## Welcome to INFO216: Knowledge Graphs Spring 2022

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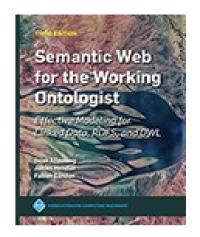
## Session 6: Enterprise Knowledge Graphs

- Themes:
  - Open Knowledge Graphs (← S05)
    - Linked Open Data resources / datasets
    - Wikidata, DBpedia, GDELT, EventKG GeoNames, WordNet, BabelNet...
  - Enterprise Knowledge Graphs (EKGs) (→ S06)
    - Google's knowledge graph
    - Amazon's product graphs
    - the News Hunter infrastructure and architecture



#### Readings

- Sources (suggested):
  - Blumauer & Nagy (2020):
    Knowledge Graph Cookbook Recipes that Work (parts 2 and 4)
- Material at http://wiki.uib.no/info216:
  - Introducing the Knowledge Graph: Things not Strings,
    Amit Singhal, Google (2012).
  - A reintroduction to our Knowledge Graph and knowledge panels, Danny Sullivan, Google (2020).
  - How Amazon's Product Graph is helping customers find products more easily, Arun Krishnan, Amazon (2018).
  - lecture slides







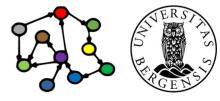




# Is anyone really using Knowledge Graphs?



• But...



#### **Tencent** 腾讯

















National Library



ANTONI

VAN LEEUWENHOEK









REUTERS













**Deloitte** 

accenture

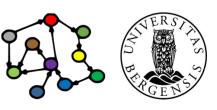






# Yes!

- But...
  - not quite as in the semantic web vision
  - not quite as in the LOD vision either
- Knowledge graphs are (additionally) becoming:
  - company internal
  - based on other technologies
    - such as general graph databases
  - not always linked to the LOD cloud



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Many of these ideas are widely adopted too, such as:

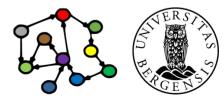
- microdata / schema.org
- RDF / SPARQL / ... for semantic data exchange
- graph representations in general

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Similar ideas, adapted to new uses and business contexts, using a combination of standard and other technologies

# Google's Knowledge Graph



### Google's Knowledge Graph

- Google Knowledge Graph (from 2012)
  - "Things, not Strings"
  - seeded from Freebase
  - facts from Wikipedia, Wikidata, CIA World Factbook
    - a growing number of other sources
  - enriched by natural-language parsing (NLP)
    - Google's Knowledge Vault
  - used internally for many purposes
  - visible in Google Search results (Knowledge Panels)
  - question answering in Google Assistant / Home

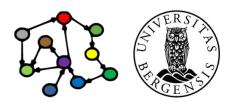
**Caution:** The public documentation is limited, so this is compiled based on presentations, technical notes, forums etc.



## Google's Knowledge Graph

- Coverage:
  - claimed
    - 18 billion facts (18G, norsk: 18 milliarder) about 570 million entities soon after start
  - 70 billion facts claimed in (2016)
  - 500 billion facts about five billion entities (2020)
    - ...perhaps 3 times the size of the LOD cloud
  - from English to multiple languages
- Critiques:
  - source attribution, incl. Wikipedia / Wikidata

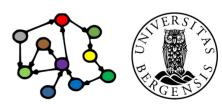
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## Google's Knowledge Vault Project

- Google Knowledge Vault
  - extends the Knowledge Graph
  - covers resources not from open semantic datasets
  - facts extracted from the whole web
    - NLP of text documents
    - HTML trees and tables
    - human annotated pages (e.g., schema.org)
  - probabilistic reasoning
    - graph-based priors
    - knowledge fusion

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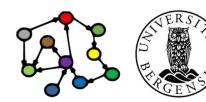


# Amazon's Knowledge Graph

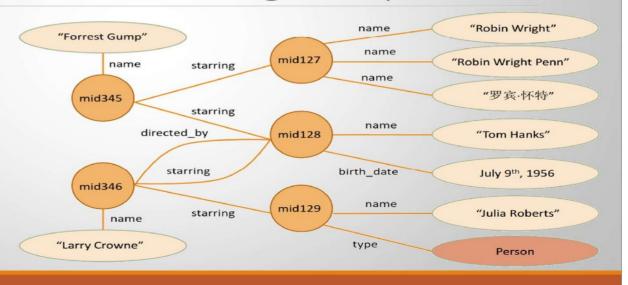


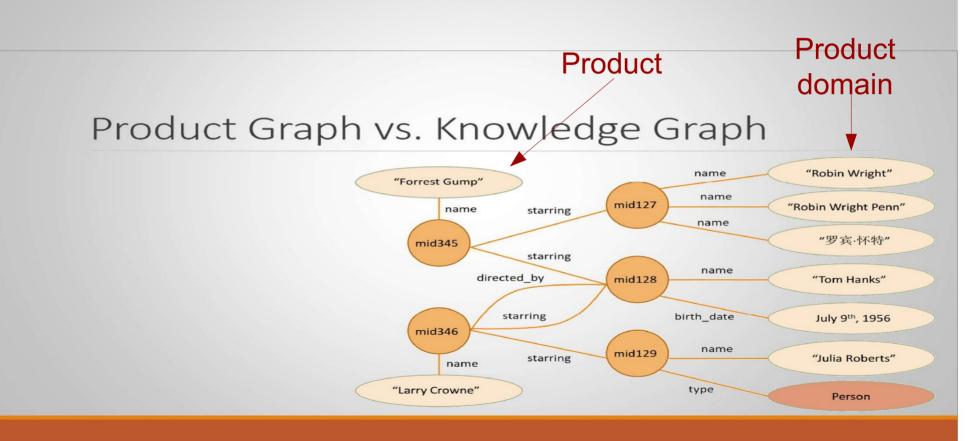
#### Amazon's ambition

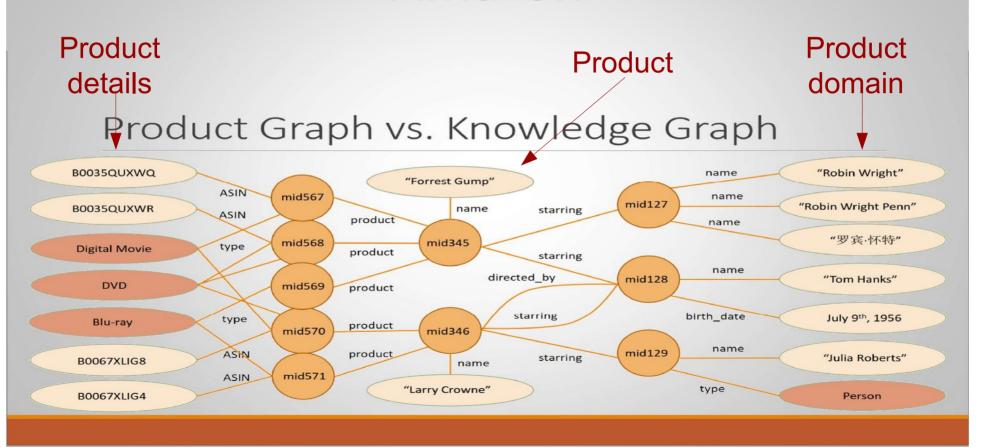
- Let shoppers find the best products that fit their needs
  - allow greater variation in search terms
  - allow complex queries
- Structure all of the world's information as it relates to everything available on Amazon
- Describe every product on Amazon
  - concrete and abstract concepts
  - products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

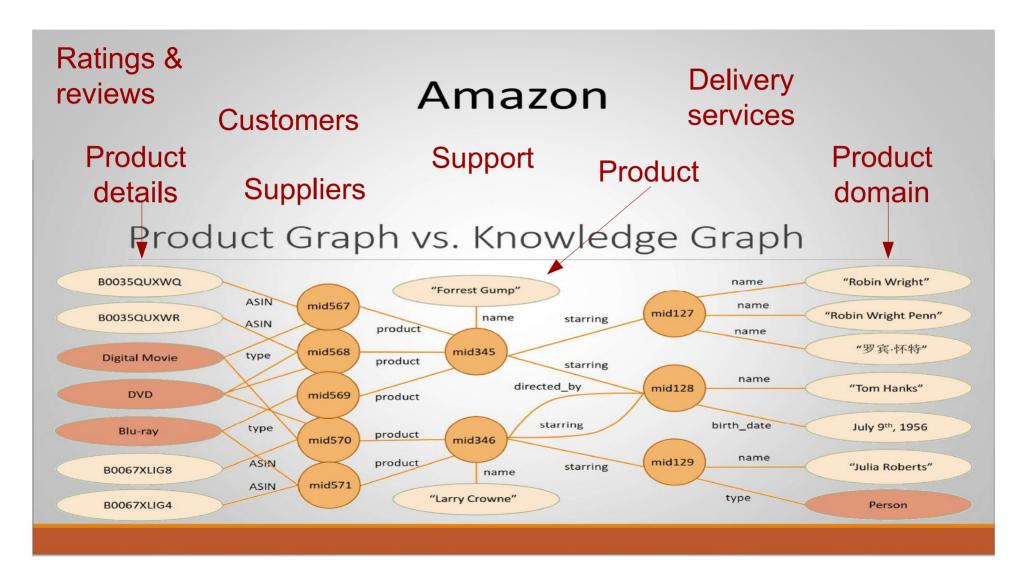


#### Product Graph vs. Knowledge Graph









Frank van Harmelen (2018): Keynote at CAiSE'18

#### Challenges

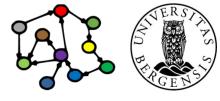
- Ingest product-related information from Amazon's detail pages and from the Internet at large
  - product information is largely unstructured
  - trustworthiness of sources
- Machine learning techniques for
  - knowledge extraction, linkage and cleaning
  - distantly supervised learning
    - train on more structured subset of data
    - run on larger unstructured data space
  - open information extraction
  - graph mining techniques to identify interesting hidden patterns (buying product-X → buying product-Y)



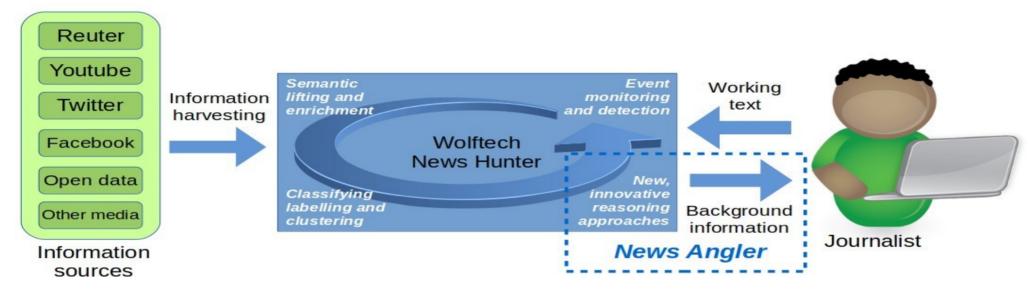
"We aim at building an authoritative knowledge graph for all products in the world"

Xin Luna Dong, Amazon, at WSDM conf. Feb 2018 Architecture Search, QA, Graph **Embedding** Recommen-Graph Querying Generation dation Conversation **Applications** Mining **Amazon Neptune Product Graph** Graph Schema Entity Knowledge Construction Knowledge Mapping Cleaning Cleaning Catalog Ontology Ingestion Knowledge Extraction Extraction Collection

# The News Hunter Platform



#### 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



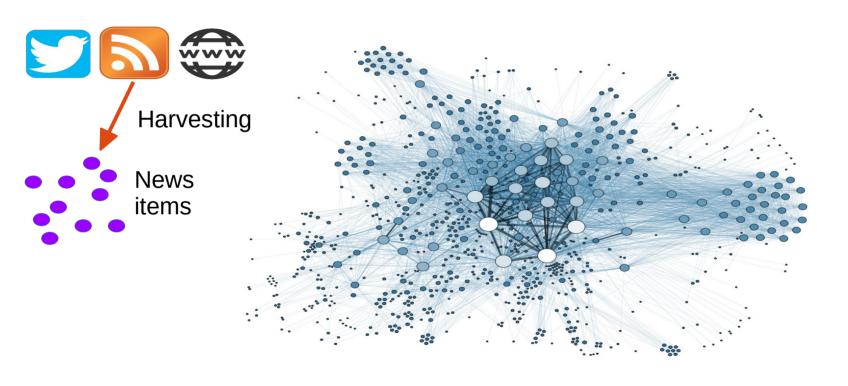




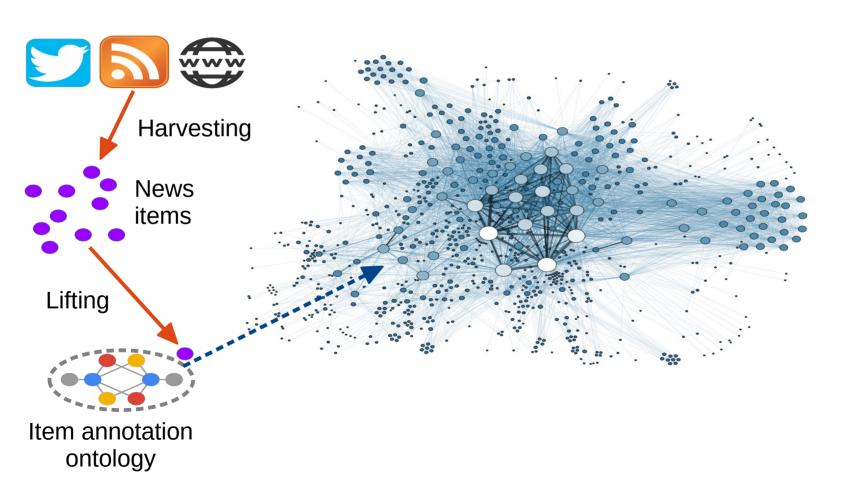




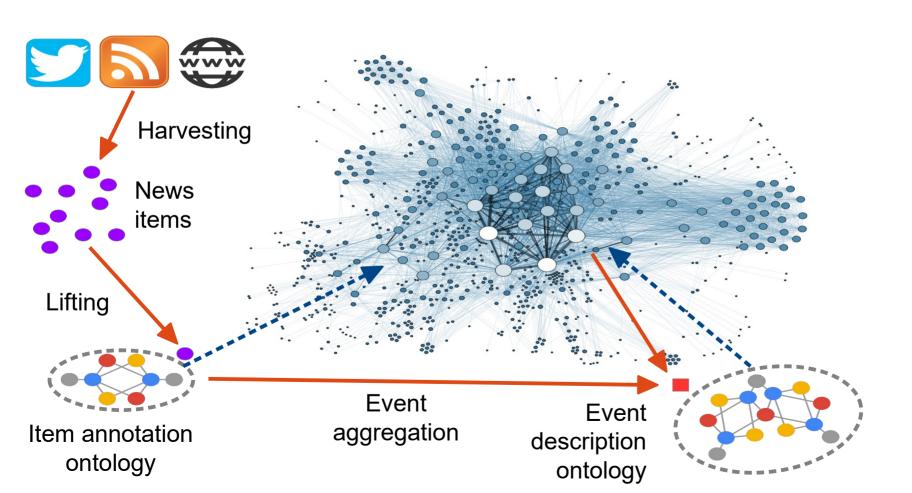




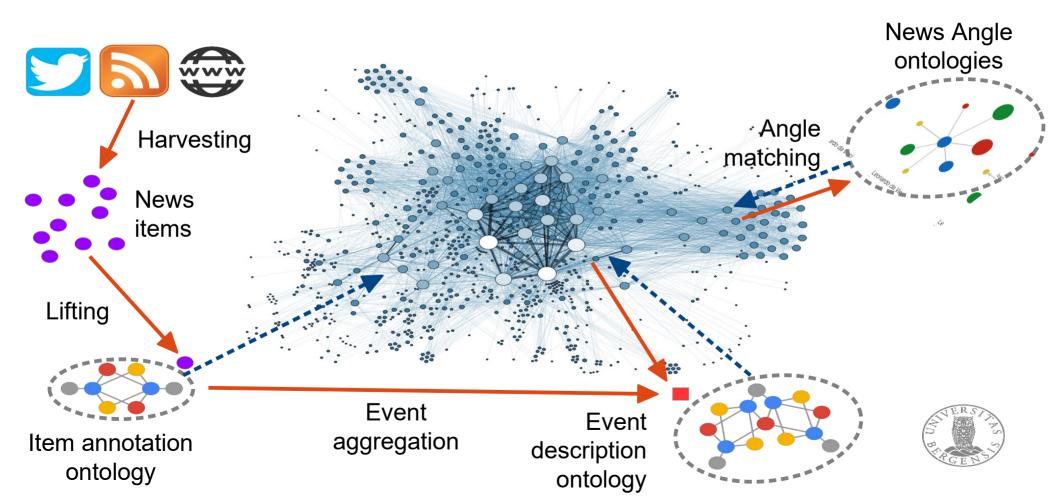










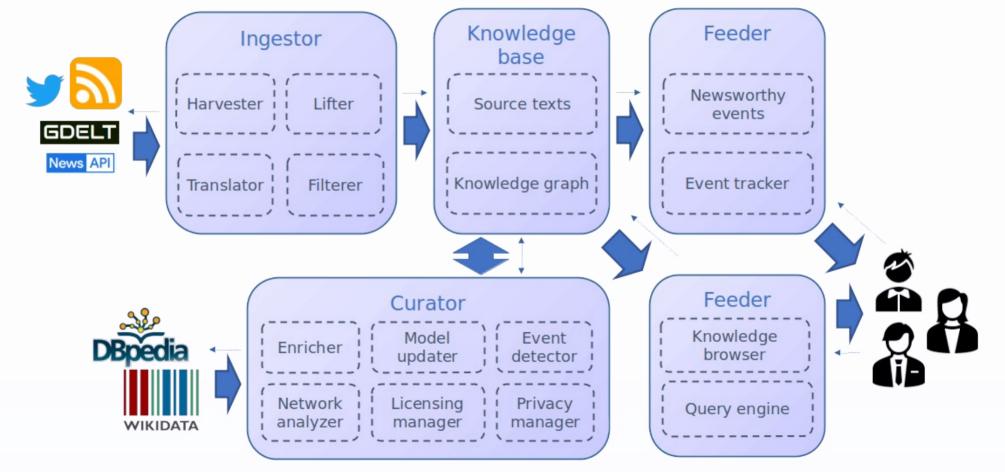




#### The News Hunter architecture

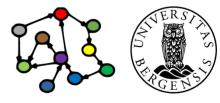
Harvesting news-related information from social media and other sources; analysing, organising, enriching and presenting news-related information to journalists. Implemented state-of-the-art big data and distributed technologies.





#### Services

- Written in Python 3.8-3.9
- All services are deployed in docker containers
- FastAPI as the main python library for writing APIs



#### Services - harvesters

- Twitter harvester: connects to the Twitter API to read streams of tweets from news organizations accounts
- RSS harvester: downloads RSS feeds from news organisations
- GDELT harvester: gets the events and GKG datasets from GDELT projects
- NewsAPI harvester: use NewsAPI.org API to get real-time feeds of news from thousands of news outlets



#### Services - lifters

Lifters for news and GDELT that use NER to represent the information into knowledge graphs

- DbpediaSpotlight NEL: using DBpediaSpotlight for named entity linking
- SpaCy NEL: using SpaCy for named entity linking
- Kolitsas NEL: using Kolitsas algorithm for named entity linking





#### The News Hunter infrastructure

#### Service nodes

Web scraping, API, user interfaces, semantic lifting processes

- · Light-to-medium processing
- Pvthon, REST API. ...

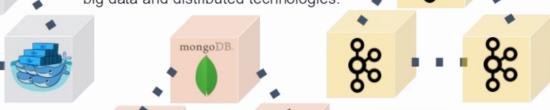


#### Management nodes

Service orchestration and monitoring

- Lighter processing
- Docker Swarm

Harvesting news-related information from social media and other sources; analysing, organising, enriching and presenting news-related information to journalists. Implemented using state-of-the-art big data and distributed technologies.



#### Computationintensive nodes

Complex AI services and training processes.

- · CPU, RAM, GPU intensive
- Python, spaCy, ...

#### Message exchange, aueueina (TBD)

Lighter processing

- Kafka

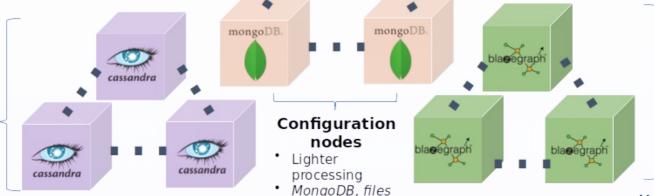
Message

queue nodes

#### Raw data nodes

Distributed storage for raw data files (textual. multimedia)

- Disk intensive
- Cassandra, ...



#### Knowledge graph nodes

News semantic representation storage.

- Disk, CPU and RAM intensive
- Blazegraph

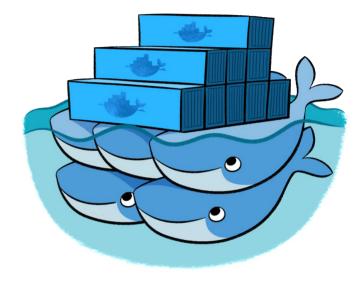
M. Gallofré Ocaña & A.L. Opdahl (2021)

## Cloud infrastructure deployment tools









Slide by Marc Gallofré Ocaña

### **Technologies**

- Docker Swarm
- Kafka (as pub/sub message queue to communicate between all services in the platform)
- Zookeeper
- Cassandra (storing raw data in a distributed cluster)
- Blazegraph (Knowledge graph with news and events representations)
- MongoDB (configuration and metadata)

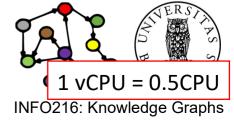
\* All of them have been deployed using Docker contained

#### **News Hunter Platform:**

- 38 vCPUs
- 152GB RAM
- 20TB Disk
- 17 Instances

+

- 1 Launcher instance for deploying the cloud infrastructure:
- 1 vCPU
- 4 GB RAM



## Next week: Rules (RDFS)