

How do Semantic Web, Linked Data, Open Data, and Knowledge Graphs interplay?

Axel Polleres

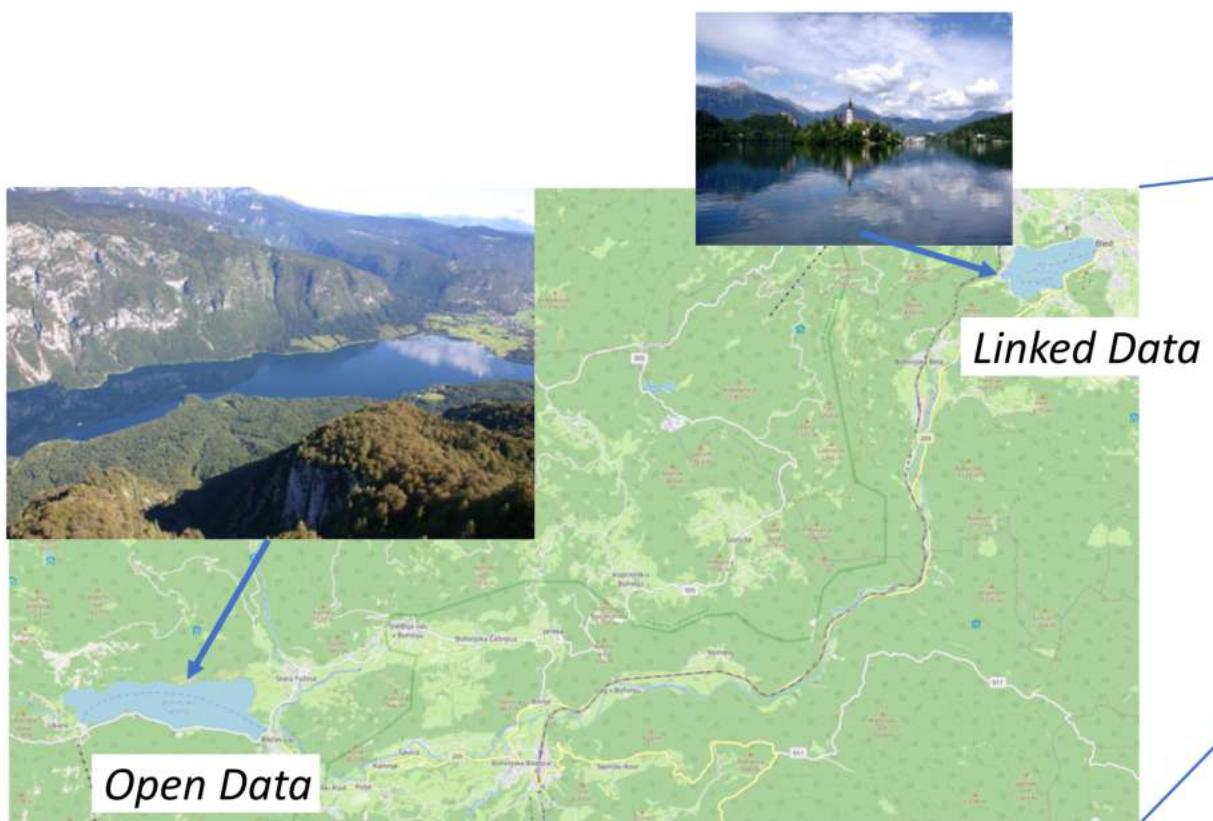
Semantic Web, Linked Data, Open Data, Knowledge Graphs ... travel report from a personal journey



Disclaimer: just like any trip report, this talk might be opinionated and influenced by personal experiences

Trip map... (incl. some “data lakes”)

- From Semantic Web to Linked Data
- From Linked Data to Open Data
- From Linked Data to Knowledge Graphs
- From Knowledge Graphs to Semantic Web



From Semantic Web to Linked Data

- (early 2000s)



The Semantic Web

A new form of Web content that is meaningful to computers will unleash a revolution of new possibilities

by [TIM BERNERS-LEE, JAMES HENDLER and ORA LASSILA](#)

The entertainment system was belting out the Beatles' "We Can Work It Out" when the phone rang. When Pete answered, his phone turned the sound down by sending a message to all the other *local* devices that had a *volume control*. His sister, Lucy, was on the line from the doctor's office: "Mom needs to see a specialist and then has to have a series of physical therapy sessions. Biweekly or something. I'm going to have my agent set up the appointments." Pete immediately agreed to share the chauffeuring. At the doctor's office, Lucy instructed her Semantic Web agent through her handheld Web browser. The agent promptly retrieved information about Mom's *prescribed treatment* from the doctor's agent, looked up several lists of *providers*, and checked for the ones *in-plan* for Mom's insurance within a *20-mile radius* of her *home* and with a *rating* of *excellent* or *very good* on trusted rating services. It then began trying to find a match between available *appointment times* (supplied by the agents of individual providers through their Web sites) and Pete's and Lucy's busy schedules. (The emphasized keywords indicate terms whose semantics, or meaning, were defined for the agent through the Semantic Web.)

In a few minutes the agent presented them with a plan. Pete didn't like it—University Hospital was all the way across town from Mom's place, and he'd be driving back in the middle of rush hour. He set his own agent to redo the search automatically assisted by supplying access certificates and shortcuts to the data it had already sorted through.

Expressing Meaning
Knowledge Representation
Ontologies
Agents
Evolution of Knowledge

<https://1n.pm/A5psO>

From Semantic Web to Linked Data

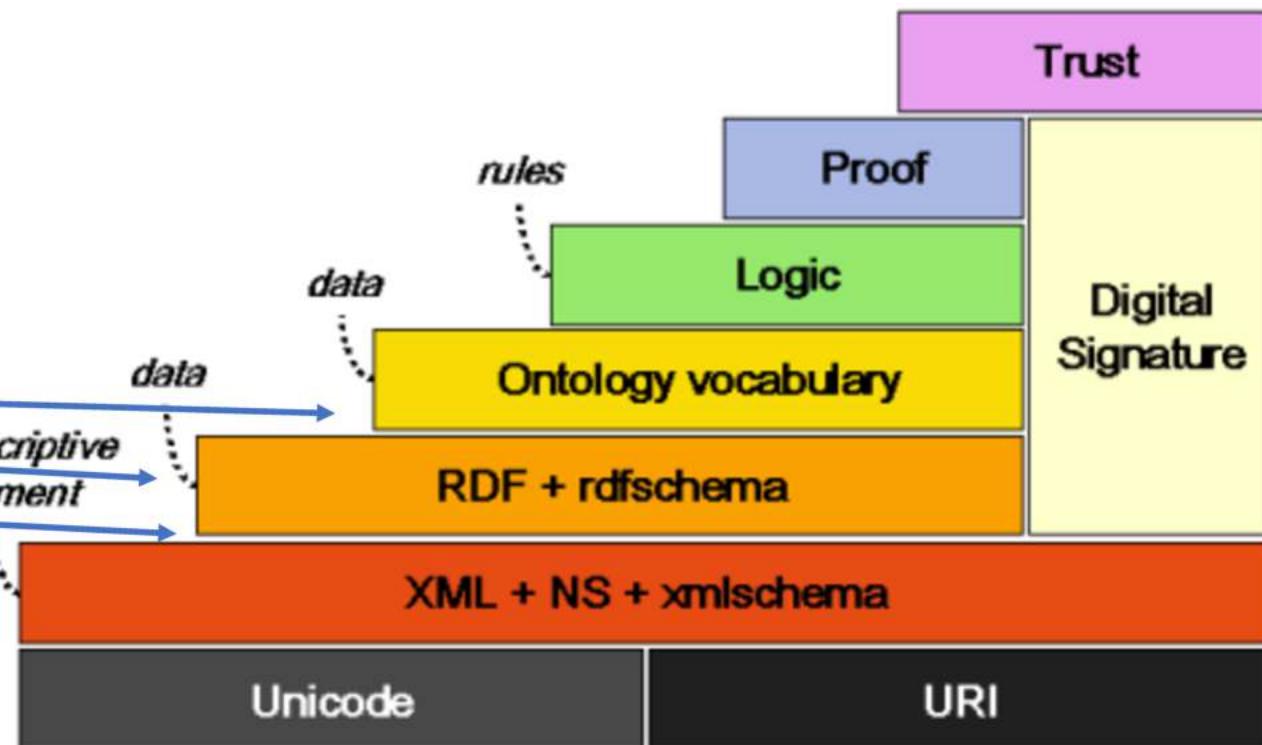
- (early 2000s)

Evolution of Knowledge Agents

Ontologies (OWL)

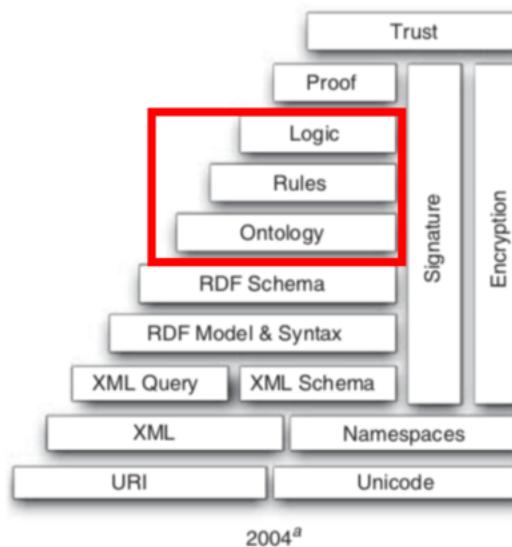
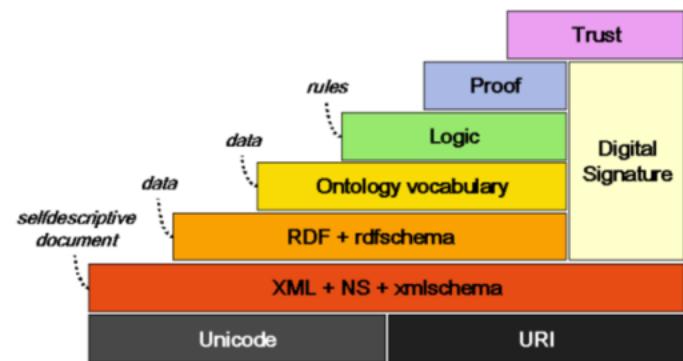
Knowledge Representation (RDFS)

Expressing Meaning (URLs and RDF)

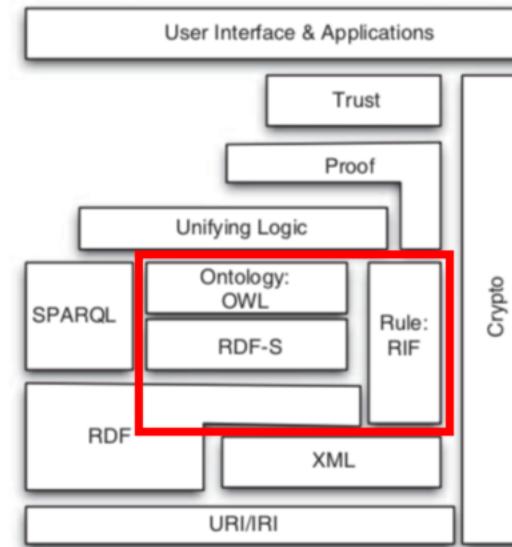


From Semantic Web to Linked Data

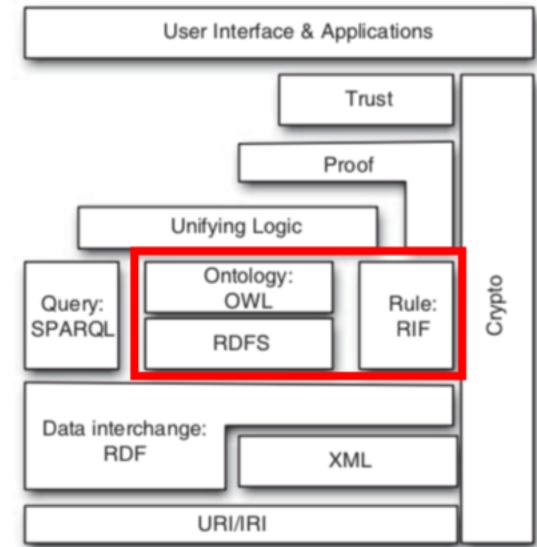
- (2000s - ca. 2009)



2004^a



2007^b



2009^c

^ahttp://www.w3.org/2004/Talks/0319-RDF-WGs/sw_stack.png

^b<http://www.w3.org/2007/Talks/0130-sb-W3CTechSemWeb/layerCake-4.png>

^c<http://www.w3c.it/talks/2009/athena/images/layerCake.png>

From Semantic Web to Linked Data

- (2000s - ca. 2009)

- Main (distracting?) question: what is the “right” knowledge representation and logic to express “Knowledge on the Web”?

- Description Logics?
- Rules? (Datalog, Deductive Databases)

- Nonmonotonic Logic Programming/Default Reasoning
- Open World Assumption vs. Closed World Assumption?
- “Local Closed–World assumption”/“Contextually scoped negation”
- Unique Names Assumption vs. Non-unique names



www.imdb.com/title/tt0175142/
vs.
en.wikipedia.org/wiki/Scary_Movie

My Web site is about *thrillers*,
Psycho is a thriller
all *thrillers* are *movies*

All horror movies
except comedies
are *thrillers*

Is *Nightmare on Elm Street* a comedy?



All horror movies *listed on my Website* except comedies *listed on IMDB* are thrillers

From Semantic Web to Linked Data

- (2000s - ca. 2009)

Semantic Web Architecture: Stack or Two Towers?

Ian Horrocks¹, Bijan Parsia², Peter Patel-Schneider³, and James Hendler²

¹ School of Computer Science
University of Manchester, UK
www.cs.man.ac.uk/~horrocks/

² Maryland Information and Network Dynamics Laboratory
University of Maryland
<http://cs.msi.cs.umd.edu/~bparsia/>

Keep DL and Rules
separate?

Abstract. We discuss language architecture for the Semantic Web, and in particular different proposals for extending this architecture with a rules component. We argue that an architecture that maximises compatibility with existing languages, in particular RDF and OWL, will benefit the development of the Semantic Web, and still allow for forms of closed world assumption and negation

Good news! Boost in KR research:

We know very well which ontological reasoning approaches are decidable and how they scale
→ OWL2, OBDA



Work on Unified Logics?

bases)
Planning/Default Reasoning
vs. Closed World Assumption
vs. Non-unique names
"option"/"Contextually sco

Rules with Contextually Scoped Negation

Axel Polleres^{1,2}, Cristina Feier¹, and Andreas Harth¹
¹ Digital Enterprise Research Institute Innsbruck, Austria and Galway, Ireland
² Universidad Rey Juan Carlos, Madrid, Spain
axel@polleres.net, {cristina.feier, andreas.harth}@deri.org

Contextualized Reasoning?

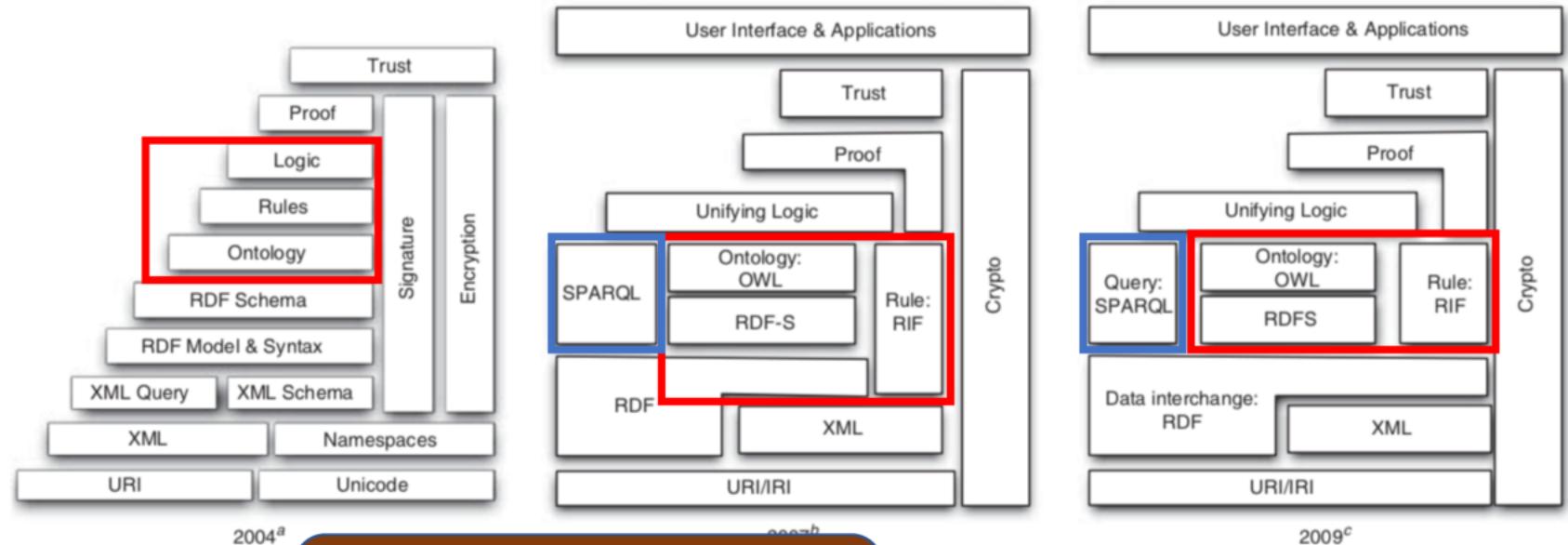
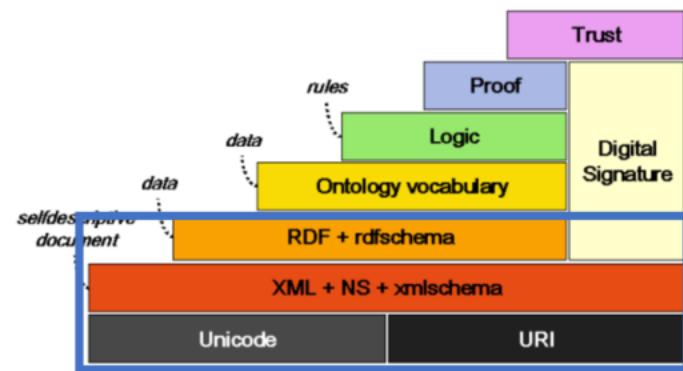
representation formalisms used on the Semantic Web. We argue that the real situation is closer to the classical closed world assumption. Therefore, nonmonotonic logics are better suited than the standard open world view, especially negation. Especially negation is notoriously hard to handle. The main idea is to extend the standard logic programs with contexts and "scoped negation". Especially negation is notoriously hard to handle. The main idea is to extend the standard logic programs with contexts and "scoped negation". We also argue that this class of logic programs can be viewed as a rule extension to a subset of RDF(S).
vs.
https://en.wikipedia.org/wiki/Scary_Movie

My Web site is about *Thrillers*, all *thrillers* are *movies*



From Semantic Web to Linked Data

- (2000s - ca. 2009)

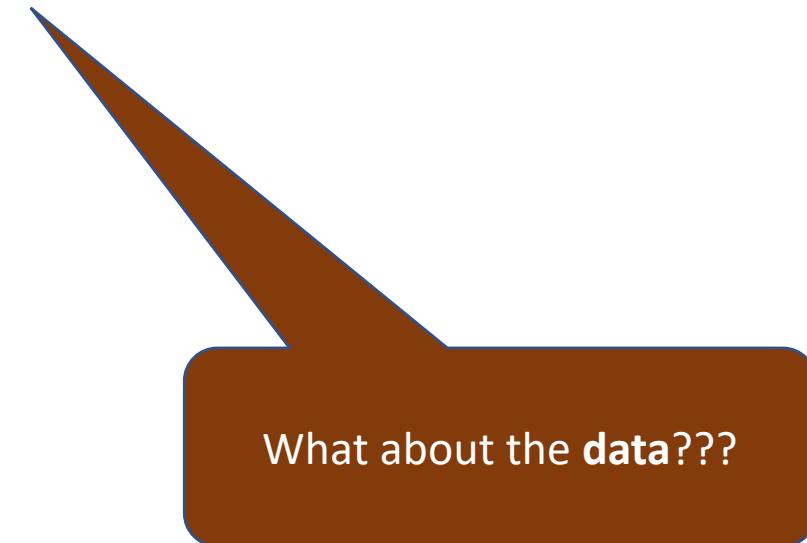


What about the **data**???

How much ontological Reasoning and Expressivity does the **Web** actually need?

From Semantic Web to Linked Data

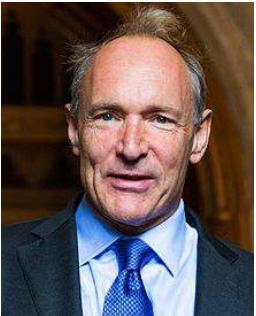
- (ca. 2006/7 – ca. 2013)
 - Main question: How can I **publish** “Knowledge on the Web” in order to enable **answering structured queries**?



What about the **data**???

From Semantic Web to Linked Data

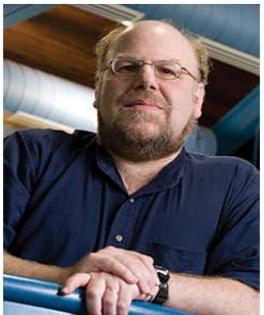
- (ca. 2006/7 – ca. 2013)
 - Main question: How can I **publish** “Knowledge on the Web” ...



Linked Data Principles

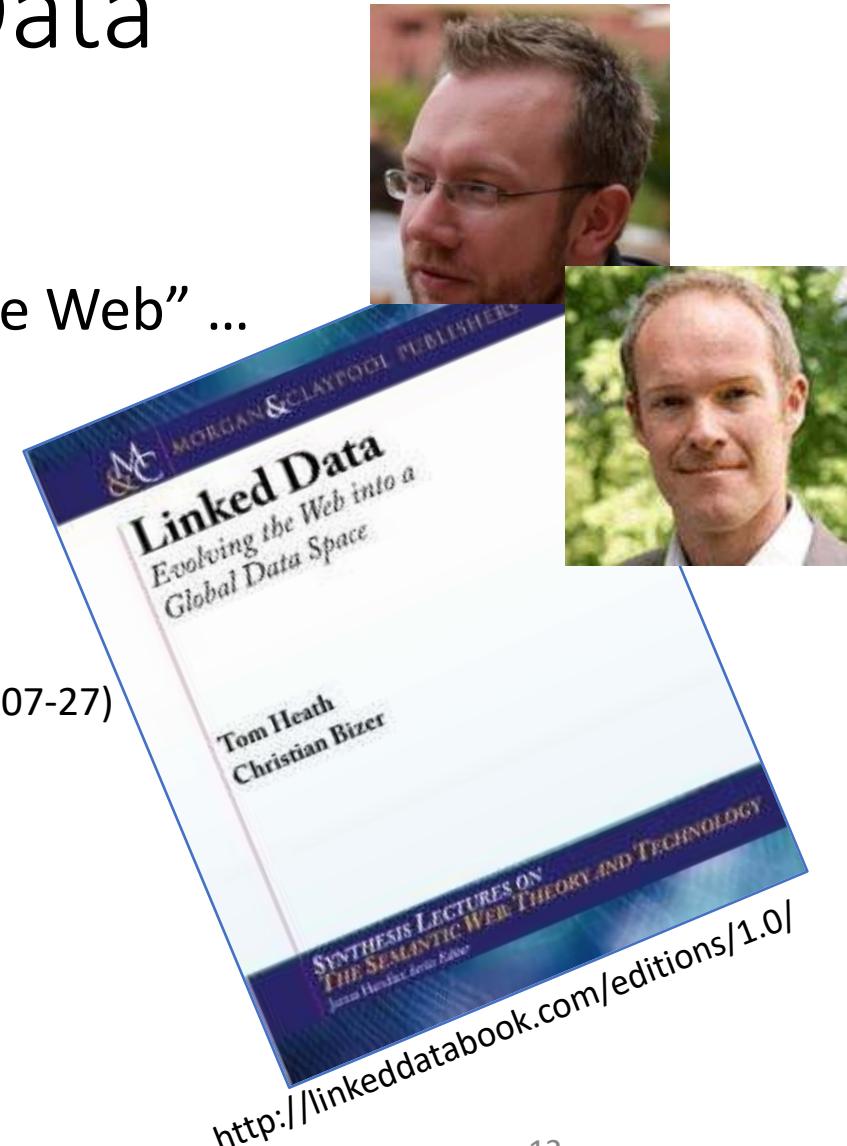
- **LDP1:** use URIs as names for things
- **LDP2:** use HTTP URIs so those names can be dereferenced
- **LDP3:** return useful – RDF? – information upon dereferencing those URIs
- **LDP4:** include links using externally dereferenceable URLs.

<https://www.w3.org/DesignIssues/LinkedData.html> (originally published 2006-07-27)



“A Little Semantics Goes a Long Way” (Jim Hendler)

<https://www.cs.rpi.edu/~hendler/LittleSemanticsWeb.html>



From Semantic Web to Linked (Open) Data

- (ca. 2006/7 – now)
 - Main question: How can I **publish** “Knowledge on the Web” ...

Linked Data Principles

- LDP1: use URIs as names for things
 - LDP2: use HTTP URIs so those names can be dereferenced
- LDP3: return useful – RDF? – information upon dereferencing those URIs
- LDP4: include links using externally dereferenceable URIs.

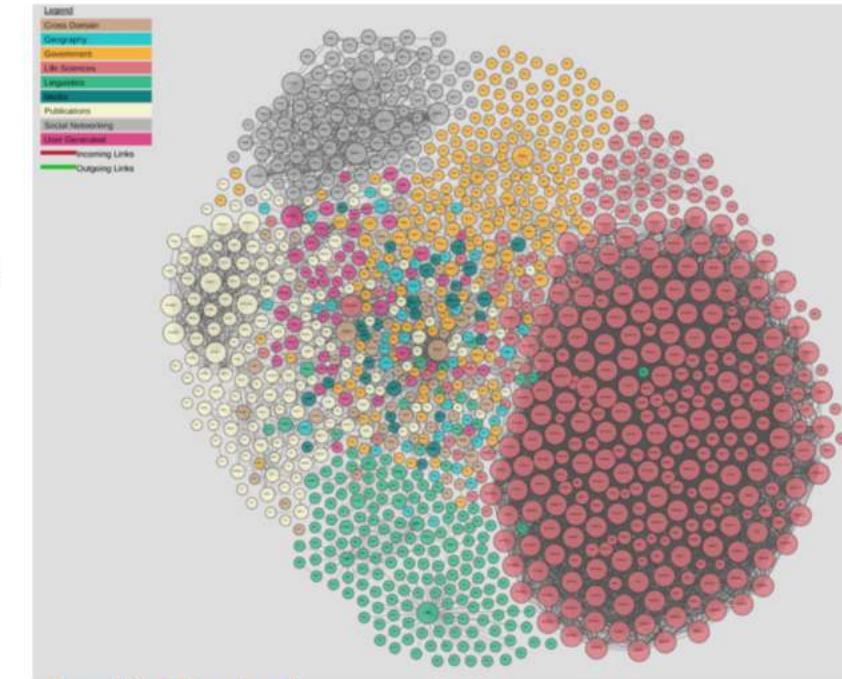
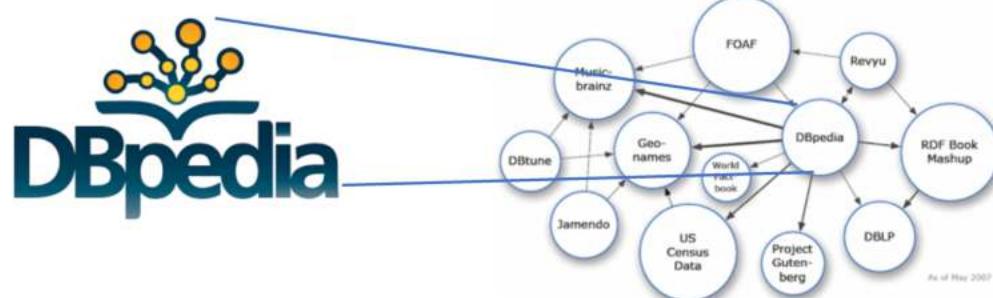
+ <https://www.w3.org/community/webize/2014/01/17/what-is-5-star-linked-data/>

★	Available on the web (whatever format) <i>but with an open licence, to be Open Data</i>
★★	Available as machine-readable structured data (e.g. excel instead of image scan of a table)
★★★	as (2) plus non-proprietary format (e.g. CSV instead of excel)
★★★★	All the above plus, Use open standards from W3C (RDF and SPARQL) to identify things, so that people can point at your stuff
★★★★★	All the above, plus: Link your data to other people's data to provide context



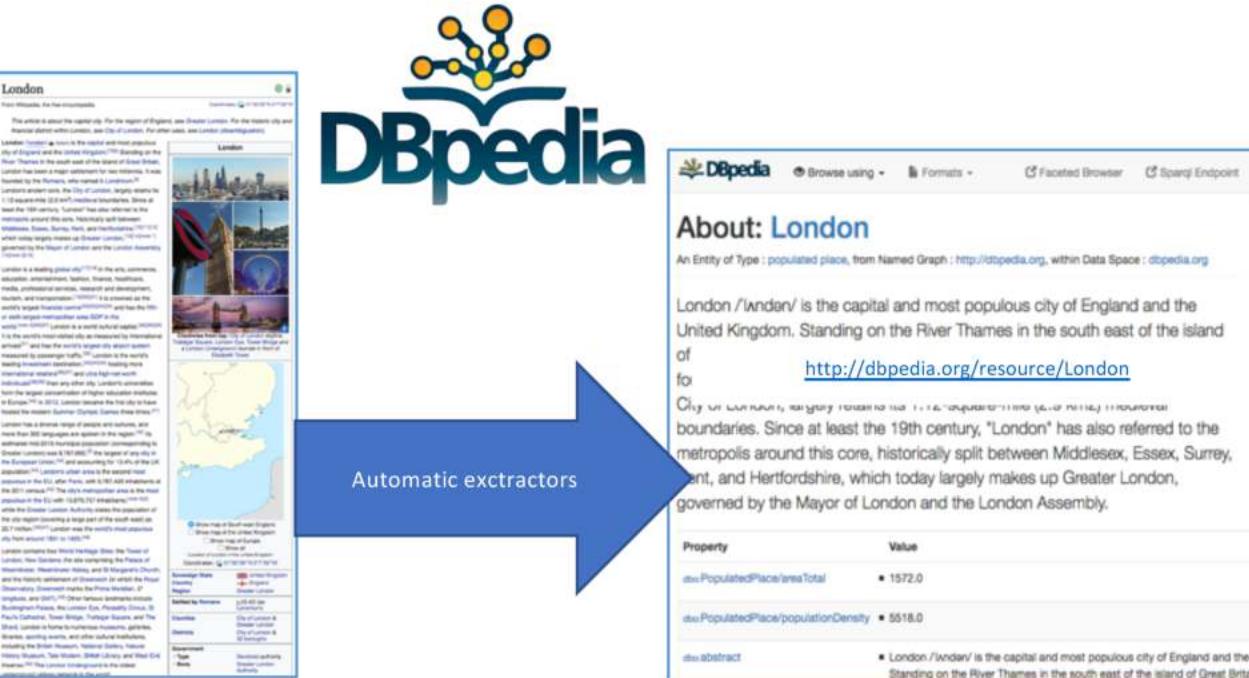
From Semantic Web to Linked (Open) Data

- (ca. 2006/7 – now)
 - Main question: How can I **publish** “Knowledge on the Web” ...
 - Linked Open Data... growth since ~10 years
 - A lot of active developments to publish and link RDF Data



From Semantic Web to Linked (Open) Data

- (ca. 2006/7 – now)
 - Main question: How can I publish “Knowledge on the Web” ...
in order to enable answering structured queries? → SPARQL



Which cities in the UK have more than 1M people?

<http://yasgui.org/short/UVQyhx8ft>

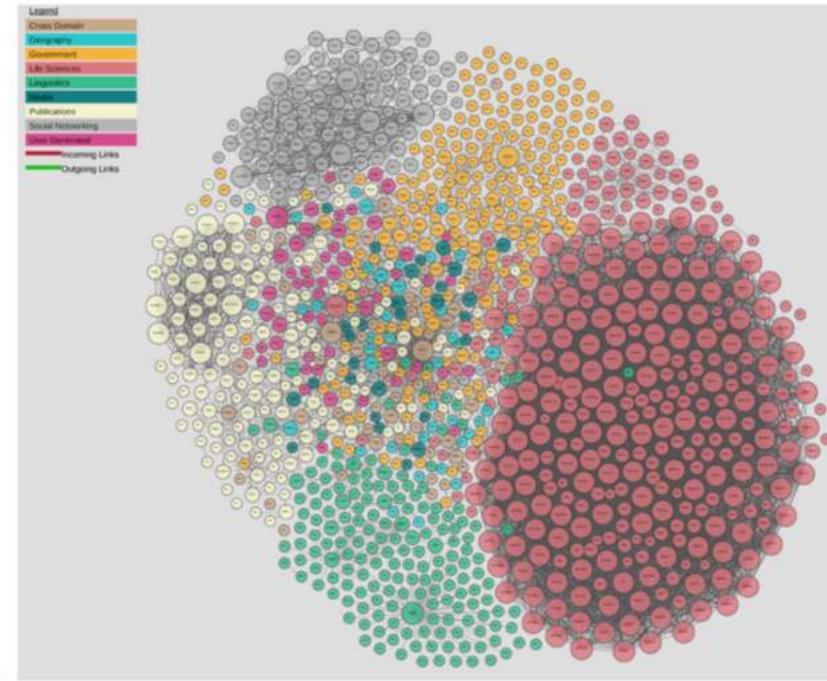
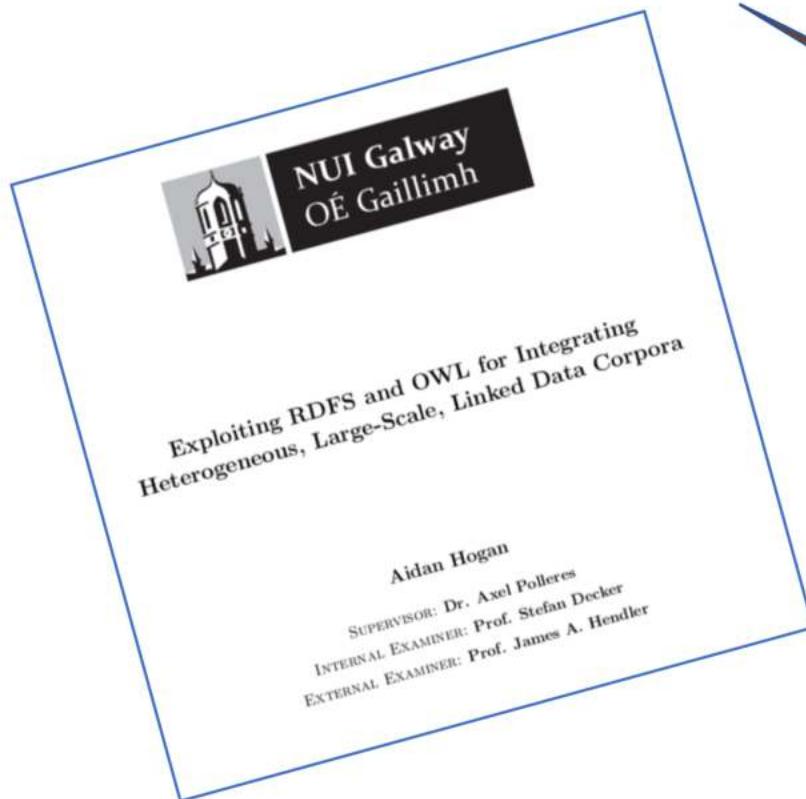
```
PREFIX : <http://dbpedia.org/resource/>
PREFIX dbo: <http://dbpedia.org/ontology/>
PREFIX yago:
<http://dbpedia.org/class/yago/>
```

```
SELECT DISTINCT ?city ?pop WHERE {
  ?city a yago:City108524735 .
  ?city dbo:country :United_Kingdom.
  ?city dbo:populationTotal ?pop}
```

```
FILTER ( ?pop > 1000000 )
}
```

From Semantic Web to Linked (Open) Data

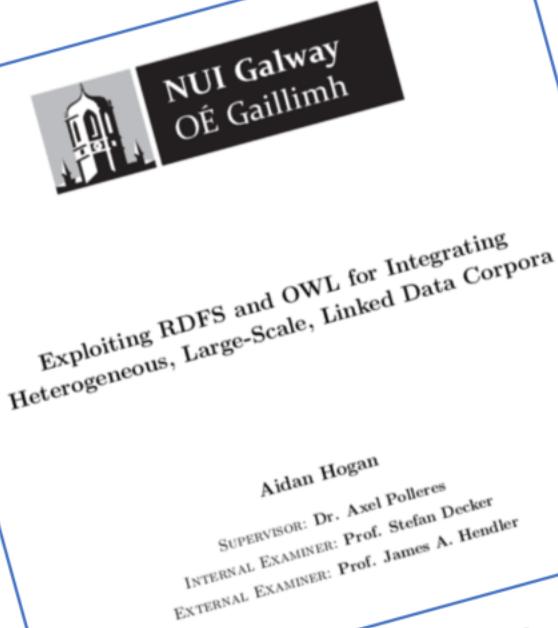
- (ca. 2006/7 – now)
 - Linked Open Data... growth since ~10 years
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How much ontological Reasoning and Expressivity does the **Web** actually need?



Main insight: OWL is (too?) hard



www.w3.org/TR/2009/REC-owl2-direct-semantics-20091027/

2.2 Interpretations

Given a datatype map D and a vocabulary V over D , an interpretation $I = (\Delta_I, \Delta_D, \cdot^C, \cdot^{OP}, \cdot^{DP}, \cdot^I, \cdot^{DT}, \cdot^{LT}, \cdot^{FA})$ for D and V is a 9-tuple with the following structure:

- Δ_I is a nonempty set called the *object domain*.
- Δ_D is a nonempty set disjoint with Δ_I called the *datatype domain* such that $(DT)^{DT} \subseteq \Delta_D$ for each datatype $DT \in V_{DT}$.
- \cdot^C is the *class interpretation function* that assigns to each class $C \in V_C$ a subset $(C)^C \subseteq \Delta_I$ such that
 - $(owl:Thing)^C = \Delta_I$ and
 - $(owl:Nothing)^C = \emptyset$.
- \cdot^{OP} is the *object property interpretation function* that assigns to each object property $OP \in V_{OP}$ a subset $(OP)^{OP} \subseteq \Delta_I \times \Delta_I$ such that
 - $(owl:topObjectProperty)^{OP} = \Delta_I \times \Delta_I$ and
 - $(owl:bottomObjectProperty)^{OP} = \emptyset$.
- \cdot^{DP} is the *data property interpretation function* that assigns to each data property $DP \in V_{DP}$ a subset $(DP)^{DP} \subseteq \Delta_I \times \Delta_D$ such that
 - $(owl:topDataProperty)^{DP} = \Delta_I \times \Delta_D$ and
 - $(owl:bottomDataProperty)^{DP} = \emptyset$.
- \cdot^I is the *individual interpretation function* that assigns to each individual $a \in V_I$ an element $(a)^I \in \Delta_I$.
- \cdot^{DT} is the *datatype interpretation function* that assigns to each datatype $DT \in V_{DT}$ a subset $(DT)^{DT} \subseteq \Delta_D$ such that
 - \cdot^{DT} is the same as D for each datatype $DT \in V_{DT}$, and
 - $(rdfs:Literal)^{DT} = \Delta_D$.
- \cdot^{LT} is the *literal interpretation function* that is defined as $(lt)^{LT} = (LV, DT)^{LS}$ for each $lt \in V_{LT}$, where LV is the lexical form of lt and DT is the datatype of lt .
- \cdot^{FA} is the *facet interpretation function* that is defined as $(F, lt)^{FA} = (F, (lt)^{LT})^{FS}$ for each $(F, lt) \in V_{FA}$.

The following sections define the extensions of \cdot^{OP} , \cdot^{DT} , and \cdot^C to object property expressions, data ranges, and class expressions.

2.2.1 Object Property Expressions

The object property interpretation function \cdot^{OP} is extended to object property expressions as shown in Table 1.

(...to learn, to understand, to implement, to compute, to teach, to represent in RDF, to publish, to parse, to use *appropriately*...)

...for Linked Data publishers

Linked Data publishers only use a little bit of OWL ...

... they still manage to make mistakes ☺

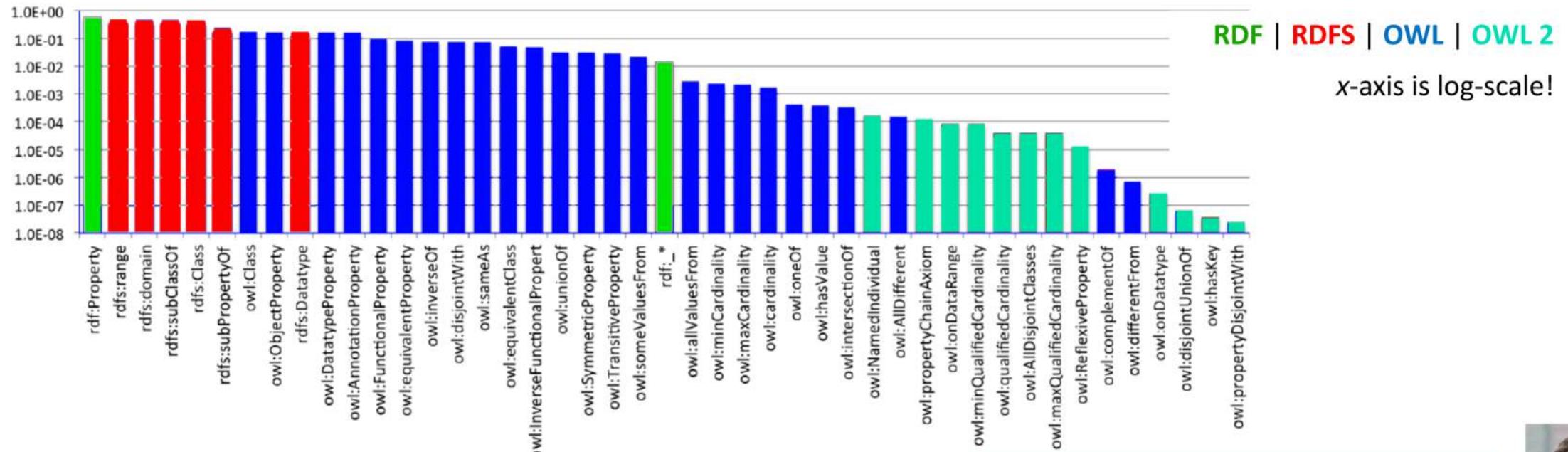
Cautious OWL inference can still help to enrich Linked Data!

How much ontological Reasoning and Expressivity does the Web actually need?



Linked Data publishers only use a little bit of OWL ...

- RDFS features amongst the most prominently used
- OWL 2 features hardly used
- The commonly used features are a fragment of OWL RL (*i.e., fragment of OWL where Description Logics and Datalog Rules coincide*)



RDF | RDFS | OWL | OWL 2

x-axis is log-scale!

How much ontological Reasoning and Expressivity does the Web actually need?



Linked Data publishers still manage to make mistakes ☺



E.g. [DBpedia](#)

Browse using - Formats - Faceted Browser Sparql Endpoint

About: European Union

An Entity of Type : populated place, from Named Graph : <http://dbpedia.org>, within Data Space : dbpedia.org

The European Union (EU) is a politico-economic union of 28 member states that are located primarily in Europe. It has an area of 4,324,782 km² (1,669,808 sq mi), and an estimated population of over 500 million. The EU is a political entity.

- rdf:type**
- [owl:Thing](#)
 - [dbo:Place](#)
 - [dbo:Location](#)
 - [wikidata:Q6256](#)
 - [dbo:Country](#)
 - [dbo:Organisation](#)
 - [dbo:PopulatedPlace](#)
 - [geo:SpatialThing](#)

DBpedia Ontology:

dbo:Agent **owl:disjointWith** dbo:Place.

dbo:Country rdfs:subClassOf dbo:Place.

dbo:Organisation rdfs:subClassOf dbo:Agent.



Other issue: **Ontology “Hijacking”**

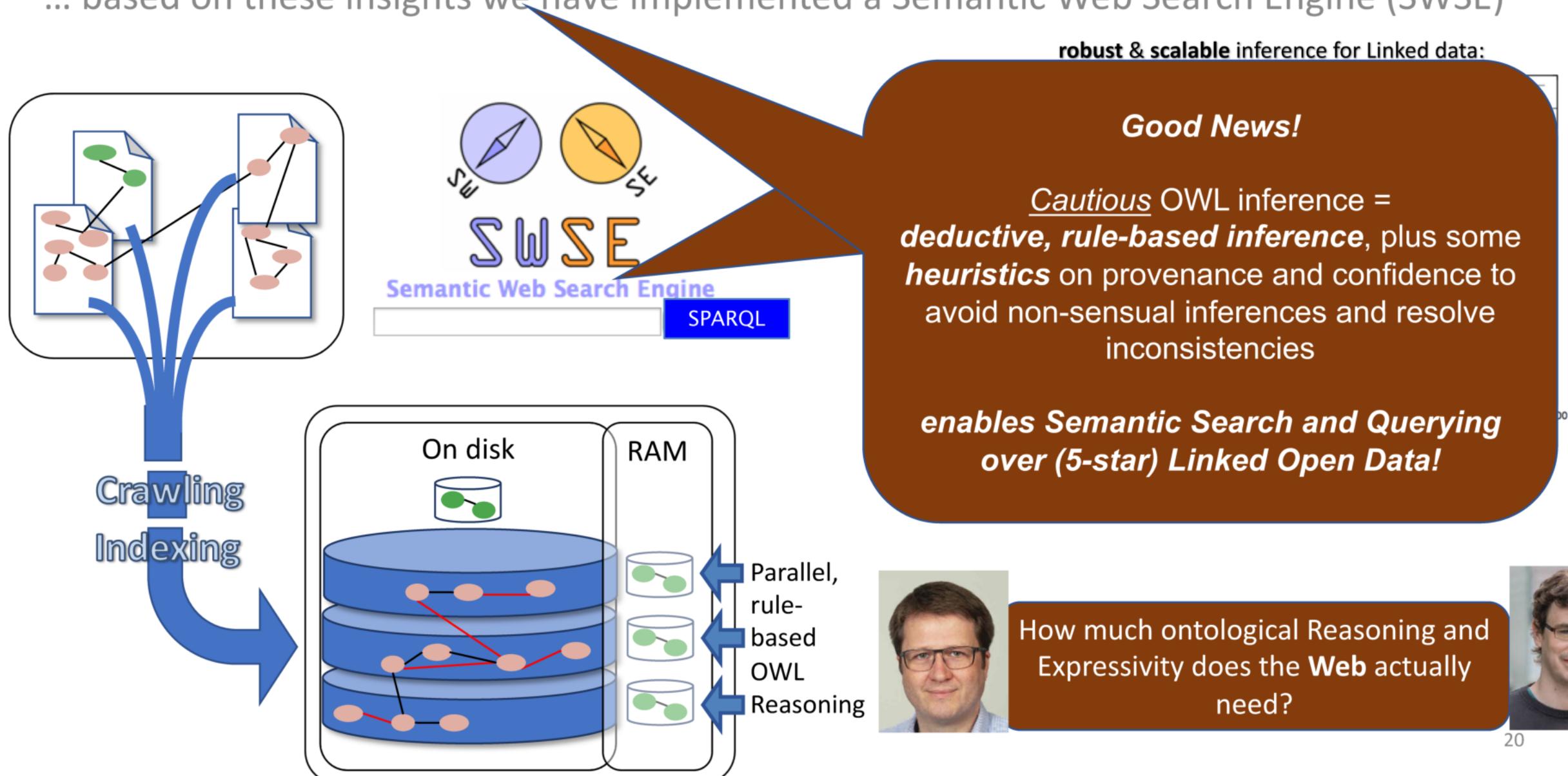
i.e., publishers re-defining others' ontologies by adding inconsistent axioms

How much ontological Reasoning and Expressivity does the **Web** actually need?



Cautious OWL inference can still help to enrich Linked Data and can be implemented in a scalable manner!

... based on these insights we have implemented a Semantic Web Search Engine (SWSE)



From Linked Data to Open Data

- (ca. 2009 - now) Open Data has become a global trend!

EU & Austria, but also the (previous) US and UK administrations are/were pushing Open Data!



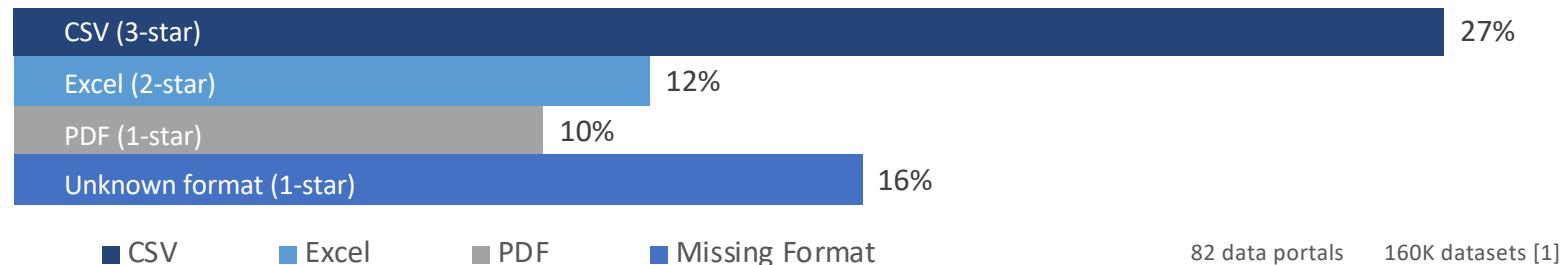
The home of the U.S. Government's open data
Here you will find data, tools, and resources to conduct research, develop web and mobile applications, design data visualizations, and more.

GET STARTED
SEARCH OVER 170,714 DATASETS

Federal Student Loan Program Data

From Linked Data to Open Data

- However... • Available data is only partially structured and not linked [1]:



Non-Linked Open Data is growing still much faster than Linked Open Data on the Web

[1] Umbrich, J., Neumaier, S., Polleres, A.: Quality assessment & evolution of open data portals. International Conference on Open and Big Data (2015)

Open Data as a Global Trend:



Country	URL	Datasets
United States	data.gov	170.7k
Canada	open.canada.ca	79.1k
UK	data.gov.uk	45.1k
France	www.data.gouv.fr	34.2k
Russia	opengovdata.ru	30.3k
Japan	data.go.jp	21k
Italy	dati.gov.it	20.4k
Germany	govdata.de	19.8k

Data portals of the G8 countries

Open Data portals...

 DATA TOPICS - IMPACT APPLICATIONS DEVELOPERS CONTACT

DATA CATALOG

 / Datasets Organizations ?

 /  Department of Housing and ... / US Department of Housing a  


Harvested

Housing Affordability Data System (HADS)

Metadata Updated: March 8, 2017

The Housing Affordability Data System (HADS) is a set of files derived from the 1985 and later national American Housing Survey (AHS) and the 2003 and later Metro AHS. This system categorizes housing units by affordability and households by income, with respect to the Adjusted Median Income, Fair Market Rent (FMR), and poverty income. It also includes housing cost burden for owner and renter households. These files have been the basis for the worst case needs tables since 2001. The data files are available for public use, since they were derived from AHSS public use files and the published income limits and FMRs. These dataset give the community of housing analysts the opportunity to use a consistent set of affordability measures.

Access & Use Information

 **Public:** This dataset is intended for public access and use.

 **License:** No license information was provided. If this work was prepared by an officer or employee of the United States government as part of that person's official duties it is considered a U.S. Government Work.

Downloads & Resources

 Comma Separated Values File  17780 views
[Download](#)

 [HTML](#)

 [Link to file](#)  100% Openness score

Dates

Metadata Created Date	March 7, 2014
Metadata Updated Date	March 8, 2017

Metadata Source

 [Data.json Metadata](#)
[Download Metadata](#)

Harvested from HUD JSON

affordability cost fmr households housing income rent renter

Additional Metadata

Resource Type	Dataset
Metadata Created Date	March 7, 2014
Metadata Updated Date	March 8, 2017
Publisher	US Department of Housing and Urban Development
Unique Identifier	HUD031
Maintainer	Shula.Markland
Maintainer Email	Shula.Markland@HUD.gov

The screenshot shows the data.gv.at website interface. At the top left is the logo "data.gv.at". To the right is a search bar with placeholder text "Suchbegriff (z.B. Finanzen, Wahlen)" and a search button labeled "Suche starten". Below the search bar are navigation links: "Datenkatalog" (highlighted in blue), "Apps & News", and "Katalog durchstöbern". The main menu includes "Startseite", "Daten", "Dokumente", and "Anwendungen".

Katalog Bildungsausgaben

Bildungsausgaben;Regionale Gliederung:Bildungseinrichtung

Daten und Ressourcen

		Entdecke	Entdecke	Entdecke	Entdecke	Entdecke
	OGD_bildungsausgaben_BILDAUS_1					
	OGD_bildungsausgaben_BILDAUS_1_HEADER					
	OGD_bildungsausgaben_BILDAUS_1_C-A10-0					
	OGD_bildungsausgaben_BILDAUS_1_C-BARG-0					
	OGD_bildungsausgaben_BILDAUS_1_C-BABEL1-0					

Titel und Beschreibung Englisch Educational expenditure

Veröffentlichende Stelle Statistik Austria

Datenverantwortliche Stelle Statistik Austria, Guglgasse 13, 1110 Wien, Austria

Kontaktseite der datenverantwortlichen Stelle http://www.statistik.at/web_de/kontakt

Datenverantwortliche Stelle - E-Mailkontakt open.data@statistik.gv.at

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Lizenz Zitat Datenquelle: CC-BY-3.0: Statistik Austria - data.statistik.gv.at

Link zur Lizenz <https://creativecommons.org/licenses/by/3.0/>

Weiterführende Metadaten - Link [\[http://staticube.at/staticube/opendatabase#-debildungsausgaben,http://www.statistik.at/web_de/statistiken/bildung_und_kultur/formales_bildungswesen/bildungsausgaben/index.html,http://www.statistik.at/web_en/statistics/education_culture/formal_education/educational_expenditure/index.html\]](http://staticube.at/staticube/opendatabase#-debildungsausgaben,http://www.statistik.at/web_de/statistiken/bildung_und_kultur/formales_bildungswesen/bildungsausgaben/index.html,http://www.statistik.at/web_en/statistics/education_culture/formal_education/educational_expenditure/index.html)

Veröffentlichende Organisation bzw. Person Statistik Austria

Kategorie Bildung und Forschung, Finanzen und Rechnungswesen, Wirtschaft und Tourismus

Schlagworte Bildungsausgaben

API - Link zu allen Metadaten /api/3/action/package_show?id=71137735-2c65-328f-b57d-be94ada765e

RSS-Feeds für Statistik Austria

geänderte Datensätze

Letzte Änderung 30.04.2018 00:59:46

Uniform metadata, accessible via common APIs (JSON), mostly not yet RDF.

What do you find on Open Data Portals?

The screenshot shows the homepage of the Open Data Portal Österreich. At the top, there's a navigation bar with links for HOME, DATEN, THEMEN, and ANWENDUNGEN. Below this, a search bar contains the text "Leopoldstadt". On the left, there are sections for "Organisationen" and "Gruppen", both stating "Für diese Suche wurden keine [Category] gefunden". The main content area displays search results for "Leopoldstadt" under three categories: Daten & Dokumente, Anwendungen, and News. The "Daten & Dokumente" section shows a single result: "Stadtplan von Anton Behsel 1825" from Wien, dated 29.03.2019, with download options for WMS, GIF, JPEG, and PNG. The "Anwendungen" and "News" sections show no results found.

The screenshot shows the homepage of data.gv.at - Open Data Österreich. The search bar at the top also contains "Leopoldstadt". The results are displayed in three columns: "Suchergebnisse - Daten & Dokumente", "Suchergebnisse - Anwendungen", and "Suchergebnisse - News". The "Daten & Dokumente" column shows the same result as the previous portal: "Stadtplan von Anton Behsel 1825" from Wien, dated 29.03.2019, with download options for WMS, GIF, JPEG, and PNG. The other two columns show no results found.



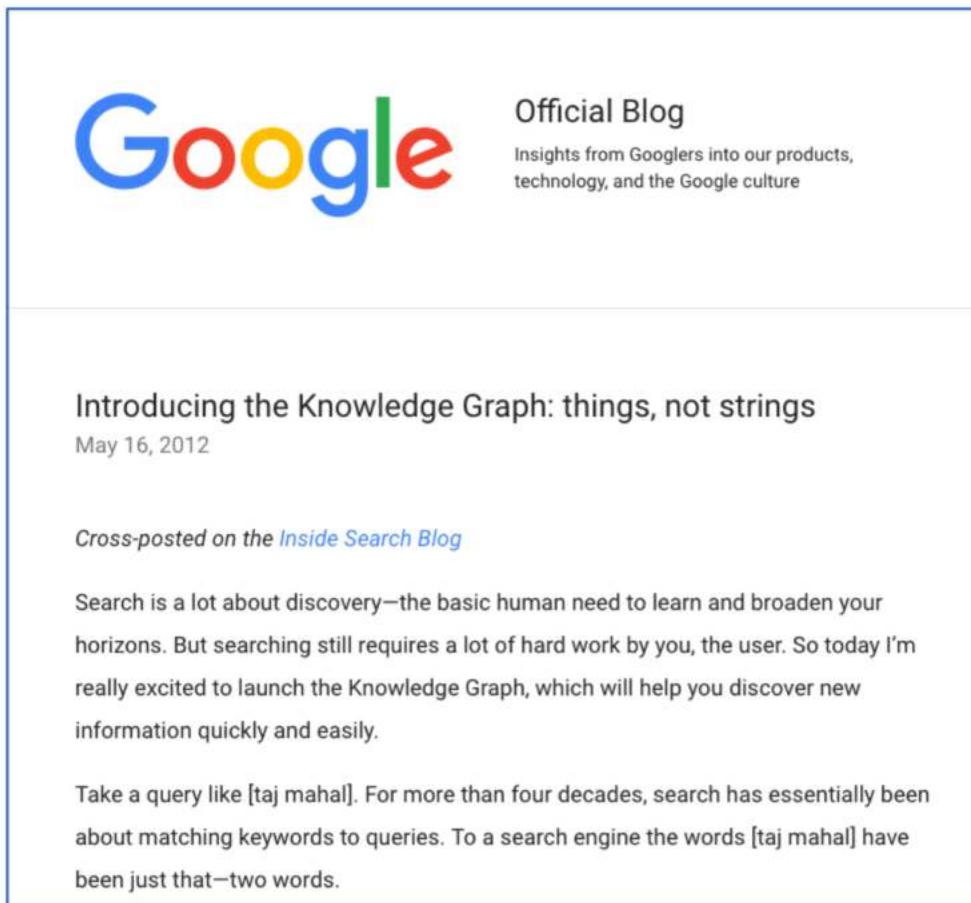
Not too much!

From (Linked) Open Data to Knowledge Graphs

- Needs some more explanation...
- Interlude: ***What is (new) a(bout) Knowledge Graph(s)?***

What is a Knowledge Graph?

- ... good question!



Google
Official Blog
Insights from Googlers into our products, technology, and the Google culture

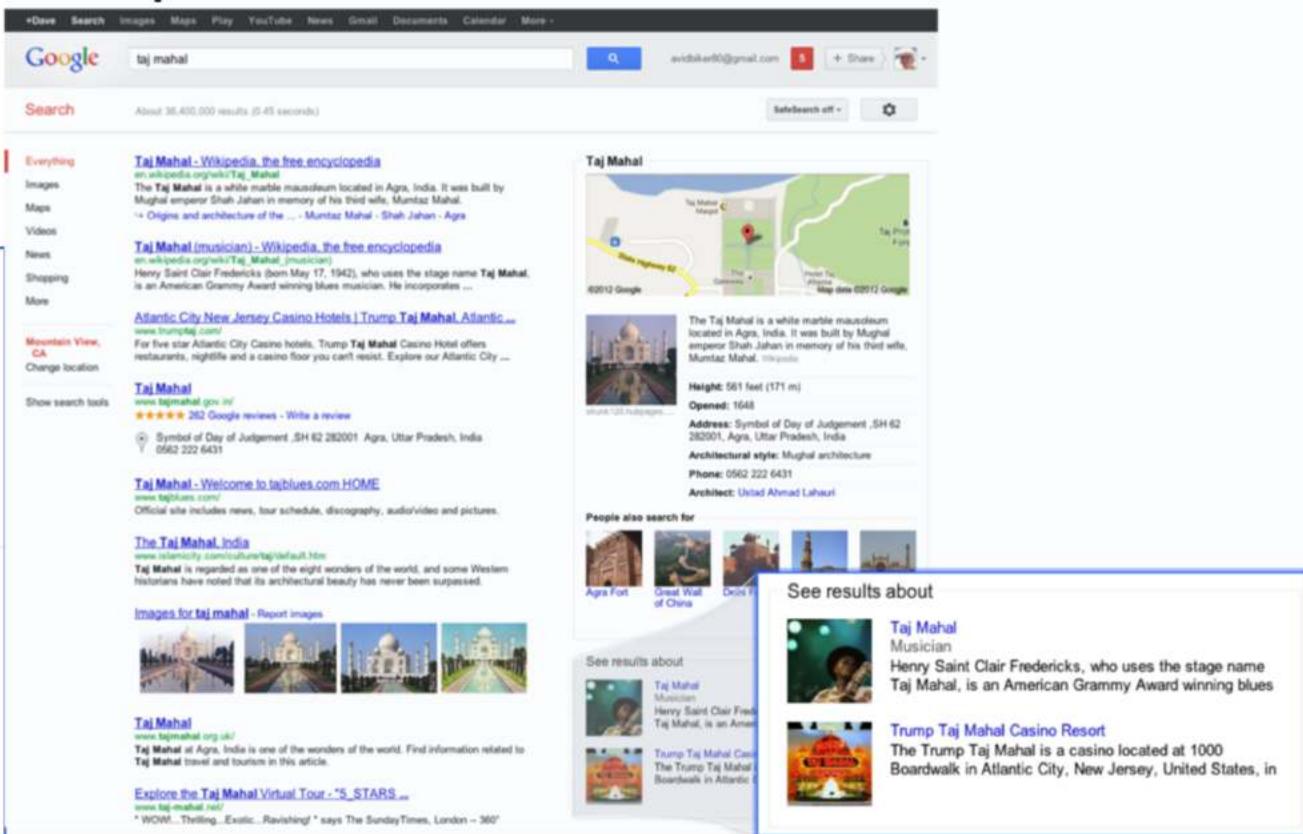
Introducing the Knowledge Graph: things, not strings

May 16, 2012

Cross-posted on the [Inside Search Blog](#)

Search is a lot about discovery—the basic human need to learn and broaden your horizons. But searching still requires a lot of hard work by you, the user. So today I'm really excited to launch the Knowledge Graph, which will help you discover new information quickly and easily.

Take a query like [taj mahal]. For more than four decades, search has essentially been about matching keywords to queries. To a search engine the words [taj mahal] have been just that—two words.



A screenshot of a Google search results page for the query "taj mahal". The results are displayed in a Knowledge Graph format. On the left, there's a sidebar with search filters like "Everything", "Images", "Maps", etc., and a location dropdown set to "Mountain View, CA". Below the sidebar, there's a section for "Taj Mahal" with a link to its Wikipedia page and a snippet about it being a white marble mausoleum in Agra, India. To the right of this is a map showing the location of the Taj Mahal in Agra. Further down, there are sections for "Taj Mahal (musician)", "Atlantic City New Jersey Casino Hotels | Trump Taj Mahal, Atlantic", and "Taj Mahal - Welcome to tajblues.com HOME". On the far right, there's a "See results about" box containing cards for "Taj Mahal Musician" (Henry Saint Clair Fredericks), "Trump Taj Mahal Casino Resort" (The Trump Taj Mahal Casino located at 1000 Boardwalk in Atlantic City, New Jersey, United States), and other related entities like "Agra Fort", "Great Wall of China", and "Delhi Palace".

The announcement says more what a KG **does** than what it **is...**
“[graph of] interesting things and [understanding their] relationships [for search]”

What is a Knowledge Graph? More examples and Definitions:

Other companies with (closed) knowledge graphs:

- [Facebook](#)
- [Bing](#)
- Yandex' Object Answer
- [Baidu](#)
- [LinkedIn](#)
- [Amazon](#)
- [NASA](#)

Some free **open knowledge graphs**:

- [DBpedia](#)
- [WikiData](#)

Some Definitions:

- [McCusker, Chastain, Erickson, and McGuinness. What is a Knowledge Graph?](#) (unpublished, 2016).
 - “principled” aggregation of **Linked Data**? p.7
- [Rospocher, van Erp, Vossen, Fokkens, Aldabe, Rigau, Soroa, Ploeger, Bogaard. Building event-centric knowledge graphs from news.](#) JWS (2016)
 - “knowledge-base of facts about entities typically [Remark: often automatically] obtained from structured repositories [such as Freebase]”
- [Lisa Ehrlinger and Wolfram Wöß Towards a Definition of Knowledge Graphs \(SEMANTiCS2016\)](#)
 - “A knowledge graph acquires and integrates information into an ontology and applies a reasoner to derive new knowledge.”

What is a Knowledge Graph? More examples and Definitions:

Other companies with (closed) knowledge graphs:

- [Facebook](#)
- [Bing](#)
- [Yandex' C](#)
- [Baidu](#)
- [LinkedIn](#)
- [Amazon](#)
- [NASA](#)

The screenshot shows the Schloss Dagstuhl website. At the top is a yellow header bar with the text "SCHLOSS DAGSTUHL" and "Leibniz-Zentrum für Informatik". Below the header is a navigation bar with tabs "About Dagstuhl", "Program" (which is highlighted in yellow), and "Publications". A sidebar on the left lists "Dagstuhl Seminars", "Dagstuhl Perspectives", "GI-Dagstuhl Seminars", "Events", "Research Guests", "Seminar Calendar", and "All Events". The main content area displays information about a seminar: "September 9 – 14, 2018, Dagstuhl Seminar 18371" titled "Knowledge Graphs: New Directions for Knowledge Representation on the Semantic Web". It lists "Organizers" including Piero Andrea Bonatti, Stefan Decker, Axel Polleres, and Valentina Presutti. Below the text is a group photo of seminar participants standing in front of a building.

Some free ones:

- [DBpedia](#)
- [WikiData](#)

Some Definitions:

10 (Re)Defining Knowledge Graphs

Aidan Hogan (IMFD, DCC, University of Chile - Santiago de Chile, CL), Dan Brickley (Google Research - Mountain View, US), Claudio Gutierrez (IMFD, DCC, University of Chile - Santiago de Chile, CL), Axel Polleres (Wirtschaftsuniversität Wien, AT), and Antoine Zimmermann (École des Mines de Saint-Étienne, FR)

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The phrase “Knowledge Graph” has recently gained a lot of attention in both industry and academia. But what is a “Knowledge Graph”? Several definitions have been proposed but – we shall argue – fall short of capturing the full generality of the usage of the term. We argue for a looser, more permissive definition that may be instantiated in various concrete ways, setting the stage for the study and practice of “Knowledge Graphs” to become a commons that unites – rather than divides – previously disparate areas of Computer Science, focused

We thus propose to define a “Knowledge Graph”, succinctly, as:

“a graph of data with the intent to compose knowledge”.

as nodes, with their relations represented as edges; technically this notion can be instantiated with a number of concrete graph models, including for example:

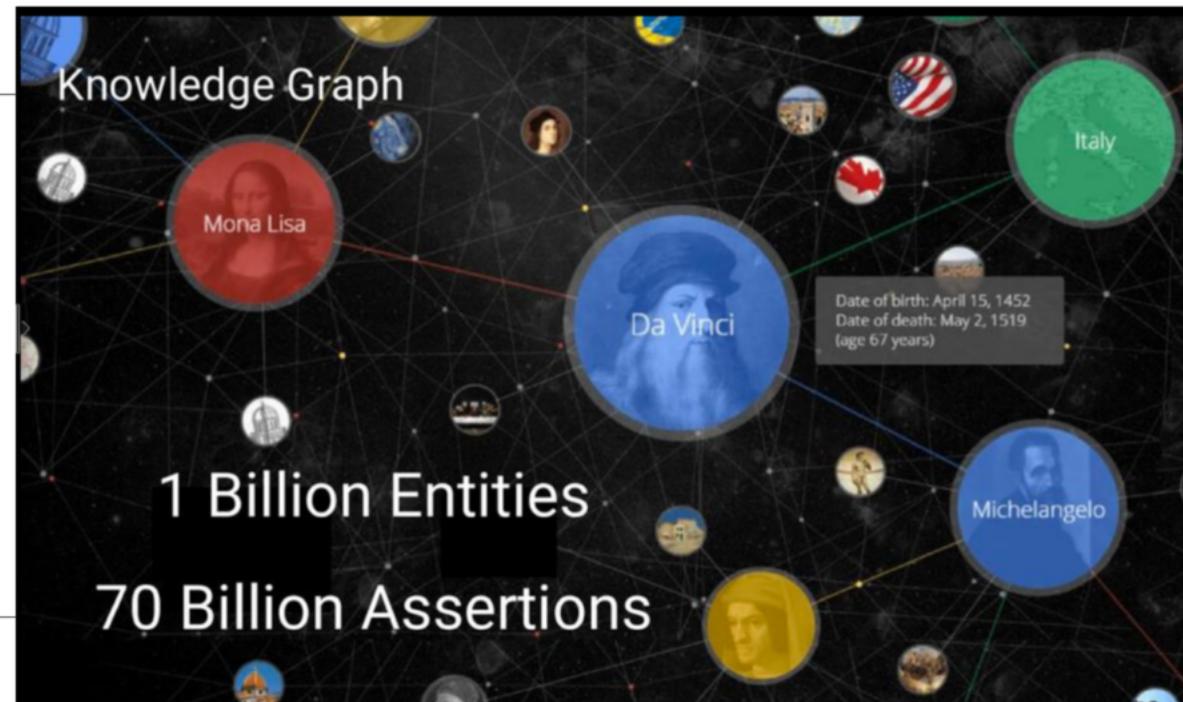
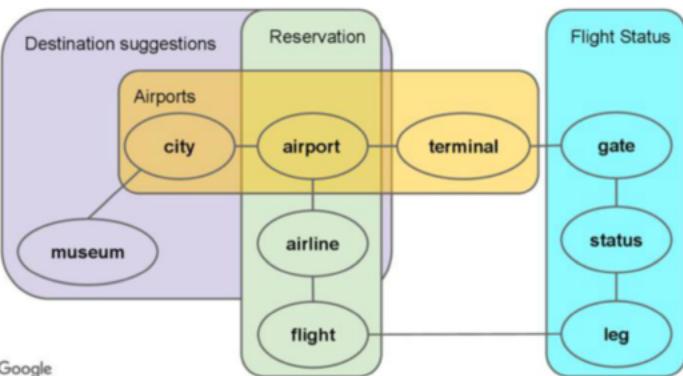
- **directed edge-labelled graphs** (aka sets of triples), composed of named binary relations (labelled edges) between entities (nodes);
- **property graphs**, which extends directed edge-labelled graphs such that both nodes and edges may be additionally annotated with sets of property-attribute pairs;
- **named graphs**, where rather than supposing one large graph, data are represented as a collection of (typically directed edge-labelled) graphs, each associated with an identifier.

What is new/different about Knowledge Graphs?

- Jamie Taylor, Google, Inc., Keynote [ISWC2017](#)



The Power of Knowledge Graph: Interlocking data



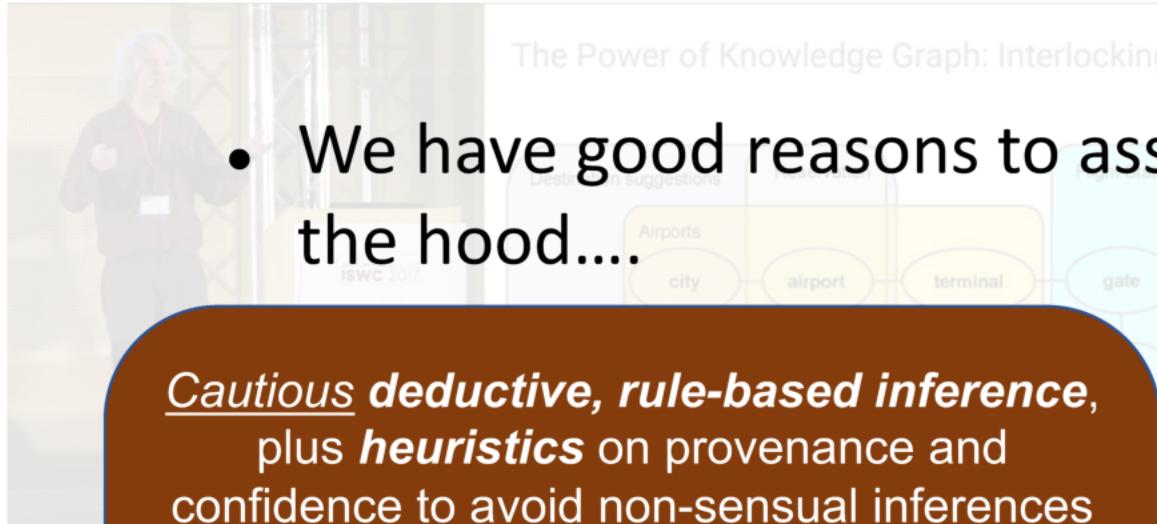
- Actors, Directors, Movies
- Art Works & Museums
- Cities & Countries
- Islands, Lakes, Lighthouses
- Music Albums & Music Groups
- Planets & Spacecraft
- Roller Coasters & Skyscrapers
- Sports Teams

[...]

*Answer whether (something like)
RDF and/or triple stores are used
under the hood answered
vaguely...*

What is new/different about Knowledge Graphs?

- Jamie Taylor, Google, Inc., Keynote [ISWC2017](#)



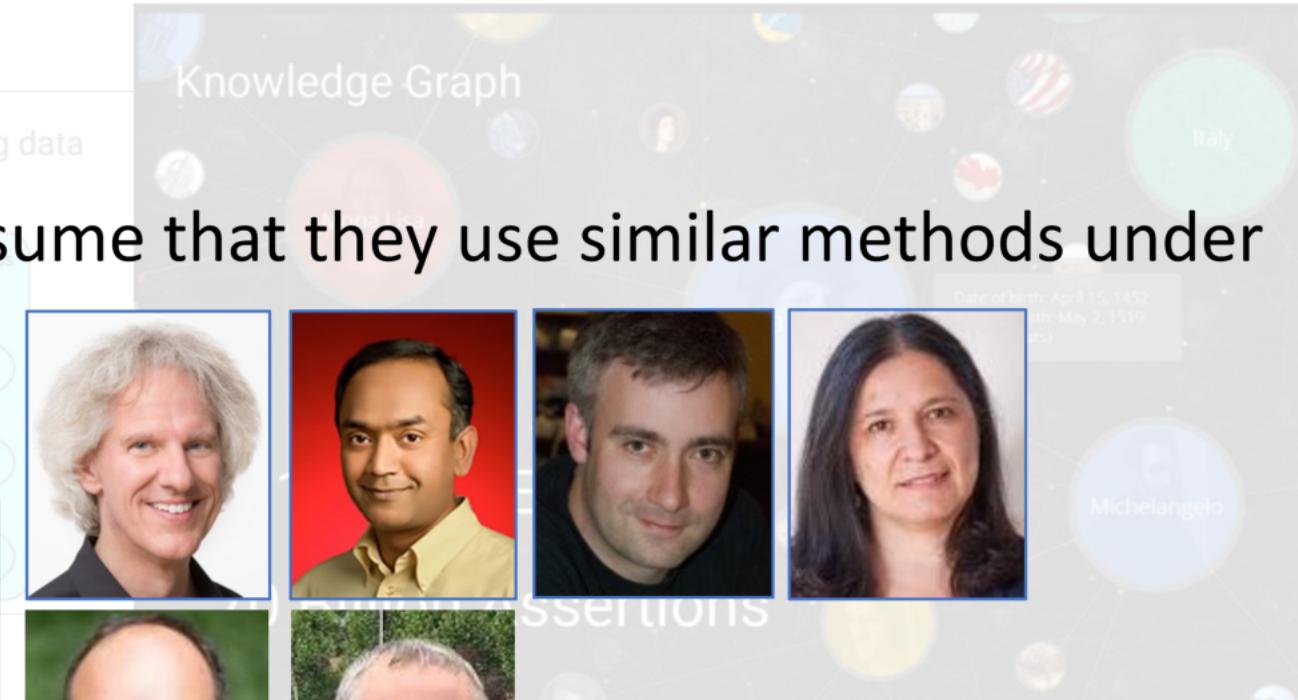
- We have good reasons to assume that they use similar methods under the hood....

Cautious deductive, rule-based inference, plus **heuristics** on provenance and confidence to avoid non-sensual inferences and resolve inconsistencies works.

What's (probably) new?

statistical methods/learning more central:

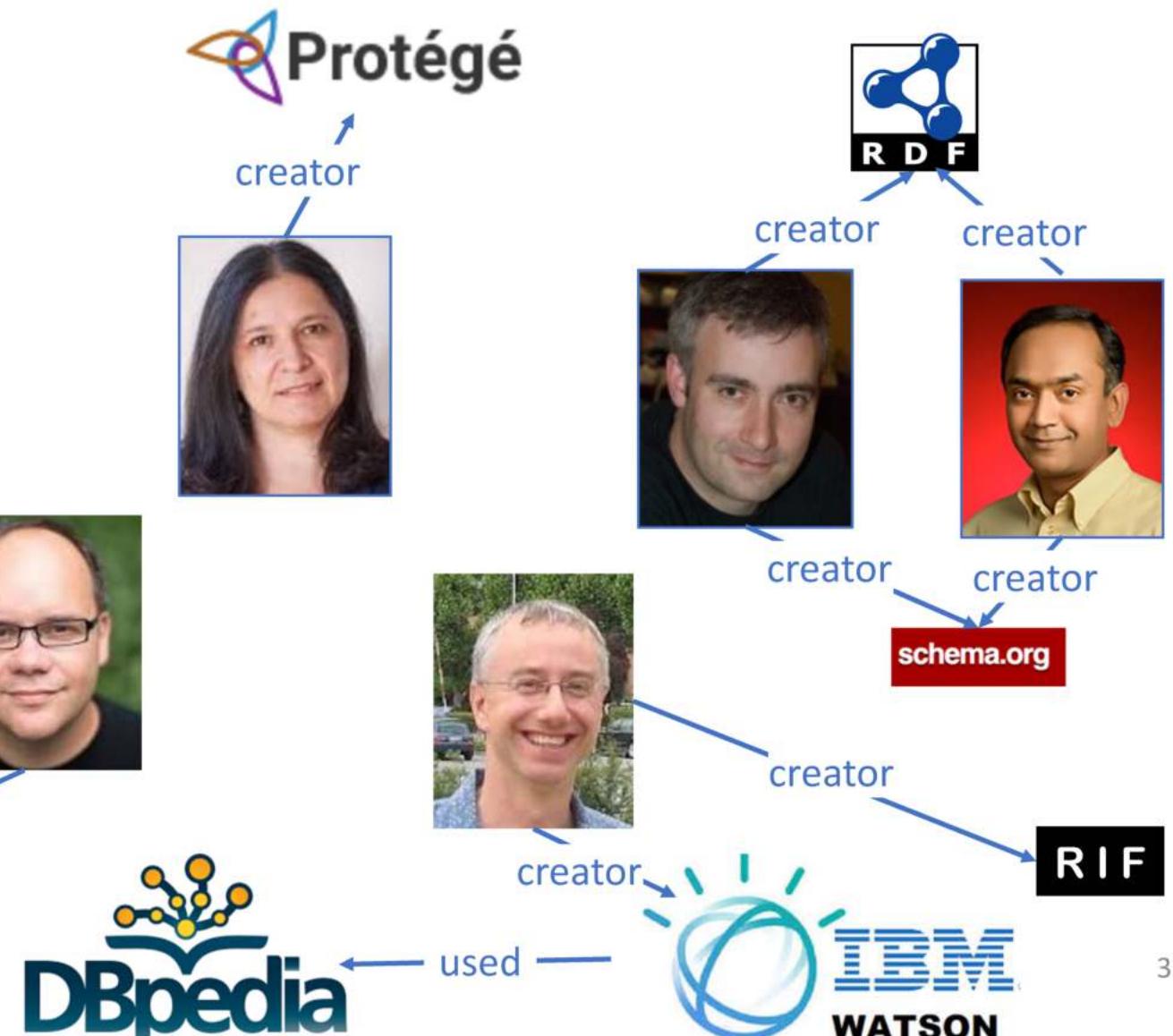
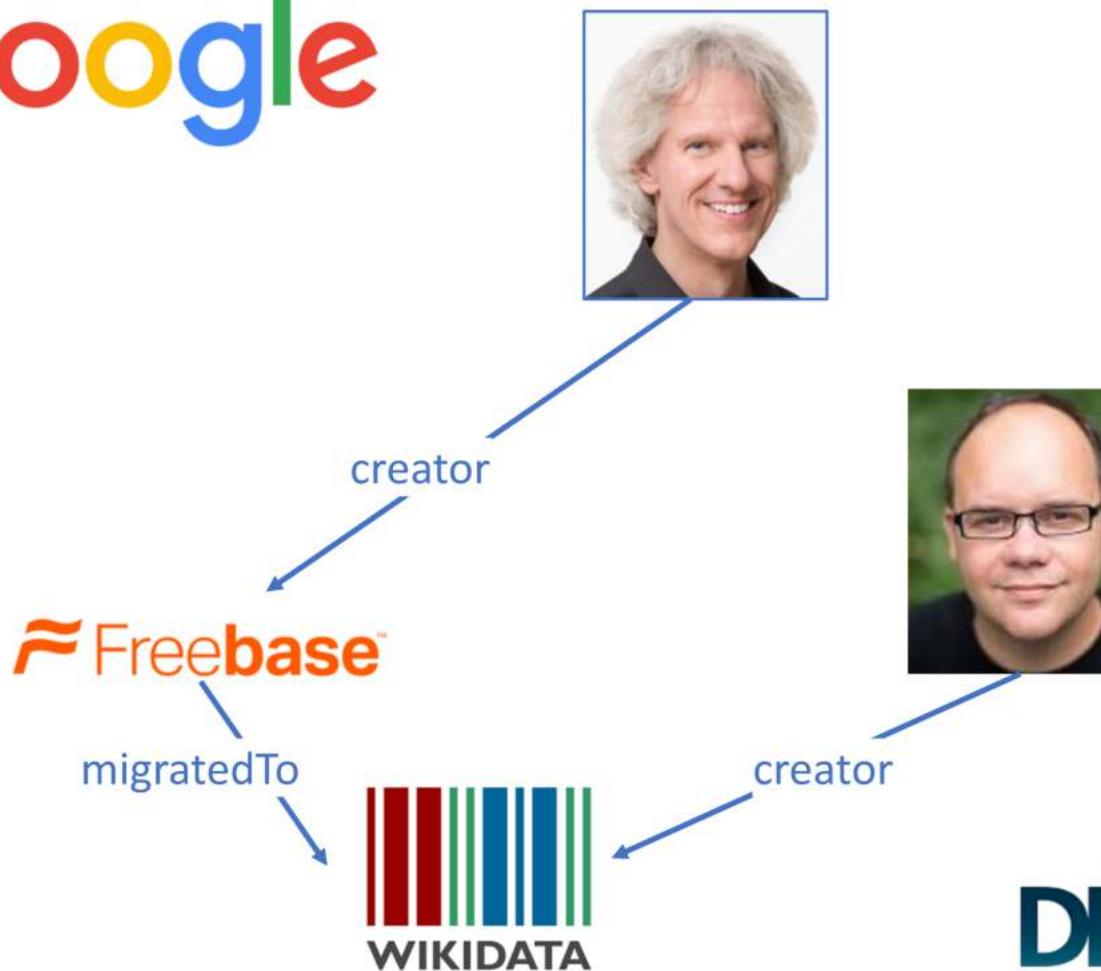
- lots of data enables
 - more accurate confidence scores
 - rule mining (not restricted to OWL)



Answer whether (something like) RDF and/or triple stores are used under the hood answered vaguely...

- We have good reasons to assume that they use similar/ the same methods under the hood...

Google



- ... and have extended/improved them!

Google search results for "cities in the UK with more than 1M people". The search bar shows the query. Below it, the navigation bar includes All, Images, News, Shopping, Videos, More, Settings, and Tools. A message indicates About 37,900,000 results found in 0.55 seconds. A sidebar on the left lists cities from citymetric.com, showing thumbnails for Birmingham, Manchester, Liverpool, Sheffield, and Glasgow. Below this, a section titled "Metropolitan areas" lists the following population figures:

- London – 13,709,000.
- Birmingham-Wolverhampton – 3,683,000.
- Manchester – 2,556,000.
- Leeds-Bradford – 2,302,000.
- Liverpool-Birkenhead – 2,241,000.
- Newcastle-Sunderland – 1,599,000.
- Sheffield – 1,569,000.
- Southampton-Portsmouth – 1,547,000.

Which cities in the UK have more than 1M people?

<http://yasgui.org/short/UV0yhX8ft>

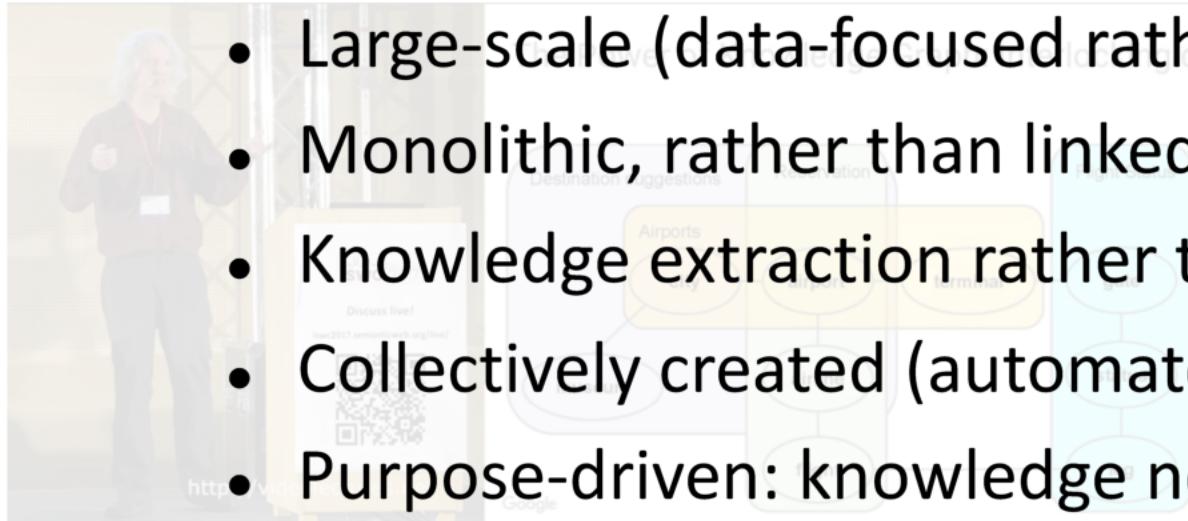
```
PREFIX : <http://dbpedia.org/resource/>
PREFIX dbo: <http://dbpedia.org/ontology/>
PREFIX yago:
<http://dbpedia.org/class/yago/>
```

```
SELECT DISTINCT ?city ?pop WHERE {
?city a yago:City108524735 .
?city dbo:country :United_Kingdom.
?city dbo:populationTotal ?pop

FILTER ( ?pop > 1000000 )
}
```

Summary: What is new/different about Knowledge Graphs?

- Jamie Taylor, Google, Inc., Keynote [ISWC2017](#)



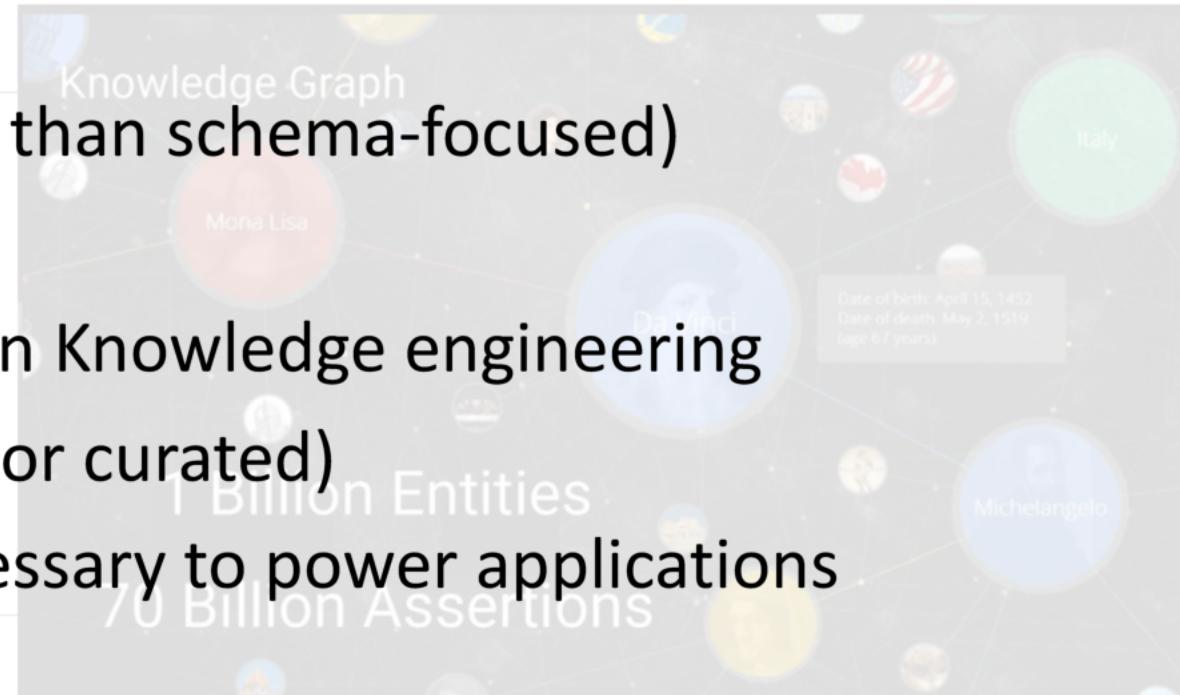
- Large-scale (data-focused rather than schema-focused)
- Monolithic, rather than linked
- Knowledge extraction rather than Knowledge engineering
- Collectively created (automated or curated)
- Purpose-driven: knowledge necessary to power applications
- (Logical) consistency not a must

• **Ontological expressivity not central – BUT: Expressing context is!**

- Art Works & Museums
- Cities & Countries
- Islands, Lakes, Lighthouses

- Planets & Spacecraft
- Roller Coasters & Skyscrapers
- Sports Teams

[...]



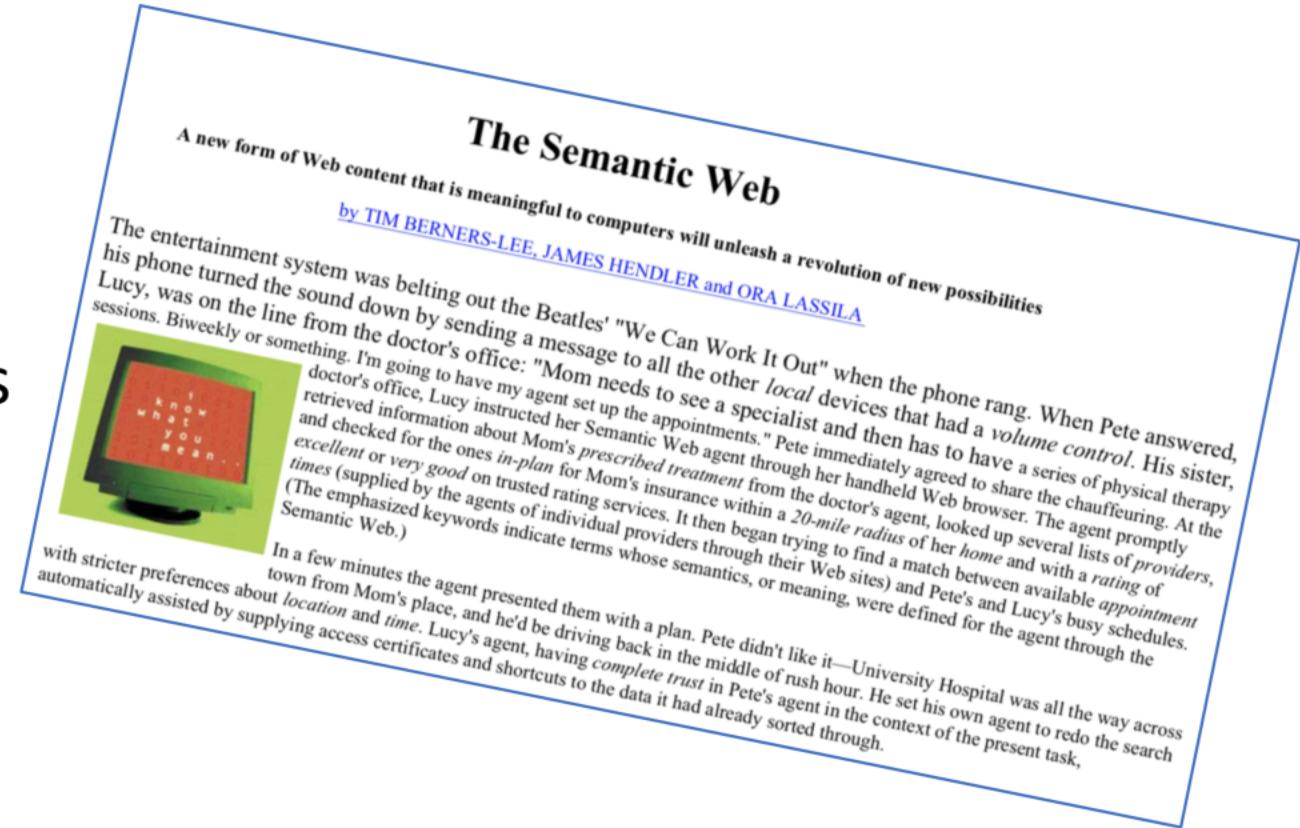
Answer whether (something like
RDF and SPARQL) can be used to
unambiguously identify the entity

- For instance:
- Provenance
 - Temporal context
 - Confidence

From Knowledge Graphs (back) to Semantic Web

What closed KGs already enable:

- Semantic Search
- Appointment detection in emails
- Ratings of products/services



However,

- most of these KGs and applications are closed
- they often cover either only generic, or domain-specific “knowledge”

From Knowledge Graphs (back) to Linked (Open) Data:

Good news:

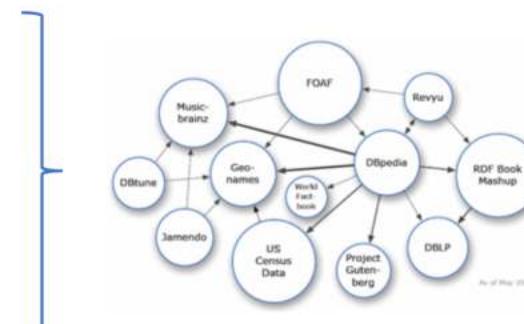
- *There are also open KGs & They use RDF, SPARQL & Linked Data!*
 - *Open Knowledge Graphs can be used to link Open Data!*
 - *Open Knowledge Graphs can be used for many other AI applications!*

Bad news:

- *Open KGs don't/only partially cover Knowledge in Open Data!*
 - *Decentralised Open KGs are hard to maintain and serve*
 - *Managing context makes things harder!*

Some free open knowledge graphs:

- [Dbpedia](#)
 - [WikiData](#)
 - [GeoNames](#)
 - ...



From Knowledge Graphs (back) to Linked Open Data:

- There are also open KGs & They use RDF, SPARQL & Linked Data!

Wikidata as RDF ... can be queried by SPARQL

- “Simple” surface [query](#):

Which cities in the UK have more than 1M people?

```
SELECT DISTINCT ?city WHERE {
    ?city wdt:P31/wdt:P279* wd:Q515 .
    ?city wdt:P1082 ?population .
    ?city wdt:P17 wd:Q38 .
    FILTER (?population > 1000000) }
```

instance of (P31)
that class of which this subject is
a particular example and
member. (Subject typically an
individual member with Proper
Name label.) Different from P279
(subclass of).

subclass of (P279)
all instances of these items are
instances of those items; this
item is a class (subset) of that
item. Not to be confused with
Property:P31 (instance of).

city (Q515)
large and permanent human
settlement

population (P1082)
number of people inhabiting the
place; number of people of
subject

country (P17)
sovereign state of this item
United Kingdom (Q145)
country in Europe

- What's this?

From Knowledge Graphs (back) to Linked Open Data:

- Managing context makes things harder! → but is also doable!

- However, Wikidata has more complex info:
(temporal context, provenance,...)

Which cities in the UK have reached 1M in which year?

... Can I query that with SPARQL?

Yes!

The screenshot shows the Wikidata Query Service interface. At the top, there are navigation links for 'Examples', 'Help', and 'More tools'. Below the header, a sidebar on the left contains icons for search, history, and other functions. The main area displays a SPARQL query and its results.

SPARQL Query:

```
1 SELECT ?city (min(?time) as ?year) WHERE {
2   ?city wdt:P31/wdt:P279* wd:Q515.
3   ?city wdt:P17 wd:Q38 .
4   ?city p:P1082 ?statement .
5   ?statement <http://www.wikidata.org/prop/statement/value/P1082> ?value .
6   ?statement <http://www.wikidata.org/prop/qualifier/P585> ?time .
7   ?value <http://wikiba.se/ontology#quantityAmount> ?population .
8   FILTER (?population > 1000000 )
9 } GROUP BY ?city
```

Results:

population	point in time	determination method	reference URL
8,416,535±0	2012	estimation	http://www.ons.gov.uk/ons/repop-estimate/population-estimates-for-england-and-wales/mid-2012/mid-2012-population-estimates-for-england-and-wales.html
1,011,157±0	1801	census	http://www.visionofbritain.org.uk/data_cube_page.jsp?data_item=e=T_POP&data_cube=N_TOT_POP&u_id=10097836&c_id=10001043&add=N_g
1,811	1811	census	http://www.visionofbritain.org.uk/data_cube_page.jsp?data_item=e=T_POP&data_cube=N_TOT_POP&u_id=10097836&

BTW, seemingly not yet doable in **Google**

Which cities in the UK have reached 1M in which year?



From Knowledge Graphs (back) to Linked Open Data:

- *(Open) Knowledge Graphs can be used to link Open Data!*

Recall: What do you find on Open Data Portals?

odp DATA PORTAL ÖSTERREICH

HOME DATEN THEMEN ANWENDUNGEN

OpenDataPortal Österreich > Datenkatalog > Datensätze

▼ Organisationen
Für diese Suche wurden keine Organisationen gefunden.

▼ Gruppen
Für diese Suche wurden keine Gruppen gefunden.

Leopoldstadt

Keine Datensätze gefunden bei der Suche "Leopoldstadt"

data.gv.at - Open Data Österreich

Startseite Daten Dokumente Anwendungen Infos News

Suche

Leopoldstadt

Suchergebnisse - Daten & Dokumente

Stadtplan von Anton Behsel 1825 29.03.2019
Wien

Stadt Wien/Magistratsabteilung 8 - Wiener Stadt- und Landesarchiv
Georeferenzierter Stadtplan der Inneren Stadt und der Vorstädte (Bezirke innerhalb des Gürtels sowie...)

WMS GIF JPEG PNG

Alle Datensätze Alle Dokumente

Suchergebnisse - Anwendungen

Keine Anwendungen gefunden!

Alle Anwendungen

Suchergebnisse - News

Keine News gefunden!

Alle News



Not too much!

From Knowledge Graphs (back) to Linked Open Data:

- (Open) Knowledge Graphs can be used to link Open Data!

The diagram illustrates the integration of knowledge graphs and linked open data. On the left, a slide titled "Google Dataset Search: Building a search engine for datasets in an open Web ecosystem" lists four individuals and their contact information. A blue line connects this slide to the Google Dataset Search interface on the right. The search results for "Leopoldstadt" show three datasets from Statista, each with a detailed description and a "Explore at de.statista.com" button. The interface includes a "schema.org" badge and a "Beta" indicator.

Google Dataset Search: Building a search engine for datasets in an open Web ecosystem

Dan Brickley
danbri@google.com
Google
Mountain View, California

Matthew Burgess
matburg@google.com
Google AI
Mountain View, California

Natasha Noy
noy@google.com
Google AI
Mountain View, California

Leopoldstadt

schema.org

16 results found

statista Mitarbeiter der Erste Group nach Ländern 2018
de.statista.com

statista Cost-Income-Ratio der Erste Group bis 2018
de.statista.com

statista Betriebserträge und -aufwendungen der Erste Group bis 2018
de.statista.com

Mitarbeiter der Erste Group nach Ländern 2018

Explore at de.statista.com

Dataset provided by
Statista

Time period covered
2018

Area covered
Österreich

Description

Die Statistik zeigt die Anzahl der Mitarbeiter der Erste Group nach Ländern im Jahr 2018. Die Erste Group Bank AG wurde im Jahr 1819 in der Leopoldstadt, einer Vorstadt Wiens, als "Erste österreichische Spar-Casse" gegründet. Heute ist sie eine der größten Bankengruppen in Zentral- und



A good start, but not much better either!

From Knowledge Graphs (back) to Linked Open Data:

- (Open) Knowledge Graphs can be used to link Open Data!

The image is a collage of screenshots from various datasets and search engines, illustrating the state of linked open data. It includes:

- A screenshot of the Google Dataset Search interface, showing a search bar and a "schema.org" logo.
- A screenshot of the data.gv.at catalog, featuring a search bar and a "Startseite" button.
- A screenshot of the eurostat website, showing a navigation menu and a featured image of a man.
- A screenshot of the HADS (Housing Affordability Data System) dataset page, showing detailed information about the dataset and access/use information.
- A screenshot of the US Department of Housing and Urban Development's dataset page, showing a sidebar with topics like BusinessUSA, Consumer, Energy, and Finance.
- A large blue diagonal banner across the top left containing text and contact information for Dan Brickley, Matthew Burgess, and Natasha Noy.
- Red arrows pointing from the eurostat and data.gv.at screenshots towards the schema.org logo in the Google Dataset Search screenshot.
- A large black outline of a sad face icon on the right side of the image.
- Text overlaid on the right side: "A good start, but not much better either!"
- A large question at the bottom: "Can we do better (using KGs)?"



From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

Example Open Data Table:

<i>federal state</i>	<i>district</i>	<i>year</i>	<i>sex</i>	<i>population</i>
Upper Austria	Linz	2013	male	98157
Upper Austria	Steyr	2013	male	18763
Upper Austria	Wels	2013	male	29730
...



From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

Open Data CSVs look more like this

NUTS2	LAU2_NAME	YEAR	SEX	P_TOTAL
AT31	Linz	2013	1	98157
AT31	Steyr	2013	1	18763
AT31	Wels	2013	1	29730
...



From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

What can we do to make Open Data better searchable?

- Particularly **temporal** and **geospatial** search requires better support [2]

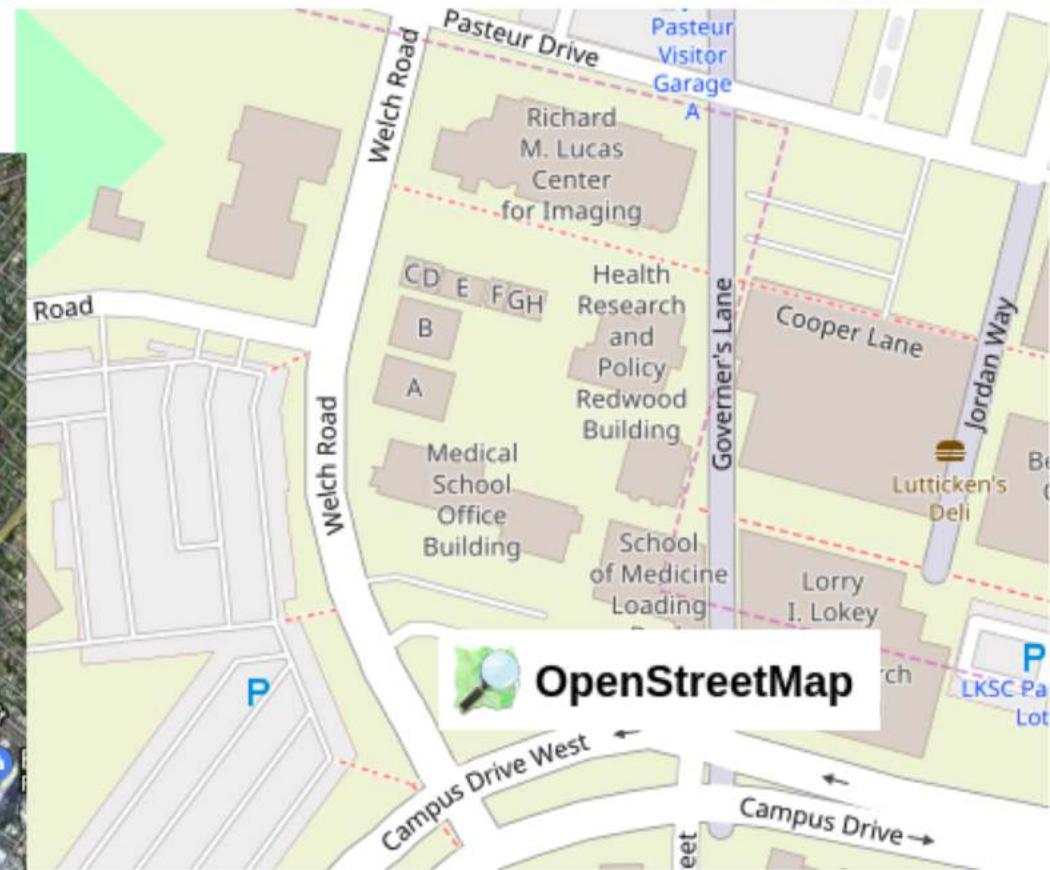
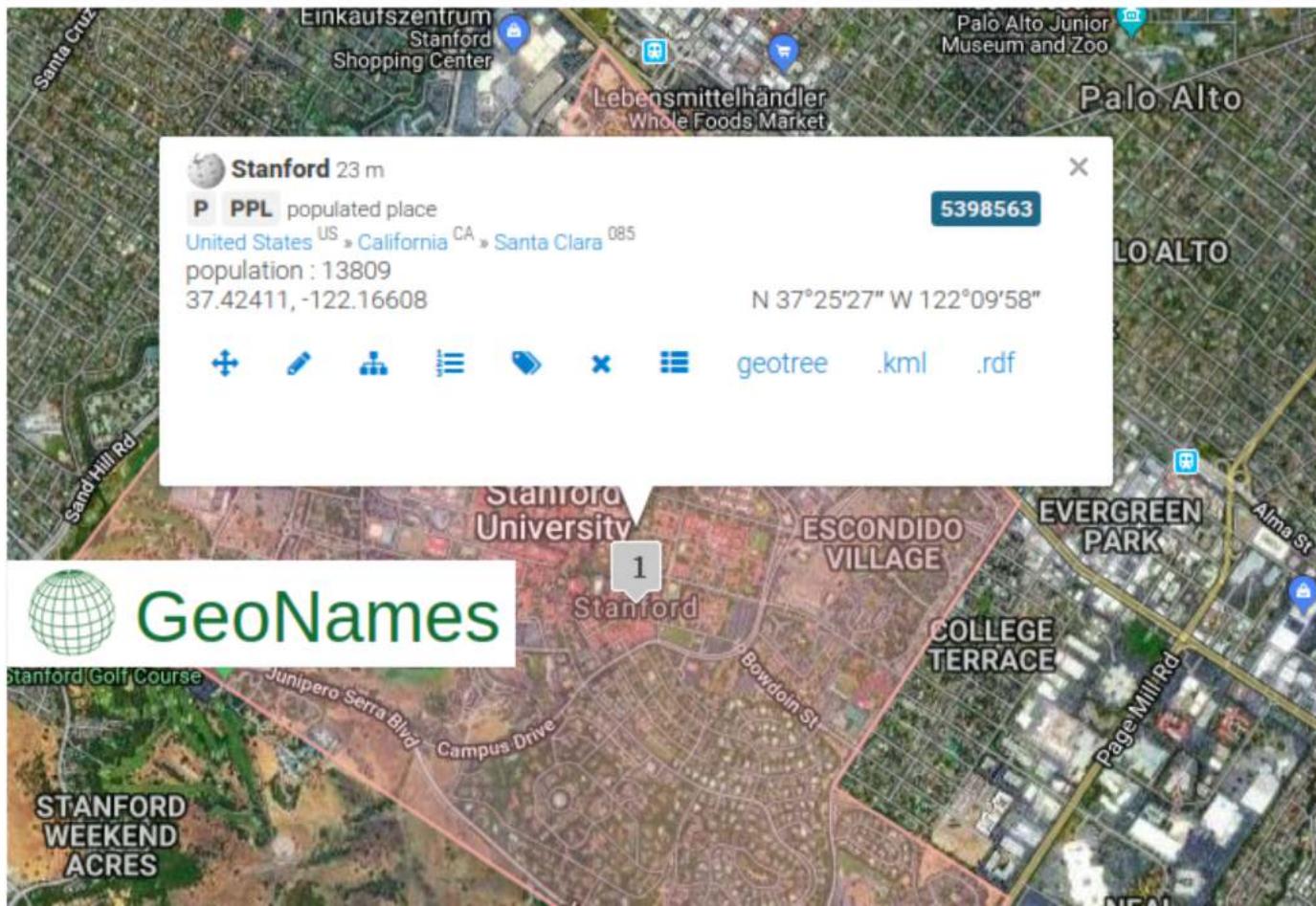
NUTS2	LAU2_NAME	YEAR	SEX	AGE_TOTAL
AT31	Linz	2013	1	98157
AT31	Steyr	2013	1	18763
AT31	Wels	2013	1	29730
...



From Knowledge Graphs (back) to Linked Open Data:

- Open Knowledge Graphs can be used to **link Open Data!**

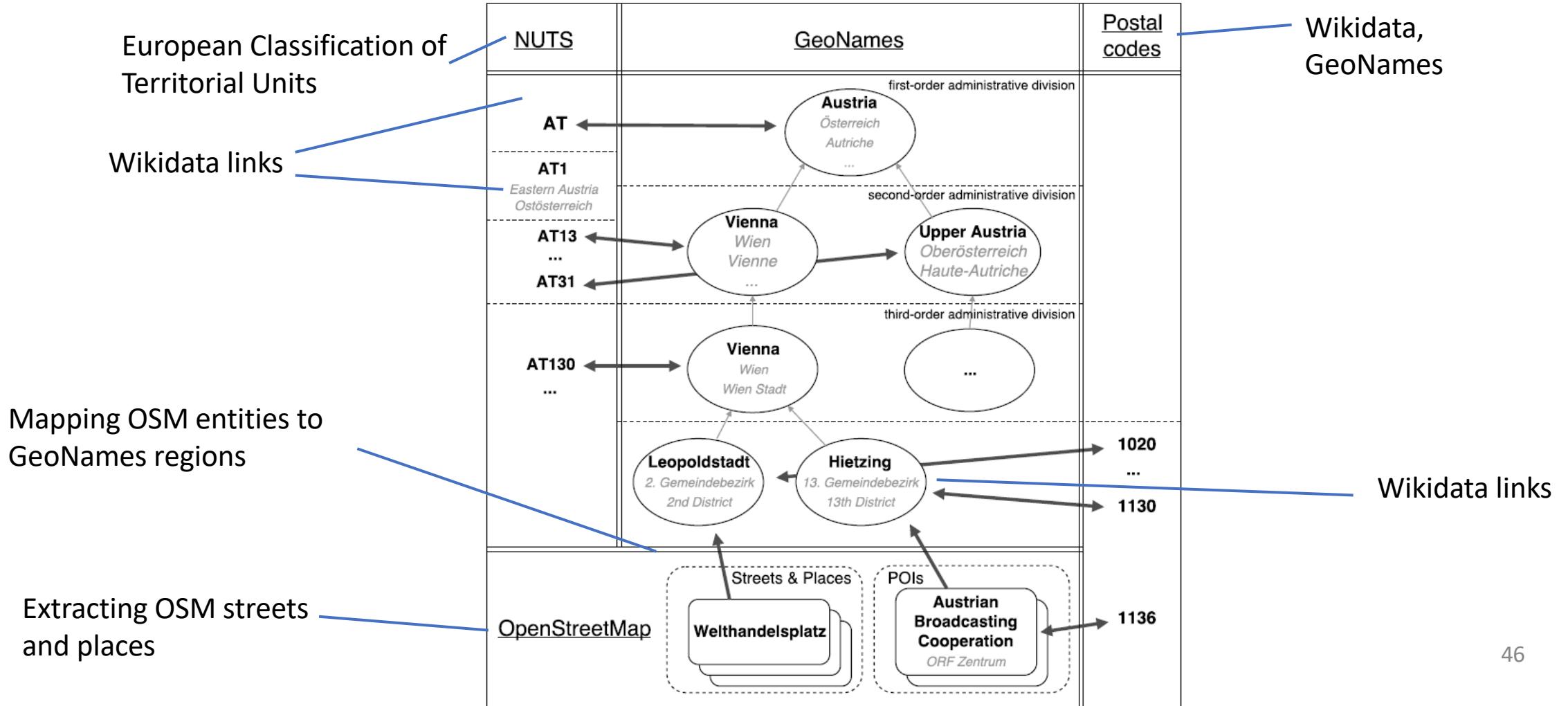
Geospatial Linked Data:





From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*





From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

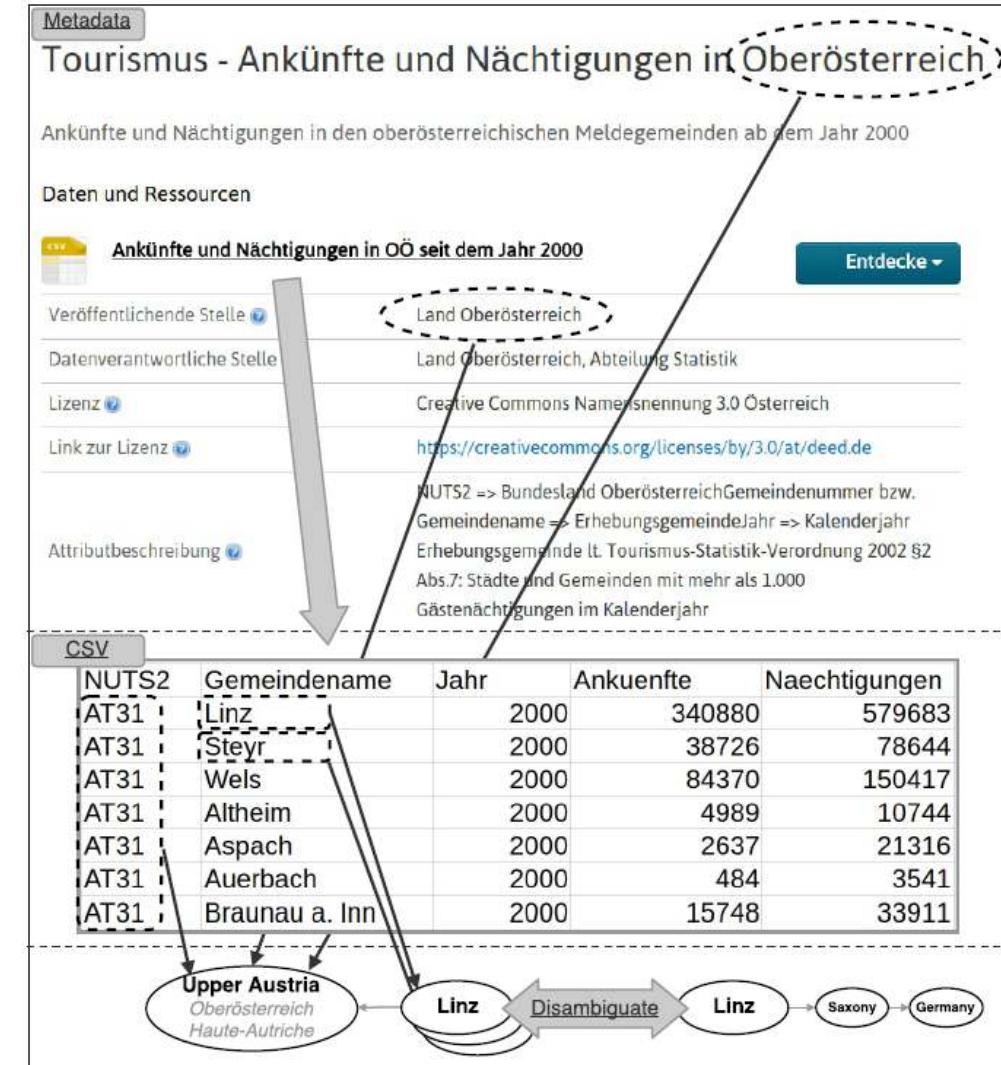
Dataset Labelling

Metadata descriptions

- Geo-entities in titles, descriptions, organizations
- „Origin“ country of the dataset (from portal)
- Temporal tagging

CSV cell value disambiguation

- Row context:
 - Filter candidates by potential parents (if available)
- Column context:
 - Least common ancestor of the spatial entities



From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

How well does it work? Indexed Datasets

<u>portal</u>	<u>datasets</u>	<u>CSVs</u>	<u>indexed</u>
<i>total</i>			15728
govdata.de	19464	10006	5646
data.gv.at	20799	18283	2791
offenedaten.de	28372	4961	2530
datos.gob.es	17132	8809	1275
data.gov.ie	6215	1194	884
data.overheid.nl	12283	1603	828
data.gov.uk	44513	7814	594
data.gov.gr	6648	414	496
data.gov.sk	1402	877	384
www.data.gouv.fr	28401	6038	258
opingogn.is	54	49	41

From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

Search Interface:

<https://data.wu.ac.at/odgraphsearch/>

Faceted query interface:

- Timespan
- Time pattern
- Geo-entities
- Full-text queries
- SPARQL endpoint

Back end:

- **MongoDB** for efficient key look-ups
- **ElasticSearch** for indexing and full-text queries
- **Virtuoso** as a triple store

The screenshot shows a search interface with a "Temporal filters" section. It includes a slider for selecting a timespan, with markers at "1/2010" and "1/2020". There are also radio buttons for "Off", "Title & description", and "CSV columns". A "Filter pattern" button and an "Apply Filter" button are also visible.

The screenshot shows search results for the query "Linz". At the top, there is a search bar with "Linz" and a magnifying glass icon. Below it, a breadcrumb navigation path is shown: Republic of Austria > Oberösterreich > Linz Stadt > Linz. There are two tabs: "Spatial entity" (which is selected) and "Full-text results".

Hotspot - Standorte - Hotspot Standorte							Stadt Linz	
POI's (Points of Interest) für Hotspot (freies, kostenloses WiFi) in der Stadt Linz. Die Koordinaten sind im im EPSG-Codes WGS84 verfügbar.							http://data.gv.at	
Nummer	Latitude	Longitude	Name	Kurztext	Start im Jahr	Ende im Jahr	Stadt	Postleitzahl
4007	48,304793	14,299414	Hotspot Linz-Rotes...	Hier ist nur einer v...	2013	0	Linz	4020

Finanzgebarung der Gemeinden in Oberösterreich - OÖ. Gemeinde-Finanzgebarung 2015					Land Oberösterreich		
Finanzdaten der 444 oberösterreichischen Gemeinden					http://data.gv.at		
Jahr	NUTS2	Gemeindenummer	Gemeindename	Ordentliche Einnahme...	Ordentliche Ausgaben	Außero. Einnahm...	
2015	AT31	40101	Linz	628704196,3	718773006,9	131859	

From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

Lessons learned

- Geospatial and Temporal scope is the most useful search feature for Open Data
- Respective Hierarchical Knowledge Graphs can be built from existing Linked Data Sources
- Our algorithms annotate CSV tables **and** their metadata descriptions

→KGs improve search (with some extra work)

- *Plus: RDF and SPARQL allow us to build application-specific KGs as “principled” aggregation(s) of Linked Data’ (cf. Def. slide 29)*



First Look
Journal of Web Semantics

Enabling Spatio-Temporal Search in Open Data

JWS: Information Retrieval

23 Pages • Posted: 20 Dec 2018 • First Look: Accepted

Sebastian Neumaier

Vienna University of Economics and Business; Vienna University of Technology

Axel Polleres

Vienna University of Economics and Business; Complexity Science Hub Vienna; Stanford University

Abstract

Intuitively, most datasets found on governmental Open Data portals are organized by spatio-temporal criteria, that is, single datasets provide data for a certain region, valid for a certain time period. Likewise, for many use cases (such as, for instance, data journalism and fact checking) a pre-dominant need is to scope down the relevant datasets to a particular period or region. Rich spatio-temporal annotations are therefore a crucial need to enable semantic search for (and across) Open Data portals along those dimensions, yet - to the best of our knowledge - no working solution exists. To this end, we (i) present a scalable approach to construct a spatio-temporal knowledge graph that hierarchically structures geographical as well as temporal entities, (ii) annotate a large corpus of tabular datasets from open data portals with entities from this knowledge graph, and (iii) enable structured, spatio-temporal search and querying over Open Data catalogs, both via a search interface as well as via a SPARQL endpoint, available at data.wu.ac.at/odgraphsearch/.

Keywords: open data, spatio-temporal labelling, spatio-temporal knowledge graph

Suggested Citation:

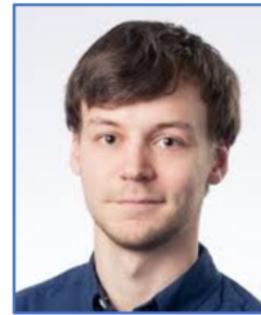


From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

How can we also use numeric values?

NUTS2	LAU2_NAME	YEAR	SEX	P_TOTAL
AT31	Linz	2013	1	98157
AT31	Steyr	2013	1	18763
AT31	Wels	2013	1	29730
...

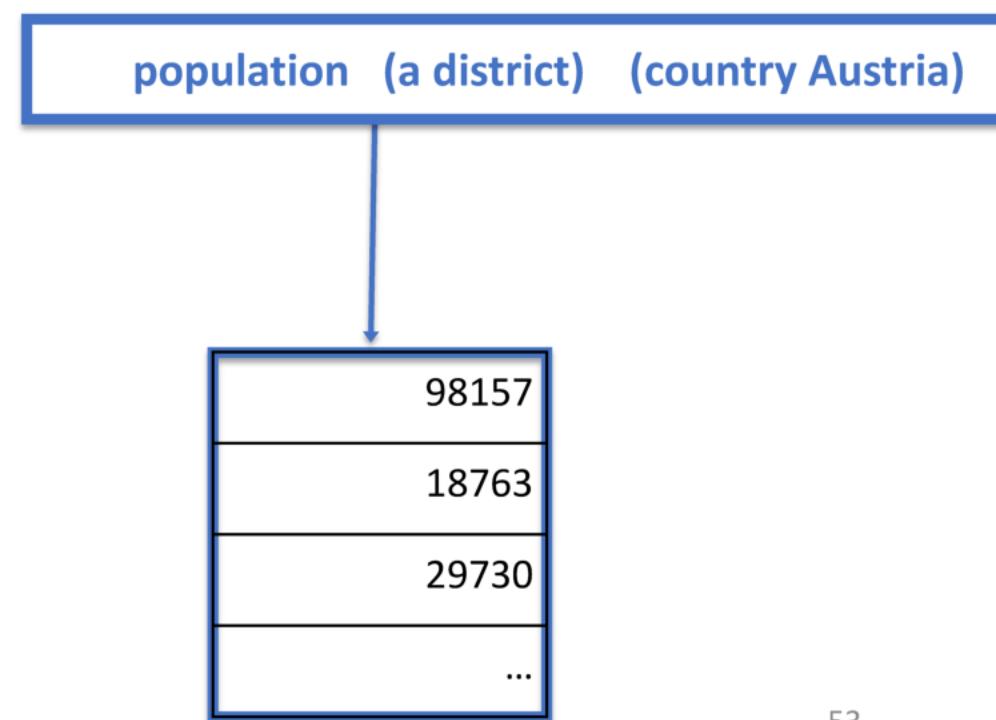


From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

How can we also use numeric values?

- Identifying the most likely semantic label for a bag of numerical values
- Deliberately ignore surroundings
- Map to labels from a KG





From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

Numeric Values in Open KGs?



What's in DBpedia?

- Cities
 - **Population**
 - **Area**
 - **Country**
 - **Location (Coordinates)**
 - Economic indicators
 - ...
- Organisations:
 - **Revenues**
 - Board members
 - ...
- Persons (e.g. celebrities, sports)
 - Name
 - Profession
 - **Height**
- Landmarks (e.g. famous buildings)
 - **Country**
 - **Location**
 - **Height**
- Events
 - **Dates**
 - **Location**

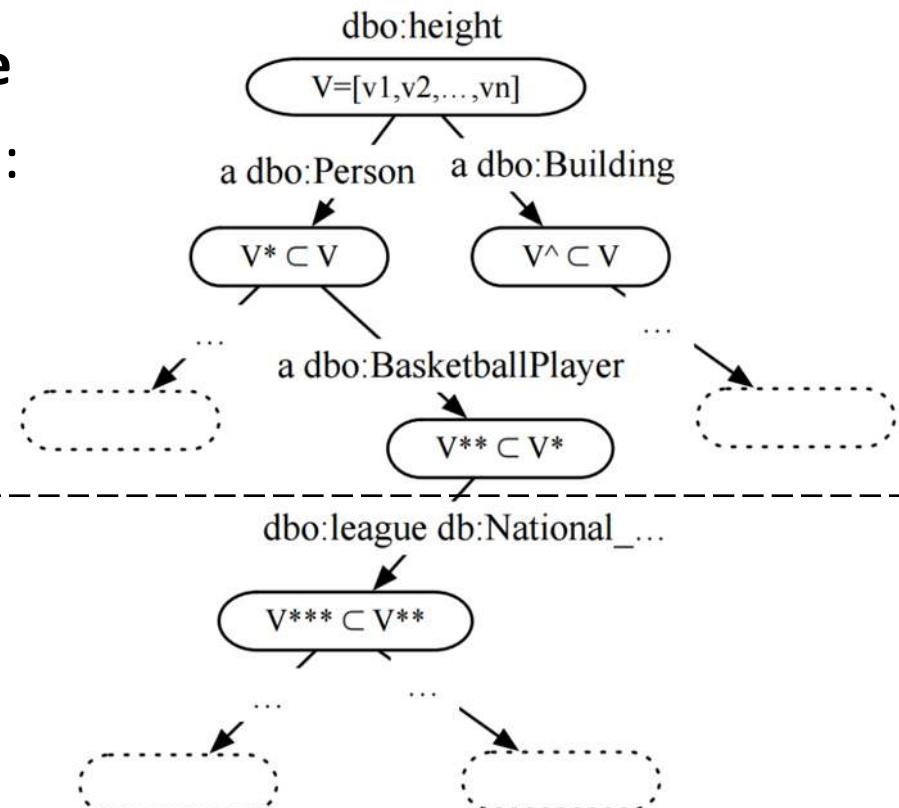


From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*

Background Knowledge Graph

- properties with **numerical range**
- Hierarchical clustering approach:
- Two hierarchical layers:
 - **Type** hierarchy
(using OWL classes)
 - **Property-object** hierarchy
(shared property-object pairs)

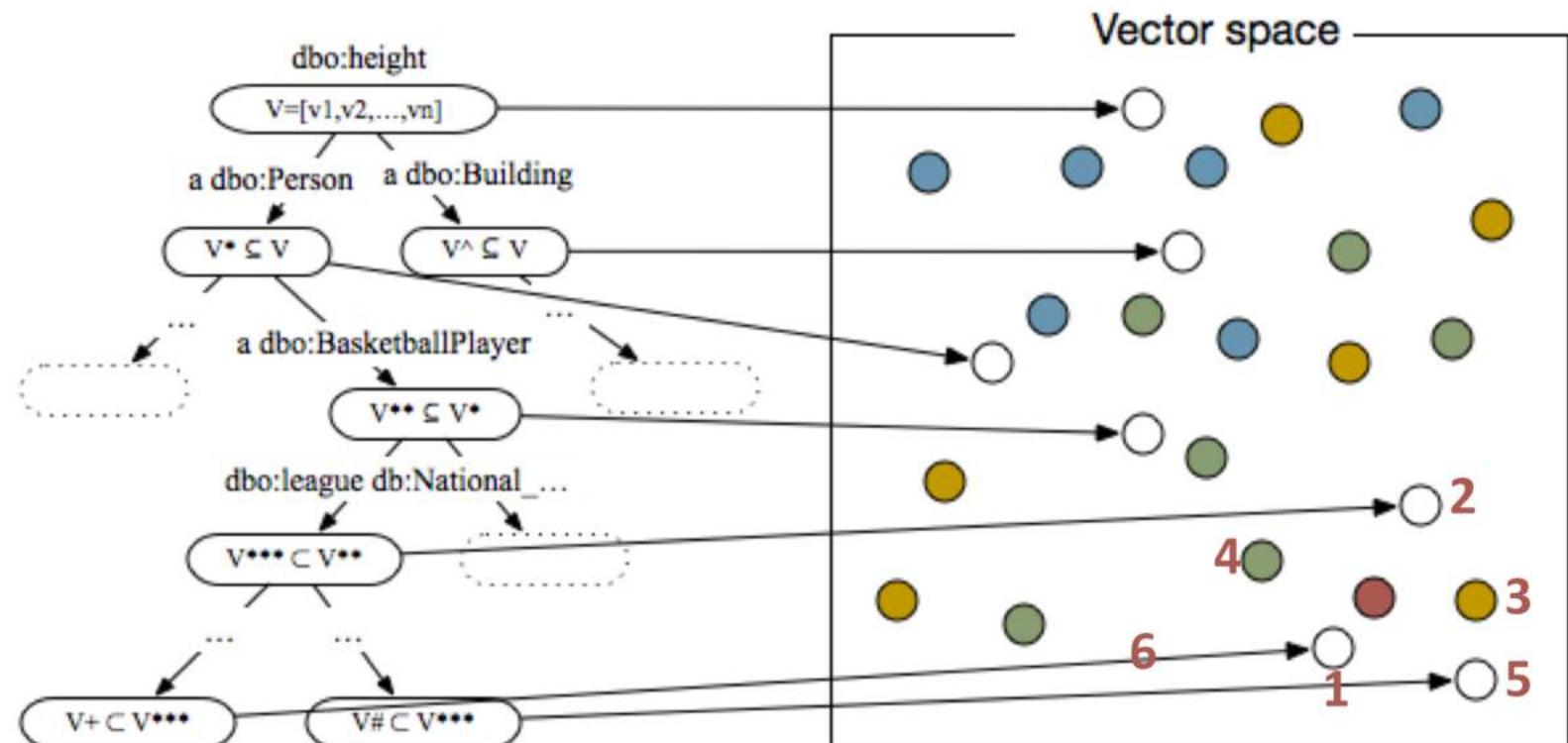




From Knowledge Graphs (back) to Linked Open Data:

- Open Knowledge Graphs can be used to **link Open Data!**

Label based on Nearest Neighbors





From Knowledge Graphs (back) to Linked Open Data:

- Open Knowledge Graphs can be used to **link Open Data!**
- Open KGs don't/only partially cover Knowledge in Open Data!

Example Labelling:

NUTS1	NUTS2	NUTS3	DISTRICT_CODE	T	WV	WK	BZ	SPR	WBER	ABG.	UNG.	OEVP	SPOE	FPOE	GRUE	BZOE	NEOS
AT1	AT13	AT130		1	9	0	0	0	1163061	503284	9386	81974	136391	89963	103249	1516	44891
AT1	AT13	AT130		2	9	1	0	0	111279	52674	774	9344	12395	6482	14154	114	5412
AT1	AT13	AT130		2	9	2	0	0	98379	51785	646	10324	10236	4700	15398	124	6569
AT1	AT13	AT130		2	9	3	0	0	110527	45483	810	5317	13304	7816	10944	115	3613
AT1	AT13	AT130		2	9	4	0	0	229521	84387	1953	10097	27922	21091	11631	256	5299
AT1	AT13	AT130		2	9	5	0	0	212262	97755	1806	18703	25314	16613	19333	324	9175
AT1	AT13	AT130		2	9	6	0	0	175288	82790	1321	17560	19059	11765	18996	242	8389
AT1	AT13	AT130		2	9	7	0	0	225805	88410	2076	10629	28161	21496	12793	341	6434
AT1	AT13	AT130	90301	3	9	1	3	0	57528	27320	412	4938	6586	3567	6969	68	2789
AT1	AT13	AT130	90401	3	9	1	4	0	21000	11027	138	2401	2253	1068	3082	26	1277
AT1	AT13	AT130	90501	3	9	1	5	0	32751	14327	224	2005	3556	1847	4103	20	1346



From Knowledge Graphs (back) to Linked Open Data:

- *Open Knowledge Graphs can be used to link Open Data!*
- *Open KGs don't/only partially cover Knowledge in Open Data!*

Lessons learned:

- We can assign fine-grained semantic labels
 - **if there is enough evidence in Background Knowledge Graph**
- *However:* Missing domain knowledge for labelling OD

Future work:

- Complementary to existing approaches (column header labeling, entity linking and relation extraction)
- Combined approaches may improve results

International Semantic Web conference 2016:

Multi-level semantic labelling of numerical values

Sebastian Neumaier¹, Jürgen Umbrich¹, Josiane Xavier Parreira², and Axel Polleres¹

¹ Vienna University of Economics and Business, Vienna, Austria

² Siemens AG Österreich, Vienna, Austria

Abstract. With the success of Open Data a huge amount of tabular data sources became available that could potentially be mapped and linked into the Web of (Linked) Data. Most existing approaches to “semantically label” such tabular

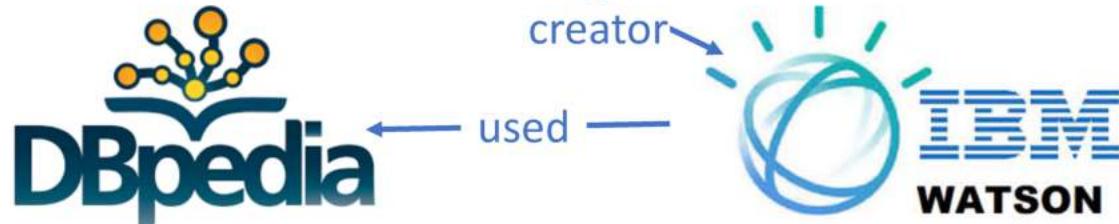
The journey doesn't end here...

Open Knowledge Graphs can be used for many other AI applications!

E.g. Open KG Question Answering



creator



Cornell University
arXiv.org > cs > arXiv:1908.06917
Computer Science > Computation and Language
Message Passing for Complex Question Answering over Knowledge Graphs
(Submitted on 19 Aug 2019)
Svitlana Vakulenko, Javier David Fernandez Garcia, Axel Polleres, Maarten de Rijke, Michael Cochez

Pre-print (accepted at CIKM2019)

This image shows a screenshot of an arXiv pre-print page. The title of the paper is "Message Passing for Complex Question Answering over Knowledge Graphs". It is submitted to the Computer Science > Computation and Language category on arXiv.org. The authors listed are Svitlana Vakulenko, Javier David Fernandez Garcia, Axel Polleres, Maarten de Rijke, and Michael Cochez. The submission date is 19 August 2019. The page also includes the Cornell University logo and navigation links like "Search..." and "Help | Ad". A large diagonal banner across the top right of the page reads "Pre-print (accepted at CIKM2019)".

Combining Graph Algorithms and ML/Neural Networks



The journey doesn't end here...

Open Knowledge Graphs can be used for many other AI applications!

E.g. Open KG Question Answering

More examples of active research topics:

- *Using ML/NN for KG population/extraction*
- *Graph embeddings for automated KG enrichment/repair*

Pre-print (accepted at CIKM2019)

Search... Help | Ad

Cornell University arXiv.org > cs > arXiv:1908.06917 Computer Science > Computation and Language Message Passing for Complex Question Answering over Knowledge Graphs (Submitted on 19 Aug 2019) Svitlana Vakulenko, Javier David Fernandez Garcia, Axel Polleres, Maarten de Rijke, Michael Cochez

We propose a novel approach for complex KGQA that uses unsupervised message passing, which propagates confidence scores obtained by parsing an input question and matching terms in the knowledge graph to a set of possible answers. First, we identify entity, relationship, and class names mentioned in a natural language question, and map these to their counterparts in the graph. Then, the confidence scores of these mappings propagate through the graph structure to locate the answer entities. Finally, these are aggregated depending on the identified question type. This approach can be efficiently implemented as a series of sparse matrix multiplications. We show that the proposed approach outperforms the state-of-the-art on the LC-QuAD dataset and achieves state-of-the-art performance on the CoQA dataset. Our error analysis reveals that the proposed approach produces a correct answer ranking. Our error analysis depends only on the quality of the question interpretation. Finally, we provide a comprehensive evaluation of the proposed approach on the CoQA dataset and a detailed analysis of the question answering components in more detail.



Combining Graph Algorithms and ML/Neural Networks

Last but not least, some bad news...

Bad news: Decentralized Open KGs are hard to maintain and serve

- Hosting federated SPARQL endpoints is (too) expensive
- Many Linked Data datasets are not maintained sustainably.
- Linked Datasets difficult to crawl and process automatically

Actually: Good news!

Still more research needed and actively being conducted!

Examples:

- *Federated SPARQL querying*
- *Client-Server Load-Balancing (LDF, SAGE, etc.)*
- *Efficient/lightweight triple stores (e.g. HDT)*
- *Graph Databases in the Cloud*
- *Automated Linked Data Quality checking*
- *(BTW, real applications often also need access-control, policies)*

A More Decentralized Vision for Linked Data
Axel Polleres^{1,2}, Maulik R. Kamdar¹, Javier D. Fernández², Tania Tudorache¹, and Mark A. Musen¹
¹ Stanford University, CA, USA
² Vienna Univ. of Economics & Business / Complexity Science Hub Vienna, Austria

Abstract. We claim that ten years into Linked Data there are still many unsolved challenges towards arriving at a truly machine-readable *and* decentralized Web of data. With a focus on the biomedical domain—currently, one of the most promising “adopters” of Linked Data, we highlight and exemplify key technical and non-technical challenges to the success of Linked Data, and we outline potential solution strategies.

DESEMWEB 2018

Time for a postcard...



Time for a postcard...

RDF, Ontologies, Linked Data are a solid basis for Knowledge graphs.

Focus has shifted from schema-centric to data-centric.

Focus has shifted from knowledge engineering to knowledge extraction.

For companies thinking about investing in KGs it makes sense to invest in open, standard technologies for KGs.

Hybrid Reasoning needed (rules + heuristics + inductive/statistical inference (ML)

It is not enough to rely on centralized/closed KGs!

Looking forward to the onward journey!
lot of research left to be done in this space (decentralization, contextualization, federation)!

Take-home Messages



Confirmed by yesterday's keynote:

+ Future of AI Research for Manufacturing

- **AI provides just sets of methods – the human centric complex solutions should be built upon them – three key research areas for future AI research** (Catalyst & AAAI report):
 - **Integrated intelligence:** combining modular AI capabilities, functionalities and skills, ability to contextualize general capabilities to suit specific uses, creation of open shared knowledge repositories, models of human cognition
 - **Meaningful interaction:** productive collaboration in mixed teams of humans and machines, combining diverse communication modalities (verbal, visual, emotional) while respecting privacy, responsible and trustworthy behaviors
 - **Self-aware learning:** developing robust and trustworthy learning, learning from small amounts of data, incorporating prior knowledge into learning, and learning real-time behaviors for intentional sensing and acting.

Some references for the mentioned works (thanks to my co-authors!):

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