

Knowledge Representation and Reasoning

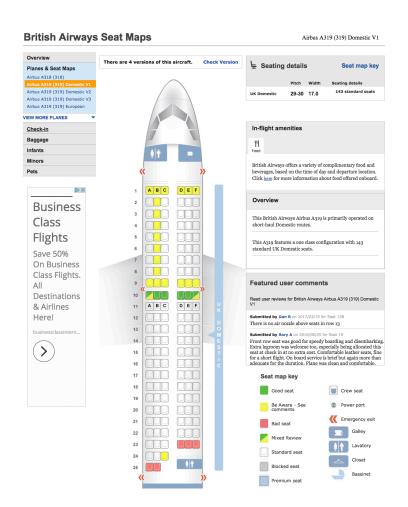
Hannes Straß

Based on slides of Bernardo Cuenca Grau, Ian Horrocks, and Przemysław Wałęga (University of Oxford)



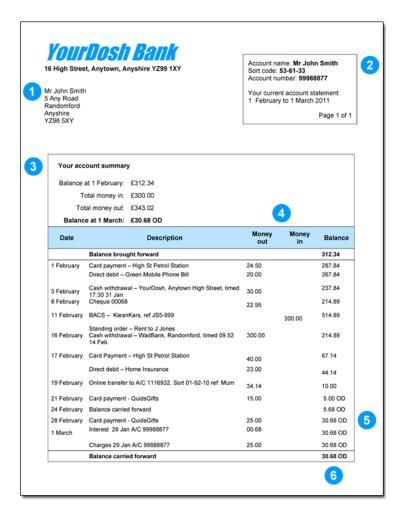


- Applications often need to represent knowledge
 - seats on an aeroplane





- Applications often need to represent knowledge
 - seats on an aeroplane
 - account transactions



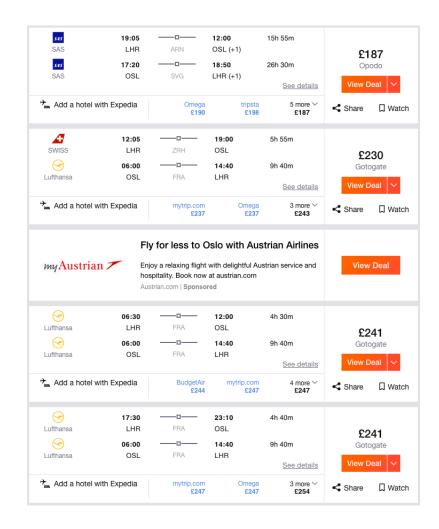


- Applications often need to represent knowledge
 - seats on an aeroplane
 - account transactions
 - tall buildings

Name \$	City ¢	State	Height +	Feet +	Floors +	Completed +
Federation Tower: East Tower	Moscow	Russia	373.7	1,226	95	2016
OKO: South Tower	Moscow	Russia	354.1	1,162	85	2015
Mercury City Tower	Moscow	Russia	338.8	1,112	75	2013
The Shard ^[1]	London	United Kingdom	309.7	1,017	87	2012
Eurasia ^[2]	Moscow	Russia	308.9	1,014	72	2014
CoC: Moscow Tower[3]	Moscow	Russia	301.6	990	76	2010
Skyland İstanbul 1 ^{[4][5]}	Istanbul	C Turkey	293.1	932	65	2017
Skyland İstanbul 2 ^{[4][5]}	Istanbul	C Turkey	293.1	932	65	2017
Metropol İstanbul ^[6]	Istanbul	C Turkey	280	919	66	2017
Emaar Square	Istanbul	C· Turkey	280	920	62	2018
Naberezhnaya Tower C	Moscow	Russia	268.4	881	59	2007
Triumph Palace ^[7]	Moscow	Russia	264.1	867	57	2005
Commerzbank Tower ^{[8][9]}	Frankfurt	Germany	258.7	848	56	1997
CoC: Saint Petersburg Tower ^[10]	Moscow	Russia	256.9	843	65	2010
Messeturm	Frankfurt	Germany	256.5	842	55	1990
Nurol Life ^{[11][12]}	Istanbul	c· Turkey	252	827	60	2017
Torre de Cristal	Madrid	Spain	249	817	45	2008
Torre Cepsa	Madrid	Spain	248.3	815	45	2008
Evolution Tower ^[13]	Moscow	Russia	245.9	807	53	2014
OKO: North Tower ^[14]	Moscow	Russia	245	804	49	2014
Federation: West Tower	Moscow	Russia	243.2	798	62	2007
Main building of Moscow State University	Moscow	Russia	240	787	36	1953
Imperia Tower	Moscow	Russia	238.7	783	60	2011
Palace of Culture and Science	Warsaw	Poland	237	777	43	1955
Torre PwC	Madrid	Spain	236	774	52	2008
1 Canada Square	London	United Kingdom	235	771	50	1991
Istanbul Sapphire ^{[15][16]}	Istanbul	C Turkey	234.9	770	54	2010
Tour First	Paris ^B	France	231	758	56	2011
Unicredit Tower	Milan	■ Italy	231	758	35	2011
Heron Tower	London	United Kingdom	230	755	46	2011

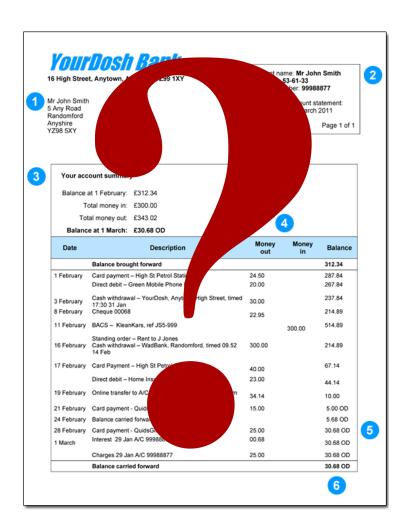


- Applications often need to represent knowledge
 - seats on an aeroplane
 - account transactions
 - tall buildings
- and to answer questions
 - seats available on flight?





- Applications often need to represent knowledge
 - seats on an aeroplane
 - account transactions
 - tall buildings
- and to answer questions
 - seats available on flight?
 - can afford to pay rent?

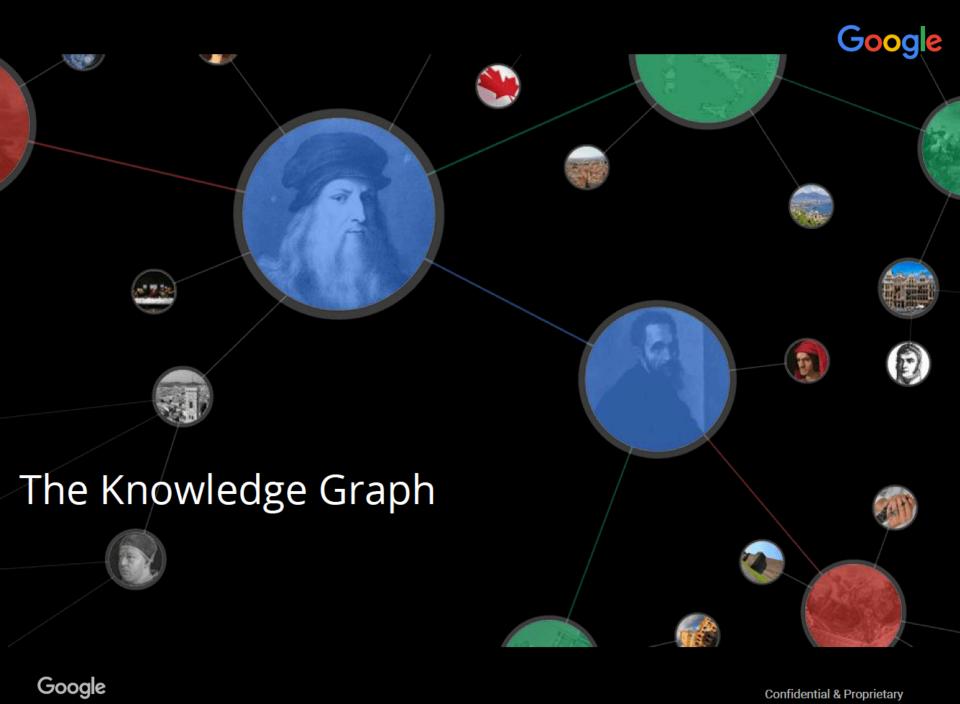




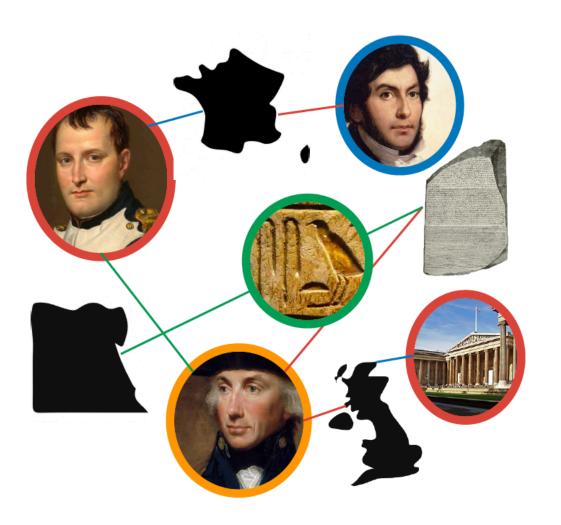
- Applications often need to represent knowledge
 - seats on an aeroplane
 - account transactions
 - tall buildings
- and to answer questions
 - seats available on flight?
 - can afford to pay rent?
 - tallest building in Europe?







The Knowledge Graph

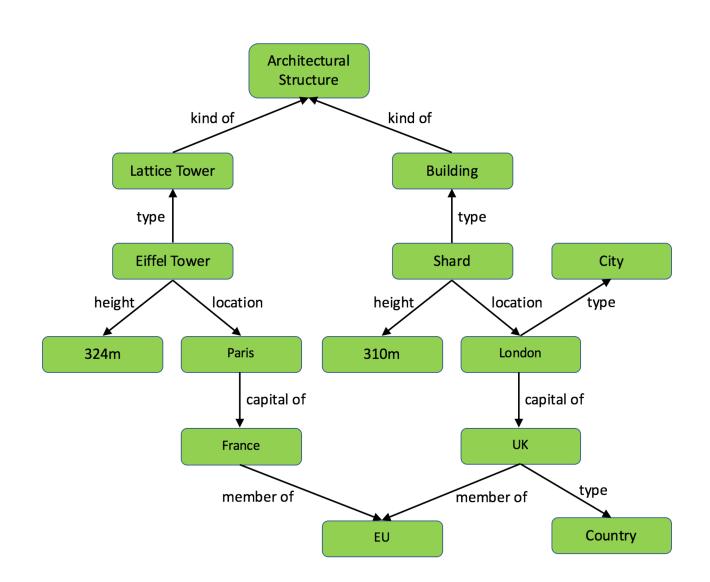


The Knowledge Graph is a comprehensive collection of real-world entities (people, places, things, and concepts) along with relationships and factual attributes that describe them.

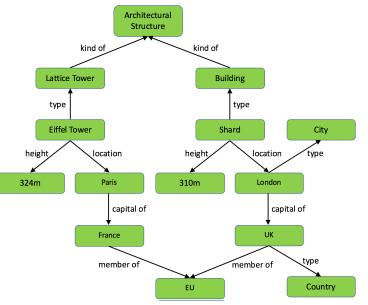


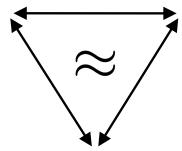












Architectural Structure					
name	location	height	kind		
Shard	London	310m	Building		
Eiffel Tower	Paris	324m	Lattice Tower		

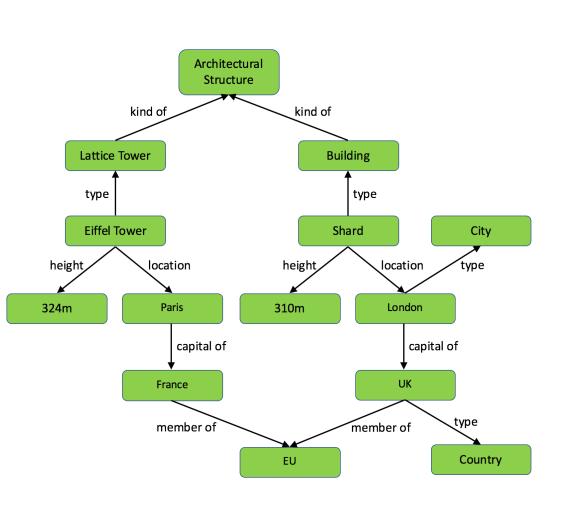
City			
name	capital of		
London	UK		
Paris	France		

member of				
country	organisation			
France	EU			
UK	EU			

Building(Shard)
City(London)
location(Shard,London)
height(Shard,310m)
capitalOf(London,UK)

.. 13

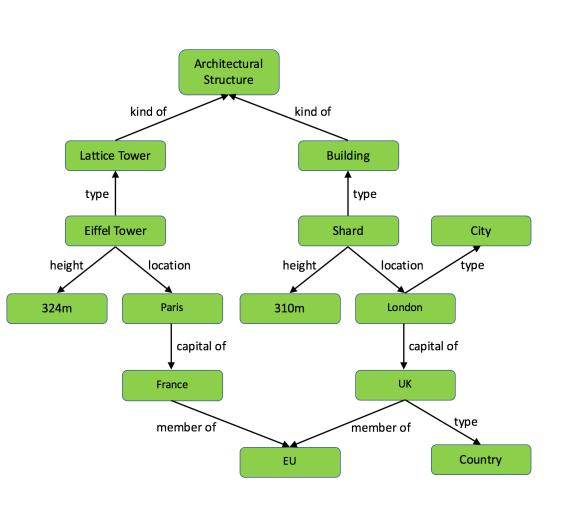




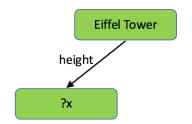
- Reasoning is the process of answering queries w.r.t. the represented knowledge
- What is the height of the Eiffel Tower?

SELECT ?x
WHERE { EiffelTower height ?x. }

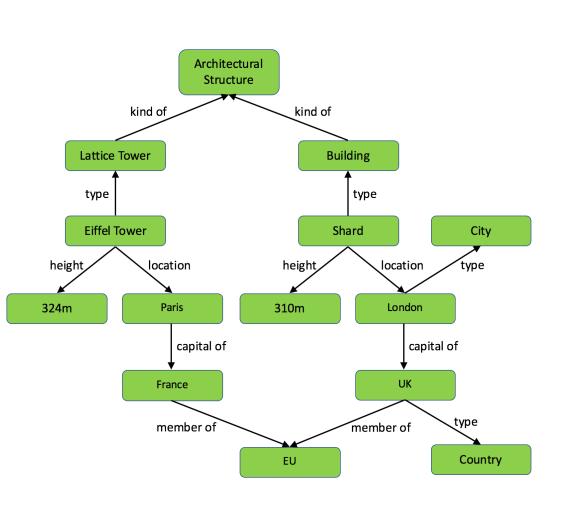




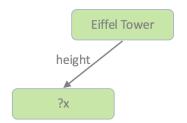
- Reasoning is the process of answering queries w.r.t. the represented knowledge
- What is the height of the Eiffel Tower?



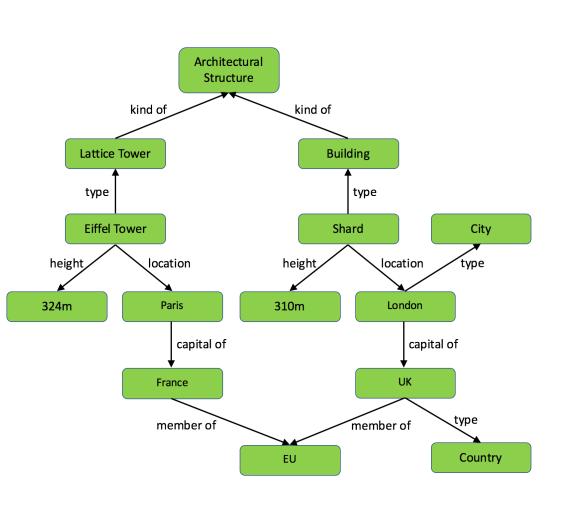




- Reasoning is the process of answering queries w.r.t. the represented knowledge
- What is the height of the Eiffel Tower?







- Reasoning is the process of answering queries w.r.t. the represented knowledge
- What is the height of the Eiffel Tower?
- 324m





how tall is the eiffel tower

Q

All

Images

Maps

Shopping

News

More

Settings

Tools

About 41,100,000 results (0.68 seconds)

Eiffel Tower / Height

300 m, 324 m to tip



People also search for



Burj Khalifa 828 m



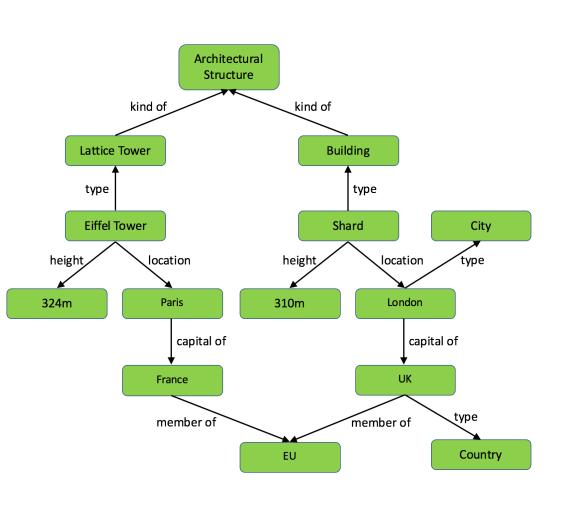
Statue of Liberty 93 m



Leaning Tower of Pisa 58 m



Problem: Vague Semantics

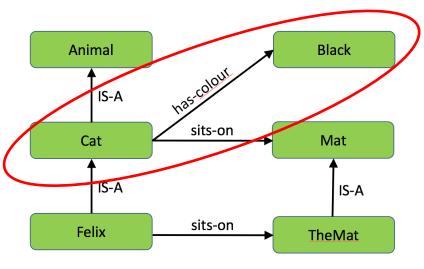


- What is the difference between a Building, a Lattice Tower and an Architectural Structure?
- Is the Eiffel Tower a Building; is it an Architectural Structure?
- Special meaning of, e.g., type and kind of edges?



Semantics and Reasoning

Semantics: the study of meaning



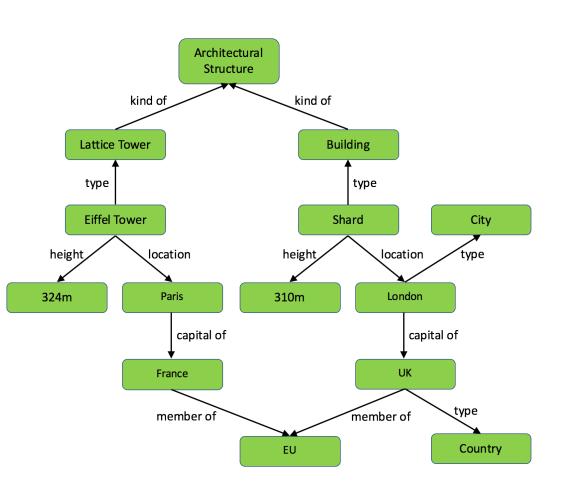


[Quillian, 1967]

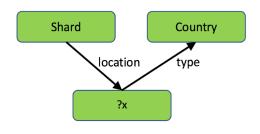
• (Precise) semantics needed in order to define what (correct) query answers should be



Problem: Implicit Knowledge

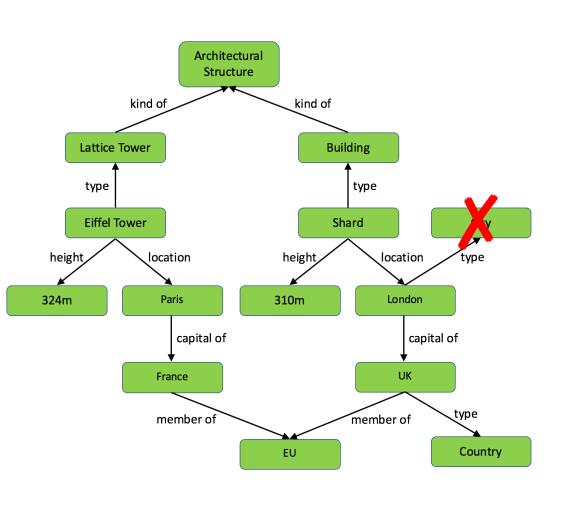


 What country is the Shard located in?



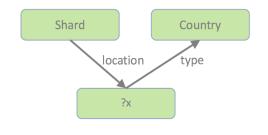


Problem: Implicit Knowledge



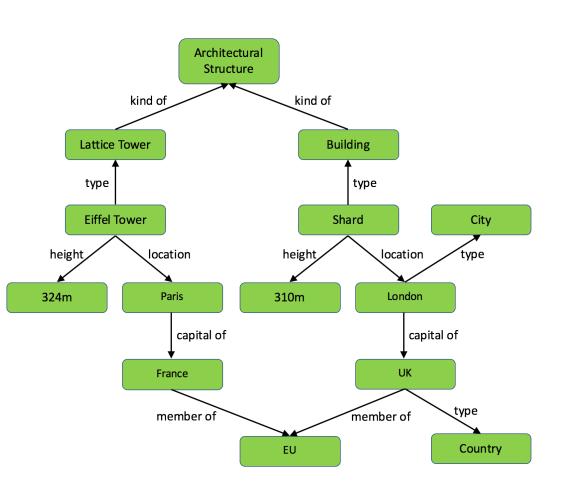
 What country is the Shard located in?

• ???

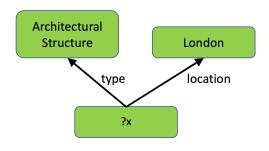




Problem: Implicit Knowledge

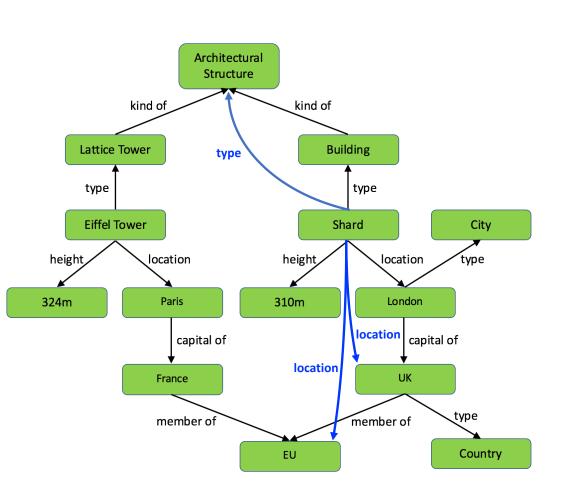


 What Architectural Structures are located in London?



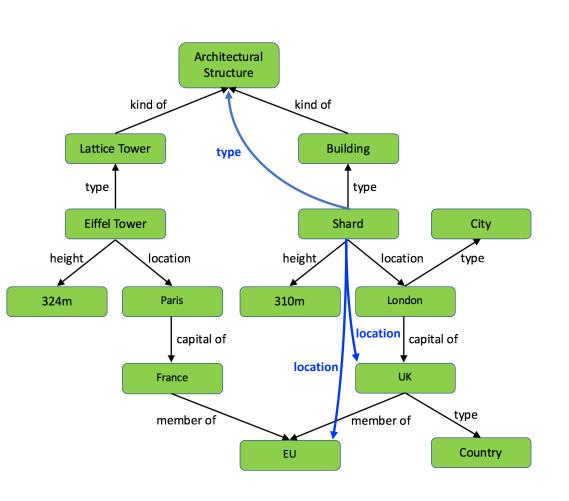


Add "Relevant" Edges?





Problem: Creation & Maintenance



- Every Building located in London is also located in
 - UK
 - EU
 - England
 - Northern Hemisphere
 - ...
- Need to add a very large number of edges



Extended representation

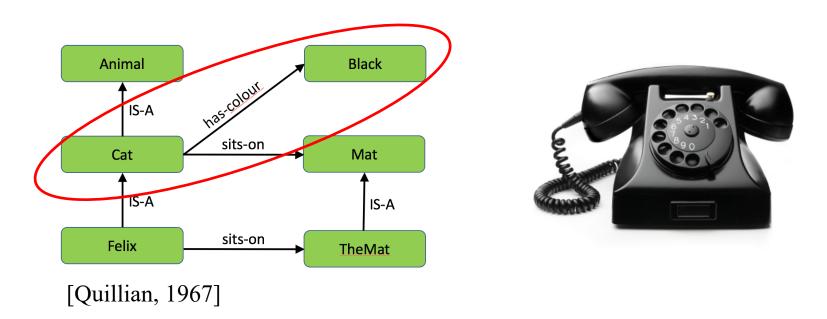
- A city that is the capital of a country is a (geographical) part of that country[†]
- A thing that is located in a city that is a (geographical) part of a country is also located in that country
- A thing that is located in a country that is a member of a supranational union is also located in that supranational union

[†] Part-whole relationships are complicated! They are the subject of a whole field of study in logic and philosophy: mereology



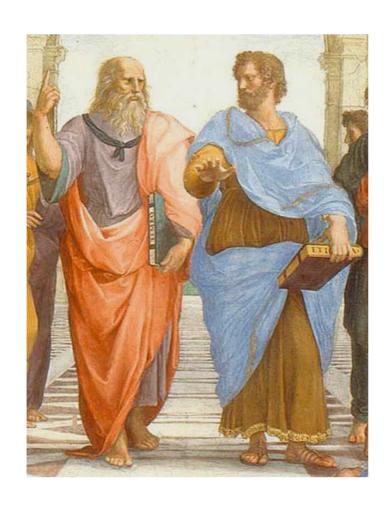
Semantics and Reasoning

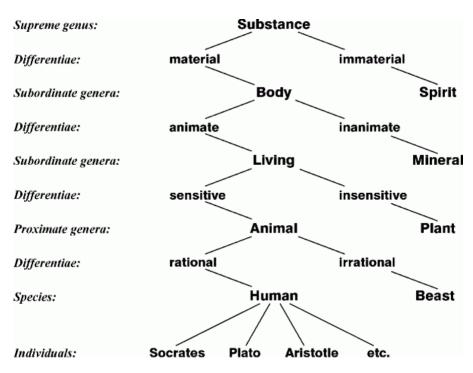
Semantics: the study of meaning



 (Precise) semantics needed in order to define what (correct) query answers should be







"All men are mortal, all Greeks are men, therefore all Greeks are mortal" (syllogism)



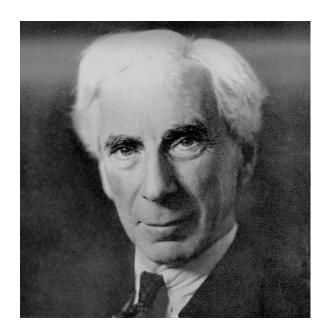
- Modern KR languages often based on logic
- Typically (subsets of) First Order Predicate Calculus



Gottlob Frege

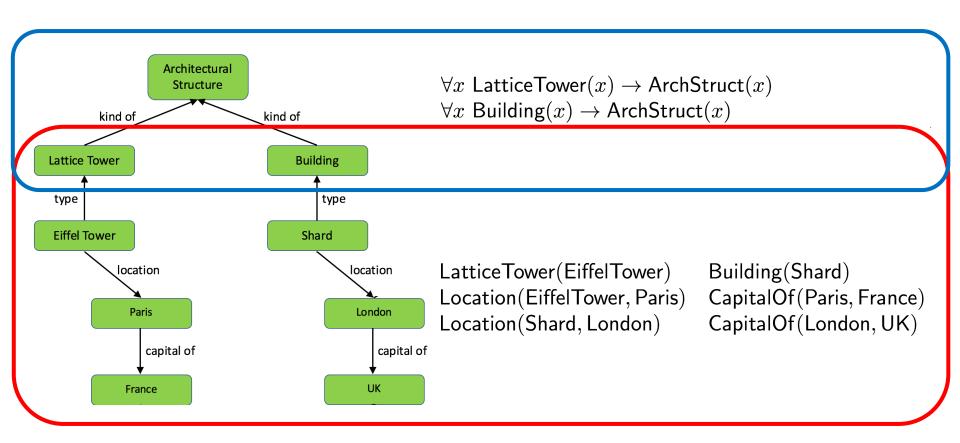


Charles Sanders Peirce



Bertrand Russell

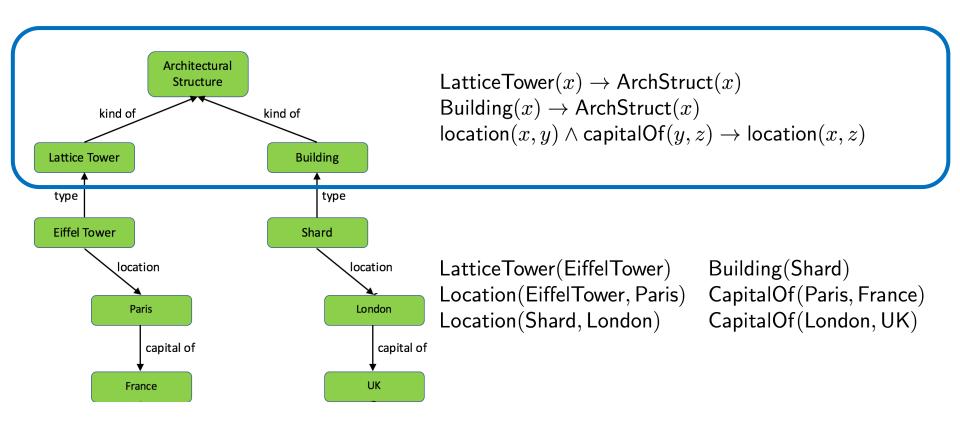




facts / data / (RDF) graph

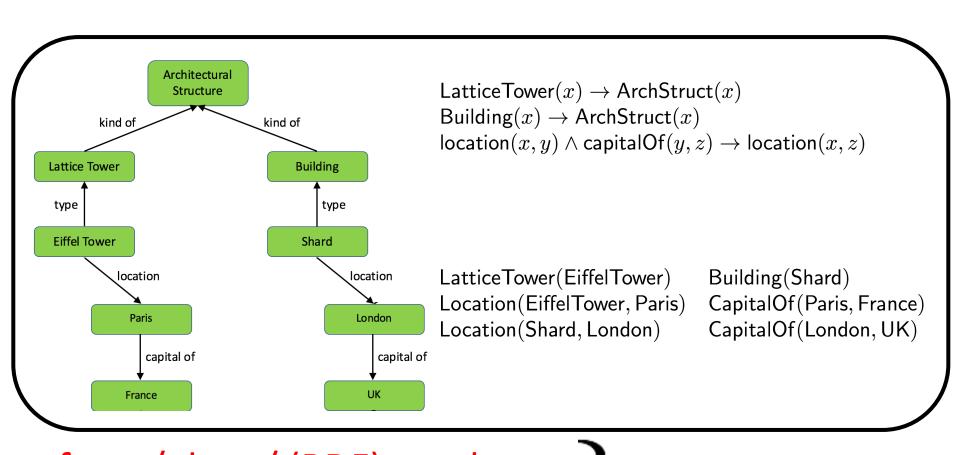
ontology / conceptual schema





facts / data / (RDF) graph
ontology / conceptual schema

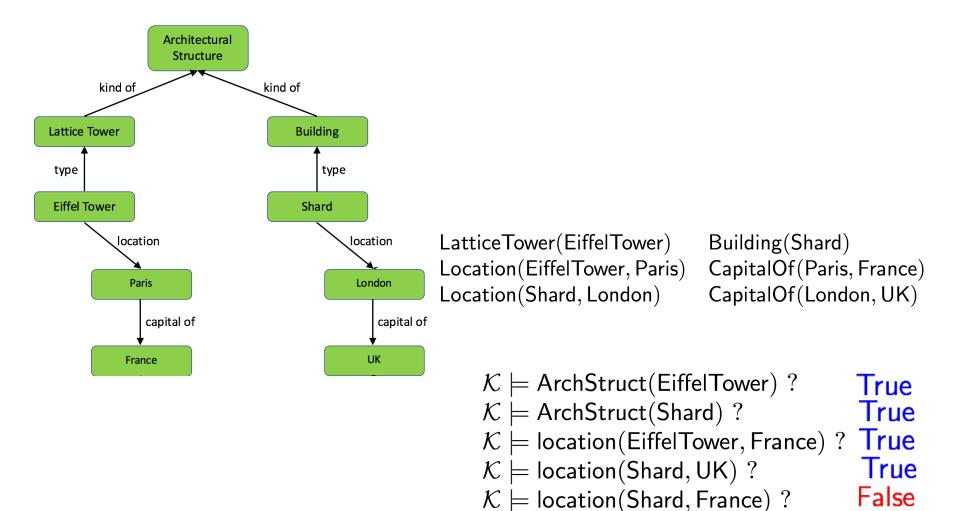




facts / data / (RDF) graph
ontology / conceptual schema

knowledge base







Reasoning Algorithms

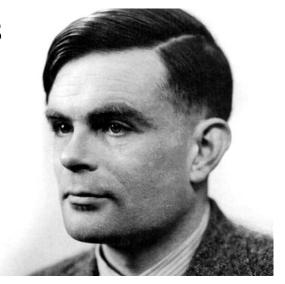
- Devise algorithms that compute query answers
- E.g., using natural deduction rules:

 Can check/prove algorithms are sound and complete w.r.t. semantics



Problem Solved?

- Turing showed that some problems cannot be completely solved using standard computational model
 - halting problem
 - FOL entailment problem



 Even if decidable, reasoning might be of inherently high complexity and so take an infeasibly long time



So what to do?

"Scruffy" approach:

- Ad-hoc representation
- Efficient but (at least) incomplete algorithms
- ✓ Can use arbitrarily powerful representation
- √ Favourable scalability properties
- Incomplete answers
 - X Degree of incompleteness unknown
 - X Incompleteness can easily become unsoundness

Computational Logic : Group

So what to do?

"Neat" approach:

- Study KR languages to find appropriate balance of expressive power and computability
- Design algorithms that work well in typical cases
- Develop highly optimised implementations
- ✓ Precisely defined semantics
- √ Formal properties well understood
- ✓ Sound and complete reasoning
- Limited representation power
- Optimisations may not offer robust scalability



Description Logics

- Family of logic-based KR languages
- Most are decidable subsets of FO logic
- Provide a range of different constructors
 - Booleans (and, or, not)
 - Restricted forms of quantification (exists, forall)
 - Counting (atmost, atleast)
 - ...
- Decidability/complexity and (efficient) algorithms known for many combinations of constructors
- Highly optimised implementations for various "sweet spot" languages



Description Logic



ABox consistency

Complexity of reasoning in Description Logics

Note: the information here is (always) incomplete and updated often

Base description logic: Attributive \mathcal{L} anguage with \mathcal{C} omplements

 $\mathcal{ALC} ::= \bot \mid A \mid \neg C \mid C \land D \mid C \lor D \mid \exists R.C \mid \forall R.C$



Concept constructors:	Role constructors:	trans reg
■ \mathcal{F} - functionality 2 : (≤1 R) ■ \mathcal{N} - (unqualified) number restrictions: (≥ n R), (≤ n R) ■ Q - qualified number restrictions: (≥ n R . C), (≤ n R . C) ■ Q - nominals: { a } or { a_1 ,, a_n } ("one-of" constructor) ■ Q - least fixpoint operator: Q - Q - Q - Q - Q - Q - Q - Q -	✓ I – role inverses: $R^ \bigcirc$ \cap – role intersection 3 : $R \cap S$ \bigcirc \cup – role union: $R \cup S$ \bigcirc \neg – role complement: full \bigcirc \bigcirc – role chain (composition): $R \cap S$	
R⊆S - role-value-maps $ f = g - agreement of functional role chains ("same-as")$	 * - reflexive-transitive closure⁴: R* id - concept identity: id(C) Forbid : complex roles⁵ in number restrictions⁶ 	
TBox is internalized in extensions of ALCIO, see [76, Lemma 4.12], [54, p.3] • Empty TBox • Acyclic TBox ($A \equiv C$, A is a concept name; no cycles) • General TBox ($C \subseteq D$ for arbitrary concepts C and D)	Role axioms (RBox): $ S = \text{Role transitivity: Trans}(R) $ $ S = \text{Role hierarchy: } R \subseteq S $ $ S = \text{Complex role inclusions: } RoS \subseteq R, RoS \subseteq S $ $ S = \text{some additional features} $	OWL-Lite OWL-DL OWL 1.1
You have selected the Description Lo	gic: SHOIN	

Complexity of reasoning problems ⁷			
Reasoning problem	easoning problem Complexity ⁸ Comments and references		
		 <u>Hardness</u> of even ALCFIO is proved in [76, Corollary 4.13]. In that paper, the result is formulated for ALCQIO, but only number restrictions of the form (≤1R) are used in the proof. 	
		 A different proof of the NExpTime-hardness for ALCFIO is given in [54] (even with 1 nominal, and ro inverses not used in number restrictions). Upper bound for SHOIQ is proved in [77, Corollary 6.31] with numbers coded in unary (for binary 	
Concept satisfiability	NExpTime-complete		

NExpTime-complete By reduction to concept satisfiability problem in presence of nominals shown in [69, Theorem 3.7].

transitive subroles) are allowed; otherwise we gain undecidability even in 3727 ; see [46].
• Remark: recently [47] it was observed that, in many cases, one can use transitive roles in number
restrictions - and still have a decidable logic! So the above notion of a simple role could be substantially
extended.

• Important: in number restrictions, only simple roles (i.e. which are neither transitive nor have a



Description Logic



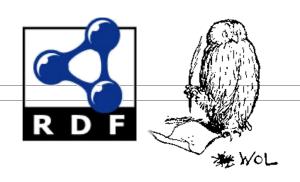
List of reasoners

Instution	Download	Publication
VIStology, Inc.	Download	Core publication
University of Ferrara	Download	Core publication
Technische Universität Dresden	Download	Core publication
The University of Manchester	Download	Core publication
Vienna University of Technology	Download	Core publication
University of Malaga	Download	Core publication
Not given	Download	Core publication
	VIStology, Inc. University of Ferrara Technische Universität Dresden The University of Manchester Vienna University of Technology University of Malaga	VIStology, Inc. Download University of Ferrara Download Technische Universität Dresden The University of Manchester Vienna University of Technology University of Malaga Download

DistEL	Wright State University	Download	Core publication
DRAOn	University of Paris 8, IUT of Montreuil	Download	Core publication
DReW	Vienna University of Technology	Download	Core publication
ELepHant	Not given	Download	Core publication
ELK	University of Ulm, Germany	Download	Core publication
ELOG	Not given	Download	Core publication
FaCT++	The University of Manchester	Download	Core publication
fuzzyDL	ISTI – CNR	Download	Core publication

HermiT	University of Oxford	Download	Core publication
jcel	Technische Universität Dresden	Download	Core publication
JFact	The University of Manchester	Download	Core publication
Konclude	University of Ulm, derivo GmbH	Download	Core publication
LiFR	Centre for Research and Technology Hellas (CERTH)	Download	Core publication
Mastro	Sapienza University of Rome	Download	Core publication
MORe	University of Oxford	Download	Core publication
ontop	Free University of Bozen-Bolzano	Download	Core publication





Standardised KR language

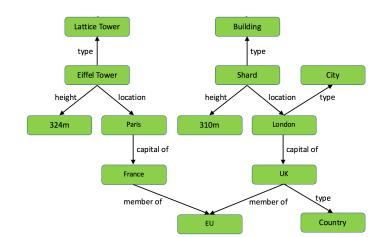
- RDF provides a graphical data model
- OWL provides a DL-based ontology language

 $\mathsf{LatticeTower}(x) \to \mathsf{ArchStruct}(x)$

OWL ontology: Building $(x) \rightarrow ArchStruct(x)$

 $\mathsf{location}(x,y) \land \mathsf{capitalOf}(y,z) \to \mathsf{location}(x,z)$

RDF data:











- Standardised KR language
 - RDF provides a graphical data model
 - OWL provides a DL-based ontology language
- Developed as part of W3C's Semantic Web project

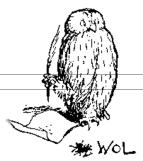
"A new form of Web content that is meaningful to computers will unleash a revolution of new possibilities" (!)





Computational Logic : Group





- Standardised KR language
 - RDF provides a graphical data model
 - OWL provides a DL-based ontology language
- Developed as part of W3C's Semantic Web project
- Now widely used in science, healthcare and Industry
- Often referred to as "semantic technology"



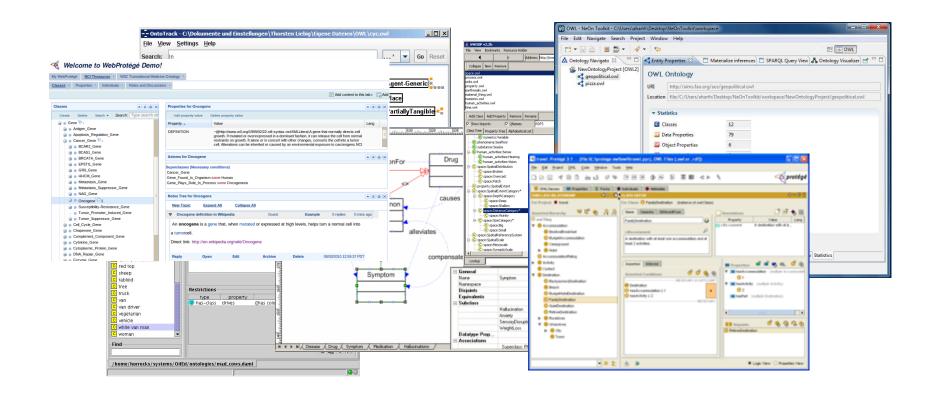


RDF + OWL



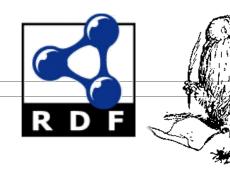


Tools:





RDF + OWL



Reasoners:

Hermit FaCT++ ORACLE















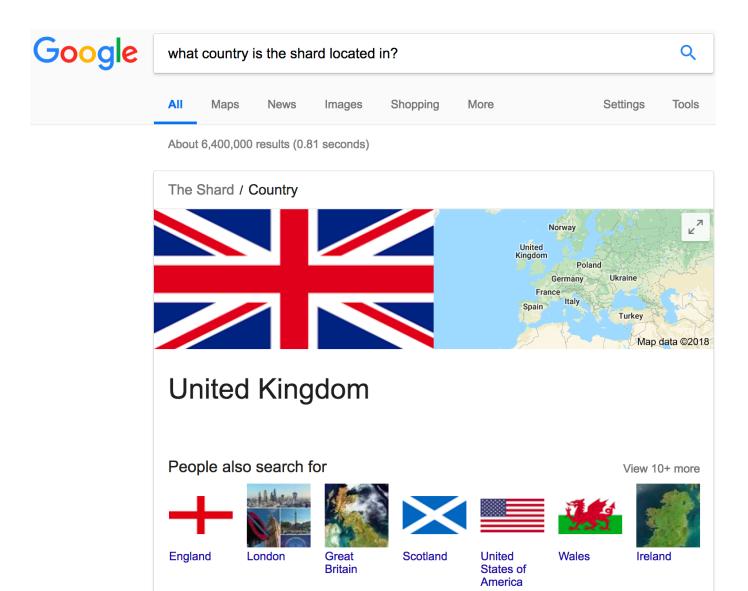


Applications: Question Answering





Applications: Question Answering





Applications: Clinical Records

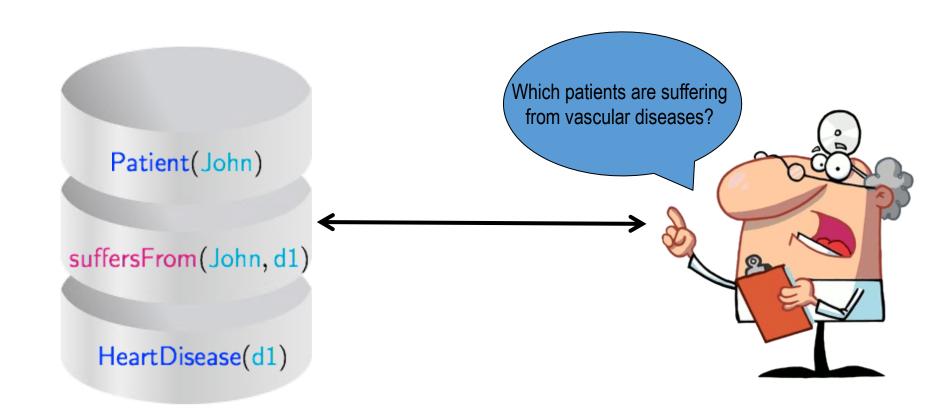
SNOMED CT The global

- SNOMED is a huge medical ontology
 - More than 500,000 terms!
- Why SNOMED? Let's ask Healthcare experts!
 - "We need a clinical ontology that is universal, so any term I use is the same as every other colleague around the country"
 - "SNOMED is the glue that binds the clinical community together and is the platform for all clinically relevant information"
- Used to annotate patient records in more than 20 countries, including UK, USA, New Zealand,

. . .

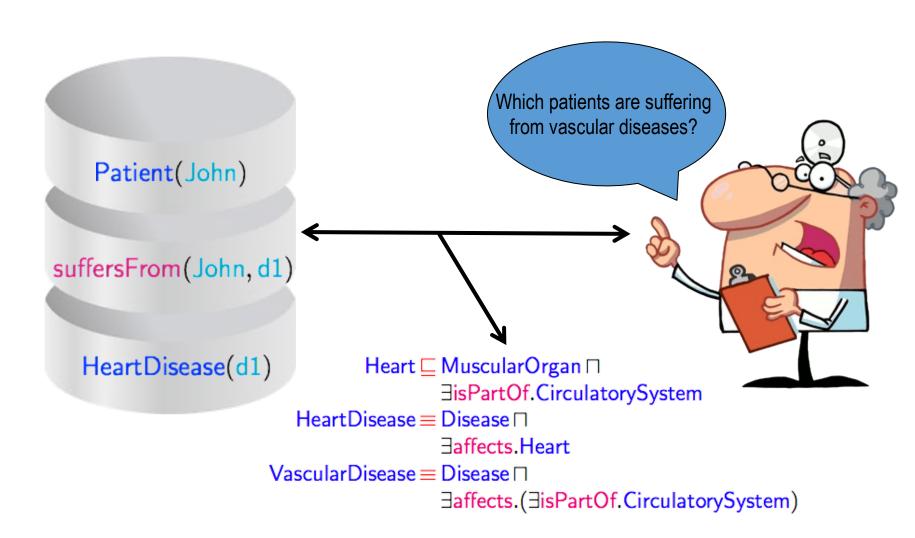


Applications: Clinical Records





Applications: Clinical Records





Course Structure

Introduction

Basic Logics for KRR

- Propositional Logic and First-Order Logic
- Horn Logics and Datalog

Description Logics for KRR

- Description Logics Syntax and Semantics I
- Description Logics Syntax and Semantics II
- Description Logics Reasoning with Data

KRR beyond Classical Logic

- Nonmonotonic Reasoning I
- Nonmonotonic Reasoning II
- Inconsistency Handling
- Argumentation
- Uncertainty



Reading List

Primary Text

 Handbook of Knowledge Representation. Frank van Harmelen, Vladimir Lifschitz and Bruce Porter (Eds). Foundations of Artificial Intelligence, 2008.

Supplementary Texts

- An Introduction to Description Logic. Franz Baader, Ian Horrocks, Carsten Lutz, Uli Sattler
- Foundations of Semantic Web Technologies. Chapman & Hall/ CRC Textbooks in Computing. Pascal Hitzler, Markus Krötzsch, and Sebastian Rudolph, 2009.