

When to Reach for the Cloud

Using Parallel Hardware for Link Discovery

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Outline

- 1 Motivation
- 2 Goal
- 3 Approach
- 4 Evaluation
- 5 Conclusion

Outline

① Motivation

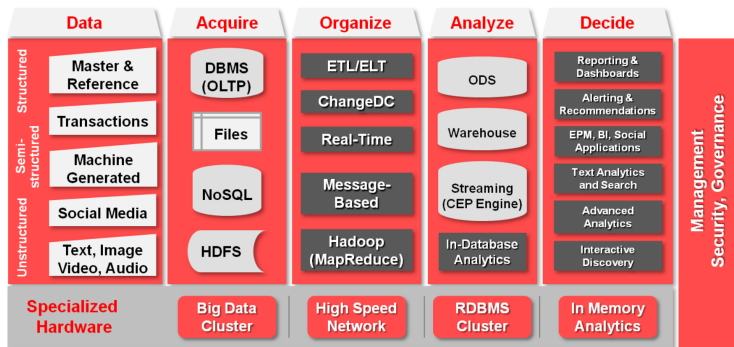
② Goal

③ Approach

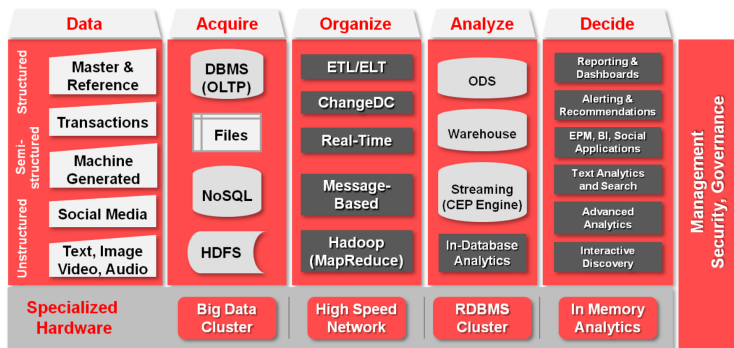
④ Evaluation

⑤ Conclusion

Big Data Processing



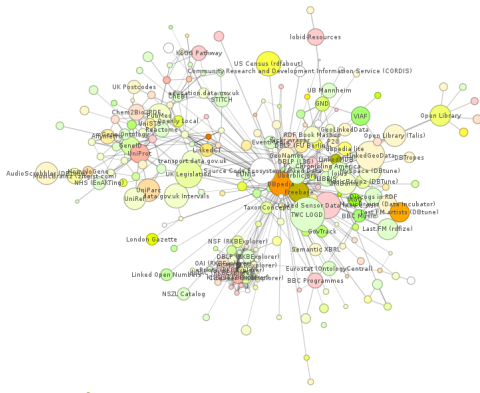
Big Data Processing



- Tendency towards porting all processing steps to the Cloud
- Full results available at <http://big-project.eu>
- **Question:** When should we reach to the cloud?

Why Link Discovery?

- 1 Fourth principle
- 2 Links are central for
 - **Data Integration**
 - Cross-ontology QA
 - Reasoning
 - Federated Queries
 - ...
- 3 Current topology of the LOD Cloud
 - 31+ billion triples
 - ≈ 0.5 billion links
 - `owl:sameAs` in most cases



Why is it difficult?

① Time complexity

- Large number of triples
- Quadratic a-priori runtime
- 69 days for mapping cities from DBpedia to Geonames



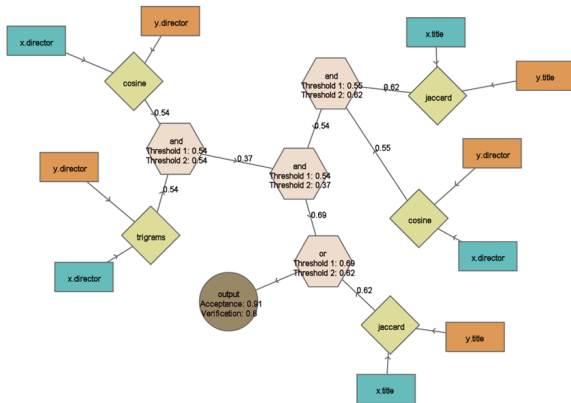
Definition (Link Discovery)

- Given sets S and T of resources and relation \mathcal{R}
- Find $M = \{(s, t) \in S \times T : \mathcal{R}(s, t)\}$
- Common approaches:
 - Find $M' = \{(s, t) \in S \times T : \sigma(s, t) \geq \theta\}$
 - Find $M' = \{(s, t) \in S \times T : \delta(s, t) \leq \theta\}$

Why is it difficult?

② Complexity of specifications

- Combination of several attributes required for high precision
- Tedious discovery of most adequate mapping
- Dataset-dependent similarity functions



Dealing with Time Complexity

1 Devise better algorithms



- Blocking
- Algorithms for given metrics
 - PPJoin+, EDJoin
 - HR^3

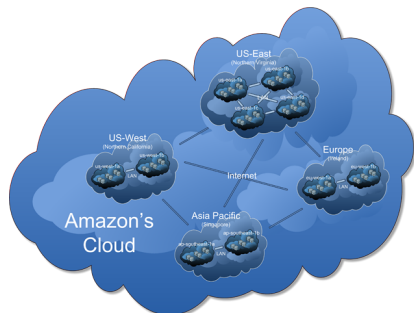
Dealing with Time Complexity

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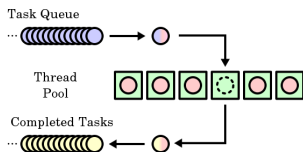
2 Use parallel hardware



- Threads pools
- MapReduce
- Massively Parallel Hardware

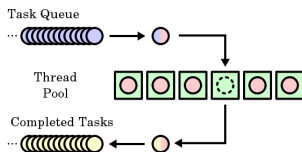
Parallel Implementations

- Different architectures
 - Memory (shared, hybrid, distributed)
 - Execution paths (different, same)
 - Location (remote, local)



Parallel Implementations

- Different architectures
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Question

- When should which use which hardware?

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Premises

- Given an algorithm that runs on all three architectures ...

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- **Note to self:** Implement one
 - Picked HR^3
 - Reduction-ratio-optimal



Premises

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- **Note to self:** Implement one
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Goals

- Compare runtimes on all three parallel architectures
- Find break-even points



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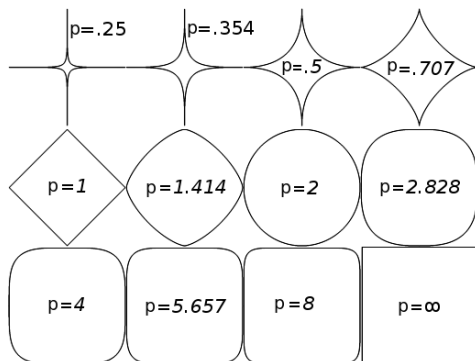
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HR³

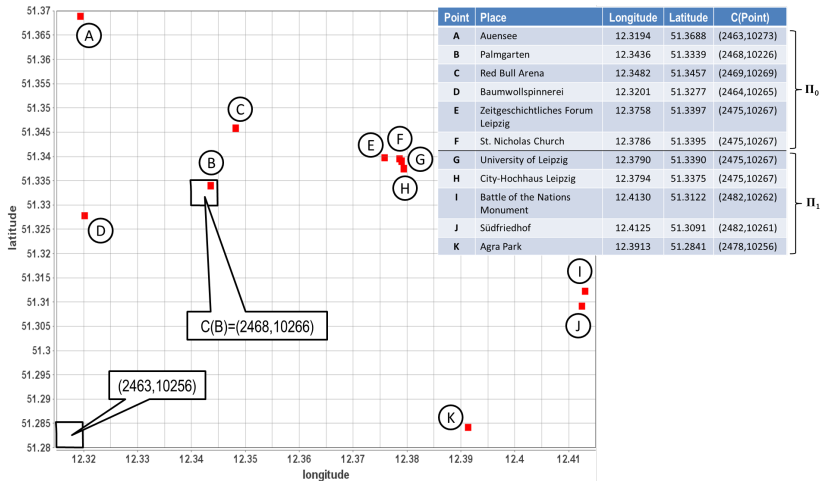
- 1 Assume spaces with Minkowski metric and $p \geq 2$

$$\delta(s, t) = \sqrt[p]{\sum_{i=1}^n |s_i - t_i|^p}$$



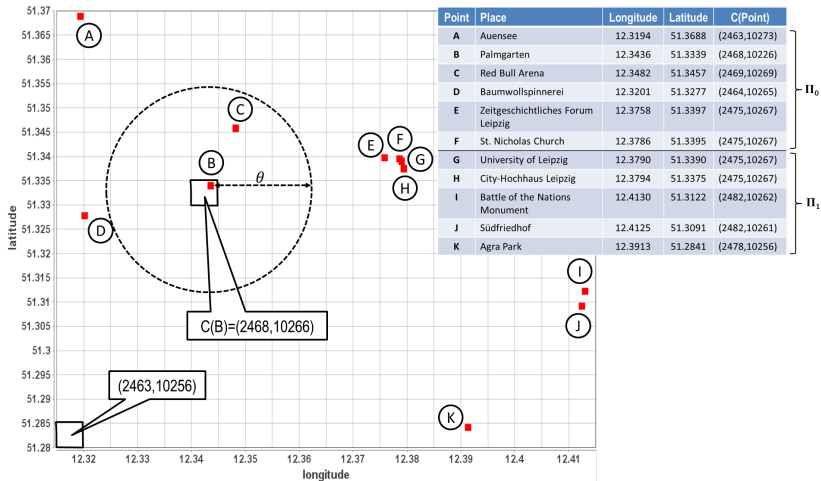
HR³

2 Create grid of width $\Delta = \theta/\alpha$



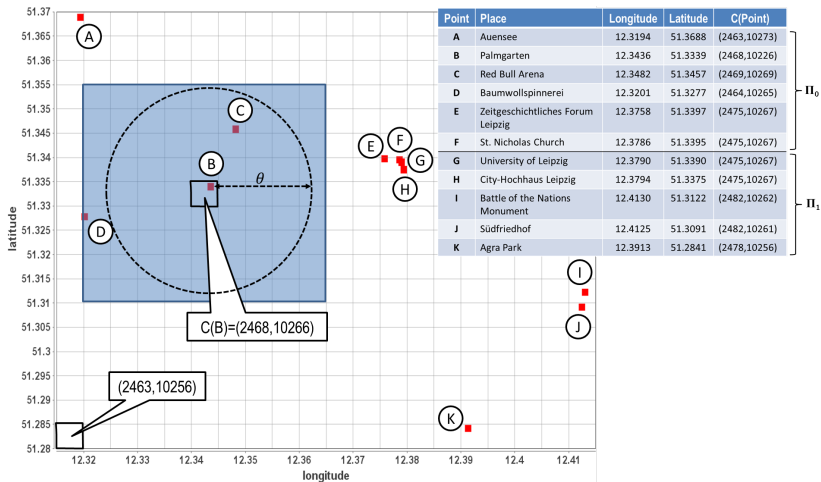
HR³

3 Link discovery condition describes a hypersphere



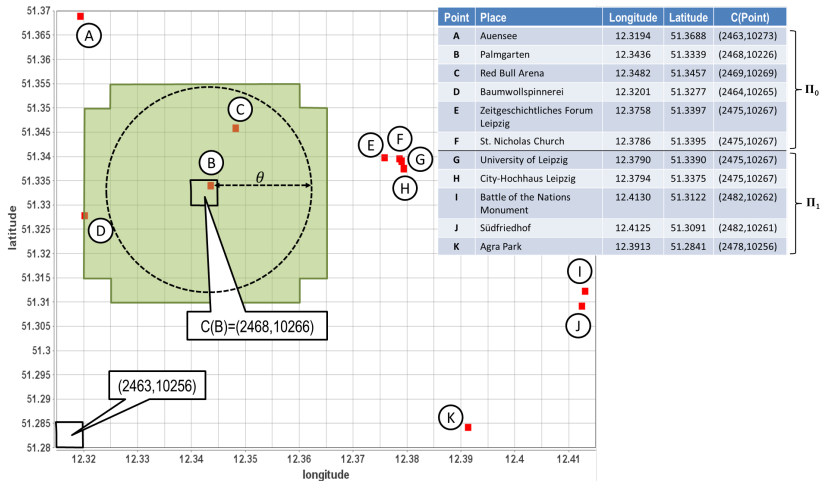
HR³

4 Approximate hypersphere with hypercube



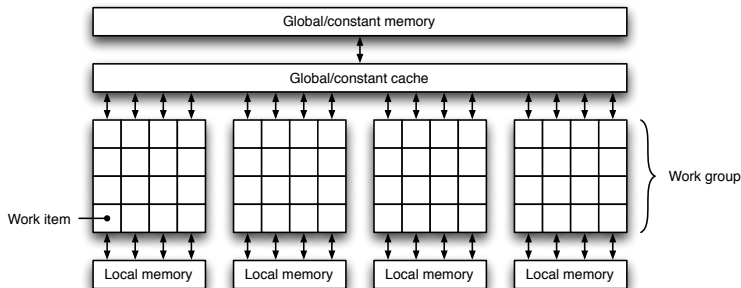
HR³

5 Use index to discard portions of hypercube



HR^3 in GPUs

- Large number of simple compute cores
- Same instruction, multiple data
- Bottleneck: PCI Express Bus
 - 1 Run discretization on CPU
 - 2 Run indexing on GPU
 - 3 Run comparisons on CPU



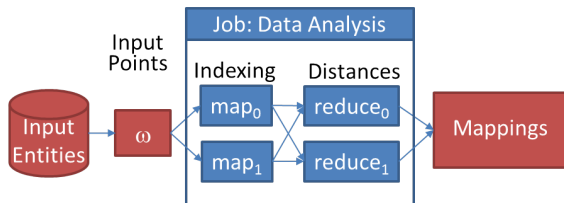
HR³ on the Cloud

- **Naive Approach**

HR^3 on the Cloud

- **Naive Approach**

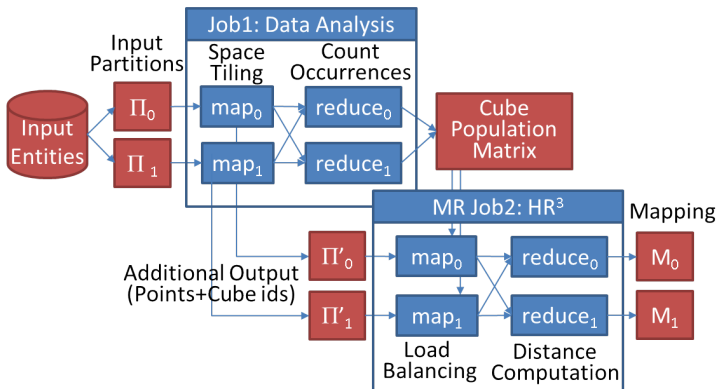
- ① Rely on Map Reduce paradigm
- ② Run discretization and assignment to cubes in map step
- ③ Run distance computation in reduce step



HR^3 on the Cloud

- **Load Balancing**

- 1 Run two jobs
- 2 Job1: Compute cube population matrix
- 3 Job2: Distribute balanced linking tasks across mappers and reducers



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Hardware

1 CPU (Java)

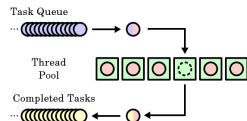
- 32-core server running Linux 10.0.4
- AMD Opteron 6128 clocket at 2.0GHz

2 GPU (C++)

- AMD Radeon 7870 with 20 compute units, 64 parallel threads
- Host program ran on Intel Core i7 3770 CPU with 8GB RAM and Linux 12.10
- Ran the Java code on the same machine for scaling

3 Cloud (Java)

- 10 c1.medium nodes (2 cores, 1.7GB) for small experiments
- 20 c1.large nodes (8 cores, 7GB) for large experiments



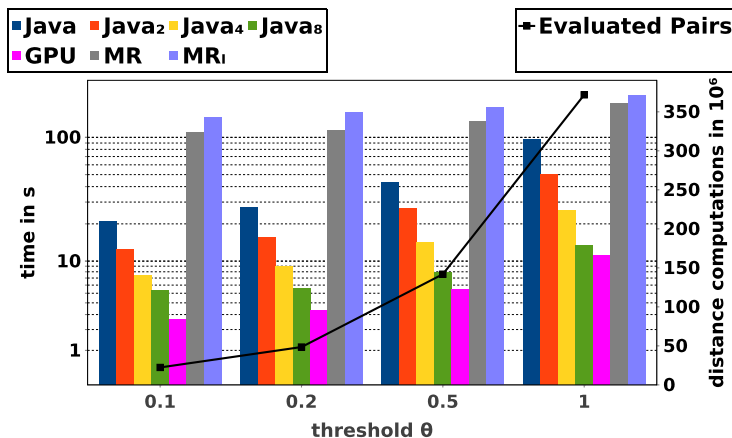
Experimental Setup

- Run deduplication task
- Evaluate behavior on different number of dimensions
- **Important:** Scale results
 - Different hardware (2-7 times faster C++ workstation)
 - Programming language
- Evaluate scalability (DS₄)

Dataset	Source	Size	Features
DS ₁	DBPedia	25,781	min/medium/max elevation
DS ₂	DBPedia	475,000	latitude, longitude
DS ₃	Linked Geo Data	500,000	latitude, longitude
DS ₄	Linked Geo Data	6,000,000	latitude, longitude

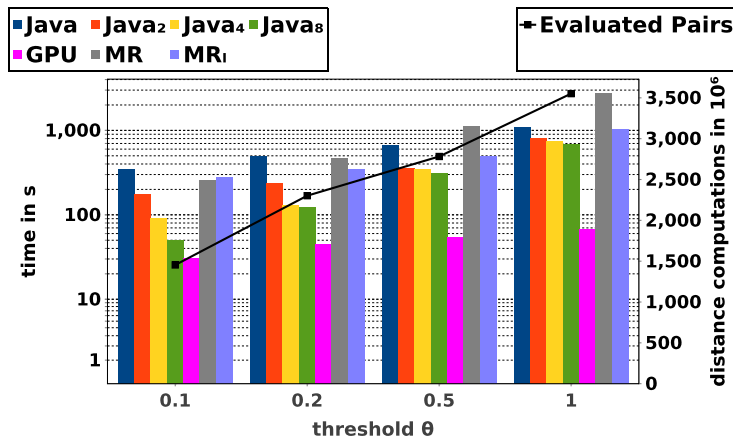
Results – DS1

- DBpedia, 3 dimensions, 26K
- CPUs scale better across different θ



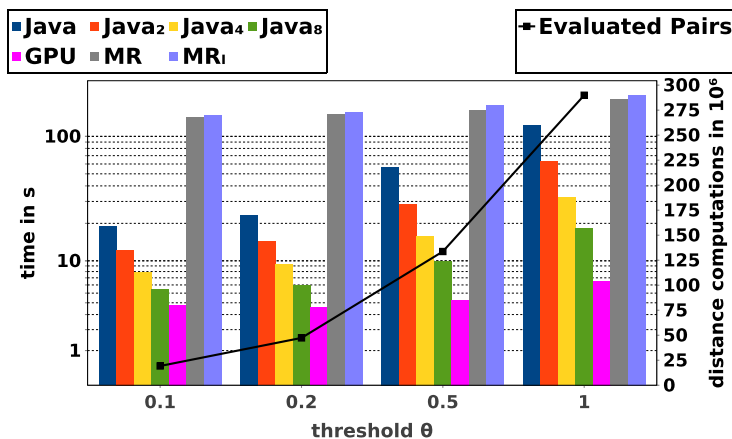
Results – DS2

- DBpedia, 2 dimensions, 475K
- GPUs scale better across different θ
- **Break-even point** $\approx 10^8$ results



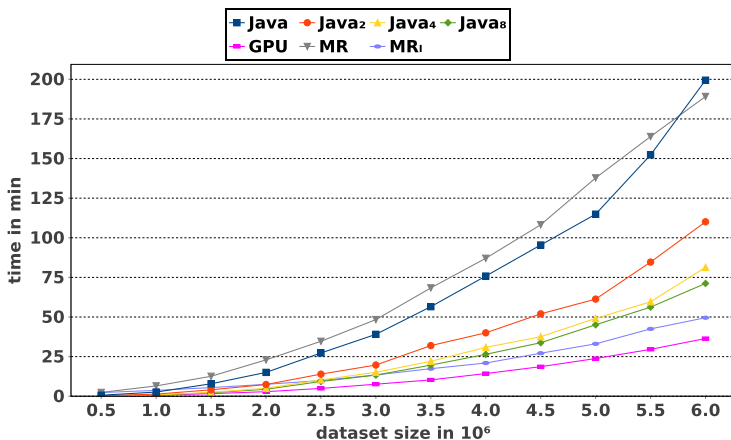
Results – DS3

- LinkedGeoData, 2 dimensions, 500K
- Similar picture



Results – Scalability

- LinkedGeoData, 2 dimensions, 6M
- Cloud (with load balancing) better for $\approx 10^{10}+$ results with 30 nodes



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Summary

- **Question**
 - When should we use which hardware for link discovery?
- **Results**
 - Implemented HR^3 on different hardware
 - Provided the first implementation of link discovery on GPUs
 - Devised a load balancing approach for linking on the cloud
 - Discovered 10^8 and 10^{10} results as break-even points



Summary

- **Question**
 - When should we use which hardware for link discovery?
- **Insights**
 - New hardware requirements for Big Data Processing
 - E.g., GPUs with buses faster than PCIe (e.g., Firewire speed)
 - **Accurate use of local resources sufficient for most of the current applications**



Vision

- **Self-adaptive link discovery frameworks**
 - ① Check for free hardware
 - ② Use free resources to be time-efficient
- Already developed approaches for predicting
 - ① Result size of algorithms
 - ② Runtimes on different hardware
- Need to develop **orchestration approach**



Thank You!

Questions?

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