Introduction

Something about me

- 2005 Studies of Computer Science, University Halle (Germany)
- 2010 PhD in Software Engineering and Programming Languages, University Halle (Germany)
- 2012 Project Lead of "Semantic Web Project" (R & D), Unister GmbH (Germany)
- 2015 Head of Research and Development Department, Unister GmbH (Germany)
- 2016 Research and Development Lead Mercateo AG (Germany)
- ☐ 11/2016 Head of Architecture, Web Technology and IT Research, DATEV eG (Germany)

DATEV eG

- Software company and IT service provider
- ☐ turnover: > 900 million euros
- age: > 50 years old
- core market: Germany
- fields: accounting, business consulting, taxation, enterprise resource planning (ERP) as well as organization and planning
- members: > 40.000
- ☐ customers: > 2.6 million companies



Qanary

Several tasks common across QA systems.

Having these tasks as components, sharing a vocabulary can help develop better QA system.

Qanary

A framework for developing QA system by integrating various component using a pre-defined **Q**uestion **An**swering vocabul**ary**.

Observations

- Limited compatibility
- Predefined pipelines
- Limited semantics

- + Interoperable infrastructure
- + Exchangeable components
- + Flexible granularity
- + Isolation of components

Goals

Easy-to-build QA systems on-top of reusable components

Establish an ecosystem of components for QA systems

Qanary Methodology and Technical Framework

Knowledge perspective - I

Requirements of Knowledge perspective

- abstract knowledge representation: qa vocabulary
- align the input/output of the each component in a QA process

QA vocabulary

Represent all the available knowledge about a question.

- representation of knowledge about question separated from process
- ☐ includes trust & provenance
- self-describing, reusable and extensible
- enables efficient collaboration on a data-level
- agnostic to question format (text, structured, audio)
- agnostic to question answering processing steps and implementation

Aligning I/O in QA processes

- ☐ required input mapped from KB
- computed output mapped into KB
- mapping on a logical and sound level

Knowledge Representation using the qa Vocabulary

Abstract Knowledge Representation (KR)

Represent all the knowledge about a question using a RDF vocabulary

KR requirements

- self-describing, sound knowledge representation
- □ represent provenance for (all) information
- represent trust for (all) information

Derived Technology stack

- Resource Description Framework (RDF)
- ☐ Web Annotation Data Model (WADM)
- Question Answering vocabulary (qa)

Resource Description Framework (RDF)

Introduction to RDF (slides by Manolis Koubarakis)

Web Annotation Data Model (WADM)

- oa:Annotation
- oa:hasTarget
- □ oa:hasBody
- oa:annotatedAt
- oa:annotatedBy

```
<myIRI> a oa:Annotation;
    oa:hasTarget <questionIRI> ;
    oa:hasBody <TextSelector> ;
    oa:annotatedBy <DBpediaSpotlight> ;
oa:annotatedAt "..."^^xsd:date ;
```

QA Vocabulary

Introducing new QA-related concepts on-top of WADM

From KR to methodology

Conclusion: Advantages of using an ontology

- □ agnostic to question format (text, structured, audio)
- □ agnostic to question answering processing steps
- agnostic to implementation
 - programming language
 - □ component granularity

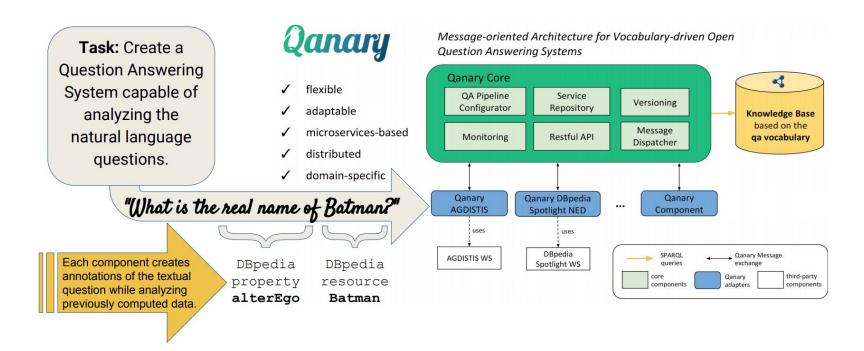
From KR to methodology II

Methodology:

- □ abstract knowledge representation
 - independent representation
- align the input/output of the each component
 - on a logical and sound level

Methodology

Qanary Architecture



It's about the components, stupid

An agile QA framework can only provide common features

central data access, logging

Any particular problem solving/algorithm needs to be separated from the pipeline

It's about the components, stupid -II

create exchangeable, isolated components only communicating via data.

component data needs to be mapped/aligned to the data of the QA process.

Component Data Alignment- CDA

Goal: Establish common ground for the research community

CDA- II normalized exchange of data (input/output) External Component Adapter Encapsulated Functionality Library External Services

CDA III

Alignment of input/output of each component with qa, if component provides output using presentation as

NER/NED DBpedia semantic data (RDF) Spotlight logical representation of Mendes et al., 2011 alignment (NIF) ontology alignment relation detection (OWL, DOL) PATTY [Nakashole SPARQL query et al., 2012 non-semantic data (API, query construction JSON, XML, CSV) SINA [Shekarpouret SPARQL query al., 2011

CDA: NED

Create component input

fetch question URI (from Qanary triplestore)

Processing

- retrieve textual question representation from URI
- compute named entities within the text

store component output

- for each named entity:
 - create a oa:TextSelector within the Qanary triplestore containing the positions of the particular Named Entity

CDA: NED Benefits

Easily replace the NED component.

Measure quality against exchangeable relation detection and query construction components.

CDA: Relation Detection

Create component input

- fetch question URI (from Qanary triplestore)
- fetch Named Entities which are already available

Processing

- retrieve textual question representation from URI
- compute relations within the text

store component output

- for each named relation:
 - create a relation resource within the Qanary triplestore (using a oa:TextSelector to mark the positions)

CDA: Relation Detection Benefits

Any improvement on the NED component (i.e., replace) will improve the quality here

Measure quality against exchangeable query construction components.

CDA: Query Construction

Create component input

- fetch Named Entities (which are already available)
- fetch Relations (which are already available)

Processing

compute SPARQL

Store component output

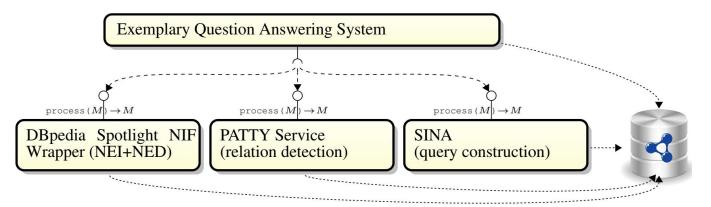
- for each created SPARQL:
 - store a resource/SPARQL in the Qanary knowledge base

CDA: Query Construction Benefits

Any improvement on the NED component (i.e., replace) will improve the quality here

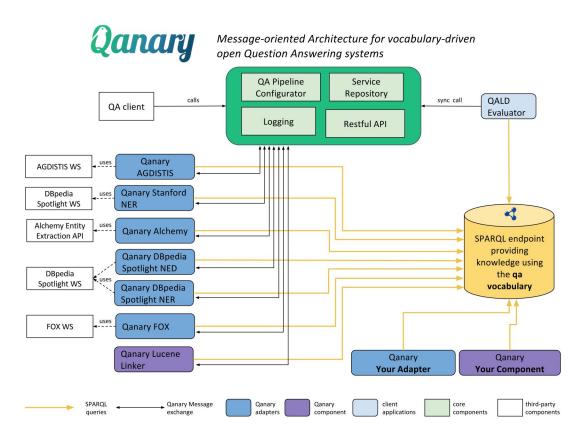
Any improvement on the Relation detection component (i.e., replace) will improve the quality here

Case Study



Component	Process within components
DBpedia Spotlight	Retrieve data from KB
PATTY	Process data
SINA query execution	Extend KB

CDA-II



Coffee Break

In the following practical session you will need:

- Internet connection
- text editor or any Ontology Designer
- Git client
- Java and Maven
- Stardog triplestore (free version)

Our Goal

Qanary Message-oriented Architecture for Vocabulary-driven Open Task: Create a Question Answering Systems **Question Answering Qanary Core 4** flexible System capable of **QA** Pipeline Service Versioning Configurator Repository adaptable analyzing the **Knowledge Base** based on the microservices-based Message natural language Restful API Monitoring qa vocabulary Dispatcher distributed questions. domain-specific Qanary Qanary DBpedia "What is the real name of Batman?" Qanary ••• **AGDISTIS** Spotlight NED Component uses uses **DBpedia SPARQL** Qanary Message **AGDISTIS WS** Each component creates DBpedia DBpedia queries exchange Spotlight WS annotations of the textual core Qanary third-party property resource components adapters components question while analyzing alterEgo Batman previously computed data.

Preparation

Given questions:

- "Name all the movies in which Robert Downey Jr acted?"
- "Is Bruce Wayne the real name of Batman?"
- "How many partners had Batman?"

Preparation - Task

Model the required annotations for answering these questions

Write the SPARQL query to retrieve the answers for these queries

Note: Typically, the result of a QA process is not a SPARQL query. Due to time constraints, we exclude the mostly following Answer Generation (e.g., using Natural Language Generation or visualizations) from this exercise. See wolframalpha.com from inspiration

Coding Session: Implement your first QA system from existing components

Implement your first QA system

https://github.com/WDAqua/Qanary/wiki/Demo:-How-to-Create-a-Question-Answering-System-capable-of-Analyzing-the-Question-%22What-is-the-real-name-of-Batman%3F%22

Implement your first QA system

- ☐ Git checkout Qanary ecosystem's components
- Run components
- ☐ Run Qanary QA system template
- Configure your pipeline
- ☐ Run the pipeline
- ☐ Test your QA system with some questions on DBpedia
- Done

Validate the quality of your QA system

Implement your first QA system

Interactive validation using TRILL front-end from Qanary Ecosystem

Use Qanary QALD validator to compute precision, recall and f-measure

Improve and revalidate your QA system

Improve and revalidate your QA system

Solve questions not implemented before

- ☐ Pick from prepared list
- Define test cases
- Extend functionality
- Validate results in triplestore

Solve new QA tasks

Extend the qa vocabulary

Choose existing QA components supporting your task

Implement new QA component for your new use case

Extend test cases and validate your work

Final Remarks

Summary

Qanary methodology, the RDF vocabulary qa and the corresponding component-oriented Qanary framework.

Advantage of the Qanary ecosystem for rapid research results

Learn to iteratively build, validate and improve your own QA system using the Qanary framework

You built a QA system capable of answering generic question in a specific domain

Take Away: Qanary methodology

Qanary: knowledge-driven methodology for QA systems

Build on-top of the qa vocabulary (i.e., knowledge-driven approach)

Agile approach for creating QA systems

Join Qanary at GitHub github.com/WDAqua/Qanary

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