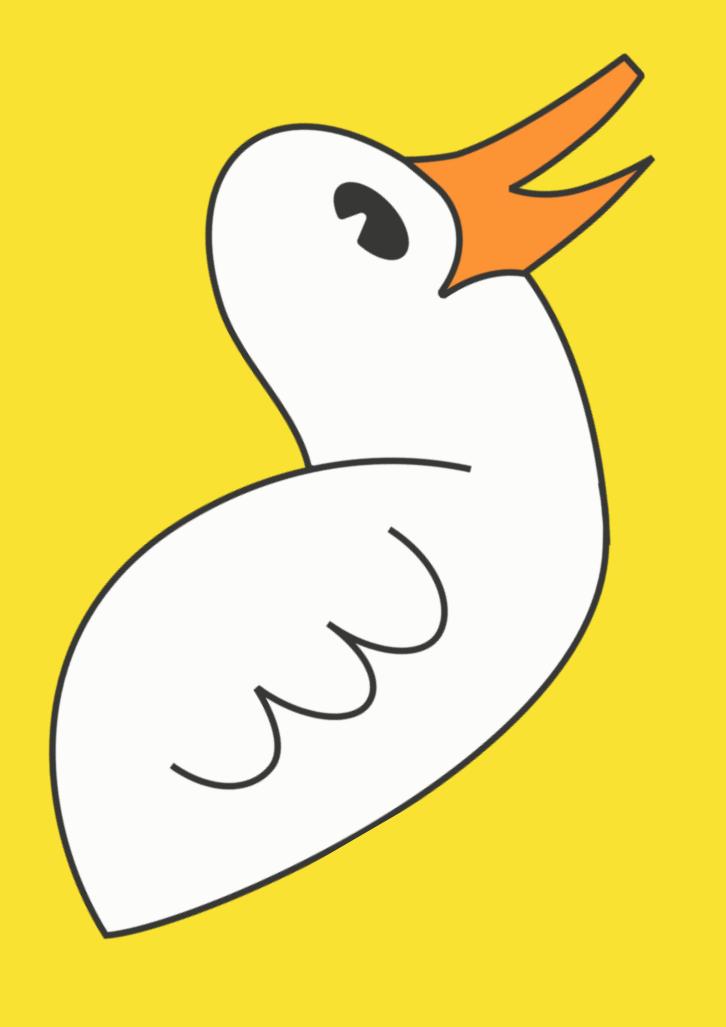
QUACK, QUACK, QUACK, QUACK,





If a Duck Quacks In the Forest, Should You Care?!?

#DuckTalk
#DuckPost
#DuckPoint

Ryan Boyd Co-founder @ MotherDuck @ryguyrg

THIS PRESENTATION IS NOT

A complete tutorial on DuckDB

THIS PRESENTATION <u>DUCKING IS</u>

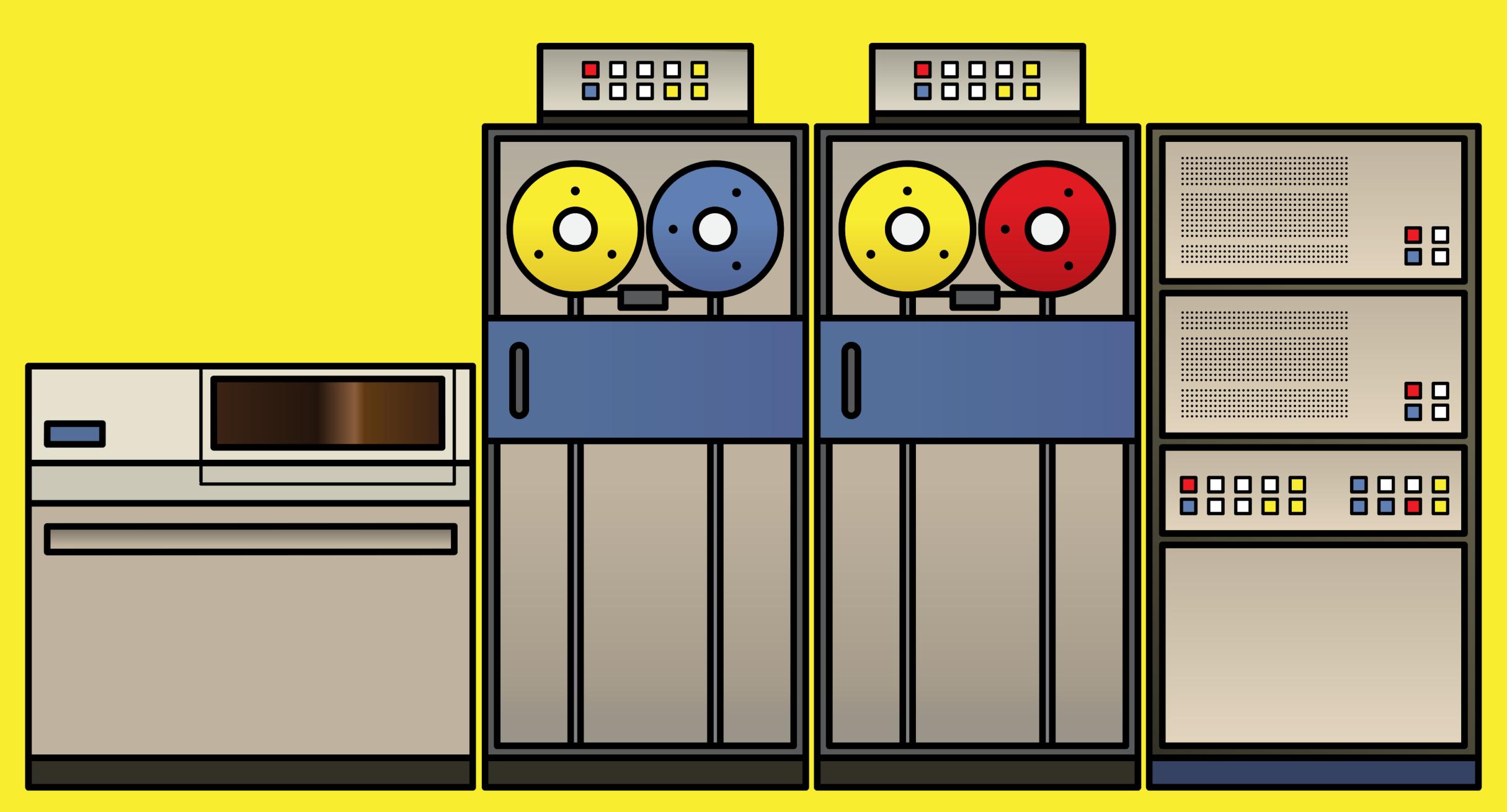
An explanation of the industry context in which DuckDB came to life

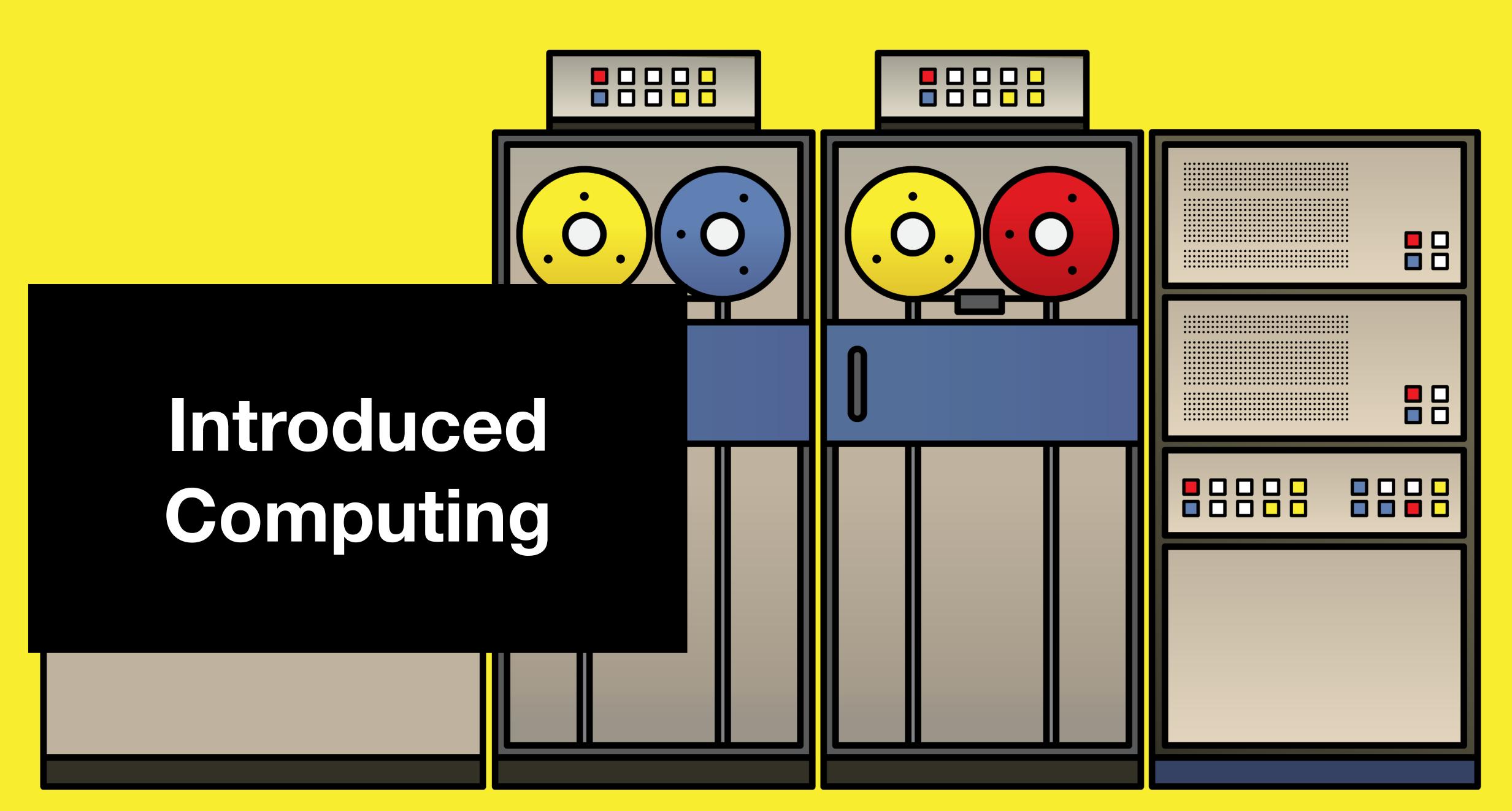
A discussion on OSS DuckDB and why it's awesome

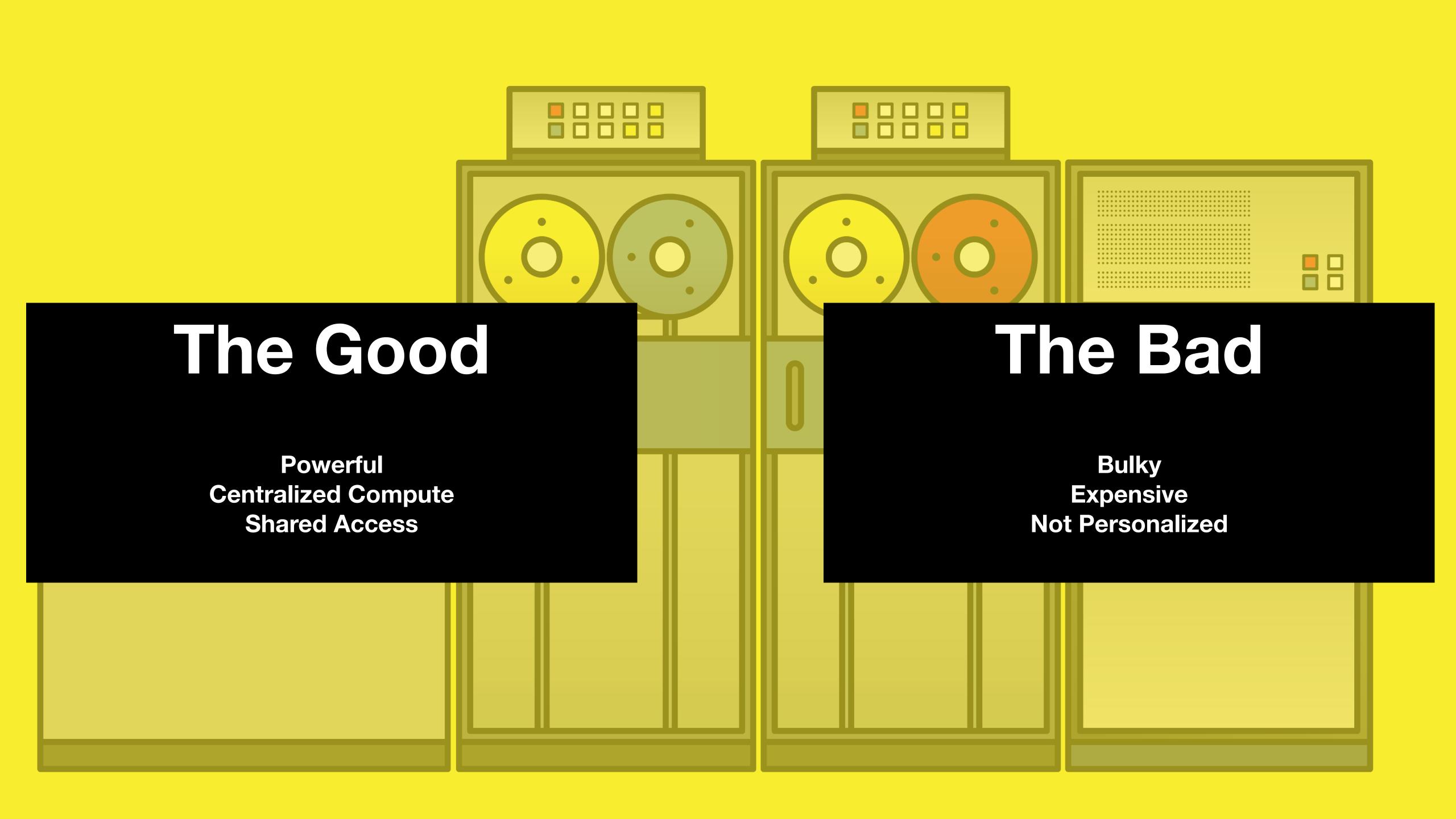
A spotlight on the internals of DuckDB and why

A demonstration of DuckDB's capabilities

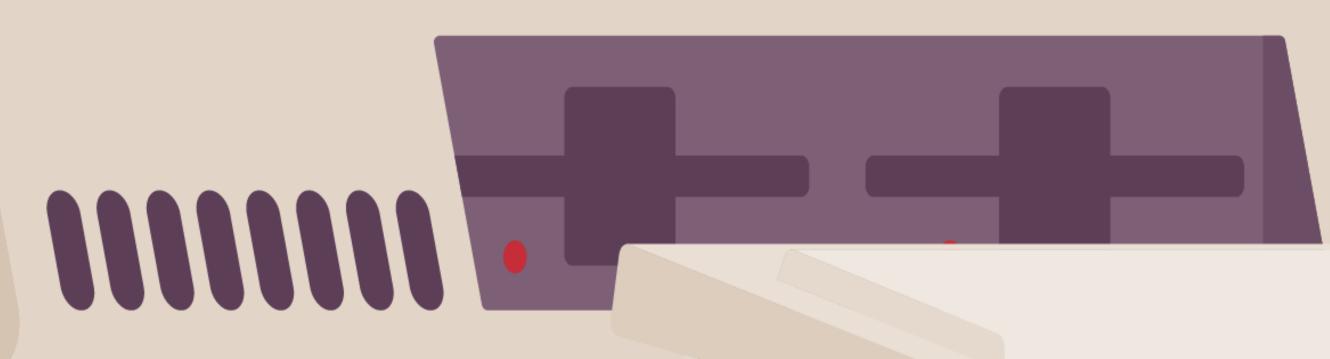
A preview of MotherDuck and its architecture





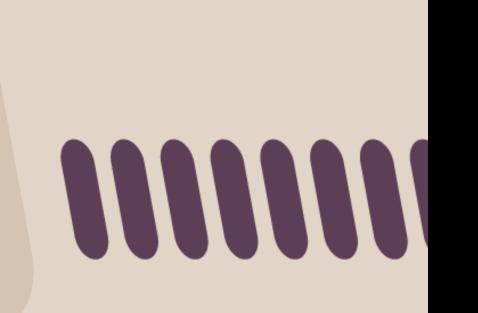






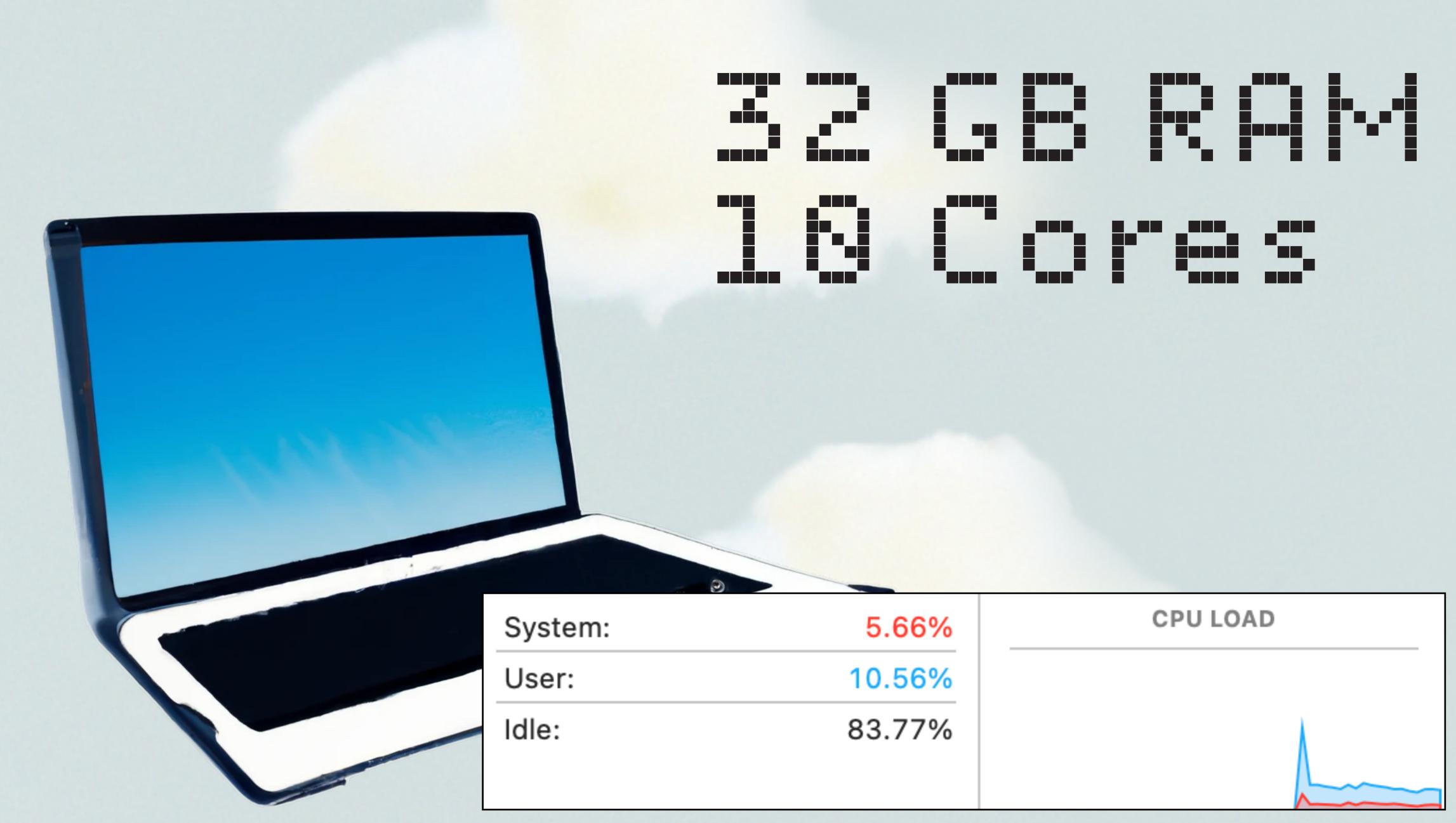
Popularized Computing

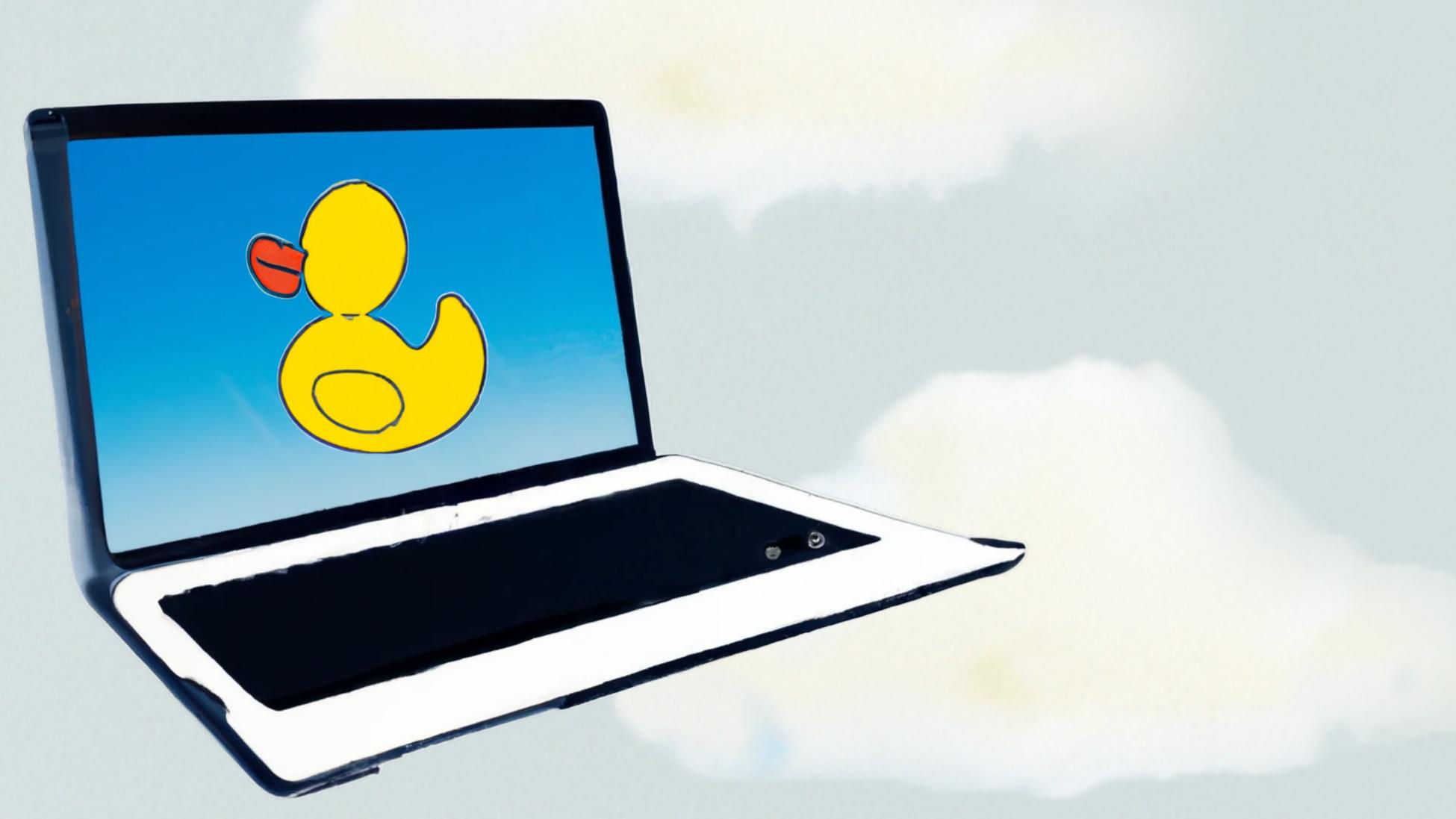




Dedicated Personalized

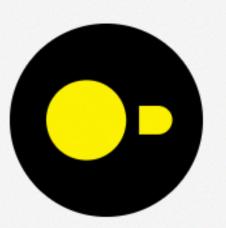






Duck DB





DuckDB is an in-process SQL OLAP database management system



Simple

- In-process, serverless
- C++11, no dependencies, single file build
- APIs for Python/R/Java/...

more →



Feature-rich

- Transactions, persistence
- Extensive SQL support
- Direct Parquet & CSV querying

more →



Fast

- Vectorized engine
- Optimized for analytics
- Parallel query processing



Free

- Free & Open Source
- Permissive MIT License

more →



created at:



created by:



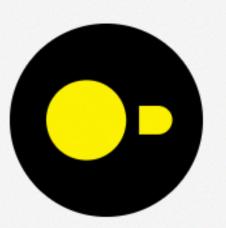
maintained by:





Community & Foundation





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- In-process, serverless
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- Direct Parquet & CSV querying

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- Vectorized engine
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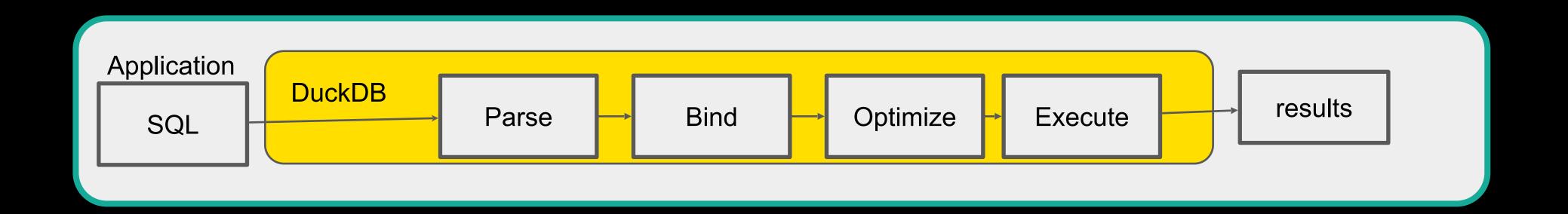
more →

Let's Break that Down

In-Process

Runs inside Python, R, C++.

No external dependencies.



Let's Break that Down

In-Process

Runs inside Python, R, C++.

No external dependencies.

OLAP

Full SQL support - things like window functions

High-performance aggregations



SQLite for Analytics



created at:

crea ea by:

aintamed by:



DuckDB Labs



Community & Foundation

• DuckDB

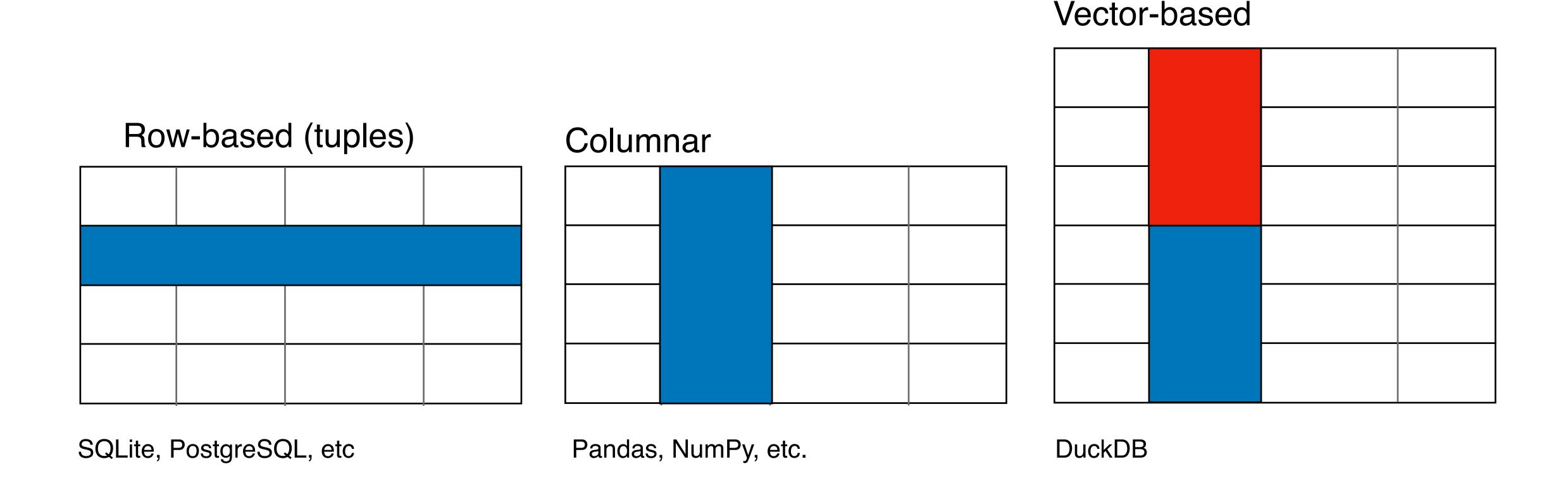
* High Performance Architecture

* Simplified Data Access

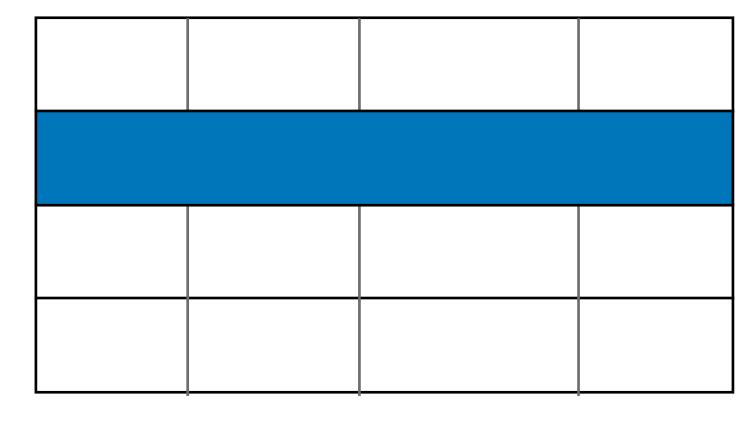
* OSS Community Flocks Together

* Fast-paced innovation

High Performance Architecture



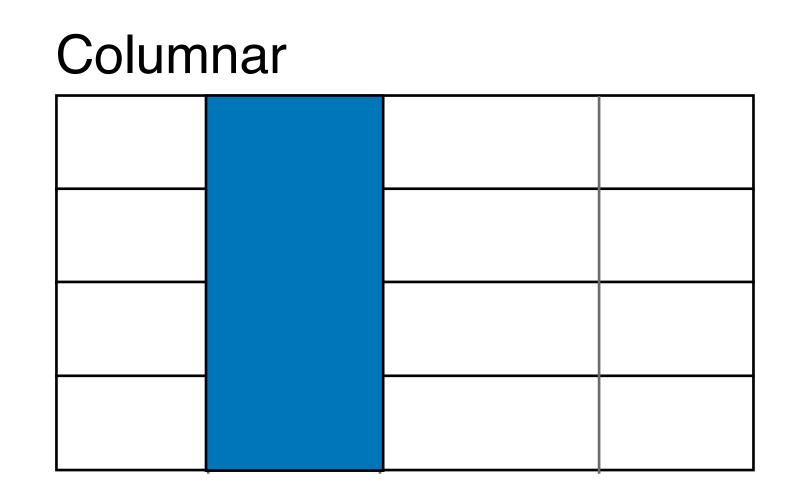
Row-based (tuples)



SQLite, PostgreSQL, etc

Optimized for:

- * low memory footprint
- * transactional workloads

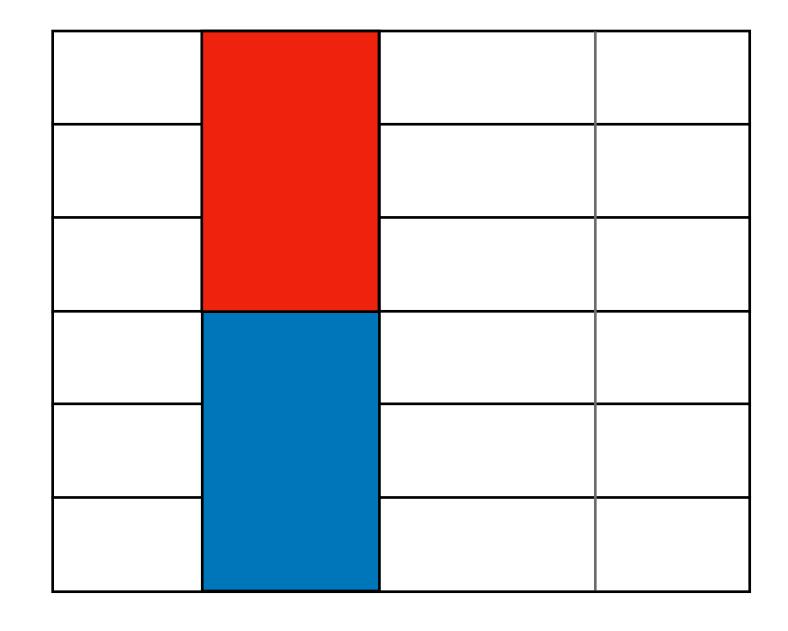


Pandas, NumPy, etc.

Optimized for:

- * analytic workloads
- * aggregations
- * data compression on like data

Vector-based

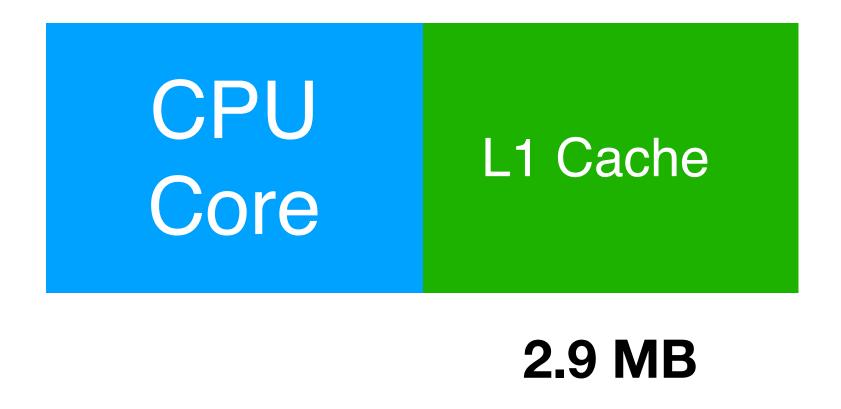


DuckDB

Optimized for:

- * analytic workloads
- * aggregations
- * CPU can do SIMD
- * CPU cache locality

Why Vectorized?



L2 Cache

L3 Cache

28 MB 48 MB

Not just in-memory

* DuckDB has a native storage format

* Supports larger-than-memory queries (spillover to disk)

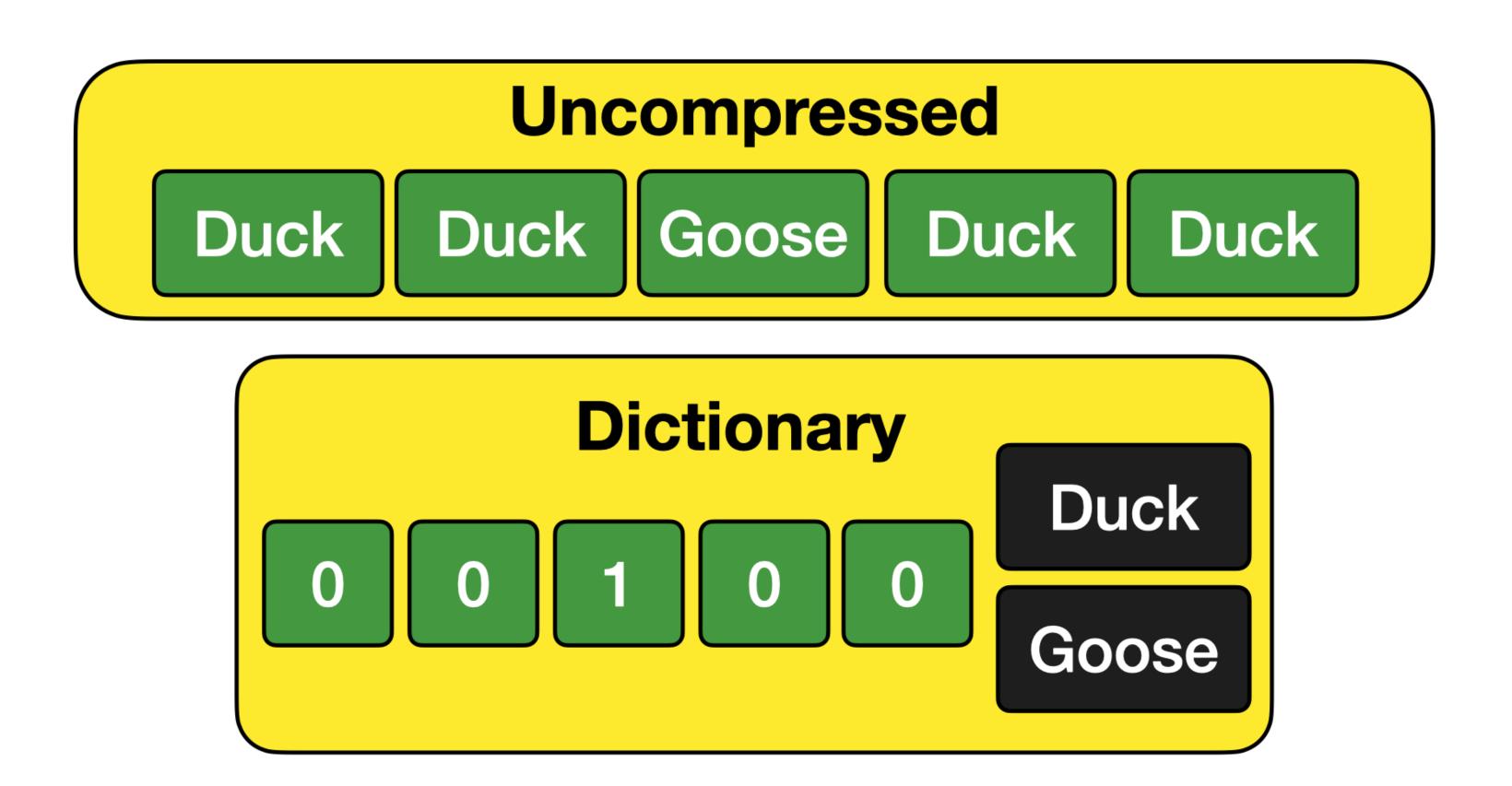
Native storage format

* Columnar, partitioned

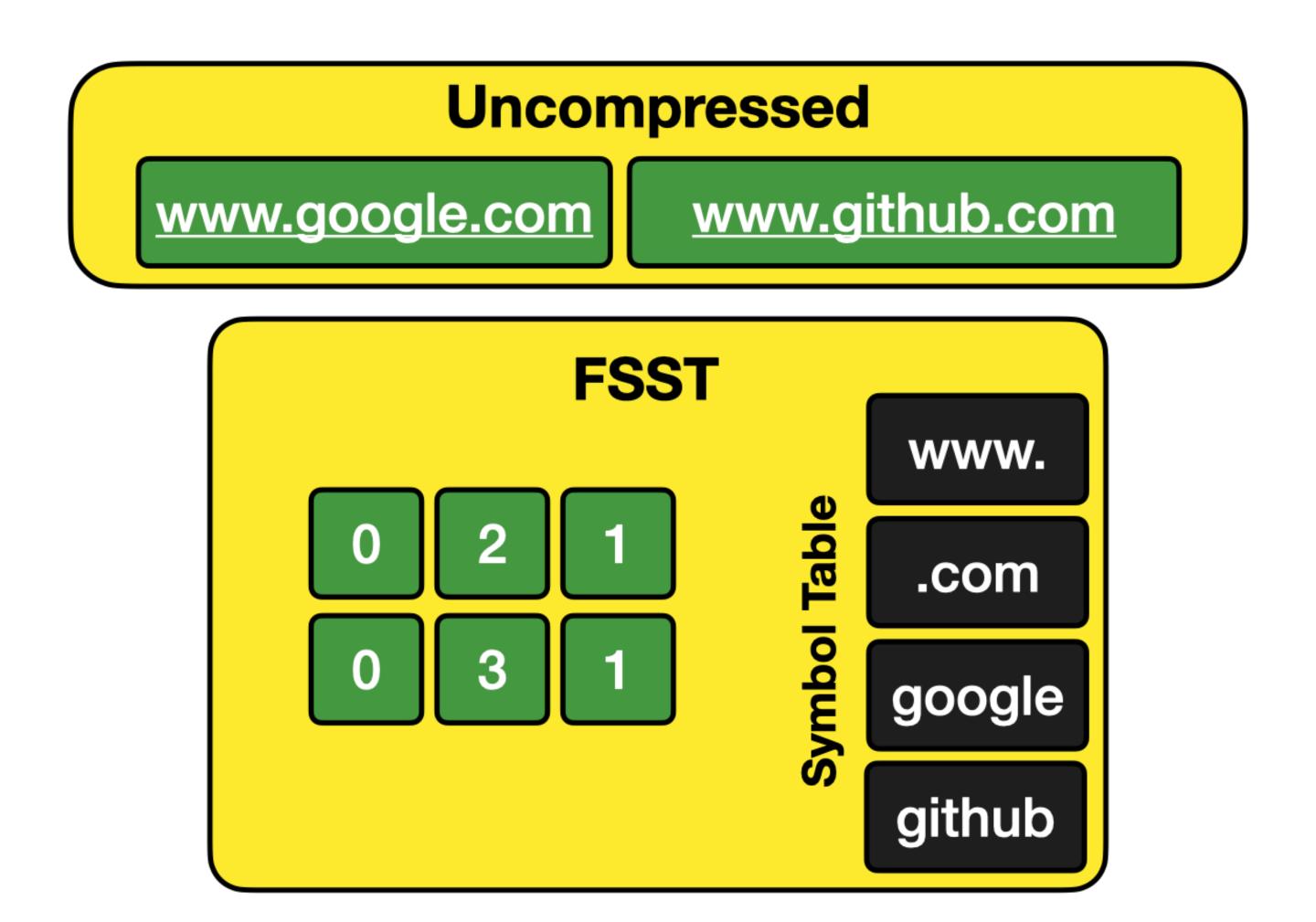
* Efficient ACID-compliant updates

* Stores an entire database, not just tables

Compression algos: dictionary



Compression algos: FSST



Compression algos: FSST

| Version | Taxi | On Time | Lineitem | Notes | Date |
|------------------------|--------|---------|----------|----------------|----------------|
| DuckDB v0.2.8 | 15.3GB | 1.73GB | 0.85GB | Uncompressed | July 2021 |
| DuckDB v0.2.9 | 11.2GB | 1.25GB | 0.79GB | RLE + Constant | September 2021 |
| DuckDB v0.3.2 | 10.8GB | 0.98GB | 0.56GB | Bitpacking | February 2022 |
| DuckDB v0.3.3 | 6.9GB | 0.23GB | 0.32GB | Dictionary | April 2022 |
| DuckDB v0.5.0 | 6.6GB | 0.21GB | 0.29GB | FOR | September 2022 |
| DuckDB dev | 4.8GB | 0.21GB | 0.17GB | FSST + Chimp | NOW() |
| CSV | 17.0GB | 1.11GB | 0.72GB | | |
| Parquet (Uncompressed) | 4.5GB | 0.12GB | 0.31GB | | |
| Parquet (Snappy) | 3.2GB | 0.11GB | 0.18GB | | |
| Parquet (ZSTD) | 2.6GB | 0.08GB | 0.15GB | | |

Compression algos: History

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Compression algos

```
SELECT *
FROM pragma_storage_info('persons')
USING SAMPLE 100 rows
```

Compression algos

| row_group_id int64 | column_name varchar | column_id int64 | segment_type varchar | count int64 | compression varchar |
|-----------------------|------------------------|--------------------|-------------------------|----------------|------------------------|
| 10 | email | 3 | VALIDITY | 122880 | Constant |
| 79 | phone | 5 | VALIDITY | 122880 | Constant |
| 341 | company | 4 | VARCHAR | 31250 | FSST |
| 1007 | email | 3 | VARCHAR | 25355 | FSST |
| 1032 | phone | 5 | VARCHAR | 29526 | FSST |
| 1212 | company | 4 | VARCHAR | 30740 | FSST |
| 1252 | company | 4 | VARCHAR | 30570 | FSST |
| 1575 | id | 0 | BIGINT | 28672 | BitPacking |
| 1595 | last_name | 2 | VARCHAR | 122880 | Dictionary |
| 1733 | id | 0 | VALIDITY | 122880 | Constant |
| 2335 | email | 3 | VARCHAR | 25033 | FSST |

Simplified Data Access

create a sample pandas data frame import pandas as pd

```
test_df = pd.DataFrame.from_dict({"i":[1, 2, 3, 4], "j":["one", "two", "three", "four"]})
```

make this data frame available as a view in duckdb conn.register("test_df", test_df)

print(conn.execute("SELECT j FROM test_df WHERE i > 1").fetchdf())

install and load httpfs

\$ duckdb

D INSTALL httpfs;

D LOAD httpfs;

query as normal

SELECT * FROM 'https://rb-tmp-public.s3.amazonaws.com//persons_1.csv.gz' LIMIT 10

and do magic like

COPY(SELECT * FROM 'https://rb-tmp-public.s3.amazonaws.com//persons_1.csv.gz') TO 'persons_from_csv.parquet' (FORMAT PARQUET)

<bucket>/<file>');

```
# install and load httpfs
$ duckdb
D INSTALL httpfs;
D LOAD httpfs;
# set S3 creds
SET s3_region='us-east-1';
SET s3_access_key_id='AKIA42DX...BV';
SET s3_secret_access_key='w7oOQ60hyGAh...VGuQ6';
# query as normal
SELECT * FROM read_parquet('s3://<bucket>/<file>');
# or copy data over
CREATE TABLE trivia AS SELECT * FROM read_parquet('s3://
```



```
# install and load postgresql
$ duckdb
INSTALL postgres;
LOAD postgres;
# attach postgresql tables as read-only views
CALL postgres_attach('dbname=postgres');
# query as normal
SELECT AVG(id)
FROM persons
```

or query both DuckDB data and postgresql
or use the awesomeness of parquet creation
COPY(SELECT * FROM postgres_scan('dbname=postgres', 'public', 'persons')) TO 'persons.parquet' (FORMAT PARQUET);

Simplified Data Access

Parquet, CSV, SQLite, PostgreSQL, Arrow Local, Remote on S3

In-process, Command-line, In-browser, ODBC/JDBC

Simplified Data Access

Parquet, CSV, SQLite, PostgreSQL, Arrow Local, Remote on S3

In-process, Command-line, In-browser, ODBC/JDBC

For "normal size" data



David is at data-folks.masto.host



@DSJayatillake

The vast majority of orgs in the world do not have > billion row datasets, but many need the benefits of data.

This is partly why I'm so excited about @duckdb; it affords the power of an expensive cloud dwh without the complexity, for smaller data for ALL.

8:10 AM · 17 Apr, 2022

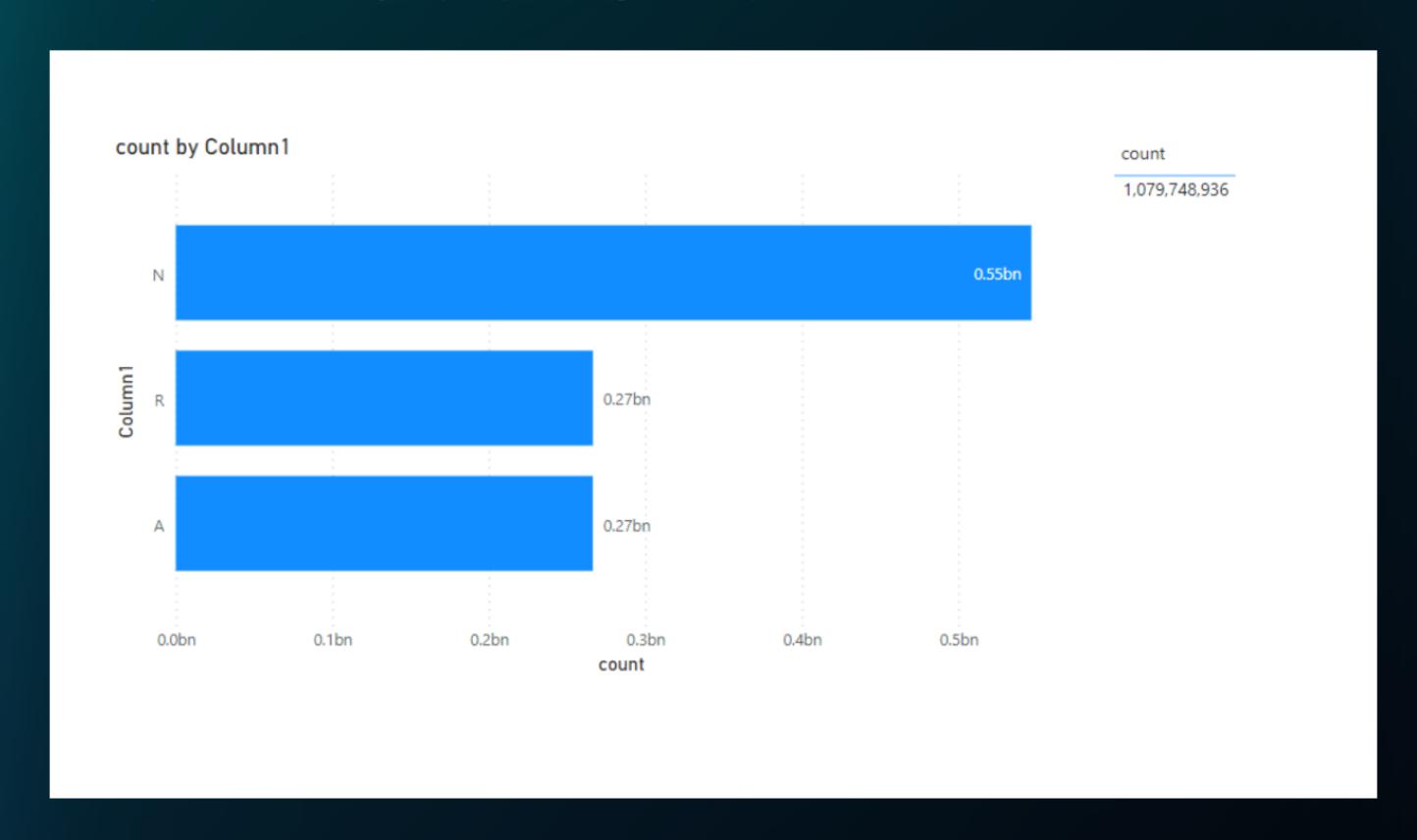
2 replies 12 likes

And larger than normal size data





the irony of life, I never done a Query on a Billion row Table, but when I did it, it was on my laptop using an in-process Database #DuckDB



5:57 AM · 17 Apr, 2022

2 replies 6 shares 43 likes



\$ duckdb

D load parquet; D SELECT AVG(trip_distance) FROM 'yellow_tripdata_20[12]*.parquet';

avg(trip_distance)

6.1801595481684

2.95s querying 1.45 billion rows No indexes

\$ duckdb taxis.ddb

D SELECT AVG(trip_distance) FROM yellow_tripdata;

avg(trip_distance)

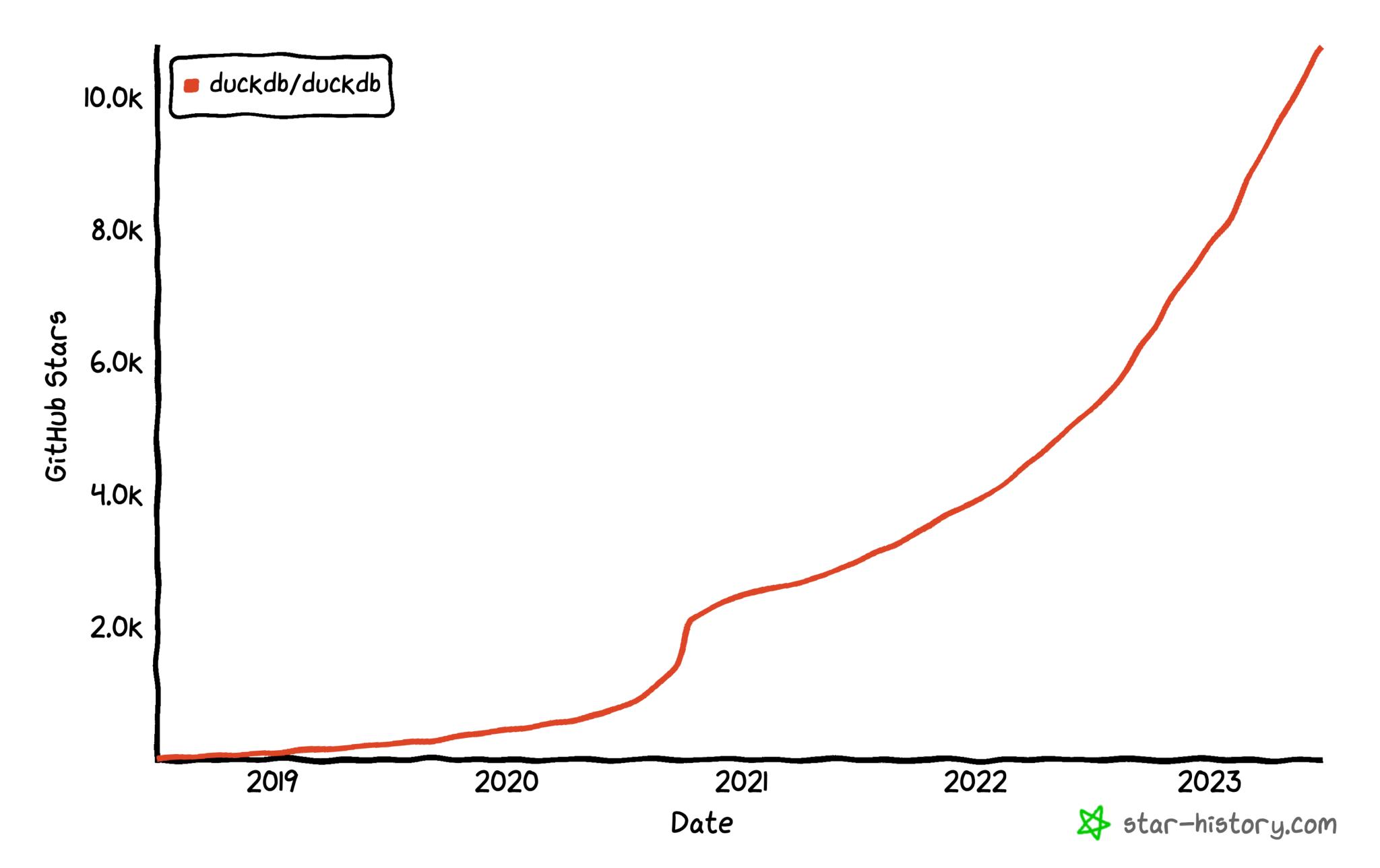
6.1801595481684

1.47s querying 1.45 billion rows No indexes

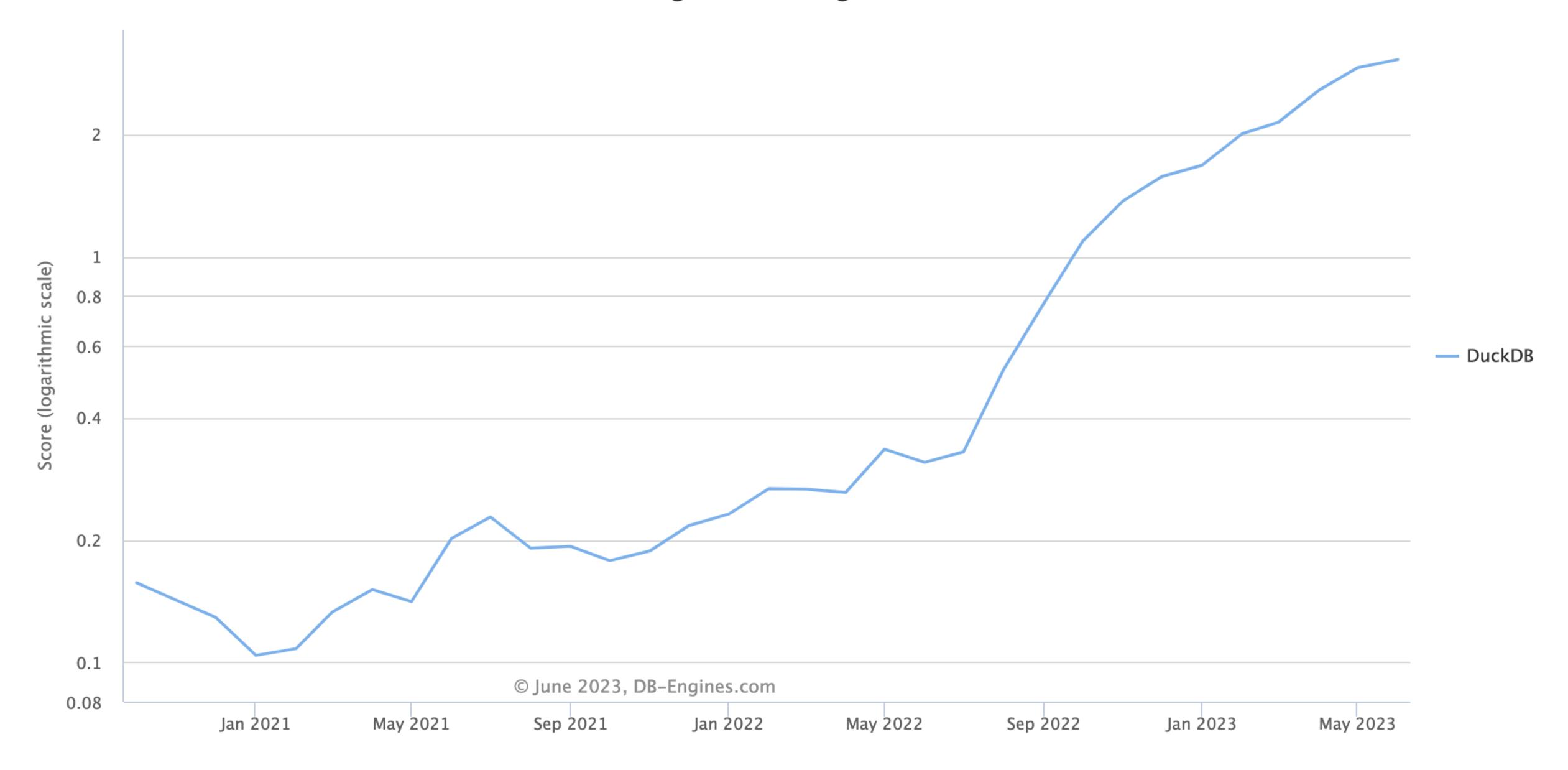
| m6i.4xlarge | 16 cores, 64GB RAM | 1.020s |
|--------------|----------------------|---------------|
| m6i.8xlarge | 32 cores, 128GB RAM | 0.509s |
| m6i.16xlarge | 64 cores, 256GB RAM | 0.268s |
| m6i.32xlarge | 128 cores, 512GB RAM | 0.145s |

OSS Community Flocks Together

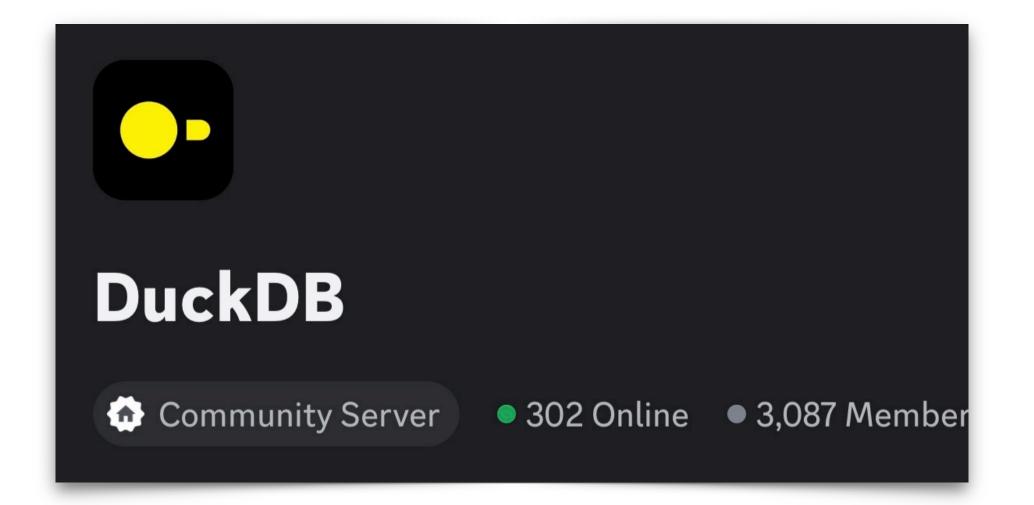




DB-Engines Ranking of DuckDB



Contributors 216 + 205 contributors



DuckDB Ecosystem Monthly

Happy new year, friend 🤲



Hi, I'm Marcos! I'm a data engineer by day at X-Team, working for Riot Games. By night, I create newsletters for a few topics I'm passionate about: helping folks find data digs and AWS graviton. After getting involved in the DuckDB community, I saw a great opportunity to partner with the MotherDuck team to share all the amazing things happening in the DuckDB ecosystem.

In this first issue of the year 2023, we wanted to share some of the incredible stuff coming out of the global DuckDB community.

-Marcos

Feedback: <u>duckdbnews@motherduck.com</u>

Featured Community Members



Jacob Matson

Jacob is the writer of the Modern Data Stack in a Box with DuckDB. A fast, free, and open-source Modern Data Stack (MDS) can now be fully deployed on your laptop or to a single machine using the combination of DuckDB, Meltano, dbt, and Apache Superset.

He is working today as the VP of Finance & Operations at Simetric, bringing IoT connectivity data into a single pane-of-glass. He also does SMB analytics consulting via his agency, Elliot Point LLC. You can find him on Twitter @matsonj

Fast-paced Innovation

Optimistic writing to disk [assuming successful COMMIT]

Parallel data loading [91.4s -> 17.2s for 150 million rows!]

Three new compression algos: FSST, Chimp, Patas

Parallel

- * CSV reading
- * Index creation
- * COUNT(DISTINCT)

SELECT clause now optional

List comprehension support [x + 1 for x in [1, 2, 3]]

JSON ingestion via read_json

Partitioned export of CSV and Parquet

Parallel CSV and Parquet writing

Multi-database ATTACH support

SQLite backend support

Positional JOIN support

Upsert support

Better Python APIs

User-defined scalar functions for Python

Support for ADBC

Swift API

Parallel JSON writing

PIVOT and UNPIVOT

Lazy-loading table metadata



created at:



created by:



maintained by:





Community & Foundation



But who the Duck is MotherDuck?



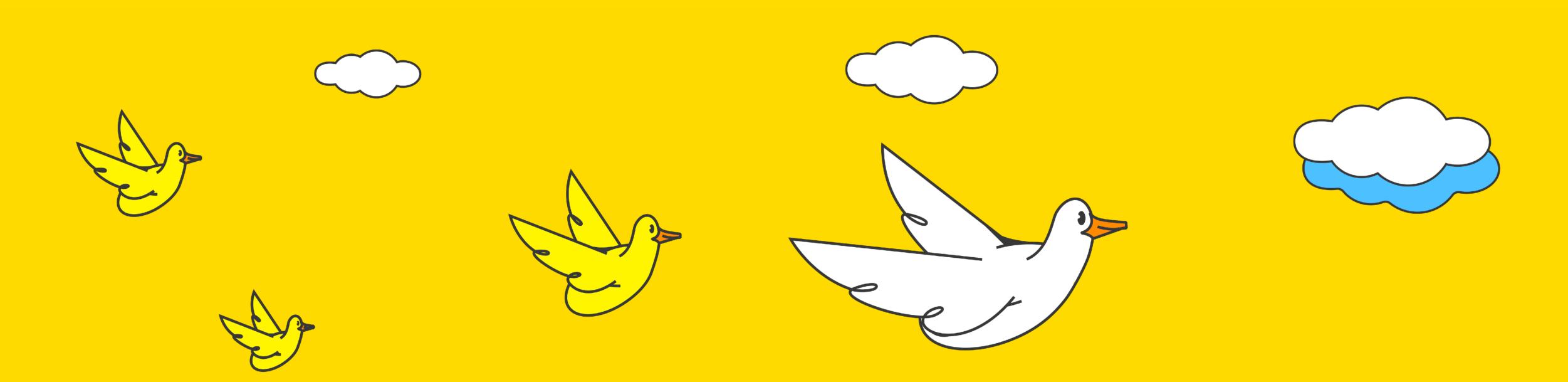
s: venture-backed startup

made up of:

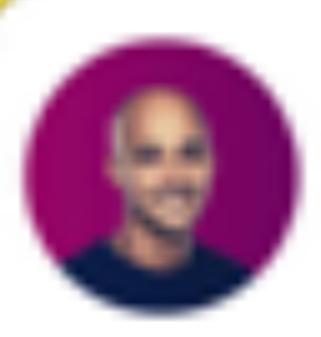
data geeks from Google BigQuery, Databricks, Snowflake, Meta, Elastic, SingleStore, ++

doing:





TEACH YOUR DUCKDB TO FLY



Marc Lamberti

@marclambertiml

Just an honest question, wasn't the purpose of DuckDB to run fast analytical queries in LOCAL? What's the need for a cloud version?

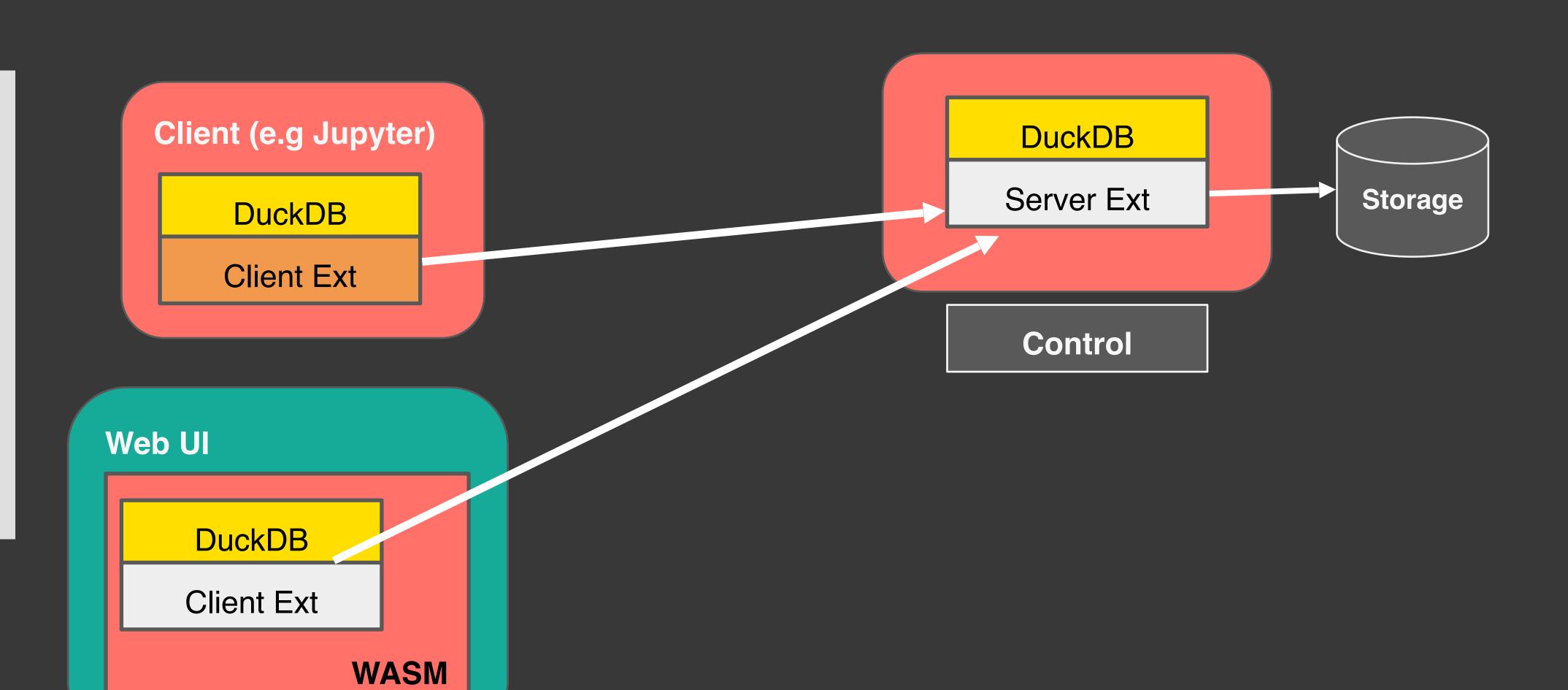
Our Beliefs @ Mother Duck



What if query plans could decide whether to bring the compute to the data, or the data to the compute?

LET'S GET REAL

Client APIs
 Overview
 Python
 R
 Java
 Julia
 C
 C++
 Node.js
 WASM
 ODBC
 CLI



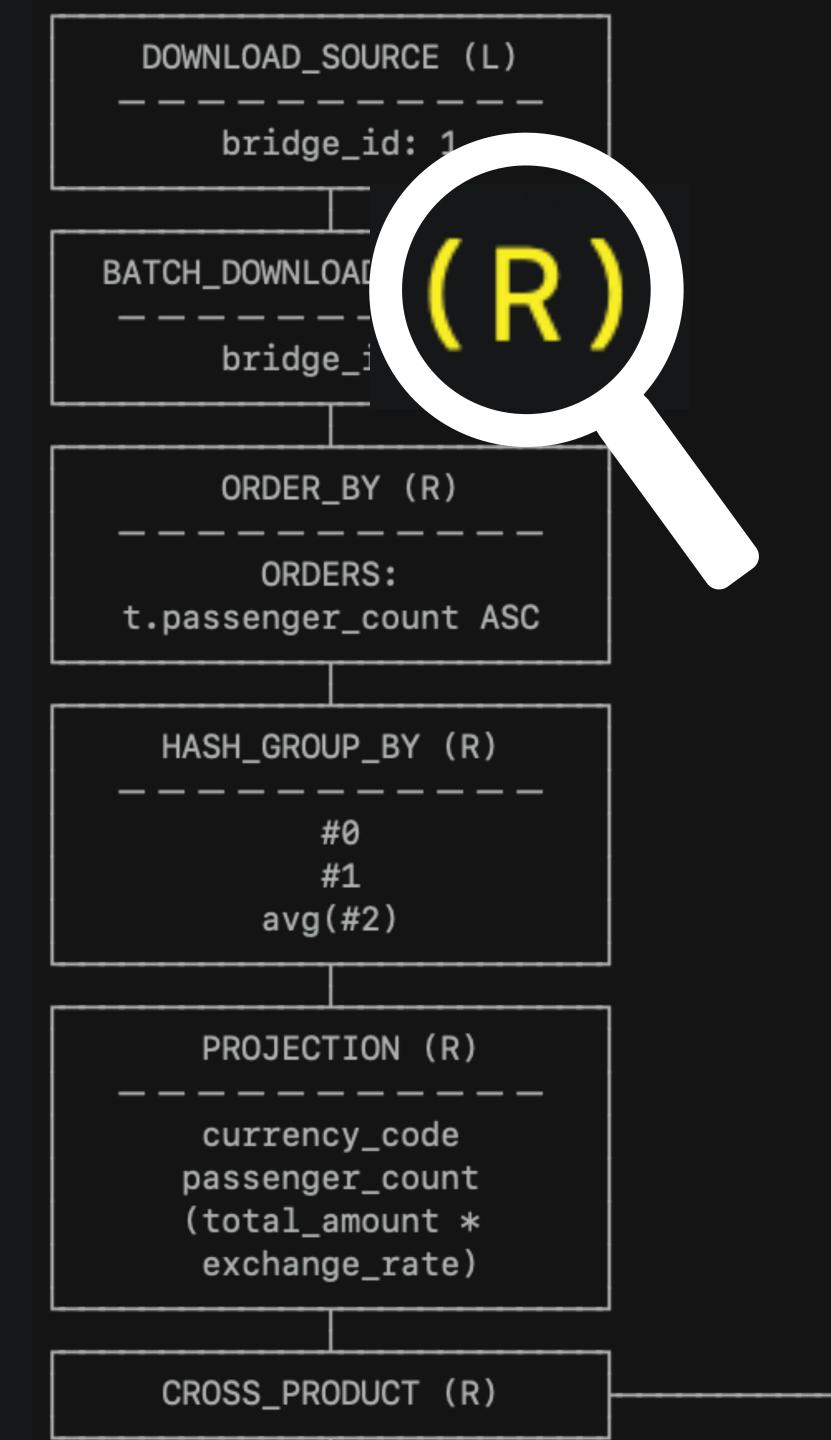
```
SELECT
   cr.currency_code,
   t.passenger_count,
   AVG(t.total_amount * cr.exchange_rate) as average_converted_amount
FROM
   sample_data.nyc.yellow_cab_nyc_2022_11 t
CROSS JOIN
   (SELECT * FROM './popular_currency_rate_dollar_20230620.csv') cr
WHERE cr.currency_code = 'EUR'
GROUP BY
   cr.currency_code, t.passenger_count
ORDER by t.passenger_count ASC;
```

```
SELECT
   cr.currency_code,
   t.passenger_count,
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```

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CROSS JOIN
   (SELECT * FROM './popular_currency_rate_dollar_20230620.csv') cr
WHERE cr.currency_code = 'EUR'
GROUP BY
   cr.currency_code, t.passenger_count
ORDER by t.passenger_count ASC;
```

```
DOWNLOAD_SOURCE (L)
     bridge_id: 1
BATCH_DOWNLOAD_SINK (R)
     bridge_id: 1
     ORDER_BY (R)
       ORDERS:
t.passenger_count ASC
  HASH_GROUP_BY (R)
          #0
          #1
       avg(#2)
    PROJECTION (R)
    currency_code
    passenger_count
    (total_amount *
    exchange_rate)
  CROSS_PRODUCT (R)
```

| SEQ_SCAN (R) | UPLOAD_SOURCE (R) |
|---------------------------------|-------------------------|
| yellow_cab_nyc_2022_11 | bridge_id: 2 |
| passenger_count total_amount | |
| EC: 3252717 | |
| | |
| | UPLOAD_SINK (L) |
| | bridge_id: 2 |
| | FILTER (L) |
| | (currency_code = 'EUR') |
| | EC: 18 |
| | |
| | READ_CSV_AUTO (L) |
| | LOCAL |
| | currency_code |
| | exchange_rate |
| | EC: 0 |



```
SEQ_SCAN (R)
                               UPLOAD_SOURCE (R)
yellow_cab_nyc_2022_11
                                  bridge_id: 2
   passenger_count
     total_amount
     EC: 3252717
                                UPLOAD_SINK (L)
                                  bridge_id: 2
                                   FILTER (L)
                            (currency_code = 'EUR')
                                     EC: 18
                               READ_CSV_AUTO (L)
                                     LOCAL
                                 currency_code
                                 exchange_rate
                                     EC: 0
```

```
DOWNLOAD_SOURCE (L)
     bridge_id: 1
BATCH_DOWNLOAD_SINK (R)
     bridge_id: 1
     ORDER_BY (R)
       ORDERS:
t.passenger_count ASC
  HASH_GROUP_BY (R)
          #0
       avg(#2)
    PROJECTION (R)
    currency_code
   passenger_count
    (total_amount *
    exchange_rate)
  CROSS_PRODUCT (R)
```

| UPLOAD_SOURCE (R) |
|--------------------------------|
| bridge_id: 2 |
| |
| |
| LIDLOAD CTAIK (L) |
| UPLOAD_SINK (L) |
| bridge_id: 2 |
| FILTE (L) |
| (currency_co |
| EC 18 |
| READ_CSV_AUTO (L) |
| LOCAL |
| currency_code exchange_rate |
| EC: 0 |
| |

| currency_code varchar | passenger_count double | average_converted_amount double |
|--------------------------|---------------------------|------------------------------------|
| EUR | 0.0 | 18.546608022179736 |
| EUR | 1.0 | 19.4198124783772 |
| EUR | 2.0 | 22.04667298565129 |
| EUR | 3.0 | 21.236435271048638 |
| EUR | 4.0 | 22.20329272449214 |
| EUR | 5.0 | 19.238401343401563 |
| EUR | 6.0 | 19.688502601081026 |
| EUR | 7.0 | 69.78117721000001 |
| EUR | 8.0 | 56.326965400000006 |
| EUR | 9.0 | 69.24747212666666 |
| EUR | | 26.81066096388576 |
| 11 rows | | 3 columns |





+ ADD FILES

My Databases

- > **l** foo
- > | localmemdb
- ✓ sample_data

∨ 庙 hn

- > \begin{align*} \hat{\text{hacker_news}}
- main main
- ✓
 ☐ nyc
 - > fhvhv_tripdata_nyc_2022_11
 - > == service_requests_311_fro...
 - > \equiv yellow_cab_nyc_2022_11
- - > **ambient_air_quality**

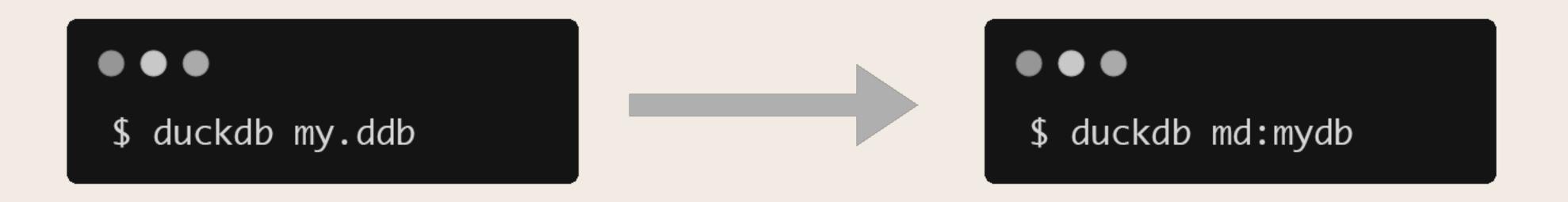
```
MONTH(timestamp) AS month,
            ROW_NUMBER()
                OVER (PARTITION BY YEAR(timestamp), MONTH(timestamp) ORDER BY score DESC)
            AS rn
10
        FROM sample_data.hn.hacker_news
11
        WHERE type = 'story'
12
13
14
    SELECT
16
        year,
17
        month,
        title,
18
        hn_url,
19
20
        score
    FROM ranked_stories
    WHERE rn = 1
    ORDER BY year, month;
```

■ Query executed in 640 ms. Row count: 11

| i₹ | year | month | title | hn_url | score |
|----|-------|-------|--|---|-----------------|
| | | 0 12 | | | 2.0 1,500 4, |
| | 2,022 | 1 | My First Impressions of Web3 | https://news.ycombinator.com/item?id=2984 | |
| | 2,022 | 2 | Google Search Is Dying | https://news.ycombinator.com/item?id=3034 | |
| | 2,022 | 3 | US Senate votes unanimously to make daylig | https://news.ycombinator.com/item?id=3068 | |
| | 2,022 | 4 | Elon Musk makes \$43B unsolicited bid to tak | https://news.ycombinator.com/item?id=3102 | |
| | 2,022 | 5 | Mechanical Watch | https://news.ycombinator.com/item?id=3126 | |
| | 2,022 | 6 | Supreme Court Overturns Roe vs. Wade | https://news.ycombinator.com/item?id=3186 | |
| | 2,022 | 7 | Ask HN: What are some cool but obscure dat | https://news.ycombinator.com/item?id=3218 | |
| | 2 222 | | | // // // // // // // // // // // // | |

Mother Duck Features

- Git-like collaboration: sharing snapshots
- If it can Duck, it can Mother Duck









ASTRONOMER









#PONDER













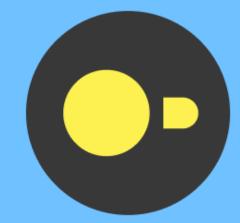


LET'S GET REAL

\$ duckdb my.db md:mydb

-- OAuth Loop for credentials

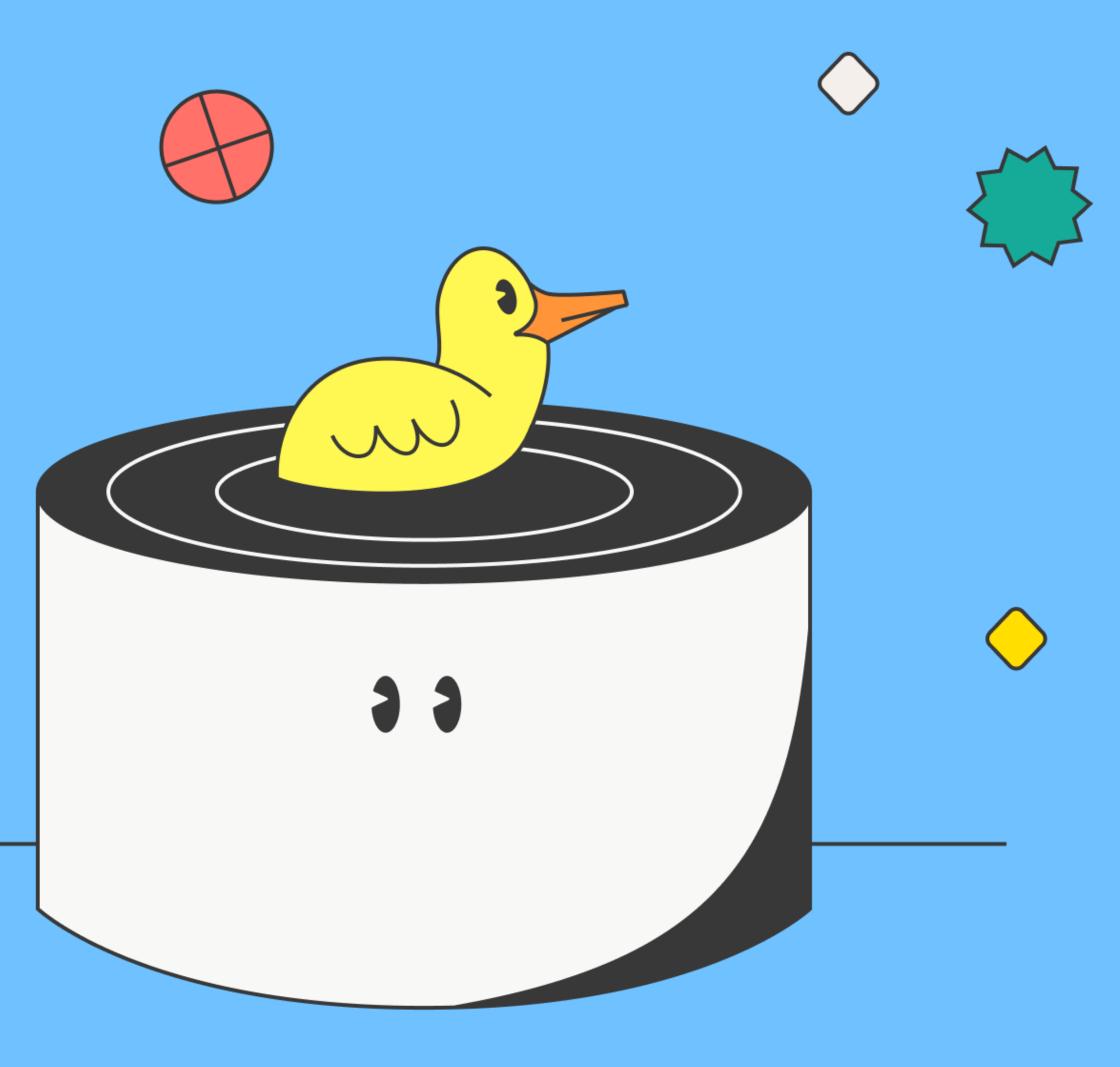
- -- Create remote database and table D CREATE DATABASE db1
- D CREATE TABLE t1 as select 'abc' as x
- -- Open local database and join to remote
- D ATTACH local.db as L
- D SELECT * from db1.t1
 JOIN L.t1 on (id)

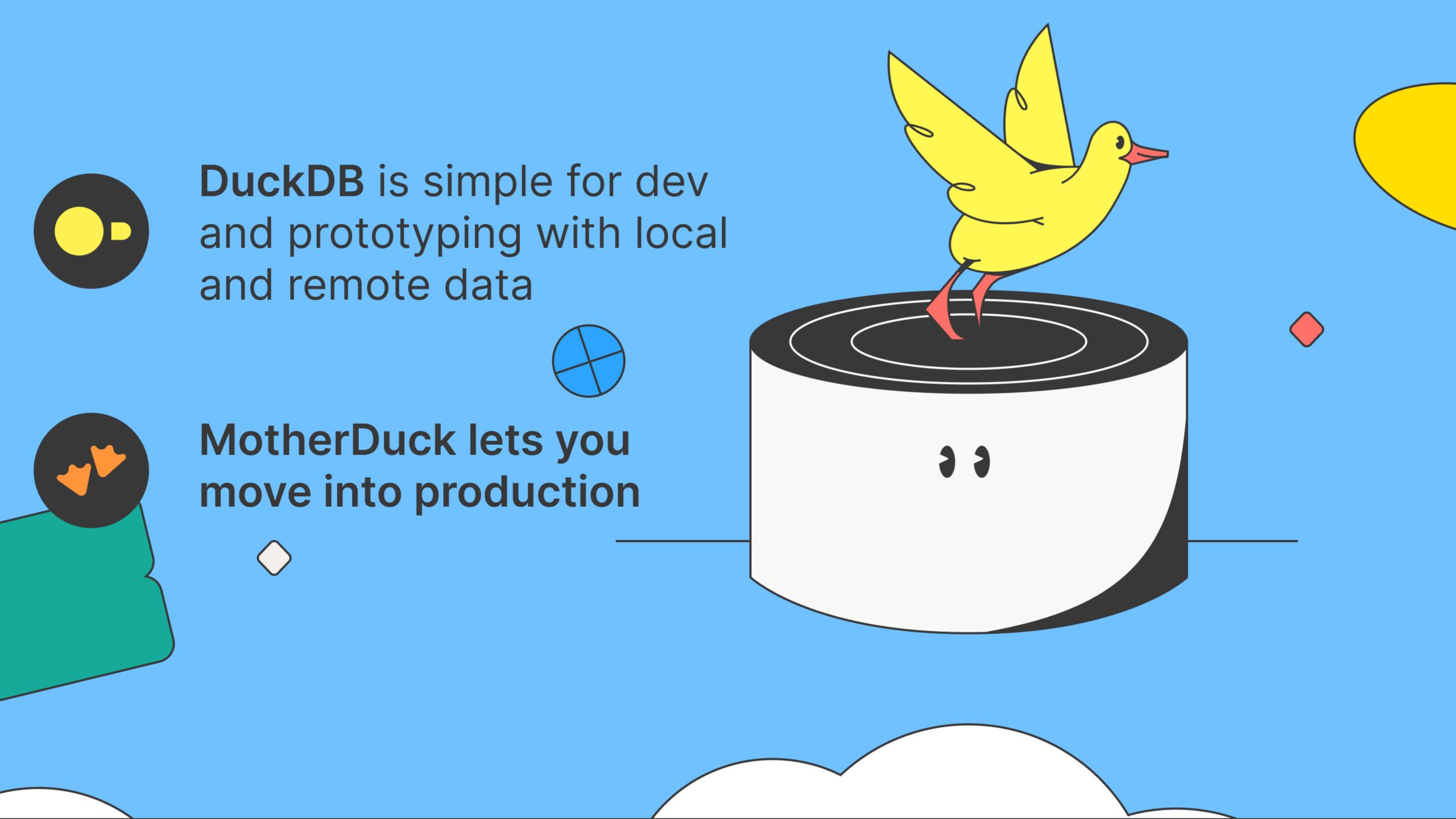


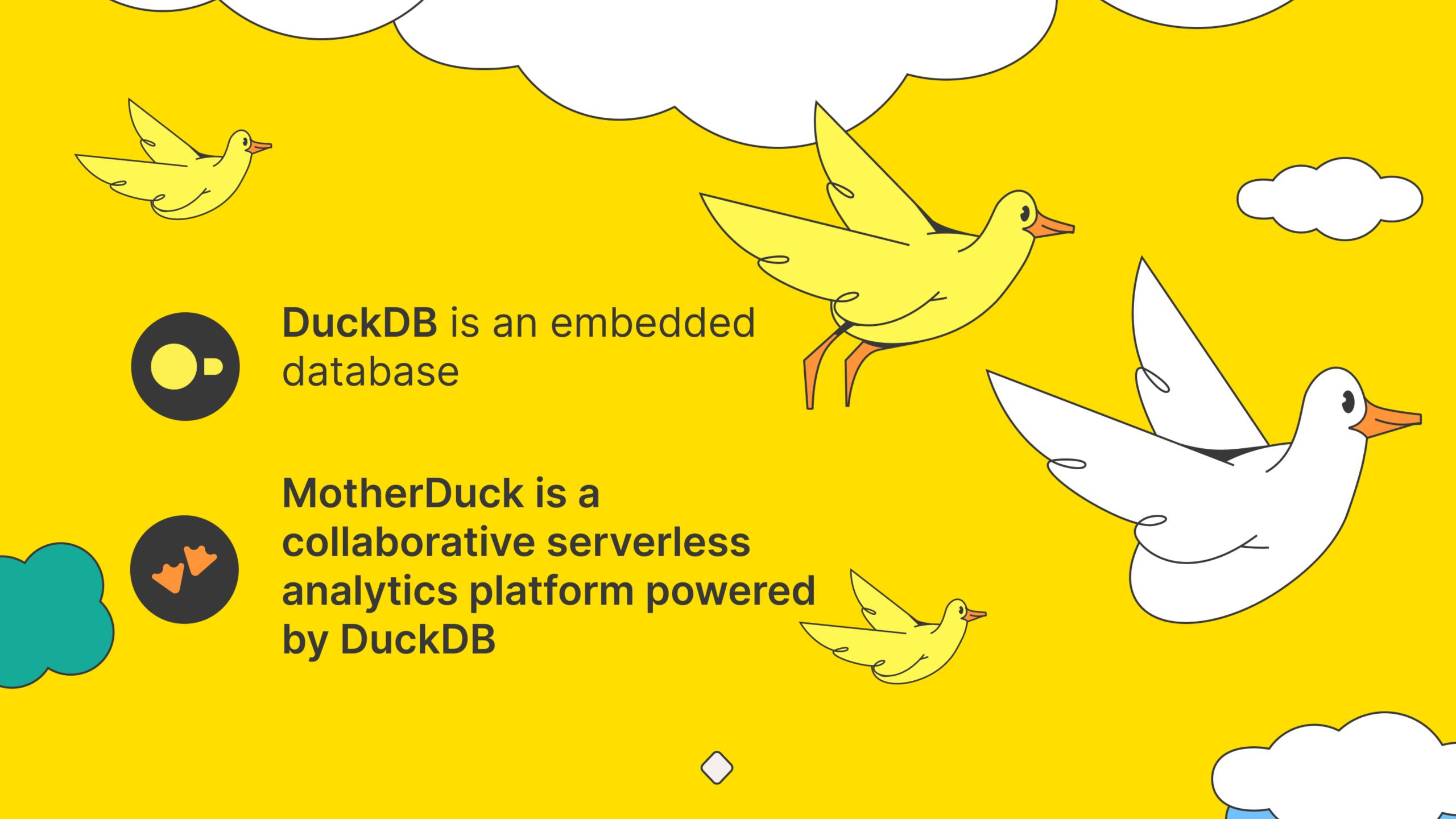
DuckDB turns your laptop into a personal analytics engine



MotherDuck scales your laptop into the cloud with Hybrid Execution



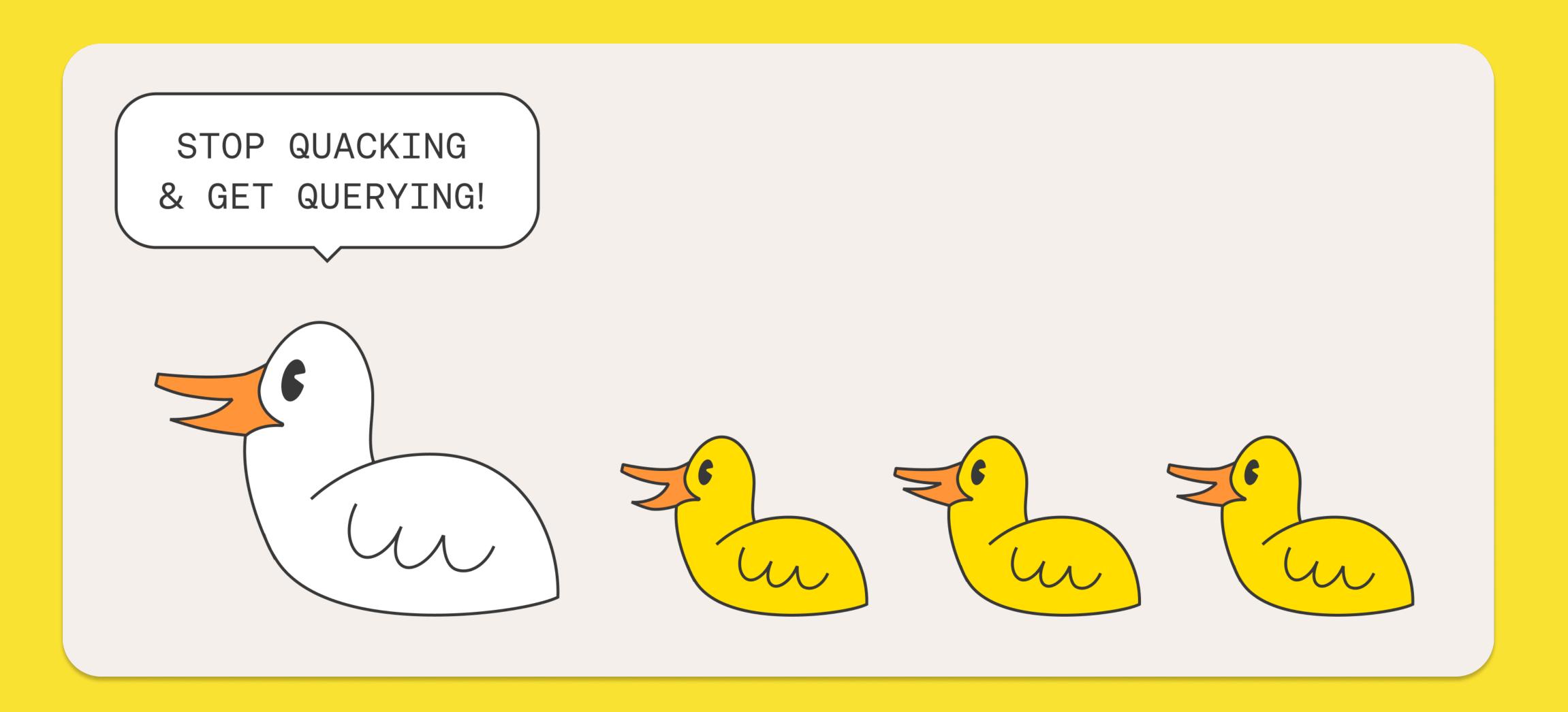




motherduck.com

DUCKCON on THURSDAY

bit.ly/duckcon-sf



Ryan Boyd Co-founder @ MotherDuck @ryguyrg

Our Beliefs @ Mother Duck



BIG DATA IS EASY DATA BIG DATA IS EASY DATA

- 1. What is Big Data really?
- 2. How much data do people have?
- 3. How much data do they query?

BIG DATA IS EASY DATA BLONG LIVE EASY

DEFINITION OF BIG DATA

- More than can fit in Excel
- More than can fit on your laptop
- More than can fit in memory
- More than can fit on disk



DEFINITION OF BIG DATA

In 2012, when I worked on BigQuery:

Largest EC2 instance was 60.5GB of RAM

Today:

Largest EC2 instance is 25TB of RAM

400X

BIG DATA IS EASY DATA
BLONG LIVE EASY

DEFINITION OF BIG DATA

In 2012, when I worked on BigQuery:

Largest MacBook Pro was 8GB of RAM

Today:

Largest MacBook Pro is 96GB of RAM

12X

BIG DATA IS EASY DATA BILONG LIVE EASY

DATA SIZES

- BigQuery: ~95% of customers with < 1TB
- SingleStore: ~80% of customers want S-00
- Gartner: Most EDWs < 100GB
- A16Z: B2B portfolio cos all had data < 1T
- A16Z: B2C portfolio cos had data < 10T

DATA RECENCY

- Most data queried is from today
- Today's data is a small fraction of the whole
- Access history gets quiet very quickly
- Most data stored is very infrequently used



QUERY SIZES

- BQ: 90% of queries < 100 MB
- Just because you have a large data doesn't mean you query it
- Effective partitioning, compression, pushdowns, etc reduce query sizes