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#### Gain information about job applications and career entry!

Company Contact Fair: Getting to know various national and international companies and making direct contact with potential employers

Career Lounge: Exclusive one-on-one conversations with company representatives
 from seven interesting companies

 Lectures und Workshops: Helpful tips from experienced coaches aspects of starting a career

CV-Checks: Get your CV checked to obtain personalised feedback

JobWall: Job advertisements from 16 companies for working students are:









## **Previously on Introduction to Linked Data...**

- We have learned about RDF and how to publish and access data
- The notion of RDF instance explains the meaning of blank nodes
- We have learned to write SPARQL queries against an RDF dataset

## **Lecture 3: Syntax of RDF Triples**

**Definition 6** (RDF Triple, RDF Graph). Let  $\mathcal{U}$  be the set of URIs,  $\mathcal{B}$  the set of blank nodes, and  $\mathcal{L}$  the set of RDF literals. A tuple  $\langle s, p, o \rangle \in (\mathcal{U} \cup \mathcal{B}) \times \mathcal{U} \times (\mathcal{U} \cup \mathcal{B} \cup \mathcal{L})$  is called an RDF triple, where s is the subject, p is the predicate and o is the object. A set of RDF triples is called RDF graph.

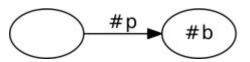
## **Lecture 3: RDF Instance Mapping**

- To understand how blank nodes are handled, we start with the notion of an instance of a graph
- For that, we need the notion of RDF Instance Mapping

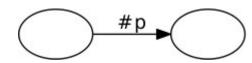
**Definition 10** (RDF Instance Mapping, RDF Instance). A partial function from blank nodes to RDF terms  $\sigma \colon \mathcal{B} \mapsto \mathcal{U} \cup \mathcal{B} \cup \mathcal{L}$  is called an RDF instance mapping. We write  $\sigma(G)$  to denote an RDF graph obtained from graph G by replacing each blank node x in G with  $\sigma(x)$ . We call  $\sigma(G)$  an instance of G.

## Lecture 3: Example RDF Instance Mapping

Graph G0: \_:a <#p> <#b> .



• Graph G1:  $\_:bn1 < \#p > <math>\_:bn2$ .



- G0 is an instance of G1, assuming the RDF instance mapping :
  - (\_:bn1) = \_:a
  - (\_:bn2) = <#b>

## Lecture 3: Subgraph

A subgraph of an RDF graph is a subset of the triples in the graph.

Note that "subgraph" is to be taken literally. Datatyped literals are compared as they are, without taking into account that the same value might have different lexical representations. For example:

```
{:a :b "01"^^xsd:integer .} is not a subgraph of {:a :b
"1"^^xsd:integer .}
```

# C06 SPARQL Query Processing What are correct solutions to queries?

Version 2021-05-12

Lecturer: Prof. Andreas Harth

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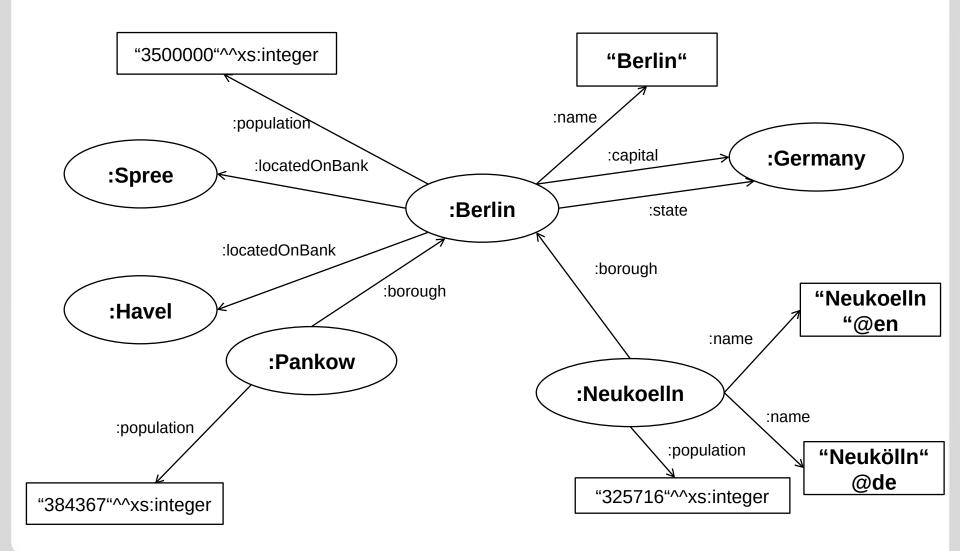
- This set of slides is part of the lecture "Semantic Web Technologies" held at Karlsruhe Institute of Technology
- The content of the lecture was prepared by PD Dr. Andreas Harth based on his book "Introduction to Linked Data"
- The initial slides were prepared by Lars Heling
- Additional work on the slides by Maribel Acosta and Andreas Harth

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## Remember: The cities.ttl Graph



## **An Example Query**

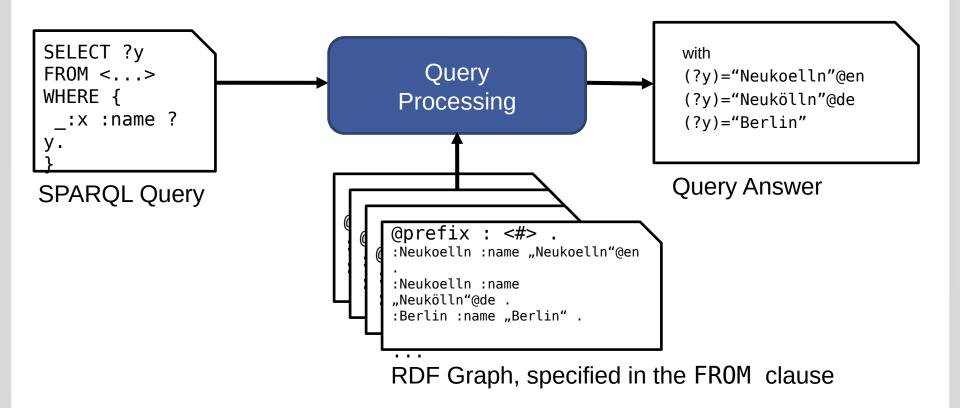
Based on the document http://example.org/cities.ttl, the following query shall be executed:

```
PREFIX : <http://example.org/cities.ttl#>
SELECT ?y
FROM <http://example.org/cities.ttl>
WHERE {
    _:x :name ?y .
}
```

What are the results of the query?

## **Processing a SPARQL Query**

Processing a SPARQL query against an RDF dataset yields the solutions (or answers) of the query.



## **How to Compute the Solutions of a Query?**

Computing the solutions to a SPARQL query against a graph specified in the FROM clause involves the following steps:

- Translation of the SPARQL query into an abstract SPARQL query, including an algebra expression.
- Evaluation of the algebra expression for the graph patterns (conditions specified in the WHERE clause of the query).
- Accessing the RDF dataset and processing the query form.

## **Agenda**

- 1. Basic Graph Pattern Matching
- 2. Translating Queries to Abstract Queries
- 3. Evaluating Graph Pattern Algebra Expressions
- 4. Processing Abstract Queries

## From SPARQL Queries to Algebra Expressions

We start with:

The algebra expression obtained for the conditions specified in the WHERE clause (graph patterns).

For the sake of simplicity, we ignore the prefix declarations.

```
PREFIX:
<http://example.org/cities.ttl#>
SELECT ?y
FROM <http://example.org/cities.ttl>
WHERE {
   ?borough :borough ?berlin .
   ?berlin :name "Berlin" .
}
```

## **Basic Graph Patterns (BGPs)**

Translate the following graph pattern into an algebra expression.

```
{ ?borough :borough ?berlin .
 ?berlin :name "Berlin" .}
```

**Solution:** Let us consider

```
tp1 = ?borough :borough ?berlin .

tp2 = ?berlin :name "Berlin" .
```

Then, the resulting algebra expression is:

```
BGP(tp1 . tp2)
```

We deal with blank nodes in BGPs later; for now, assume blank nodes are variables.

## **Definition: Basic Graph Pattern**

**Definition 12** (Variable, Triple Pattern, Basic Graph Pattern). *Let* V *be an infinite set of variables;*  $V \cap (U \cup B \cup L) = \emptyset$ . *A triple*  $p \in (U \cup B \cup V) \times (U \cup V) \times (U \cup B \cup L \cup V)$  *is called a triple pattern. A set of triple patterns is called basic graph pattern (BGP).* 

Example of triple patterns:

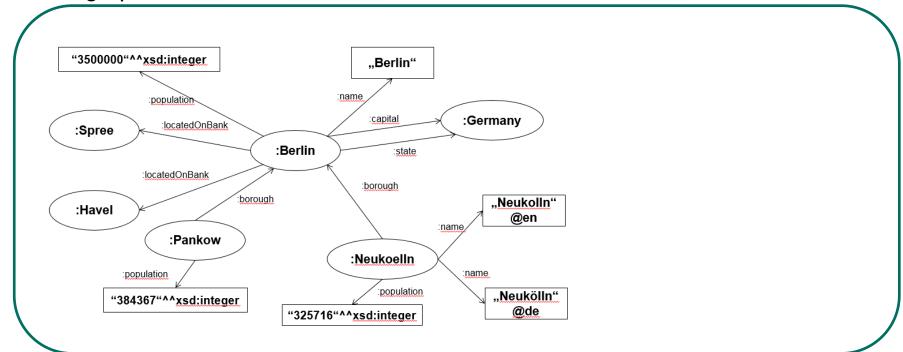
```
tp1 = ?borough :borough ?berlin .
tp2 = ?berlin :name "Berlin" .
```

The entire BGP is:

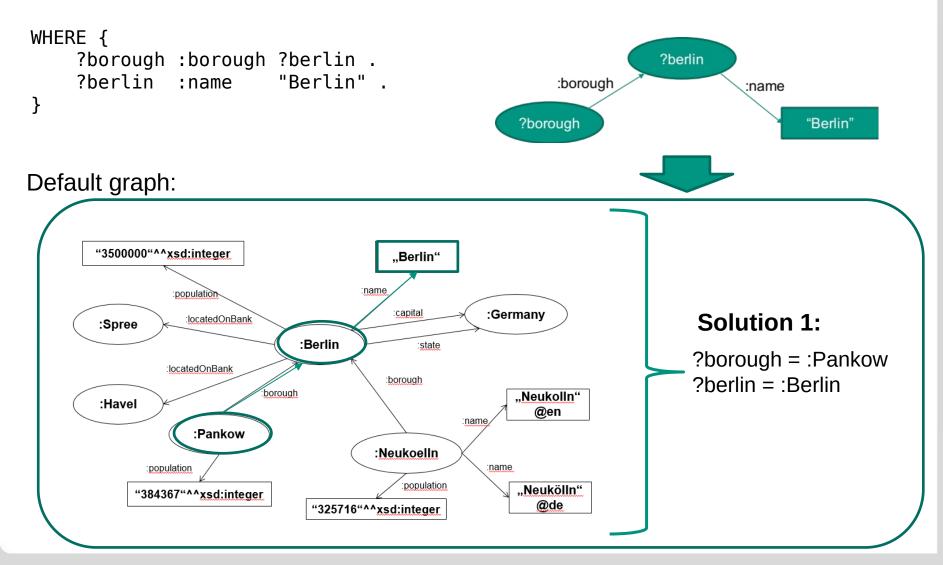
```
?borough :borough ?berlin .
?berlin :name "Berlin" .
```

## Pattern Matching: Intuition (1)

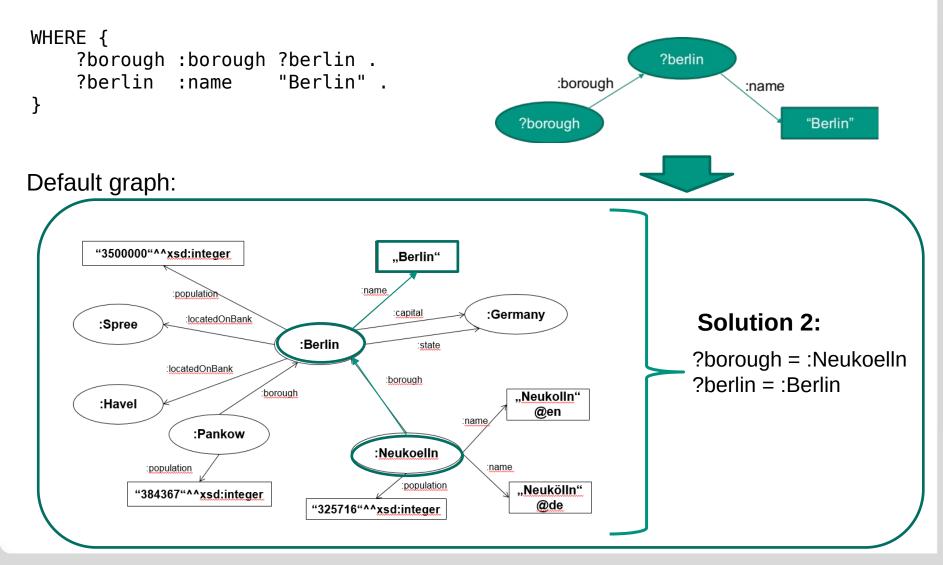
#### Default graph:



## Pattern Matching: Intuition (2)

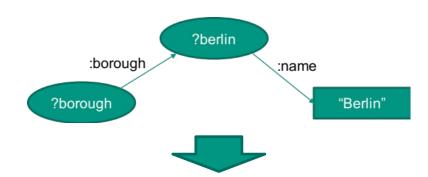


## Pattern Matching: Intuition (3)

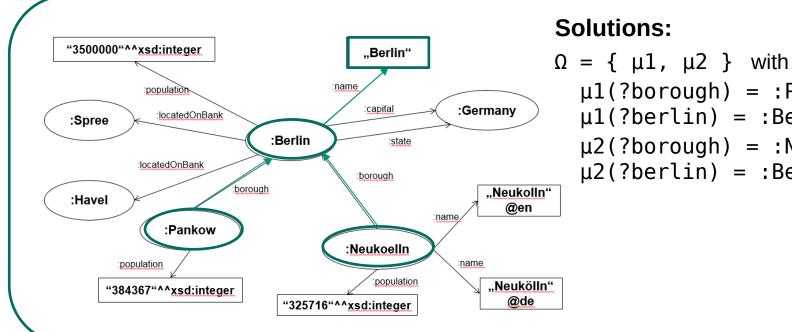


## Pattern Matching: Intuition (4)

```
WHERE {
   ?borough :borough ?berlin .
    ?berlin :name
                     "Berlin" .
```



#### Default graph:



```
\mu 1(?borough) = :Pankow,
\mu 1(?berlin) = :Berlin
\mu 2(?borough) = :Neukoelln,
\mu 2(?berlin) = :Berlin
```

## **Definition: Solution Mapping**

**Definition 13** (Solution Mapping, Solution Sequence). A partial function from variables to RDF terms  $\mu \colon \mathcal{V} \mapsto \mathcal{U} \cup \mathcal{B} \cup \mathcal{L}$  is called a solution mapping. A solution sequence  $\Omega$  is a collection of solution mappings.

- A collection is either:
  - a set (no duplicates allowed, elements not ordered)
  - a bag (duplicates allowed, elements not ordered) or
  - a list (duplicates allowed, elements are ordered)
- We assume that the solution sequence is a set, unless otherwise noted

## **Solution Mapping: Example**

**Definition 13** (Solution Mapping, Solution Sequence). A partial function from variables to RDF terms  $\mu \colon \mathcal{V} \mapsto \mathcal{U} \cup \mathcal{B} \cup \mathcal{L}$  is called a solution mapping. A solution sequence  $\Omega$  is a collection of solution mappings.

Consider the solution sequence  $\Omega$  consisting of the following solution mappings  $\mu 1$  and  $\mu 2$ :

```
\mu1(?borough) = :Pankow, \mu1(?berlin) = :Berlin 
 <math>\mu2(?borough) = :Neukoelln, \mu2(?berlin) = :Berlin
```

The domain of  $\mu 1$  and  $\mu 2$  are as follows:

```
dom(\mu 1) = \{?borough, ?berlin\}

dom(\mu 2) = \{?borough, ?berlin\}
```

## **Definition: Basic Graph Pattern Matching**

**Definition 14** (Basic Graph Pattern Matching). Let P be a basic graph pattern and G be an RDF graph. A partial function  $\mu$  is a solution of matching P on the graph G if:

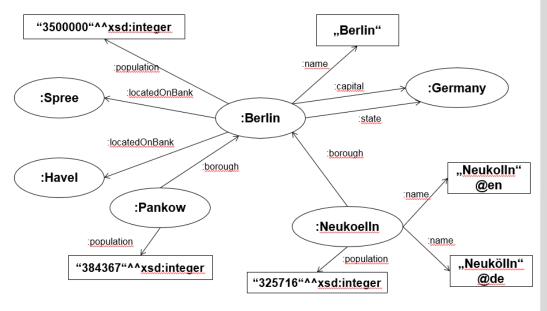
- the domain of  $\mu$  is the set of variables in P, and
- there exists a mapping  $\sigma$  of blank nodes in P to RDF terms ( $U \cup B \cup L$ ) in G, so that
- the graph  $\mu(\sigma(P))$  is a subgraph of G.

We write  $\mu(P)$  to denote an RDF graph obtained from BGP P by replacing each variable x in P with  $\mu(x)$ .

## **Example of BGP Matching (1)**

- Consider BGP P = ?borough :borough ?berlin .
  ?berlin :name "Berlin" .
- Consider solution  $\mu 1$  $\mu 1$ (?borough) = :Pankow,  $\mu 1$ (?berlin) = :Berlin
- Conditions:
- the domain of μ1 is the set of variables in P
- 2. no blank nodes, so empty mapping σ
- 3. the graph  $\mu(P)$  is a subgraph of G

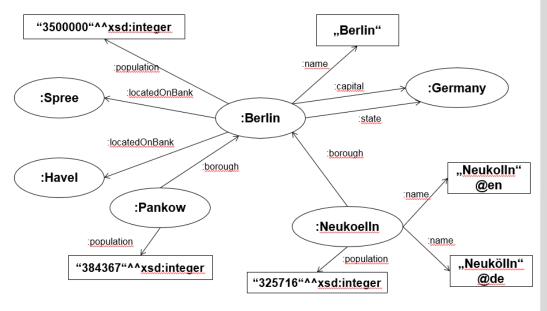
#### RDF Graph G:



## **Example of BGP Matching (2)**

- Consider P = ?borough :borough ?berlin .
  ?berlin :name "Berlin" .
- Consider solution  $\mu$ 2  $\mu$ 2(?borough) = :Neukoelln,  $\mu$ 2(?berlin) = :Berlin
- Conditions:
- the domain of μ1 is the set of variables in P
- 2. no blank nodes, so empty mapping  $\sigma$
- 3. the graph  $\mu(P)$  is a subgraph of G

#### RDF Graph G:



### Think-Pair-Share

RDF graph G:



**Definition 14** (Basic Graph Pattern Matching). Let P be a basic graph pattern and G be an RDF graph. A partial function  $\mu$  is a solution of matching P on the graph G if:

- ullet the domain of  $\mu$  is the set of variables in P, and
- there exists a mapping  $\sigma$  of blank nodes in P to RDF terms ( $U \cup B \cup L$ ) in G, so that
- the graph  $\mu(\sigma(P))$  is a subgraph of G.

## **Agenda**

- 1. Basic Graph Pattern Matching
- 2. Translating Queries to Abstract Queries
- 3. Evaluating Graph Pattern Algebra Expressions
- 4. Processing Abstract Queries

## From SPARQL Queries to Abstract Queries

For processing entire SPARQL queries we require a more structured representation of a query string:

- The algebra expression obtained for the conditions specified in the WHERE clause (graph patterns) and the solution modifiers.
- The query form.
- The RDF dataset over which the query is evaluated.

For the sake of simplicity, we ignore the prefix declarations.

```
PREFIX:
<a href="http://example.org/citie"><a href="http://example.org/citie">http://example.org/citie</a><a href="http://example.org/citie">http://example.org/citie</a><a href="http://example.org/citie">http://example.org/cit
```

## **Named Graphs and RDF Dataset**

**Definition 15** (Named Graph, RDF Dataset). Let  $\mathcal{G}$  be the set of RDF graphs and  $\mathcal{U}$  be the set of URIs. A pair  $\langle g, u \rangle \in \mathcal{G} \times \mathcal{U}$  is called a named graph. An RDF dataset consists of a (possibly empty) set of named graphs (with distinct names) and a default graph  $g \in \mathcal{G}$  without a name.

## **SPARQL Abstract Query**

**Definition 16** (SPARQL Abstract Query). A SPARQL abstract query is a tuple  $\langle M, D, QF \rangle$ , where M is a SPARQL algebra expression, D is an RDF dataset and QF is a query form.

- We start with algebra expressions of the graph pattern in the WHERE clause of a query.
- We consider query forms and RDF dataset descriptions later.

## **Example: Basic Graph Pattern (BGP)**

Translate the following graph pattern into an algebra expression.

```
{ _:b :borough :Berlin .
   _:b :name ?y .}
```

Solution: Let us consider

```
tp1 = \_:b :borough :Berlin .

tp2 = \_:b :name ?y .
```

```
BGP(tp1 . tp2)
```

## **Example: JOIN**

Translate the following graph pattern into an algebra expression.

```
{{ ?x :borough :Berlin .
    ?x :name ?y .}
    { ?x rdf:type ?t .}}
```

#### Solution: Let us consider

```
tp1 = ?x : borough : Berlin .

tp2 = ?x : name ?y .

tp3 = ?x : rdf : type ?t .
```

```
JOIN(BGP(tp1 . tp2), BGP(tp3))
```

## **Example: UNION**

Translate the following graph pattern into an algebra expression.

```
{ ?x :borough :Berlin .
  { ?x :name ?y . }
  UNION
  { ?x :population ?z . } }
```

Solution: Let us consider

```
tp1 = ?x : borough : Berlin .

tp2 = ?x : name ?y .

tp3 = ?x : population ?z .
```

```
JOIN(BGP(tp1), UNION(BGP(tp2), BGP(tp3)))
```

## **Example: OPTIONAL**

Translate the following graph pattern into an algebra expression.

```
{ ?x :borough :Berlin .
  { ?x :name ?y . }
  OPTIONAL
  { ?x :population ?z . } }
```

Solution: Let us consider

```
tp1 = ?x : borough : Berlin .

tp2 = ?x : name ?y .

tp3 = ?x : population ?z .
```

```
JOIN(BGP(tp1), LEFTJOIN(BGP(tp2), BGP(tp3), true))
```

## **Example: FILTER**

Translate the following graph pattern into an algebra expression.

```
{ ?x :borough :Berlin .
    { ?x :name ?y . }
    UNION
    { ?x :population ?z . FILTER (?z > 350000)} }
```

Solution: Let us consider

```
tp1 = ?x : borough : Berlin .

tp2 = ?x : name ?y .

tp3 = ?x : population ?z .
```

```
JOIN(BGP(tp1), UNION(BGP(tp2), FILTER(?z>350000, BGP(tp3))))
```

## **Example: EXTEND**

Translate the following graph pattern into an algebra expression.

```
{ ?x :borough :Berlin .
   ?x :population ?z .
   BIND (?z/1000000 AS ?m) }
```

Solution: Let us consider

```
tp1 = ?x : borough : Berlin .

tp2 = ?x : population ?z .
```

```
EXTEND(BGP(tp1 . tp2), ?z/1000000, ?m)
```

### **Example: GRAPH**

Translate the following graph pattern into an algebra expression.

```
{ GRAPH ?g {
    ?x :borough :Berlin .
    ?x :population ?z . } }
```

Solution: Let us consider

```
tp1 = ?x : borough : Berlin .

tp2 = ?x : population ?z .
```

Then, the resulting algebra expression is:

```
GRAPH(?g, BGP(tp1 . tp2))
```

## **Selected Algebra Expressions for Graph Patterns**

Symbol	Description
BGP(P)	Adjacent triple patterns <i>P</i> form a Basic Graph Pattern.
JOIN(M1, M2)	Conjunctive combination of algebra expression M1, M2 (M1 and M2).
UNION(M1, M2)	Alternative combination of algebra expressions M1, M2 (M1 or M2).
LEFTJOIN(M1, M2, expr)	Combination of algebra expression M1 with optional algebra expression M2 and filter condition <i>expr</i> .
FILTER(expr, M)	Use filter condition $expr$ on algebra expression $M$ .
EXTEND(M, expr, var)	Add extension expression $expr$ to algebra expression $M$ for .
GRAPH(g, M)	Apply the algebra expression <i>M</i> to graphs from

# **Translating Query Forms**

Selected query forms.

Symbol	Description
SELECT(vars)	Return only variables from graph pattern solutions.
CONSTRUCT(pat)	Construct an RDF graph from the template pattern pat and the graph pattern solutions.
ASK	Return true if there exist at least one solution to the graph pattern. Return false otherwise.

### **Translating Dataset Clauses to Dataset Descriptions**

Translate the RDF dataset clause into a dataset description.

## **Translating Dataset Clauses to Dataset Descriptions**

Translate the RDF dataset clause into a dataset description.

■ **Solution:** The RDF dataset description consists of the default graph with the RDF merge of http://example.org/cities.ttl and http://example.org/munich.ttl. The dataset has no named graphs.

#### **Think-Pair-Share**

Translate the WHERE clause of the following SPARQL query into an SPARQL algebra expression:

```
PREFIX : <http://example.org/cities.ttl#>
SELECT ?x ?p
FROM <http://example.org/cities.ttl>
WHERE {
  ?x :population ?p .
  { ?x :borough :Berlin . }
  UNTON
  { ?x :borough :Muenchen . }
  FILTER (?p > 100000)
```

## **Agenda**

- 1. Basic Graph Pattern Matching
- 2. Translating Queries to Abstract Queries
- 3. Evaluating Graph Pattern Algebra Expressions
- 4. Processing Abstract Queries

## **Evaluating Algebra Expressions**

- To evaluate the algebra expression, the eval() function is introduced
- The input of eval() is a graph and an algebra expression
- The output of eval() is a solution sequence Ω (a set of solution mappings)

# The eval() Function

- Let D be an RDF dataset
- Let D(G) be an RDF dataset with the active graph G
- Let M be a SPARQL algebra expression
- **Let**  $\Omega$  be a solution sequence
- The SPARQL recommendation defines *eval()* as: Ω = eval(D(G), M)
- The eval() function is applied recursively over the algebra expression to evaluate each the expression (defined as algebra operators)

#### **Recursive Procedure**

- Let P be a basic graph pattern
- Let M, M1, M2 be SPARQL algebra expressions
- Let expr be a function expression
- The initially active graph is the default graph of D
- Now, apply eval() recursively:
  - $eval(D(G), BGP(P)) = \Omega$  (see definition Basic Graph Pattern Matching)
  - $\bullet$  eval(D(G), JOIN(M1, M2)) = Join(eval(D(G), M1), eval(D(G), M2))
  - $\bullet$  eval(D(G), UNION(M1, M2)) = Union(eval(D(G), M1), eval(D(G), M2))
  - eval(D(G), LEFTJOIN(M1, M2, expr)) = LeftJoin(eval(D(G), M1), eval(D(G), M2), expr)
  - eval(D(G), FILTER (expr, M)) = Filter(expr, eval(D(G), M), D(G))
  - eval(D(G), EXTEND(M, expr, var)) = Extend(eval(D(G), M), var, expr)
  - eval(D(G), GRAPH(g, M)) = change the active graph G

# **Compatibility of Solution Mappings**

**Definition** (Solution Mapping Compatibility). Two mappings  $\mu_1$ ,  $\mu_2$  are compatible, written  $\mu_1 \sim \mu_2$ , if  $\forall x \in dom(\mu_1) \cap dom(\mu_2) : \mu_1(x) = \mu_2(x)$ .

In other words: Two solution mappings are compatible if variables with the same name are bound to the same RDF term.

#### Example:

Consider the following solution mappings  $\mu 1$  and  $\mu 2$ :

```
\mu 1(?x) = :Neukoelln, \mu 1(?y) = "Neukoelln"@en 
 <math>\mu 2(?x) = :Neukoelln
```

#### $\mu$ 1 ~ $\mu$ 2?

```
We have that dom(\mu 1) = \{?x, ?y\} and dom(\mu 2) = \{?x\}. The variables in common are: dom(\mu 1) = dom(\mu 2) = \{?x\} \mu 1(?x) = :Neukoelln \text{ and } \mu 2(?x) = :Neukoelln Therefore, \mu 1 \sim \mu 2.
```

# **Think-Pair-Share: Compatibility of Solution Mappings**

μ1	μ2	μ1 ~ μ2? Why?
$\mu1(?x) = :Sonne,$ $\mu1(?y) = :Erde$	$\mu 2(?y) = :Erde, \\ \mu 2(?z) = :Mond$	
μ1(?x) = :Sonne, μ1(?y) = :Erde	$\mu2(?y) = :Mars,$ $\mu2(?z) = :Deimos$	
$\mu$ 1(?s1) = :Mars	$\mu 2(?s2) = :Venus$	
μ1(?s1) = :Mars, μ1(?s2) = :Erde	μ2(?s1) = :Mars, μ2(?s2) = :Erde, μ2(?s3) = :Sonne	 

# The Empty Solution Mapping $\mu_{\emptyset}$

- The empty solution mapping, denoted  $\mu_g$ , is the mapping with empty domain, i.e., dom( $\mu_g$ ) =  $\emptyset$ .
- By definition of *compatibility*, we obtain that  $\mu_g$  is compatible with all solution mappings.

#### Example:

Consider the following solution mappings  $\mu 1$  and  $\mu_{g}$ :

```
\mu 1(?x) = :Neukoelln, \ \mu 1(?y) = "Neukoelln"@en \ \mu_\emptyset = \{ \}
```

 $\mu 1$  and  $\mu_{\alpha}$  are compatible, i.e.,  $\mu 1 \sim \mu_{\alpha}$ 

# **Definition of Operators**

Operator	Definition
$Union(\Omega_1,\Omega_2)$	$\{\mu \mid \mu \in \Omega_1 \text{ or } \mu \in \Omega_2\}$
$Join(\Omega_1,\Omega_2)$	$\{\mu_1 \cup \mu_2 \mid \mu_1 \in \Omega_1, \mu_2 \in \Omega_2 \text{ and } \mu_1, \mu_2 \text{ are compatible}\}$
$Diff(\Omega_1,\Omega_2,F)$	$\{\mu \mu\in\Omega_1 \text{ such that for all }\mu'\in\Omega_2, \text{ either }\mu \text{ and }\mu' \text{ are not compatible }\}$
	or $\mu$ and $\mu'$ are compatible and $F(merge(\mu, \mu'))$ is false }
$Left Join(\Omega_1, \Omega_2, F)$	$Filter(F, Join(\Omega_1, \Omega_2)) \cup Diff(\Omega_1, \Omega_2, F)$
$Filter(expr,\Omega)$	$\{\mu \mid \mu \in \Omega \text{ and } expr(\mu) \text{ is an expression that evaluates to true}\}$
$Extend(\mu, var, expr)$	$\mu \cup \{(var, value) \mid var \notin dom(\mu) \text{ and } value = expr(\mu)\}$
	$\mu$ if $var \notin dom(\mu)$ and $expr(\mu)$ is an error
	undefined when $var \in dom(\mu)$
Extend $(\Omega, var, expr)$	$\{Extend(\mu, var, expr) \mid \mu \in \Omega\}$

### **Example:** Join Operator

Consider the following sets of solution mappings:

```
\Omega 1 = \{ \mu 1, \mu 2, \mu 3 \} with \mu 1(?x) = :Pankow, \mu 1(?y) = :Berlin \mu 2(?x) = :Neukoelln, \mu 2(?y) = :Berlin \mu 3(?x) = :Barcelona, \mu 3(?y) = :Catalonia <math>\Omega 2 = \{ \mu 4 \} with \mu 4(?y) = :Berlin, \mu 4(?z) = :Germany
```

**Compute Join** (Ω1, Ω2)

Compatible mappings!

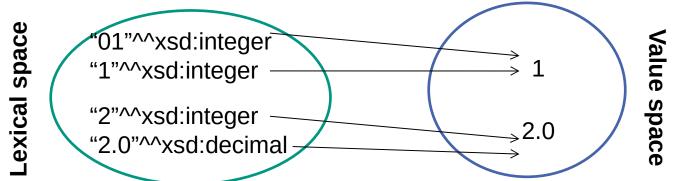
```
Join(\Omega_{l},\Omega_{r}):=\{\mu_{l}\cup\mu_{r}\mid\mu_{l}\in\Omega_{l},\mu_{r}\in\Omega_{r}:\mu_{l}\sim\mu_{r}\} 
 Join(\Omega 1,\ \Omega 2)=\{ \mu 14(?x)=:Pankow,\ \mu 14(?y)=:Berlin,\ \mu 14(?z)=:Germany \ \mu 24(?x)=:Neukoelln,\ \mu 24(?y)=:Berlin,\ \mu 24(?z)=:Germany \}
```

# Filter Operator: expr(μ)

- expr(μ) evaluates the instantiation of variables in expr with the values of μ. Returns TRUE, FALSE, or error.
- For example:

```
\mu1(?b) = "01"^^xsd:integer and F = (?b = 1), then F(<math>\mu1) evaluates ("01"^^xsd:integer = 1)
```

• Unlike pattern matching,  $expr(\mu)$  considers the value space of RDF literals (and not the lexical space).



 $\blacksquare$  Taking into account the value space, we conclude that F(µ1) from the example evaluates to TRUE.

## **Type Conversion**

\* SPARQL does automatic type casting between xsd:float and xsd:decimal.

- Some datatypes (e.g., xsd:float and xsd:decimal) share a lexical space (although they have a different value space!)
- To be able to compare literals of different (incompatible) datatypes, we can use casting.
- For example, consider foo.ttl:

```
@prefix : <foo#> .
:s :p "9.90" ; :q 9.90 .
```

Now, we can compare the two literals as follows:

### **Example:** *Filter* Operator

Consider the following set of solution mappings:

```
\Omega 1 = \{ \mu 1, \mu 2 \} with \mu 1(?a) = :Pankow, \mu 1(?b) = "384367"^^xsd:integer <math>\mu 2(?a) = :Neukoelln, \mu 1(?b) = "325716"^^xsd:integer
```

• Compute Filter((?b > 350000),  $\Omega1$ )

```
Filter(expr, \Omega) := \{\mu | \ \mu \in \Omega \ \ and \ expr(\mu) \ evaluates \ to \ true \}
```

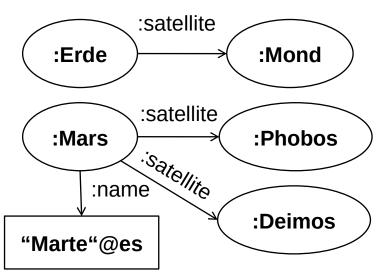
```
For \mu1: ("384367"^^xsd:integer > 350000) TRUE
```

For  $\mu2$ : ("325716"^^xsd:integer > 350000) FALSE

Only  $\mu 1$  is part of the solution:

```
Filter((?b > 350000), \Omega1) = { \mu1 } with \mu1(?a) = :Pankow, \mu1(?b) = "384367"^^xsd:integer
```

#### Given is D(G):



Evaluate the following SPARQL algebra expression against D(G):

```
JOIN(BGP(P1), BGP(P2))
With
```

```
P1 = ?p :satellite ?s1 .
```

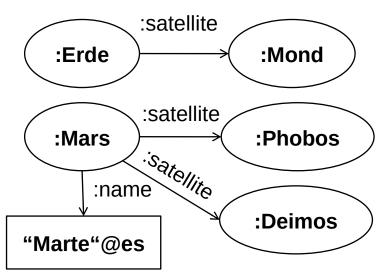
$$P2 = ?p : satellite ?s2$$
.

**Solution:** We have to compute  $\Omega = \text{eval}(D(G), JOIN(BGP(P1), BGP(P2))).$ 

$$eval(D(G), JOIN(P_1, P_2))$$
 :=  $Join(eval(D(G), P_1), eval(D(G), P_2))$ 

We have to compute  $\Omega1=eval(D(G), P1)$  and  $\Omega2=eval(D(G), P2)$  and then Join the results  $\Omega1$  and  $\Omega2$ .

#### Given is D(G):



Evaluate the following SPARQL algebra expression against D(G):

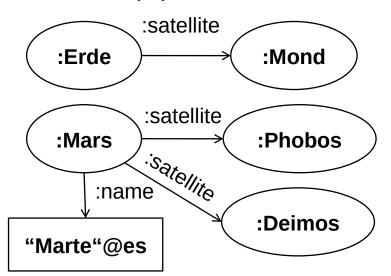
```
JOIN(BGP(P1), BGP(P2))
With
P1 = ?p :satellite ?s1 .
P2 = ?p :satellite ?s2 .
```

Solution (Cont): Let's compute  $\Omega 1 = eval(D(G), P1)$ .

P1 is a triple pattern. The definition of *eval* for triple patterns says that we have to apply pattern matching:  $\Omega 1 = \{ \mu 1, \mu 2, \mu 3 \}$  with

```
\mu1(?p) = :Erde, \ \mu1(?s1) = :Mond \ \mu2(?p) = :Mars, \ \mu2(?s1) = :Phobos \ \mu3(?p) = :Mars, \ \mu3(?s1) = :Deimos
```

#### Given is D(G):



Evaluate the following SPARQL algebra expression against D(G):

P1 J0IN P2

With

P1 = ?p : satellite ?s1.

P2 = ?p : satellite ?s2.

**Solution (Cont):** Now we compute  $\Omega 2 = eval(D(G), P2)$ .

P2 is a triple pattern. The definition of *eval* for triple patterns says that we have to apply pattern matching:  $\Omega 2 = \{ \mu 4, \mu 5, \mu 6 \}$  with

 $\mu 4(?p) = :Erde, \mu 4(?s2) = :Mond$ 

 $\mu 5(?p) = :Mars, \mu 5(?s2) = :Phobos$ 

 $\mu6(?p) = :Mars, \mu6(?s2) = :Deimos$ 

**Solution (Cont):** Now we compute  $\Omega = Join(\Omega 1, \Omega 2)$ , with:

```
\Omega 1 = \{ \mu 1, \mu 2, \mu 3 \} with \mu 1(?p) = :Erde, \mu 1(?s1) = :Mond \mu 2(?p) = :Mars, \mu 2(?s1) = :Phobos \mu 3(?p) = :Mars, \mu 3(?s1) = :Deimos <math>\Omega 2 = \{ \mu 4, \mu 5, \mu 6 \} with \mu 4(?p) = :Erde, \mu 4(?s2) = :Mond \mu 5(?p) = :Mars, \mu 5(?s2) = :Phobos \mu 6(?p) = :Mars, \mu 6(?s2) = :Deimos <math>\Omega = \{ m \}
```

**Solution (Cont):** Now we compute  $\Omega = Join(\Omega 1, \Omega 2)$ , with:

```
\Omega 1 = \{ \mu 1, \ \mu 2, \ \mu 3 \} with \mu 1(?p) = :Erde, \ \mu 1(?s1) = :Mond \ \mu 2(?p) = :Mars, \ \mu 2(?s1) = :Phobos \ \mu 3(?p) = :Mars, \ \mu 3(?s1) = :Deimos Compatible! \Omega 2 = \{ \mu 4, \ \mu 5, \ \mu 6 \} with \mu 4(?p) = :Erde, \ \mu 4(?s2) = :Mond \ \mu 5(?p) = :Mars, \ \mu 5(?s2) = :Phobos \ \mu 6(?p) = :Mars, \ \mu 6(?s2) = :Deimos
```

Only the compatible mappings are part of  $\Omega$ :

```
\Omega = \{ \mu 14, \dots \} with \mu 14(?p) = :Erde, \ \mu 14(?s1) = :Mond, \ \mu 14(?s2) = :Mond
```

**Solution (Cont):** Now we compute  $\Omega = Join(\Omega 1, \Omega 2)$ , with:

Only the compatible mappings are part of  $\Omega$ :

```
\Omega = { \mu14, \mu25, ... } with \mu14(?p) = :Erde, \mu14(?s1) = :Mond, \mu14(?s2) = :Mond \mu25(?p) = :Mars, \mu25(?s1) = :Phobos, \mu25(?s2) = :Phobos
```

**Solution (Cont):** Now we compute  $\Omega = Join(\Omega 1, \Omega 2)$ , with:

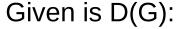
Only the compatible mappings are part of  $\Omega$ :

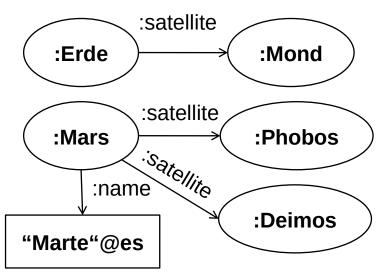
```
Ω = { μ14, μ25, μ26, ... } with μ14(?p) = :Erde, μ14(?s1) = :Mond, μ14(?s2) = :Mond μ25(?p) = :Mars, μ25(?s1) = :Phobos, μ25(?s2) = :Phobos μ26(?p) = :Mars, μ26(?s1) = :Phobos, μ26(?s2) = :Deimos
```

**Solution (Cont):** Now we compute  $\Omega = Join(\Omega 1, \Omega 2)$ , with:  $\Omega 1 = \{ \mu 1, \mu 2, \mu 3 \}$  with  $\mu 1(?p) = :Erde, \ \mu 1(?s1) = :Mond$  $\mu 2(?p) = :Mars, \mu 2(?s1) = :Phobos$  $\mu 3(?p) = :Mars, \mu 3(?s1) = :Deimos$  $\Omega 2 = \{ \mu 4, \mu 5, \mu 6 \}$  with Compatible!  $\mu 4(?p) = :Erde, \mu 4(?s2) = :Mond$  $\mu 5(?p) = :Mars, \mu 5(?s2) = :Phobos$  $\mu6(?p) = :Mars, \mu6(?s2) = :Deimos$ Only the compatible mappings are part of  $\Omega$ :  $\Omega = \{ \mu 14, \mu 25, \mu 26, \mu 35, ... \}$  with  $\mu 14(?p) = :Erde, \ \mu 14(?s1) = :Mond, \ \mu 14(?s2) = :Mond$  $\mu 25(?p) = :Mars, \ \mu 25(?s1) = :Phobos, \ \mu 25(?s2) = :Phobos$  $\mu 26(?p) = :Mars, \mu 26(?s1) = :Phobos, \mu 26(?s2) = :Deimos$ 

 $\mu$ 35(?p) = :Mars,  $\mu$ 35(?s1) = :Deimos,  $\mu$ 35(?s2) = :Phobos

**Solution (Cont):** Now we compute  $\Omega = Join(\Omega 1, \Omega 2)$ , with:  $\Omega 1 = \{ \mu 1, \mu 2, \mu 3 \}$  with  $\mu 1(?p) = :Erde, \ \mu 1(?s1) = :Mond$  $\mu 2(?p) = :Mars, \mu 2(?s1) = :Phobos$  $\mu 3(?p) = :Mars, \mu 3(?s1) = :Deimos$  $\Omega 2 = \{ \mu 4, \mu 5, \mu 6 \}$  with  $\mu 4(?p) = :Erde, \mu 4(?s2) = :Mond | Compatible!$  $\mu 5(?p) = :Mars, \mu 5(?s2) = :Phobos$  $\mu6(?p) = :Mars, \mu6(?s2) = :Deimos$ Only the compatible mappings are part of  $\Omega$ :  $\Omega = \{ \mu 14, \mu 25, \mu 26, \mu 35, \mu 36 \}$  with  $\mu 14(?p) = :Erde, \ \mu 14(?s1) = :Mond, \ \mu 14(?s2) = :Mond$  $\mu 25(?p) = :Mars, \ \mu 25(?s1) = :Phobos, \ \mu 25(?s2) = :Phobos$  $\mu 26(?p) = :Mars, \mu 26(?s1) = :Phobos, \mu 26(?s2) = :Deimos$  $\mu 35(?p) = :Mars, \mu 35(?s1) = :Deimos, \mu 35(?s2) = :Phobos$  $\mu$ 36(?p) = :Mars,  $\mu$ 36(?s1) = :Deimos,  $\mu$ 36(?s2) = :Deimos





Evaluate the following SPARQL algebra expression against D(G):

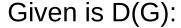
```
JOIN(BGP(P1), BGP(P2))
With
P1 = ?p :satellite ?s1 .
```

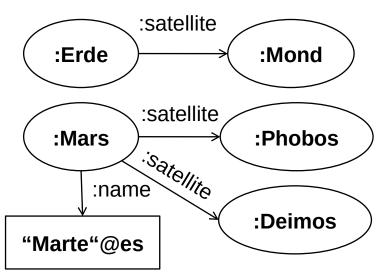
P2 = ?p : satellite ?s2.

**Solution (Cont):** The result of evaluating JOIN(BGP(P1), BGP(P2)) against D(G) is:

```
Ω = \{ μ14, μ25, μ26, μ35, μ36 \} with μ14(?p) = :Erde, μ14(?s1) = :Mond, μ14(?s2) = :Mond μ25(?p) = :Mars, μ25(?s1) = :Phobos, μ25(?s2) = :Phobos μ26(?p) = :Mars, μ26(?s1) = :Phobos, μ26(?s2) = :Deimos μ35(?p) = :Mars, μ35(?s1) = :Deimos, μ35(?s2) = :Phobos μ36(?p) = :Mars, μ36(?s1) = :Deimos, μ36(?s2) = :Deimos
```

## **Example: Evaluating a FILTER Expression (1)**





Evaluate the following SPARQL algebra expression against D(G):

**Solution:** We have to compute  $\Omega = \text{eval}(D(G), \text{FILTER}(E, \text{JOIN}(BGP(P1), BGP(P2)))).$ 

```
Ry definition of eval for FTI TFR expressions, we obtain that eval(D(G), FILTER(expr, P_1)) := Filter(expr, eval(D(G), P_1))
```

We have to compute  $\Omega1=eval(D(G), JOIN(BGP(P1), BGP(P2)))$  and then apply Filter of the expression E over the results  $\Omega1$ .

## **Example: Evaluating a FILTER Expression (2)**

Solution (Cont): We computed  $\Omega$ =eval(D(G), J0IN(BGP(P1), BGP(P2))), in the previous example:

```
\Omega = \{ \mu 14, \mu 25, \mu 26, \mu 35, \mu 36 \} with \mu 14(?p) = :Erde, \mu 14(?s1) = :Mond, \mu 14(?s2) = :Mond \mu 25(?p) = :Mars, \mu 25(?s1) = :Phobos, \mu 25(?s2) = :Phobos \mu 26(?p) = :Mars, \mu 26(?s1) = :Phobos, \mu 26(?s2) = :Deimos \mu 35(?p) = :Mars, \mu 35(?s1) = :Deimos, \mu 35(?s2) = :Phobos \mu 36(?p) = :Mars, \mu 36(?s1) = :Deimos, \mu 36(?s2) = :Deimos
```

Now we have to compute  $\Omega$ =**Filter(E, \Omega)**, with **E= (?s1!=?s2).** For each solution mapping  $\mu$  in  $\Omega$ , we check if **?s1!=?s2**. The solution mappings  $\mu$  for which  $E(\mu)$  evaluates to TRUE are part of  $\Omega$ .

### **Example: Evaluating a FILTER Expression (3)**

**Solution (Cont)**: We computed  $\Omega = \text{eval}(D(G), J0IN(BGP(P1), BGP(P2))}$ , in the previous example:  $\Omega = \{ \mu 14, \mu 25, \mu 26, \mu 35, \mu 36 \}$  with  $\mu 14(?p) = :Erde, \mu 14(?s1) = :Mond, \mu 14(?s2) = :MonfαLSE \mu 25(?p) = :Mars, \mu 25(?s1) = :Phobos, \mu 25(?s2) = :PhofasE \mu 26(?p) = :Mars, \mu 26(?s1) = :Phobos, \mu 26(?s2) = :Deimos \mu 35(?p) = :Mars, \mu 35(?s1) = :Deimos, \mu 35(?s2) = :PhofasE \mu 36(?p) = :Mars, \mu 36(?s1) = :Deimos, \mu 36(?s2) = :Deimos \mu$ 

Now we have to compute  $\Omega$ =**Filter(E, \Omega)**, with **E= (?s1!=?s2).** For each solution mapping  $\mu$  in  $\Omega$ , we check if **?s1!=?s2**. The solution mappings  $\mu$  for which  $E(\mu)$  evaluates to TRUE are part of  $\Omega$ .

```
\Omega = \{ \mu 26, \mu 35 \} \text{ with } \\ \mu 26(?p) = :Mars, \mu 26(?s1) = :Phobos, \mu 26(?s2) \\ = :Deimos \\ \mu 35(?p) = :Mars, \mu 35(?s1) = :Deimos, \mu 35(?s2) \\ \Delta Deimos = :Deimos = :Deimos
```

<sup>67</sup>= : Phobo S<sup>ntroduction to Linked Data – Chapter 5: SPARQL Query Processing</sup>

### **Agenda**

- 1. Basic Graph Pattern Matching
- 2. Translating Queries to Abstract Queries
- 3. Evaluating Graph Pattern Algebra Expressions
- 4. Processing Abstract Queries

## **SPARQL Query Processing: End-to-End**

- Evaluating algebra expressions via eval() is at the core of the definitions around SPARQL query processing.
- However, getting the solution sequence Ω is only a part of query processing.
- Other features need consideration, for example the implementation of BGP matching (the definition 14 of BGP matching was not an operational one).
- Handling of RDF datasets is another feature, also handling of GRAPH.
- We also have not considered solution modifiers (LIMIT, ORDER BY...)
- Finally, the final results of the have to be serialised, e.g., as TSV for SELECT queries.

# **User Agent Algorithm for SELECT Query Processing**

- input: SPARQL SELECT query *q*
- **output**: solution sequence with result to *q*
- set default := URIs in FROM clause of q
- set named := URIs in FROM NAMED clause of q
- <sup>5</sup> RDF dataset D := access(default, named)
- SPARQL algebra expression P := translate(q)
- solution sequence  $\Omega := \mathsf{eval}(D, P)$
- $\Omega := \text{project out selected variables from } \Omega$
- $_{ extsf{9}}$  return  $\Omega$

## Recap (1)

#### Pattern Matching:

- Pattern matching: How to match triple patterns against an RDF graph. Considers the lexical space of RDF literals (exact match)
- The SPARQL algebra:
  - Solution mappings (μ): from variables to RDF terms
  - $\blacksquare$  RDF instance mappings ( $\sigma$ ): from blank nodes to RDF terms
  - Solution sequences  $(\Omega)$ : set of solution mappings
  - Algebra operators to combine sets of solution mappings
- Translating SPARQL queries into algebra expressions:
  - Algebra expressions of SPARQL graph patterns: to translate the conditions specified in the WHERE clause to algebra expressions
  - Handling of SPARQL query forms: to translate the query form (SELECT, ASK, CONSTRUCT)

# Recap (2)

- Evaluating Algebra Expressions of SPARQL Graph Patterns
  - The eval() function for algebra expressions of SPARQL graph patterns
  - eval() for FILTER expressions considers the value space of RDF literals
- Processing Abstract Queries
  - Practical considerations for user agents that process queries
- We left out a lot of SPARQL features (intentionally!)

### **Learning Goals**

- G 5.1 Explain the formal definition of basic graph pattern matching show how to match basic graph patterns to graphs.
- G 5.2 Translate a given SPARQL WHERE clause, including UNION, OPTIONAL, FILTER and BIND AS clauses, to a SPARQL algebra expression.
- G 5.3 Given a query, explain the handling of the RDF dataset with FROM and FROM NAMED clauses in conjunction with GRAPH.
- G 5.4 Evaluate a SPARQL algebra expression on a given RDF dataset and specify the solutions of the entire algebra expression and also of partial expressions.
- G 5.5 Generate the final results to a SPARQL abstract query, taking into account the solution sequence of the graph pattern algebra expression and the query form.

### **Outlook – Chapter 6**

- Chapter 6 is concerned with building data integration systems.
- Many organisations (EU, countries, cities) provide data under open licenses.
- We consider what architectures are suitable for accessing Linked Data published under open licenses on the web.



https://5stardata.info/en/