SLA 2021 ANNUAL CONFERENCE DESTINATION EVERYWHERE



Using Semantic Resources to Enrich Content Services

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Agenda

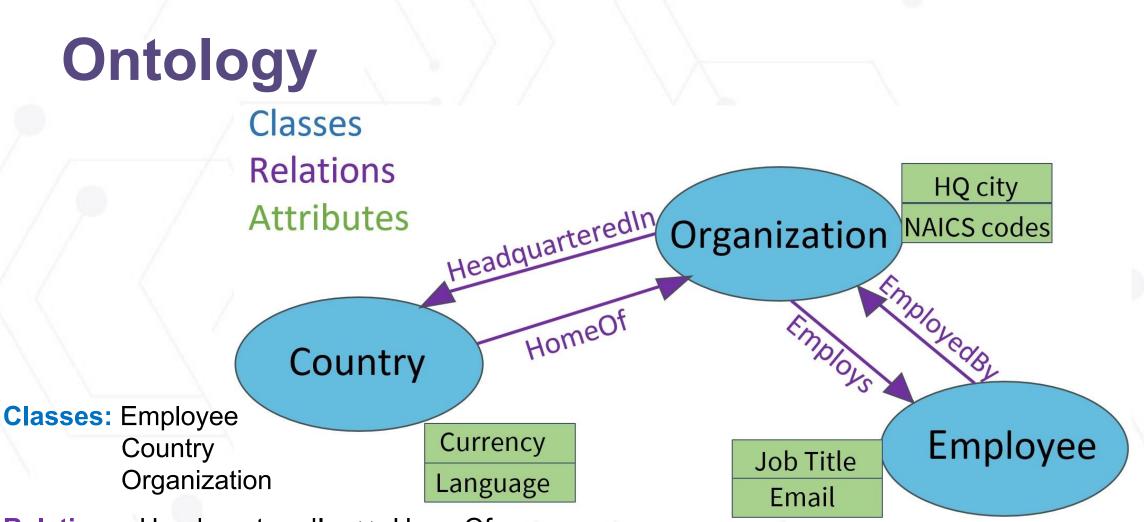
- Introductions
- Value of Semantic Resources
- Types of Semantic Resources (KOS)
- Knowledge Graph Use Cases
- JPL's Knowledge Graph Use Cases
- Knowledge Graph Implementation Examples



Types of Semantic Resources



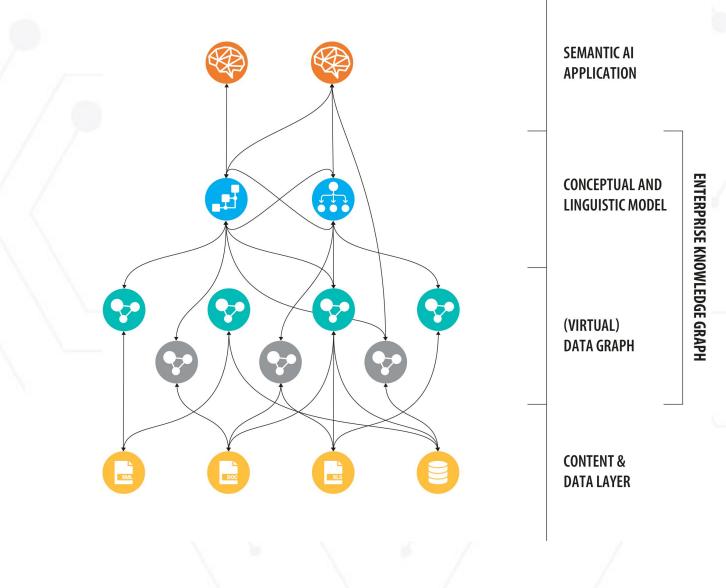
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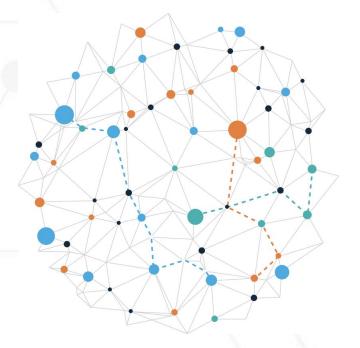


Relations: HeadquarteredIn < > HomeOf EmployedBy < > Employs

Attributes: Email address, Job title, HQ city, NAICS codes, Currency, Language

What is an Enterprise Knowledge Graph (EKG)?





An Enterprise Knowledge Graph (EKG) contains business objects and topics that are closely linked, classified, semantically enriched, and connected to existing data and documents. Knowledge Graph Implementation Examples



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Knowledge Graph Implementations

Systems that provide information without standard search for non-expert users:

Recommendation Systems

- Users don't know how best to look for information that would benefit them
- Users don't know that the information is there or how to find it
- Use cases for suggesting content: cross-selling, networking, learning

Question Answering Systems

- Natural language interface over enterprise data
- Concise and comprehensible answers to user questions
- Fast access to enterprise data for non-technical users
- It is already everywhere (Alexa, Siri, Google Assistant, etc.)

Knowledge Graph Implementations

Advantages of *knowledge graph*-based systems over other methods:

For recommendation systems:

- Based on explicit knowledge of content stored in database
- Does not require prior interactions, large numbers of users, or trained data, as in content filtering or collaborative filtering systems
- Does not have cold start problem due to insufficient data

For question answering systems:

- Requires little or no training data
- Robust in handling user mistakes
- Efficient and scales with the increasing size of the knowledge graph
- Can include explanations for how the question was interpreted
- Can deal with questions and answers in different languages





Knowledge Graph Implementations

Example solutions:

Recommendation system:

HR Recommender

Prototype application developed by Semantic Web Company with knowledge graph built in PoolParty software and user interface in Drupal https://hr-recommender.poolparty.biz

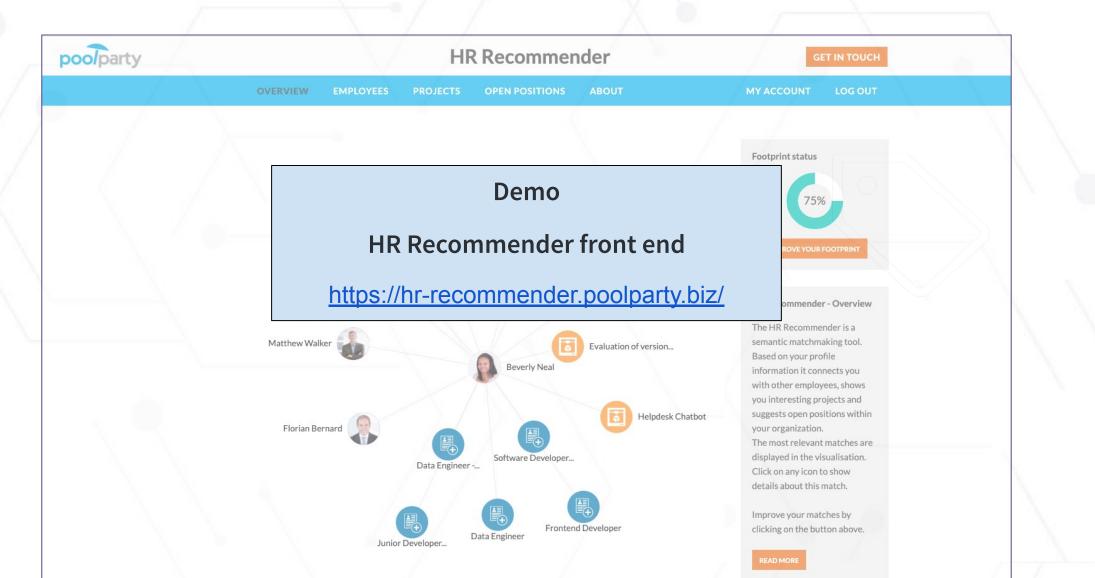
Question answering system:

QAnswer

Developed in a strategic partnership between QA Company and Semantic Web Company with knowledge graph built in PoolParty. https://ganswer.poolparty.biz



Recommendation System Example



Recommendation System Example: HR Recommender

PROFILE	EXTRACT CONCEPTS	STORE IN GRAPHSEARCH	
		SPARQL	
SHOW RESULTS	FIND MATCHES	ENRICH FOOTPRINT	

Recommendation System Example: HR Recommender

HR Recommender Components

Semantic resources

- Taxonomies containing concepts and labels
- Ontology of semantic relations

Content that is text-mined

• CVs, personal profiles, job descriptions, project descriptions

Stored data

• Knowledge graph and a search index (Solr)

Recommender application

- Algorithms for calculating similarities and recommendations to *enrich* the semantic footprint
- Web application user interface on top of an API

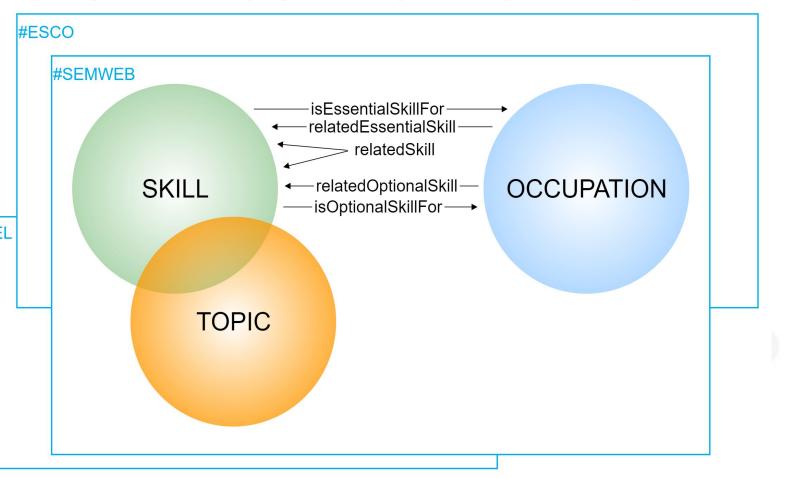


Recommendation System Example: HR Recommender

Taxonomy sources:

- Skills & Occupations Topics:
 SEMWEB custom taxonomy
- Skills & Occupations: ESCO Classification
- Taxonomy enriched with text mining (entity extraction) #PROPEL of Topics:
 Propel corpus of industry conference content:
 - submitted papers, speakers

Ontology layer: Adds semantic relationships



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	advocate for healthcare users' needs (0) advocate for others (0)	relatedEssentialSkill						
	advocate health (0) aeronomy (0) aerospace engineering (1)	relatedOptionalSkill						

Question Answer System Example

poolparty & OQAnswer How to Use Contact QAnswer Demo what is r who is program how man **QAnswer front end** W e which https://ganswer.poolparty.biz

This demo demonstrates a natural language interface over data included in a typical **Human Resources Use Case**. The data is stored in a **Knowledge Graph** that has been created from structured (Excel files) and unstructured data (Text files) using **PoolParty Semantic Suite**. The user can pose questions, get answers in the form of tables, infoboxes, and plain text, and provide feedback, in order to improve the Question Answering (QA) system. The questions are answered using **QAnswer**.

Download our White Paper

Have a look at our guide to experience the full spectrum of possibilities of this demo.

QAnswer System Example

QAnswer System Components

Semantic resources

- Taxonomies containing concepts and labels
- Ontology of semantic relations

Data sources

• Databases, spreadsheets, document repositories

Pipelines

• To automatically extract and semantify the data

Machine learning models

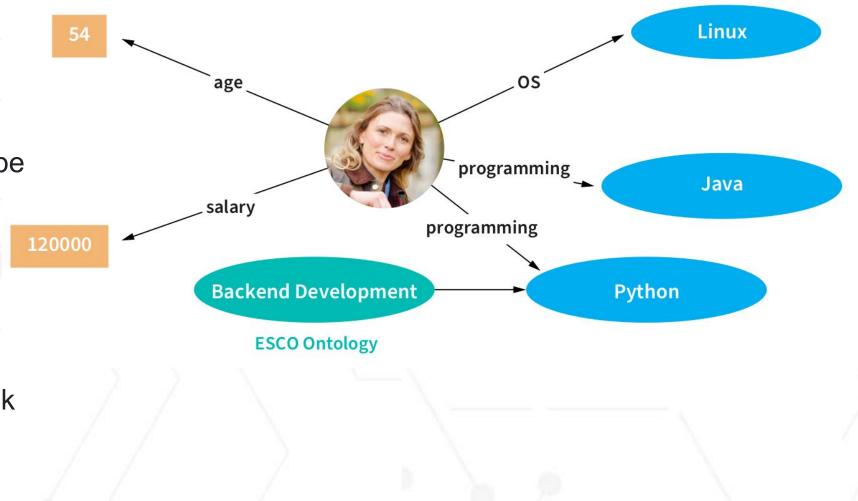
- Pre-trained machine learning models
- New machine learning models generated and trained





QAnswer System Example

- Create a semantic model using taxonomies and ontologies to describe the domain.
- Establish new
 relationships using
 logical inferences,
 graph analysis or link
 discovery.



QAnswer System Example

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Contact



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