

Big Graph Processing Systems

Organisation and Motivation

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This presentation is an adaption of slides from Angela Bonifati



- ▶ Motivation
- ▶ Part I: Graph Query Paradigms and their Semantics
- ▶ Part II: Property Graphs
 - ▶ Chapter 1: An introduction to a concrete query language
 - ▶ Chapter 2: PG Schema Languages and PG Constraints
 - ▶ Chapter 3: Schema Discovery and Property Graph Transformations
- ▶ Part III: RDF and Sparql

Tentative Schedule

January	06/01	CM + CM	Part I
	13/01	CM + TP	Part II
	20/01	CM + TP	Part II
	27/01	CM + TP	Part II
February	03/02	TP + TP	Part II
	10/02	TP + TP	Part II
	17/02	TP + TP	Part II
	24/02	CM + CM	Part III
	25/02	CM + TP	Part III
March	03/03	CM (Exam)	

- ▶ 15 CM + 15 TP
- ▶ Please ask (and answer) questions

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- ▶ 15 CM + 15 TP
- ▶ Please ask (and answer) questions
- ▶ Grading
 - ▶ 50% Practical lab (TP)
 - ▶ 50% Exam (CM)

Hands-on Part – Preparations

- ▶ The warm-up project consists of analysing a dataset containing genomic information
- ▶ Please install [Neo4j](#) and import the dataset before the first hands-on lesson

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1. Install Neo4j

- ▶ Make sure you are using [version 5.26](#) or newer
- ▶ Make sure you have [enough memory](#), at least 20GB
- ▶ A simple option is to install Neo4j Desktop by following the official instructions:
<https://neo4j.com/docs/desktop-manual/current/installation/>

2. Download and import the database

- ▶ Download the database dump file from
<https://partage.liris.cnrs.fr/index.php/s/LoEtp24fk38P6n5>
- ▶ Import the dump file by following the official instructions:
<https://neo4j.com/docs/desktop-manual/current/operations/create-from-dump/>

Everything is Data

[<https://www.acronis.com/en-us/blog/posts/data-everything-8-noble-truths/>]



A New Era of Science

New Realities

- ▶ Everything is data
- ▶ Rise of data-driven culture
- ▶ High-performant data analytics
- ▶ Exploit sophisticated statistical methods

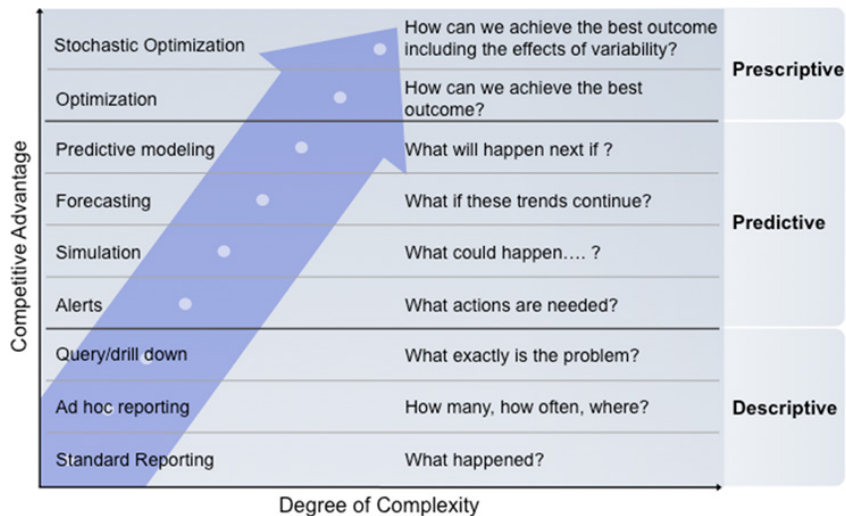
The quest for knowledge used to begin with grand theories. Now it begins with massive amounts of data.

Welcome to the Peta**Zetta**byte Age.

How do we structure/implement/live with this trend?



Level of Analytics



Based on: Competing on Analytics, Davenport and Harris, 2007

Properties of Entities

- ▶ Captured/measured values
- ▶ What are the sales figures/temperatures/etc.?
- ▶ Multidimensional data/time series/matrices



Connections Between Entities

- ▶ Network structure
- ▶ What do the friends of your customers buy?
- ▶ Graph data



Connections Manifest in Many Different Ways, Facets, Values, Scopes, and Scales

Cause: interaction, proximity, information, relationship, affiliation, ownership, allegiance, force, repulsion, ...

Nature: social, cultural, economical, physical, chemical, ...

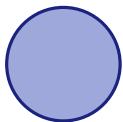
Quality: existential, essential, inconsequential, or indiscernible; persistent or temporary; positive, neutral, or negative

Scale: from interacting cells, such as neurons in our brain, to high-scale entities, such as families, groups, clubs, companies, organizations, states, nations, ... or stars and galaxies

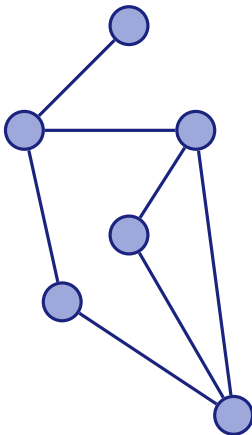
(graphs) – [: ARE] -> (everywhere)

Graph Building Blocks

Nodes (Dots, Circles)



- ▶ Like an entity in conceptual models
- ▶ Exist on their own
- ▶ Have an object identity



Edges (Lines)



- ▶ Like a relationship in conceptual models
- ▶ Exist only between nodes
- ▶ Identity depends on the nodes they connect

Graphs as Unifying Abstractions

- ▶ Graphs are **natural abstractions** for representing interconnected objects when encoding, explaining and predicting real-world and digital-world phenomena.
- ▶ Graphs are underpinning several **data management ecosystems**, in societal, scientific, RDF, product and digital domains.
- ▶ There is **no unique killer application** for graphs, but several exist.
- ▶ Nevertheless, the **data models, query languages and system requirements** needed for graphs are constantly evolving.

The Ubiquity of Large Graphs and Surprising Challenges of Graph Processing: Extended Survey

Siddhartha Sahu · Amine Mhedhbi · Semih Salihoglu · Jimmy Lin · M. Tamer Özsu

CONTRIBUTED ARTICLES

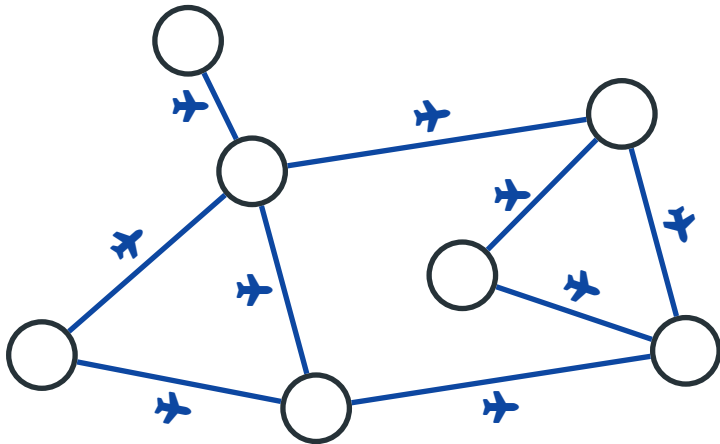
The Future Is Big Graphs: A Community View on Graph Processing Systems

By Sherif Sakr, Angela Bonifati, Hannes Voigt, Alexandru Iosup, Khaled Ammar, Renzo Angles, Walid Aref, Marcelo Arenas, Maciej Besta, Peter A. Boncz, Khuzaima Daudjee, Emanuele Della Valle, Stefania Dumbrava, Olaf Hartig, Bernhard Haslhofer, Tim Hegeman, Jan Hidders, Katja Hose, Adriana Iamnitchi, Vasiliki Kalavri, Hugo Kapp, Wim Martens, M. Tamer Özsu, Eric Peukert, Stefan Plantikow, Mohamed Ragab, Matei R. Ripeanu, Semih Salihoglu, Christian Schulz, Petra Selmer, Juan F. Sequeda, Joshua Shinavier
Communications of the ACM, September 2021, Vol. 64 No. 9, Pages 62-71
10.1145/3434642

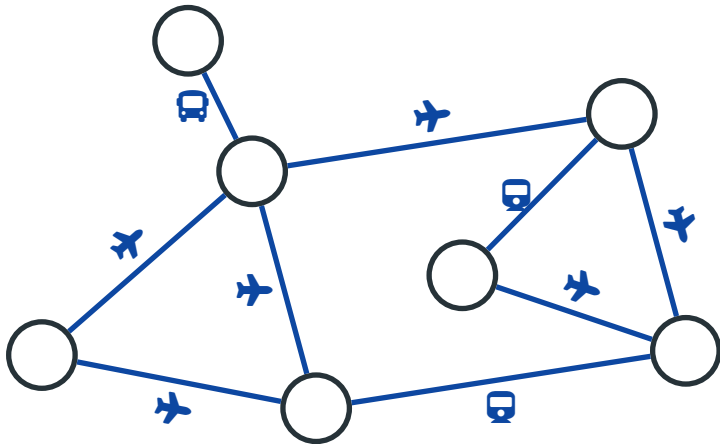
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ARTICLE CONTENTS:
Introduction
Key Insights
Abstractions
Ecosystems
Performance

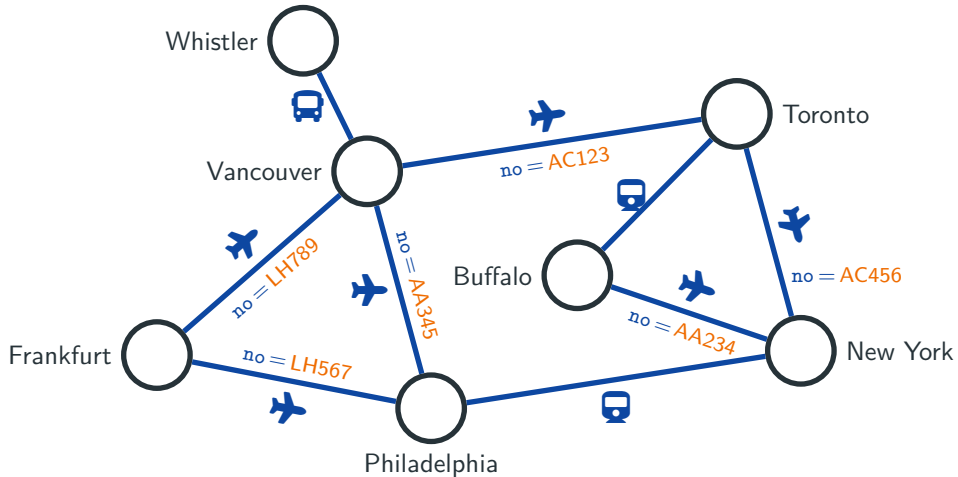
Graph Data – Transportation Network



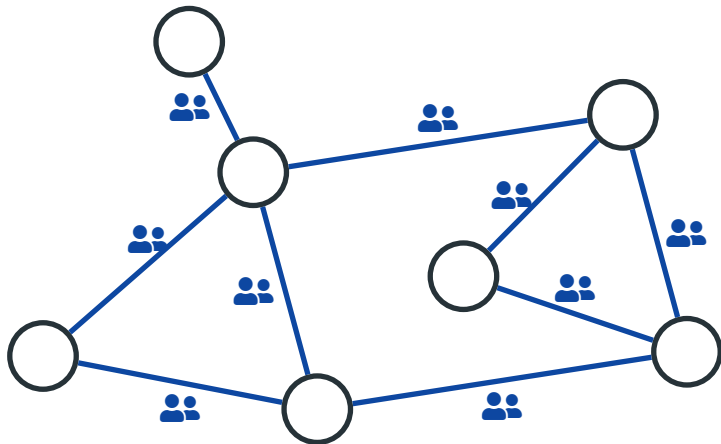
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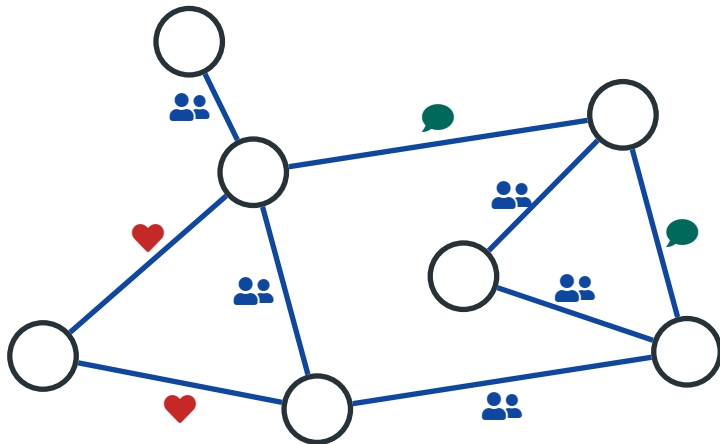
Graph Data – Transportation Network



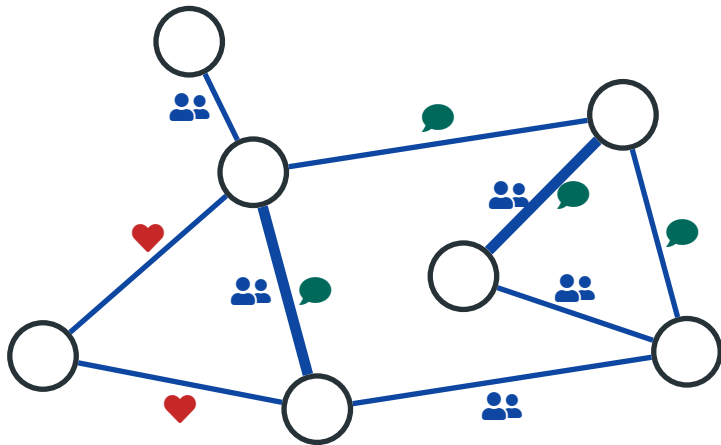
Graph Data – Social Network



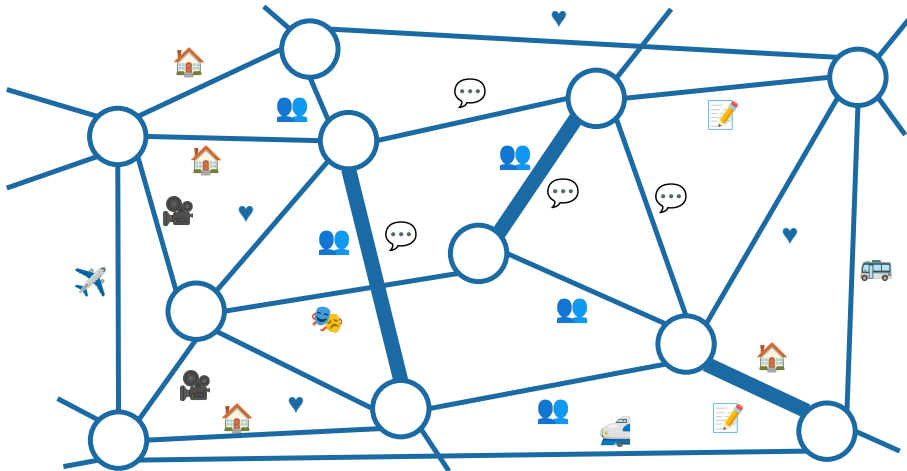
Graph Data – Social Network



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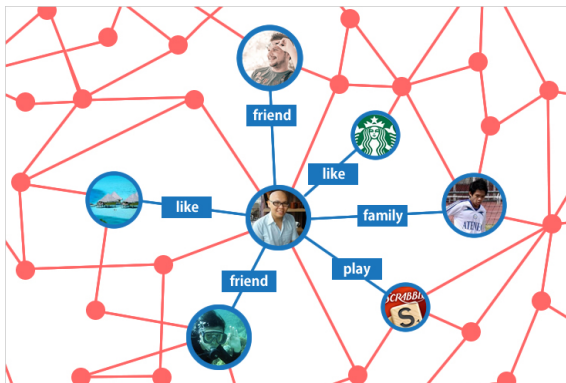
Graph Data – Heterogeneity



Structured Search

Example (Facebook Graph Search)

- ▶ Finding subgraph structures
- ▶ Very natural way of formulating queries

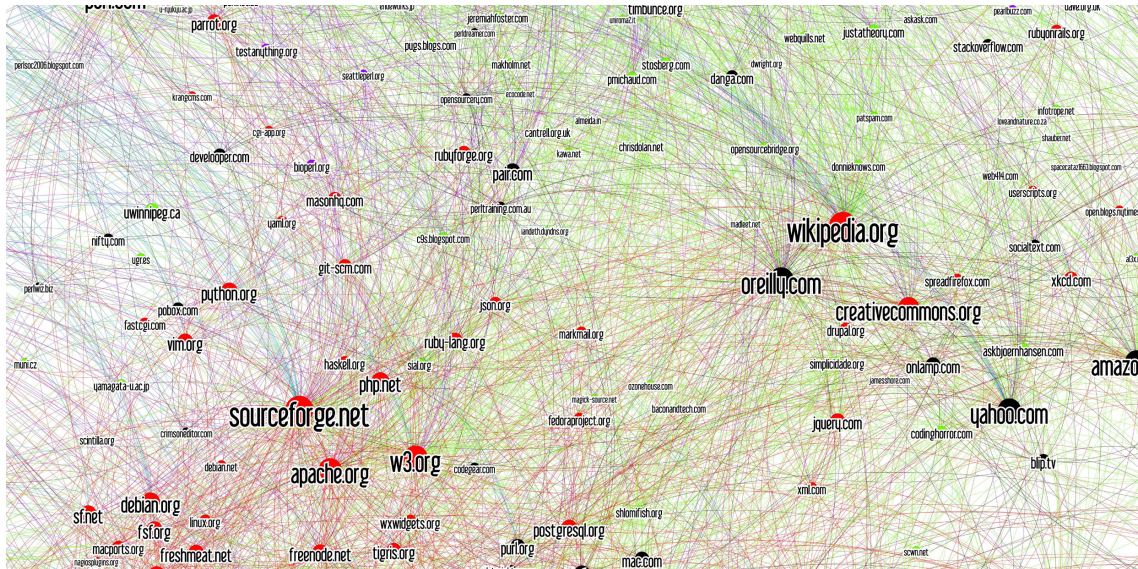


Search bar: People who like Michael Jackson

- People who like **Michael Jackson** (Musician/Band · 55,105,238 like this)
- People who like **Michael Jackson** (Interest)
- People who like **Michael Jackson's This Is It**
- People who like **Michael Jackson The Experience**
- People who like **Michael Jackson Legend Never Die**
- People who like **Michael Jackson** and live in **Pune, Maharashtra**

SEE MORE

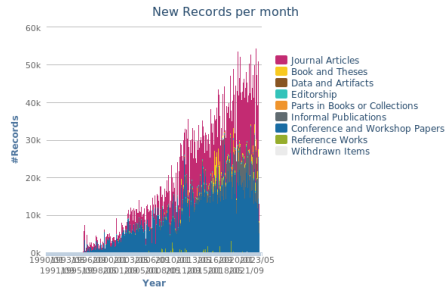
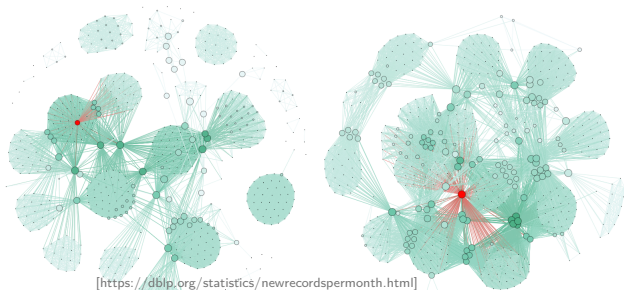
Web Graph



Bibliography

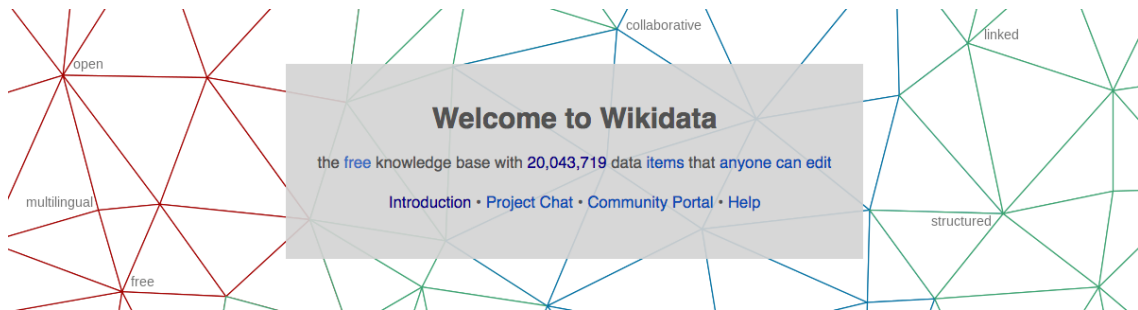
Example (DBLP)

- ▶ Open bibliographic information on major computer science journals and proceedings
- ▶ >6.5 million publication
- ▶ >46000 new publication per month
- ▶ >1.7 million authors



Example (Graph to capture world-knowledge)

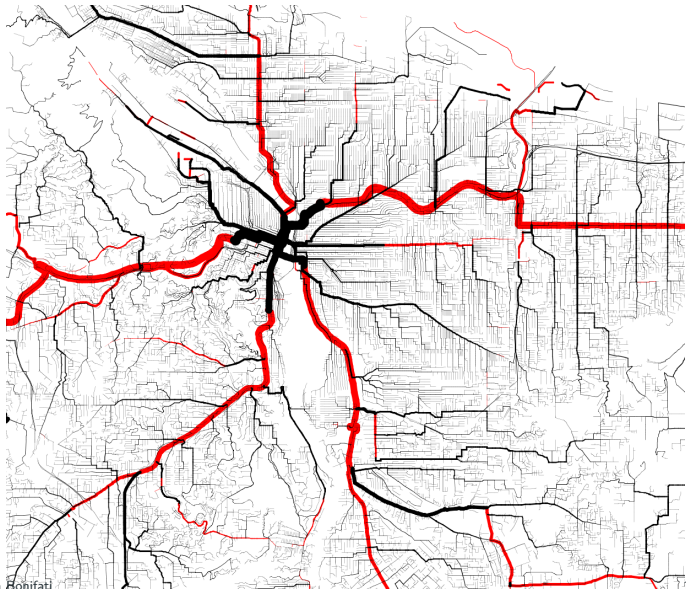
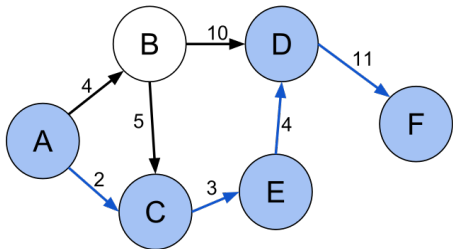
- ▶ Open knowledge base that can be read and edited by humans and machines
- ▶ Structured data of Wikipedia, Wikivoyage, Wikisource, etc.



Shortest Paths

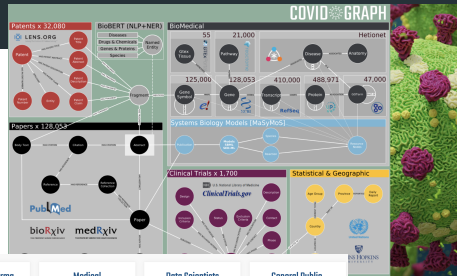
Example (Transportation Network)

- ▶ Only hop-distances (#edges)
 - ▶ in friendship network
- ▶ With weights (distance, travel time, etc.)
 - ▶ in road networks, transportation connections



A Plethora of Applications

- ▶ Among which, the covidgraph.org initiative aiming at building the Covid19 knowledge graph
 - ▶ Collecting patents, publications about the human coronaviruses
 - ▶ Biomedical data (genomics and omics)
 - ▶ Experimental data about clinical trials
 - ▶ Key demographic indicators
- ▶ Practical use case in many data-oriented tasks
 - ▶ Property graph schema discovery
 - ▶ Threshold queries in Theory and in the Wild



Threshold graph queries on the Covid19 graph

- Find each country that does not have three reports for some age group

```
MATCH (c:Country)
      -[e:CURRENT_FEMALE | CURRENT_MALE | CURRENT_TOTAL]->
      (a:AgeGroup)
WITH c, a, COUNT(type (e)) AS ecount
WHERE ecount < 3 RETURN c, a
```

Adapted from

[Bon22a] Angela Bonifati, Stefania Dumbrava, George Fletcher, Jan Hidders, Matthias Hofer, Wim Martens, Filip Murlak, Joshua Shinavier, Slawek Staworko, Dominik Tomaszuk: Threshold Queries in Theory and in the Wild. Proc. VLDB Endow. 15(5): 1105-1118 (2022)

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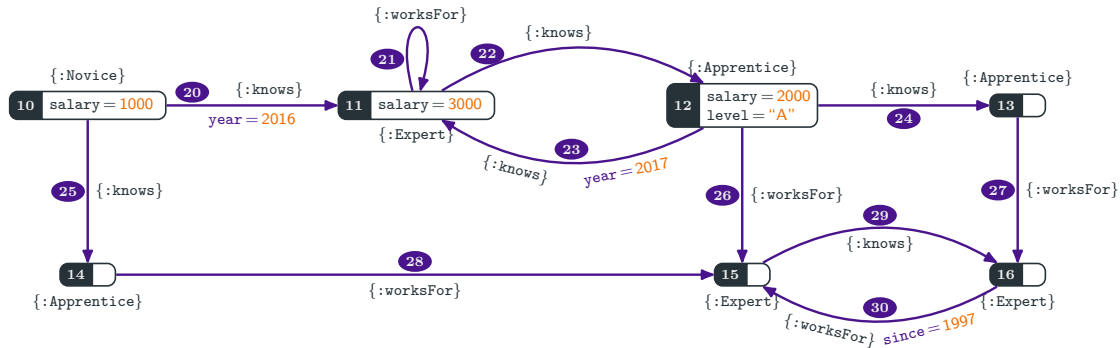
- Find each protein that has more than 43917 associated geneontology terms

```
MATCH (p:Protein )-[:MAPS]->*( )-[:HAS_ASSOCIATION]->( )
      (( )-[:IS_A]->*( ) | ( )-[:PART_OF]->*( ))(t:GOTerm)
WITH p, COUNT (DISTINCT t) AS count_go
WHERE count_go > 43917 RETURN p
```

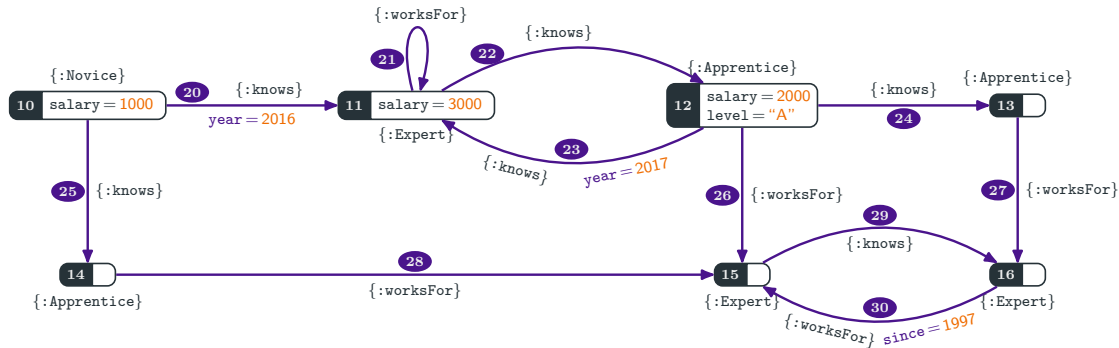
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Property Graphs – Example



Property Graphs – Example



Nodes and edges have

- ▶ IDs: 10, 11, ...
- ▶ labels, e.g., `:Novice`, `:Apprentice`, `:worksFor`, ...
- ▶ properties, e.g. `"salary = 3000"`, `"year = 2016"`, ...

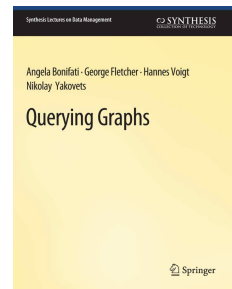
Property Graphs: A formal definition

Assume pairwise disjoint sets of
 \mathcal{O} (objects), \mathcal{L} (labels), \mathcal{K} (property keys), and \mathcal{N} (values)

Definition

A **property graph** is a structure $(V, E, \eta, \lambda, \vartheta)$ where

- ▶ $V \subseteq \mathcal{O}$ is a finite set of vertices,
- ▶ $E \subseteq \mathcal{O}$ is a finite set of edges,
- ▶ $\eta: E \rightarrow V \times V$ assigns an ordered pair of vertices to each edge,
- ▶ $\lambda: V \cup E \rightarrow \mathcal{P}(\mathcal{L})$ assigns a finite set of labels to each vertex and edge,
- ▶ $\vartheta: (V \cup E) \times \mathcal{K} \rightarrow \mathcal{N}$ assigns values for properties to vertices and edges.



A Lattice of Graph Data Models

- ▶ A data model per use case
- ▶ How expressive and human-friendly is a data model?
- ▶ Need of making different data models interoperable via mappings or direct translations

