Question Answering on Knowledge Graphs

Part 1.3 of the Tutorial
"Neuro-Symbolic Representations
for Information Retrieval"
ECIR 2023, Dublin (April 6)

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Knowledge Graphs 1/8

L D E L D E

- Resource Description Framework
 - RDF is based on a strikingly simple principle: ALL DATA, whatever it is, is modelled as a **set of triples** subject predicate object
 - For example:

```
<Dublin>
                <capital of>
                                  <Ireland>
<Dublin>
                 <coordinates>
                                  "53°21'N 6°15'E"
<Ireland>
                 <language>
                                  <English>
                                                      Dublin
<Ireland>
                 <language>
                                  <Gaelic>
                                                        53°21'N 6°15'E
                                              Ireland
Graph view: each distinct subject or object
is a node, each triple is a directed edge from
                                                 English
                                                           Gaelic
subject to object, with the predicate as label
```

Knowledge Graphs 2/8

wdt:P17 ~ country wd:Q27 ~ Ireland wdt:P625 ~ coordinates

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- International Resource Identifier (IRI)
 - Names have globally unique identifiers, so-called IRIs

```
An IRI is like an address in a browser (URL), only that Unicode characters like ä, ö, ü, ... are allowed
```

- The IRI is often not human-readable but ID-like; instead there are dedicated triples for its names and aliases
- For example, the Wikidata IRI for Dublin is

```
<a href="https://www.wikidata.org/wiki/Q1761">https://www.wikidata.org/wiki/Q1761">
```

And some example triples (with **IRI prefixes**) look like this:

```
wd:Q1761 rdfs:label "Baile Átha Cliath"@ga
```

wd:Q1761 **wdt:**P17 **wd:**Q27

wd:Q1761 wdt:P625 "Point(53.35 -6.26)"^^geo:wktLiteral

Knowledge Graphs 3/8

- Some widely used knowledge graphs
 - Wikidata: Started 2012, successor of Freebase (bought by Google in 2010 for 99M\$), crowd-sourced, amazing coverage
 - 18 **B** triples, 99 M entities, 10K predicates
 - UniProt: Started 2002, protein sequences, genes, all kinds of metadata, ...
 - 94 **B** triples, 16 B entities, 520 predicates
 - PubChem: Started 2004, chemical compounds, substances, proteins, genes + interrelations
 - 124 **B** triples, 4 B entities, 414 predicates







Knowledge Graphs 4/8

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- And more ...
 - OpenStreetMap: Started 2004, all the geo-data of the world, crowd-sourced, amazing coverage and quality
 - 11 B triples, 1.1 B entities, 87K predicates
 - DBLP: Started 1993, publication meta-data for computer science and adjacent fields
 1 B triples, 150 M entities, 75 predicates





Many more datasets available as RDF, of high quality and maintained with great care

Knowledge Graphs 5/8

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- Historical knowledge graphs
 - Freebase: Started 2007 by Metaweb,
 acquired by Google in 2010, closed down
 in 2015 (migrated to Wikidata)



- 3.1 B triples, 47 M entities, 785 K predicates
- YAGO: first released 2008 at the MPI for Informatics, derived from Wikipedia infoboxes combined with WordNet



121 M triples, 15 M entities, 100 predicates



 DBpedia: all kinds of structured data extracted from Wikipedia

These are still mentioned in a lot in papers (because benchmarks were based on them)

Knowledge Graphs 6/8

- SPARQL
- The standard query language for RDF is SPARQL
 - SPARQL is a variant of SQL, adapted to the (simple) RDF

```
SELECT ?title ?author ?year WHERE {
    ?paper dblp:title ?title .
    ?paper dblp:authoredBy ?author.
    ?paper dblp:yearOfPublication ?year .
    FILTER (?year <= 1940)
}</pre>
```

All papers in DBLP published before 1940

Query on QLever

– The result of a SPARQL query is always a table, in the example:

All assignments to ?title ?author ?year such that the triples exist in the knowledge graph and the FILTER condition is true

Note: if a paper has **k** authors, there will be **k** rows for it

Knowledge Graphs

wd:Q183 ~ Germany osmrel:51477 ~ Germany

wd:P37 ~ population

wdt:Q9142 ~ Gaelic

Interoperability

 One of the great strengths of RDF and SPARQL is the great ease, with which different datasets can be combined

For standard databases, this is a nightmare: different and often complex schemas, different identifiers, etc.



In RDF, all you need are additional triples relating the IDs from the two datasets you want to combine



osmrel:62773 osm:wikidata **wd:**Q27

osmrel:62773 geo:hasGeometry "MULTIPOLYGON(...)"^^geo:...

wdt:Q9142 **wd:**Q27 wdt:P37

The first two triple are from OSM, the third is from Wikidata (OSM knows the geometric shapes, Wikidata knows the language)

Knowledge Graphs 8/8



A variety of example queries

Birth places of people with first name X
 Wikidata

Notable events that happened on April 6

Wikidata

Genes for human diseasesUniProt

NSAID drugs with small molecular weight
 PubChem

All streets in Ireland
 OpenStreetMap

All countries with official language X
 Wikidata+OSM

Average number of authors per year

DBLP

Side note: These are all running with the same system, with very little configuration per dataset, that is the power of RDF+SPARQL

SPARQL Autocompletion 1/4



- Finding the right SPARQL query is hard
 - Consider the following simple search request
 Oscars of Meryl Streep with corresponding movie
 - The result we are looking for is

Academy Award for Best Supporting Actress Kramer vs. Kramer Academy Award for Best Actress Sophie's Choice Academy Award for Best Actress The Iron Lady

 On the next slide, you see the correct SPARQL query on Wikidata

SPARQL Autocompletion 2/4

```
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```

```
PREFIX rdfs: <a href="http://www.w3.org/2000/01/rdf-schema#">http://www.w3.org/2000/01/rdf-schema#</a>
PREFIX wdt: <http://www.wikidata.org/prop/direct/>
PREFIX pq:
               <a href="http://www.wikidata.org/prop/qualifier/">http://www.wikidata.org/prop/qualifier/>
PREFIX ps: <http://www.wikidata.org/prop/statement/>
PREFIX p: <a href="http://www.wikidata.org/prop/">
PREFIX wd: <http://www.wikidata.org/entity/>
SELECT ?movie ?award WHERE {
 wd:Q873
              p:P166 ?statement .
 ?statement ps:P166 ?award id .
 ?statement pq:P1686 ?movie_id .
 ?award_id wdt:P31 wd:Q19020 .
 ?award_id rdfs:label ?award . FILTER (LANG(?award) = "en")
 ?movie_id rdfs:label ?movie . FILTER (LANG(?movie) = "en")
```

SPARQL Autocompletion 3/4



- What's hard about finding this query?
 - Knowing the right prefix definitions

```
PREFIX rdfs: <a href="http://www.w3.org/2000/01/rdf-schema">http://www.w3.org/2000/01/rdf-schema#>
```

Knowing the right entity names

```
wd:Q873 ("Meryl Streep"), wd:Q19020 ("Academy Award")
```

Knowing the right predicate names very hard, even for experts

```
p:P166 "won award" from awardee to statement
ps:P166 "won award" from statement to award entity
pq:P1686 "for work" from statement to movie
wdt:P31 "instance of" from "instance" entity to "class" entity
```

Knowing the syntax for filtering by language

```
FILTER (LANG(?award) = "en")
```

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SPARQL Autocompletion 4/4

- Let's look at two query building tools
 - Tool 1: Wikidata's own query builder
 - A simple (context-insensitive) autocompletion
 - Requires **deep expert knowledge** of the prefixes and the right predicate names
 - Tool 2: QLever's query builder
 - Context-sensitive suggestions as you type
 - Requires only relatively basic knowledge of RDF/SPARQL, and little or no knowledge about the dataset
 - Fun fact: the suggestions are themselves computed via SPARQL queries, on the same knowledge graph on which we are trying to formulate a query

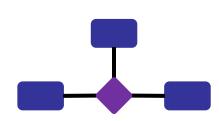
QA on Knowledge Graphs 1/6

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Goal

- Ask questions in **natural language**:
 Which character did Ellen DeGeneres play in Finding Nemo?
- Or even more informally / telegraphically:who did ellen play in finding nemo?
- Goal: automatically translate such a natural-language or keyword question into the corresponding SPARQL query

```
SELECT ?x WHERE {
  ?m actor Ellen DeGeneres .
  ?m film Finding Nemo .
  ?m character ?x
}
```



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Natural-Language Queries 2/6

- Challenge 1: Linguistic variation
 - The same question can be asked in dozens of ways: which character did ellen degeneres play in finding nemo which character did ellen play in finding nemo who did ellen play in finding nemo ellen's role in finding nemo whose voice did ellen do in finding nemo role ellen nemo

This rules out simple pattern-based approaches

Natural-Language Queries 3/6



- Challenge 2: Ambiguous entity names
 - Ellen could mean

Ellen DeGeneres

Ellen Page

Ellen Burstyn

anyone called "Ellen"

The Ellen Show

The Ellen DeGeneres Show

. . .

Over 100 different entities named "ellen" in Freebase

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Natural-Language Queries 4/6

- Challenge 3: Ambiguous relation names
 - Like for entity names, but worse, because the relation can be implicit in the question, for example:

```
Question: who is the ceo of apple

Query: SELECT ?x WHERE {
    ?m job-title "Managing Director".
    ?m company "Apple Inc.".
    ?m person ?x.
}
```

None of the relation words "job title", "company", "person" appear in the question ... nor synonyms of them

Natural-Language Queries 5/6

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- General Approach 1: Generate candidate +rank them
 - Step 1: Entity linking in the query
 - No need to commit to a perfect solution at this stage, having multiple candidates is OK
 - Step 1: Generate candidate queries
 Generate all possible SPARQL queries that connect those entity (candidates)
 - Step 3: Rank these candidates by their similarity to the question using a learned model
 - Can be trained only from question-answer pairs (no need to have also have the correct SPARQL query for each question)

Natural-Language Queries 6/6



- General Approach 2: Use a language model
 - Step 1: Write a prompt

Write a SPARQL query to the Wikidata Query Service for the following question: Which Oscars did Meryl Streep win and for which movie?

Step 2: Send the query

Send the query to the respective API

For example: OpenAI Codex

A descendant of GPT-3 trained on both natural language and code, see https://openai.com/blog/openai-codex

References

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Links

- https://query.wikidata.org
- https://github.com/ad-freiburg/qlever
- https://qlever.cs.uni-freiburg.de
- https://aqqu.cs.uni-freiburg.de

Papers

TODO: will be added after the tutorial