

## RDFS Semantics and OWL 2 Profiles

#### Ernesto Jiménez-Ruiz

Lecturer in Artificial Intelligence

Before we start...

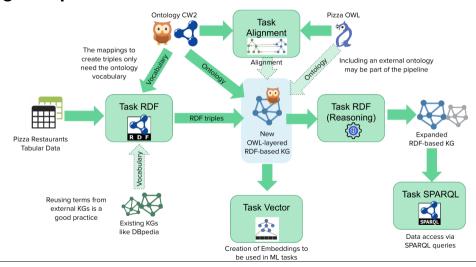
### **Drop-in sessions**

- Today from **3pm** to **5pm** (shifted by 1h).
- Tuesdays, online 10am.

### Where are we? Module organization.

- ✓ Introduction: Becoming a knowledge scientist.
- RDF-based knowledge graphs.
- ✓ OWL ontology language. Focus on modelling.
- SPARQL 1.0 Query Language.
- From tabular data to KG.
- 6. RDFS Semantics and OWL 2 profiles.(Today)
- 7. SPARQL 1.1, Rules and Graph Database solutions.
- 8. Ontology Alignment.
- 9. Ontology (KG) Embeddings and Machine Learning.
- 10. (Large) Language Models and KGs. (Seminar)

### The global picture



### Coursework part 2

- Sunday, 12 May 2024, 5:00 PM
- Team registration March 17
- Components:
  - ✓ Tabular Data to Knowledge Graph: 40% (Weeks 2 and 5)
  - SPARQL and Reasoning: 20% (Weeks 4, 7 and 8)
  - Ontology Alignment: 10% (Week 9)
  - Ontology Embeddings: 10% (Week 10)

### Learning outcomes for today

- Light understanding of how reasoning works.
- Usefulness of OWL 2 Profiles (specially OWL 2 RL)
- Learning how to do reasoning programmatically (coursework): basically 1 line of code.

# Recap

#### London is a city in England called Londres in Spanish

```
dbr:london a dbo:City .
```

dbr:london dbo:locationCountry dbr:england .

dbr:london rdfs:label "Londres"@es .

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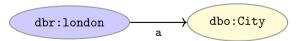
```
{\tt dbr:london\ dbo:locationCountry\ dbr:england\ .}
```

dbr:london rdfs:label "Londres"@es .

dbr:london

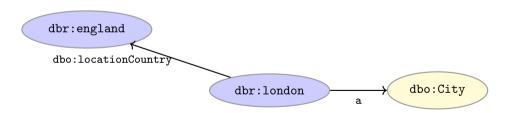
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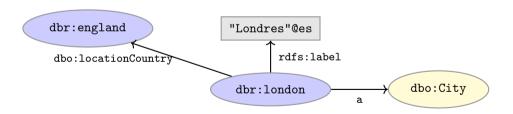


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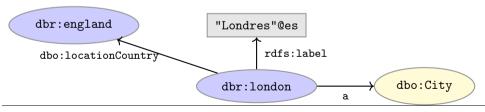
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### Recap: SPARQL Example (i)

#### **Return all Cities:**

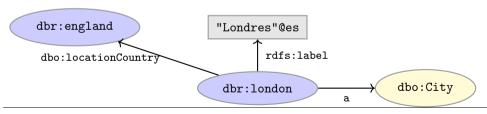
```
PREFIX dbo: <http://dbpedia.org/ontology/>
PREFIX rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#>
SELECT DISTINCT ?city WHERE {
    ?city rdf:type dbo:City .
}
```



### Recap: SPARQL Example (i)

#### Return all Cities: Query Result= {dbr:london}

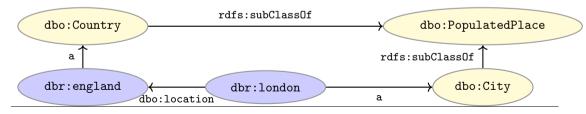
```
PREFIX dbo: <http://dbpedia.org/ontology/>
PREFIX rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#>
SELECT DISTINCT ?city WHERE {
     ?city rdf:type dbo:City .
}
```



### Recap: SPARQL Example (ii)

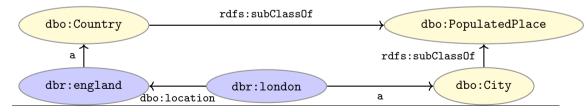
#### **Return all Populated Places:**

```
PREFIX dbo: <http://dbpedia.org/ontology/>
PREFIX rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#>
SELECT DISTINCT ?place WHERE {
     ?place rdf:type dbo:PopulatedPlace .
}
```



### Recap: SPARQL Example (ii)

#### Return all Populated Places: Query Result= {}



### **Recap: RDF Grammar for triples**

 RDF imposes a basic grammar. A triple consists of subject, predicate, and object

- URI references may occur in all positions
- Literals may only occur in object position
- Blank nodes can not occur in predicate position
- But one could still define:

```
dbr:london\ rdf:type\ "some\ string"^^xsd:string\ .
```

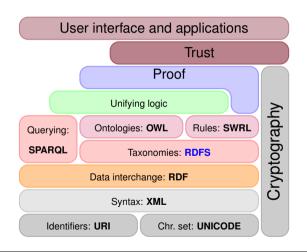
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- URI references may occur in all positions
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- But one could still define: dbr:london rdf:type "some string"^^xsd:string .
- RDF Schema (RDFS) extends the grammar for the "expected" triples (e.g., rdf:type rdf:range rdfs:Resource .), extends the vocabulary, and include a set of inference rules.

## RDF Schema (RDFS)

### **Semantic Web Technology Stack**



#### **RDF Schema**

- RDF Schema (RDFS) is a vocabulary defined by W3C.
  - https://www.w3.org/TR/rdf-schema/
  - https://www.w3.org/TR/rdf11-mt
- Namespace:

```
rdfs: http://www.w3.org/2000/01/rdf-schema#
```

- Originally designed as a "schema language" like XML Schema.
  - Not strictly doesn't describe "valid" RDF graphs.
- A very simple modeling language for RDF data → Taxonomies
- (For our purposes) it can be seen as a language less expressive than OWL, useful to understand how reasoning works.

#### **RDFS Semantics**

- RDFS is a semantic extension (adds semantics/meaning) that
  - proposes some syntactic conditions on RDF graphs,
  - comes with some (non-ambiguous) inference rules, and
  - includes some (default) triples as part of the specification.

#### **RDFS Semantics**

- RDFS is a semantic extension (adds semantics/meaning) that
  - proposes some syntactic conditions on RDF graphs,
  - comes with some (non-ambiguous) inference rules, and
  - includes some (default) triples as part of the specification.
- For example, RDFS expects as range of rdf:type a resource/IRI (e.g., rdf:type rdf:range rdfs:Resource .)
  - dbr:london rdf:type "some string"^^xsd:string .
  - RDFS: Not expected triple, but not prohibited by specification.
  - OWL: a prohibited triple will lead to an error (validation).

### **RDFS Vocabulary**

- RDFS adds the concept of "classes" which are sets of resources.

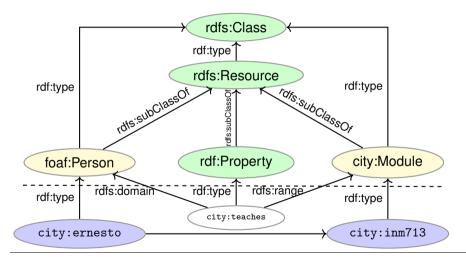
### **RDFS Vocabulary**

- RDFS adds the concept of "classes" which are sets of resources.
- Defined resources:
  - rdfs:Resource: The class of resources, everything.
  - rdfs:Class: The class of classes. (similar to owl:Class)
  - rdfs:Literal: The class of all literal values.
  - rdfs:Datatype: The class of all datatypes.

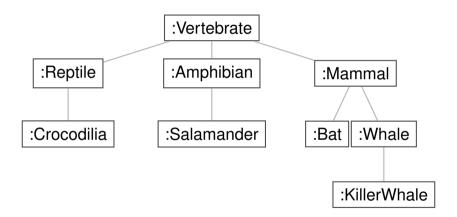
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  - rdfs:Literal: The class of all literal values.
  - rdfs:Datatype: The class of all datatypes.
- Defined properties:
  - rdfs:domain: The domain (sources) of a relation.
  - rdfs:range: The range (targets) of a relation.
  - rdfs:subClassOf: Class inclusion.
  - rdfs:subPropertyOf: Property inclusion.

### **Example RDF graph and RDF Schema**

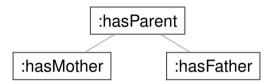


### Class Taxonomy (via rdfs:subClassOf)



(\*) As in OWL.

### Property Taxonomy (via rdfs:subPropertyOf)



(\*) As in OWL.

### **Expected RDF/RDFS resources**

### Types of resources or elements:

- Object Properties like foaf:knows
- Datatype Properties like dc:title, foaf:name
- Classes like foaf:Person
- Built-ins, a fixed set including rdf:type, rdfs:domain, etc.
- Individuals (all the rest, "usual" resources) like city:ernesto
- Datatypes like xsd:integer
- Literals like "ernesto", "39"
- (\*) Not real split of properties into object and data properties in RDFS. This comes in OWL

### **Expected RDF/RDFS triple grammar**

```
Triples
indi o-prop indi .
indi d-prop "lit" .
indi rdf:type class .
class rdfs:subClassOf class .
o-prop rdfs:subPropertvOf o-prop .
d-prop rdfs:subPropertyOf d-prop .
o-prop rdfs:domain class .
o-prop rdfs:range class .
d-prop rdfs:domain class .
d-prop rdfs:range datatype .
```

### (Default) RDFS axiomatic triples (excerpt)

- Indeed RDF and RDFS include a set of default triples to guide the above grammar of expected triples.
- Only resources have types:

```
rdf:type rdfs:domain rdfs:Resource .
```

– Types are classes:

```
rdf:type rdfs:range rdfs:Class .
```

– Ranges apply only to properties:

```
rdfs:range rdfs:domain rdf:Property .
```

RDFS Entailment Rules: https://www.w3.org/TR/rdf-mt/#rdfs\_interp

### (Default) RDFS axiomatic triples (excerpt)

Ranges are classes:

```
rdfs:range rdfs:range rdfs:Class .
```

Only properties have subproperties:

```
rdfs:subPropertyOf rdfs:domain rdf:Property .
```

– Only classes have subclasses:

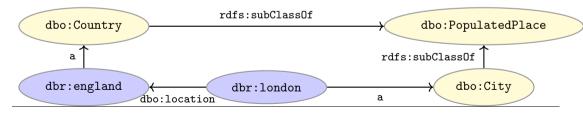
```
rdfs:subClassOf rdfs:domain rdfs:Class .
```

– ... (another 30 or so)

### Entailment via Model-Theoretic Semantics

### **SPARQL Example**

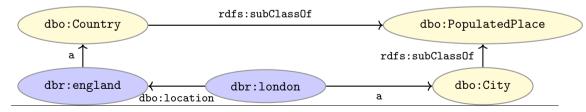
#### **Return all Populated Places:**



#### **SPARQL Example**

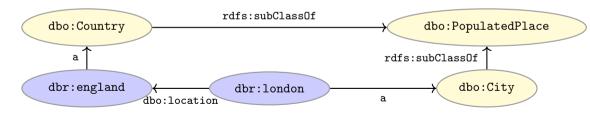
#### Return all Populated Places: Query Result= {}

```
PREFIX dbo: <http://dbpedia.org/ontology/>
PREFIX rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#>
SELECT DISTINCT ?place WHERE {
    ?place rdf:type dbo:PopulatedPlace .
}
```



#### **Entailment in RDFS**

- Given a set of triples  $\mathcal{G}$  (i.e., a Graph) can we entail a triple t ( $\mathcal{G} \models t$ )?
- Can we entail the triple: dbr:london rdf:type dbo:PopulatedPlace and add it to the graph below? (Graph expansion via reasoning).
- Similarly for dbr:england



- Interpretations might be conceived as potential "realities" or "worlds".
- Interpretations assign values to elements.
  - (The **intuitions** behind set-theory are **formally represented**.)

- Interpretations might be conceived as potential "realities" or "worlds".
- Interpretations assign values to elements.
  - (The intuitions behind set-theory are formally represented.)
- Given an interpretation  $\mathcal I$  and a set of triples  $\mathcal G$
- $-\mathcal{G}$  is valid in  $\mathcal{I}$  (written  $\mathcal{I} \models \mathcal{G}$ ), iff  $\mathcal{I} \models t$  for all  $t \in \mathcal{G}$ .
- Then  $\mathcal{I}$  is also called a **model** of  $\mathcal{G}$ .

– The following interpretation  $\mathcal{I}$  is a model of our example  $\mathcal{G}$ :

```
- dbo:City<sup>I</sup> = {dbr:london}
- dbo:Country<sup>I</sup> = {dbr:england}
- dbo:PopulatedPlace<sup>I</sup> = {dbr:london, dbr:england}
- dbo:location<sup>I</sup> = {\dbr:london, dbr:england\}
```

- The following interpretation  $\mathcal{I}$  is a model of our example  $\mathcal{G}$ :
  - $dbo:City^{\mathcal{I}} = \{dbr:london\}$
  - $dbo:Country^{\mathcal{I}} = \{dbr:england\}$
  - dbo:PopulatedPlace $^{\mathcal{I}} = \{dbr:london, dbr:england\}$
  - dbo:location $^{\mathcal{I}} = \{ \langle dbr:london, dbr:england \rangle \}$
- $-\mathcal{I} \models \mathcal{G}$  (is a model of  $\mathcal{G}$ ) as the following holds:
  - $dbo:City^{\mathcal{I}} \subseteq dbo:PopulatedPlace^{\mathcal{I}}$
  - $\mathtt{dbo}$ :  $\mathtt{Country}^{\mathcal{I}} \subseteq \mathtt{dbo}$ :  $\mathtt{PopulatedPlace}^{\mathcal{I}}$
  - $dbr:london^{\mathcal{I}} \in dbo:City^{\mathcal{I}}$

- -t = dbr:london rdf:type dbo:PopulatedPlace
- Does  $\mathcal{I} \models t$  ?

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- Does  $\mathcal{I} \models t$  ?
  - Yes:  $dbo:PopulatedPlace^{\mathcal{I}} = \{dbr:london, dbr:england\}$
- Does  $\mathcal{G}$   $\models t$  ?

- -t = dbr:london rdf:type dbo:PopulatedPlace
- Does  $\mathcal{I} \models t$  ?
  - Yes:  $dbo:PopulatedPlace^{\mathcal{I}} = \{dbr:london, dbr:england\}$
- Does  $\mathcal{G}$   $\models t$  ?
  - if and only if
    - For any interpretation  $\mathcal{I}$  with  $\mathcal{I} \models \mathcal{G}$
    - $-\mathcal{I} \models t$ .
  - (Yes, in this case too.)

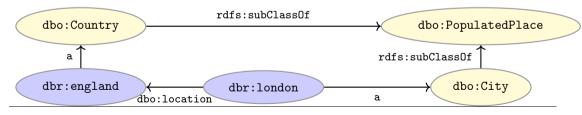
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- Does  $\mathcal{G} \models t_2$  ( $t_2$ =dbr:london rdf:type dbo:Country)?

- -t = dbr:london rdf:type dbo:PopulatedPlace
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  - (Yes, in this case too.)
- Does  $\mathcal{G} \models t_2$  ( $t_2$ =dbr:london rdf:type dbo:Country)?
  - No: Our  $\mathcal{I}$  is a counter example.  $\mathcal{I} \models \mathcal{G}$  but  $\mathcal{I} \not\models t_2$

#### **SPARQL Example: with entailment**

#### **Return all Populated places:**

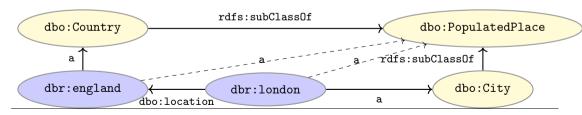
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}
```



#### **SPARQL Example: with entailment**

#### Return all Populated places: Query Result= {dbr:england, dbr:london}

```
PREFIX dbo: <http://dbpedia.org/ontology/>
PREFIX rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#>
SELECT DISTINCT ?place WHERE {
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}
```



#### **Model-Theoretic Semantics in practice**

- Model-theoretic semantics yields an unambigous notion of entailment.
- In principle, all interpretations need to be considered.
- However there are infinitely many such interpretations,
- An algorithm should terminate in finite time.

Foundations of Semantic Web Technologies. Chapter 3.

# Entailment via Inference Rules

# **Syntactic Reasoning**

- From the computation point of view, we need means to decide entailment syntactically.
- Syntactic methods operate
  - only on the form of a statement, that is on its concrete grammatical structure (i.e., triples),
  - without recurring to interpretations.
- Syntactic methods should justify that their so-called operational semantics are expected with respect to model-theoretic semantics.

# Inference rules (i)

- Inference rules (also known as deduction rules or derivation rules) is an option to describe syntactic solutions.
- The general form of an inference rule is:

$$\frac{P_1,\ldots,P_n}{P}$$

- the  $P_i$  are **premises** (body)
- and P is the **conclusion** (head).
- An inference rule may have,
  - any number of premises (typically one or two),
  - but only one conclusion.

#### Inference rules (ii)

- Recall that syllogisms (i.e., inference) can be traced back to Aristotle
- Example:

All human are mortal
Socrates is a human
Therefore. Socrates is mortal

# Inference rules (iii)

- The whole set of inference rules given for a logic is called **deduction** calculus.
- ⊢ is the inference relation, while ⊨ was the entailment relation using model theoretic semantics.
  - − We write  $\Gamma \vdash P$  if we can deduce P from the premises  $\Gamma$ .

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- The whole set of inference rules given for a logic is called **deduction** calculus.
- ⊢ is the inference relation, while ⊨ was the entailment relation using model theoretic semantics.
  - − We write  $\Gamma \vdash P$  if we can deduce P from the premises  $\Gamma$ .
- In our setting:
  - the **premises**  $\Gamma$  are a **set of triples** (*i.e.*, a (sub)graph  $\mathcal{G}$ ),
  - the conclusion P is a new triple t
  - After applying the rules to  $\mathcal{G}$  we will get an **expanded graph**  $\mathcal{G}'$

#### **RDFS Inference Rules**

RDFS supports several rules. Organized into three groups:

#### 1. Type propagation:

- "London is a City, all Cities are populated places, so. . . "

#### 2. Property propagation:

- "London is the capital of England, anything that is capital of a country is also located in that country, so..."

#### 3. Domain and range propagation:

- "Everything that has a capital is a country, so England is a..."
- "Everything that is a capital is a city, so London is a..."

### Type propagation

- Members of superclasses:

(\*) rdfs9, rdfs10, rdfs11 are the names of the inference rules in the W3C standard. A, B are classes; x is an instance.

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#### Type propagation: Examples

– Members of superclasses:

```
:City rdfs:subClassOf :PopulatedPlace . :london rdf:type :City . :london rdf:type :PopulatedPlace . rdfs9
```

Reflexivity of sub-class relation:

```
:City rdf:type rdfs:Class .
:City rdfs:subClassOf :City .
```

Transitivity of sub-class relation:

```
:City rdfs:subClassOf :PopulatedPlace . :PopulatedPlace rdfs:subClassOf :Place . :City rdfs:subClassOf :Place . rdfs11
```

#### **Property Propagation**

- Transitivity:

```
P rdfs:subPropertyOf Q . Q rdfs:subPropertyOf R .

P rdfs:subPropertyOf R . rdfs5
```

(\*) P, Q are properties; u, v are instances.

### **Property Propagation**

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#### **Property Propagation: Examples**

– Transitivity:

```
:has_writer rdfs:subPropertyOf :has_author . :has_author rdfs:subPropertyOf :has_creator . rdfs:subPropertyOf :has_creator .
```

- Reflexivity:

```
:has_writer rdf:type rdf:Property .
:has_writer rdfs:subPropertyOf :has_writer .
```

Property transfer:

```
:has_author rdfs:subPropertyOf :has_creator . :Hamlet :has_author :Shakespeare . rdfs
```

# Domain and range propagation

Typing triggered by the use of properties.

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# Domain and range propagation

Typing triggered by the use of properties.

– Domain propagation:

- Range propagation:

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#### **Domain and Range Propagation: Examples**

– Domain propagation:

```
:capitalOf rdfs:domain :City . :london :capitalOf :england . :clondon rdf:type :City . rdfs2
```

– Range propagation:

# Properties of RDFS Semantics

#### **Entailment and Inference**

- Both have the monotonic property.
  - If a graph  $\mathcal{G} \models t$  (or  $\mathcal{G} \vdash t$ ),
  - then adding more triples  $(e.g., t_1)$  does not alter the entailment  $\mathcal{G} \cup \{t_1\} \models t$  (or derivation  $\mathcal{G} \cup \{t_1\} \vdash t$ )

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- The set of RDFS rules we have seen are **sound**.
  - If  $\mathcal{G} \vdash t$  then  $\mathcal{G} \models t$
- But not complete.
  - Not always applies that If  $\mathcal{G} \models t$  then  $\mathcal{G} \vdash t$
  - Due to how RDFS deals with datatypes.

# (Non) Validation in RDFS (i)

- RDFS was conceived of as a schema language for RDF
- However, the statements in an RDFS graph never trigger inconsistencies.
- Reasoning will not lead to a "contradiction", "error", "non-valid document"
- Inference rules in RDFS add more triples, but do not detect errors.

### (Non) Validation in RDFS (ii)

- RDFS has no notion of negation
  - For instance, the two triples

```
city:ernesto rdf:type ex:Smoker .
city:ernesto rdf:type ex:NonSmoker .
are not inconsistent.
```

### (Non) Validation in RDFS (ii)

- RDFS has no notion of negation
  - For instance, the two triples

```
city:ernesto rdf:type ex:Smoker .
city:ernesto rdf:type ex:NonSmoker .
are not inconsistent.
```

There is also not clear notion of disjointness among RDF resources:
 Object Properties, Datatype Properties, Classes, Built-in properties,
 Individuals, Datatypes and Literals.

### (Non) Validation in RDFS (ii)

- RDFS has no notion of negation
  - For instance, the two triples

```
city:ernesto rdf:type ex:Smoker .
city:ernesto rdf:type ex:NonSmoker .
are not inconsistent.
```

- are not inconsistent.
- There is also not clear notion of disjointness among RDF resources:
   Object Properties, Datatype Properties, Classes, Built-in properties,
   Individuals, Datatypes and Literals.
- OWL (and OWL 2) includes additional vocabulary and also consistency-checks.

# OWL 2 Reasoning and Profiles

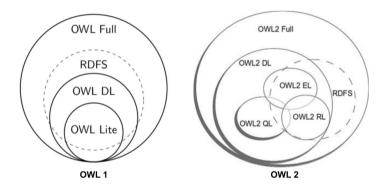
# Recap: OWL (The Web Ontology Language)

- A W3C recommendation:
  - OWL 1 (2004): http://www.w3.org/TR/owl-ref/
  - OWL 2 (2009): https://www.w3.org/TR/owl2-overview/

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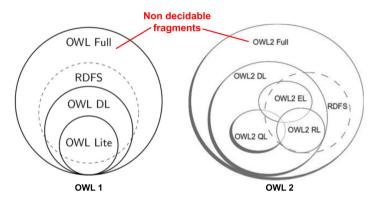
- A W3C recommendation:
  - OWL 1 (2004): http://www.w3.org/TR/owl-ref/
  - OWL 2 (2009): https://www.w3.org/TR/owl2-overview/
- OWL semantics based on **Description Logics (DL)**.
  - Family of knowledge representation languages
  - Decidable subset of First Order logic (FOL)
  - Original called: Terminological language or concept language

### Recap: OWL 1, OWL 2 (profiles) and RDFS



Olivier Cure and Guillaume Blin. RDF Database Systems (Chapter 3). 2015. Elsevier.

### Recap: OWL 1, OWL 2 (profiles) and RDFS



† **Reasoning in OWL 2** will partially get the same consequences as in the RDFS inference rules and many more.

### **Automated Reasoning in OWL**

- Formal semantics allows the automatic deduction of **new facts** (i.e., not explicit but derived from others).
- Also allows us to perform checks that aim to detect the correctness of the designed model (e.g., is :dolphin a :Fish?). New respect to RDFS.

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  - :dolphin cannot be an individual (or a subclass) of both :Mammal and :Fish.

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- In the form of logical errors:
  - : Mammal and : Fish are disjoint classes.
  - :dolphin cannot be an individual (or a subclass) of both :Mammal and :Fish.
- Extremely valuable for designing correct ontologies/KGs, specially when working collaboratively and integrating various sources.

# OWL 2 Reasoning

### **Recap: OWL 2 Axioms into Boxes**

- Traditionally OWL 2 axioms are put in boxes.
- The **TBox** (terminological knowledge)
  - Typically independent of any actual instance data.
  - Property axioms are also referred to as RBox
- The ABox (assertional knowledge)
  - Contains facts about concrete instance.

### (Standard) Reasoning tasks that use only the TBox $\mathcal{T}^{\dagger}$

- Concept **unsatisfiability**: Given C, does  $\mathcal{T} \models C \sqsubseteq \bot$ ? (i.e.,  $\mathcal{C}^{\mathcal{I}} = \emptyset$ )

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- Concept **equivalence**: Given C and D, does  $T \models C \equiv D$ ? (i.e.,  $C^{\mathcal{I}} = D^{\mathcal{I}}$ )
- Concept **disjointness**: Given C and D, does  $\mathcal{T} \models C \sqcap D \sqsubseteq \bot$ ? (i.e.,  $\mathcal{C}^{\mathcal{I}} \cap \mathcal{D}^{\mathcal{I}} \subseteq \emptyset$ )

#### (Standard) Reasoning tasks that involve both the TBox ${\mathcal T}$ and Abox ${\mathcal A}$

- Consistency:

Is there a model for  $(\mathcal{T}, \mathcal{A})$ ? i.e., is there an interpretation  $\mathcal{I}$  such that  $\mathcal{I} \models (\mathcal{T}, \mathcal{A})$ ?

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- **Retrieval**: Given C, find all a such that  $(\mathcal{T}, \mathcal{A}) \models C(a)$ .
- Conjunctive Query Answering (SPARQL).
- † (Model-Theoretic Semantics) The answer to 'does  $(\mathcal{T}, \mathcal{A}) \models \alpha$ ?' will be positive if for each interpretation  $\mathcal{I}$  such that  $\mathcal{I} \models (\mathcal{T}, \mathcal{A}), \mathcal{I} \models \alpha$  too.

### **OWL 2 Reasoning Algorithms**

- Reasoning in OWL 2 is typically based on (Hyper)Tableau Reasoning Algorithms (tableau = truth tree)
- Reasoning tasks reduced to (un)satisfiability.
- Algorithm tries to construct an abstraction of a model.

Chapter 5: Foundations of Semantic Web Technologies. CRC Press 2009
Seminars by Prof. lan Horrocks: http://www.cs.ox.ac.uk/people/ian.horrocks/Seminars/seminars.html

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- Reasoning tasks reduced to (un)satisfiability.
- Algorithm tries to construct an abstraction of a model.
- State-of-the-art algorithms:
  - e.g., HermiT (default option in Protégé).
  - Implement a number of (search) optimisations.
  - Effective with many realistic ontologies

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### **Tractability Problems with OWL 2 Reasoning**

- Problems with very large and/or cyclical ontologies.
  - Ontologies may define hundred of thousands of terms (e.g., SNOMED CT)
  - Large number of tests for classification (each test can lead to the construction of very large models).

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- Problems with very large and/or cyclical ontologies.
  - Ontologies may define hundred of thousands of terms (e.g., SNOMED CT)
  - Large number of tests for classification (each test can lead to the construction of very large models).
- Problems with medium/large data sets (ABoxes)
  - OWL 2 Reasoners typically optimized for TBox reasoning tasks.
  - Data also brings additional complexity.

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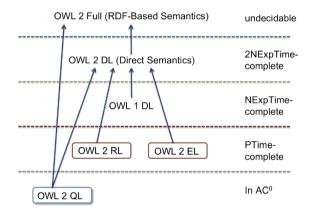
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- These profiles have very interesting computational properties.
  - OWL 2 QL:
    - Specifically designed for efficient database integration.
  - OWL 2 EL:
    - A lightweight language with polynomial time reasoning.
  - OWL 2 RL:
    - Designed for compatibility with rule-based inference tools.
    - Efficient reasoning with large datasets.

### **Data Complexity OWL 2 Profiles**



https://www.w3.org/TR/owl2-profiles/

# **OWL EL profile (i)**

### Based on the DL $\mathcal{EL}^{++}$ . Concept descriptions (simplified)

#### **Axioms**

- $C \sqsubseteq D$  and  $C \equiv D$  for concept descriptions D and C.
- $-P \sqsubseteq Q$  and  $P \equiv Q$  for roles P, Q. Also Domain and Range.
- -C(a) and R(a,b) for concept C, role R and individuals a,b.

### **OWL EL Profile (ii)**

- Standard reasoning tasks in P time
- Very good for large ontologies.
- ✓ Used in many biomedical ontologies (e.g., SNOMED CT).

### Not supported features, simplified:

- $\nearrow$  negation (but  $C \sqcap D \sqsubseteq \bot$  possible)
- disjunction
- universal quantification and cardinalities
- inverse roles and some role characteristics
- reduced list of datatypes

- Reasoning can be performed via saturation<sup>†</sup> (i.e., inference rules).
- For example:

$$\begin{array}{c|c} A \sqsubseteq B & B \sqsubseteq C \\ \hline A \sqsubseteq C \\ \hline A \sqsubseteq \exists R.B & \exists S.B \sqsubseteq C & S \sqsubseteq R \\ \hline A \sqsubseteq C \\ \end{array}$$

† Using a saturation-based approach over an OWL 2 ontology is not possible.

ELK reasoner (also available as Protégé plugin): https://github.com/liveontologies/elk-reasoner/wiki

#### Based on DL-Lite<sub>R</sub>. Concept descriptions (simplified)

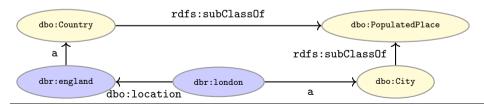
#### **Axioms**

- $C \sqsubseteq D$  for concept descriptions D and C (and  $C \equiv C'$ ).
- $-P \sqsubseteq Q$  and  $P \equiv Q$  for roles P,Q. Also Domain and Range.
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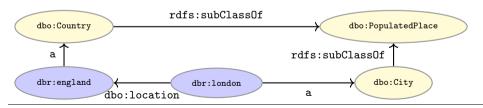
- ✓ Required language so that queries can be rewritten using the TBox.
- ✓ Used in Ontology Based Data Access (OBDA) where SPARQL queries are translated to SQL

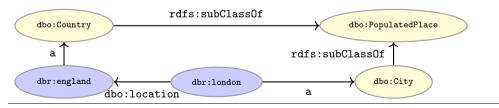
#### Not supported, simplified:

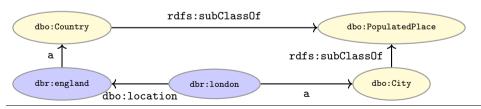
- disjunction
- universal quantification, cardinalities, and functional roles
- X = (SameIndividual)
- enumerations (closed classes)
- subproperties of chains, transitivity
- reduced list of datatypes



```
Q: SELECT DISTINCT ?place WHERE {?place rdf:type dbo:PopulatedPlace . }
```







# OWL 2 RL Profile (i)

#### Based on Description Logic Programs (DLP). Concept descriptions:

#### **Axioms**

- $-C \sqsubseteq D, C \equiv C', \top \sqsubseteq \forall R.D, \top \sqsubseteq \forall R^-.D \ R \sqsubseteq P, R \equiv P^- \text{ and } R \equiv P \text{ for roles } R, P \text{ and concept descriptions } C, C' \text{ and } D. \text{ Also Domain and Range.}$
- -C(a) and R(a,b) for concept C, role R and individuals a,b.

#### OWL 2 RL Profile (ii)

- Puts syntactic constraints in the way in which constructs are used (i.e., syntactic subset of OWL 2).
- Imposes a reduced list of allowed datatypes
- ✓ OWL 2 RL axioms can be directly translated into datalog rules
- Enables desirable computational properties using rule-based reasoning engines.

Reasoning via full materialisation of the graph, similarly to RDFS inference rules. e.g.,:

See W3C specification for further inference rules in OWL 2 RL.

# Practical Examples: OWL Reasoning

# **Necessary conditions and primitive classes**

Hawaiian pizza **implies** having pineapple as ingredient (among others); but not the other way round.



#### Sufficient conditions and defined classes

Meat pizza implies having meat as ingredient (and being pizza).

A pizza with meat as ingredient **implies** being a meat pizza.

Hawaiian pizza has ham as ingredient and thus is a meat pizza.



# **Detecting modelling errors**

Ice cream **implies** having fruit as topping Ice cream is **disjoint with** Pizza

The **domain** of has topping is pizza, that is, having any topping **implies** being a pizza (domain is a type of sufficient condition).



#### Detecting modelling errors: part-of VS subclass-of

:City rdfs:subClassOf :Country . ?

Members of superclasses (rdfs9 rule):

```
:City rdfs:subClassOf :Country . :london rdf:type :City . rdfsC
```

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```

:City rdfs:subClassOf :locatedIn some :Country

#### **Detecting modelling errors: property hierarchy**

:isLocatedInCity rdfs:subPropertyOf :isLocatedInCountry . ?

Property transfer (rdfs7 rule):

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```

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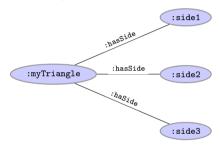
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```

:isLocatedInCity rdfs:subPropertyOf :isLocatedIn . :isLocatedInCountry rdfs:subPropertyOf :isLocatedIn .

#### **OWL 2 and Open World Assumption**

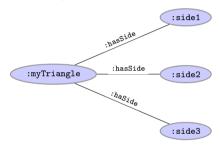
- :Triangle EquivalentTo :hasSide exactly 3 :Side



- is :myTriangle a :Triangle?

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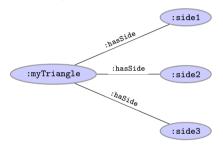
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#### **OWL 2 and Open World Assumption**

- :Triangle EquivalentTo :hasSide exactly 3 :Side



- is :myTriangle a :Triangle? I don't know because of OWA and NUNA.
- Solution: reasoning in OWL can be complemented with SPARQL queries (in this case with aggregates) → SPARQL 1.1

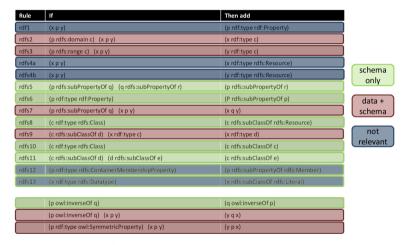
Laboratory: RDFS Semantics and OWL 2 RL

#### **Tasks**

- Manually checking inferences using RDFS and OWL 2 RL semantics.
- Extracting inferences and expanding the graph/model programmatically and checking via SPARQL.
- Python: We are using the OWL-RL library owlrl.DeductiveClosure(owlrl.RDFS\_Semantics).expand(g) owlrl.DeductiveClosure(owlrl.OWLRL\_Semantics).expand(g)
- Java: Jena API

```
Reasoner reasoner = ReasonerRegistry.getRDFSReasoner();
Reasoner reasoner = ReasonerRegistry.getOWLMiniReasoner();
InfModel inf_model = ModelFactory.createInfModel(reasoner, model);
```

#### RDFS inference rules (cheatsheet)



(\*) Figure adapted from "Towards Efficient Schema-Enhanced Pattern Matching over RDF Data Streams". ISWC'11 slides.