ax}_{\sqsubseteq\forall}(a, r, b) &\rightarrow \text{Ex}(C, r, D \cup \{b\}) \end{aligned}$$

**Theorem.** Consider a Horn- $\mathcal{ALC}$  ontology  $\mathcal{O}$  and an axiom of the form  $A \sqsubseteq B$ . Then,  $\mathcal{O} \models A \sqsubseteq B$  if and only if  $\mathcal{R}_{\text{HALC}} \cup \text{Facts}(\mathcal{O}) \models \text{SC}(c_A, c_B)$ .

# Implementing the Classification Calculus with VLog

Alas, VLog does not support Datalog(S) reasoning. There maybe some other rule-based language that we can use...

The following result is a recent finding by (Krötzsch, Marx, and Rudolph 2019):

**Theorem.** The data complexity of fact entailment over rule sets that terminate with respect to the restricted chase is ExpTime-hard.

Moreover, (Carral et al. 2019) have proposed a translation from Datalog(S) into existential rule programs such that:

- The resulting programs terminate with respect to the restricted chase.
- Fact entailment is preserved.



# From Datalog(S) to Existential Rules

$\text{Person}(x) \rightarrow \text{LikesAll}(x, \emptyset)$        $\text{LikesAll}(x, X) \wedge \text{Likes}(x, y) \rightarrow \text{LikesAll}(x, X \cup \{y\})$

$$\rightarrow \exists V. \text{empty}(V) \tag{1.1}$$

$$\text{person}(x) \wedge \text{empty}(Y) \rightarrow \text{likesAll}(x, Y) \tag{1.2}$$

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# From Datalog(S) to Existential Rules

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*person(eve)*

*likes(eve, a)*

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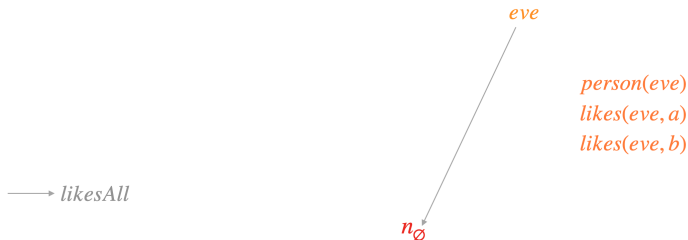
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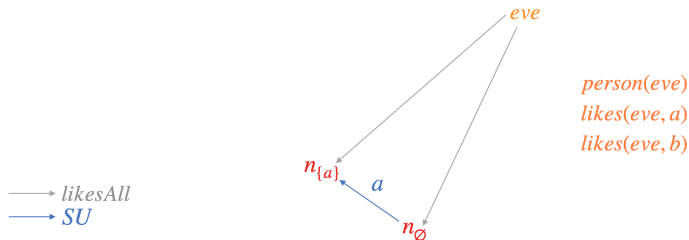
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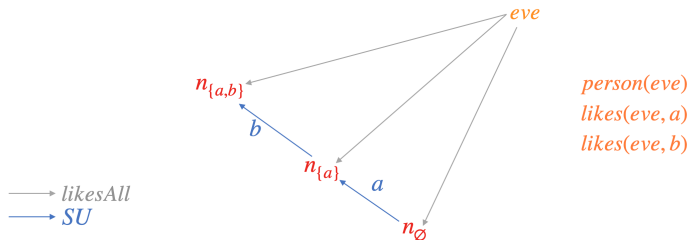
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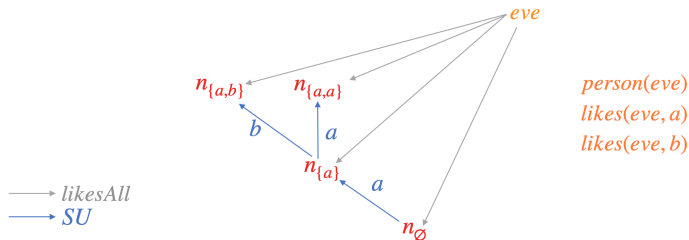
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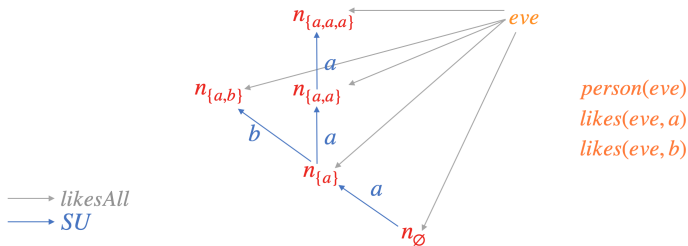
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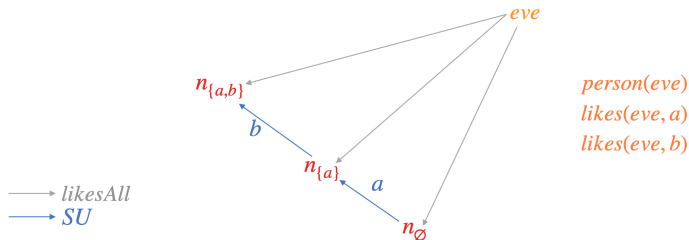
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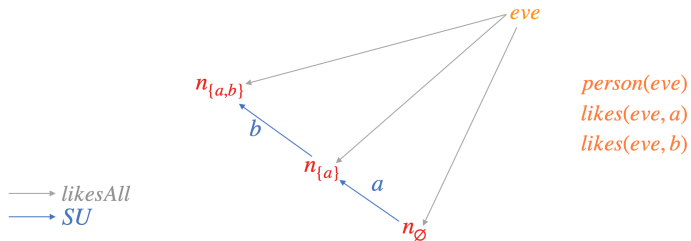
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$$\rightarrow \exists V. \text{empty}(V) \quad (1.1)$$

$$\text{person}(x) \wedge \text{empty}(Y) \rightarrow \text{likesAll}(x, Y) \quad (1.2)$$

$$\text{likesAll}(x, S) \wedge \text{likes}(x, y) \rightarrow \exists V. \text{likesAll}(x, V) \wedge \text{SU}(S, y, V) \wedge \text{SU}(V, y, V) \quad (2.1)$$



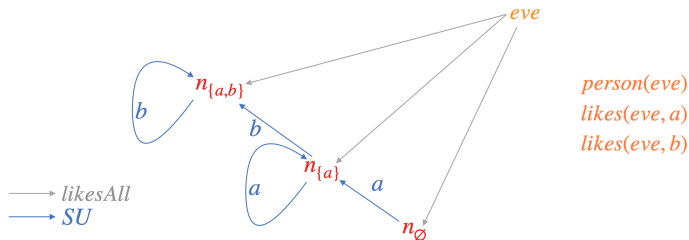
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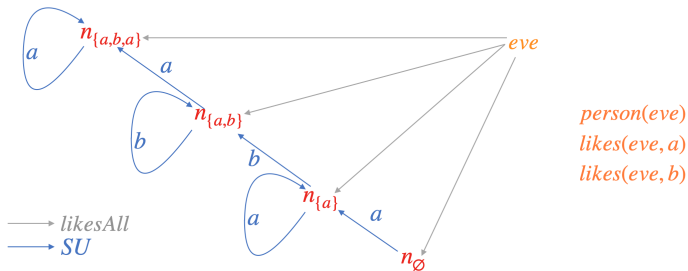
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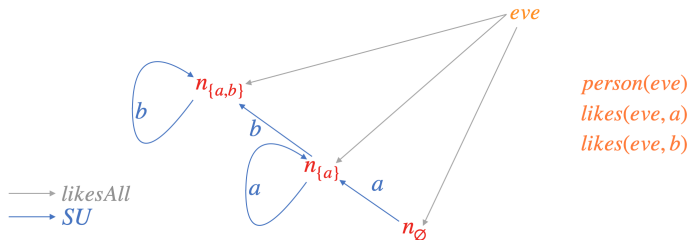
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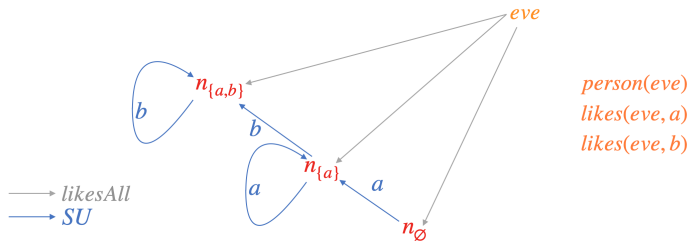
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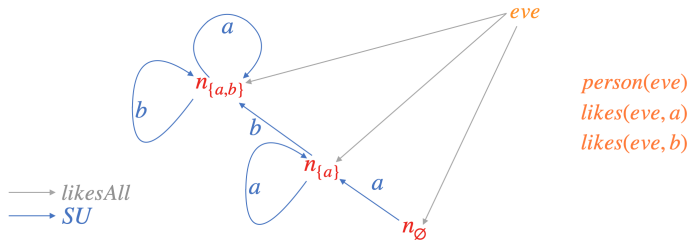
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## Experimental Evaluation: Solving Classification

ID	#Ax.	#Set	#SC	#Ex	VLog	Konclude
00040	223K	2K	1051K	334K	432s	5s
00048	142K	19	718K	171K	387s	3s
00477	318K	0	162K	167K	1s	3s
00533	159K	0	965K	351K	132s	2s
00786	152K	12K	2283K	978K	549s	14s

Figure: Ontologies and results for classification (A) showing: axiom count; number of non-singleton “set terms” introduced (#Set); number of SC and Ex facts derived; reasoning time in VLog and Konclude

# Experimental Evaluation: Assertion Retrieval

**Definition.** A Horn- $\mathcal{ALC}$  ontology is a set of Horn- $\mathcal{ALC}$  axioms:

$$\begin{array}{cccc} A \sqsubseteq \perp & \top \sqsubseteq B & A \sqsubseteq B & A \sqcap E \sqsubseteq B \\ \exists R.A \sqsubseteq B & A \sqsubseteq \forall R.B & A \sqsubseteq \exists R.B & A(a) \quad R(a,b) \end{array}$$

In the above;  $A$ ,  $B$ , and  $E$  are concept names (i.e., unary predicates); and  $R$  is a role name (i.e., binary predicate).

**Definition.** Assertion Retrieval is the reasoning task of computing all axioms of the form  $A(a)$  or  $R(a,b)$  that are logically entailed by some input ontology  $\mathcal{O}$ .

**Remark.** The Horn- $\mathcal{ALC}$  classification calculus can be extended with 3 rules (as done by (Carral et al. 2019)) to solve assertion retrieval.

# Experimental Evaluation: Assertion Retrieval

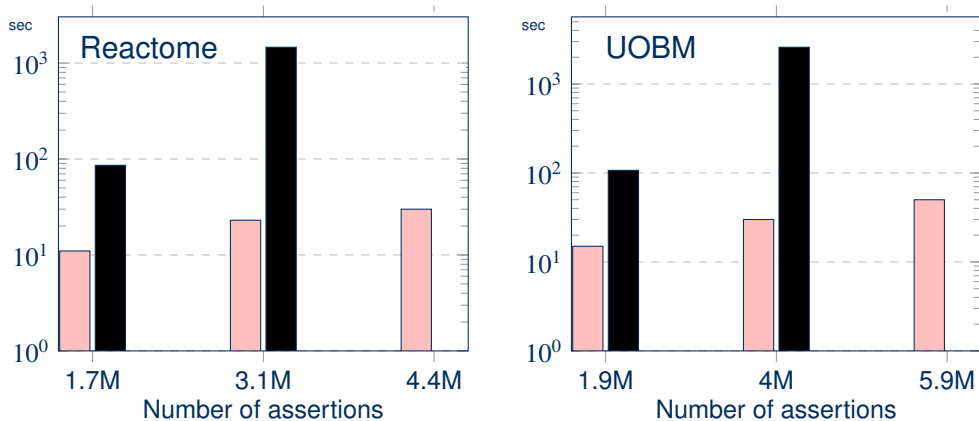


Figure: Experimental results for class retrieval (B) in VLog (pink/grey) and Konclude (black); note the log scale

## Conclusions and Future Work






**Remark.** We can use VLog to solve ExpTime-hard problems!

Future work:

- Rulewerk Extension: translate Datalog(S) to existential rules
- VLog Extension: native support for Datalog(S)
- Implement existing calculi using our approach

Hands on Session!

## References I

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