



A Geospatial Join Optimization for Federated GeoSPARQL querying

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- Background: Federated query processing
- Optimization of federated within-distance queries
- Evaluation (using a real-world use case)
- Conclusions and future work



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Federated query processing

- Federated query processors are systems that seamlessly integrate data from multiple remote dataset servers
- receive a query, issue the necessary subqueries in the remote servers, combine the intermediate results accordingly, and presents the result to the client.
- used thoroughly in Linked Data; there exist many data providers that publish their (thematic) datasets through public SPARQL endpoints
- the technology is not yet mature for Geospatial Linked Data and GeoSPARQL endpoints



Federated Geospatial Joins

GeoSPARQL specification:

- geospatial operations are denoted using functions between geographic literals (e.g., geof:sfWithin, geof:sfIntersects, geof:distance)
- geographic literals are denoted using WKT serializations (e.g., "POINT(21.814 38.422)"^^geo:wktLiteral)
- features are linked with corresponding geographic literals using geo:hasGeometry/geo:asWKT chains

```
SELECT * WHERE
{
    ?s1 geo:hasGeometry ?g1 .
    ?g1 geo:asWKT ?w1 .
    ?s2 geo:hasGeometry ?g2 .
    ?g2 geo:asWKT ?w2 .
    FILTER ( geof:sfIntersects(?w1, ?w2) )
}
```

- A geospatial join is a *cross product* filtered by a *geospatial function*.
- A federated geospatial join is a *cross product* filtered by a *geospatial function* comparing shapes coming from different endpoints.

Federated Join Implementations

Federated Thematic Joins

- Many algorithms and implementations exist (*bind join*, *hash join*, *adaptive join*, etc.)
- Bind Join!
 - issue a query to the "left" endpoint, then pass its results as bindings to the "right" endpoint.
 - reduces the communication cost by reducing intermediate results.

Federated Geospatial Joins

- No specialized algorithms for federated geospatial joins exist
- Bind Join with a FILTER pushdown!
 - fetch "left" shapes to partially bind the geospatial function, then push the filter to the "right" endpoint
 - exploits the fact that geospatial functions are evaluated faster in the sources (spatial index).



"Given 2 endpoints (administrative divisions, hotels), fetch all hotels within Hersonisos"





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Federated Within-Distance Queries

- We reduce focus on **federated within-distance** queries:
 - o shapes from different endpoints that their distance is less than d
 - without requiring the exact distance

FILTER (geof:distance(?x, ?y, uom) < d).</pre>

- **Problem:** The evaluation such filters is computationally expensive:
 - it cannot be answered from the spatial index.
 - every shape is a potential match and its distance should be compared with the threshold.



Example: The process of finding all red shapes within distance d from the given green shape, is slow. For each shape we have to calculate the distance from the given shape and compare it with the threshold d.

Optimizing Federated Within-Distance Queries

- **Solution**: We augment the subquery to be issued to the "right-hand" endpoint with an additional FILTER:
 - keeps only shapes that do not intersect with a constructed rectangle
 - o used to prune all shapes that are "too-far away"
 - o can be answered from the spatial index of the source.
- We efficiently refine the set of candidate shapes before starting to actually compute distances.



Example: To speed up the process of finding all red shapes within distance d from the given green shape, we insert a condition that filters out all shapes that do not intersect with the blue rectangle.



"Given 2 endpoints (administrative divisions, hotels), fetch all hotels within 1km distance from Hersonisos"



Contributions

• Optimization technique

- We provide a pseudocode of our approach in the paper
- The actual algorithm is slightly more complex (designed to work with multiple bindings per query).

Correctness proof

 We show that the additional FILTER does not change the semantics of the original query (it does not prune any unwanted shapes)

• Implementation

- We provide an open source implementation of the technique
- The implementation is integrated within the *Semagrow* federation engine



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Experimental Evaluation Validating Crop type data using Ground Observations



- A federation with 2 GeoSPARQL endpoints:
 - INVEKOS (field parcels, owners' self declaration)
 - LUCAS (Ground observations of crop type data)
- 14.1 million triples, ~4GB of data in N-triples format
- Task: Estimate the crop-type accuracy of the INVEKOS

	Queries		
Q1	given a ground observation, return the <i>closest</i> field if it is within 10 meters and the crop types match	positive	
Q2	given a ground observation return the <i>closest</i> field if it is within 10 meters and their crop types do not match	negative	
Q3	given a ground observation, return it if there is <i>no</i> field within 10 meters	irrelevant	

Experimental Results Crop-type data validation task

	#queries in the workload	naive	optimized	
		query exec. time (average per query)	query exec. time (average per query)	
Q1	2488	120 sec	2.6 sec	
Q2	2488	119 sec	2.4 sec	
Q3	2488	117 sec	1.8 sec	
		query exec. time (total workload)	query exec. time (total workload)	
Q1-3	7494	10 days & 6 hours	4 hours & 39 min	

- We evaluate the full workload for the data validation task.
- **optimized** is faster than **naive** by 2 orders of magnitude.
- The queryset has several complex characteristics, but the bottleneck is the within-distance operation.
- Possible reason: the distance parameter is quite small (10 meters).

Experimental Results (cont.) Additional query

	distance	naive	optimized	
		query exec. time	query exec. time	shapes pruned
Q4	10 m	58 sec	0.1 sec	>99%
Q4	100 m	57 sec	0.1 sec	>99%
Q4	1 km	58 sec	0.1 sec	>99%
Q4	10 km	57 sec	1.2 sec	99%
Q4	50 km	72 sec	26 sec	84%
Q4	100 km	110 sec	86 sec	60%

Q4 return all fields within-distance D from a specific ground observation (D = 10m-100km)

- **naive**: ~1 min to calculate the result, remaining time to fetch the result.
- optimized: smaller distance parameter → higher amount of pruning → faster query execution time. The additional filter does not introduce any time overheads.
- huge time difference for distance ≤1km, less pronounced for distance ≥50km.



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Conclusions

• We proposed an optimization for federated geospatial within-distance joins

- Augments the subqueries prepared for each source with additional filters that can be answered from the spatial index of the federated sources.
- Does not change the semantics of the query
- Implemented within the Semagrow federation engine.

• We show in our evaluation that the optimization substantially speeds-up query execution

- we used datasets and queries from a practical use-case from the agro-environmental domain
- very effective, especially for small distance limits (2 orders of magnitude for 1km)
- o can be useful for real-world applications (restrictions used to limit results to a local scope)



• Develop a GeoSPARQL extension where within-distance is expressed with a single function.

• Develop similar rewriting techniques for optimizing queries with other GeoSPARQL functions.



Thank you!

Visit us at: https://github.com/semagrow/



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