

FOUNDATIONS OF SEMANTIC WEB TECHNOLOGIES

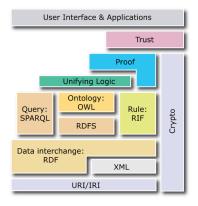
RDFS Rule-based Reasoning

Sebastian Rudolph



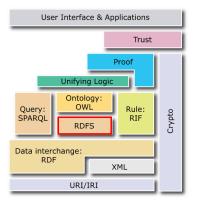


RDFS Rule-based Reasoning





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Agenda

- Rules
 - Llyod-Topor Transformation
- Datalog
 - Characterizations of Datalog Program Semantics
- Evaluating Datalog Programs
 - Naïve Evaluation
 - Semi-naïve Evaluation
- Rules for RDFS via a Triple Predicate
- Rules for RDFS via Direct Translation



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Constituents of Rules

- · basic elements of rules are atoms
 - ground atoms without free variables
 - non-ground atoms with free variables



What are Rules?

- 1 logic rules (fragments of predicate logic):
 - F → G equivalent to $\neg F \lor G$
 - logical extension of knowledge base → static
 - open world
 - declarative (describing)



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- logic rules (fragments of predicate logic):
 - F → G equivalent to $\neg F \lor G$
 - logical extension of knowledge base → static
 - open world
 - declarative (describing)
- procedural rules (e.g. production rules):
 - "If X then Y else Z"
 - executable commands → dynamic
 - operational (meaning = effect caused when executed)



What are Rules?

- logic rules (fragments of predicate logic):
 - F → G equivalent to $\neg F \lor G$
 - logical extension of knowledge base → static
 - open world
 - declarative (describing)
- 2 procedural rules (e.g. production rules):
 - "If X then Y else Z"
 - executable commands → dynamic
 - operational (meaning = effect caused when executed)
- 3 logic programming et al. (e.g. PROLOG, F-Logic):
 - man(X) <- person(X) AND NOT woman(X)
 - approximation of logical semantics with operational aspects, built-ins are possible
 - often closed-world
 - semi-declarative



Predicate Logic as a Rule Language

• rules as implication formulae in predicate logic:

$$\underbrace{H}_{\text{head}} \leftarrow \underbrace{A_1 \wedge A_2 \wedge \ldots \wedge A_n}_{\text{body}}$$



Predicate Logic as a Rule Language

rules as implication formulae in predicate logic:

$$H$$
 $\leftarrow \underbrace{A_1 \wedge A_2 \wedge \ldots \wedge A_n}_{\text{body}}$

→ semantically equivalent to disjunction:

$$H \vee \neg A_1 \vee \neg A_2 \vee \ldots \vee \neg A_n$$

- implications often written from right to left(← or :-)
- constants, variables and function symbols allowed
- quantifiers for variables are often omitted: free variables are often understood as universally quantified (i.e. rule is valid for all variable assignments)



Rules – Example

Example:

```
hasUncle(x, z) \leftarrow hasParent(x, y) \land hasBrother(y, z)
```

- we use short names (hasUncle) instead of IRIs like http://example.org/Example#hasUncle
- we use x,y,z for variables



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Lloyd-Topor Transformation

• multiple heads in atoms are usually understood as conjunction

$$H_1, H_2, \dots, H_m \leftarrow A_1, A_2, \dots, A_n$$
 equivalent to
$$H_1 \leftarrow A_1, A_2, \dots, A_n$$

$$H_2 \leftarrow A_1, A_2, \dots, A_n$$

$$\dots$$

$$H_m \leftarrow A_1, A_2, \dots, A_n$$

• such a rewriting is also referred to as Lloyd-Topor transformation



Disjunctive Rules

- some rule formalisms allow for disjunction
- → several atoms in the head are conceived as alternatives:

$$H_1, H_2, \ldots, H_m \leftarrow A_1, A_2, \ldots, A_n$$

equivalent to

$$H_1 \vee H_2 \vee \ldots \vee H_m \leftarrow A_1 \wedge A_2 \wedge \ldots \wedge A_n$$

equivalent to

$$H_1 \vee H_2 \vee \ldots \vee H_m \vee \neg A_1 \vee \neg A_2 \vee \ldots \vee \neg A_n$$



- clause: disjunction of atomic and negated atomic propositions
 - Woman(x) ∨ Man(x) ← Person(x)



- clause: disjunction of atomic and negated atomic propositions
 - Woman $(x) \vee \text{Man}(x) \leftarrow \text{Person}(x)$
- Horn clause: clause with at most one non-negated atom
 - $\leftarrow \mathsf{Man}(x) \wedge \mathsf{Woman}(x)$



- clause: disjunction of atomic and negated atomic propositions
 - Woman $(x) \vee \text{Man}(x) \leftarrow \text{Person}(x)$
- Horn clause: clause with at most one non-negated atom
 - $\leftarrow \mathsf{Man}(x) \wedge \mathsf{Woman}(x)$
- definite clause: Horn clause with exactly one non-negated atom
 - Father(x) \leftarrow Man(x) \wedge hasChild(x, y)



- clause: disjunction of atomic and negated atomic propositions
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- definite clause: Horn clause with exactly one non-negated atom
 - Father(x) \leftarrow Man(x) \wedge hasChild(x, y)
- fact: clause containing just one non-negated atom
 - Woman(gisela)



Rules may also contain function symbols:

```
\mathsf{hasUncle}(x,y) \leftarrow \mathsf{hasBrother}(\mathsf{mother}(x),y)\mathsf{hasFather}(x,\mathsf{father}(x)) \leftarrow \mathsf{Person}(x)
```

- → new elements are dynamically generated
- → not considered here
- → see logic programming



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Datalog

Horn rules without function symbols → Datalog rules

- logical rule language, originally basis of deductive databases
- knowledge bases ("programs") consisting of Horn clauses without function symbols
- decidable
- efficient for big datasets, combined complexity ExpTime
- a lot of research done in the 1980s



Datalog as Extension of the Relation Calculus

Datalog can be conceived as Extension of the relation calculus by recursion

$$T(x, y) \leftarrow E(x, y)$$

 $T(x, y) \leftarrow E(x, z) \wedge T(z, y)$

→ computes the transitive closure (T) of the binary relation E, (e.g. if E contains the edges of a graph)

• a set of (ground) facts is also called an instance



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Semantics of Datalog

three different but equivalent ways to define the semantics:

- model-theoretically
- proof-theoretically
- via fixpoints



Model-theoretic Semantics of Datalog

rules are seen as logical sentences:

$$\forall x, y. (T(x, y) \leftarrow E(x, y))$$
$$\forall x, y. (T(x, y) \leftarrow E(x, z) \land T(z, y))$$

- not sufficient to uniquely determine a solution
- \rightsquigarrow interpretation of T has to be minimal



Model-theoretic Semantics of Datalog

in principle, a Datalog rule

$$\rho: R_1(u_1) \leftarrow R_2(u_2), \ldots, R_n(u_n)$$

represents the FOL sentence

$$\forall x_1,\ldots,x_n.(R_1(u_1)\leftarrow R_2(u_2)\wedge\ldots\wedge R_n(u_n))$$

- x_1, \ldots, x_n are the rule's variables and \leftarrow is logical implication
- an instance I satisfies ρ , written $I \models \rho$, if and only if for every instantiation

$$R_1(\nu(u_1)) \leftarrow R_2(\nu(u_2)), \ldots, R_n(\nu(u_n))$$

we find $R_1(\nu(u_1))$ satisfied whenever $R_2(\nu(u_2)), \ldots, R_n(\nu(u_n))$ are satisfied



Model-theoretic Semantics of Datalog

- an instance I is a model of a Datalog program P, if I satisfies every rule in P (seen as a FOL formula)
- the semantics of P for the input I is the minimal model that contains I (if it exists)
- Question: does such a model always exist?
- If so, how can we construct it?



based on proofs for facts:

given:
$$E(a,b), E(b,c), E(c,d)$$

 $T(x,y) \leftarrow E(x,y)$ (1)
 $T(x,y) \leftarrow E(x,z) \wedge T(z,y)$ (2)

- (a) E(c, d) is a given fact
- (b) T(c, d) follows from (1) and (a)
- (c) E(b, c) is a given fact
- (d) T(b, d) follows from (c), (b) and (2)
- (e) ...



- programs can be seen as "factories" that produce all provable facts (deriving new facts from known ones in a bottom-up way by applying rules)
- alternative: top-down evaluation; starting from a to-be-proven fact, one looks for lemmata needed for the proof (→ Resolution)



a fact is provable, if it has a proof, represented by a proof-tree:

Definition

A proof tree for a fact A for an instance I and a Datalog program P is a labeled tree in which

- every node is labeled with a fact
- 2 every leaf is labeled with a fact from I
- the root is labeled with A
- \P for each internal leaf there exists an instantiation $A_1 \leftarrow A_2, \ldots, A_n$ of a rule in P, such that the node is labeled with A_1 and its children with A_2, \ldots, A_n

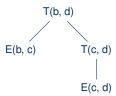


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Fixpoint Semantics

defines the semantics of a Datalog program as the solution of a fixpoint equation

- procedural definition (iteration until fixpoint reached)
- given an instance I and a Datalog program P, we call a fact A a direct consequence for P and I, if
 - A is contained in I or
 - $A \leftarrow A_1, \dots, A_n$ is the instance of a rule from P, such that $A_1, \dots, A_n \in I$
- then we can define a "direct consequence"-operator that computes, starting from an instance, all direct consequences
- similar to the bottom-up proof-theoretic semantics, but shorter proofs are always generated earlier than longer ones



Semantics of Rules

- compatible with other approaches that are based on FOL (e.g. description logics)
- conjunctions in rule heads and disjunction in bodies unnecessary
- other (non-monotonic) semantics definitions possible
 - well-founded semantics
 - stable model semantics
 - answer set semantics
- · for Horn rules, these definitions do not differ
- production rules/procedural rules conceive the consequence of a rule as an action "If-then do"
 - → not considered here



Extensional and Intensional Predicates

- from the database perspective (and opposed to logic programming) one distinguishes facts and rules
- within rules, we distinguish extensional and intensional predicates
- extensional predicates (also: extensional database edb) are those not occurring in rule heads (in our example: relation E)
- intensional predicates (also: intensional database idb) are those occurring in at least one head of a rule (in our example: relation T)
- semantics of a datalog program can be understood as a mapping of given instances over edb predicates to instances of idb predicates



Datalog in Practice

Datalog in Practice:

- several implementations available
- ullet some adaptations for Semantic Web: XSD types, URIs (e.g. ightarrow IRIS)

Extensions of Datalog:

- disjunctive Datalog allows for disjunctions in rule heads
- non-monotonic negation (no FOL semantics)



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Evaluating Datalog Programs

- top-down or bottom-up evaluation
- direct evaluation versus compilation into an efficient program
- here:
 - Naïve bottom-up EvaluierungSemi-naïve bottom-up Evaluierung



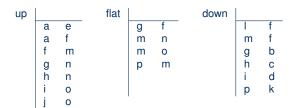
Reverse-Same-Generation

given Datalog programm:

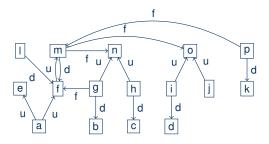
$$rsg(x,y) \leftarrow flat(x,y)$$

$$rsg(x,y) \leftarrow up(x,x_1), rsg(y_1,x_1), down(y_1,y)$$

given data:

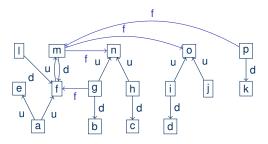






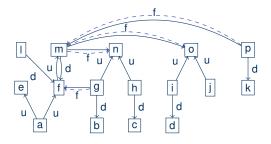
$$rsg(x, y) \leftarrow flat(x, y) rsg(x, y) \leftarrow up(x, x_1), rsg(y_1, x_1), down(y_1, y)$$





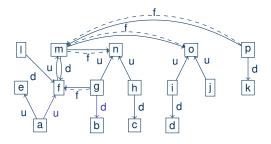
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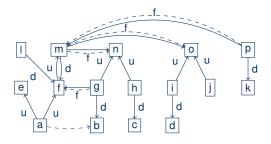
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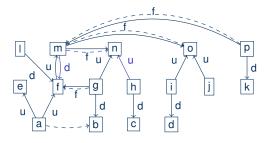




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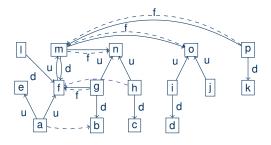
 $rsg(x, y) \leftarrow up(x, x_1), rsg(y_1, x_1), down(y_1, y)$





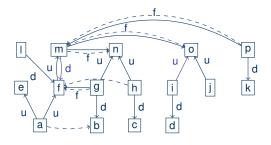
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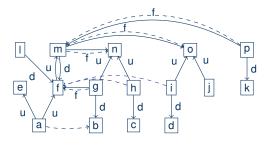
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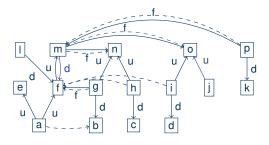
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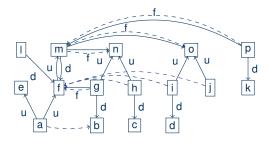
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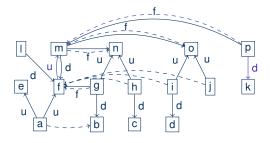




$$rsg(x, y) \leftarrow flat(x, y)$$

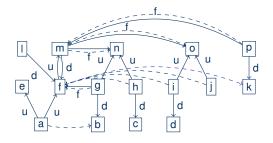
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Naïve Algorithm for Computing rsg

```
rsg(x, y) \leftarrow flat(x, y)

rsg(x, y) \leftarrow up(x, x_1), rsg(y_1, x_1), down(y_1, y)
```

Algorithm 1 RSG

```
rsg := \emptyset
repeat
rsg := rsg \cup flat \cup \pi_{16}(\sigma_{2=4}(\sigma_{3=5}(up \times rsg \times down)))
until fixpoint reached
```

```
 \begin{array}{l} \mathit{rsg}^{i+1} := \mathit{rsg}^{i} \cup \mathit{flat} \cup \pi_{16}(\sigma_{2=4}(\sigma_{3=5}(\mathit{up} \times \mathit{rsg} \times \mathit{down}))) \\ \mathsf{Level} \ 0 \colon \quad \emptyset \\ \mathsf{Level} \ 1 \colon \quad \{(g,f),(m,n),(m,o),(p,m)\} \\ \mathsf{Level} \ 2 \colon \quad \{\mathsf{Level} \ 1\} \cup \{(a,b),(h,f),(i,f),(j,f),(f,k)\} \\ \mathsf{Level} \ 3 \colon \quad \{\mathsf{Level} \ 2\} \cup \{(a,c),(a,d)\} \\ \mathsf{Level} \ 4 \colon \quad \{\mathsf{Level} \ 3\} \\ \mathsf{TU} \ \mathsf{Dresden} \qquad \qquad \mathsf{Foundations} \ \mathsf{of} \ \mathsf{Semantic} \ \mathsf{Web} \ \mathsf{Technologies} \\ \end{aligned}
```



Naïve Algorithm for Evaluating Datalog Programs

- redundant computations (all elements of the preceding level are taken into account)
- on each level, all elements of the preceding level are re-computed
- monotone (rsg is extended more and more)



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Semi-Naïve Algorithm for Computing rsg

focus on facts that have been newly computed on the preceding level

Algorithm 2 RSG'

```
\Delta_{rsg}^{1}(x, y) := flat(x, y) 
 \Delta_{rsg}^{i+1}(x, y) := up(x, x_1), \Delta_{rsg}^{i}(y_1, x_1), down(y_1, y)
```

- not recursive
- no Datalog program (set of rules is infinite)
- for each input I and Δ_{rsg}^{i} (the newly computed instances on level i),

$$rsg^{i+1} - rsg^i \subseteq \Delta_{rsg}^{i+1} \subseteq rsg^{i+1}$$

- RSG(I)(rsg) = $\bigcup_{1 \leq i} (\Delta_{rsp}^i)$
- less redundancy



An Improvement

```
But: \Delta_{rsg}^{i+1} \neq rsg^{i+1} - rsg^{i}

e.g.: (g,f) \in \Delta_{rsg}^{2}, (g,f) \notin rsg^{2} - rsg^{1}

\Rightarrow rsg(g,f) \in rsg^{1}, because \mathit{flat}(g,f),

\Rightarrow rsg(g,f) \in \Delta_{rsg}^{2}, because \mathit{up}(g,n), rsg(m,f), down(m,f)
```

• idea: use $rsg^i - rsg^{i-1}$ instead of Δ^i_{rsg} in the second "rule" of RSG'

Algorithm 3 RSG"

```
\begin{split} & \Delta_{rsg}^{1}(x,y) := \mathit{flat}(x,y) \\ & rsg^{1} := \Delta_{rsg}^{1} \\ & \mathit{tmp}_{rsg}^{i+1}(x,y) := \mathit{up}(x,x_{1}), \Delta_{rsg}^{i}(y_{1},x_{1}), \mathit{down}(y_{1},y) \\ & \Delta_{rsg}^{i+1}(x,y) := \mathit{tmp}_{rsg}^{i+1} - rsg^{i} \\ & rsg^{i+1} := rsg^{i} \cup \Delta_{rsg}^{i+1} \end{split}
```



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Datalog Rules for RDFS (no Datatypes & Literals)

problem: no strict separation between data and schema (predicates)

$$\frac{\text{a rdfs:domain x . u a y .}}{\text{u rdf:type x .}} \ \text{rdfs2}$$

$$\text{rdf:type}(u,x) \leftarrow \text{rdfs:domain}(a,x) \land a(u,y)$$

• solution: use a triple predicate



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$$\frac{\text{a rdfs:domain x . u a y .}}{\text{u rdf:type x .}} \text{ rdfs2}$$

$$Triple(u, rdf: type, x) \leftarrow Triple(a, rdfs: domain, x) \wedge Triple(u, a, y)$$

- usage of just one predicate reduces optimization potential
- all (newly derived) triples are potential candidates for any rule
- rules change when the data changes, no separation between schema and data



Datalog Rules for RDFS (no Datatypes & Literals)

• solution 2: introduce specific predicates

$$\frac{\text{a rdfs:domain x . u a y .}}{\text{u rdf:type x .}} \text{ rdfs2}$$

$$\text{type}(u,x) \leftarrow \text{domain}(a,x) \land \text{rel}(u,a,y)$$



Axiomatic Triples as Facts

```
type(rdf:type, rdf:Property)
type(rdf:subject, rdf:Property)
type(rdf:predicate, rdf:Property)
type(rdf:object, rdf:Property)
type(rdf:first, rdf:Property)
type(rdf:rest, rdf:Property)
type(rdf:value, rdf:Property)
type(rdf:l, rdf:Property)
type(rdf:2, rdf:Property)
type(rdf:l, rdf:Property)
type(rdf:l, rdf:Property)
type(rdf:l, rdf:Property)
type(..., rdf:Property)
type(rdf:nil, rdf:List)
...(plus RDFS axiomatic triples)
```



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type(rdf:rest, rdf:Property)
type(rdf:value, rdf:Property)
type(rdf:_1, rdf:Property)
type(rdf:_2, rdf:Property)
type(..., rdf:Property)
type(rdf:nil, rdf:List)
...(plus RDFS axiomatic triples)
```

 \rightsquigarrow only needed for those rdf:_i that occur in the graphs G_1 and G_2 , if $G_1 \models^? G_2$ is to be decided



```
\frac{\text{u a y}}{\text{a rdf:type rdf:Property}} \text{ rdf1}
\Rightarrow \text{type(a, rdf:Property)} \leftarrow \text{rel(u, a, y)}
```

```
a, b IRIs x, y IRI, blank node or literal u, y IRI or blank node 1 literal :n blank nodes
```



```
\begin{array}{c} u \text{ a } y \\ \hline \text{a rdf:type rdf:Property} \\ \hline & \text{rdf:type rdf:Property} \leftarrow \text{rel}(u, a, y) \\ \hline \\ \frac{\text{a rdfs:domain } x \cdot u \text{ a } y \cdot r\text{dfs2}}{u \text{ rdf:type } x \cdot r\text{dfs2}} \\ \hline \\ & \text{$\leftrightarrow$ type}(u, x) \leftarrow \text{domain}(a, x) \land \text{rel}(u, a, y) \\ \end{array}
```

```
a, b IRIs x, y IRI, blank node or literal u, y IRI or blank node 1 literal :n blank nodes
```



```
\begin{array}{c} u \text{ a } y \\ \hline \text{a rdf:type rdf:Property} \\ \hline \rightarrow \text{type(a, rdf:Property)} \leftarrow \text{rel(u, a, y)} \\ \hline \\ a \text{ rdfs:domain } x \text{ . } u \text{ a } y \text{ .} \\ \hline \\ u \text{ rdf:type } x \text{ .} \\ \hline \\ \hline \rightarrow \text{type(u, x)} \leftarrow \text{domain(a, x)} \land \text{rel(u, a, y)} \\ \hline \\ a \text{ rdfs:range } x \text{ . } u \text{ a } v \text{ .} \\ \hline \\ v \text{ rdf:type } x \text{ .} \\ \hline \\ \hline \\ v \text{ rdf:type } x \text{ .} \\ \hline \\ \rightarrow \text{type(v, x)} \leftarrow \text{range(a, x)} \land \text{rel(u, a, v)} \\ \hline \end{array}
```

```
a, b IRIs x, y IRI, blank node or literal u. y IRI or blank node 1 literal .:n blank nodes
```



```
u a y rdf1 a rdf:type rdf:Property

→ type(a, rdf:Property) ← rel(u, a, y)
a rdfs:domain x . u a y . rdfs2
       u rdf:type x .
\rightsquigarrow type(u, x) \leftarrow domain(a, x) \land rel(u, a, y)
a rdfs:range x . u a v . rdfs3
\rightsquigarrow type(v, x) \leftarrow range(a, x) \land rel(u, a, v)
u a x .
u rdf:type rdfs:Resource . rdfs4a

→ type(u, rdfs:Resource) ← rel(u, a, x)

a, b IRIs x, y IRI, blank node or literal
u. v IRI or blank node 1 literal _:n blank nodes
```



```
\frac{\text{u a v.}}{\text{v rdf:type rdfs:Resource.}} rdfs4b
\Rightarrow type(\text{v, rdfs:Resource}) \leftarrow rel(\text{u, a, v})
```

```
a, b IRIs x, y IRI, blank node or literal u, v IRI or blank node 1 literal _:n blank nodes
```



```
\label{eq:continuous_propertyOf(u, x)} \begin{array}{c} u \text{ a v .} \\ \hline v \text{ rdf:type rdfs:Resource .} \\ \hline & \text{type(v, rdfs:Resource)} \leftarrow \text{rel}(u, a, v) \\ \\ \hline u \text{ rdfs:subPropertyOf v . } v \text{ rdfs:subPropertyOf x .} \\ \hline u \text{ rdfs:subPropertyOf x .} \\ \hline \\ & \text{subPropertyOf(u, x)} \leftarrow \text{subPropertyOf(u, v)} \wedge \text{subPropertyOf(v, x)} \\ \end{array}
```

```
a, b IRIs x, y IRI, blank node or literal u, y IRI or blank node 1 literal .:n blank nodes
```



```
u a v .
v rdf:type rdfs:Resource . rdfs4b

→ type(v, rdfs:Resource) ← rel(u, a, v)
u rdfs:subPropertyOf v . v rdfs:subPropertyOf x . rdfs5
                u rdfs:subPropertyOf x .
\rightsquigarrow subPropertyOf(u, x) \leftarrow subPropertyOf(u, v) \land subPropertyOf(v, x)
u rdf:type rdf:Property .
u rdfs:subPropertyOf u .rdfs6

→ subPropertyOf(u, u) ← type(u, rdf:Property)

a, b IRIs x, y IRI, blank node or literal
u. v IRI or blank node 1 literal _:n blank nodes
```



```
u a v .
v rdf:type rdfs:Resource . rdfs4b

→ type(v, rdfs:Resource) ← rel(u, a, v)
u rdfs:subPropertyOf v . v rdfs:subPropertyOf x . rdfs5
                u rdfs:subPropertyOf x .
\rightsquigarrow subPropertyOf(u, x) \leftarrow subPropertyOf(u, v) \land subPropertyOf(v, x)
u rdf:type rdf:Property .
u rdfs:subPropertyOf u .rdfs6

→ subPropertyOf(u, u) ← type(u, rdf:Property)

a rdfs:subPropertyOf b . u a y . rdfs7
\rightsquigarrow rel(u, b, v) \leftarrow subPropertyOf(a, b) \land rel(u, a, v)
a, b IRIs x, y IRI, blank node or literal
u. v IRI or blank node 1 literal _:n blank nodes
```



```
\frac{\text{u rdf:type rdfs:Class.}}{\text{u rdf:subClassOf rdfs:Resource.}} \cdot rdfs8
\Rightarrow subClassOf(u, rdfs:Resource) \leftarrow type(u, rdfs:Class)
```

```
a, b IRIs x, y IRI, blank node or literal u, y IRI or blank node 1 literal _:n blank nodes
```



```
u rdf:type rdfs:Class .
u rdf:subClassOf rdfs:Resource .

subClassOf(u, rdfs:Resource) ← type(u, rdfs:Class)

u rdfs:subClassOf x . v rdf:type u .
v rdf:type x .

type(v, x) ← subClassOf(u, x) ∧ type(v, x)
```

```
a, b IRIs x, y IRI, blank node or literal u, y IRI or blank node 1 literal _:n blank nodes
```



```
u rdf:type rdfs:Class . rdfs8

w rdf:subClassOf rdfs:Resource . rdfs8

subClassOf(u, rdfs:Resource) ← type(u, rdfs:Class)

u rdfs:subClassOf x . v rdf:type u . rdfs9

v rdf:type x . rdfsype(v, x) ← subClassOf(u, x) ∧ type(v, x)

u rdf:type rdfs:Class . rdfs10

w rdfs:subClassOf(u, u) ← type(u, rdfs:Class)
```

```
a, b IRIs x, y IRI, blank node or literal u, v IRI or blank node 1 literal _:n blank nodes
```



```
u rdf:type rdfs:Class . rdfs8

→ subClassOf(u, rdfs:Resource) ← type(u, rdfs:Class)

u rdfs:subClassOf x . v rdf:type u . rdfs9
           v rdf:type x .
\rightsquigarrow type(v, x) \leftarrow subClassOf(u, x) \land type(v, x)
u rdf:type rdfs:Class . rdfs10
u rdfs:subClassOf v . v rdfs:subClassOf x . rdfs11
            u rdfs:subClassOf x .
\rightsquigarrow subClassOf(u, x) \leftarrow subClassOf(u, v) \land subClassOf(v, x)
a, b IRIs x, y IRI, blank node or literal
u, v IRI or blank node 1 literal _:n blank nodes
```



```
u rdf:type rdfs:ContainerMembershipProperty · rdfs12
u rdfs:subPropertyOf rdfs:member .

→ subPropertyOf(u, rdfs:member) ← type(u, rdfs:ContainerMembershipProperty)

a, b IRIs
u, v IRI or blank node 1 literal .:n blank nodes
```



Agenda

- Rules
 - Llyod-Topor Transformation
- Datalog
 - Characterizations of Datalog Program Semantics
- Evaluating Datalog Programs
 - Naïve Evaluation
 - Semi-naïve Evaluation
- Rules for RDFS via a Triple Predicate
- Rules for RDFS via Direct Translation