



**Heiko Paulheim** 

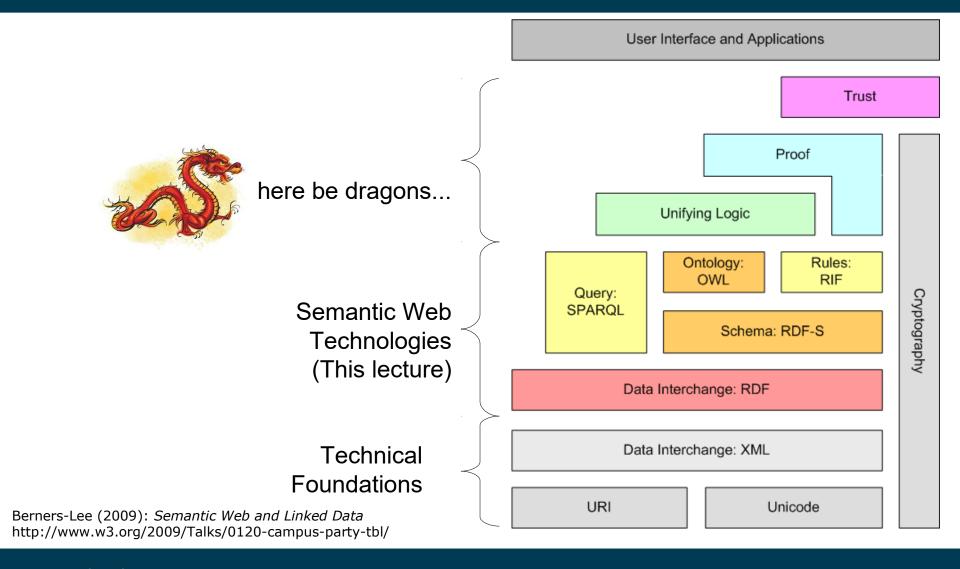
# **Previously on Semantic Web Technologies**

 What we would like to have: daughterOf(X,Y) ← childOf(X,Y) ∧Woman(X).

- Rules are flexible
- There are rules in the Semantic Web, e.g.
  - Semantic Web Rule Language (SWRL)
  - Rule Interchange Format (RIF)
  - Some more
- Some reasoners do (partly) support rules

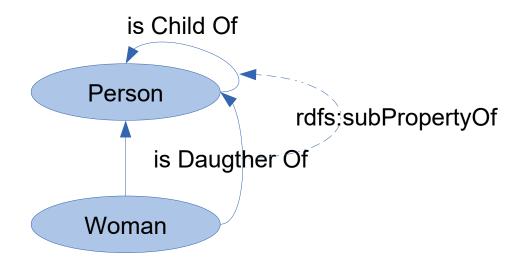


#### **Semantic Web – Architecture**



#### **Towards Rules for the Semantic Web**

- What we would like to have:
  - daughterOf(X,Y) ← childOf(X,Y)  $\land$ Woman(X).
- OWL only gives an approximation:



#### **SWRL**

- Semantic Web Rule Language
  - a rule language for the Semantic Web
  - built to be combined with OWL
- W3C Member Submission (2004)
  - not a standard in a strict sense
  - but widely adopted
- Tool support
  - many reasoners
  - Protégé



# **SWRL Building Blocks**

- Classes are defined as unary predicates
  - :Peter a :Person . ↔ Person(Peter)
- Properties are defined as binary predicates
  - :Peter :hasMother :Julia . ↔ hasMother(Peter,Mary)
  - :Peter :hatAge 24^^xsd:integer . ↔ hasAge(Peter,24)

#### **SWRL Rules**

- Basic form:
  - Head (Consequence) ← Body (Condition)
- Body and head are conjunctions of predicates
- Variables are introduced by ?
- Example:
  - daughterOf(?X,?Y) ← childOf(?X,?Y) ∧ Woman(?X)
- There is no
  - disjunction (logical or)
  - negation
  - undbound variables in the rule head
- ...but there are some ways out

## **Disjunctions in Rule Body**

- There is no disjunction
- Example for disjunction in rule body:
  - Female faculty members are students or staff of the faculty
- Intuitive:
  - FemaleFacultyMember(?X) ← Woman(?X) ∧ Faculty(?Y) ∧
     (worksAt(?X,?Y) ∨ studentAt(?X,?Y))

# **Disjunctions in Rule Body**

- Solution
  - first step: convert body to disjunctive normal form
    - i.e., disjunction of conjunctions
  - second step: split into individual rules

## **Disjunctions in Rule Body**

- FemaleFacultyMember(?X) ← Woman(?X) ∧ Faculty(?Y) ∧
   (worksAt(?X,?Y) ∨ studentAt(?X,?Y))
- turns into
  - FemaleFacultyMember(?X) ←
     (Woman(?X) ∧ Faculty(?Y) ∧ worksAt (?X,?Y))
     ∨ (Woman(?X) ∧ Faculty(?Y) ∧ worksAt (?X,?Y))
- ...which turns into
  - FemaleFacultyMember(?X) ←
     Woman(?X) ∧Faculty(?Y) ∧worksAt (?X,?Y)
  - FemaleFacultyMember(?X) ←
     Woman(?X) ∧Faculty(?Y) ∧studentAt (?X,?Y)

## Disjunctions in Rule Head

- Disjunctions in rule head
  - are not so easy to get rid off
- Example
  - Every faculty member is a student or an employee
     Student(?X) ∨ Employee(?X) ← FacultyMember(?X)
- On the other hand: what should a reasoner conclude?
  - → disjunction in rule head does not make as much sense!

## Disjunctions in Rule Head

- SWRL is meant to be used together with OWL
- Idea: build an artificial class for the rule head StudentOrEmployee owl:unionOf (Student Employee)
   StudentOrEmployee(?X) ← FacultyMember(?X)
- This way, we can conclude that ?X is in the union of both classes
  - Further reasoning on other axioms might rule out one option

#### **Negation**

- Negation can be simulated with a similar trick
- Example:
  - Creatures living in the water are not human.
- Intuitive:
  - $\neg$ Human(?X)  $\leftarrow$  Creature(?X)  $\land$  habitat(?X,Water)

## **Simulating Negation**

- Again: combining SWRL and OWL
  - NonHuman owl:complementOf Human .
- New Rule:
  - NonHuman(?X) ← Creature(?X) ∧ habitat(?X,Water)
- Now, a reasoner can find a contradiction between
  - :Nemo a :Creature; habitat :Water .
- and
  - :Nemo a :Human .

## **Simulating Negation**

Negation in the rule body:

```
FlightlessBird(?X) \leftarrow Bird(?X) \land \neghabitat(?X,Air)
```

Define class:

```
NotAirHabitat owl:equivalentClass [
    a owl:Restriction;
    owl:onProperty :habitat;
    owl:allValuesFrom [
        owl:complementOf [ owl:oneOf (:Air) ] ] ]
```

FlightlessBird(?X) ← Bird(?X) ∧ NotAirHabitat(?X)

#### **Unbound Variables**

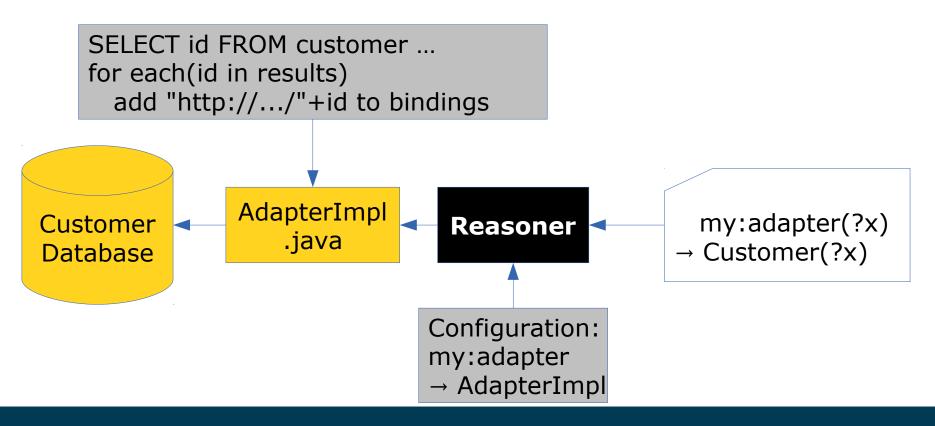
- All variables appearing in the rule head must also appear in the body
  - those are bound variables
- Example: every human has a (human) father
  - Human(?Y) ∧ hasFather(?X,?Y) ← Human(?X)
- In that case, the reasoner would have to create new instances for Y
  - Possible issue: termination
  - No easy solution in SWRL+OWL

#### **SWRL Extensions and Built-Ins**

- Comparison
  - olderThan(?X,?Y) ← hasBirthdate(?X,?BX) ∧ hasBirthdate(?Y,?BY) ∧ swrlb:lessThan(?BX,?BY)
- Arithmetics
  - twiceAsOld(?X,?Y) ← hasAge(?X,?AX) ∧ hasAge(?Y,?AY) ∧ swrlb:multiply(?AX,?AY,2)
- String operations
  - PeopleWithS(?X) ← hasName(?X,?N) ∧ swrlb:startsWith(?N,"S")

#### **SWRL Extensions and Built-Ins**

- Some reasoners also allow for custom built-ins
- E.g., for wiring a reasoner to external systems



#### **SWRL Extensions and Built-Ins**

- More use cases for custom built-ins
- Live data
  - Weather
  - Stock exchange
  - Product availability
- Complex computations
  - Trip duration from A to B (e.g., Google Maps API)
  - Simulations and predictions

**–** ...

# Monotonic Reasoning with SWRL

- Recap: monotonous vs. non-monotonous reasoning
  - monotonous: every consequence derived is true forever
  - non-monotonous: consequences may be revoked
- SWRL is monotonous
  - i.e., consequences of all rules add up
  - allows for efficient reasoning
  - may lead to contradictions

- Termination guarantee of reasoning
- So far
  - no new instances, classes, and properties are generated
- This constrains the set of consequences which can be derived:
  - C\*I type assertions
  - I\*O\*I object property assertions
  - I\*D\*L datatype property assertions
  - → in monotonous reasoning, the reasoner eventually terminates

Consider this example:

```
:Person rdfs:subClassOf [
   a owl:Restriction ;
   owl:onProperty :hasFather ;
   owl:cardinality 1^^xsd:integer ] .
:hasFather rdfs:range :Person .
:Grandchild rdfs:subClassOf :Person .

hasFather(?x,?y) ∧ hasFather(?y,?z) → Grandchild(?x)
```

Given

:Peter a :Person .

Do we derive GrandChild(Peter)?

- Possible solution:
  - We know that each person has a father
  - therefore:

```
:Peter :hasFather _:p0 . _:p0 :hasFather _:p1 . :p1 ...
```

and thus

```
:Peter a :Grandchild .
```

- What is the price of that solution?
  - We allow for the creation of new instances
  - i.e., we sacrifice guaranteed termination

- DL safe rules:
  - Variables are only bound to existing instances
  - No new instances are created
- Thus, we cannot derive

```
:Peter a :Grandchild .
```

- Once more: trading off
  - expressivity
  - decidability

#### **Production Rules**

- Sometimes, monotonous rules are not desirable
  - consider: if a student passes SWT, his/her credit increases by 6 ECTS
- A first attempt with SWRL + built-ins:

```
Student(?X) ∧hasPassed(?X,:SWT)
```

- ∧hasCredits(?X,?C) ∧swrlb:add(?NC,?C,6)
- $\rightarrow$  hasCredits(?X,?NC).

#### **Production Rules**

Consider:

```
:Peter a :Student .
:Peter :hasCredits 26^^xsd:integer .
:Peter :hasPassed :SWT .
```

After applying the rule:

```
:Peter :hasCredits 32^^xsd:integer .
```

• But rules are monotonous, so the following holds as well:

```
:Peter :hasCredits 26^^xsd:integer .
```

...and the reasoner is done yet

#### **Production Rules**

What happens:

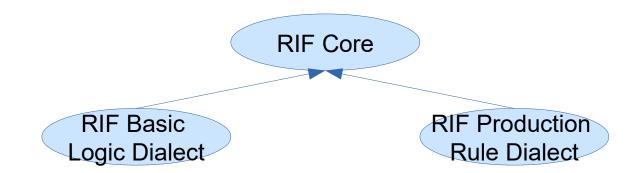
```
:Peter :hasCredits 26^^xsd:integer .
:Peter :hasCredits 32^^xsd:integer .
:Peter :hasCredits 38^^xsd:integer .
:Peter :hasCredits 44^^xsd:integer .
:Peter :hasCredits 50^^xsd:integer .
```

- We need to
  - revoke/overwrite statements
    - in contrast to monotonous reasoning!
  - define new criteria for termination

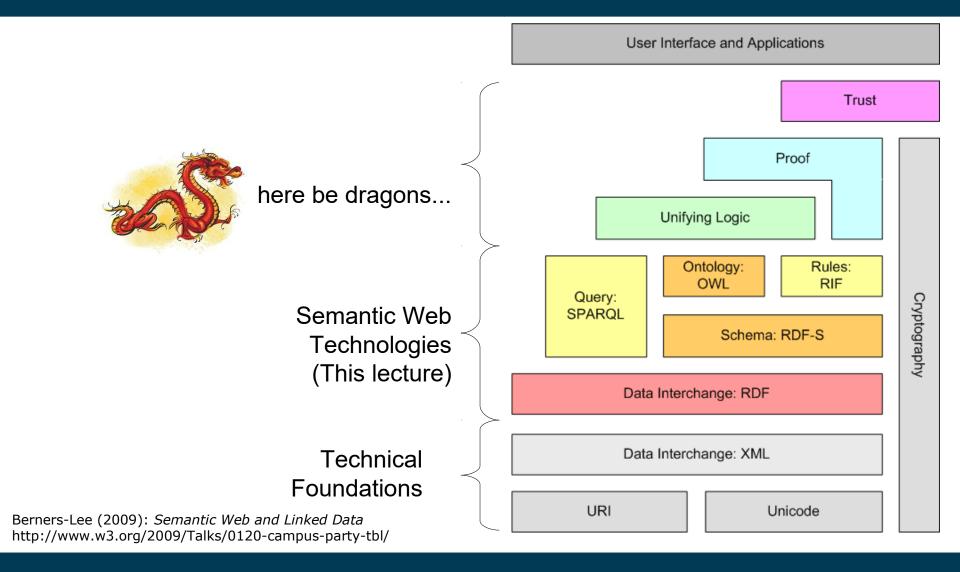
# Rule Interchange Format

- Rule Interchange Format (RIF)
- Unification of
  - Basic Logic Rules (such as SWRL)
  - Production Rules (e.g., JENA rules)
- Standardized by W3C in 2010





#### **Semantic Web – Architecture**



## Other Semantic Web Languages

What else is out there?

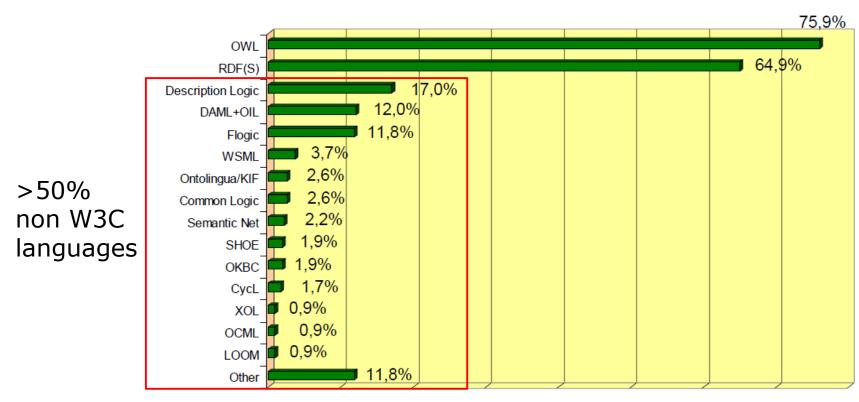


Figure 5. Ontology languages currently used by users.

Cardoso (2006): The Semantic Web Vision - Where are We?

## Other Semantic Web Languages

- There is a wild mix
  - of old and new languages
  - of different paradigms
  - of sophisticated languages and pure, low-level logic
- We will look at one example of a radically different language

# F-Logic

- Main concept: frames
  - collection of properties of a class
  - similar to class and database models

Person	Mother (Person)	Father (Person)	Age (int)
:Paul	:Martha	:Hans	24
:Martha	:Johanna	:Karl	47

## F-Logic: A First Glance

- First observation:
  - relations are bound to class
  - in RDFS/OWL: first class citizens
- Inheritance
  - Relations are inherited to subclasses.
  - Domain and range cannot be restricted any further
- Semantics
  - Closed world semantics
  - Negation can be used

#### F-Logic: Rules

- Almost everything is expressed in rules
- e.g., property chains:

- Datalog-like syntax
- :- is used for implication ←
- Variables are denoted with ?

## F-Logic: Quantifiers

- There are extensional and universal quantifiers
- Authors are persons who have written at least one book

```
?X:Author :- ?X:Person
AND (EXIST ?Y ?Y:Book and ?X[hasWritten->?Y]).
```

A non-author is a person who has not written any book

```
?X:NonAuthor :- ?X:Person
AND NOT(EXIST ?Y ?Y:Book and ?X[hasWritten ->?Y]).
```

A star author is an author who as only written bestsellers

#### F-Logic: Negation

- Negation may have unwanted consequences
- Consider this example:

```
• ?X[hates->?Y] :-
   not(?X[likes->?Y] or ?X[doesntCare->?Y])) .
?X[likes->?Y] :- ?X[knows->?Y] and not(?X[hates->?Y]) .
```

- Assume, the reasoner wants to prove ?X[likes->Stefan] .
- Possible plan:

- F-Logic ontologies with negations can be undecidable
- Underyling problem:
  - Cycles of rules containing negations
- Simplest case

```
- p(X) :- not(p(X)).
```

- Test: Stratification
  - lat. Stratum (pl.: Strata): Layer
- Divide ontology into layers
- Each predicate is assigned to a layer
  - Classes are treated as unary predicates

- Assign a layer S(p) to each predicate p
- Two conditions must be fulfilled:
  - for all rules which have p in their head and a non-negated predicate q in the body:
     S(q)≤S(p)
  - for all rules which have p in their head and a negated predicate q in their body S(q)<S(p)</li>
- If such an assignment can be found, the ontology is decidable

- Simple case:
- ?X[hates->?Y] :- not(?X[likes->?Y]) .
  ?X[knows->?Y] :- ?X[likes->?Y] .
- We have to ensure
  - S(likes) < S(hates)</p>
  - S(likes) ≤ S(knows)
- For those two rules, we can assign
  - S(likes) = 0
  - S(hates) = 1
  - S(knows) = 0

We obtain the following layers

```
?X[hates->?Y] :- not(?X[likes->?Y]) . Layer 1

?X[knows->?Y] :- ?X[likes->?Y] . Layer 0
```

- Trivial observation
  - For ontologies without negation, one layer is enough!

Back to the original example

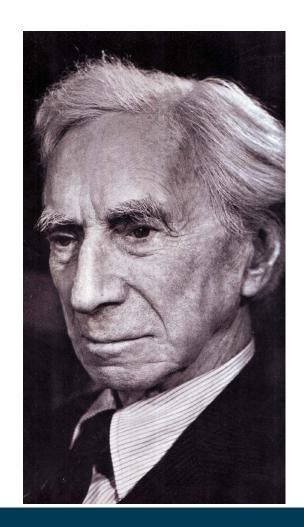
```
?X[hates->?Y] :-
  not(?X[likes->?Y] or ?X[doesntCare->?Y])) .
?X[likes->?Y] :- ?X[knows->?Y] and not(?X[hates->?Y])
.
```

- Can we find a stratification?
- We would need
  - S(likes) < S(hates)</p>
  - S(hates) < S(likes)</p>
- This is not possible!
  - → The ontology cannot be stratified, i.e., it is undecidable!

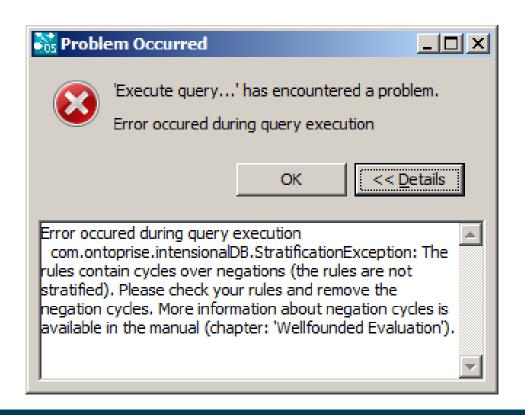
# Recap: Russell's Paradox

- A classic paradox by Bertrand Russell, 1918
- In a city, there is exactly one barber who shaves everybody who does not shave themselves.

Who shaves the barber?



- Russell's paradox in F-Logic: theBarber[shaves->?X] :- not(?X[shaves->?X]) .
- We would need
   S(shaves) < S(shaves)</li>



# Validating Datasets with RDF Shapes

- Ontology reasoning is good for semantic validation
  - but sometimes problematic due to semantic properties
  - i.e., closed world assumption, non unique name assumption
- To validate data quality
  - we want to ensure certain data is there
    - e.g., every person has a name
  - we want to ensure that data is not duplicated
    - e.g., every person has exactly one birth place

etc.

### Validating Datasets with RDF Shapes

Example dataset:

```
:Mary a :Person .
:Mary :birthPlace :Mannheim .
:Mary :birthPlace :Berlin .
```

Constraints in OWL:

```
Person rdfs:subClassOf [
   a owl:Restriction .
   owl:onProperty :name .
   owl:minCardinality 1 . ]

Person rdfs:subClassOf [
   a owl:Restriction .
   owl:onProperty :birthPlace .
   owl:maxCardinality 1 . ]
```

# **Shapes Constraint Language (SHACL)**

- A W3C Standard since 2017
- For RDF validation
- Differences to reasoning
  - Closed world evaluation
  - Counting is possible
  - More fine-grained checks (see later)



# **Shapes Constraint Language (SHACL)**

Example dataset: :Mary a :Person . :Mary :birthPlace :Mannheim . :Mary :birthPlace :Berlin . Constraints in SHACL: :PersonShape a sh:NodeShape; sh:targetClass:Person; sh:property [ sh:path:name; sh:minCount 1; sh:datatype xsd:string ]; sh:property [ sh:path:birthPlace; sh:maxCount 1:

sh:class:City].

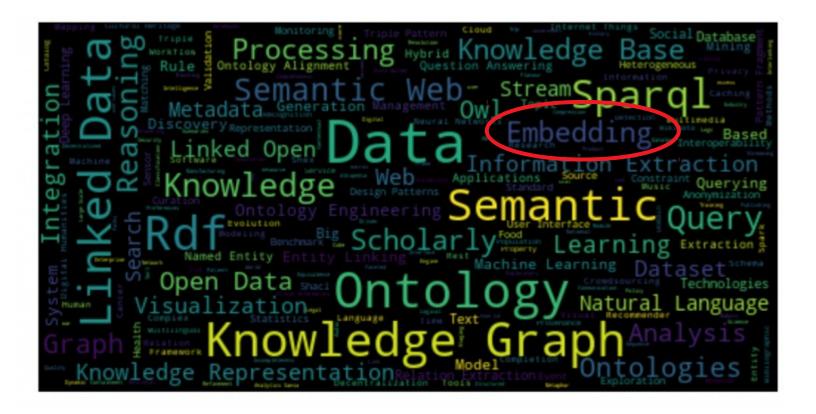
### **Shapes Constraint Languages**

- Further possibilities
  - Dependencies between attributes
    - e.g., given name and first name are equivalent
  - Complex expressions involving paths and even SPARQL queries
  - Checking strings against regex patterns (e.g., phone numbers)

**–** ...

### RDF Embeddings

- One of the current hot topics in Semantic Web research:
  - Embeddings

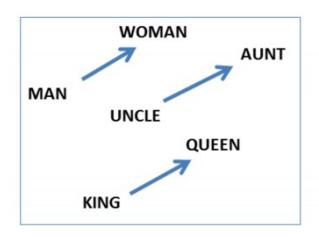


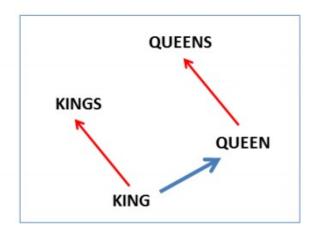
# RDF Embeddings

- Challenge in RDF/OWL etc.:
  - How similar are two entities?
  - e.g., is Mannheim more similar to Karlsruhe than to Heidelberg?
- Application scenarios:
  - Recommender systems
  - Information retrieval

### **Excursion: word2vec**

- Such approaches exist for words
  - aka, word embeddings
  - each word becomes a vector in a low-dimensional vector space
  - similar words are close in that vector space
  - semantic relations have a similar direction and length
    - allows for arithmetics, e.g., King Man + Woman = Queen

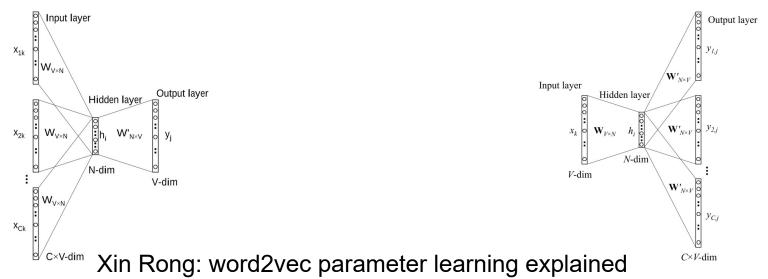




(Mikolov et al., NAACL HLT, 2013)

### word2vec

- General idea: similar words appear in similar contexts
- Training set: sequences from a text corpus
- Training method: neural network
- Training variants:
  - Continuous bag of words (CBOW): predict a word from its context
  - Skip-Gram: predict context from a word

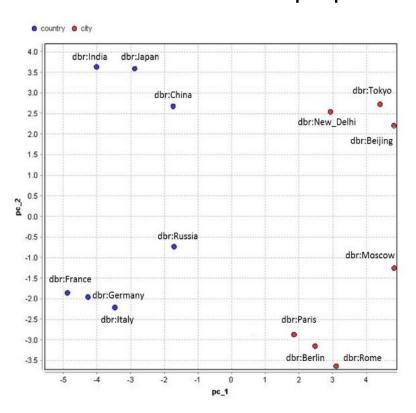


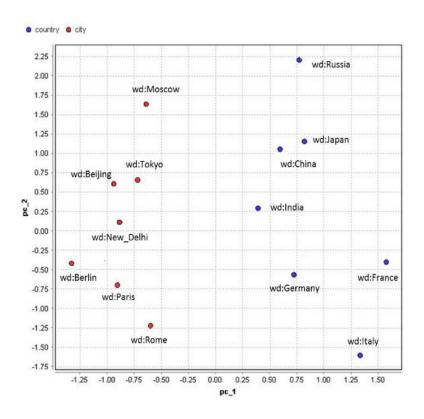
#### From word2vec to RDF2vec

- Generating sequences from an RDF dataset
  - by starting random walks from each entity
- Example:
  - dbr:Germany dbo:capital dbr:Berlin dbo:mayor dbr:Michael Mueller
- Those are fed into a word2vec training engine
- Variants (Cochez et al., 2017)
  - replace "random" by "semi-random" walk
  - e.g., weight edges by frequency, PageRank, ...

#### From word2vec to RDF2vec

Observation: similar properties hold for RDF2vec





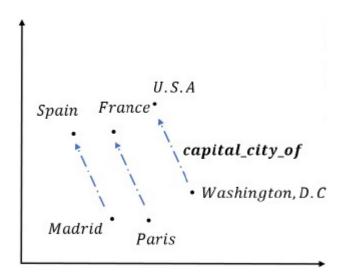
a) DBpedia vectors

b) Wikidata vectors

Ristoski & Paulheim: RDF2vec: RDF Graph Embeddings for Data Mining, 2016

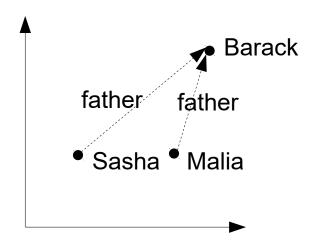
#### **TransE and Descendants**

- In RDF2vec, relation preservation is a by-product
- TransE: direct modeling
  - Formulates RDF embedding as an optimization problem
  - Find mapping of entities and relations to R<sup>n</sup> so that
    - across all triples <s,p,o>
       Σ ||s+p-o|| is minimized



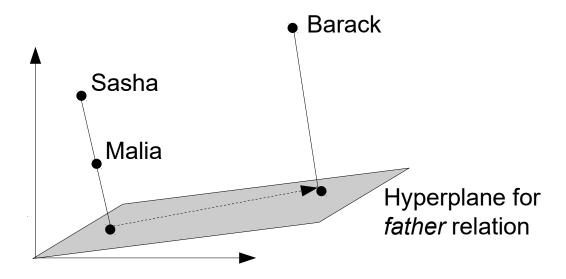
#### **Limitations of TransE**

- TransE works fine if we have 1:1 relations
  - But what in case of 1:n or n:m relations?
- Example below:
  - We have to miminize||Malia + father Barack|| + ||Sasha + father Barack||
  - This is minimized if ||Malia|| = ||Sasha||
  - i.e., they become indistinguishable



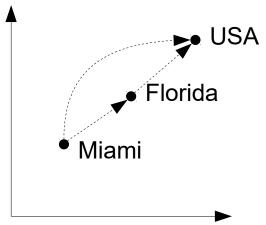
#### **Extension: TransH**

- In TransH, there is a hyperplane per relation
  - Subject and object are projected to that hyperplane
  - On each hyperplane, a TransE-like optimization is conducted
- Sasha and Malia become indistinguishable on father hyperplane
  - But are still distinguishable in the vector space



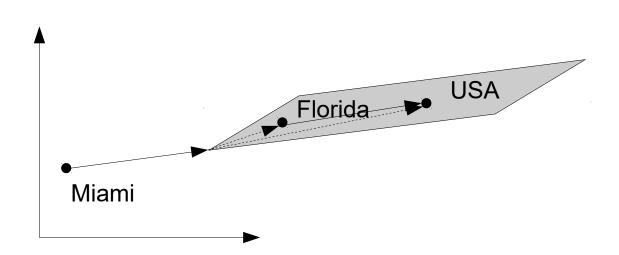
#### **Limitations of TransE**

- Transitive Properties
  - we have to minimize
     ||Miami + partOf Florida|| and ||Florida + partOf USA||, but also
     ||Miami + partOf USA||
  - ideally, Miami + partOf = Florida, Florida + partOf = USA,Miami + partOf = USA
    - Again: all three become infinitely close
    - partOf becomes 0 vector



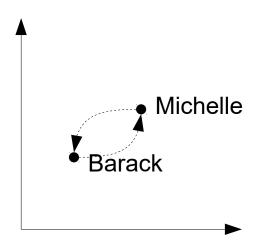
# **Extension: TransE-DT (2017)**

- Entities and relations can be hyperplanes
  - Represented as the vector of the hyperplane plus an individual vector
  - Relations can hold between a point and a hyperplane or two hyperplanes
  - Relations may be hyperplanes themselves



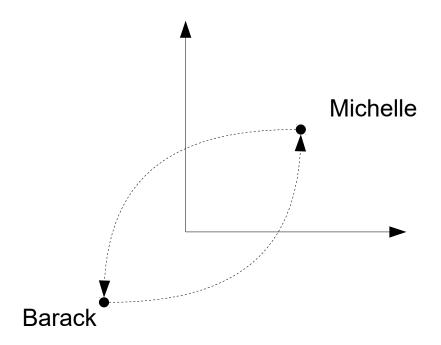
#### **Limitations of TransE**

- Symmetric properties
  - we have to minimize
     ||Barack + spouse Michelle|| and ||Michelle + spouse Barack||
     simultaneously
  - ideally, Barack + spouse = Michelle and Michelle + spouse = Barack
    - Michelle and Barack become infinitely close
    - spouse becomes 0 vector



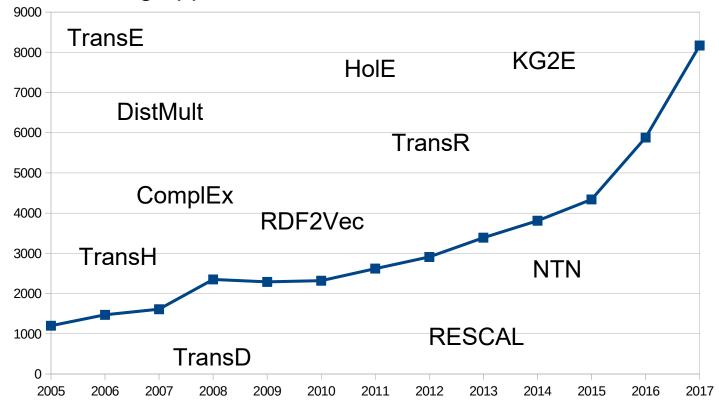
# Extension: RotatE (2019)

- Relations are not represented as straight vectors, but rotations
  - This allows for symmetric relations
  - Also works for reflexive relations (they become a rotation by 360°)



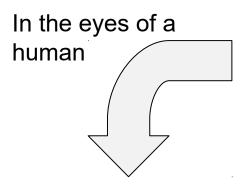
### **Limitations of TransE**

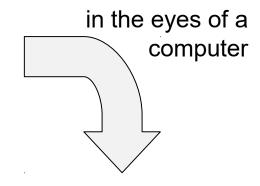
- Numerous variants of TransE have been proposed to overcome limitations (e.g., TransH, TransR, TransD, ...)
- Plus: embedding approaches based on tensor factorization etc.



#### A Look Back

 This is where we embarked on our Semantic Web Technologies journey in September:





```
Dr. Mark Smith
Physician
Main St. 14
Smalltown
Mon-Fri 9-11 am
Wed 3-6 pm
```

Print in bold: "hmf298hmmhudsa"
Print in italics: "mj2i9ji0"
Print normal: "fdsah
02hfadsh0um2m0adsmf0ihm
asdfjköfdsa298ndsfmij32mio
lk2mjpoimjiofdpmsajiomjm"

### A Look Back

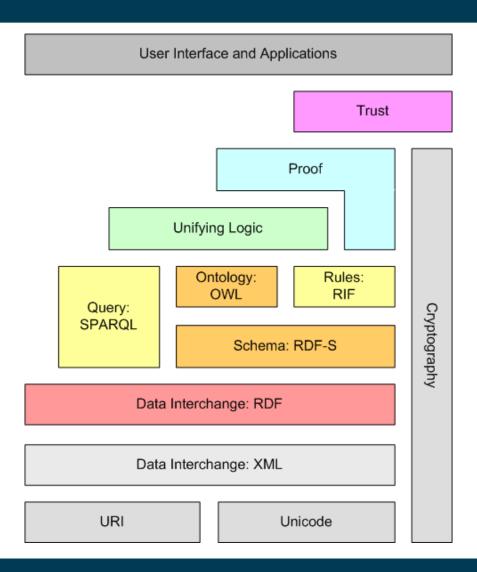
#### Formal Semantics

- Every entity has classes and relations to other entities
- Those are defined in an ontology
- Humans and computers can interpret those semantics
- Computers give justification on reasoning results

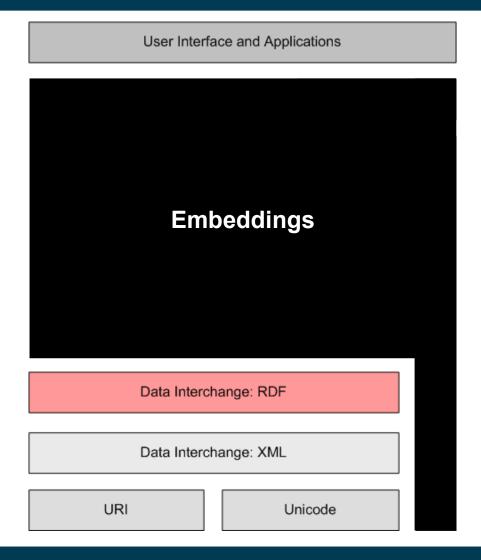
#### Embeddings

- Every entity is an n-dimensional vector
- We do not know about the meaning of the dimensions
- Results are often good, but hard to justify

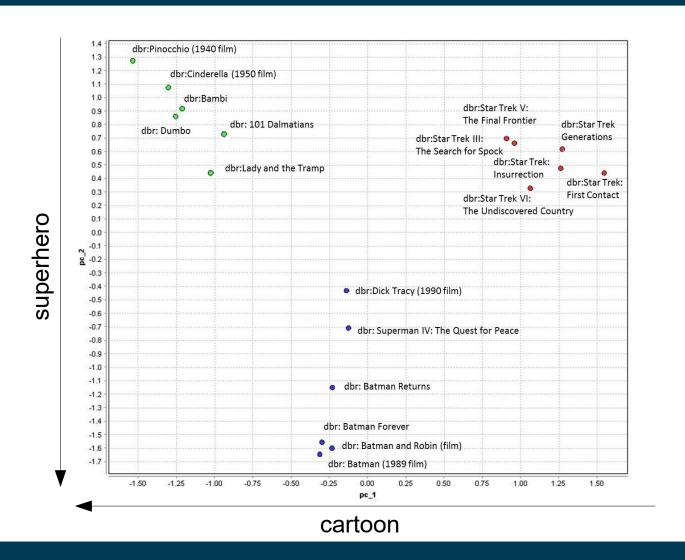
# The 2009 Semantic Web Layer Cake



# The 2018 Semantic Web Layer Cake



# **Towards Semantic Vector Space Embeddings**



# The Holy Grail

- Combine semantics and embeddings
  - e.g., directly create meaningful dimensions
  - e.g., learn interpretation of dimensions a posteriori

**–** ...



# **Summary**

- OWL and OWL 2 are not the end
  - Rules create more possibilities
- Other (non W3C standard) languages have also been also proposed
  - different semantic paradigms (e.g., F-Logic)
  - different problem setting (e.g., SHACL)
- Recent trend
  - using vector space embeddings
  - challenge: combine interpretable semantics and embeddings

# **Questions?**

