A Practical Framework for Evaluating the Quality of Knowledge Graph

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Presentation

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Structure

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Introduction

Knowledge graphs

- A graph representation of knowledge using entities, edges, and attributes.
- Entities represent real-world objects; edges represent relationships; attributes define entities.

Importance of knowledge graph quality

Quality impacts system effectiveness and usability.

Quality criteria

Correctness, comprehensiveness, and freshness

Limitations of current frameworks

- Some criteria may not be necessary for all applications
- Validation of certain criteria might occur during application development
- Handling low-quality items is often the responsibility of application developers

Purpose

- Identify essential quality requirements for knowledge-based systems
- Maps quality criteria to select metrics that can be used for evaluating basic quality standard of a knowledge graph
- Investigate existing evaluation approaches and tools to propose one practical approach for evaluating

Quality of Knowledge Graph

Quality Metrics for Knowledge Graph

Zaveri's framework

- Based on prior frameworks by Wang & Strong and Stvilia.
- Includes 18 quality dimensions and 69 metrics.
- Categories:
 - Accessibility: availability, licensing, interlinking, security, performance.
 - Intrinsic: syntactic validity, semantic accuracy, consistency, conciseness, completeness.
 - Contextual: relevancy, trustworthiness, understandability, timeliness.
 - Representational: representational conciseness, interoperability, interpretability, versatility.

Examples of metrics:

- Availability: "Accessibility of RDF dumps"
- Security: "Usage of digital signatures", "Authenticity of the dataset."
- Semantic Accuracy: "No outliers", "No inaccurate values", "No misuse of properties."
- Interoperability: "Compliance with standards", "Usability in multiple systems."

Challenges in defining metrics

- No complete set of metrics for every dimension.
- Metrics must be adapted for specific domains or applications.

Approaches and tools for evaluating quality of knowledge graph

Types of evaluation

- Manual
- Automated

Review by Zaveri

- Systematic review of 30 approaches and 12 tools
- Dimensions like syntactic validity and semantic accuracy received significant attention
- Dimensions like security and performance were less emphasized

Evaluation approaches

Test-Driven evaluation

- Generates test cases using data schema constraints, patterns, and requirements
- Checks queried results against specific quality metrics

Sampling for accuracy

Efficiently evaluates accuracy with strong statistical guarantees

A recommended gramework for quality evaluation of knowledge graph

Categories of knowledge graph applications:

- Semantic search
- Decision making
- Knowledge management
- Data mining
- Prediction

Example Use Cases:

Recommendation systems:

- Represent semantic relationships between items.
- Include diverse relationships to enhance recommendation variety.
- Provide user history for explainability.

• Question answering:

- Ambiguity of entity names and partial names.
- Scalability for adding new entities and relationships.

Information retrieval:

Entities associated with various names for better matching.

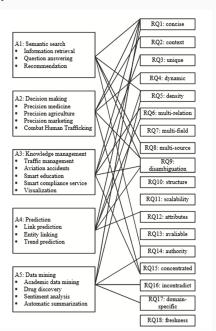
- Key quality dimensions for different applications:
 - Semantic search: Syntactic Validity, Semantic Accuracy, Conciseness
 - Decision making: Trustworthiness, Accuracy, Relevancy
 - Prediction: Accuracy, Timeliness, Consistency

Quality requirements for knowledge graphs

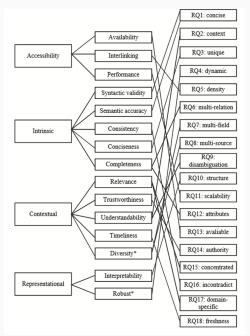
- 1. Triples should be concise.
- 2. Contextual information of entities should be captured.
- 3. Redundant triples should be eliminated.
- 4. Knowledge graph should support dynamic updates.
- 5. Entities should be densely connected.
- Relationships among different types of entities should be included.
- 7. Data sources should span multiple fields.
- 8. Data for constructing the knowledge graph should come from diverse types and resources.
- 9. Synonyms should be mapped, and ambiguities resolved.

- 10. Structured triples should enable easy machine processing.
- 11. Scalability to accommodate large graph sizes.
- 12. Attributes of entities should not be omitted.
- 13. Knowledge graph should be publicly available and proprietary.
- 14. Authority of data should be ensured.
- 15. The graph should remain concentrated on relevant entities.
- 16. Triples should not contradict each other.
- 17. Domain-specific tasks require domain-relevant data.
- 18. Freshness of resources to ensure up-to-date information.

Anowledge graph requirements produced from representative applications



Mapping knowledge graph requirements quality dimensions



Related Work

Wang and Strong (1996)

Proposed a hierarchical framework for data quality dimensions

Stvilia et al. (2007)

- Investigated causes and effects of information quality changes
- Developed a taxonomy of quality dimensions and problems

Zaveri et al. (2016)

- Evaluated DBPedia with a user-driven approach
- Reviewed frameworks for linked data

Paulheim (2017)

 Evaluated methods based on target, method type, evaluation metrics, and computational performance Combines Zaveri's framework with theoretical insights from Stvilia

Focuses on balancing

- Comprehensive evaluation with measurable dimensions.
- Scalability for large knowledge graphs.

Summary and future directions

Findings

Syntax and semantics errors significantly impact the quality of knowledge graphs.

Proposed framework applications

- Basic quality evaluation for knowledge discovery
- Practical for building knowledge-based applications

Future work

Test framework effectiveness on large-scale knowledge graphs