Interactive Knowledge Graph Querying through Examples and Facets





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Motivating Example

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Lucy is a criminologist looking for insights in large knowledge graphs.

I am looking for interesting criminal profiles, but where to start?

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"Marrying" Examples and Facets

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Model

Facts as RDF triples:

Saddam_Hussein type Leader. Saddam_Hussein birthYear 1937. \in Ent \in Pred \in Ent/Lit

```
Basic graph patterns (BGPs):
```

?x	type	?у.	?x		birthYear 1937.	
<pre>sSPARQL (SPARQ SELECT DISTI WHERE {?e ty MINUS {?e de MINUS {?e co</pre>	<u>Lfragment,</u> NCT ?e pe Person athPlace nvictedOf	simplified . ?e co []} WarCri): nvictedOf mes}.}	[]	stands for an undistinguished variable, allow any value to be assigned to i	ring t

User Interaction

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sSPARQL (SPARQL fragment, simplified):

SELECT DISTINCT ?e

- WHERE {?e type Person. ?e convictedOf []
- MINUS {?e deathPlace []}

MINUS {?e convictedOf WarCrimes}.}



Saddam Hussein is not an answer

No matching

answers

Silvio Berlusconi

and

Angela Merkel

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"Negative" patterns must never hold.

• Closed world:

"positive" patterns must hold

 Step-wise semantics: Angela Merkel maximal subset of "positive" patterns must hold

• Weighted semantics:

maximal subset of "positive" patterns with maximal weight must hold Only Silvio Berlusconi Silvio Berlusconi

Open world:

"positive" patterns may or may not hold

Example query

```
SELECT DISTINCT ?e
WHERE {?e type Person. ?e convictedOf []
MINUS {?e deathPlace []}
MINUS {?e convictedOf WarCrimes}.}
```

Example Knowledge Graph:

Saddam_Hussein deathPlace Baghdad. Saddam_Hussein type Person. Silvio_Berlusconi convictedOf Fraud. Angela_Merkel type Person.

Multiple Semantics - Results

- **Proposition** [containment between semantics]: for any Q and G, it holds that $Q_{Closed-world}(G) \subseteq Q_{weighted}(G) \subseteq Q_{step-wise}(G) \subseteq Q_{open-world}(G)$
- **Proposition** [monotonicity]: for any sequence of questions and answers, it holds for the queries that encode them that $Q_{closed-world}^{i+1}(G) \subseteq Q_{closed-world}^{i}(G)$ and $Q_{open-world}^{i+1}(G) \subseteq Q_{open-world}^{i}(G)$,

i.e., answer candidate set monotonically decreases

• **Proposition** [encoding in closed-world semantics]: for every sSPARQL query Q and a KG G, there exist sSPARQL queries Q^1, Q^2, Q^3 such that

$$- Q_{step-wise}(G) = Q^{1}_{Closed-world}(G)$$

 $- Q_{weighted}(G) = Q_{Closed-world}^{2}(G)$

$$- Q_{open-world}(G) = Q^3_{Closed-world}(G)$$

Problem definition

For a single question: choose a question $q_{p,o}$ that maximizes the number of eliminated candidate answers

$$E[|E_{\varphi'} - E_{\varphi}|]$$

= $Pr["must"|q_{p,o}]|E_{\varphi} - E_{p,o}| + Pr["must not"|q_{p,o}]|E_{\varphi} \cap E_{p,o}|$

- Depends on the semantics
- In the paper: deriving answer probabilities
- In the paper: selecting an example entity with multiple properties

Summary

- An interactive framework for knowledge graph exploration
- Selecting example entities with example properties to obtain feedback
 - Adding constraints to an sSPARQL query
- Multiple semantics for dealing with incompleteness and avoiding "dead-ends"
- See the paper for more details





Lord George Gordon

