GEO-SERIES | EPISODE #13

Handling large spatial data using PostgreSQL and DuckDB



Speaker



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Kshitij Raj Sharma is a product owner for map data access services at HOTOSM from Nepal, He is a passionate spatial developer with a love for open-source software and open data. His expertise in spatial data and deep interest in mapping led him to explore the potential of Al. Kshitij has been experimenting and advocating for FOSS and contributing to open data initiatives since past seven years. He is also a maintainer of several opensource tools including Free and open source Al tool: fAlr, OSM Export Tool, Raw Data API, geojson2osm, OSMSG, etc and an integral member of our own OSGEO Nepal

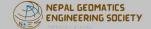
Moderator



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Handling Large Spatial Data using PostgreSQL and DuckDB

Kshitij Raj Sharma





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PostgreSQL

- Most popular open source database
- Reliable
- Feature Rich
- Powerful indexes
- Large dev support group
- Production ready database support in multiple commercial cloud service provider
- Rich in spatial extensions like postgis , pg_raster



DuckDB

https://duckdb.org/

DuckDB at a glance



Simple

DuckDB is easy to install and deploy. It has zero external dependencies and runs in-process in its host application or as a single binary.

Read more →



Portable

DuckDB runs on Linux, macOS, Windows, and all popular hardware architectures. It has idiomatic client APIs for major programming languages.

Read more →



Feature-rich

DuckDB offers a rich SQL dialect. It can read and write file formats such as CSV, Parquet, and JSON, to and from the local file system and remote endpoints such as S3 buckets.

Read more →



Fast

DuckDB runs analytical queries at blazing speed thanks to its columnar engine, which supports parallel execution and can process larger-than-memory workloads.

Read more →



Extensible

DuckDB is extensible by third-party features such as new data types, functions, file formats and new SQL syntax.

Read more →



Free

DuckDB and its core extensions are open-source under the permissive MIT License.

Read more →

- Spatial Support



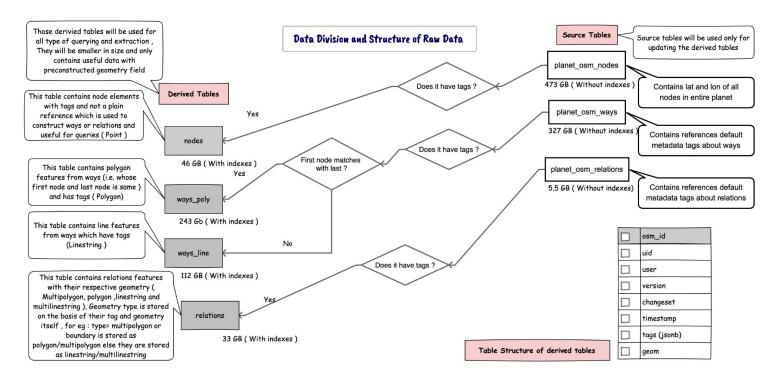
Handling Large spatial data

Tips and Tricks:

- Understanding and organizing data
- Processing raw data
- Choice of database based on need
- Schema
- Managing Indexes
- Query plan
- Scalability



Experiments



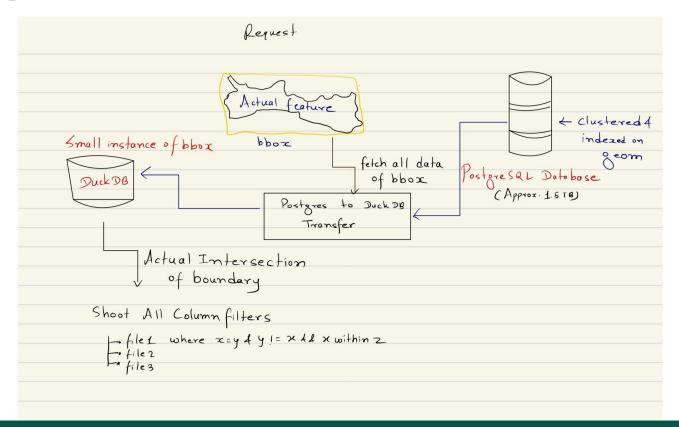


DuckDB Usage

- DuckDB is really useful when you want to do quick spatial operations in large files in your disk without having to setup whole database
- Faster columnar query as compared to postgresql without indexes
- Relatively new and doesn't contain lots of spatial datasets
- Currently focused on vector datasets
- Supports both spatial and non spatial query
- Has support of OGR formats enabling larger GIS format export support directly from the database
- Can be useful on google buildings, overture datasets / your custom csv datasets



PostgreSQL + DuckDB





Walkthrough & Demo

```
▶ 🕸 Insight - localhost:6060
▼ ¶localhost - localhost:5432

▼ Databases

    ▼ ≥ postgres

▼ Im Schemas

         ▼ 🗐 public

▼ In Tables

              ▶ ■ nodes
              ▶ = relations
              ▶ ■ spatial ref sys
             ▶ = users
              ways_line
             123 osm id (int8)
                    123 uid (int4)
                    asc user (text)
                    123 version (int4)
                    123 changeset (int4)
                    timestamp (timestamp)
                    tags (jsonb)
                    geom (public.geometry(polygon, 4326))
                    country (_int4)
                ▶ © Constraints
                ▶ Foreign Keys
                ▼ Indexes
```

```
explain analyze
                 select count(*) from ways poly wp
                 where st intersects(geom, (select geom from relations where osm id=1942586));
       >-
       Results 1 X
636M
       683M
                 COUFRY PLAN
  7M
 24K
3.8G
                 -> Gather (cost=1000.00..34049.42 rows=1 width=3510) (actual time=30.090..62.937 rows=1 loops=1)
                     Workers Planned: 2
  4G
         5
                     Workers Launched: 2
          6
                     -> Parallel Seg Scan on relations (cost=0.00..33049.32 rows=1 width=3510) (actual time=20.540..31.452 rows=0 loops=3)
          7
                        Filter: (osm id = 1942586)
          8
                        Rows Removed by Filter: 63252
                 -> Gather (cost=28788.95..28789.06 rows=1 width=8) (actual time=360.089..368.834 rows=2 loops=1)
          10
                    Workers Planned: 1
          11
                    Params Evaluated: $1
          12
                    Workers Launched: 1
          13
                    -> Partial Aggregate (cost=27788.95..27788.96 rows=1 width=8) (actual time=281.178..281.179 rows=1 loops=2)
                       -> Parallel Bitmap Heap Scan on ways_poly wp (cost=44.02..27786.74 rows=881 width=0) (actual time=16.799..275.260 rows=120447 loops=2)
          14
          15
                          Filter: st intersects(geom, $1)
          16
                          Rows Removed by Filter: 81978
          17
                          Heap Blocks: exact=4857
          18
                          -> Bitmap Index Scan on ways poly geom idx (cost=0.00..43.64 rows=1497 width=0) (actual time=30.070..30.071 rows=404849 loops=1)
          19
                             Index Cond: (geom && $1)
               Planning Time: 0.089 ms
               Execution Time: 368,955 ms
```

Contact Me:





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