

Implementing a Database System with DuckDB for Local Processing and MotherDuck for Scalable Cloud Storage



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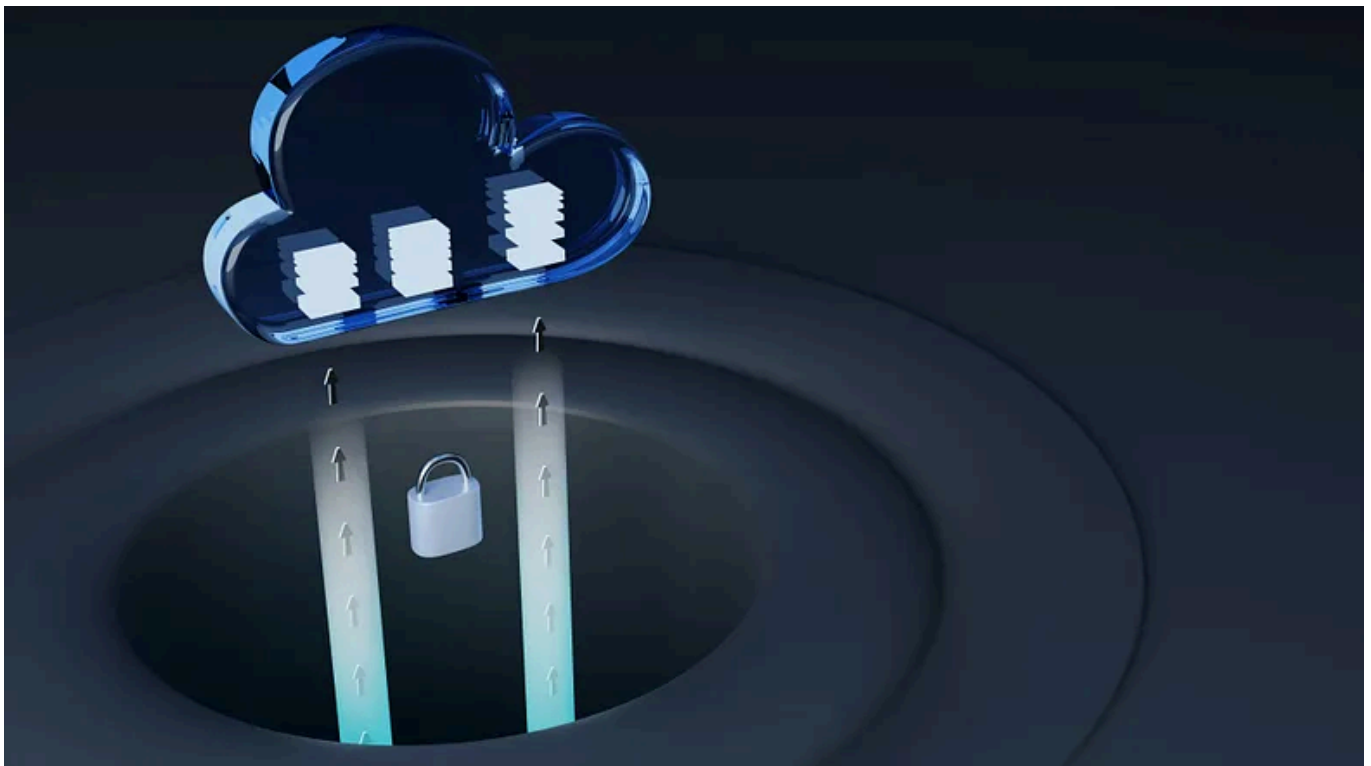


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Overview

This project explores how to leverage the strengths of DuckDB and MotherDuck to build a robust data processing and storage solution. DuckDB excels at fast in-memory analytics, while MotherDuck provides a scalable and cost-effective cloud data warehouse. By combining these technologies, you can achieve optimal performance for both local and cloud-based data operations.

Environment settings

```
import polars as pl
import duckdb as db
import glob
```

Extraction from data sources

- csv files

```
csv_files = glob.glob('./datasets/*.csv')
```

```
list(enumerate(csv_files))
```

```
[(0, './datasets/watercollection.csv'),  
 (1, './datasets/ContainsNull.csv'),  
 (2, './datasets/sales_info.csv'),  
 (3, './datasets/cdmx-subway.csv'),  
 (4, './datasets/airports.csv'),  
 (5, './datasets/colors.csv'),  
 (6, './datasets/sets.csv'),  
 (7, './datasets/appl_stock.csv'),  
 (8, './datasets/sales.csv')]
```

- json files

```
json_files = glob.glob('./datasets/*.json')
```

```
list(enumerate(json_files))
```

```
[(0, './datasets/prevalencia.json'), (1, './datasets/people.json')]
```

- database tables

```
db_files = glob.glob('datasets/*.db')
```

```
list(enumerate(db_files))
```

```
[(0, 'datasets/retail_db.db'), (1, 'datasets/restaurants.db')]
```

Data warehouse creation

```
conn = db.connect('my_database.db')
```

Data warehouse load

```
conn.sql(f"create or replace table water_collection as  
        select * from '{csv_files[0]}' ")
```

```
conn.sql(f"create or replace table contains_null as  
        select * from '{csv_files[1]}' ")
```

```
conn.sql(f"create or replace table sales_info as  
        select * from '{csv_files[2]}' ")
```

```
conn.sql(f"create or replace table cdmx_subway as  
        select * from '{csv_files[3]}' ")
```

```
conn.sql(f"create or replace table airports as  
        select * from '{csv_files[4]}' ")
```

```
conn.sql(f"create or replace table colors as  
select * from '{csv_files[5]}' ")
```

```
conn.sql(f"create or replace table sets as  
select * from '{csv_files[6]}' ")
```

```
conn.sql(f"create or replace table appl_stock as  
select * from '{csv_files[7]}' ")
```

```
conn.sql(f"create or replace table sales as  
select * from '{csv_files[8]}' ")
```

```
conn.sql(f"create or replace table prevalencia as  
select * from '{json_files[0]}' ")
```

```
conn.sql(f"create or replace table people as  
select * from '{json_files[1]}' ")
```

```
retail = db.connect('./datasets/retail_db.db')  
retail.sql('show tables')
```

name varchar
retail_sales

```
retail_sales_pl = retail.sql('select * from retail_sales').pl()
```

```
conn.execute("create or replace table retail_sales as from retail_sales_pl");
```

```
restaurants = db.connect('./datasets/restaurants.db')
restaurants.sql('show tables')
```

name varchar
restaurants

```
restaurants_pl = restaurants.sql('select * from restaurants').pl()
```

```
conn.execute("create or replace table restaurants as from restaurants_pl");
```

Data retrieval

```
conn.sql('show databases')
```

database_name varchar
my_database

```
conn.sql('show tables')
```

name varchar
airports
appl_stock
cdmx_subway
colors
contains_null
people
prevalencia
restaurants
retail_sales
sales
sales_info
sets
water_collection
13 rows

```
conn.sql('select * from restaurants limit 5').pl()
```

name	rating_count	cost	city	cuisine	rating
str	i64	f64	str	str	i64
"The Golden Wok"	1477	33.620488	"Berlin"	"American"	5
"Greek Gyros"	770	68.388874	"New York"	"French"	1
"Taste of Italy"	4420	88.23168	"Amsterdam"	"Chinese"	0
"Midnight Diner"	2155	12.965985	"Lisbon"	"Mexican"	1
"Taste of Italy"	3375	52.785226	"Sydney"	"Chinese"	1

Cloud Data Warehouse with MotherDuck

```
dw = db.connect('md')
```

```
dw.sql('select current_database()').show()
```

current_database() varchar
my_portfolio

Convert queries from local database to polars

```
airports_pl = conn.sql('select * from airports').pl()
appl_stock_pl = conn.sql('select * from appl_stock').pl()
cdmx_subway_pl = conn.sql('select * from cdmx_subway').pl()
colors_pl = conn.sql('select * from colors').pl()
contains_null_pl = conn.sql('select * from contains_null').pl()
people_pl = conn.sql('select * from people').pl()
prevalencia_pl = conn.sql('select * from prevalencia').pl()
restaurants_pl = conn.sql('select * from restaurants').pl()
retail_sales_pl = conn.sql('select * from retail_sales').pl()
sales_pl = conn.sql('select * from sales').pl()
sales_info_pl = conn.sql('select * from sales_info').pl()
sets_pl = conn.sql('select * from sets').pl()
water_collection_pl = conn.sql('select * from water_collection').pl()
```

Upload dataframes to MotherDuck

```
dw.sql(f"create or replace table airports as select * from airports_pl");
```

```
dw.sql(f"create or replace table appl_stock as  
      select * from appl_stock_pl");
```

```
dw.sql(f"create or replace table cdmx_subway as  
      select * from cdmx_subway_pl");
```

```
dw.sql(f"create or replace table colors as select * from colors_pl");
```

```
dw.sql(f"create or replace table contains_null as  
      select * from contains_null_pl");
```

```
dw.sql(f"create or replace table people as select * from people_pl");
```

```
dw.sql(f"create or replace table prevalencia as  
      select * from prevalencia_pl");
```

```
dw.sql(f"create or replace table restaurants as  
      select * from restaurants_pl");
```

```
dw.sql(f"create or replace table retail_sales as  
      select * from retail_sales_pl");
```

```
dw.sql(f"create or replace table sales as  
      select * from sales_pl");
```

```
dw.sql(f"create or replace table sales_info as  
      select * from sales_info_pl");
```

```
dw.sql(f"create or replace table sets as  
      select * from sets_pl");
```

```
dw.sql(f"create or replace table water_collection as  
      select * from water_collection_pl");
```

Check uploaded tables

```
dw.sql('show tables')
```

name varchar
airports
appl_stock
cdmx_subway
colors
contains_null
people
prevalencia
restaurants
retail_sales
sales
sales_info
sets
water_collection
13 rows

Close all database connections

```
# close db connections
conn.close()
retail.close()
restaurants.close()
dw.close()
```

Conclusions

By combining DuckDB's in-memory processing capabilities with MotherDuck's cloud-based data warehousing, you can create a powerful and flexible data

processing and storage solution. This approach allows you to efficiently handle both local and cloud-based data operations, optimize performance,

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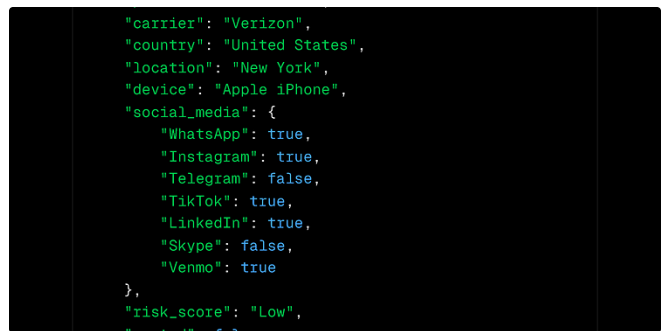
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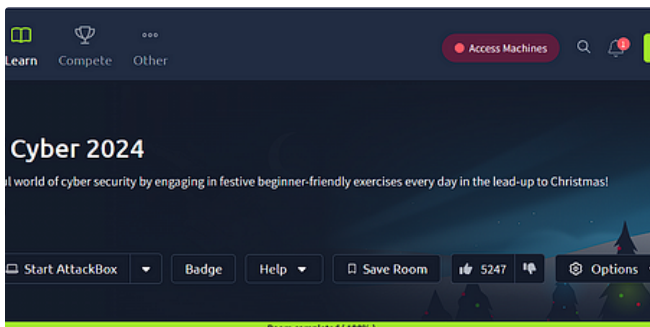
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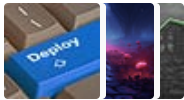
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