

On Data Placement Strategies in Distributed RDF Stores

Int. Workshop on Semantic Big Data (SBD 2017)

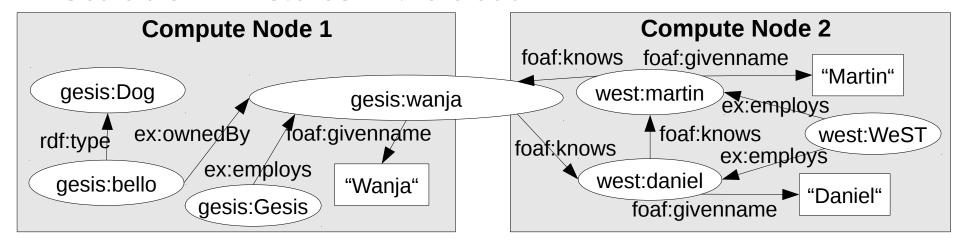
Daniel Janke, Steffen Staab, Matthias Thimm 19.05.2017





Distributed RDF Stores

- Requirement for trillion triples stores arose in the last years
- Scalable RDF stores in the cloud



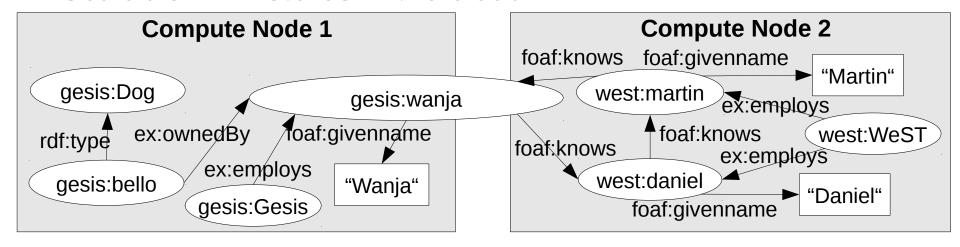
Challenges:

- Data placement strategies
- Distributed query processing
- Handling failures of compute nodes



Distributed RDF Stores

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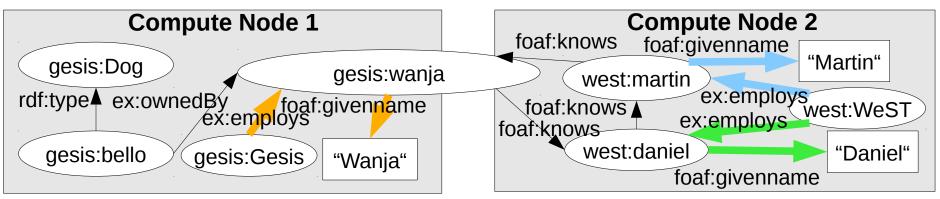


Challenges:

- Data placement strategies
- Focus of our research
- Distributed query processing
- Handling failures of compute nodes



Data Placement Strategies and Scalability



Horizontal containment

- Computation of individual query results on local data
- Indicator for robust query processing when scaling horizontally

Vertical parallelization

- Parallel computation of different query results on different compute nodes
- Indicator for query processing scaling with growing result set sizes when scaling horizontally



Data Placement Strategies and Scalability

Compute Node 1 desis:Dog Compute Node 2 foaf:knows foaf:givenname "Martin" Commonly held belief: Horizontal containment dominates query processing effort

(cf. [Huang2011SSQ, Lee2013EDP, Zhang2013ETS, ...])

Computation of individual query results on local data

Indicator for robust query processing when scaling horizontally

Vertical parallelization

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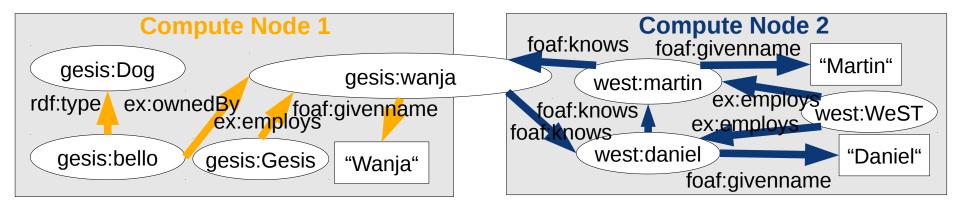


Outline

- 1) Data Placement Strategies
- 2) Benchmark methodology showing the interdependencies of data placement strategies and query processing
- 3) Analysis indicating that vertical parallelization may dominate horizontal containment
- 4) Conclusion



Graph Cover



Graph cover

Assignment of each triple to at least one compute node

Graph chunk

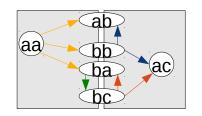
Set of triples assigned to a single compute node



Common Graph Cover Strategies

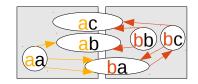
Hash cover [e.g. Harth2007YAF]

 Triple placement bases on subject hash modulo number of compute nodes

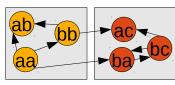


Hierarchical cover [Lee2013SQO]

 Triple placement bases on hash of subject IRI prefixes



Minimal edge-cut cover [Karypis1998AFA]



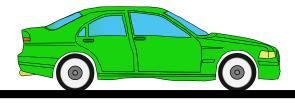
- Assign vertices (subjects and objects) to partitions such that
 - Number of edges between vertices of different partitions is minimized and
 - Each partition contains approximately $\frac{|V_G|}{|C|}$ ve



Common Evaluation Strategies

1) Evaluations of graph cover strategies using different databases => other components might bias evaluation results e.g. [Wu2014SAS, Zeng2013ADG]

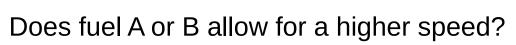






Car 1 using fuel A

Car 2 using fuel B

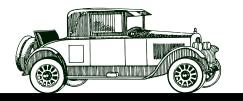






Common Evaluation Strategies

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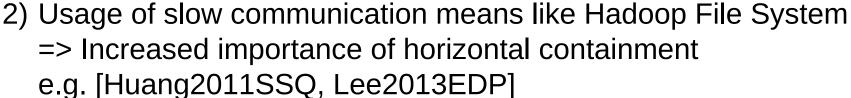




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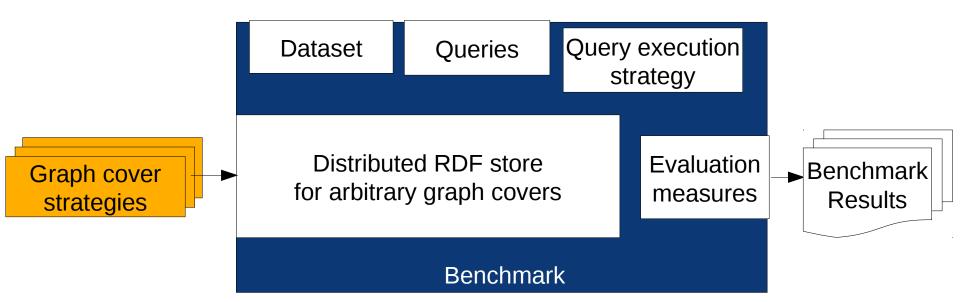


Images from https://openclipart.org



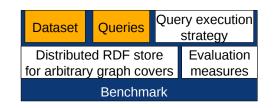
Benchmark Methodology

Goal: Investigating effect of graph cover on the scalability





Strategy for Generating Queries

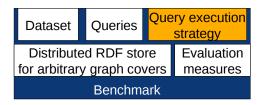


Query Generator: **SPLODGE** [Görlitz2012SSG]

- Generates SPARQL queries for arbitrary datasets
- Generates queries based on
 - Number of joins
 - Join pattern
 - Selectivity
 - Number of data sources

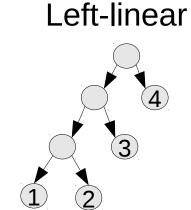


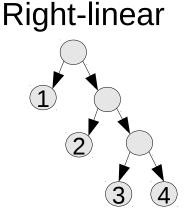
Query Execution Strategy



- Query optimizers fitting for arbitrary graph covers difficult
- Execution of several query execution trees:

Bushy 1 2 3 4





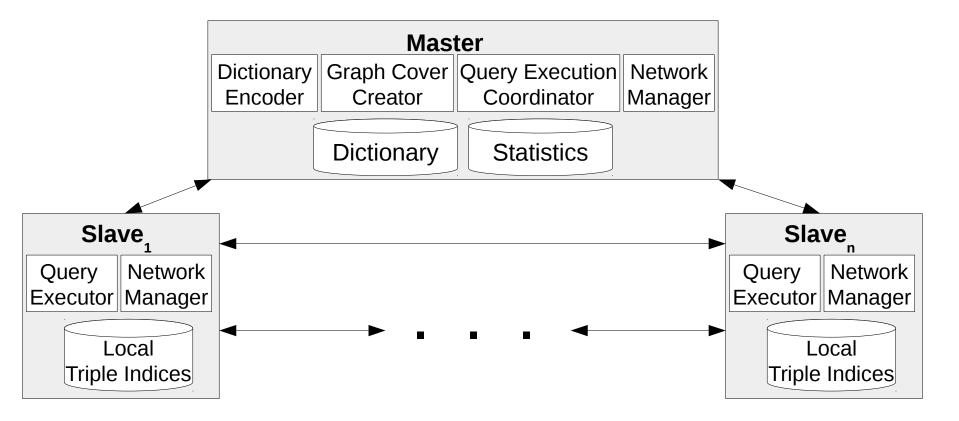


Koral

- Dataset Queries Query execution strategy

 Distributed RDF store for arbitrary graph covers

 Benchmark
- Graph cover independent distributed RDF store
- Inspired by TriAD [GurajadaTheobald2014TAD]





Evaluation Measures

Dataset Queries Query execution strategy Distributed RDF store for arbitrary graph covers Benchmark Query execution strategy Evaluation measures

Overall performance

Query execution time

Horizontal Containment

• Data transfer T: variable bindings transferred between compute nodes

Vertical Parallelization (VP)

• Workload Entropy W: entropy of join comparisons on each compute node

	T low	T high
W low	low VP	low VP
W high	high VP	low-medium VP



Experimental Setup

Compared graph cover strategies:

Hash, hierarchical hash and minimal edge-cut cover

Dataset:

1 trillion triple subset of BTC2014 [Käfer2014BTC]

Queries:

Number of joins: 2 and 8 triple patterns

Join pattern: path-shaped and star-shaped

• *Selectivity*: 0.001% and 0.01% (1M and 10M triples)

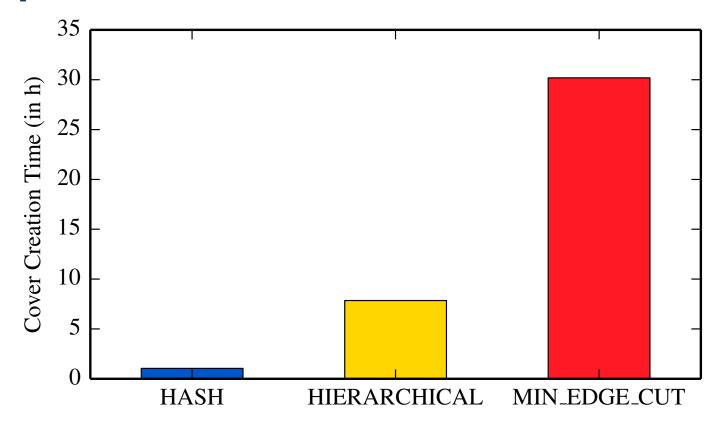
Number of data sources: 1 and 3

Computer environment:

- 1 Master à 4 cores, 8 GB RAM, 1 TB HDD
- 20 Slaves à 1 core, 2 GB RAM, 300 GB HDD
- 1 Gbit ethernet



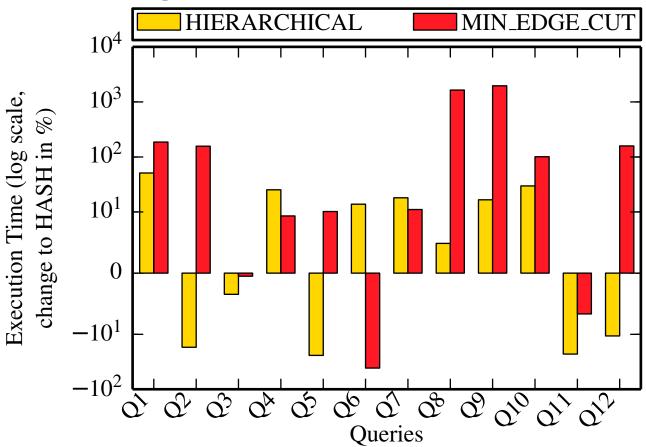
Graph Cover Creation Time



- Minimal edge-cut cover requires most time for creation
- Hash cover is created the fastest



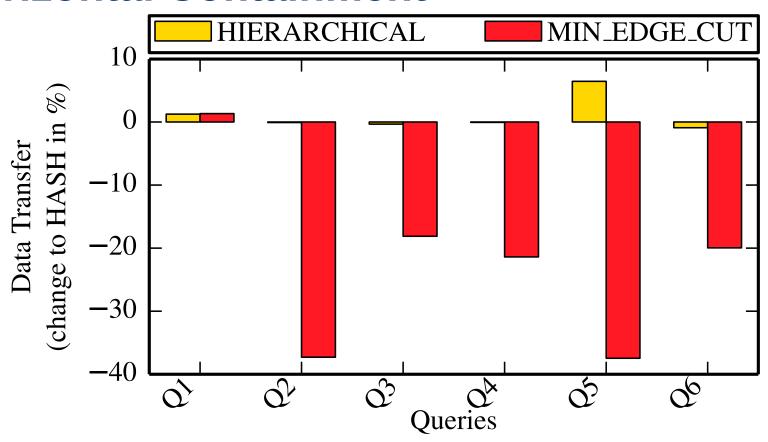
Overall Query Performance



- Bushy query execution outperforms other execution strategies
- Minimal edge-cut causes slowest query execution in most cases
- None of the hash-based covers is faster in general



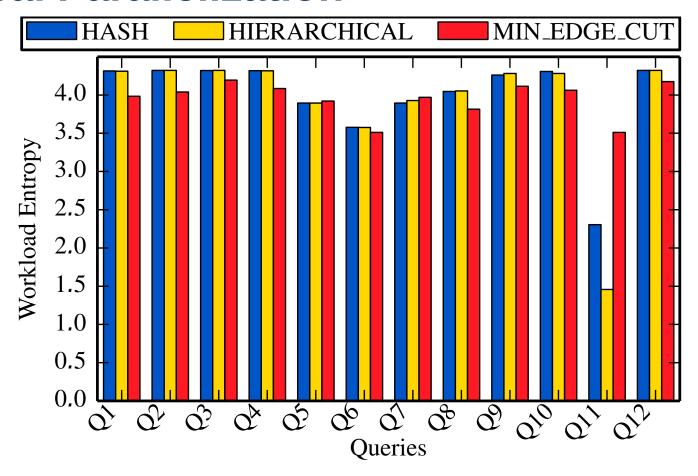
Horizontal Containment



- Star-shaped queries produce no data transfer
- Minimal edge-cut covers produces less data transfer
- Hash-based covers similar data transfer



Vertical Parallelization



- Minimal edge-cut cover has the least balanced workload
- Hash-based covers have similar balanced workloads



Conclusion

- Minimal edge-cut cover
 - Longest cover creation time
 - Lowest data transfer => high horizontal containment
 - Lowest workload balance => low vertical parallelization
 - Overall performance worse than hash-based covers
- Hash-based covers have similar performance
- Vertical parallelization might be more important than horizontal containment

Future work:

Benchmarking of workload-aware graph cover strategies



Thank you for your Attention!

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Contributions:

- 1) Benchmark methodology showing the interdependencies of graph cover strategies and query processing
- 2) A flexible open-source platform for performing the benchmark
- 3) Analysis indicating that vertical parallelization may dominate horizontal containment





References

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Hash Cover [e.g. Harth2007YAF]

Triple placement bases on subject hash modulo number of compute nodes

$$\operatorname{hashCover}(\langle s, p, o \rangle) := \operatorname{hash}(s) \mod |C|$$



Hierarchical Hash Cover [Lee2013SQO]

Triple placement bases on prefixes of subject IRIs

$$\operatorname{hashCover}(\langle s, p, o \rangle) := \begin{cases} \operatorname{hash}(\operatorname{prefix}(s)) \mod |C| &, \text{ if } s \in I \\ \operatorname{hash}(s) \mod |C| &, \text{ otherwise} \end{cases}$$

- IRI: http://www.w3.org/1999/02/22-rdf-syntax-ns#type
- Path hierarchy: org/w3/www/1999/02/22-rdf-syntax-ns/type
- Determine path hierarchy prefix such that
 - There exist at least |C| hierarchy prefixes
 - That are shared by at least $\omega\%$ of all triples



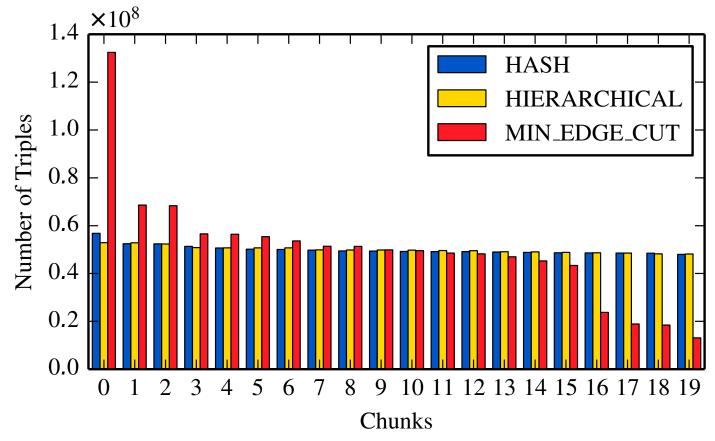
Minimal Edge-Cut Cover

Tries to solve the k-way graph partitioning problem [Karypis1998AFA]

- 1) Assign vertices (subjects and objects) to partitions such that
 - Number of edges between vertices of different partitions is minimized and
 - Each partition contains approximately $\frac{|V_G|}{|C|}$ vertices
- 2) Assign triple to the partition its subject is assigned to



Chunk Sizes



- Minimal edge-cut cover has most unbalanced chunks
- Hash-based covers have equally-sized chunks