



CSCE 590-1: From Data to Decisions with Open Data: A Practical Introduction to Al

Lecture 5: Data Preparation, Knowledge Representation/ Graph, Ontology

PROF. BIPLAV SRIVASTAVA, AI INSTITUTE 26TH JAN 2021

Carolinian Creed: "I will practice personal and academic integrity."

Organization of Lecture 5

- Introduction Segment
 - Announcements
 - Recap of Lecture 4
 - Course Project
- Main Segment
 - Data preparation
 - Knowledge representation/ graph
 - Ontology
- Concluding Segment
 - About Next Lecture Lecture 6
 - Ask me anything

Introduction Segment

Announcements

- Office hour on Monday and Wednesdays
 - 11:30 am 12:30 pm
 - Please join using Blackboard Ultra
 - Send email if you do not see me/ or am inactive
- Quiz in next class
 - From material covered in Lectures 1-5
 - First class (~45 mins) and then Quiz

Course Project

- (<u>Undergraduate</u>) Project: 50% + 10%:
 - Do a significant project of your choice following the framework presented in Lecture 3
 - Deliverable: Project report and code (50%), 1-slide elevator presentation to class (10%)
- Framework: Value of decision: before and after, Data-needed, Method, Evaluation, Integrating with overall process

• Illustration:

- Data analysis project
- Dataset must be from given catalog Lecture 1 (US: https://www.data.gov/ or any US state; Text of legislations LegiScan, https://legiscan.com/; Kaggle datasets: https://www.kaggle.com/datasets; Google datasets search: https://datasetsearch.research.google.com/) OR discussed with instructor
- Use analytical methods to present new insights
- Problem (method) to be discussed with instructor
- Examples: SC traffic deaths, COVID responses of a US states, economic growth and tax rates

Course Project

- (**Graduate**) Project: 50% + 10%:
 - Do a significant project of your choice following the framework presented in Lecture 3
 - Deliverable: Project report and code (50%), 1-slide elevator presentation to class (10%)
- Framework: Value of decision: before and after, Data-needed, Method, Evaluation, Integrating with overall process
- Illustrations:
 - Project on data analysis (like for undergraduates) but bigger scope
 - Project on creating or exploring new methods (preferred)
 - Problem (method) to be discussed with instructor
 - Examples: Estimate crowd from sound, find mask adherence from photos, find algal bloom from images of water

Rubric for Evaluation of Course Project

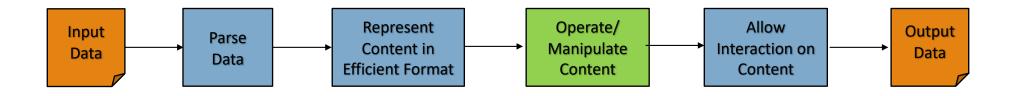
- Project
 - Project plan
 - Challenging nature of project
 - Actual achievement
 - Report
 - Sharing of code
- Presentation
 - Motivation
 - Coverage of related work
 - Results and significance
 - Handling of questions

Recap of Lecture 4

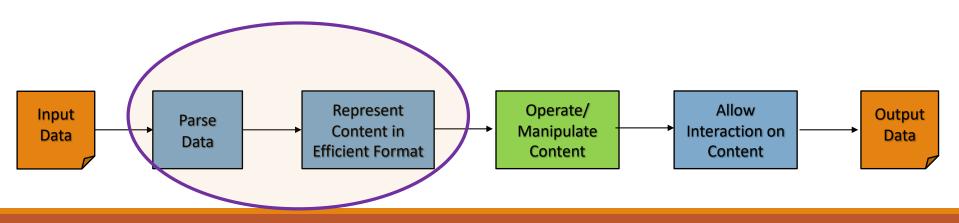
- We covered a background in water
 - Looked at common problems
- Looked at decision-support situations from US and a few solutions
- Looked at decision-support situations from India and a few solutions
- We explored the scope of AI

Main Segment

Data Processing Pipeline

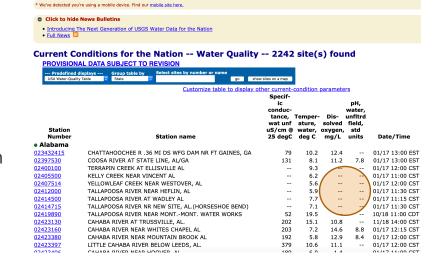


Data Preparation



Common Problem: Missing Value

- Occurrence
 - Missing completely at random
 - Missing at random (a group not wanting to participate)
 - Missing not at random (a group not able to participate)
- What does it mean?
 - The value was not provided
 - The value does not exist or has no practical interpretation
 - The value is being hidden (redaction)
 - Others: The value is not reliable, ...



- How to detect it?
 - By checking for specific values: NA, Not applicable, out-of-range value, 0, -1, "".

Missing Value – Handling

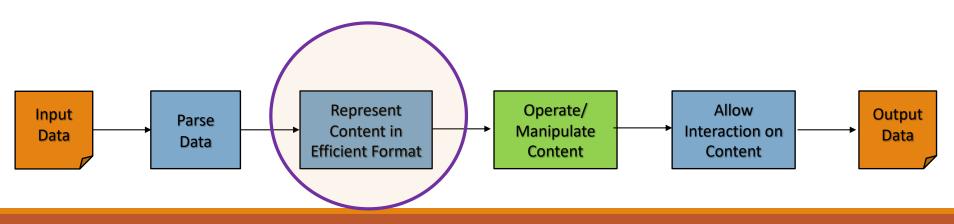
- Ignoring missing value (Omission)
 - Reduces available data
- Impute new value (Imputation)
 - Mean or median
 - Default value
- Analysis techniques which are robust against missing value
 - Expectation maximization

Code Examples

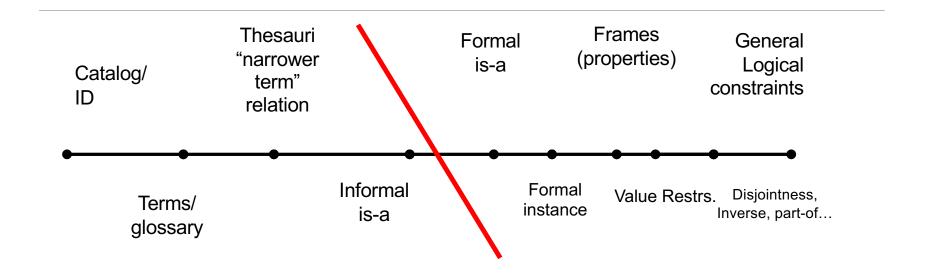
https://github.com/biplav-s/course-d2d-ai/tree/main/sample-code/l4-dataprep

- Basic concepts: **DataPreparation-Numeric.ipynb**
- An illustration: Clean-RealSample.ipynb

Annotation: Knowledge Graphs and Ontology



The Spectrum of Annotation Methods



Ontologies Come of Age McGuinness, 2001, and From AAAI Panel 99 – McGuinness, Welty, Uschold, Gruninger, Lehmann Plus basis of Ontologies Come of Age – McGuinness, 2003

Thesaurus – Authoritative Entities and Relationships

Countries: https://en.wikipedia.org/wiki/List_of-ISO_3166 country codes

ISO 3166 ^[1]				ISO 3166-1 ^[2]		ISO 3166-2 ^[3]	
Country name ^[5] +	Official state name ^[6] +	Sovereignty ^[6] [7][8]	Alpha-2 code ^[5]	Alpha-3 code ^[5]	Numeric code ^[5]	Subdivision code links ^[3]	Internet ccTLD ^[9] +
Marie Afghanistan	The Islamic Republic of Afghanistan	UN member state	AF	AFG	004	ISO 3166-2:AF	.af
Akrotiri and Dhekelia – See United Kingdom, The							
Aland Islands	Åland	Finland	AX	ALA	248	ISO 3166-2:AX	.ax
Albania	The Republic of Albania	UN member state	AL	ALB	008	ISO 3166-2:AL	.al
Algeria	The People's Democratic Republic of Algeria	UN member state	DZ	DZA	012	ISO 3166-2:DZ	.dz
American Samoa	The Territory of American Samoa	United States	AS	ASM	016	ISO 3166-2:AS	.as
■ Andorra	The Principality of Andorra	UN member state	AD	AND	020	ISO 3166-2:AD	.ad
Angola	The Republic of Angola	UN member state	AO	AGO	024	ISO 3166-2:AO	.ao
™ Anguilla	Anguilla	United Kingdom	AI	AIA	660	ISO 3166-2:AI	.ai
Antarctica ^[a]	All land and ice shelves south of the 60th parallel south	Antarctic Treaty	AQ	АТА	010	ISO 3166-2:AQ	.aq
Antigua and Barbuda	Antigua and Barbuda	UN member state	AG	ATG	028	ISO 3166-2:AG	.ag
- Argentina	The Argentine Republic	UN member state	AR	ARG	032	ISO 3166-2:AR	.ar

(Unique) US Counties Information

In COVID sample code: https://github.com/biplav-s/course-d2d-ai/blob/main/sample-code/l3-health/CovidExploration.ipynb,

reference made to FIPS code

References:

- https://www.nrcs.usda.gov/wps/portal/nrcs/detail/national/ home/?cid=nrcs143 013697
- https://github.com/kjhealy/fipscodes/blob/master/county_fips_master.csv

Question: how many Richland counties are there in US?

Answer: 14

County FIPS Codes				
FIPS	Name	e	Sta	
01001	Autauga		AL	
01003	Baldwin		AL	
01005	Barbour		AL	
01007	Bibb		AL	
01009	Blount		AL	
01011	Bullock		AL	
01013	Butler		AL	
01015	Calhoun		AL	
01017	Chambers		AL	
01019	Cherokee		AL	
01021	Chilton		AL	
01023	Choctaw		AL	
01025	Clarke		AL	
01027	Clay		AL	
01029	Cleburne		AL	
01031	Coffee		AL	
01033	Colbert		AL	
01035	Conecuh		AL	

Is-a Relationship

List of Countries, States, ... (County), City

- United Nations: https://unece.org/trade/cefact/unlocode-code-list-country-and-territory
- US Source: https://github.com/grammakov/USA-cities-and-states

Schema.org

- Website: https://schema.org/docs/about.html
- GitHub: https://github.com/schemaorg/schemaorg
- An organization of metadata information for entities found on the web. Mostly backed by web search companies.
- Explore
 - Thing: https://schema.org/Thing
 - Product:

```
Example 2 (a)

No Markup | Microdata RDFa JSON-LD Structure

Example notes or example HTML without markup.

<imp src="dell-30in-lcd.jpg" alt="A Dell UltraSharp monitor" /> Dell UltraSharp 30" LCD Monitor

87 out of 100 based on 24 user ratings

$1250 to $1495 from 8 sellers

Sellers:

<a href="save-a-lot-monitors.com/dell-30.html"> Save A Lot Monitors - $1250</a>
<a href="jondoe-gadgets.com/dell-30.html"> Jon Doe's Gadgets - $1350</a>
...
```

Example 2 Page No Markup Microdata RDFa JSON-LD Structure Structured representation of the JSON-LD example. @tvpe Product Dell UltraSharp 30" LCD Monitor offers @type AggregateOffe | offerCount \$1250 highPrice \$1495 @type http://example.org/jondoe-gadgets.com/dell-30.html offers http://example.org/save-a-lot-monitors.com/dell-30.html image http://example.org/dell-30in-lcd.jpg aggregateRating @type AggregateRating ratingValue 87 24 ratingCount bestRating 100

No structure

Structure in JSON-LD format

Induced Structure

Schema.org - 2

Schema.org - continued

Exploration Exercise

- Services: https://schema.org/Service
- Event: https://schema.org/Event
- Benefit:
 - Easy to incorporate annotations
 - Uses popular development tools and technologies (JSON, Microformat)
- Disadvantage
 - Cannot perform deep inferencing
 - Popular in certain communities

Formalizing Knowledge in an Ontology

Sources:

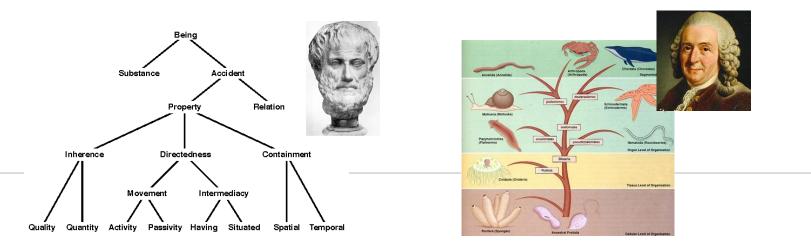
Achille Fokoue, Anastasios Kementsietsidis Tutorial SCRIBE presentation by Rosario Usceda Sosa, Biplav Srivastava, Bob Schloss

- https://github.com/rschloss/ismp ,
- https://researcher.watson.ibm.com/researcher/view_group.php?id=2505

What is an ontology, anyway?

In Computer Science, "An ontology is a formal explicit description of concepts in a domain of discourse (classes (sometimes called concepts)), **properties** of each concept describing various features and **attributes** of the concept (slots (sometimes called roles or properties)), and **restrictions** on slots (facets (sometimes called role restrictions)). An ontology together with a set of individual instances of classes constitutes a knowledge base. In reality, there is a fine line where the ontology ends and the knowledge base begins." [Noy, 2000]

Not to be confused with ontologies (and/or taxonomies) in Philosophy or Life Sciences



In a Smart City domain, we're concerned with modeling the *city data* (city activity data, city departments, assets, KPIs), not the city itself (the full set of spatial and temporal relations between people and objects in the city) Ontologies help us to structure and reason about city *events*, *entities* and *services*.

Ontology = Class + Relations + Constraints

Knowledge Base = Ontology + instances + (Standard) Inference and rules

RDF / Turtle Example

```
vcard:FN

John Smith

John Smith

John Smith
```

```
<rdf:RDF
   xmlns:rdf="http://www.w3.org/1999/02/22-rdf-
syntax-ns#"
   xmlns:vcard="http://www.w3.org/2001/vcard-
rdf/3.0#" >
   <rdf:Description rdf:nodeID="A0">
        <vcard:Given>John</vcard:Given>
        <vcard:Family>Smith</vcard:Family>
   </rdf:Description>
   <rdf:Description
rdf:about="http://somewhere/JohnSmith">
        <vcard:FN>John Smith</vcard:FN>
        <vcard:N rdf:nodeID="A0"/>
   </rdf:Description>
</rdf:RDF>
```

OWL extends RDF...

RDF-schema

- Class, subclass
- Property, subproperty

+ Restrictions

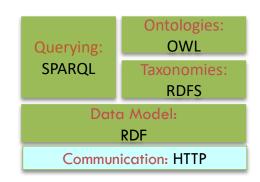
- Range, domain
- Local, global
- Existential
- Cardinality

+ Combinators

- Union, Intersection
- Complement
- Symmetric, transitive

+ Mapping

- Equivalence
- Inverse



Source: Achille Fokoue, Anastasios Kementsietsidis Tutorial

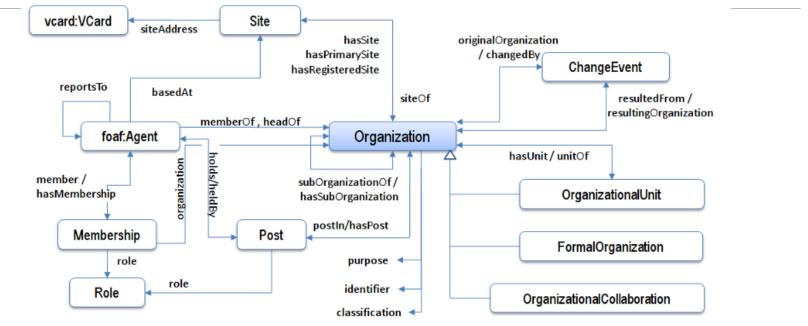
Not all ontologies are created equal

In practice, ontologies are used -together with inferencing engines and rules-, for a variety of purposes. If we think of them as schemas, there are different ways

	<u> </u>		Purpose	Instances	Inferencing	Examples
Norm sc	ative hema	As a deductive system	Deductive System (axioms + deductive rules)	Part of the knowledge base	Defined by rules.	Expert systems, Planning, Optimization.
		As a data blueprint	Constrain a domain	Must conform to the normative schema determined by the ontology	Subsumption, class inferencing	Biomedical and life sciences (FMA, Radlex)
		As a data classifier	Classify open data	Unknown formats	Subsumption, class inferencing	Tag ontologies (MOAT, Echarte, SCOT, NAO, etc.)
		As a data integrator	Integrating pre-defined model to existing data sources	Instances are mapped, no constraint enforcement.	Subsumption, class, entity inferencing	SCRIBE
S	egrativ Schema depend stance	, As data mapping	Mapping to/from existing data sources	Mined instances determine the ontology/schema.	Subsumption, class inferencing	D2RQ (a tool)

SCRIBE belongs to the **fourth** category: It has no constraints and was designed to support the programming of tools that allow domain experts to deal with entities natural to them (even if the recorded data is actually distributed).

Larger Example: Organization Ontology



Ontology description: http://www.w3.org/TR/vocab-org/

Ontology: http://www.w3.org/ns/org.ttl

Larger Ontology

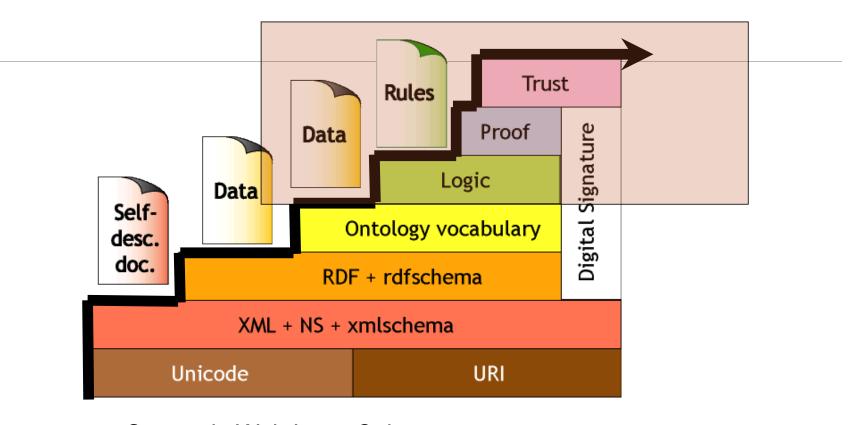
http://www.w3.org/ns/org.ttl

```
@prefix rdf:
                   <http://www.w3.org/1999/02/22-rdf-syntax-ns#> .
@prefix rdfs:
                   <http://www.w3.org/2000/01/rdf-schema#> .
@prefix owl:
                   <http://www.w3.org/2002/07/owl#> .
@prefix xsd:
                   <http://www.w3.org/2001/XMLSchema#>
@prefix skos:
                   <http://www.w3.org/2004/02/skos/core#> .
                   <http://xmlns.com/foaf/0.1/> .
@prefix foaf:
@prefix :
                  <http://www.w3.org/ns/org#> .
# -- Meta data ------
<http://www.w3.org/ns/org#>
   a owl:Ontology;
   owl:versionInfo "0.7";
   rdfs:label "Core organization ontology"@en;
   rdfs:comment "Vocabulary for describing organizational structures, specializable to a
broad variety of types of organization. "@en;
   dct:created "2010-05-28"^^xsd:date;
   dct:modified "2010-06-09"^^xsd:date;
   dct:modified "2010-10-08"^^xsd:date;
   rdfs:seeAlso <http://www.w3.org/TR/vocab-org/>;
# -- Organizational structure ------
org:Organization a owl:Class, rdfs:Class;
   rdfs:subClassOf foaf:Agent;
   owl:equivalentClass foaf:Organization;
   rdfs:label "Organization"@en;
   rdfs:label "Organisation"@fr;
   owl:hasKey (org:identifier) ;
   rdfs:comment """Represents a collection of people organized together into a community
or other social, commercial or political structure. ... Alternative names: Collective
Body__Org Group """@en;
   rdfs:comment """Représente un groupe de personnes organisées en communauté où tout
autre forme de structure sociale, commerciale ou politique. ... code provenant d'une liste
   rdfs:isDefinedBy <ahttp://www.w3.org/ns/org>;
```

```
<rdf:RDF xmlns:rdf="http://www.w3.org/1999/02/22-rdf-syntax-ns#"</pre>
 xmlns:skos="http://www.w3.org/2004/02/skos/core#" xmlns:foaf="http://xmlns.com/foaf/0.1/" xmlns:org="http://www.w3.org/ns/org#" xmlns:gr="http://purl.org/goodrelations/v1#"
 xmins:owl="http://www.w3.org/2002/07/owl#" xmins:dct="http://purl.org/dc/terms/" xmins:prov="http://www.w3.org/ns/prov#" xmins:owlTime="http://www.w3.org/2006/time#
 xmlns:xsd="http://www.w3.org/2001/XMLSchema#" xmlns:vcard="http://www.w3.org/2006/vcard/ns#"
 xmlns:rdfs="http://www.w3.org/2000/01/rdf-schema#">
+ <owl:Ontology rdf:about="http://www.w3.org/ns/org#">
+ <rdfs:Class rdf:about="http://www.w3.org/ns/org#Organization">
 <rdfs:Class rdf:about="http://www.w3.org/ns/org#Role">
   <rdfs:label xml:lang="fr">Rôle</rdfs:label>
 - <owl:disjointWith>
     <owl:Class rdf:about="http://www.w3.org/ns/org#ChangeEvent" />
   <rdfs:subClassOf rdf:resource="http://www.w3.org/2004/02/skos/core#Concept" />
   <owl:disjointWith>
     <owl:Class rdf:about="http://www.w3.org/ns/org#Site" />
   <rd>s:comment xml:lang="fr">Indique le rôle qu'une Personne ou un autre Agent peut avoir dans une
     Organisation. Les instances de cette classe décrivent le rôle dans l'absolu; pour indiquer une personne
     ayant ce rôle spécifique dans une Organisation, utilisez une instance de `org:Membership`. Il est
      courant que les rôles soient organisés dans une sorte de taxonomie, ce qui peut être représenté avec
     SKOS. Les propriétés de libellés standards de SKOS devraient être utilisées pour libeller le Rôle.
     D'autres propriétés additionnelles pour ce rôle, comme une fourchette de Salaire peuvent être ajoutées
     par une extension de ce vocabulaire.</rdfs:comment>
   <owl:disjointWith>
      <owl:Class rdf:about="http://www.w3.org/ns/org#Membership" />
    </owl:disjointWith>
   crdfs:label vml:lang="en" Role c/rdfs:label
   <rdfs:isDefinedBy rdf:resource="http://www.w3.org/ns/org"/>
   <rdf:type rdf:resource="http://www.w3.org/2002/07/owl#Class"/>
    <rdfs:comment xml:lang="en">Denotes a role that a Person or other Agent can take in an organization.
     Instances of this class describe the abstract role; to denote a specific instance of a person playing that role in a specific organization use an instance of `org:Membership`. It is common for roles to be
```

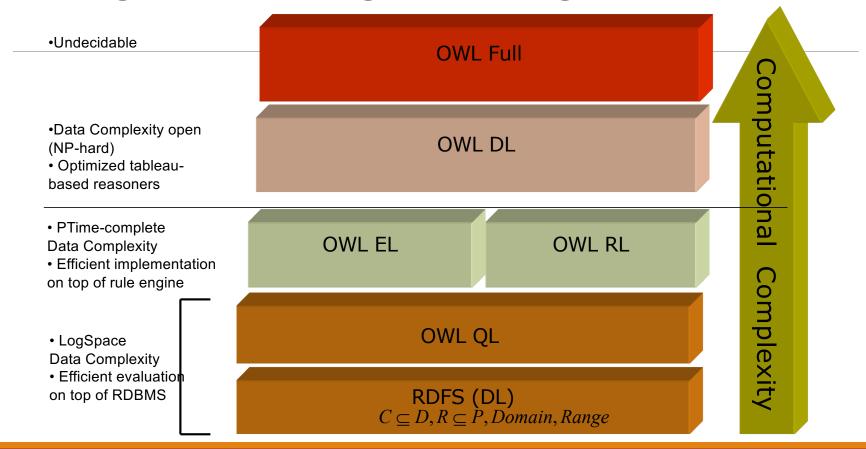
http://www.w3.org/ns/org

Moving to the future of the web



Semantic Web LayerCake (Berners-Lee, 99;Swartz-Hendler, 2001)

Challenge of Reasoning on Ontologies



What makes a good ontology for data integration?

A *good* ontology is a *useful* ontology, an ontology that *both* humans and systems can process.

Human Usability	Communicable. Naming, natural language support, etc.		
	Concise . A simple way to describe the key entities of the model and yet able to infer many facts		
	Consistent . Naming conventions and modeling patterns		
	Authoritative to domain experts		
	Documented , not just descriptions, but also provenance		
	Managed and maintained by people throughout the model lifecycle.		
	Reusable in similar domains, for similar instances.		
System Usability	Scalable so large amounts of data can be parsed, stored and retrieved.		
	Efficient query and inferencing		
	Programmable solutions, both in open and closed data paradigms.		
	Open infrastructure and tools community		

- Formal representation of knowledge in a particular domain
- ☐ Formally defines key **concepts** and **relations** in the domain
- Specifies relationships between those key concepts and relations
- Supports automated reasoning about entities in the domain

RAL LANGUAGE

Using Ontology

- Visually via tools like Protégé https://protege.stanford.edu/
- Programmatically with APIs like
 - Jena (Java) https://jena.apache.org/documentation/ontology/
 - OwlReady2 (Python) https://bitbucket.org/jibalamy/owlready2/src/master/
 - Rdflib (Python) https://github.com/RDFLib/OWL-RL
- A compendium of resources https://github.com/totogo/awesome-knowledge-graph

Code Illustration

On Github:

https://github.com/biplav-s/course-nl/blob/master/l11-ontology/Exploring%20ontologies.ipynb

Knowledge Graph

- No clear definition
 - <u>"Towards a Definition of Knowledge Graphs," by Lisa Eherlinger and Wolfram Wöß, CEURWorkshop Proceedings.</u> 2016, http://ceur-ws.org/Vol-1695/paper4.pdf
 - For practical purposes, concepts and their relationships; not constraints
 - Driven by applications in search and information integration
 - See discussion at: http://accidental-taxonomist.blogspot.com/2019/05/knowledge-graphs-and-ontologies.html
- But ontology as knowledge graph widely used in industries
 - Industry-Scale Knowledge Graphs: Lessons and Challenges, CACM 2019, https://cacm.acm.org/magazines/2019/8/238342-industry-scale-knowledge-graphs/fulltext

Development stage Data model Size of the graph Microsoft The types of entities, relations, ~2 billion primary entities, Actively used in and attributes in the graph are ~55 billion facts products defined in an ontology. Google Strongly typed entities, 1 billion entities. Actively used in relations with domain and 70 billion assertions products range inference ~50 million primary entities, Facebook All of the attributes and Actively used in ~500 million assertions relations are structured and products strongly typed, and optionally indexed to enable efficient retrieval, search, and traversal. eBay Expect around 100 million Early stages of Entities and relation, wellproducts, >1 billion triples structured and strongly typed development and deployment **IBM** Entities and relations Various sizes. Proven on Actively used in with evidence information scales documents >100 products and million, relationships >5 by clients associated with them. billion, entities >100 million

KG Usage

Figure courtesy: Industry-Scale Knowledge Graphs: Lessons and Challenges, CACM 2019

Lecture 5: Concluding Comments

- We looked at how to improve the quality of data
- Data preparation: handling missing values
- Importance of annotation and methods
 - Glossary
 - Taxonomy, Is-a relationship
 - Ontology

Concluding Segment

About Next Lecture – Lecture 6

Lecture 6: Analysis

- Machine Learning
- Structured Data: Supervised Methods
- Quiz 1