

# KNOWLEDGE GRAPHS

Lecture 12: Knowledge Graph Quality

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TU Dresden, 14th Jan 2020

# Review: The Cypher Query Language

```
Example: Find pairs of siblings:
MATCH
  (parent)-[:HAS_CHILD]->(child1),
  (parent)-[:HAS_CHILD]->(child2)
WHERE id(child1) <> id(child2)
RETURN child1, child2
```

#### **Cypher features**

- · Graph patterns that can be matched to (sets of) paths in graphs
- Querying for and counting of paths
- Various filter conditions (WHERE)
- Expressions for accessing aspects of the Property Graph model
- · Grouping and aggregates, subqueries, unions, optional matches

# Knowledge Graph Quality

# Motivation

The quality of KGs is an aspect of utmost importance:

- Almost any data can be turned into a graph format and called "knowledge graph"
- · When switching to graphs, we need to know if it was a success or failure
- As time passes, the quality of the KG must be monitored

What does quality mean in this context? How can it be measured? How can it be monitored automatically?

# Can quality be measured?

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#### However, the world is not that simple:

- Quality is multi-dimensional and non-linear
- Usually not metric ("We improved by 12.6%") but merely ordinal ("We improved a lot")
- Relationship to (metric) cost/revenue numbers all but clear

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#### Functional and non-functional requirements are rarely independent

# Example: TimBL's Open Data Quality proposal

Tim Berners-Lee in 2006 proposed a 5-star quality metric for open data:

- $\star$  make your stuff available on the Web (whatever format) under an open license
- ★ ★ make it available as structured data (e.g., Excel instead of image scan of a table)
- \* \* \* make it available in a non-proprietary open format (e.g., CSV instead of Excel)
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#### Notes:

- \* is not about data quality
- All other criteria are non-functional
- The criteria are ordinal, not metric, and no means of estimating partial progress is given (esp. for \* \* \* \* \*)
- The term Linked Open Data refers to 5-★ RDF data with resolvable URIs

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But there are important differences:

- development process (many editors, often weakly coordinated; data imports; unclear development life cycle)
- lack of modularisation/interfaces/separation of concerns (integration vs. separation of knowledge)
- KGs have no fully self-contained function: they need to be used by some software
- KGs may be intended for multiple or yet unknown functions

And measuring software quality is already a difficult issue ....

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#### Way forward:

- We will focus on methods of checking specific quality criteria
- The outcomes of many checks can be quantified to obtain measures
- One can aggregate measures by (subjective) weighting functions, or analyse them as multi-dimensional aspects of the KG's status

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  - Operational: Ad hoc implementation of quality checks

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 Declarative: Specification of quality criteria in some formal language that can be interpreted by (standard) tools

**Example:** Schema document that constrains syntactic form of KG.

### Declarative quality checks for KGs

#### The distinction between "operational" and "declarative" is often fuzzy

- A testing script is operational: implementation-specific meaning; not portable
- A schema document in a standard language (e.g., XML Schema) is declarative: meaning standardised and understood by many tools
- Many other approaches are in between:
  - Graph database constraints: possibly proprietary language; may or may not be portable
  - Wikidata property constraints: community-developed approach for expressing schema information in data
  - SPARQL test queries: declarative queries used in some operational wrapper that validates results
  - Business rules: rule-based programs that interpreted by proprietary software

- ...

#### $\rightsquigarrow$ declarativity is not an rigorously defined feature, but an ideal to strive for

# Competency questions

A classical approach of knowledge model evaluation are so-called competency questions

**Definition 12.1:** A competency question is a (usually application-related) question towards the KG that is formalised in a query language, together with a formal specification of how an acceptable answer may look.

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**Definition 12.1:** A competency question is a (usually application-related) question towards the KG that is formalised in a query language, together with a formal specification of how an acceptable answer may look.

A competency question does not need to pre-determine the KG data in all detail.

**Example 12.2:** We can specify that Wikidata should "know" that humans (Q5) are mammals (Q7377) by requiring that the query SELECT \* WHERE { wd:Q5 wdt:P171\* wd:Q7377 } returns a non-empty result ("true"). This query leaves empty how the taxonomic hierarchy is modelled.

### Competency questions

Competency questions focus on functional metrics:

- coverage/completeness (but cannot check all cases)
- correctness
- accessibility (using query answering software)

They can be used in several situations:

- To define the initial scope (requirements) of a new KG project
- To formalise data modelling decisions (how should knowledge be encoded to be accessible)
- For regression testing (ensure that KG does not break in the future)

However, there are also costs: modelling effort, maintenance, ...

Competency questions take a content-oriented view (application- and domain-specific), but the approach can be generalised to set up unit testing:

- Define a test suite of queries + (constraints on) expected answers
- Automatically run queries to detect problems

Unit tests can also validate non-functional criteria.

### Schema languages

The most formal way of defining quality criteria is by specifying structural requirements in a formal schema language

**Example 12.3:** XML Schema is a classical schema language to constrain the syntactic form of XML documents (DTD is another, older, approach with similar goals).

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For RDF, there are mainly two schema languages today:

- SHACL, a W3C standard (since 2017)
- ShEx, a W3C member submission and community group effort

**Note:** RDF Schema, despite its name, is a lightweight ontology language rather than a schema language.

# SHACL

SHACL is the W3C Shapes Constraint Language

#### **Basic principles:**

- Overall approach similar to query-based unit testing
- SHACL shapes specify constraints by defining two aspects:
  - (1) a pattern that RDF graph a nodes may match (akin to simple queries),
  - (2) and the set of target nodes that should match the pattern
- SHACL-SPARQL extension allows using SPARQL for pattern specification
- Shapes can have meta-data to define, e.g., error messages and severity levels
- Shapes and sets of shapes are encoded in RDF as shape graphs

### Further reading:

- W3C SHACL Recommendation at https://www.w3.org/TR/shacl/
- Labra Gayo, Prud'hommeaux, Boneva, Kontokostas: Validating RDF Data (Morgan Claypool 2018); see https://book.validatingrdf.com/

# SHACL by Example

The following RDF graph (in Turtle, without prefixes) defines a shape ex:PersonShape:

```
ex:PersonShape rdf:type sh:NodeShape ;
 sh:targetClass ex:Person ; # Applies to all persons
 sh:property [ # Declare a constraint on property usage
   sh:path ex:ssn ; # ... for property ex:ssn
   sh:maxCount 1 ; # ... at most one value
   sh:datatype xsd:string ; # ... having type string
   sh:pattern "^\\d{3}-\\d{2}-\\d{4}$" # ... matching this regexp
 ];
 sh:property [ # Declare another property constraint
   sh:path ex:worksFor ; # ... for property ex:worksFor
   sh:nodeKind sh:IRI ; # ... values are IRIs
   sh:class ex:Company # ... of rdf:type ex:Company (or a subclass)
 1:
 sh:closed true ; # No other properties are allowed
 sh:ignoredProperties ( rdf:type ) . # ... except for rdf:type
```

# Shapes in SHACL

#### ... are identified by IRIs and may optionally include:

- a specification of target nodes they apply to (sh:targetClass, sh:targetSubjectsOf, sh:targetObjectsOf, or sh:targetNode)
- a set of property shapes that define constrains on values reached through (paths of) properties
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#### There is a rich vocabulary for specifying property constraints, including:

- (SPARQL-like) property paths instead of single properties
- minimal and maximal cardinalities
- resource types, datatypes, or RDF classes for values
- lists of admissible values
- ways to say that one property's values are disjoint or equal to another's
- (possibly recursive) reference to the NodeType of property values
- Boolean combinations of constraints (and, or, not)

# ShEx

ShEx is the Shape Expressions Language as proposed by a W3C Community Group

#### **Basic principles:**

- Overall approach similar matching a grammar description to a graph
- ShEx shapes specify constraints by defining a pattern that refers to
  - (1) required features of the RDF graph
  - (2) required patterns matched by adjacent nodes (recursively)
- Validation tries to consistently map nodes in an RDF graphs to types as required (based on some initial map)
- Sets of shapes form a schema, encoded in an RDF-inspired own syntax

#### Further reading:

- ShEx community homepage at http://shex.io/
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# ShEx by Example

The following defines a shape a:PersonShape (without prefix declarations), which is functionally equivalent to the previous SHACL example:

```
a:PersonShape CLOSED EXTRA rdf:type {
    ex:ssn xsd:string /^\\d{3}-\\d{2}-\\d{4}$/?;
    ex:worksFor IRI @a:CompanyShape *;
}
a:CompanyShape [ ex:Company ] OR { rdfs:subClassOf @:CompanyShape }
```

#### Notes:

- CLOSED and EXTRA play the role of sh:closed and sh:ingoredProperties
- "property shapes" are compactly expressed in single lines
- taking indirect typing (instances of subclasses) into account requires the use of recursive shape definitions

As for SHACL, there are many further features. For example, shape expressions can be combined with boolean operators AND, OR, and NOT.

Markus Krötzsch, 14th Jan 2020

# Validating SHACL and ShEx

#### Both approaches support recursive constraints and disjunctions:

- Node types are not part of the data: their recursive use means that validation has to extend shapes to new target nodes
- Disjunction means that this assignment might be non-determinstic
- → worst-case NP-complete complexity in the size of the graph (data complexity!)

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#### However:

- In SHACL, recursive assignments are a minor feature, and the specification does not specify their semantics. Selection of target nodes mostly governed by conditions on RDF.
- In ShEx, type maps are the only mechanism for selecting targets, and recursive assignments are necessary to check indirect class membership
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 Since SHACL is mostly deterministic, it can also provide detailed error reports in case of failing constraints (this is hard if many assignments need to be considered which may all fail, but for different reasons)

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 Knowledge Graphs
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Defining and measuring knowledge graph quality is difficult; there are many criteria

Competency questions and unit tests are basic approaches for automatic quality checks

RDF constraint languages like SHACL and ShEx can declaratively specify constraints

#### What's next?

- Ontological modelling
- Graph analysis
- Examinations