Open Data Foundations across Hudi, Iceberg, and Delta Lake

Data Council

Speaker Bio

Kyle Weller

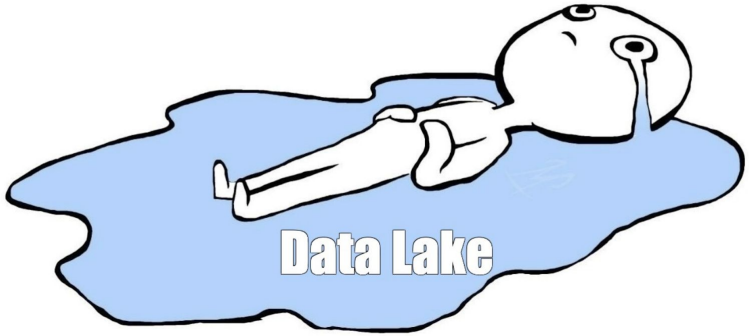
Head of Product @ ONEHOUSE

https://www.linkedin.com/in/lakehouse/

10+ years of building data platforms and data products

- Currently building Onehouse.ai
- Product lead for Azure Databricks 0 -> 9-fig ARR
- Built Azure Machine Learning services inside SQL Server
- Ran data and growth strategy for Cortana (MSFT AI assistant)
- Worked on PB scale data lake platform for Bing Search
- Designed v1 TB scale data lake for MSFT Office

Data Lakes…



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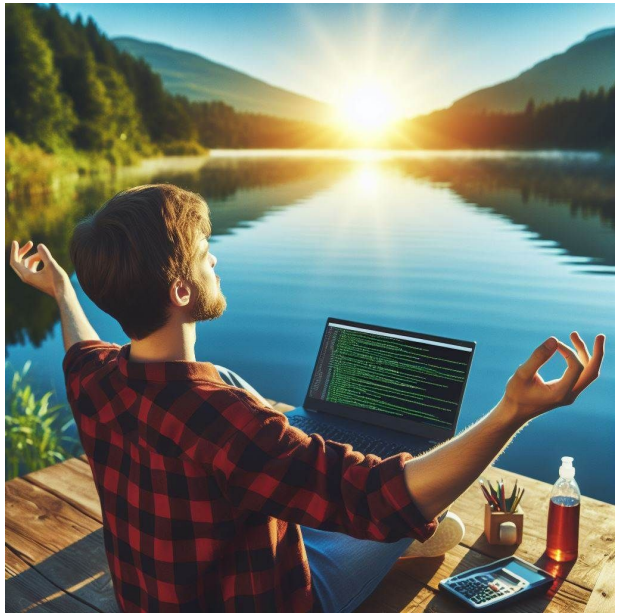
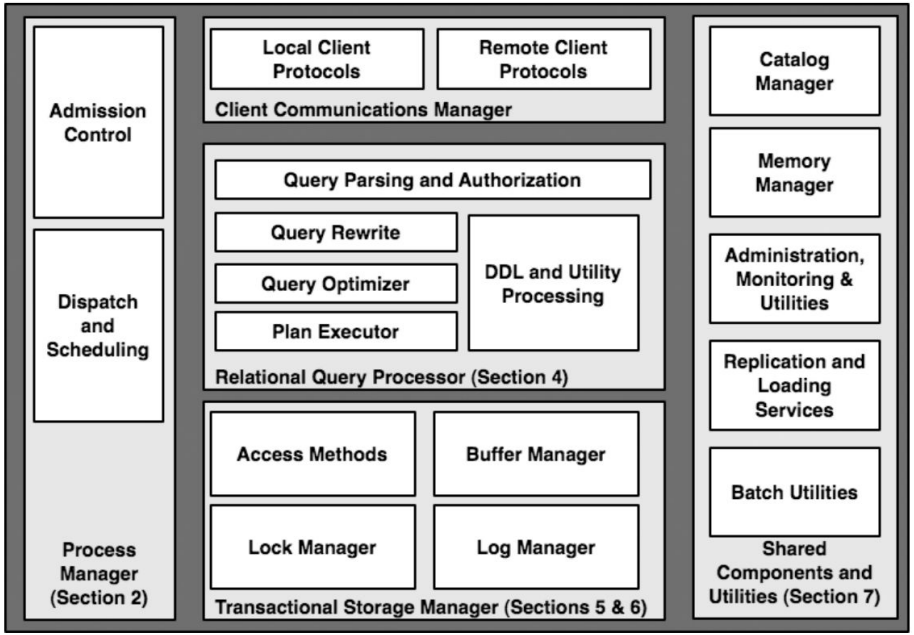


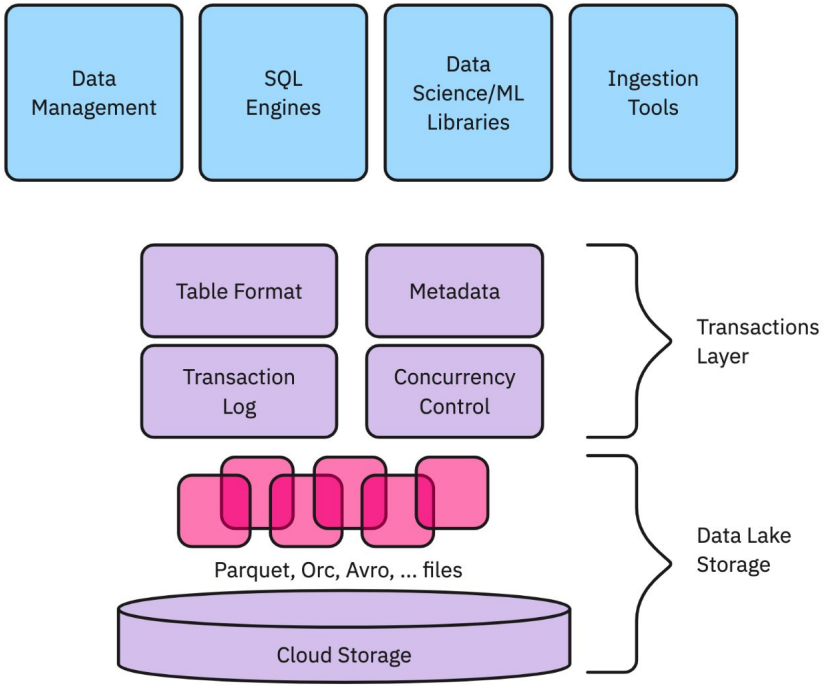
Architecture of a Database System



https://dsf.berkeley.edu/papers/fntdb07-architecture.pdf

S3 Data Lake Storage

Data Lakehouse - Unbundling of the DBMS



Apache hudi DELTA LAKE ICEBERG

A lakehouse has the following key features:

- **Transaction support:** In an enterprise lakehouse many data pipelines will often be reading and writing data concurrently. Support for ACID transactions ensures consistency as multiple parties concurrently read or write data, typically using SQL.
- **Schema enforcement and governance:** The Lakehouse should have a way to support schema enforcement and evolution, supporting DW schema architectures such as star/snowflake-schemas. The system should be able to reason about data integrity, and it should have robust governance and auditing mechanisms.
- **BI support:** Lakehouses enable using BI tools directly on the source data. This reduces staleness and improves recency, reduces latency, and lowers the cost of having to operationalize two copies of the data in both a data lake and a warehouse.
- **Storage is decoupled from compute:** In practice this means storage and compute use separate clusters, thus these systems are able to scale to many more concurrent users and larger data sizes. Some modern data warehouses also have this property.
- **Openness:** The storage formats they use are open and standardized, such as Parquet, and they provide an API so a variety of tools and engines, including machine learning and Python/R libraries, can efficiently access the data **directly**.
- **Support for diverse data types ranging from unstructured to structured data**: The lakehouse can be used to store, refine, analyze, and access data types needed for many new data applications, including images, video, audio, semi-structured data, and text.
- **Support for diverse workloads:** including data science, machine learning, and SQL and analytics. Multiple tools might be needed to support all these workloads but they all rely on the same data repository.
- **End-to-end streaming:** Real-time reports are the norm in many enterprises. Support for streaming eliminates the need for separate systems dedicated to serving real-time data applications.

Origin Stories

2017 — open sourced — Uber

2018 — open sourced — Netflix

2019 — open sourced

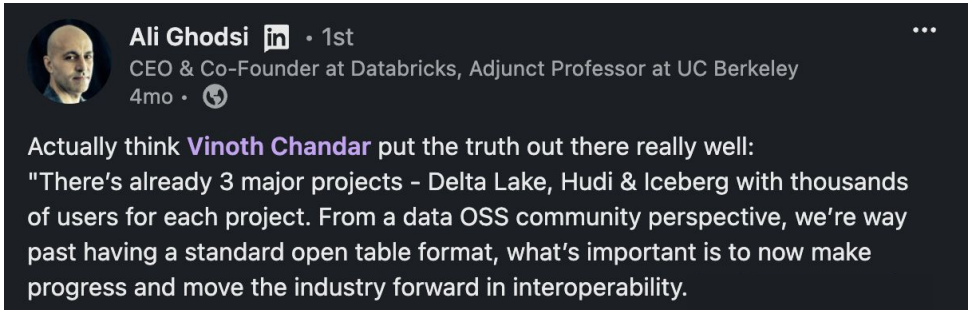
```
+ # Hudi
+ Hudi (pronounced Hoodie) stands for `Hadoop Upserts anD Incrementals`. Hudi manages storage of large analytical datasets on [HDFS](http://hadoop.apache.org/docs/stable/hadoop-project-dist/hadoop-hdfs/HdfsDesign.html) and serve them out via two types of tables

  * **Read Optimized Table** - Provides excellent query performance via purely columnar storage (e.g. [Parquet](https://parquet.apache.org/))
  * **Near-Real time Table (WIP)** - Provides queries on real-time data, using a combination of columnar & row based storage (e.g Parquet + [Avro](http://avro.apache.org/docs/current/mr.html))
```

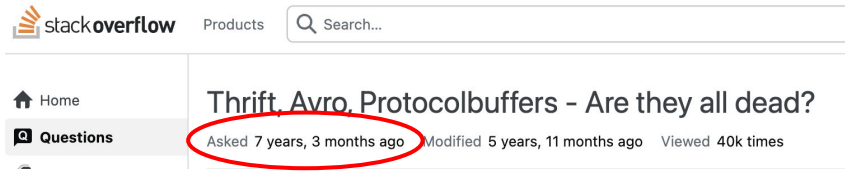
```
+ ## Iceberg
+
+ Iceberg is a new table format for storing large, slow-moving tabular
  data. It is designed to improve on the de-facto standard table layout
  built into Hive, Presto, and Spark.
+
```

```
- Delta Lake Core is .... (copy text from delta docs)
+ Delta Lake is a next-generation engine built on top of Apache Spark. Delta Lake
  provides ACID transactions, optimized layouts and indexes, and execution engine
  improvements for building data pipelines to support big data use cases: batch
  and streaming ingests, fast interactive queries, and machine learning.
  Specifically, Delta offers:
```

🌶️ My Hot Take - They are divergent!

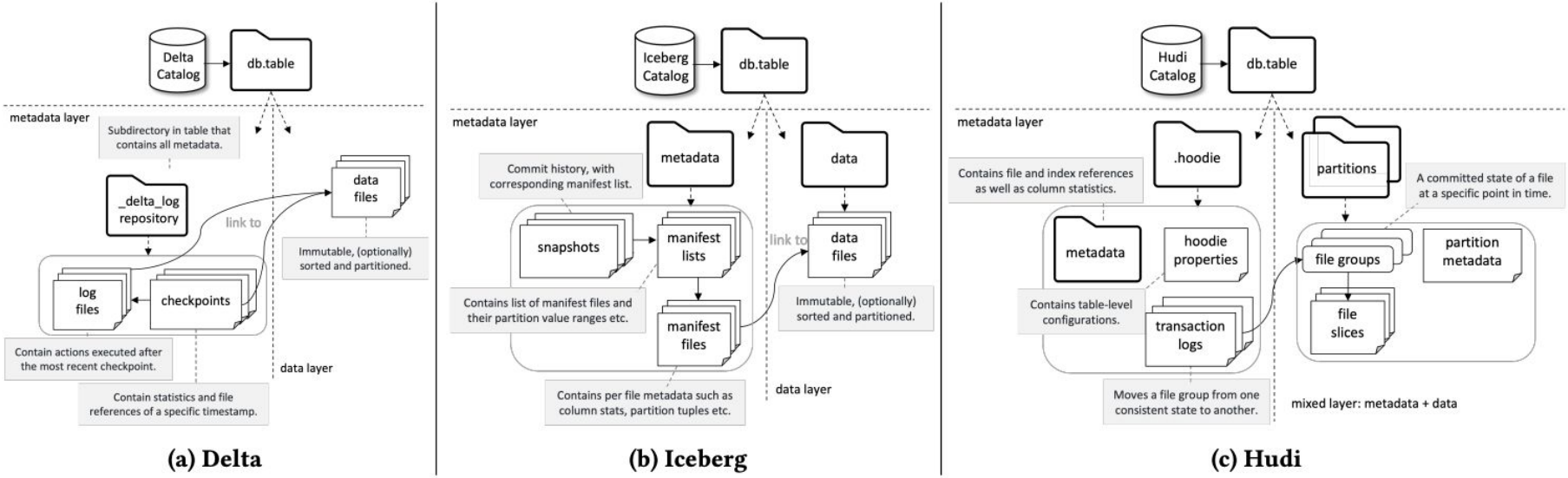
- Technical vision and goals are divergent
- The community needs are specialized
- All three projects are on fast growth trajectories
- New table formats are gaining traction: Apache Paimon, YOHB?





Technical Fundamentals

- Metadata abstractions on files in cloud object storage
- Tables with SQL semantics and schema evolution
- ACID transactions
- Updates and deletes (merge/upsert)
- Data layout optimizations for performance tuning



- How it looks on cloud storage

- Fundamentals of table formats Hudi, Delta, Iceberg are not that different
- Each has a special metadata layer on top of parquet files

Apache hudi

```
s3_bucket/my_table/
 |- .hoodie/
 |    |- hoodie.properties
 |    |- metadata/
 |- file_1.parquet
 |- file_2.parquet
 |- file_N.parquet
```

DELTA LAKE

```
s3_bucket/my_table/
 |- _delta_log/
 |    |- 000000.json
 |- file_1.parquet
 |- file_2.parquet
 |- file_N.parquet
```

ICEBERG

```
s3_bucket/my_table/
 |- metadata/
 |    |- v1.metadata.json
 |    |- snap-9fa1-2-16c3.avro
 |    |- 0d9a-98fa-77.avro
 |- file_1.parquet
 |- file_2.parquet
 |- file_N.parquet
```

- Which Format Should I Choose?

Choose Apache Hudi if:

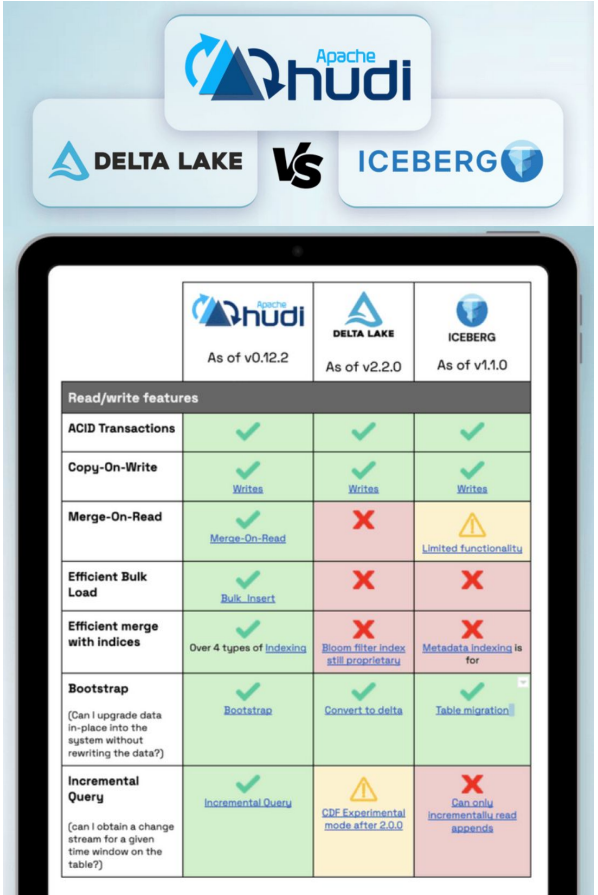
1. Mutable data - GDPR Deletes, Updates
2. CDC workloads
3. Low latency requirements
4. Large ETL pipelines - perf/cost w/ incremental ETL

Choose DELTA LAKE if:

1. Best Databricks experience
2. Needs fastest premium Spark with Photon
3. Wants an “easy-to-get-started” table format

Choose ICEBERG if:

1. Trino or Athena writes
2. Snowflake writes
3. Not sensitive to performance
4. Partition evolution



	Apache hudi As of v0.12.2	DELTA LAKE As of v2.2.0	ICEBERG As of v1.1.0
Read/write features			
ACID Transactions	✓	✓	✓
Copy-On-Write	✓ Writes	✓ Writes	✓ Writes
Merge-On-Read	✓ Merge-On-Read	✗	⚠ Limited functionality
Efficient Bulk Load	✓ Bulk_insert	✗	✗
Efficient merge with indices	✓ Over 4 types of Indexing	✗ Bloom filter index still proprietary	✗ Metadata indexing is for
Bootstrap (Can I upgrade data in-place into the system without rewriting the data?)	✓ Bootstrap	✓ Convert to delta	✓ Table migration
Incremental Query (can I obtain a change stream for a given time window on the table?)	✓ Incremental Query	⚠ CDF Experimental mode after 2.0.0	✗ Can only incrementally read appends

- Which Format Should I Choose?

Choose Apache Hudi if:

1. Mutable data - GDPR Deletes, Up[illegible]s
2. [illegible]DC workloads
3. L[illegible] latency requirements
4. Larg[illegible]TL pipelines - perf/cost w/ increme[illegible]al ETL

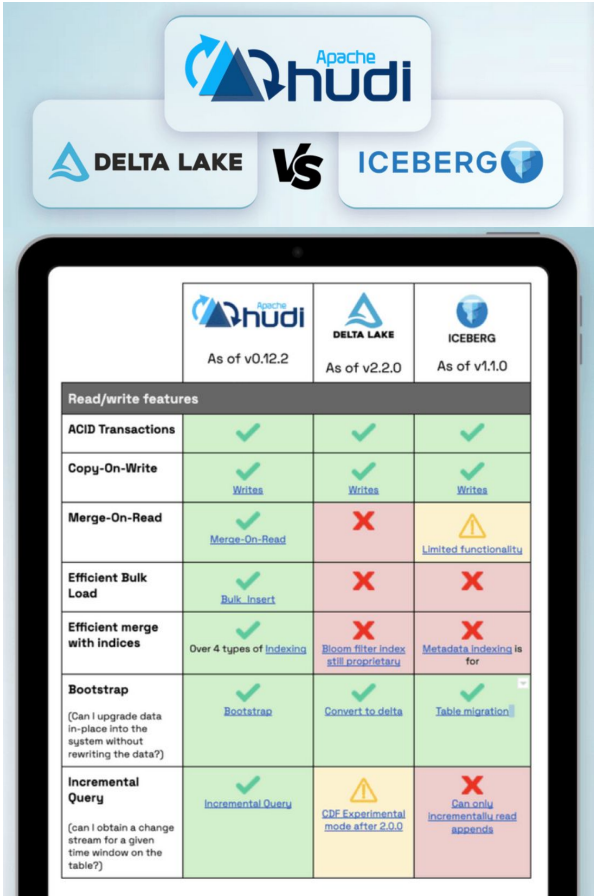
Choose DELTA [illegible] if:

1. [illegible]
2. Needs fastest pre[illegible] Spark with Photon
3. Wants an "easy-to-ge[illegible]rted" table format

What if you could work across all 3?

Choose ICEBERG if:

1. Trino or Athena writes
2. Snowflake writes
3. [illegible]ensitive to performance
4. Parti[illegible]tion



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- Example benefits of mix-and-match

Writing

Choose Apache Hudi writing w/ EMR (Spark)

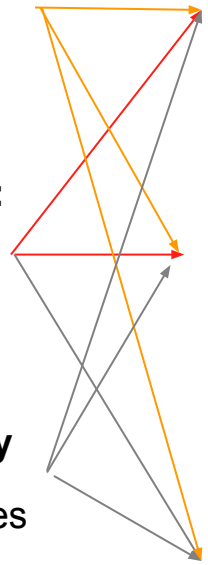
1. Fastest writes for mutable workloads
2. Most flexible tuning parameters for ingestion

Choose Delta Lake writing w/ Fabric:

1. Easy-to-get-started out of the box
2. Makes data available to the entire Azure portfolio

Choose Iceberg writing w/ BigQuery

1. Only table format supported for writes
2. Partition evolution



Reading

Choose Delta Lake reading w/ Databricks

1. Get fastest queries with Photon acceleration
2. Great experience for Data Science

Choose Iceberg reading w/ Snowflake

1. Only supported table format in Snowflake
2. Decouple data storage using external tables

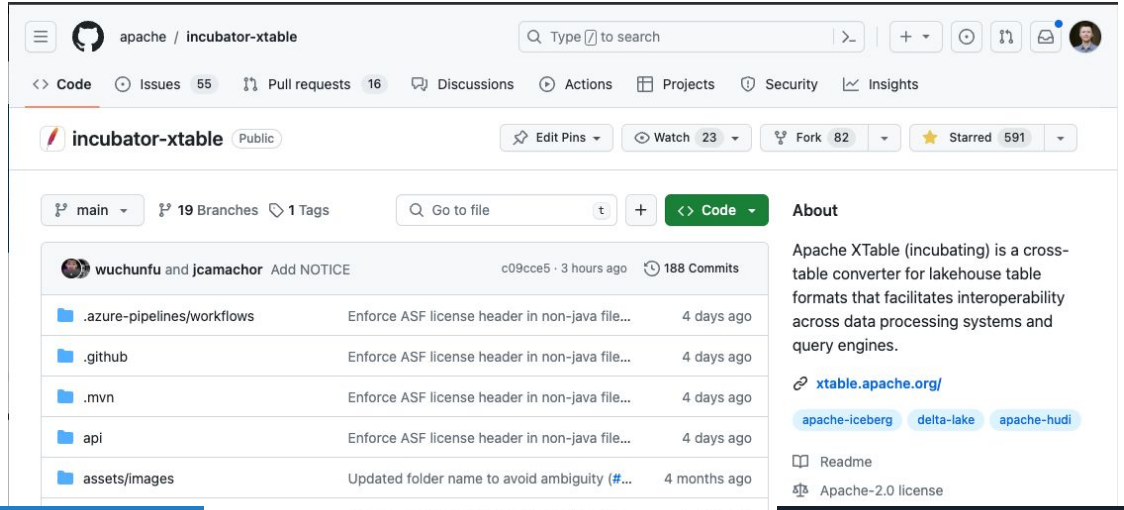
Choose Apache Hudi reading w/ DataProc (Spark)

1. Fast record level indexes for point queries
2. Powerful secondary indexing capabilities for Spark

Introducing:

Apache XTable™ Incubating

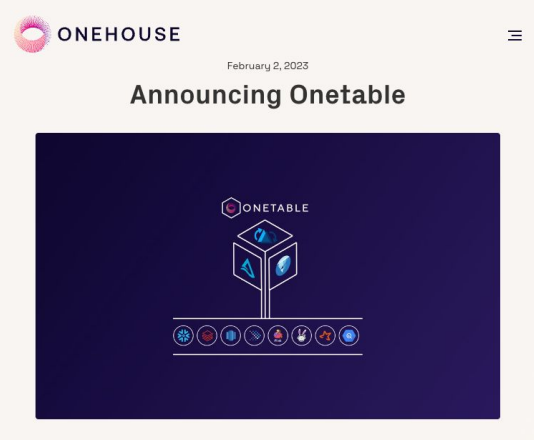
⭐Celebrate by adding a little star ⭐
https://github.com/apache/incubator-xtable



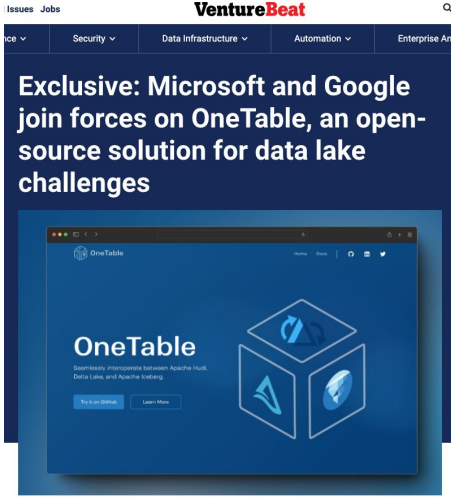
>600 GH Stars ⭐

>80 Forks

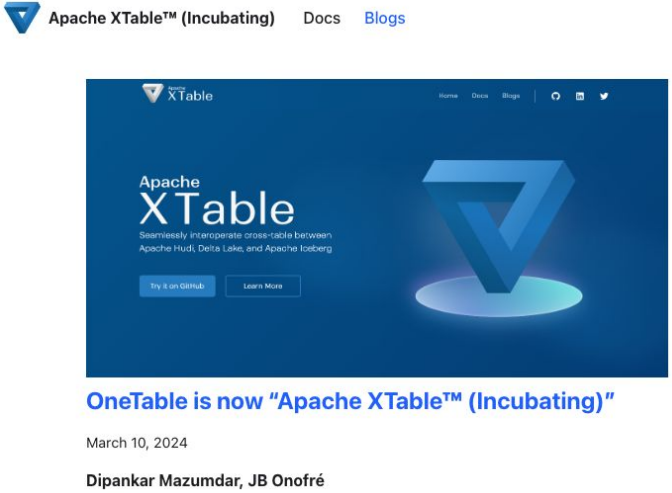
Apache XTable™ - Timeline



Onehouse announces OneTable



OSS Co-Launch with Microsoft, Google, Onehouse



Donation to ASF and incubation as Apache XTable

Feb 2023 | **Nov 2023** | **Mar 2024**

ONEHOUSE

Microsoft Google ONEHOUSE

Microsoft CLOUDERA Walmart ONEHOUSE dremio Adobe Google

Apache XTable™ - How it Works

1: Choose your “source” format
2: Choose your “target” format(s)
3: XTable translates the metadata layers

Read your table as any of the formats

Hudi | Delta | Iceberg

yaml

```yaml
sourceFormat: HUDI
targetFormats:
  - DELTA
  - ICEBERG
datasets:
  -
    tableBasePath: s3://path/to/hudi-dataset/people  # replace this with gs://path/to/hudi-dataset/people if yo
    tableName: people
    partitionSpec: city:VALUE
```

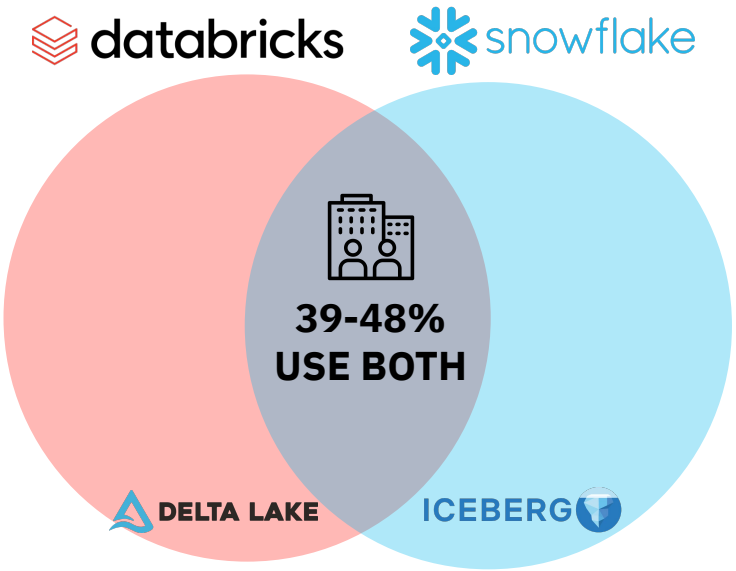
Apache hudi

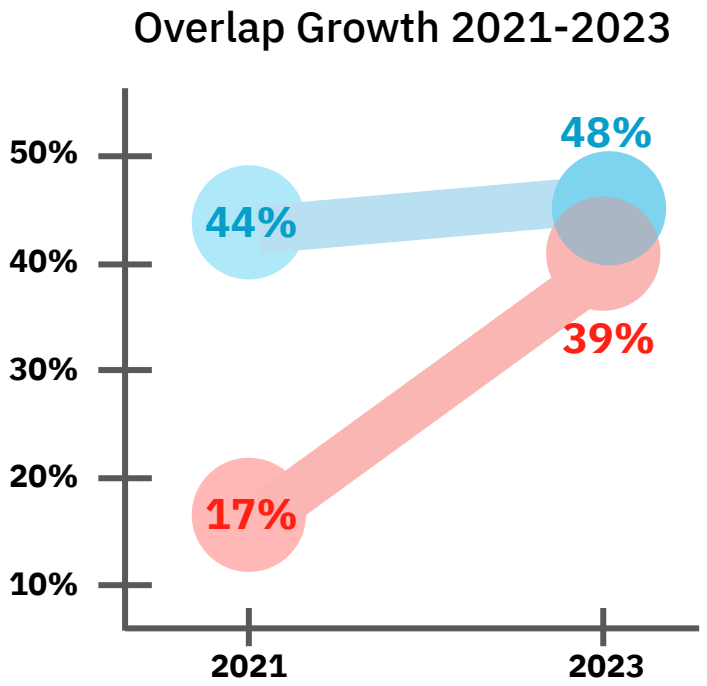
```
s3_bucket/my_table/
 |- .hoodie/
 |     |- hoodie.properties
 |     |- metadata/
 |- file_1.parquet
 |- file_2.parquet
 |- file_N.parquet
```

```
s3_bucket/my_table/
 |- .hoodie/
 |     |- hoodie.properties
 |     |- metadata/
 |- _delta_log/
 |     |- 000000.json
 |- metadata/
 |     |- v1.metadata.json
 |     |- snap-9fa1-2-16c3.avro
 |     |- 0d9a-98fa-77.avro
 |- file_1.parquet
 |- file_2.parquet
 |- file_N.parquet
```

```
# any of these work on the same table
spark.read.format("hudi")
spark.read.format("delta")
spark.read.format("iceberg")
```

A tale of two...



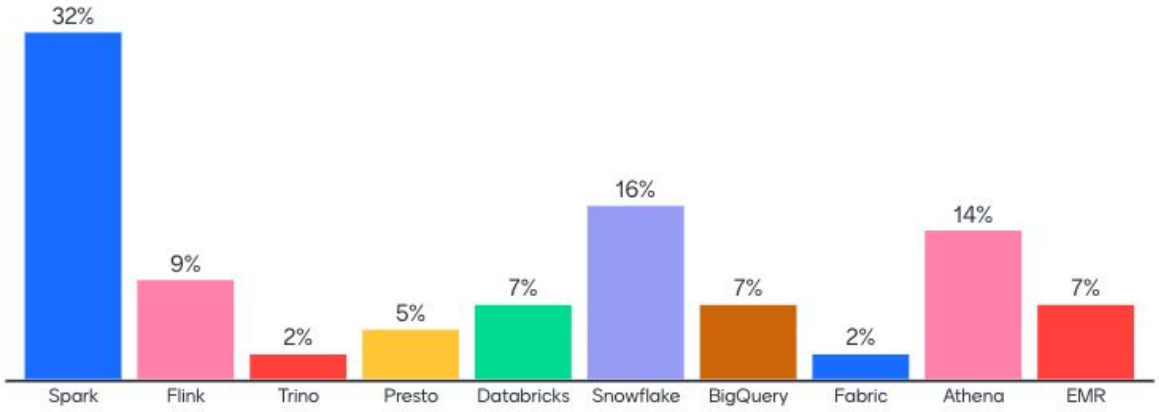


https://siliconangle.com/2023/06/30/connecting-dots-snowflakes-data-cloud-ambitions/

Join at menti.com | use code 5780 3382

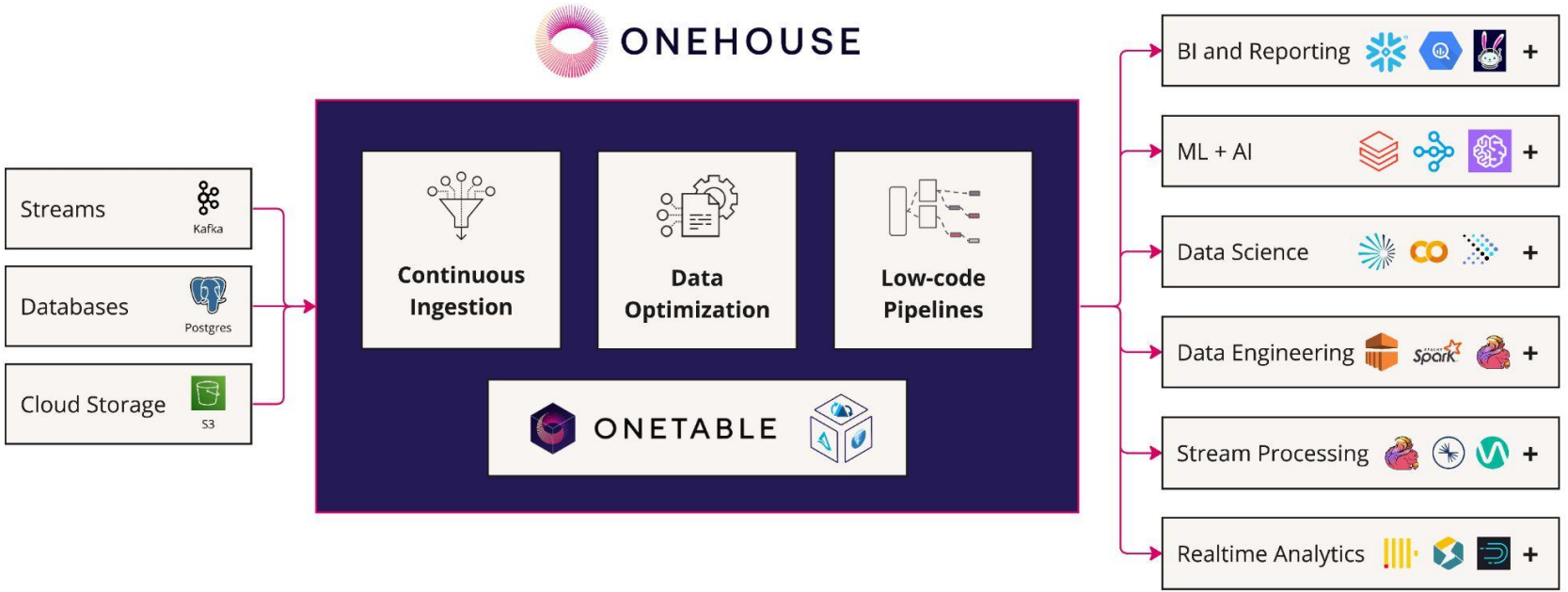
Mentimeter

Which Query Engines Do You Use?



Apache XTable™ - Why Build It?

Whitepaper



Our Goal = Universal Data Lakehouse

Demo Time!

Apache XTable™ - Vision for the Future

Goals

Seamless and efficient interoperability
Eliminate data silos
Project sustainability and evolution

Features

Real-time and transparent replication in any direction
Accurate and lossless model
Extensibility and flexibility

Community

Neutral and inclusive: Vendors, Cloud providers, Users
Graduate ASF Incubation

Apache XTable™ - Join the community

Initial Committers

- Tim Brown : ***Onehouse***
- Vamshi Gudavarthi : ***Onehouse***
- Ashvin Agrawal : ***Microsoft***
- Jesus Camacho Rodriguez : ***Microsoft***
- Anoop Johnson : ***Google***
- Stamatis Zampetakis : ***Cloudera***
- Hitesh Shah : ***Adobe***
- Jean-Baptiste Onofré : ***Dremio***
- Baljinder Singh : ***Walmart***
- Vinish Reddy: ***Onehouse***
- Vinoth Chandar: ***Onehouse***

CLOUDERA

Microsoft

ONEHOUSE

Google

Adobe

dremio

Walmart

APACHE INCUBATOR

Apache XTable™ - Roadmap

Current Status	Roadmap (6-12 months)	Roadmap (long term)
➢ Supported formats: ***Apache Hudi, Apache Iceberg, and Delta Lake***	➢ ***Merge-on-Read*** (delete vectors)	➢ Multi-writer (duplex)
➢ Tested with: Apache Spark, Trino, Microsoft Fabric, Databricks, BigQuery, Snowflake, Redshift, and more	➢ Apache Paimon (incubating)	➢ Synchronized commit timestamp
➢ Features: on-demand incremental conversion, copy-on-write, catalog integration, change-history	➢ ***Performance***, ***efficiency***, and resiliency	➢ Feature parity (superset)
	➢ Deployment: as-a-service and in-memory	➢ New technology stack
	➢ Native engine integration	➢ Support new formats & versions

Apache XTable™ - Let's Build Together


Github: https://github.com/apache/incubator-xtable

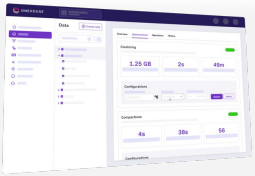

Docs : https://xtable.apache.org/docs/how-to


Twitter : https://twitter.com/apachextable


LinkedIn : https://www.linkedin.com/company/apache-xtable/

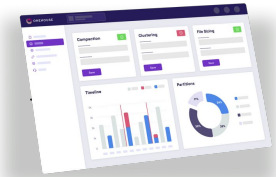

Mailing List : dev-subscribe@xtable.apache.org

ONEHOUSE - 3 Ways to Engage with Onehouse



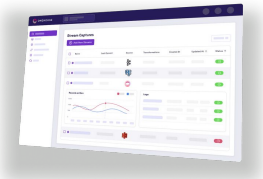
Lakehouse Monitoring - ($0 Free)

- No install, no permissions
- Monitoring and tuning insights for your Hudi, Iceberg, and Delta Lake pipelines



Lakehouse Table Services

- Keep your existing Hudi, Iceberg, Delta pipelines
- Onehouse will automate advanced Table Optimizations for 10x faster analytics



Lakehouse Streaming Ingest/ELT

- 10x Faster/Cheaper vs existing OSS Hudi, Iceberg, Delta pipelines
- Fully managed auto-scaling infrastructure w/ serverless experience in your VPC
- Simple UI + APIs for programmatic and templated CI/CD devops integration

Contact me: kyle@onehouse.ai

Apache Hudi DELTA LAKE ICEBERG

Thank You!