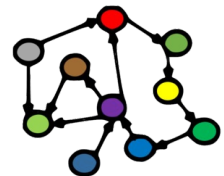


Welcome to INFO216:  
Knowledge Graphs  
Spring 2023

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



## *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



SAN DIEGO STATE  
UNIVERSITY



# VESTLANDSFORSKING

*Recent project:*



# Ubiquitous Data-Driven Urban Mobility

WESTERN NORWAY RESEARCH INSTITUTE  
**VESTLANDSFORSKING**



NTNU  
Norwegian University of  
Science and Technology



**toi**

Transportøkonomisk institutt  
Stiftelsen Norsk senter for samferdselsforskning

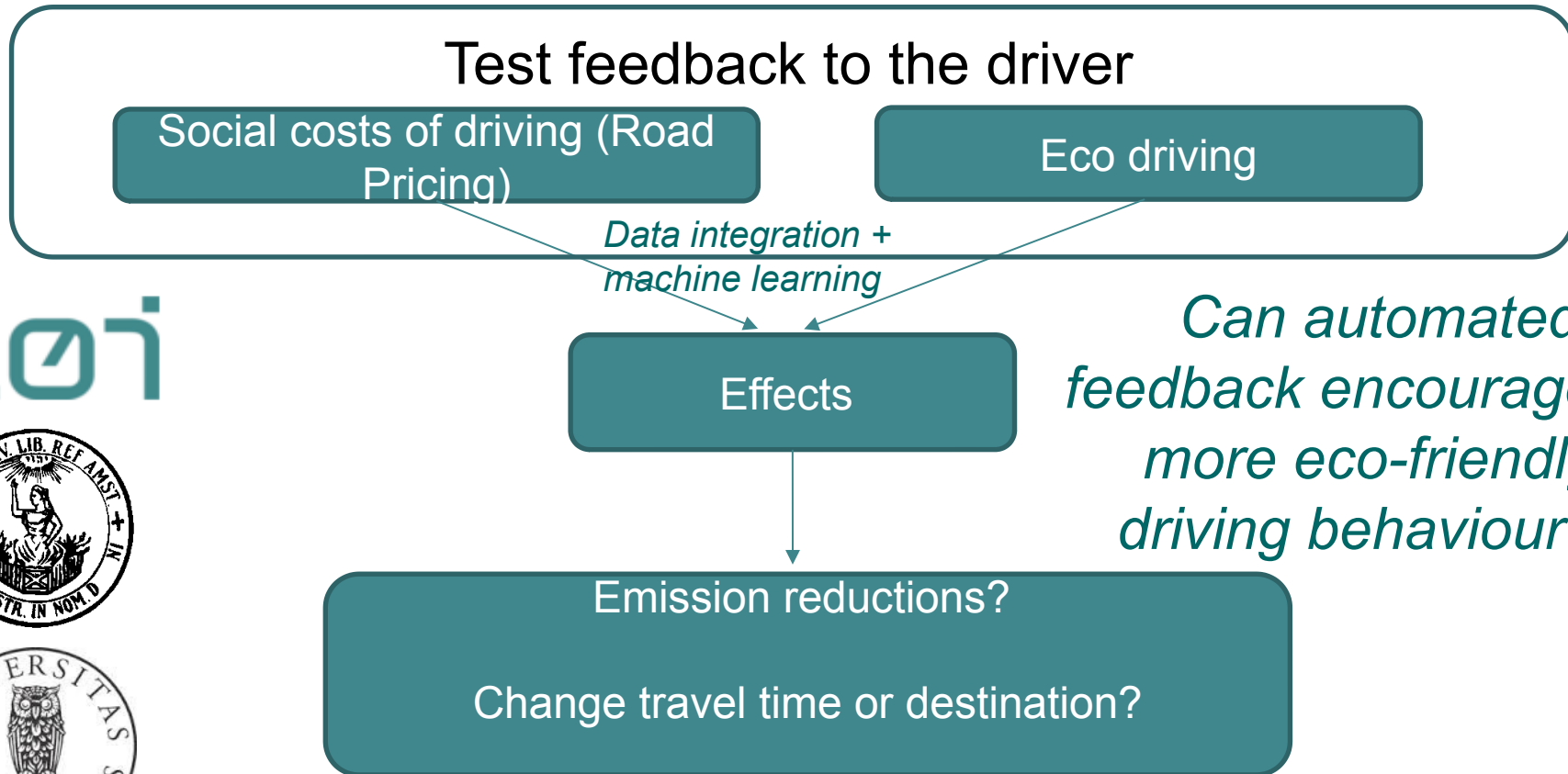


The  
University  
Of  
Sheffield.



telenor

# Recent project: Transfeed

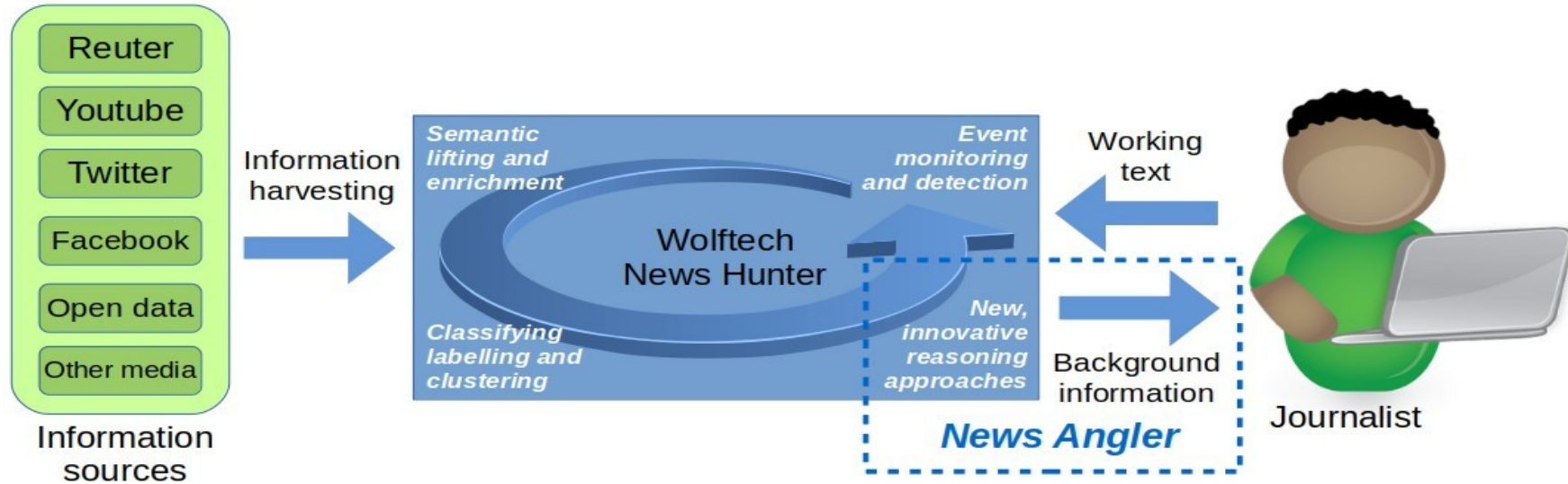


*Can automated feedback encourage more eco-friendly driving behaviour?*

toi



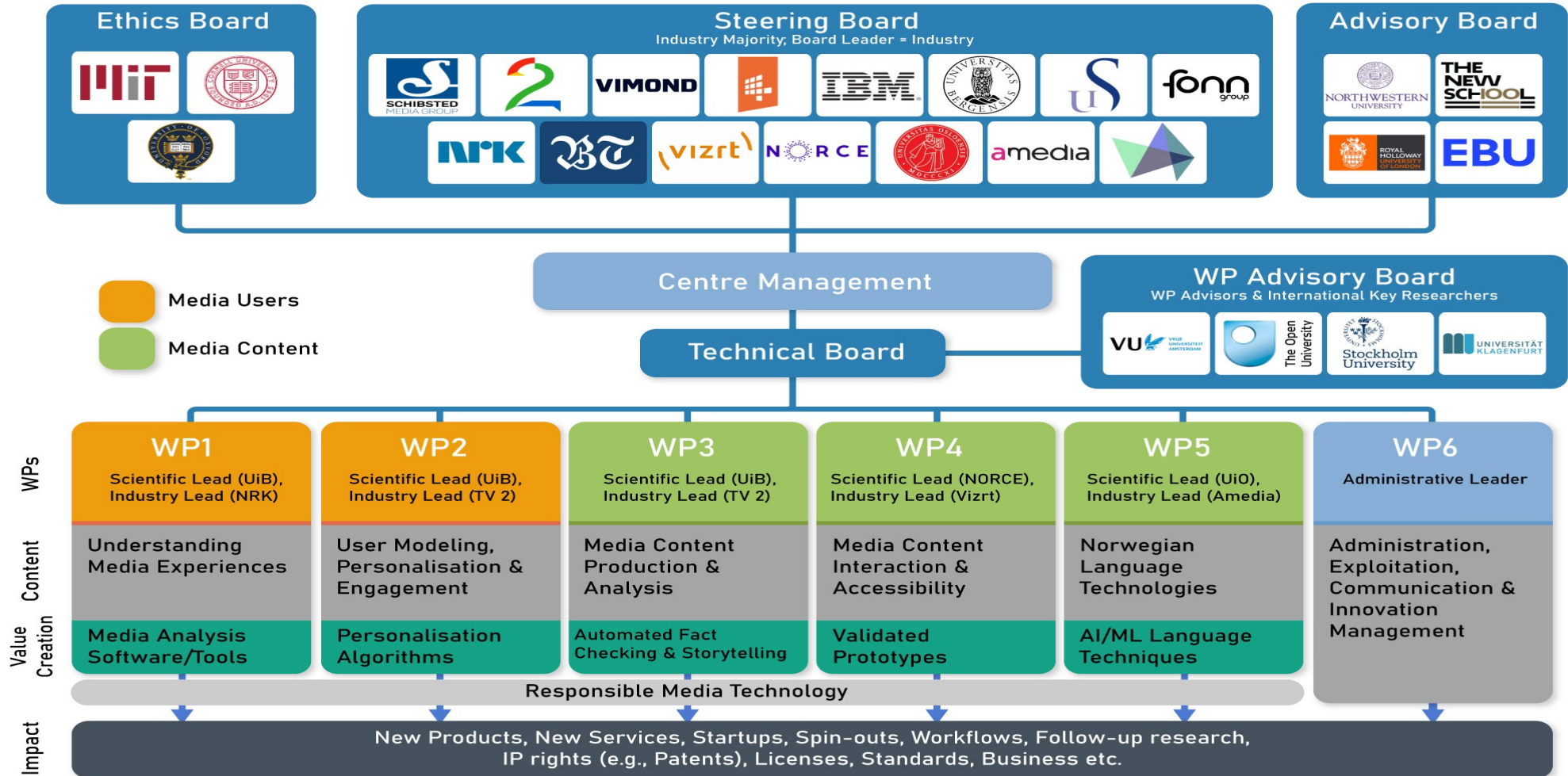
## Ongoing project: News Angler



*“Wolftch 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



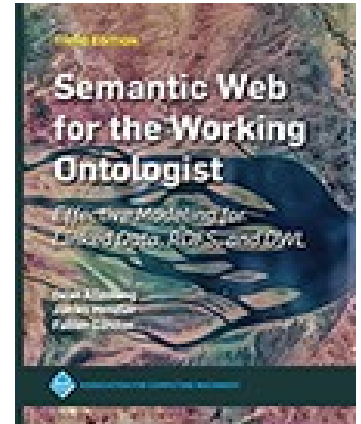
# Session 1: Introduction to Knowledge Graphs

- 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

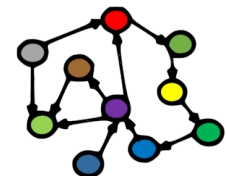


THE KNOWLEDGE GRAPH  
**COOKBOOK**  
RECIPES THAT WORK



ANDREAS BLUMAUER  
AND HELMUT NAGY

1st edition, 2020



What are  
knowledge graphs?

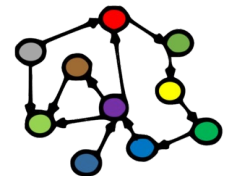
# What is the problem?

- *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>>

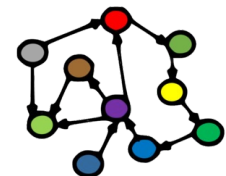


Tim Berners-Lee  
Inventor of the  
World Wide Web  
(WWW, 1989)



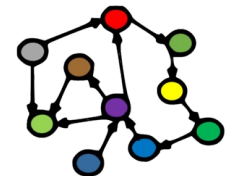
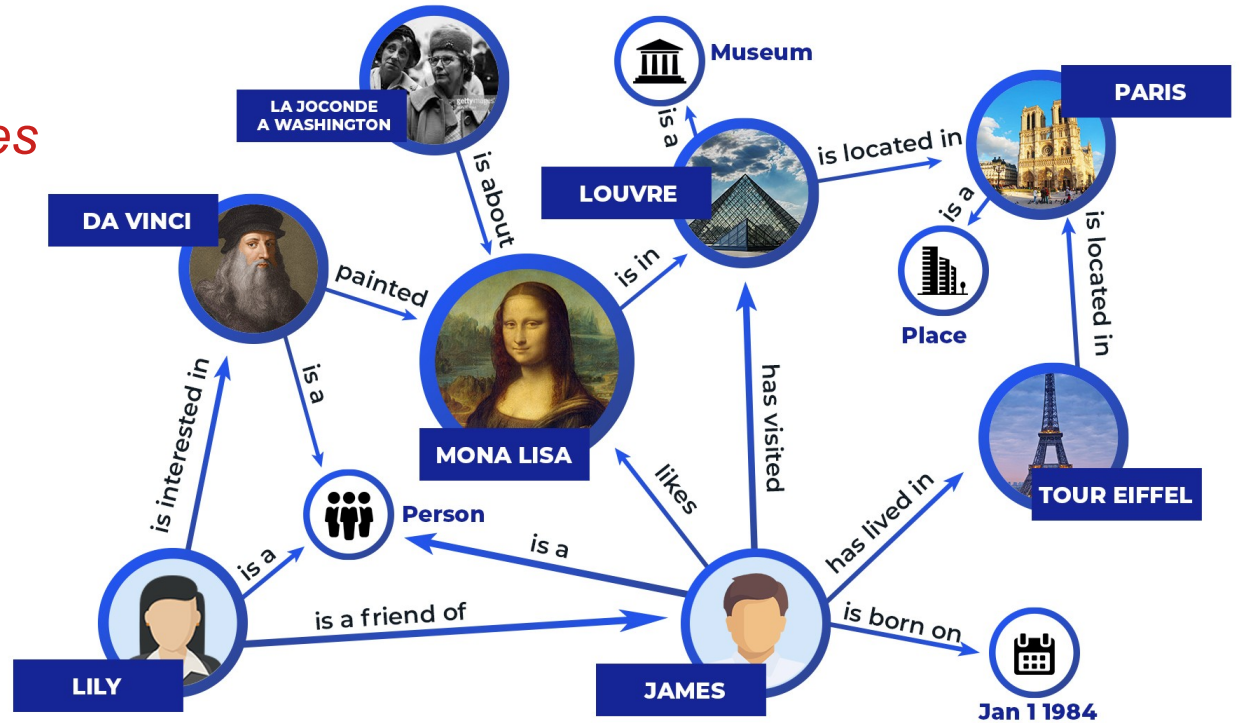
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  - but data are mostly not *linked* (think of a world wide web without document links!)
- *What if linkable data could always be linked automatically?*
- The *semantic web vision* (ca 2000)
- Need *standard ways of representing data and knowledge*:
  - technical
    - standard formats, languages, and techniques to *share data*
  - semantic
    - standard identifiers and terms to *share meaning*
  - formal
    - support formal *rules and reasoning*



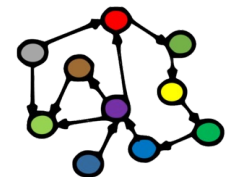
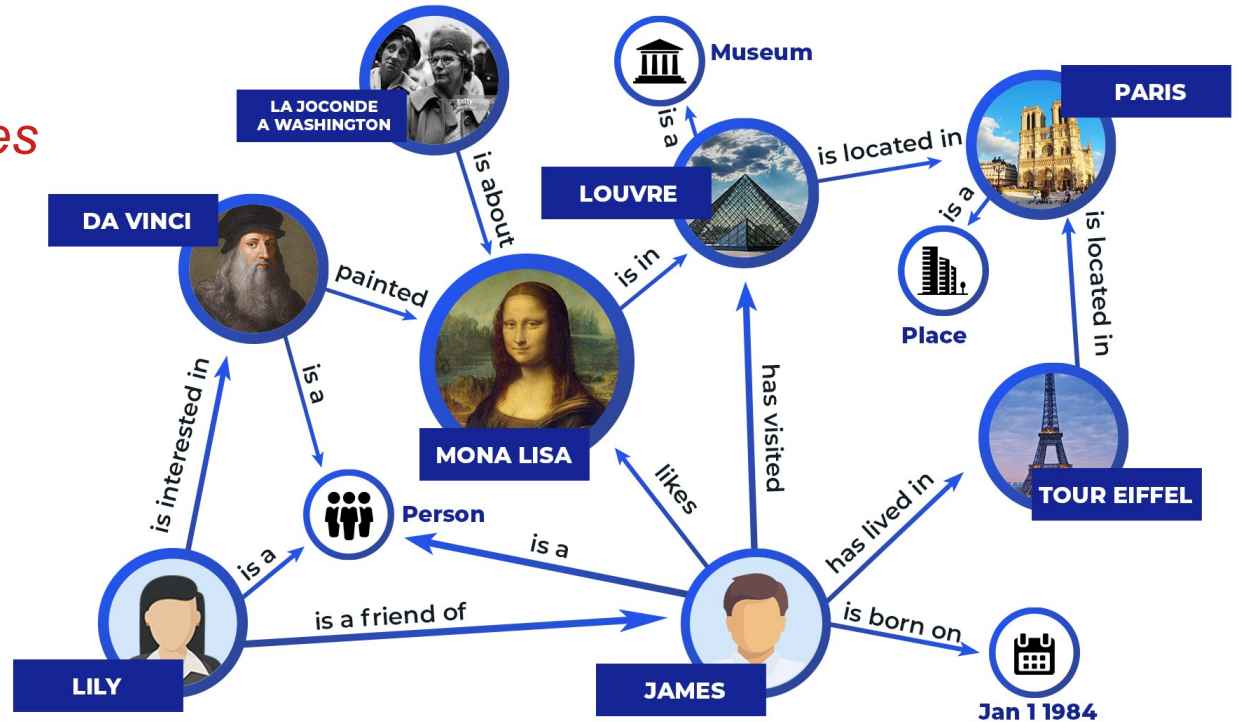
# Knowledge graph

- A *graph* of *nodes* connected by directed *edges*



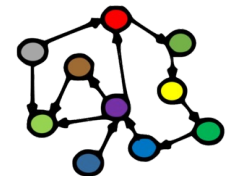
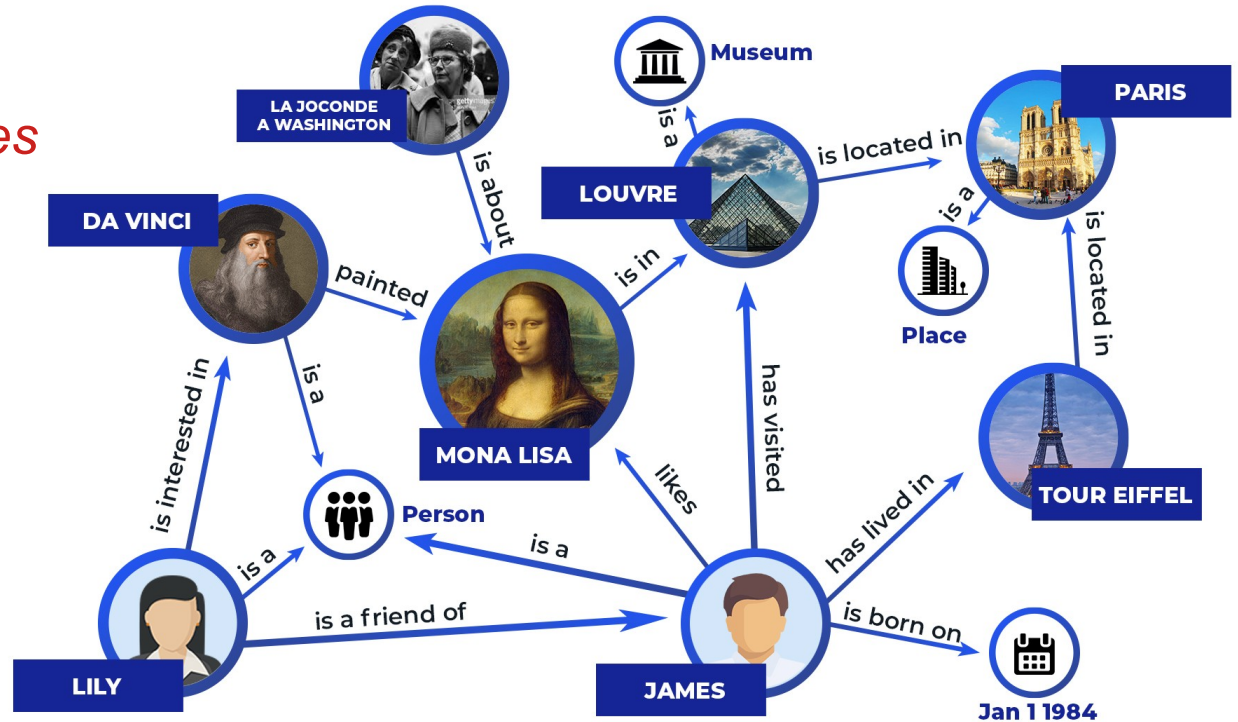
# Knowledge graph

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



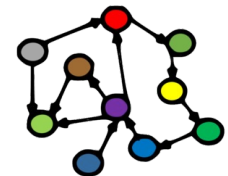
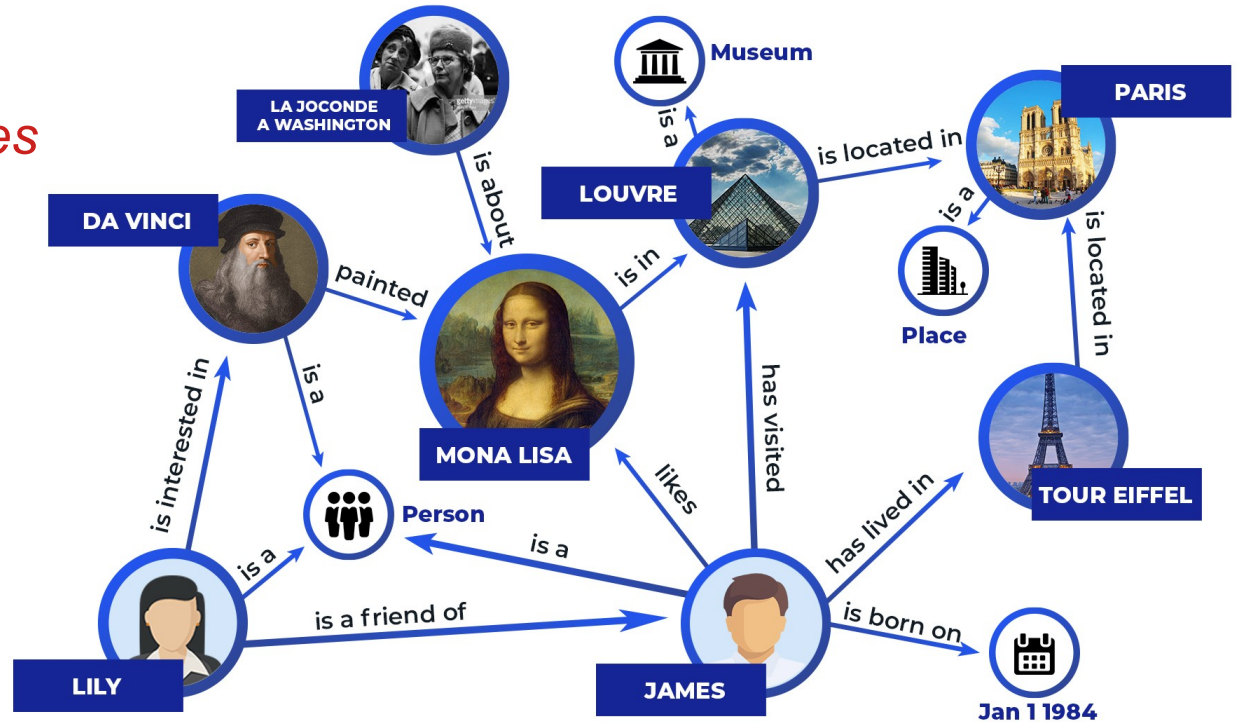
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# Knowledge graph

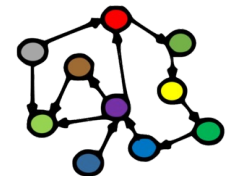
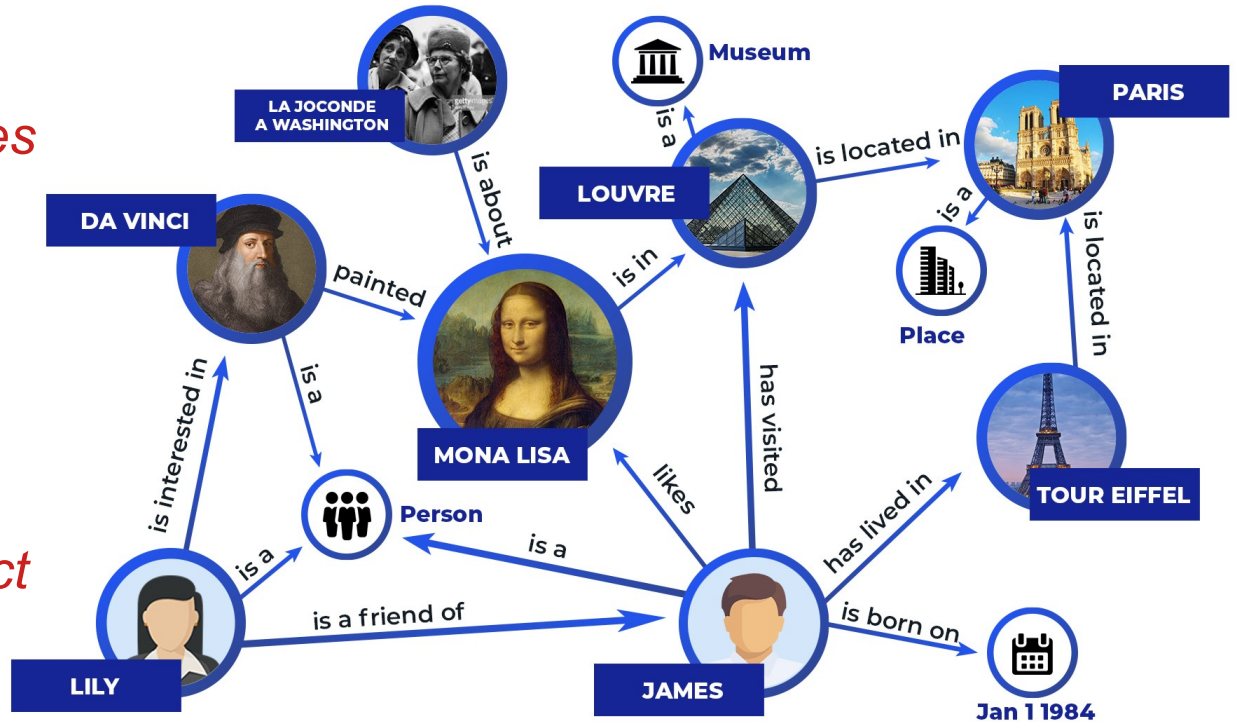
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- Edges represent *relations*
- Each node–edge–node *triple* represents a *fact*





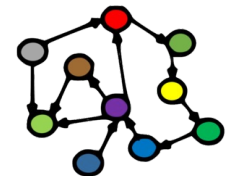
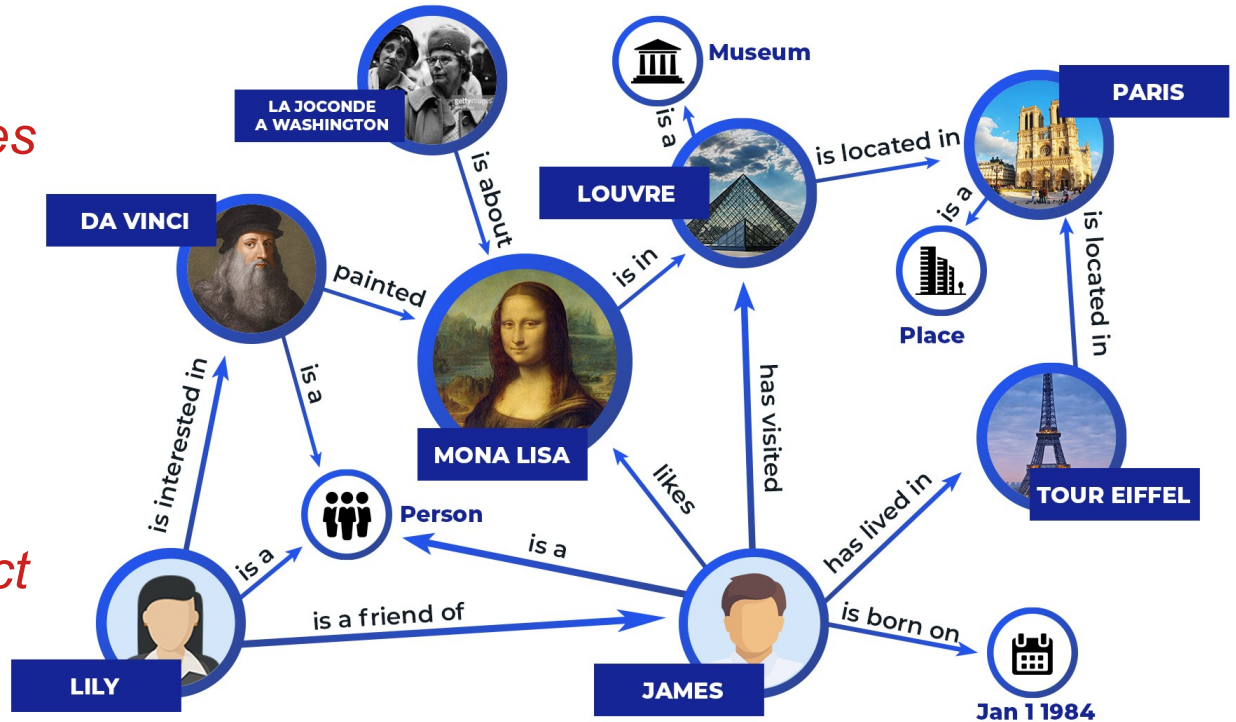
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  - *subject–predicate–object*



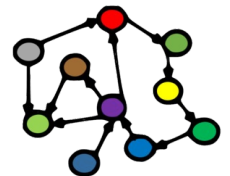
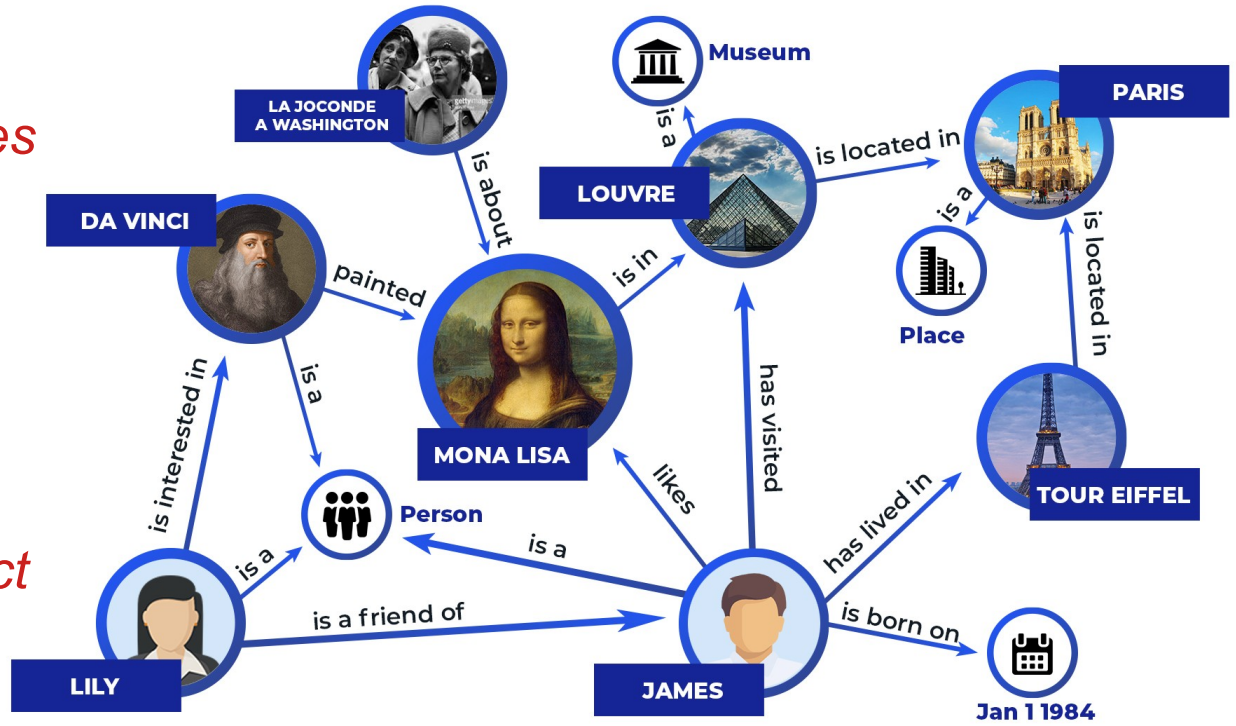
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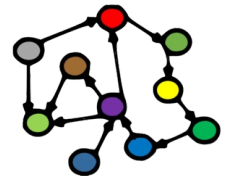


# Knowledge graph

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- 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*

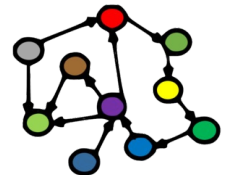


And there is more...



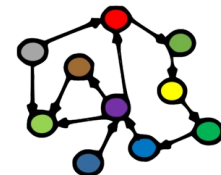
# And there is more...

- 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



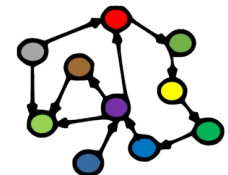
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- Semantic:
  - large repositories of *unique identifiers* for individual resources
  - *vocabularies* with unique identifiers for resource types and relations
  - *graph embeddings* and *graph neural networks* for machine learning



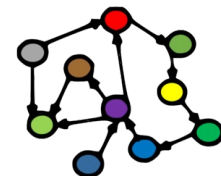
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  - large repositories of *unique identifiers* for individual resources
  - *vocabularies* with unique identifiers for resource types and relations
  - *graph embeddings* and *graph neural networks* for machine learning
- Formal:
  - rule languages and *inference engines*
  - *formal logic* systems and reasoning engines



# 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!*





# Who is using this?

- *All the big players!*
- Google's Knowledge Graph
- Microsoft's Satori
- Amazon's Product Graph
- ...and (almost) everyone else

Tencent 腾讯

UniProt USGS

Google  
Bing

Alibaba.com

Baidu 百度

PubMed

facebook

DEUTSCHE  
NATIONAL  
BIBLIOTHEK

ANTONI  
VAN  
LEEUVENHOEK  
FOUNDATION  
MAASTRO

The  
New York  
Times

CE  
ITY

europæana

NXP

BBC



REUTERS

BIBLIOTEKET  
National Library  
of Sweden

RENAULT

EPA  
United States  
Environmental Protection  
Agency

IOS  
Press

Walmart

SIEMENS



BEST  
BUY

Deloitte.



SPRINGER NATURE

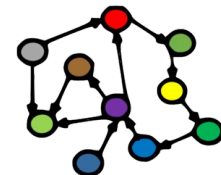
amazon.com

accenture

# Knowledge graphs at Amazon (→S06)

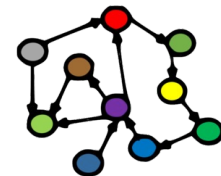


- Let shoppers find the best products that fit their needs
  - allow greater variation in search terms
  - allow complex queries
- Ambition: *to structure all of the world's information as it relates to everything available on Amazon*
- Describe every product on Amazon
  - both products and non-products
  - 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



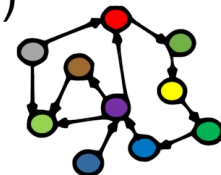
# Knowledge graphs everywhere

- 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



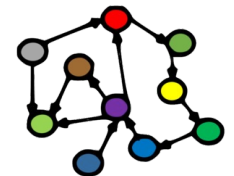
# Some knowledge graphs we will look at in INFO216

- 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



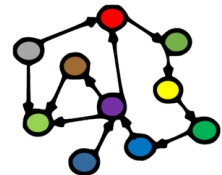
# Exercise 1:

## Getting started with VSCode, Python and RDFlib



# How can we represent semantic KGs?

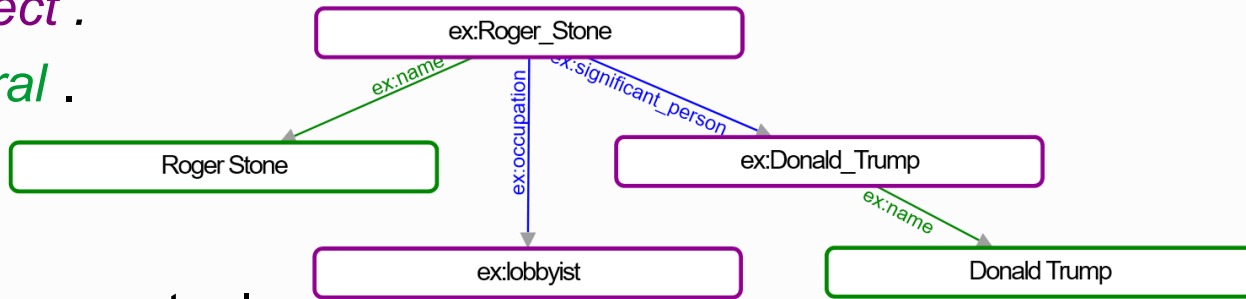
- Resource Description Framework (RDF → S02)
- RDF models (KGs) consist of statements (triples)
  - of *subject predicate object* .
  - or *subject predicate literal* .



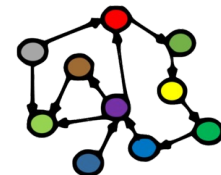
# How can we represent semantic KGs?

- Resource Description Framework (RDF → 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...

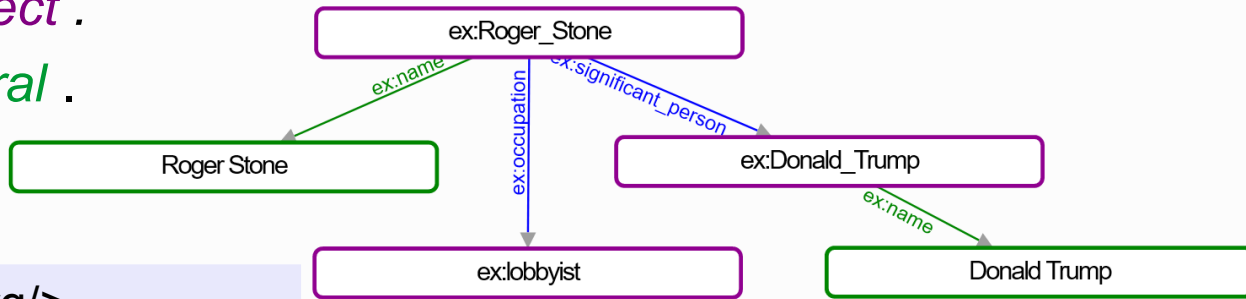


- The predicate:
  - must be a *property* (subtype of *resource*)
- The object:
  - is either a *resource*
  - or a *literal* (or a *value*: string, number... – *not a resource*)



# How can we represent semantic KGs?

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



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

ex:Roger\_Stone

ex:name "Roger Stone" ;

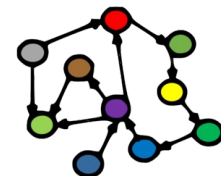
ex:occupation ex:lobbyist ;

ex:significant\_person ex:Donald\_Trump .

ex:Donald\_Trump

ex:name "Donald Trump" .

Uniform Resource Identifiers (URIs) identify resources, including types and relations

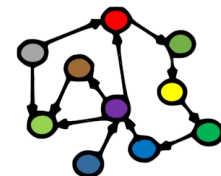




# RDFLib graphs

- **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
>>> g = Graph() # the RDF model
```



# RDFLib resources

- **URIRef:**

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

```
>>> from rdflib import URIRef
```

```
>>> donaldTrump = URIRef('http://example.org/Donald_Trump')
```

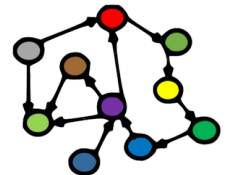
- **Namespace:**

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

```
>>> from rdflib import Namespace
```

```
>>> ex = Namespace('http://example.org/')
```

```
>>> donaldTrump = ex.Donald_Trump
```



# RDFLib triples

- Triples / statements:

- ordinary 3-item Python tuples
  - immutable sequences

```
>>> triple = (s, p, o) # creates a triple
```

```
>>> s[0] # returns the subject, etc.
```

- add/remove triples to/from graph

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

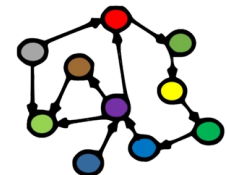
```
or: >>> g.add( (res, prop, lit) )
```

```
>>> g.remove( (res1, prop, res2) )
```

```
or: >>> g.remove( (res, prop, lit) )
```

- close persisted model:

```
>>> g.close()
```



# RDFLib literals

- Literal:

- a typed value

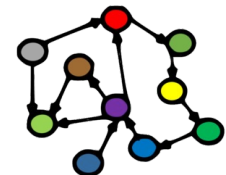
- >>> from rdflib import Literal

- >>> 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')



# Serialising and parsing

- Serialising:

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

```
>>> ttl_str = g.serialize(format='ttl').decode()
```

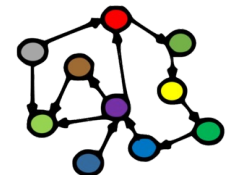
```
>>> ttl_str = g.serialize(format='json-ld').decode()
```

- Parsing:

```
>>> g.parse(location=fileNameStr, format='ttl')
```

```
>>> g.parse(source=webURLStr, format='ttl')
```

```
>>> g.parse(data=pythonStr, format='ttl')
```



# 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)
```

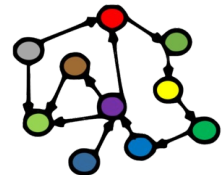
```
>>> for s, p, o in g.triples( (sub, pred, obj) ):
```

```
>>>     do_something(s, p, o) # sub, pred, obj can be None
```

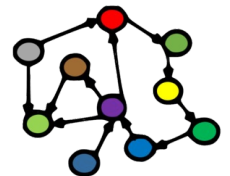
```
>>> for s, p, o in g[ sub : pred : obj ]:
```

```
>>>     do_something(s, p, o) # sub, pred, obj can be empty:  
                                # s, p, o must match
```

Python overloading!

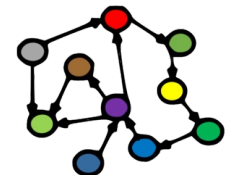


# About INFO216



# Purpose

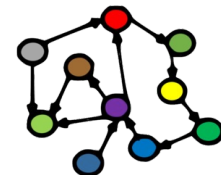
- To learn theories, techniques, tools, and best practices for managing knowledge graphs.
- To acquire understanding and skills for programming applications that use and produce such data and metadata.
- To learn about existing sources of and standards for big, open, and semantic data.
- To gain practical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





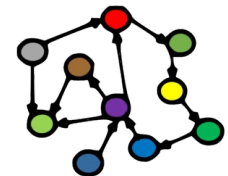
# Curriculum

- 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.



# Practical

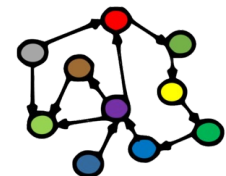
- 14 lectures:
  - Tuesdays 1215-1400
- 14 lab weeks:
  - 2 hours of weekly lab groups
  - starting this week, no labs week 10 and 14 (Easter)
  - seminar/lab leader: [Robin Johansen Bøe <Robin.Boe@student.uib.no>](mailto:Robin.Boe@student.uib.no)
- Evaluation:
  - individual, written 4-hour exam
- Requirements: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



# 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!***



# Next week: Representing KGs (more about RDF)

