JURION IPG USE-CASE RE-ENGINEERING A COMPLEX RELATIONAL DATABASE APPLICATION

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IPG USE CASE INTRODUCTION

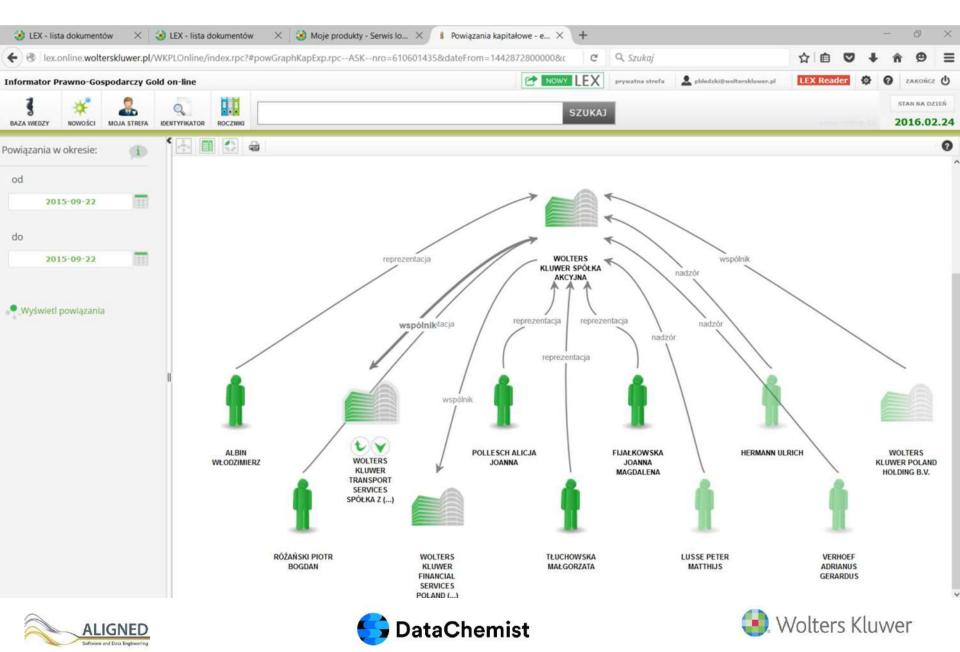






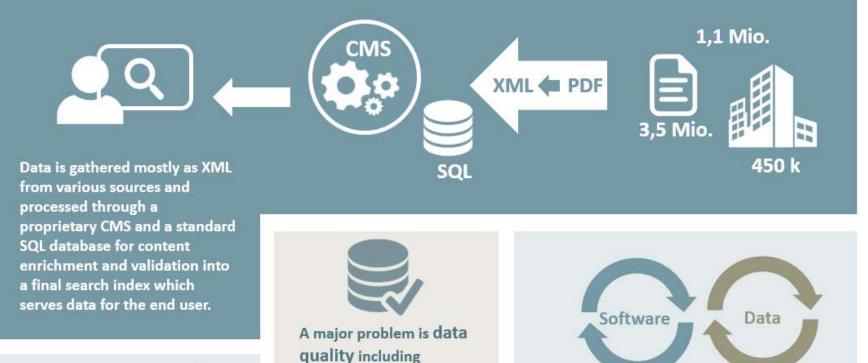
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Legal-Commercial Information System (IPG Gold) product - graph view



IPG Problem Statement

IPG – a Commercial Intelligence System by Wolters Kluwer Poland with ...



Data is also expected to be enriched with data originating from new sources including publicly available repositories and third party datasets.

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corrected by using

missing or incoherent data as well as semantic

inconsistencies which

could be detected and

The software development lifecycle ist mostly autonomous from the data lifecycle. Both of them not changing very often, but are expected to change in the next few years as major upgrades are planned.





Data Complexity







Spatial data and administrative division data (2,5k counties)

Legacy DB model in Polish

Complex Schema

50 types of Companies

20 types of relations between companies & people

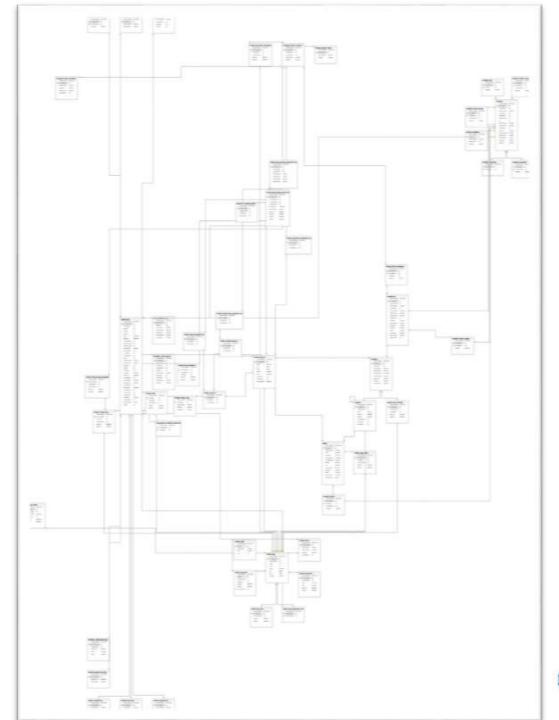
70 types of events/ documents related to companies or people

30 Types of roles









Data Complexity example





32 Unsolvable Scenarios

Basic Datatype Errors

9, 25, 26, 27, 28

Invalid email address

Missing Mandatory Properties

1, 3, 4, 5, 6, 7, 16, 18, 19, 20, 23, 24

No trustee in bankruptcy

Temporal Constraints

10, 11, 12, 13, 14, 15, 17, 21

Same receiver and trustee

Inconsistent Data

2, 8, 22, 30

Multiple shareholders in sole shareholder company

Data Model Complexity

29

Relationship model in main table is incomprehensible

Temporal Queries

31

Find relationship at any time between any 2 entities

Recursive Queries

32

Subsidiarity Loop: Company A owns B, owns C, owns A.





IPG – DATA CHEMIST SOLUTION







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DataChemist

Closed World OWL Reasoning Engine Fast ACID in-memory datastore with integrated logic engine

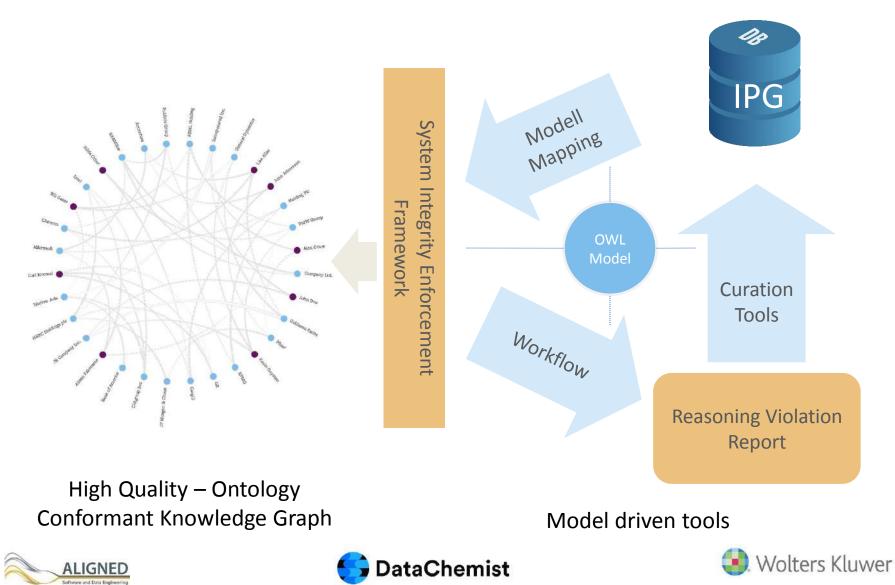
WOQL – model aware query language



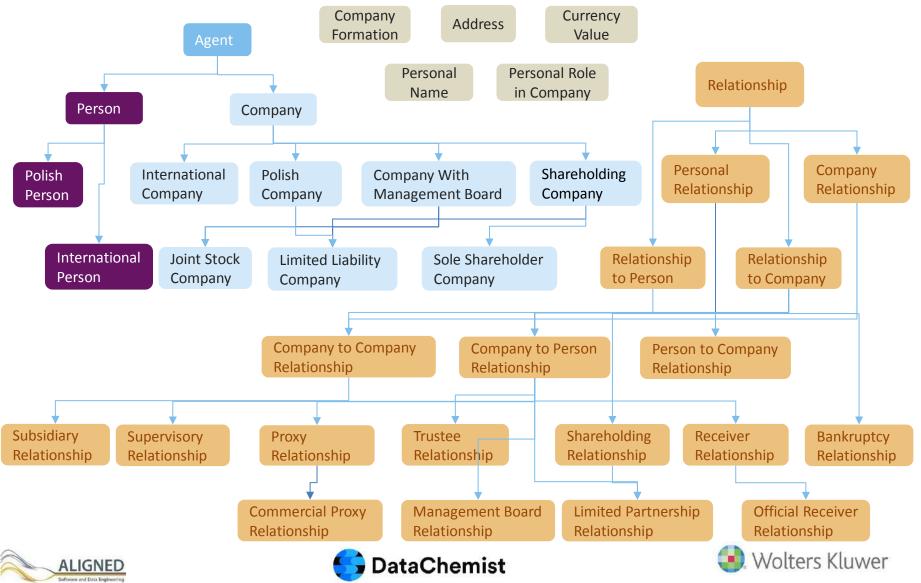




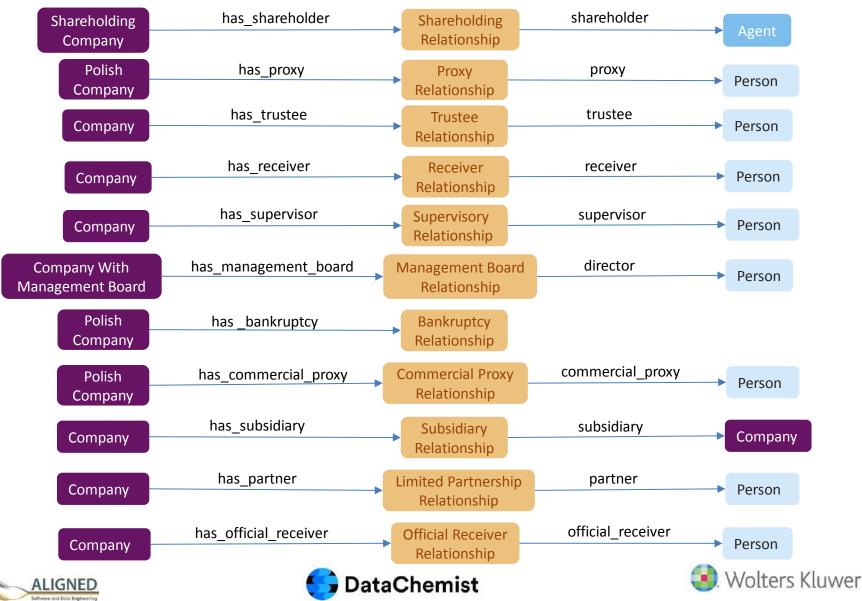
How it works



Semantic Model – 36 classes



Semantic Model – 21 relationship properties



Semantic Model – 36 simple properties

Co	mpany	Comp
company_name	String	formation_metl formation_circu
Polish	Company	
annual_report formation krs region nip	string Company Formation integer integer integer	email website postal_street postal_number postal_locality postal_code
	Agent	Personal
address	Address	personal_role_r
Р	erson	Cui
personal_name given_name	Personal Name Personal Name	currency_value currency_unit
Dolis	sh Person	Sharehol
pesel	pesel	wholly_owned liability

Company Formati	on	Trustee Relatio	onship
formation_method formation_circumstances	string string	legal_basis appointment_date	string dateTime
Address		Management Board	Relationship
website u postal_street s postal_number	email url string string	board_type management_board_role management_role_susper	
/	string string	Limited Partne	ership
		Relationsh	ір
Personal Role in Com	ipany	liability	decimal
personal_role_name	string		
		Bankruptcy Relat	tionship
Currency Value		announcement	string
urrency_value	float	termination	string
currency_unit	string	method repeal	string string

Proxy Relationship	
proxy_info	string
proxy_type	string







decimal, string

string

string

Shareholding Relationship

ShareholdingCompany Currency Value

number of shares

Ontology Editing & Visualisation

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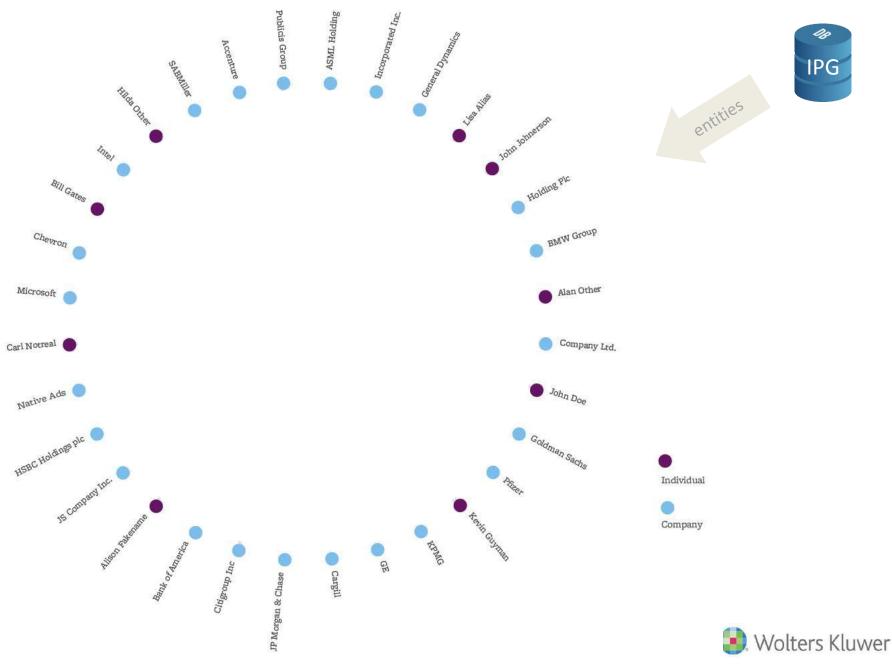
KNOWLEDGE GRAPH CONSTRUCTION

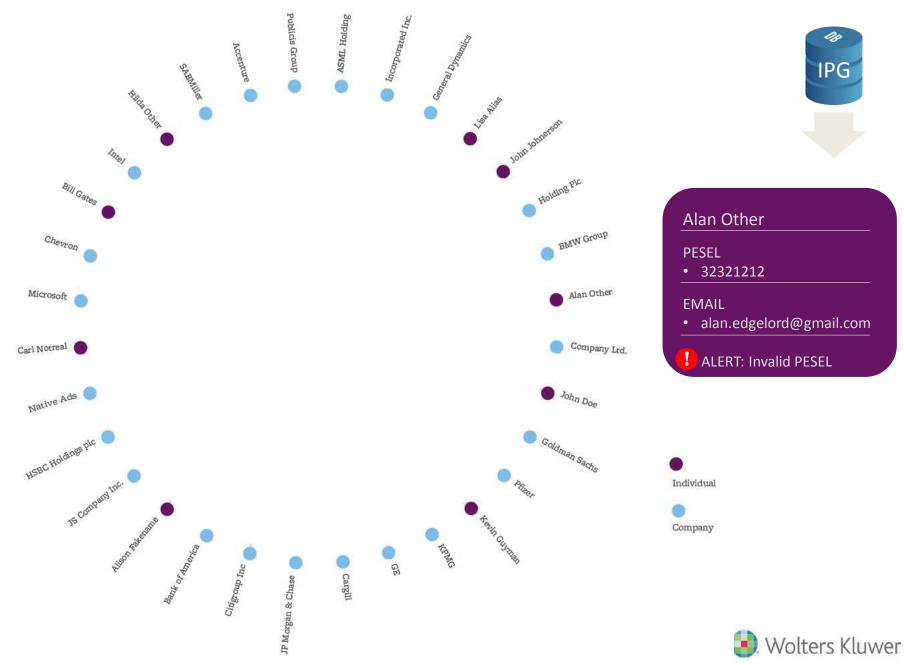


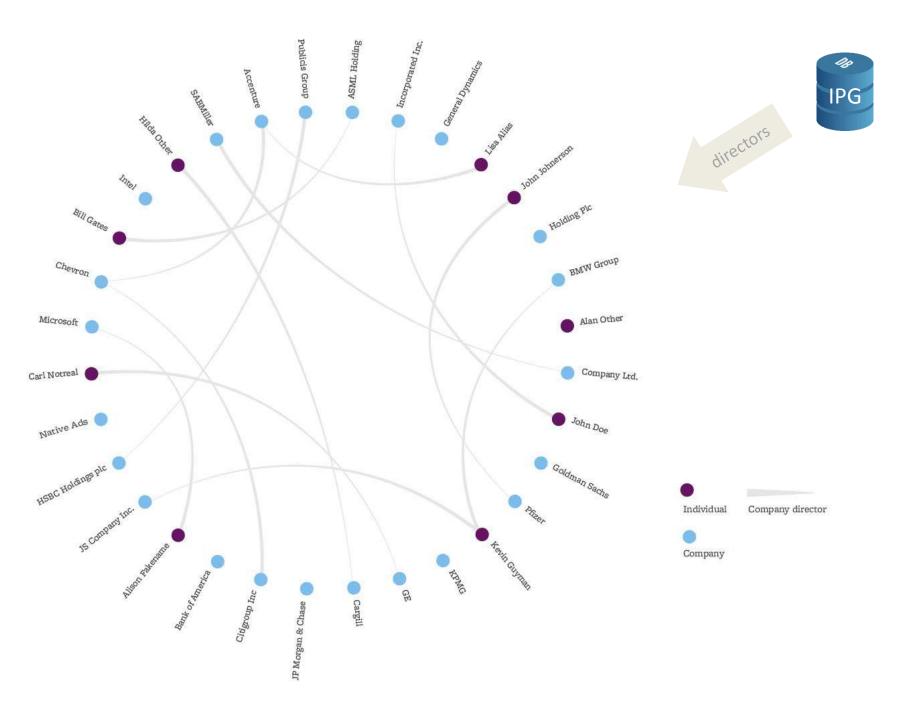


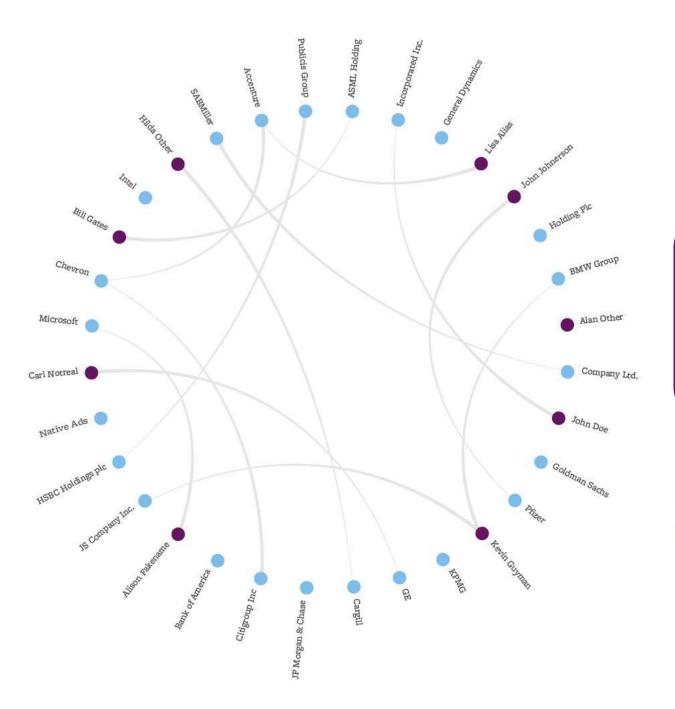


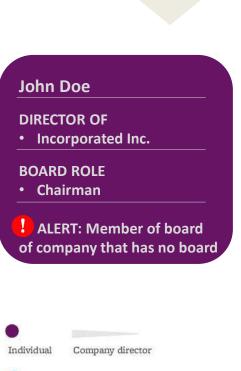
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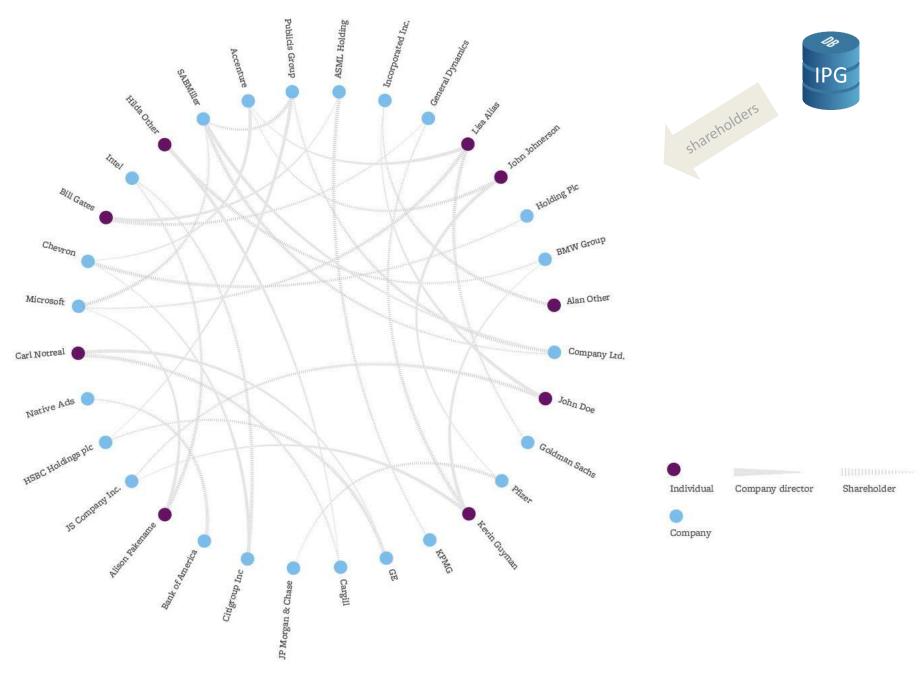


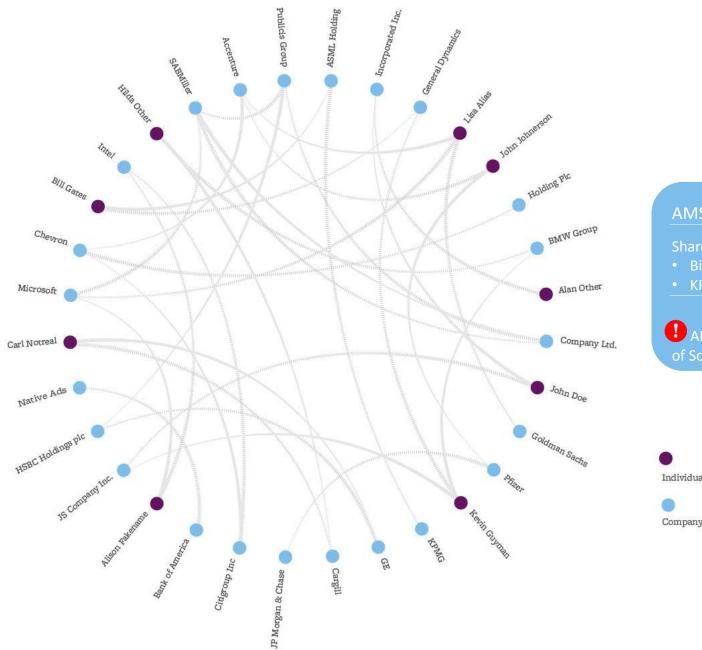


Company

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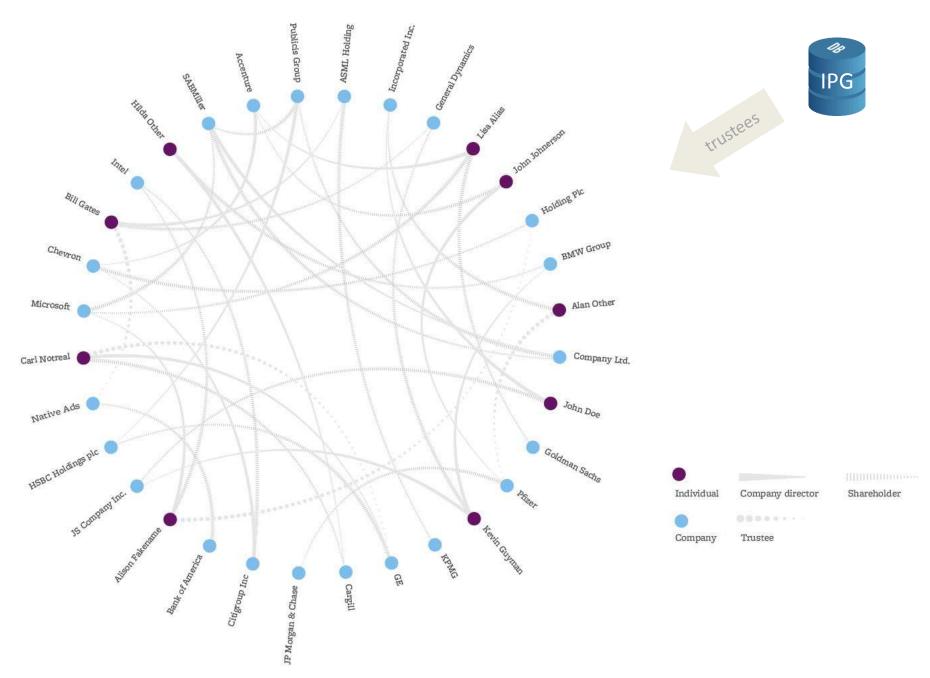
IPG

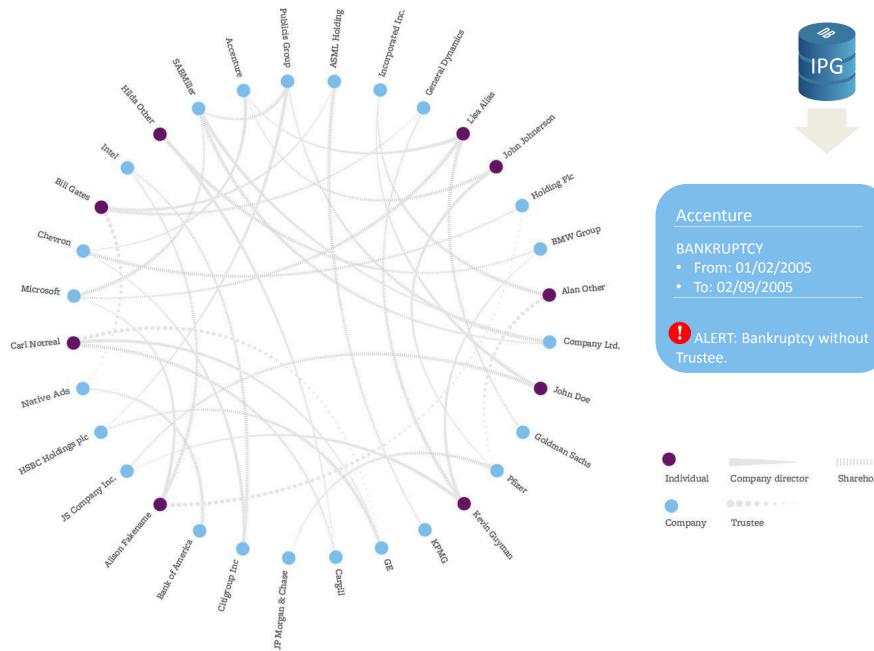






IPG



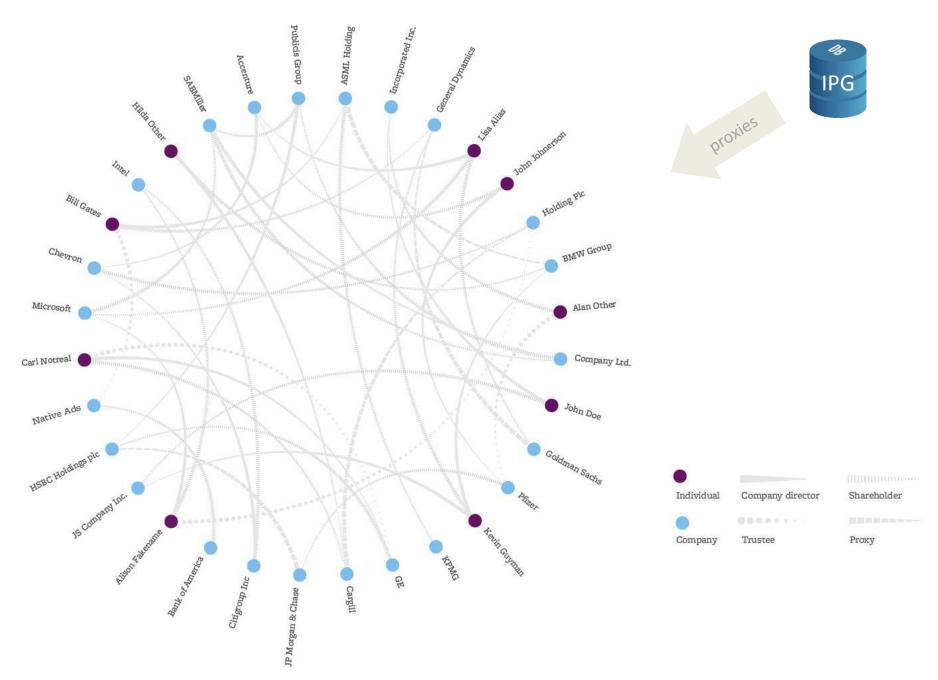




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IPG

Shareholder





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SHARE • GE	HOLDINGS	1
	RT: commercial it proxy type	proxy
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Individual	Company director	Shareholder
	0.	

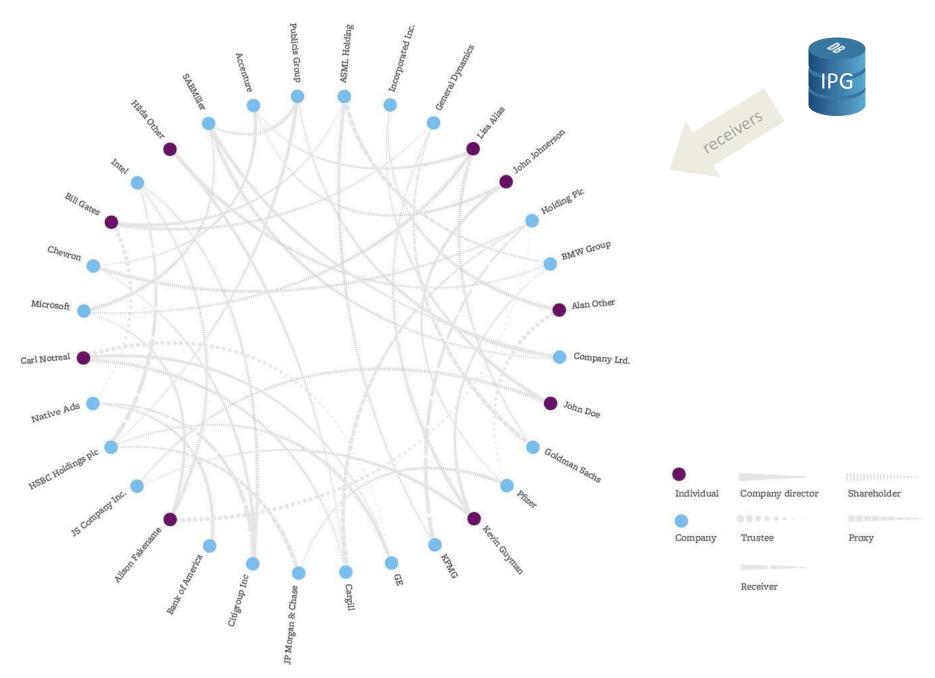
Company

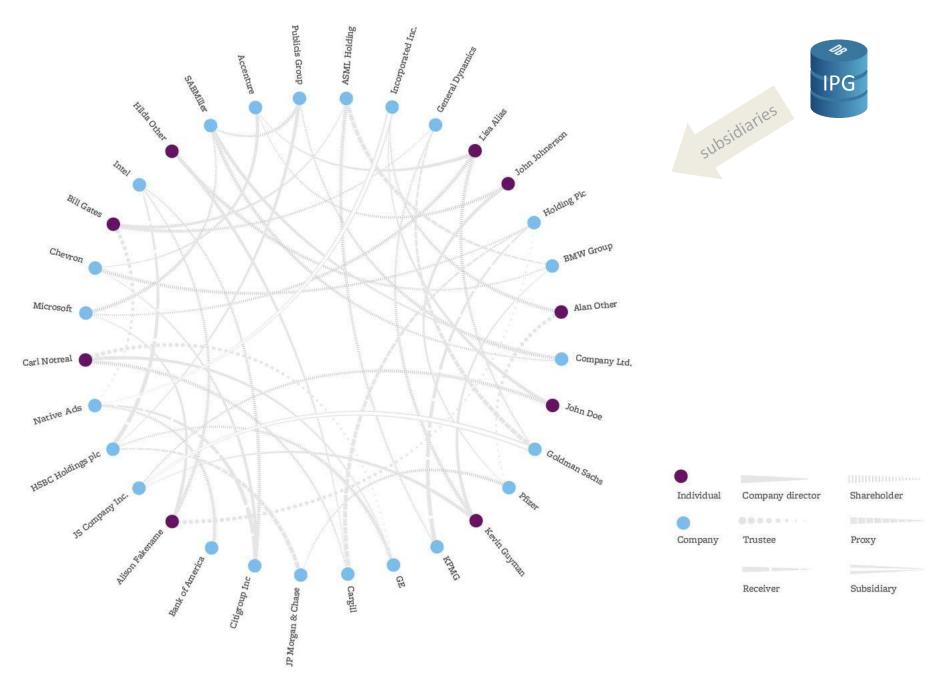
Trustee

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IPG

Proxy





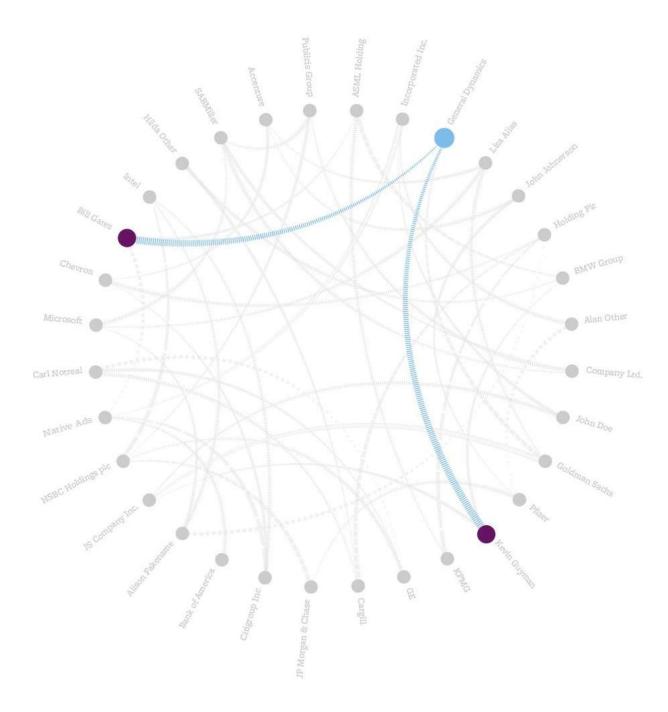
KNOWLEDGE GRAPH QUERYING







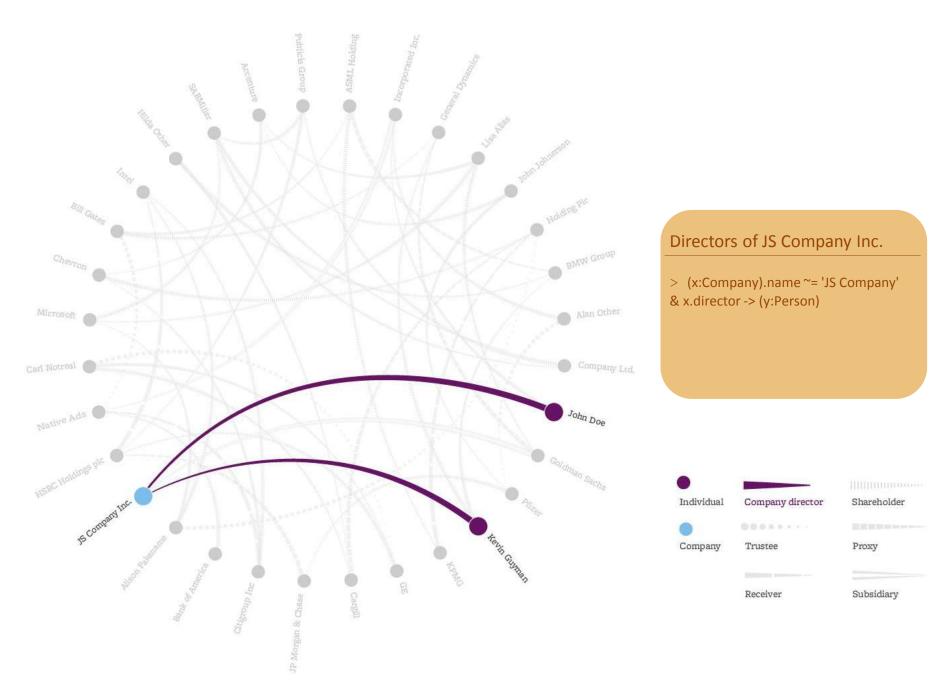
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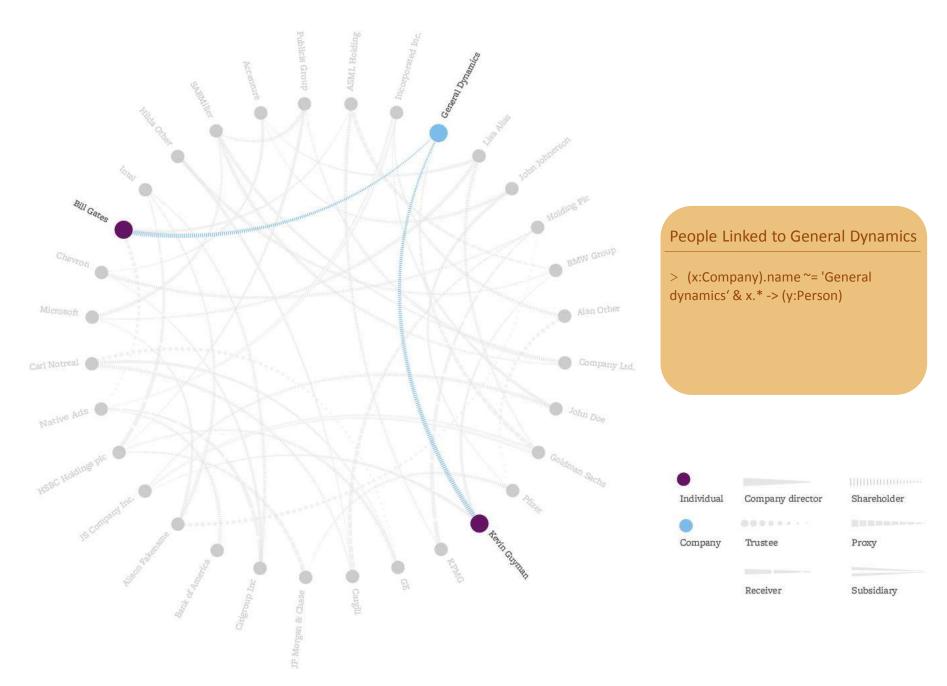


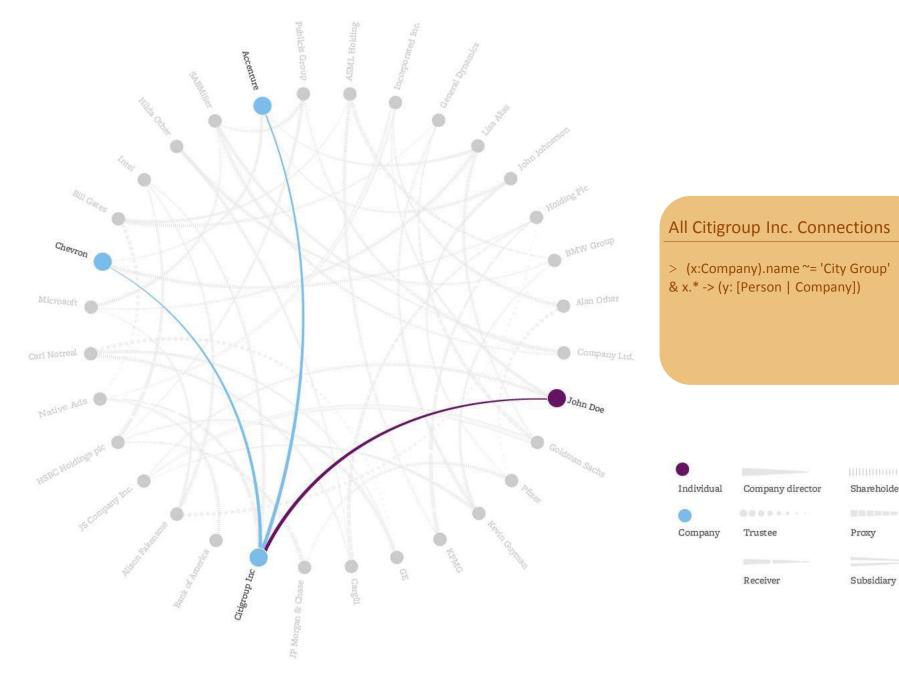
General Dynamics Shareholders

> (x:Company).name ~= 'General dynamics' & x.shareholder -> (y:Person)



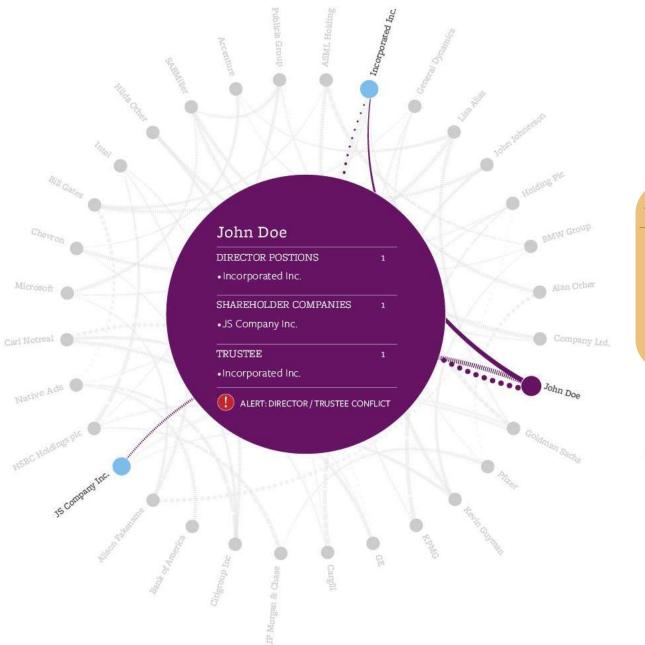






Shareholder

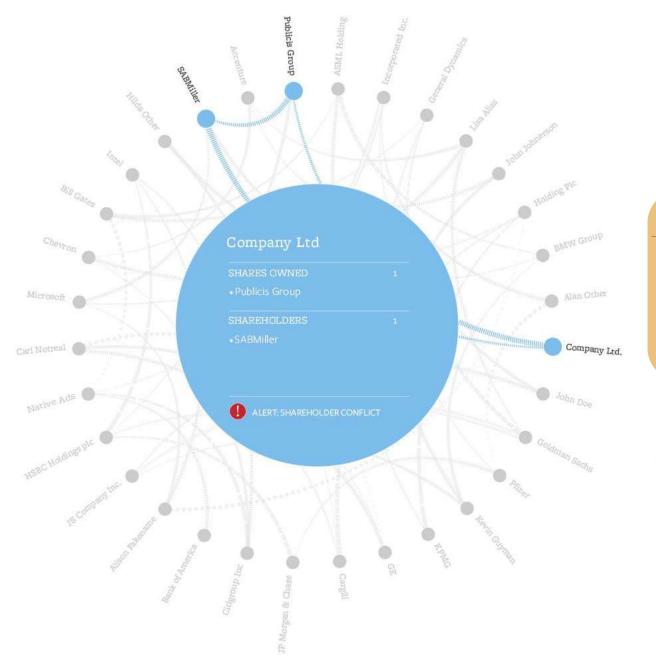
Proxy



Temporal Constraints

> (x:Person) = (y:Company).director &
(x:Person) = y.trustee &
(_.drector.lifespan) >< (_.trustee.lifespan)</pre>

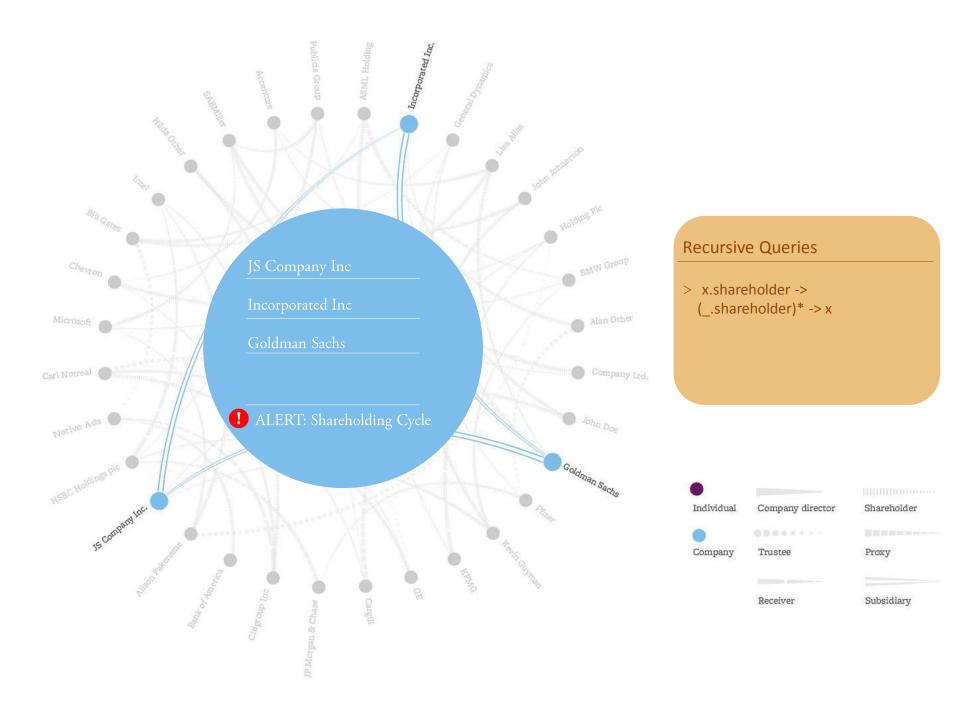




Cross Shareholding Restrictions

> (x:Company).name ~= 'Publicis Group'
 & (y:Company).name ~= 'SABMiller'
 & x.shareholder -> (z:Company)
 & y.shareholder -> z





Live Demonstration







RESULTS







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Solving the Unsolvables

Error Type	Scenarios	Solved	Partially	Unsolved	Errors Detected
Basic Datatype	5	5	0	0	8,500
Missing Mandatory Properties	12	10	2	0	10,032
Temporal Constraints	8	4	4	0	12,320
Inconsistent Data	4	2	1	1	1,000
Temporal Queries	1	1	0	0	5,324
Recursive Queries	1	1	0	0	909
Model Complexity	1	0	1	0	NA
Total	32	23	8	1	~40,000







Other Findings

- The IPG Use-Case was added in the second half of the project. The work described here began in July 2017
- IPG is a very large dataset: 100 million SQL rows. This translated into 2 billion triples with provenance information included. We had to handle files > 100GB
 - Dealing with the scale of the data was by far the largest challenge every piece of our tool-chain had to be rebuilt to deal with the size and speed requirements. Even ed breaks at that scale.
- IPG has a schema that has evolved over >15 years in response to immediate business needs.
 - The second largest challenge was deciphering the schema.
- A very large number of errors were found beyond the 32 unsolvables many referential integrity violations, duplicates, inconsistent dates, typos....
 - The third largest challenge was parsing inconsistent formats used for the same field
- The first complete demonstration of running queries over the entire 2 billiontriple dataset was delivered on 5/3/2018
 - work is ongoing to complete the partial solutions.
- We estimate that our solution is 1-2 orders of magnitude faster and cheaper than existing methods. With the scaling work, we required 10 person months; without, 3 person months.







Significance to Wolters Kluwer

- Creating domain specific knowledge models that drive new business and applications are at the core of our global WK corporate strategy (LegalTech, FinTech, Health, etc.)
- These applications are all over the place
- We need to semantify our data and we do not have the resources to build everything from scratch again
- This approach addresses several major challenges that we have to solve

Knowledge graphs as a necessary ingredient in AI applications are now at the core of interest for companies. WKD can tell from its own industry, but also SWC from their customer side.







Questions?

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