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# Big Data Integration

Philippe Cudré-Mauroux

[eXascale Infolab](#), University of Fribourg  
Switzerland



**eXascale Infolab**

CSNDSP, Budapest  
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# Instant Quiz

- n-Vs of Big Data?
- Yarn?
- Spark
- Knowledge Graph?



# eXascale Infolab (XI)

- New lab @ U. of Fribourg–Switzerland
- **Big Data/AI infrastructures** for social / semantic / scientific data

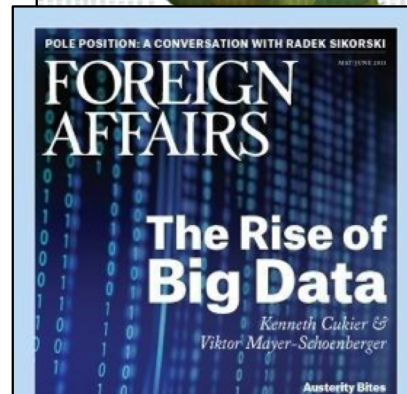


<https://exascale.info/>



# Exascale Data Deluge

- Web companies
  - Google
  - Ebay
  - Yahoo
- Science
  - Biology
  - Astronomy
  - Remote Sensing
- Financial services,  
retail companies  
governments, etc.



New data formats  
New machines  
Peta & exa-scale datasets  
Obsolescence of traditional  
information infrastructures

# Data is the new Oil

- Data + Algorithms → Actionable Insight → \$\$

Big Data /  
Data Science

Machine Learning /  
“Dumb” A.I.

Model  
(Prediction /  
Classification)

Optimized  
Services



## Big data can generate significant financial value across sectors



### US health care

- \$300 billion value per year
- ~0.7 percent annual productivity growth



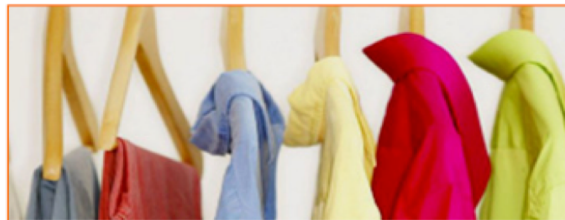
### Europe public sector administration

- €250 billion value per year
- ~0.5 percent annual productivity growth



### Global personal location data

- \$100 billion+ revenue for service providers
- Up to \$700 billion value to end users



### US retail

- 60+% increase in net margin possible
- 0.5–1.0 percent annual productivity growth



### Manufacturing

- Up to 50 percent decrease in product development, assembly costs
- Up to 7 percent reduction in working capital

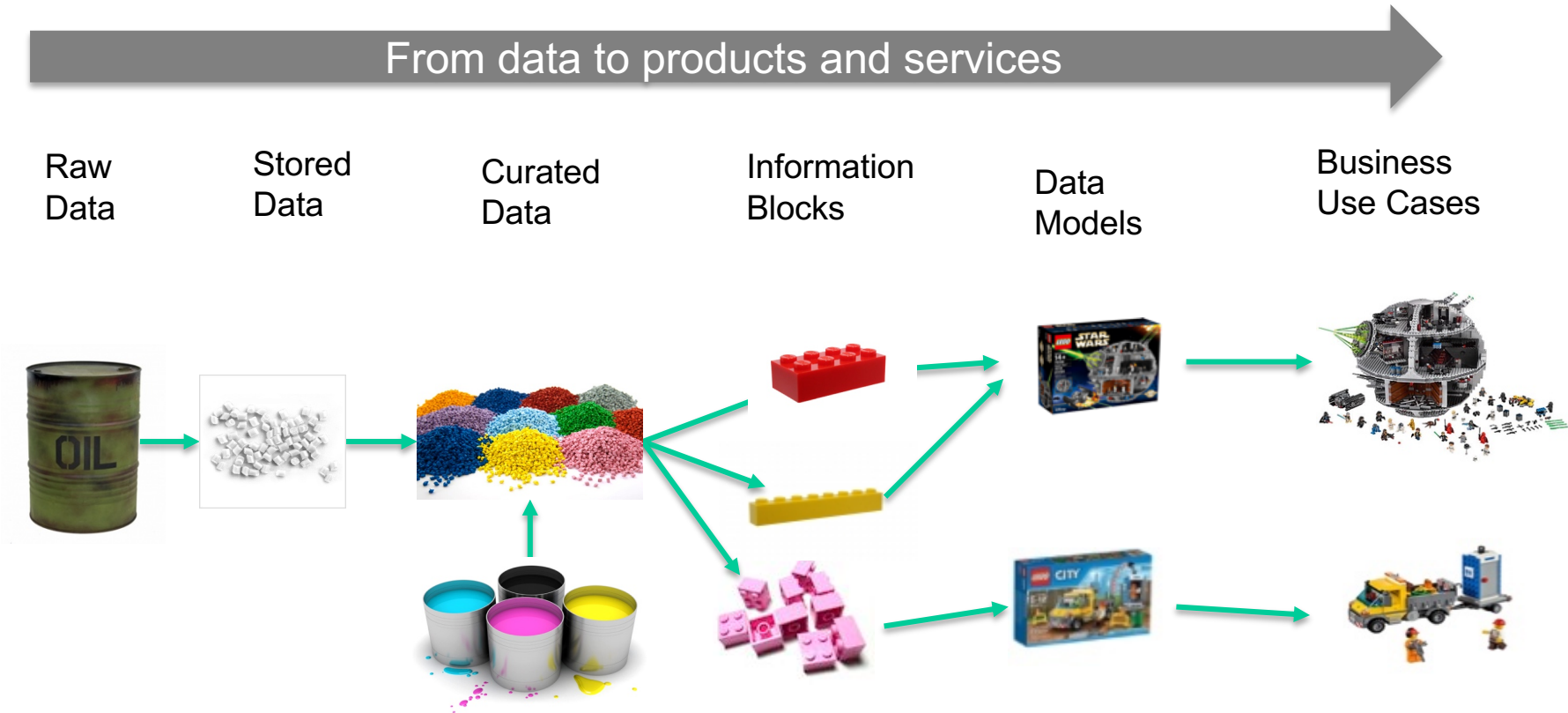
SOURCE: McKinsey Global Institute analysis

# The *n*-*V*s of Big Data

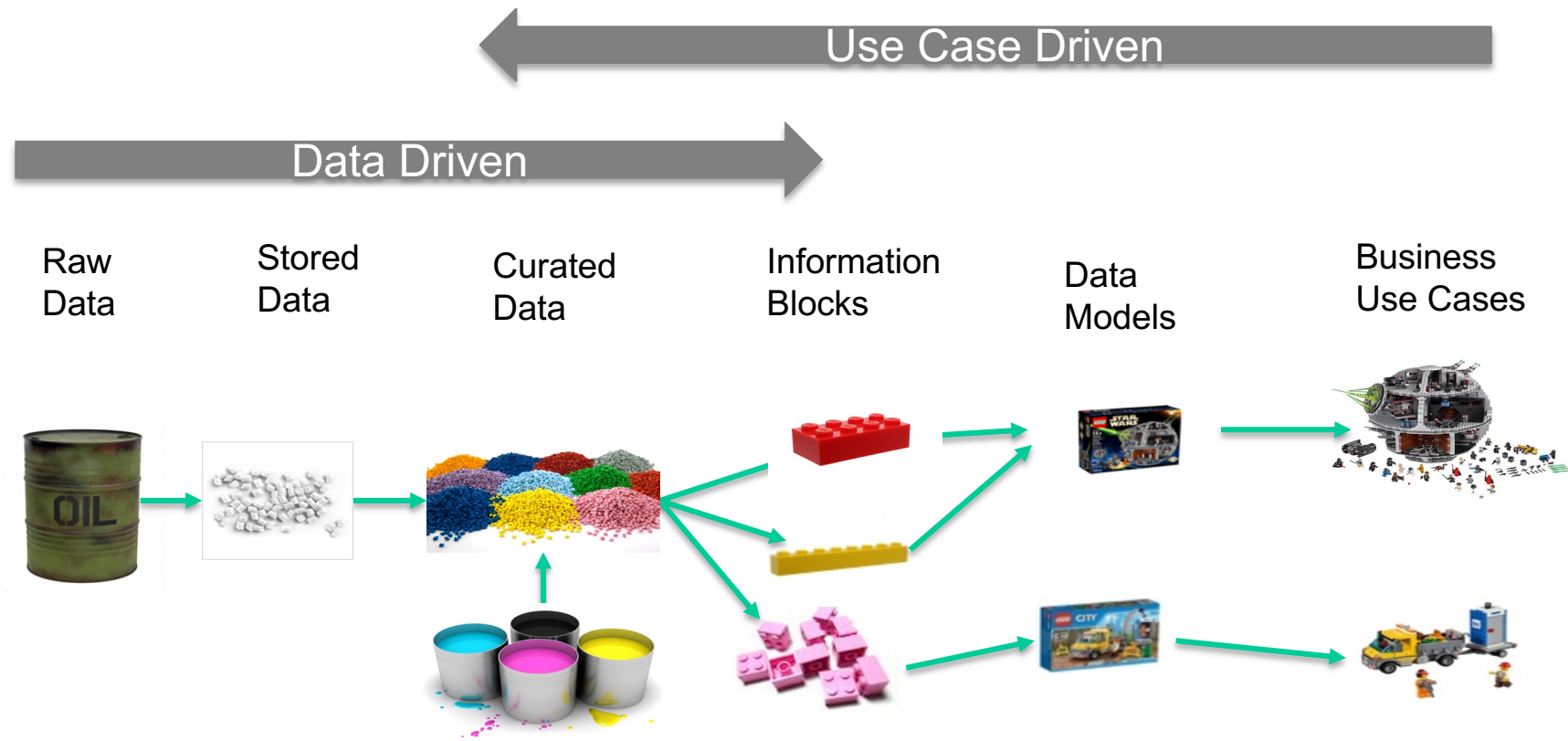
- *V*olume
  - amount of data (scale *out* not *up*)
- *V*elocity
  - speed of data in and out
- *V*ariety
  - range of data types and sources

[Gartner 2012] *"Big Data are high-volume, high-velocity, and/or high-variety information assets that require new forms of processing to enable enhanced decision making, insight discovery and process optimization"*

# Data vs. Traditional Assets

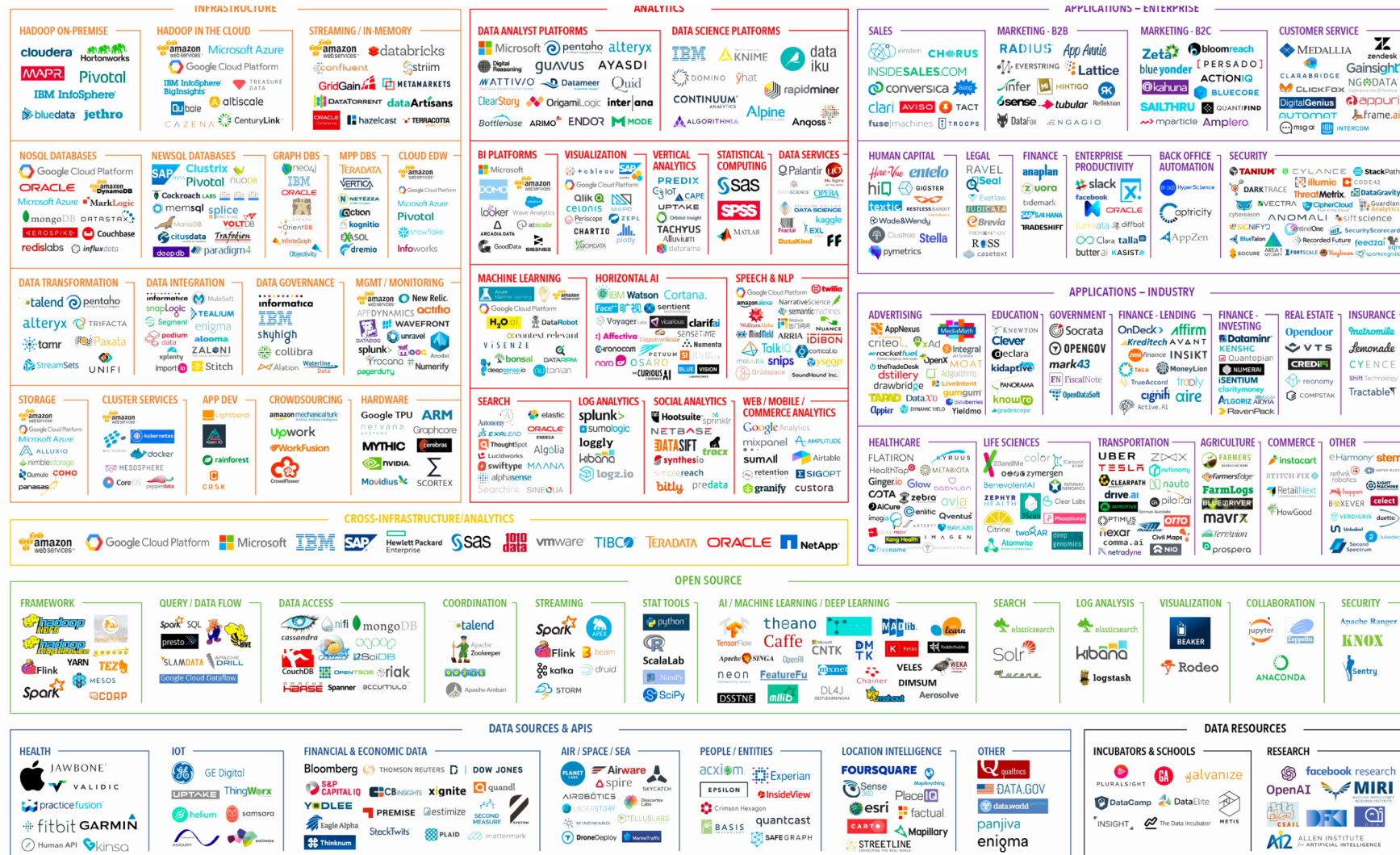


# Use Case or Data Driven



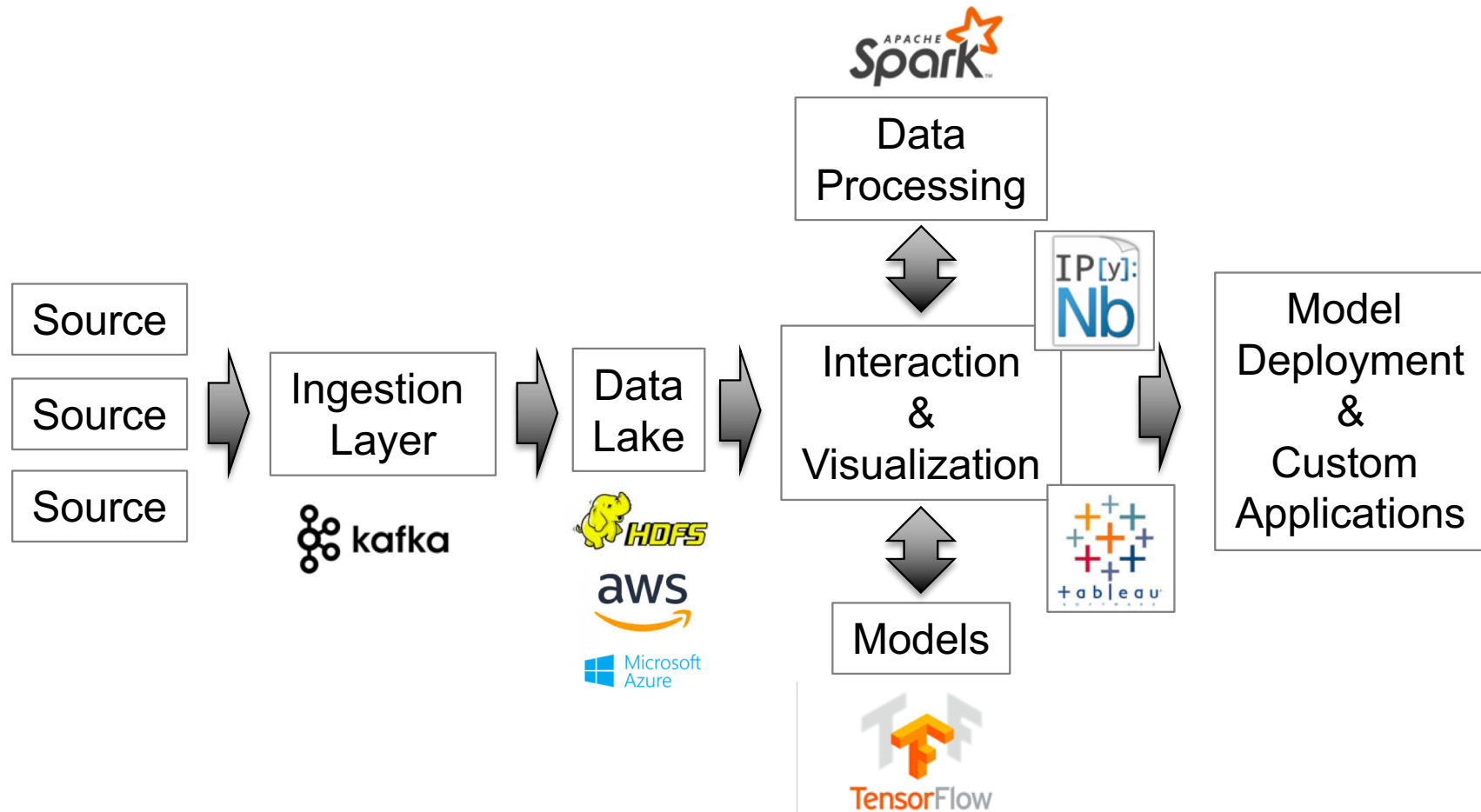


# Big Data Landscape





# Typical Big Data Architecture (circa 2018)



# The *n*-*V*s of Big Data

- *V*olume
  - amount of data
- *V*elocity
  - speed of data in and out
- *Variety* (*fusing n data sources as an input to a model*)
  - range of data types and sources

# Entity-Centric Data Fusion

Higher-level apps

Captures both direct and indirect relationships

Knowledge Graph

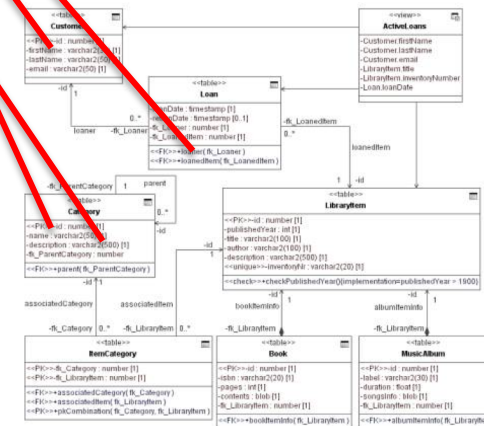
Soldered STD W/A			
Size	Rth <sub>sp</sub> (°C/W)	PCB sCu Rth <sub>sp</sub> (°C/W)	PCB Mcu Rth <sub>sp</sub> (°C/W)
0603	27	200	67
1206	20	110	60
2010	12	95	52
2512	11	95	51

Soldered Enlarged W/A			
Size	Rth <sub>sp</sub> (°C/W)	PCB sCu Rth <sub>sp</sub> (°C/W)	PCB Mcu Rth <sub>sp</sub> (°C/W)
1206	5	95	45
2010	2	85	42
2512	1	85	41

Glued STD W/A			
Size	Rth <sub>sp</sub> (°C/W)	PCB sCu Rth <sub>sp</sub> (°C/W)	PCB Mcu Rth <sub>sp</sub> (°C/W)
1206	33	123	73
2010	18	101	58
2512	16	100	56

Wire Bonding on Back Side Soldered Chip Resistor			
Size	Rth <sub>sp</sub> (°C/W)	PCB sCu Rth <sub>sp</sub> (°C/W)	PCB Mcu Rth <sub>sp</sub> (°C/W)
1206	5	95	45
2010	2	85	42
2512	1	85	41

Wire Bonding on Back Side Glued Chip Resistor			
Size	Rth <sub>sp</sub> (°C/W)	PCB sCu Rth <sub>sp</sub> (°C/W)	PCB Mcu Rth <sub>sp</sub> (°C/W)
1206	10	100	50
2010	4	87	44
2512	2	86	42



# Three Big Data Fusion Applications

1. Anomaly Detection for Smart Cities
2. Crime Prediction using Data Fusion
3. Infrastructure Monitoring

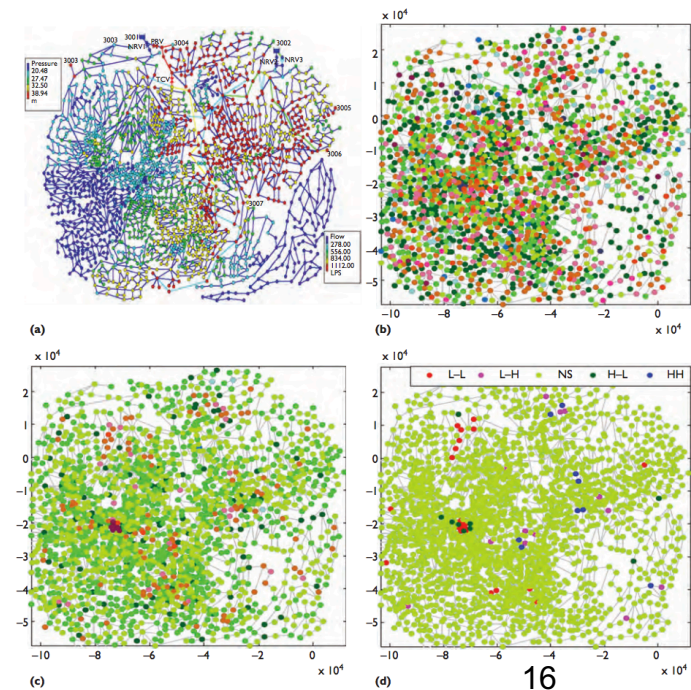
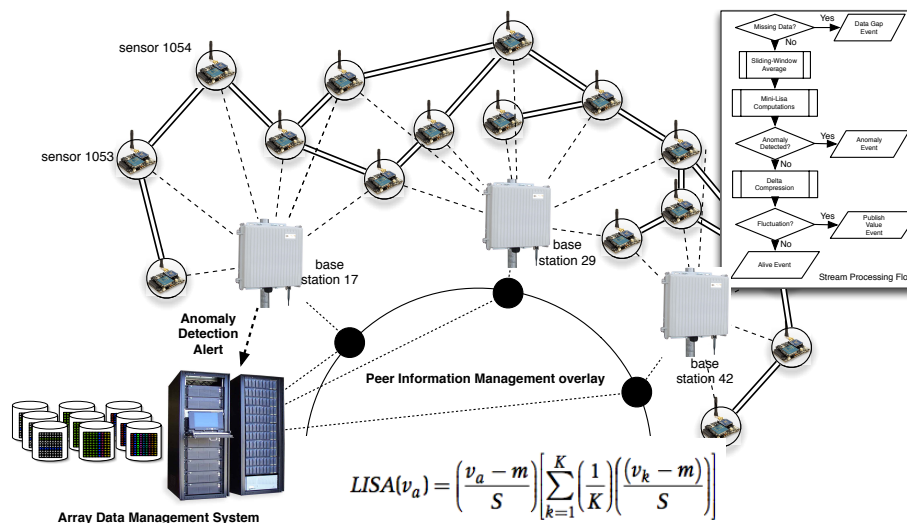
# 1. Anomaly Detection for Smart Cities

- Detecting leaks / pipe bursts / contamination in real-time for water distribution networks



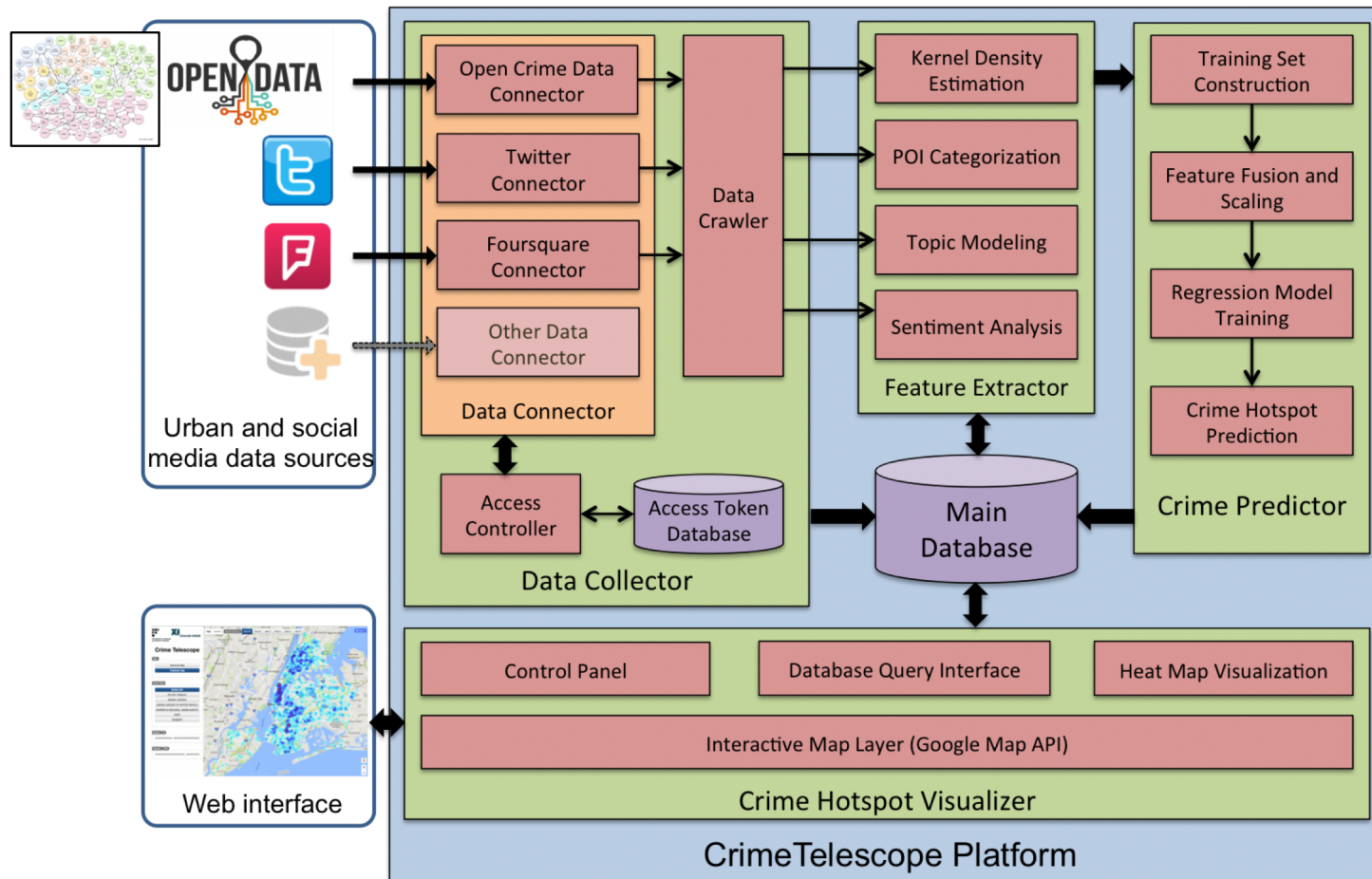
# Sensors installed in the water pipes!

- Spatial + temporal statistical processing (mini-Lisas)
- **Stream** processing (Storm) + **Array** processing (SciDB)





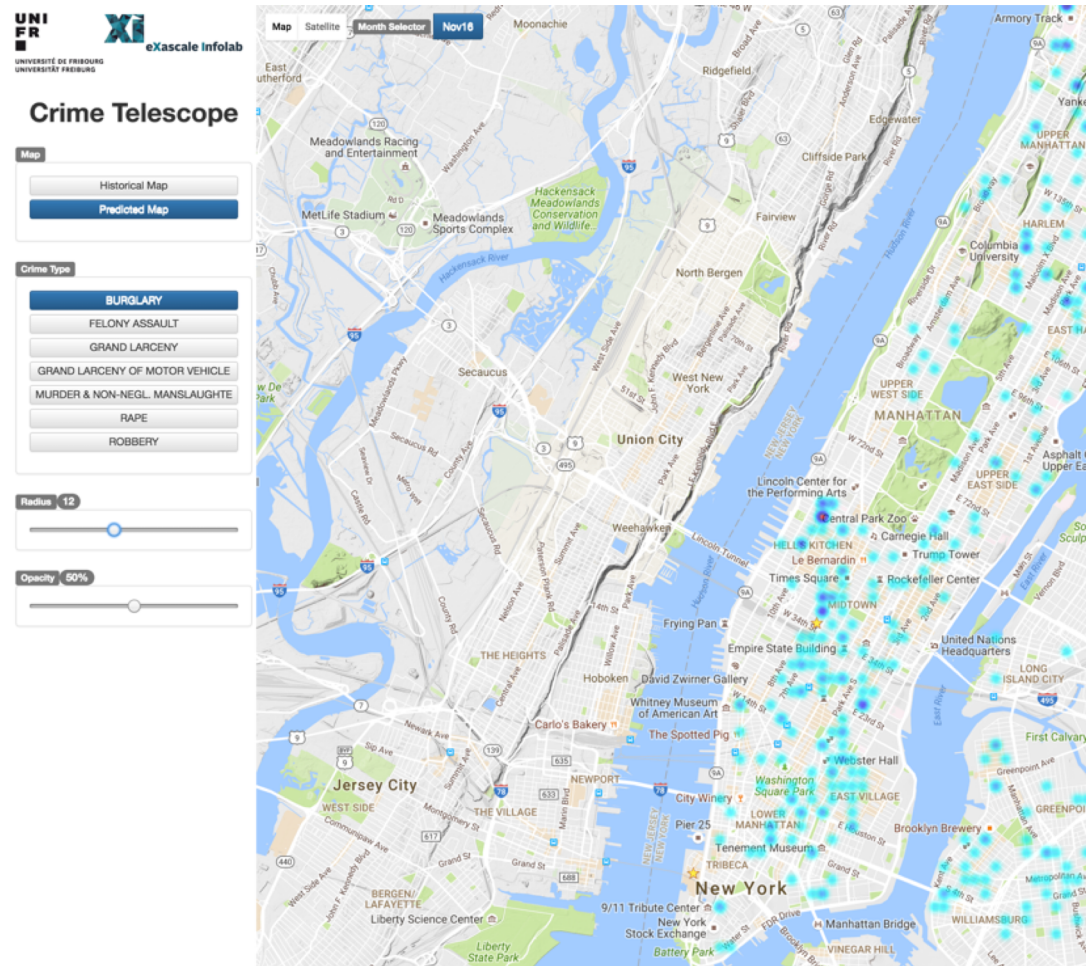
## 2. Crime Prediction using Big Data Fusion



- How to predict crime hotspots more accurately?
- **Fusion** of historical, urban & social data

CrimeTelescope: Crime Hotspot Prediction based on Urban and Social Media Data Fusion. D. Yang, T. Heaney, A. Tonon, L. Wang, P. Cudre-Mauroux. WWWJ 2017.

# Crime Prediction using Data Fusion





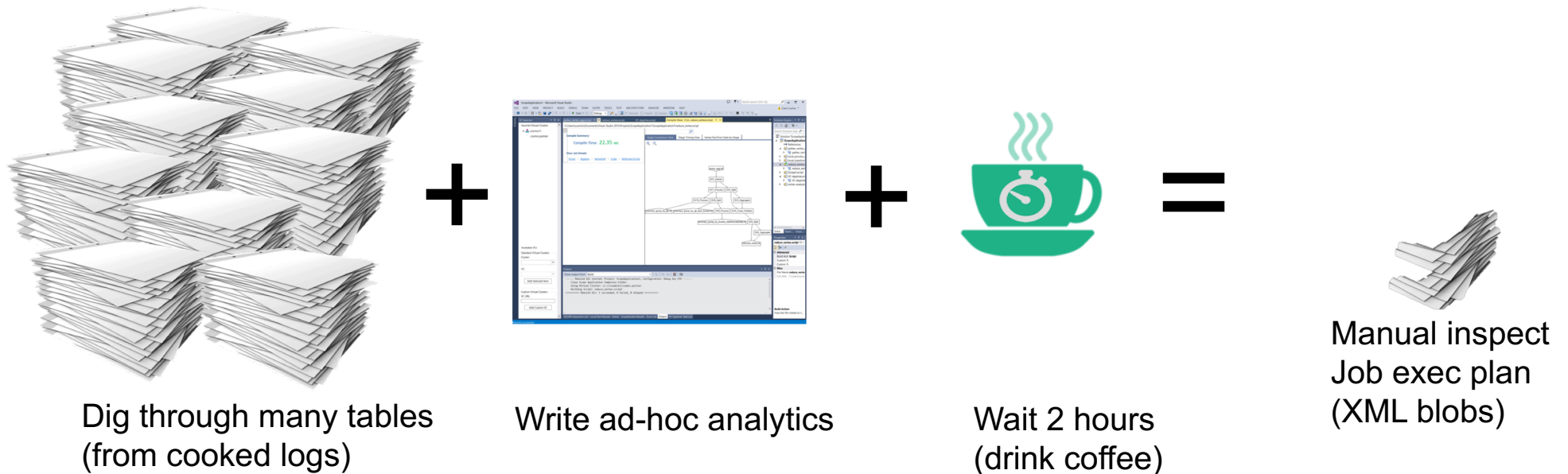
### 3. Infrastructure Monitoring

Microsoft's own Metadata Lake...

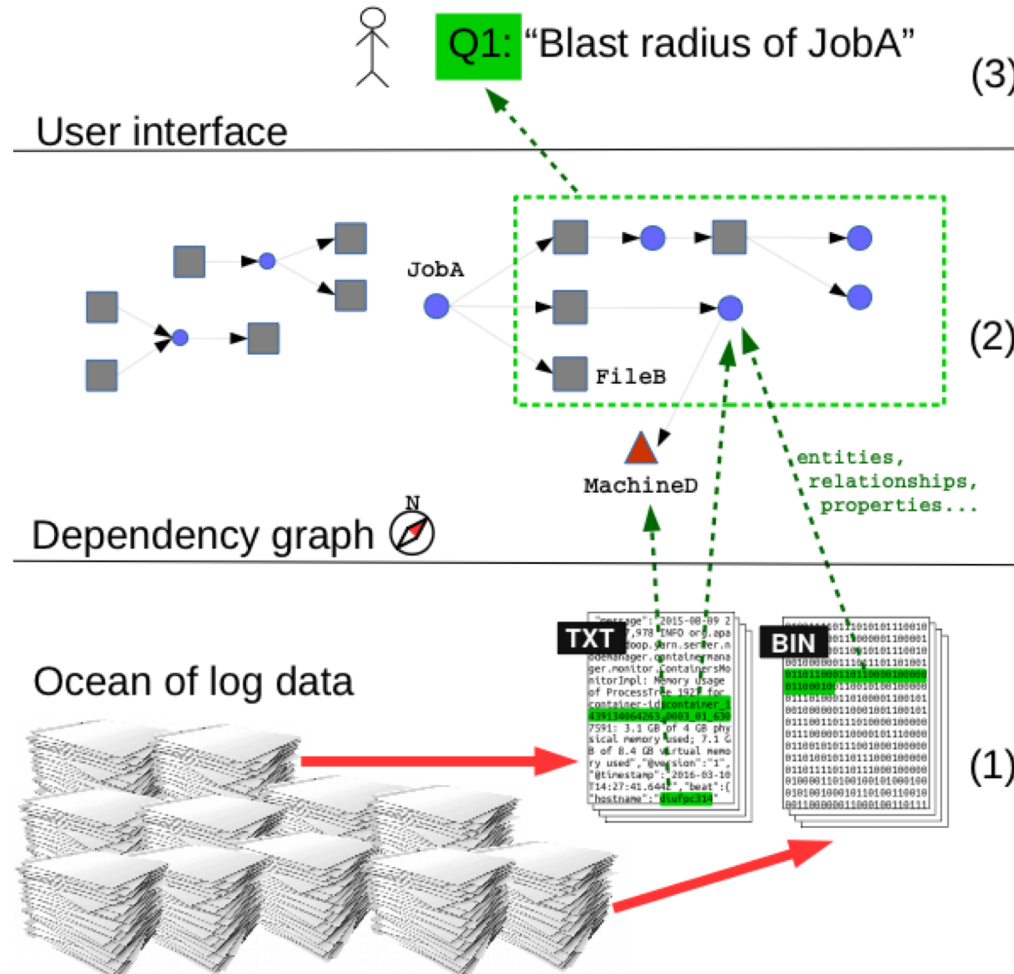


# Example 1: Job pipeline analysis (state-of-the-art)

- **User:** *“I need help with my ML experiment processing Clicklogs”*
- **Ops / Engineer:**



# Our Solution: Guider

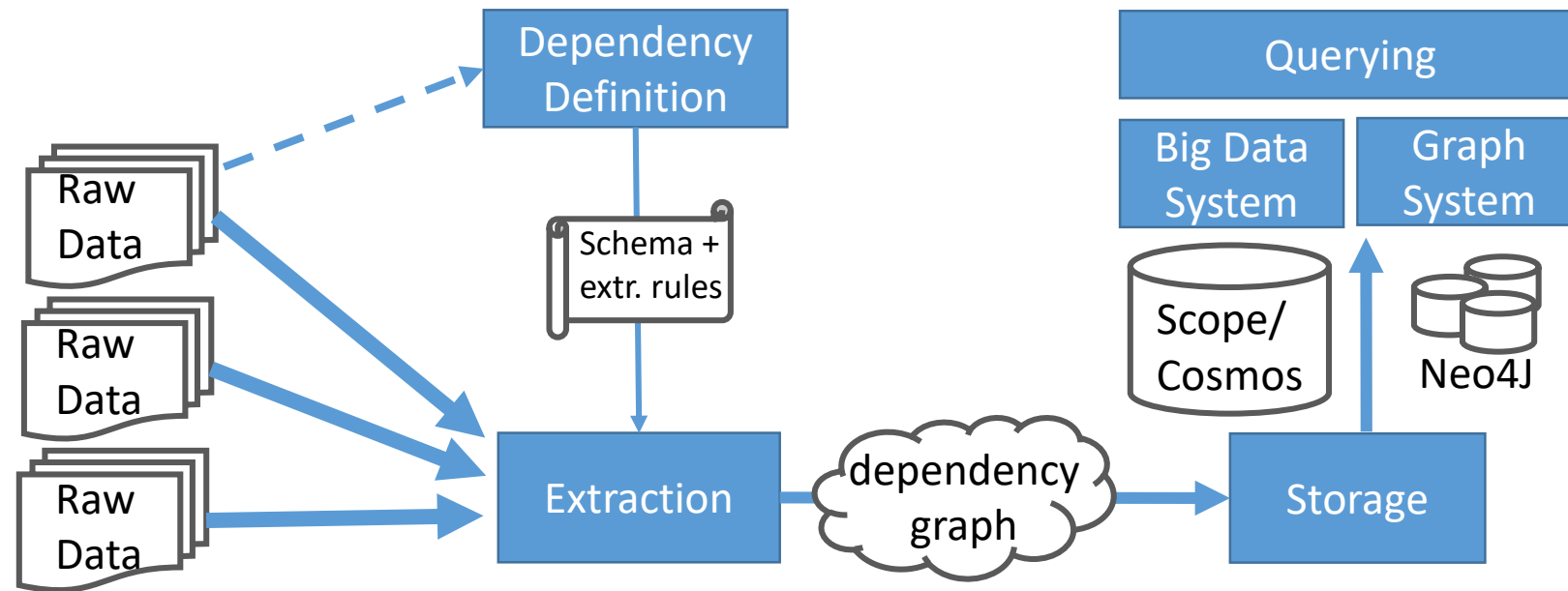


(3) User-level queries return bytes of aggregated data.

(2) Entity graph that represents a lightweight “skeleton” of the logs used for navigation

(1) Petabytes of daily logs

# Guider Architecture

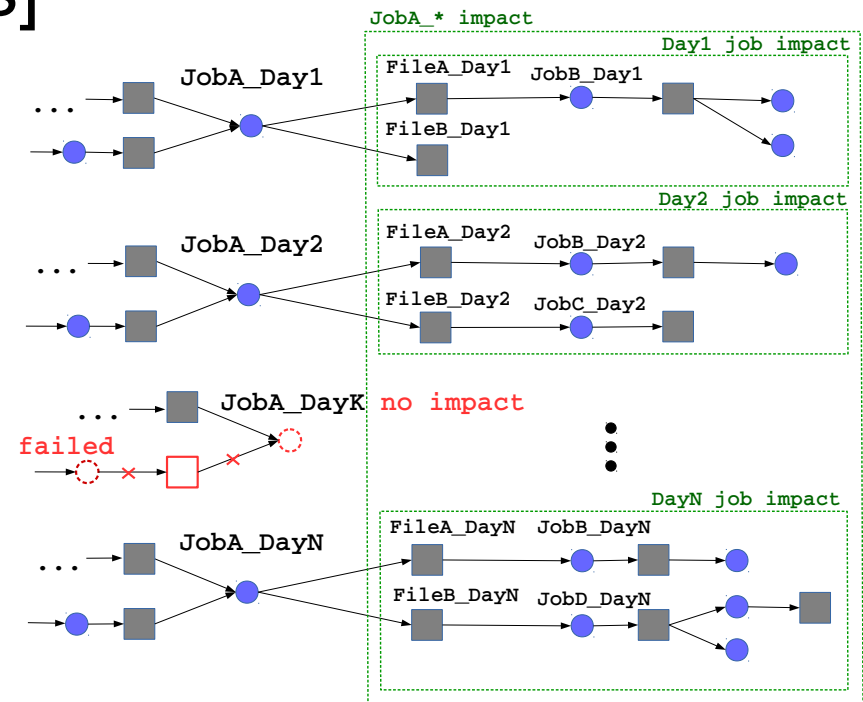


Dependency-Driven Analytics: a Compass for Uncharted Data Oceans.

Ruslan Mavlyutov, Carlo Curino, Boris Asipov, and Phil Cudre-Mauroux. CIDR 2017

# Guider Use-Cases

1. Auditing and Compliance [in production]
2. Job Scheduling [Morpheus]
3. Global Job Ranking
4. Datacenter migration



# Thanks for your Attention!



<https://exascale.info>