

NECINA

2016 AI & Machine Learning Conference

Machine Learning with GraphDB

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Agenda

- Background
- Challenges of Preparing Data for Analytics
- Graph DB – An Effective Alternative for Data Mining
- Use Cases
- Q & A

"According to our estimate, 47 percent of total US employment is in the high risk category, meaning that associated occupations are potentially automatable "

**CARL BENEDIKT FREY AND
MICHAEL A. OSBORNE**, Oxford
Martin School & Faculty of
Philosophy, UK

Image: Oxford Martin School & Faculty of
Philosophy

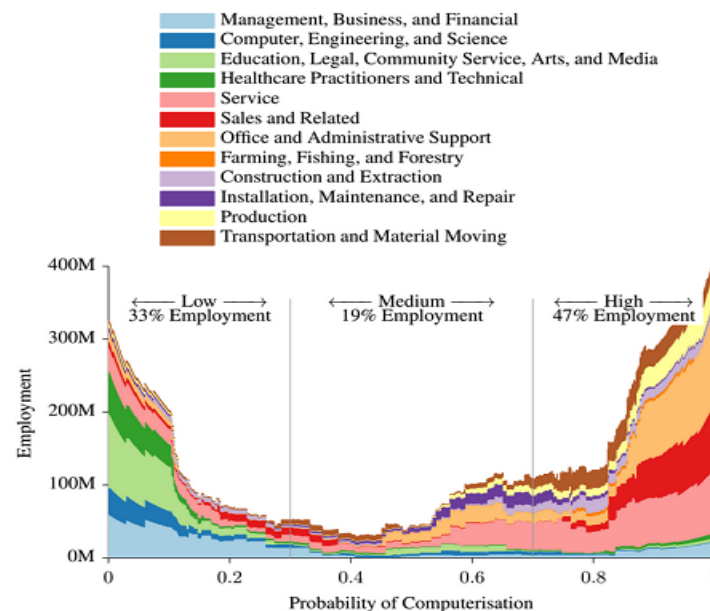


Figure 1. Employment Affected by Computerisation.

The Washington Post:- Brian Fung 10/13/2016 China has now eclipsed USA in AI research

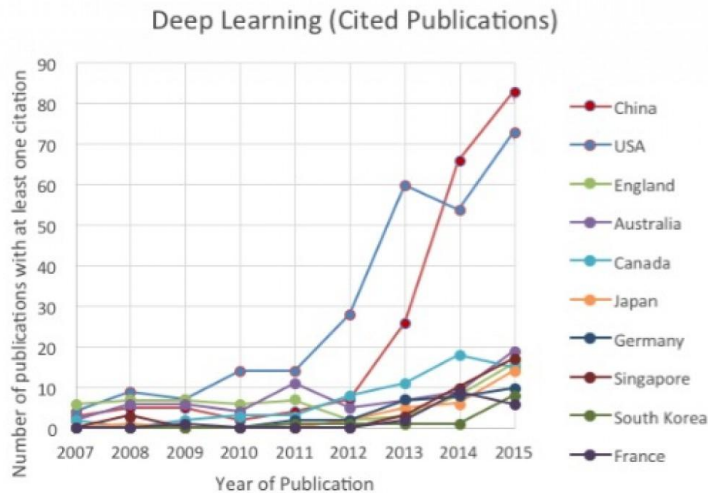


Figure 2: Journal articles cited at least once, mentioning "deep learning" or "deep neural network", by nation.⁶³

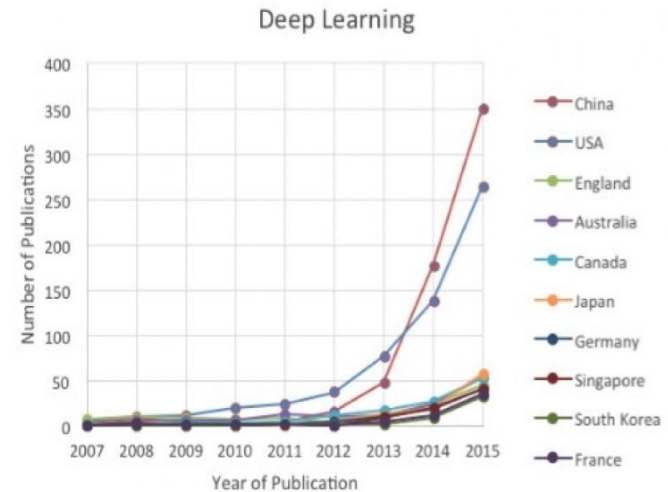


Figure 1: Journal articles mentioning "deep learning" or "deep neural network", by nation.⁶²

Researches indicated that **80%** of effort were spent on preparing data with the remaining 20% on modeling

Key Challenges

1. Poor data quality
2. Don' t understand the data deep enough
3. Factors like data variety and velocity prolong times in data preparation
4. Dealing with huge data sets
5. Don' t know what business problems to solve

Why GraphDB ? Some Quotes:-

“Machine Learning is really the umbrella and graph technology is a way of representing data when using machine learning.” - *Claus Jepsen, Chief Architect, R&D of Unit4*

“Graph technology can be considered a type or technique of machine learning, or, at a minimum, aspects of graph technology have strong application to machine learning.” - *David Thompson, Sr. Director of Production Management of LightCyber*

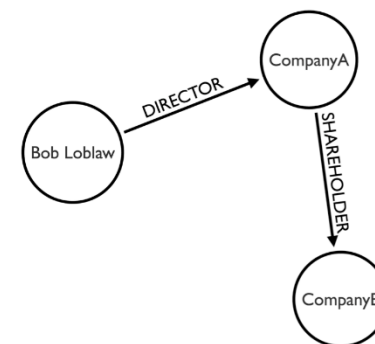
“Automatically assembling a graph of connectivity between data points is a powerful addition to learning.” - *Simon Crosby, co-founder & CTO of Bromium*

“Machine learning algorithms help data scientists discover meaning in data sets, and these insights can be expressed as relationships between nodes in a graph. Graph databases enable efficient storage and traversal of information about relationships. Therefore, graph data can either be the input or the output of machine learning processing.” - *Jim Webber, Chief Scientist of Neo Technology*

The Panama Papers - 11 Million records in over one terabyte data.

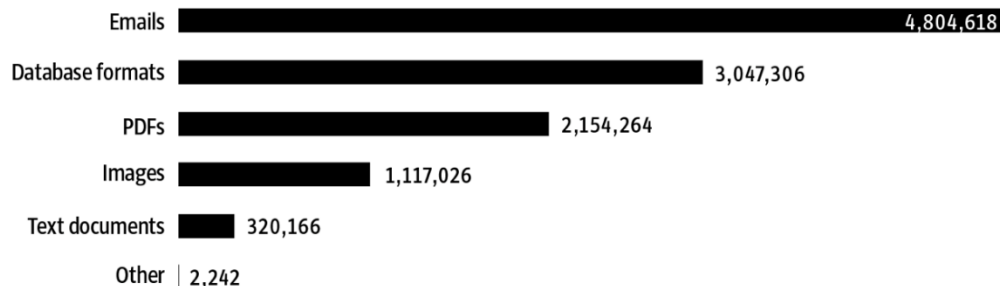
Steps:

1. Acquire documents
2. Classify documents
 - a) Scan / OCR
 - b) Extract document metadata
3. Whiteboard domain
 - a) Determine entities and their relationships
 - b) Determine potential entity and relationship properties
 - c) Determine sources for those entities and their properties
4. Work out analyzers, rules, parsers and named entity recognition for documents
5. Parse and store document metadata and document and entity relationships
 - a) Parse by author, named entities, dates, sources and classification
6. Infer entity relationships
7. Compute similarities, transitive cover and triangles
8. Analyze data using graph queries and visualizations

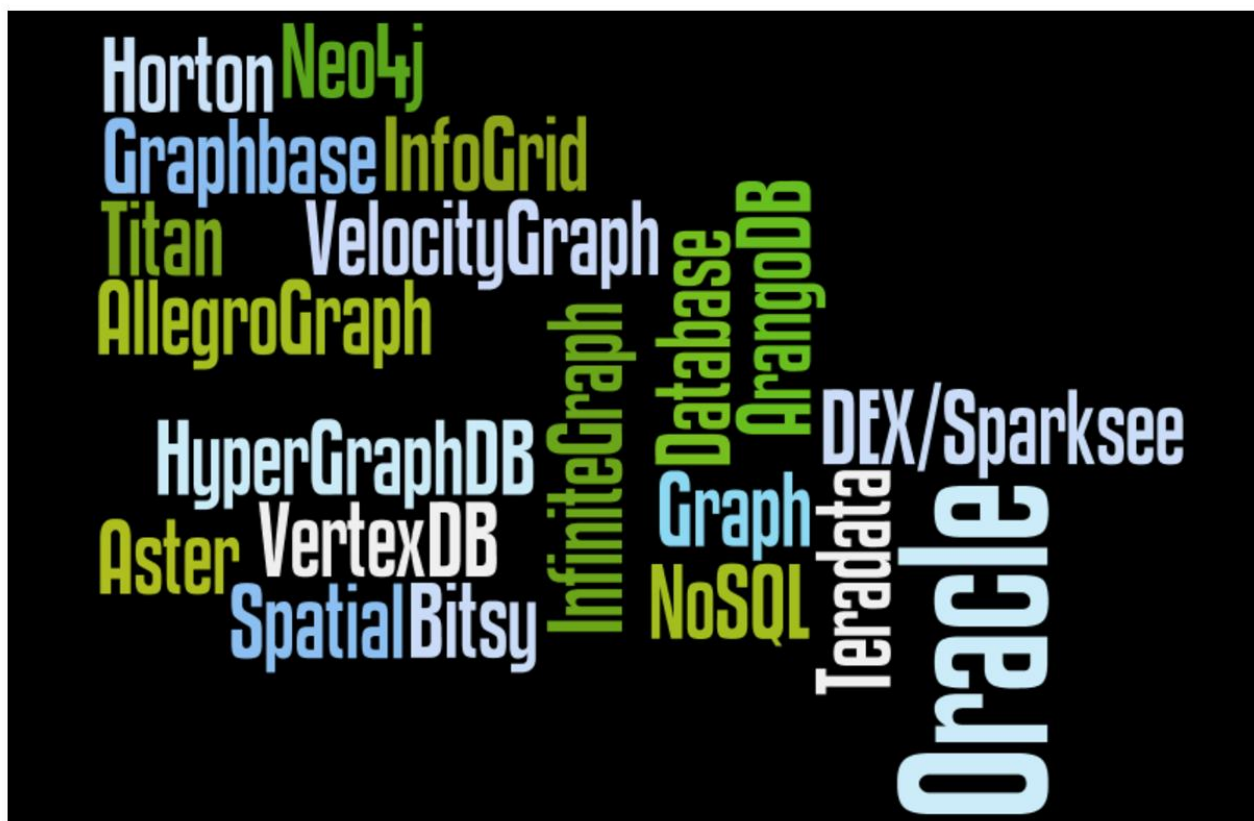


The structure of the leak

The 11,5 millionen contain the following file types



TOP 31 GRAPH DATABASES



Source: <http://www.predictiveanalyticstoday.com/top-graph-databases/>

Code Signature (Vector)

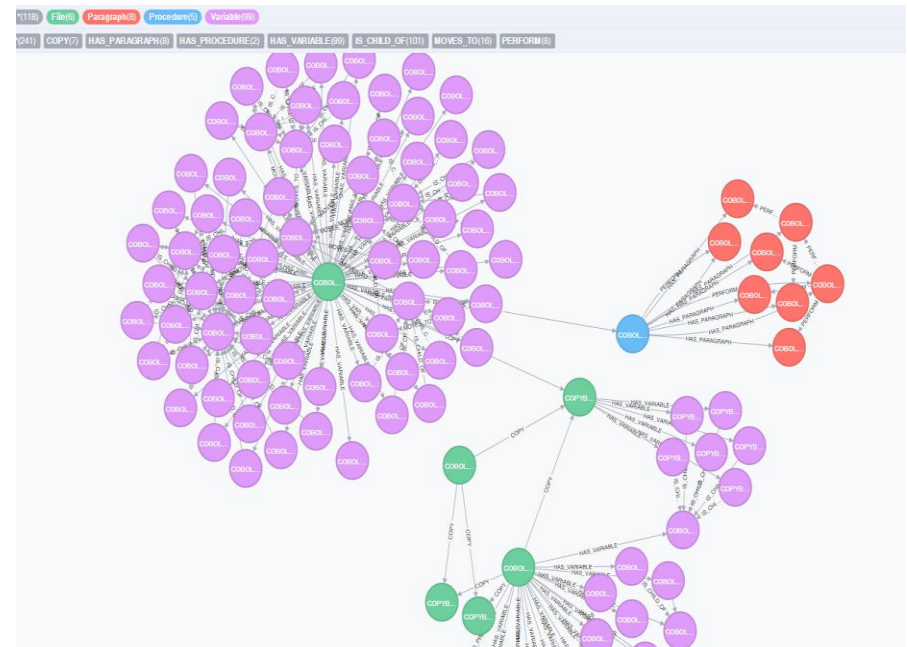
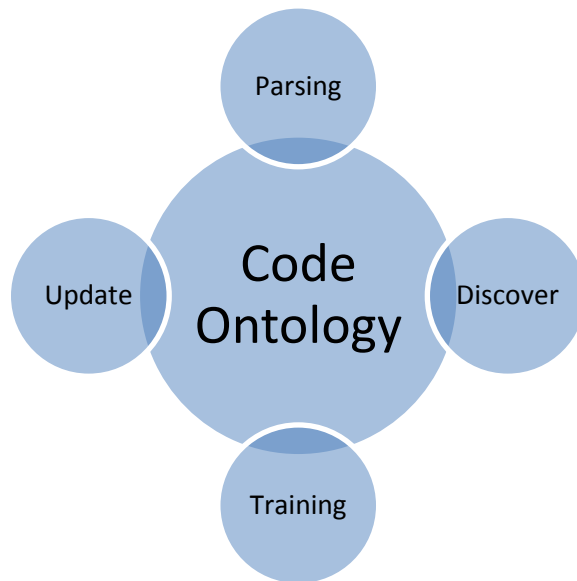
Proc: Proc A -> Proc B -> Proc C

BizDef: Private Bank Account Opening

SysDef: PBAO Main

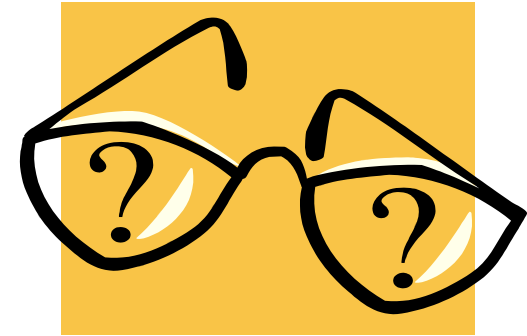
Loop: Proc A -> Loop A-> Condition (VarA < VarB)-> PassedValue (VarA)

Dataflow: VarA ->Proc A (???) -> Proc B (???); VarB -> ProcC(???) -> ProcA(???)



Pros	Cons
<ul style="list-style-type: none">• Highly applicable for solving relationship-driven business problems, like fraud detection, money laundering, consumer behavior etc.• Strong visualization capability provides better insight to the data• Finding relationships amongst data is the easiest way to qualify data sets• Persist results in GraphDB provides high flexibility and extensibility of the ontological view• Most products support sophisticated SQL- liked scripting languages• Linkage to other data types with semantic RDF	<ul style="list-style-type: none">• Not all business cases are applicable to use GraphDB (e.g. image recognition)• May take time and effort to design and discover the data relationships• Learning curve with a new programming language• Major machine / deep learning libraries are not natively working with GraphDB yet• Better integration to the Big Data infrastructure is expected

- <http://www.idgconnect.com/abstract/18124/the-wave-disruption-graph-machine-learning>, posted by [Kathryn Cave](#) on July 12, 2016
- <http://www.idgconnect.com/abstract/15609/database-helped-reporters-follow-panama-papers-money>
- <http://info.salford-systems.com/blog/bid/305673/9-Data-Mining-Challenges-From-Data-Scientists-Like-You>
- <http://www.cs.uvm.edu/~icdm/10Problems/10Problems-06.pdf>
- http://www.kdd.org/exploration_files/V13-02-09-Munson.pdf



Questions/Comments?

