Welcome to INFO216: Knowledge Graphs Spring 2023

Andreas L Opdahl <Andreas.Opdahl@uib.no>

### About me

- Background:
  - siv ing, dr ing from NTNU
  - University of Bergen (since early 1990-ies)
  - part-time programmer / consulting for industry
  - several "Forskningsråd" and EU projects and networks
- Central research interest:
  - modelling of information systems and enterprises
  - semantic modelling and modelling languages
  - semantic technologies: ontologies and knowledge graphs
  - knowledge graphs for the news





INFO216: Knowledge Graphs

### Recent project: BDEM

- Leveraging *Big Data for Emergency Management* 
  - how can semantic technologies play a part?
  - developed a new Master course: INFO319 on Big Data



### Recent project: **PROB**







Norwegian University of Science and Technology







telenor The University Of Sheffield.

### Recent project: Transfeed



#### http://newsangler.uib.no

### **Ongoing project: News Angler**



"Wolftech News supports and improves the workflows in a newsroom through mobile solutions for field work that are integrated with central systems for news monitoring, resource management, news editing, and multi-platform publishing"

- 1) Harvesting and analysing messages
- 2) Growing a semantic news graph
  - concepts, named entities, context...
- 3) Analysing working texts (stories)
- 4) Identifying background information
- 5) Prioritising and preparing
- 6) Journalistic and editorial preferences

*Research:* graph, searches, preparation, preferences, language, scaling

### Active centre: Media Futures



### Session 1: Introduction to Knowledge Graphs

- Themes:
  - what are knowledge graphs (KGs)?
    - what is the problem?
    - who uses them?
    - examples of important open KGs
  - exercise 1: Getting started ...
  - about INFO216
    - organisation of the course
    - practical information

### Readings

- Sources:
  - Allemang, Hendler & Gandon (2020): Semantic Web for the Working Ontologist, 3<sup>rd</sup> edition: chapters 1-2
  - Blumauer & Nagy (2020): Knowledge Graph Cookbook – Recipes that Work: pages 27-55, 105-122 (*supplementary*)
- Materials in the wiki <http://wiki.uib.no/info216>:
  - Tim Berners-Lee talks about the semantic web
  - links to a few important open KGs



THE KNOWLEDGE GRAPH COOKBOOK RECIPES THAT WORK



AND HELMUT NACY



INFO216: Knowledge Graphs

# What are knowledge graphs?

### What is the problem?

- Lots of data around
  - internal, public, social, open
- Enormous potential to solve, simplify, speed up many critical information handling problem
  - but data are mostly not *linked* (think of a world wide web without document links!)
- What if linkable data could always be linked automatically?
- Tim Berners-Lee's semantic web vision (ca 2000)

Tim Berners-Lee: <http://www.youtube.com/watch?v=HeUrEh-nqtU>



Tim Berners-Lee Inventor of the World Wide Web (WWW, 1989)



INFO216: Knowledge Graphs

### What is the problem?

- Lots of data around
  - internal, public, social, open
- Enormous potential to solve, simplify, speed up many critical information handling problem
  - but data are mostly not *linked* (think of a world wide web without document links!)
- What if linkable data could always be linked automatically?
- The *semantic web vision* (ca 2000)

- Need standard ways of representing data and knowledge:
  - technical
    - standard formats, languages, and techniques to share data
  - semantic
    - standard identifiers and terms to share meaning
  - formal
    - support formal *rules and reasoning*



INFO216: Knowledge Graphs

 A graph of nodes connected by directed edges





INFO216: Knowledge Graphs

- A graph of nodes connected by directed edges
- Nodes can represent resources or values





INFO216: Knowledge Graphs

- A graph of nodes connected by directed edges
- Nodes can represent resources or values
- Edges represent *relations*





INFO216: Knowledge Graphs

- A graph of nodes connected by directed edges
- Nodes can represent resources or values
- Edges represent *relations*
- Each node–edge–node triple represents a fact





INFO216: Knowledge Graphs

- A graph of nodes connected by directed edges
- Nodes can represent resources or values
- Edges represent *relations*
- Each node–edge–node triple represents a fact
  - subject-predicate-object





INFO216: Knowledge Graphs

- A graph of nodes connected by directed edges
- Nodes can represent resources or values
- Edges represent *relations*
- Each node–edge–node triple represents a fact
  - subject–predicate–object
  - head-relation-tail





INFO216: Knowledge Graphs

- A graph of nodes connected by directed edges
- Nodes can represent resources or values
- Edges represent *relations*
- Each node–edge–node triple represents a fact
  - subject–predicate–object
  - head-relation-tail
- A knowledge graph represents knowledge as triples connected by nodes





INFO216: Knowledge Graphs



INFO216: Knowledge Graphs

- Technical:
  - standard *formats* for storing and exchanging graphs
    - including types of values (strings, numbers, times, dates, etc.)
  - specialised databases and standard query languages
  - APIs for all major programming languages



(c) Andreas L Opdahl, 2023

- Technical:
  - standard *formats* for storing and exchanging graphs
    - including types of values (strings, numbers, times, dates, etc.)
  - specialised databases and standard query languages
  - *APIs* for all major programming languages
- Semantic:
  - large repositories of *unique identifiers* for individual resources
  - vocabularies with unique identifiers for resource types and relations
  - graph embeddings and graph neural networks for machine learning



- Technical:
  - standard *formats* for storing and exchanging graphs
    - including types of values (strings, numbers, times, dates, etc.)
  - specialised databases and standard query languages
  - *APIs* for all major programming languages
- Semantic:
  - large repositories of *unique identifiers* for individual resources
  - vocabularies with unique identifiers for resource types and relations
  - graph embeddings and graph neural networks for machine learning
- Formal:
  - rule languages and inference engines
  - formal logic systems and reasoning engines

(c) Andreas L Opdahl, 2023



### Why knowledge graphs?

- Ease of exchanging, reusing information
  - inherent semantics become clearer
  - less dependency on context
- Ease of *interlinking*, *enriching* information
  - semantic data can be combined in new ways
  - open reference datasets
  - general and specialised knowledge bases
- Ease of extending
  - no pre-defined data schemas ("schema-on-read")
  - easy to add new types of resources and new relations
- Well-matched with the needs of big data and machine learning!



INFO216: Knowledge Graphs

#### Google **Tencent** 腾讯 Who is > Bing Bai de 百度 Pub Med using this? Alibaba.com facebook All the big • ANTONI LEEUWENHOEK players! The TSCHE New Hork Google's • MAASTRO OTHEK BBC Times Knowledge europeana Graph REUTERS Microsoft's • **€)EPA** Satori Environmental Protection National Library of Sweden Amazon's • RENAULT IOS Product Graph European SIEMENS Press Walmar Commission ...and (almost) • BEST Deloitte everyone else SPRINGER NATURE amazon.com accenture ELSEVIER

From Frank van Harmelen's keynote at CAiSE 2018

### Knowledge graphs at Amazon ( $\rightarrow$ S06)



- Let shoppers find the best products that fit their needs
  - allow greater variation in search terms
  - allow complex queries
- Ambition: to structure all of the world's information as it relates to everything available on Amazon
- Describe every product on Amazon
  - both products and non-products
  - both concrete and abstract concepts
  - link related entities, both internal and external
- Enhanced customer experience
  - visit Amazon to see what's new or interesting
  - discover ways to simplify and enrich their lives



INFO216: Knowledge Graphs

### Knowledge graphs everywhere

- BBC's content management, ontologies, BBC Things
- Google, Bing, Yahoo... (schema.org) (2011)
- Google's Knowledge Graph (2012), Microsoft's Satori
- Facebook's Open Graph and Graph Search (2013)
- Thomson Reuters, Bloomberg...
- Uber Eats' food graph



(c) Andreas L Opdahl, 2023

### Some knowledge graphs we will look at in INFO216

- You have already seen Google's KG many times:
  - the "knowledge panels" in search results
- Wikidata (https://www.wikidata.org/)
  - part of the Wikimedia family, feeds factual information to Wikipedia
- DBpedia (https://www.dbpedia.org, https://dbpedia.org/page/Bergen)
  - extracts factual information from Wikipedia
- GeoNames (https://www.geonames.org/)
  - global database of place names (toponyms), relations and types
- BabelNet (https://babelnet.org/)
  - a multi-lingual dictionary and thesaurus
- Linked Open Vocabularies (LOV, https://lov.linkeddata.es/dataset/lov/)
  - a collection of knowledge graphs that describe vocabularies (also called ontologies) for other knowledge graphs



INFO216: Knowledge Graphs

### Exercise 1: Getting started with VSCode, Python and RDFlib



INFO216: Knowledge Graphs

### How can we represent semantic KGs?

- Resource Description Framework (RDF  $\rightarrow$  S02)
- RDF models (KGs) consist of statements (triples)
  - of subject predicate object.
  - or subject predicate literal.



(c) Andreas L Opdahl, 2023

### How can we represent semantic KGs?

- Resource Description Framework (RDF  $\rightarrow$  S02)
- RDF models (KGs) consist of statements (triples)
  - of subject predicate object.
  - or subject predicate literal.
- The subject:
  - must be a resource
  - physical, informational, conceptual...
- The predicate:
  - must be a property (subtype of resource)
- The object:
  - is either a resource
  - or a *literal* (or a *value*: string, number... not a resource)





### How can we represent semantic KGs?

- Resource Description Framework (RDF  $\rightarrow$  S02)
- RDF models (KGs) consist of statements (triples)
  - of subject predicate object.
  - or subject predicate literal.
- Serialisations, e.g., in *Turtle*:



@prefix ex: <http://example.org/> .

ex:Roger\_Stone ex:name "Roger Stone" ; ex:occupation ex:lobbyist ; ex:significant\_person ex:Donald\_Trump .

ex:Donald\_Trump ex:name "Doald Trump" . Uniform Resource Identifiers (URIs) identify resources, including types and relations



INFO216: Knowledge Graphs

### RDFLib graphs

- RDFLib:
  - an API for programming RDF and SPARQL in Python
  - simple, powerful and pythonic
  - parsers and serialisers for most RDF formats
  - a Graph interface:
    - a graph holds an RDF model
    - is a Python collection (set) of triples
    - supports adding, removing, listing, and searching for triples
    - supports writing to and reading from RDF files
    - >>> from rdflib import Graph
    - >>> model = Graph()



INFO216: Knowledge Graphs

### **RDFLib** resources

- URIRef:
  - a node with a URI (represents resources, types, and properties)

>>> from rdflib import URIRef
>>> donaldTrump = URIRef('http://example.org/Donald\_Trump')

- Namespace:
  - a more compact way to create resources, types, and properties

>>> from rdflib import Namespace
>>> ex = Namespace('http://example.org/')
>>> donaldTrump = ex.Donald\_Trump



INFO216: Knowledge Graphs

### **RDFLib** triples

- Triples / statements: •
  - ordinary 3-item Python tuples
    - immutable sequences

>>> triple = (s, p, o) # creates a triple # returns the subject, etc. >>> s[0]

add/remove triples to/from graph

>>> g.add( (res1, prop, res2) )

- close persisted model:

>>> q.close()

or: >>> g.add( (res, prop, lit) )





INFO216: Knowledge Graphs

### **RDFLib** literals

- Literal:
  - a typed value
    - >>> from rdflib import Literal
    - >>> lit = Literal('President of the United States')
  - strings can be language-tagged

>>> lit = Literal('President of the United States', 'en') >>> lit = Literal('美利堅合眾國的國家元首、政府首腦兼三軍統帥' , 'zh-ch')



(c) Andreas L Opdahl, 2023

### Serialising and parsing

• Serialising:

>>> g.serialize(destination=*fileNameStr,* format='ttl')

>>> ttl\_str = g.serialize(format='ttl').decode()

>>> ttl\_str = g.serialize(format='json-ld').decode()

• Parsing:

>>> g.parse(location=fileNameStr, format='ttl')
>>> g.parse(source=webURLStr, format='ttl')
>>> g.parse(data=pythonStr, format='ttl')



(c) Andreas L Opdahl, 2023

### Listing statements

- Retrieving statements (triples):
  - >>> for triple in g:
  - >>> do\_something(triple) # s = triple[0], p = triple[1], o = triple[2]
    >>> for s, p, o in g:
  - >>> do\_something(s, p, o)
  - >>> for s, p, o in g.triples( (sub, pred, obj) ):
  - >>> do\_something(s, p, o) # sub, pred, obj can be None
  - >>> for s, p, o in g[ sub : pred : obj ]:
  - >>> do\_something(s, p, o) # sub, pred, obj can be empty:

# s, p, o must match

Python overloading!



(c) Andreas L Opdahl, 2023

## About INFO216



INFO216: Knowledge Graphs

### Purpose

- To learn theories, techniques, tools, and best practices for managing knowledge graphs.
- To acquire understanding and skills for programming applications that use and produce such data and metadata.
- To learn about existing sources of and standards for big, open, and semantic data.
- To gain practical experience in developing knowledge graph-based applications using technologies such as RDF, RDFS, OWL, SPARQL, and JSON-LD.



### Curriculum

- Course book (*the whole book is mandatory*):
  - Allemang, Hendler & Gandon (2020).
     Semantic Web for the Working Ontologist,
     Effective Modeling for Linked Data, RDFS and OWL (Third Edition)
- Supplementary course book (*suggested, not mandatory*):
  - Blumauer & Nagy (2020).
     The Knowledge Graph Cookbook Recipes that Work
- Additional readings (both *mandatory* and *suggested*) will be made available in the course wiki: https://wiki.uib.no/info216
- The lectures and lectures notes are also *mandatory* parts of the curriculum.



INFO216: Knowledge Graphs

### **Practical**

- 14 lectures:
  - Tuesdays 1215-1400
- 14 lab weeks:
  - 2 hours of weekly lab groups
  - starting this week, no labs week 10 and 14 (Easter)
  - seminar/lab leader: Robin Johansen Bøe <Robin.Boe@student.uib.no>
- Evaluation:
  - individual, written 4-hour exam
- Requirements:
  - participation in 75% of labs
- Course wiki:
  - http://wiki.uib.no/info216

(c) Andreas L Opdahl, 2023



### Lecture plan (tentative)

- 1. Introduction to KGs
- 2. Representing KGs (RDF)
- 3. Querying and updating KGs (SPARQL)
- 4. Open KGs 1
- 5. Open KGs 2
- 6. Enterprise KGs
- 7. Rules (RDFS)

- 8. Ontologies (OWL)
- 9. Vocabularies
- 10. Reasoning about KGs (DL)
- 11. Formal ontologies (OWL-DL)
- 12. KG embeddings 1
- 13. KG embeddings 2
- 14. Knowledge engineering

### You learn KGs best through practice: do the lab exercises thoroughly!



(c) Andreas L Opdahl, 2023

# Next week: Representing KGs (more about RDF)



INFO216: Knowledge Graphs