



# How is Google Cloud Platform Different



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Google Cloud



# Today's Agenda

GCP DA

# Databricks Timeline

# Databricks Architecture

# Databricks Deployment Model

@GCP Architecture Reference

## DBX and BQ and

## why hybrid make sense



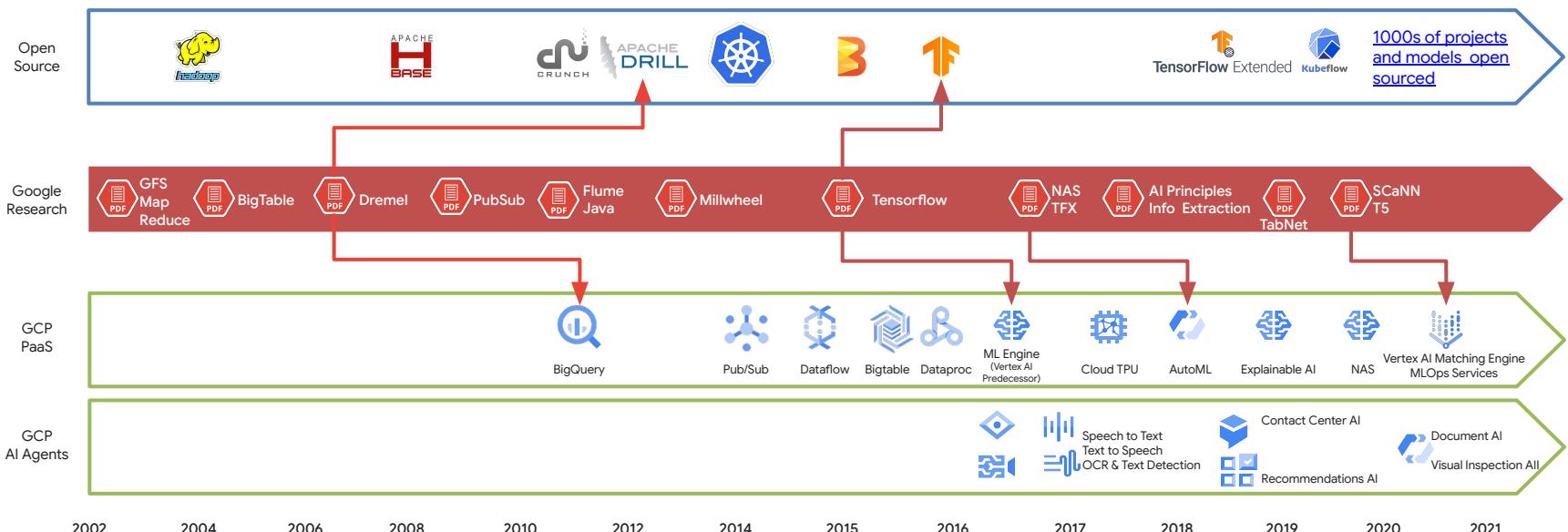


# State-of-the-Art AI

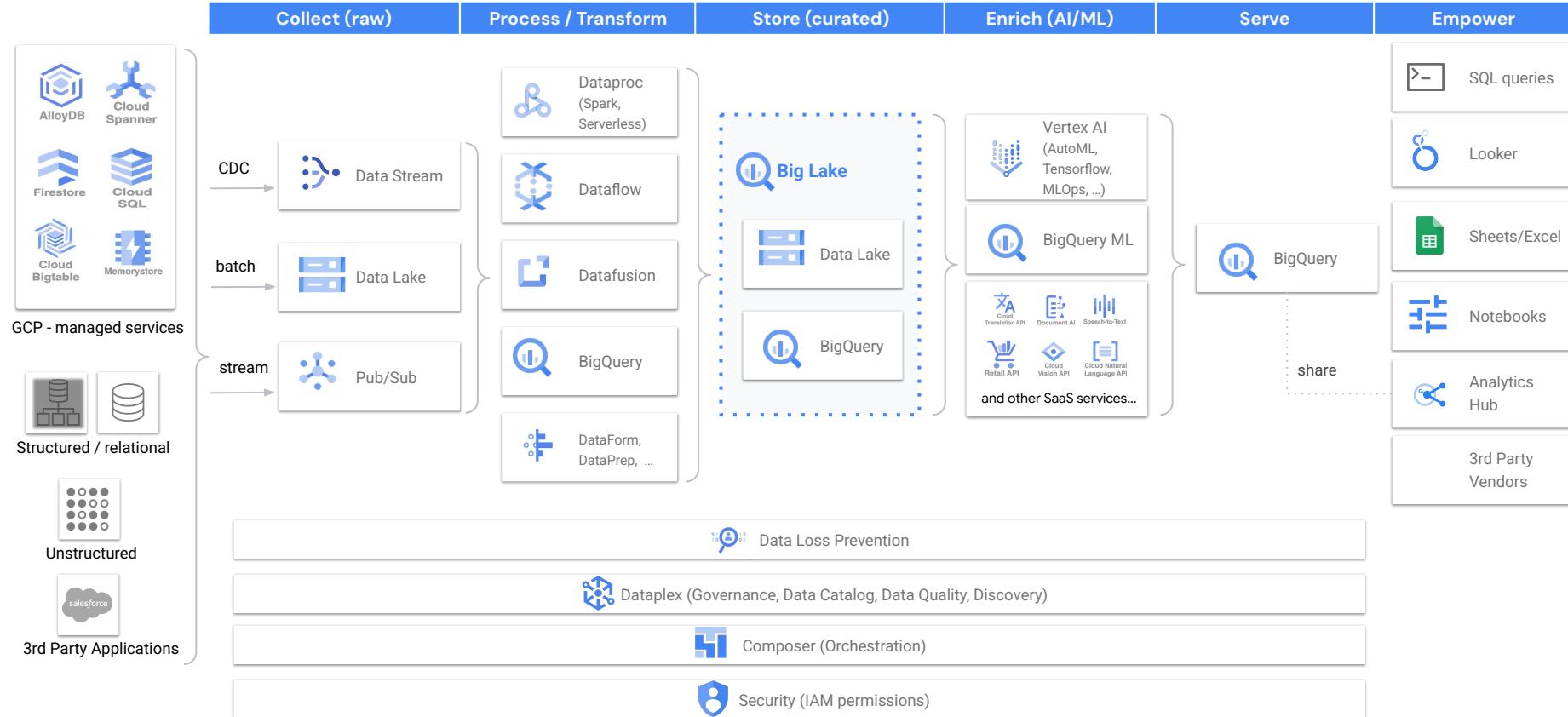
Building the next generation of **Data to AI**, **E2E MLOps**, and **Scalable Infrastructure**

3,000+ Google / Deepmind researchers and 7,000+ published publications drive a product pipeline

that results in **better product performance** and **lower costs** to reach that performance for Vertex AI customers.



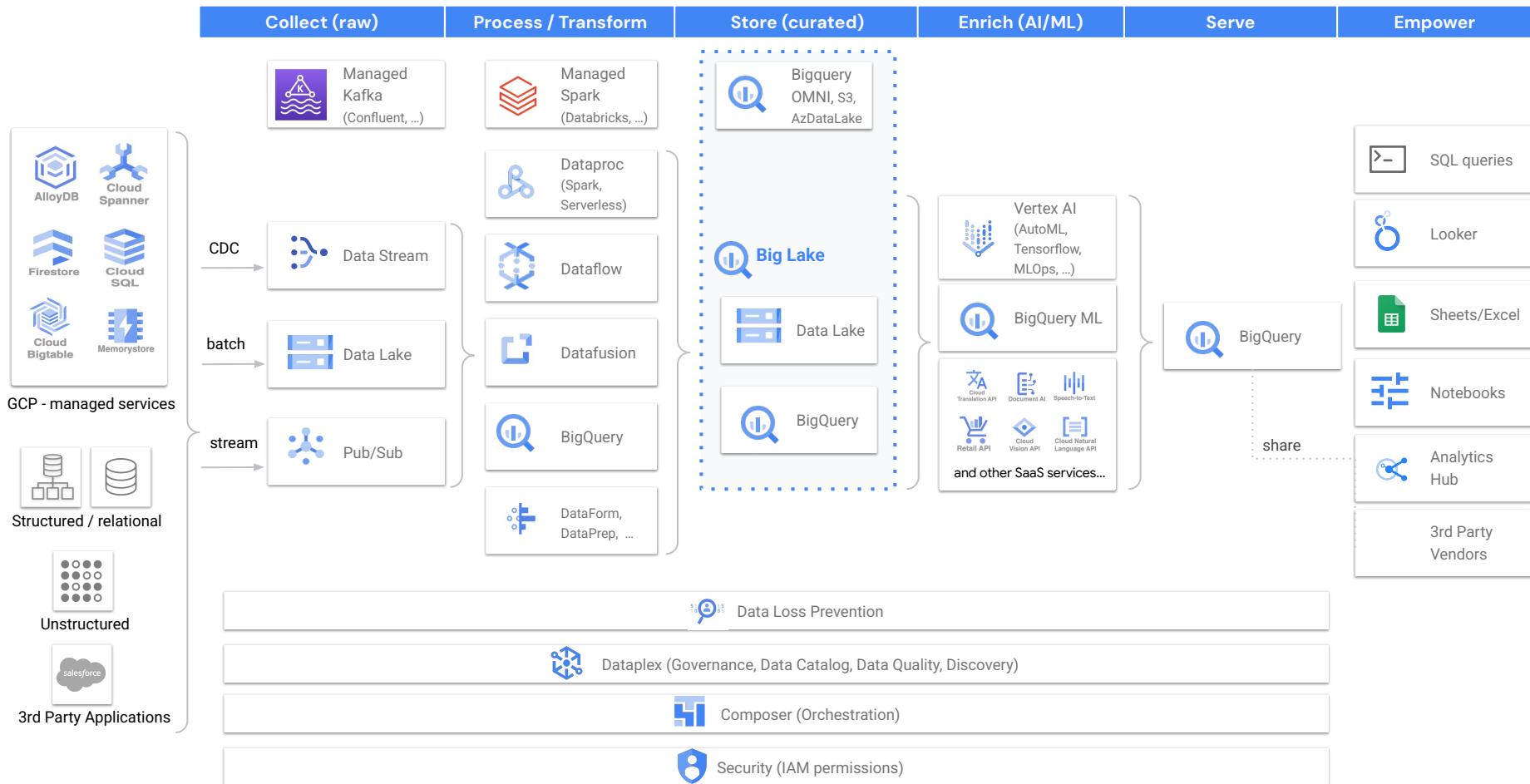
# GCP main data services

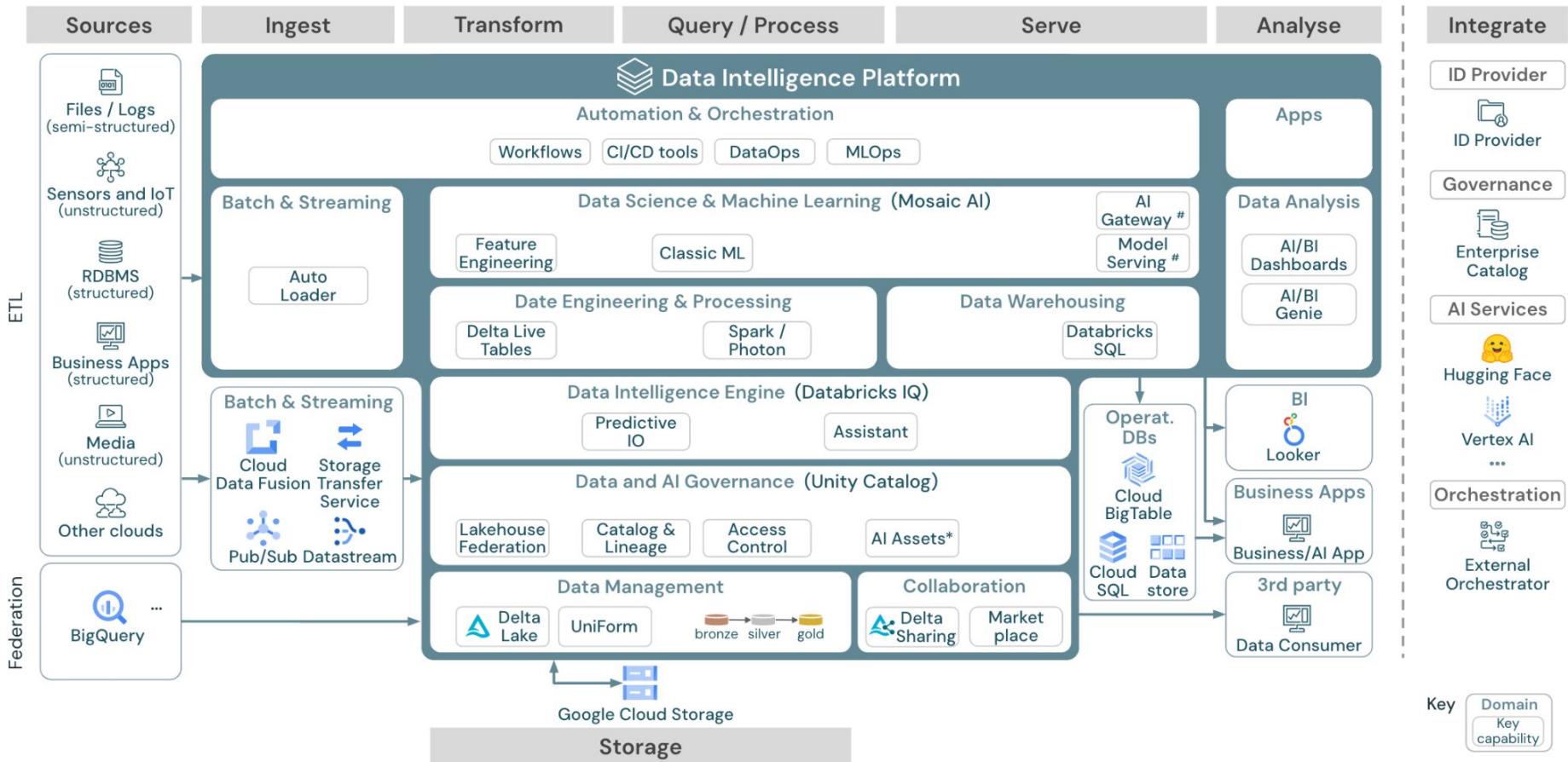


# GCP main data services



# GCP - main data services + partner services (exemple)

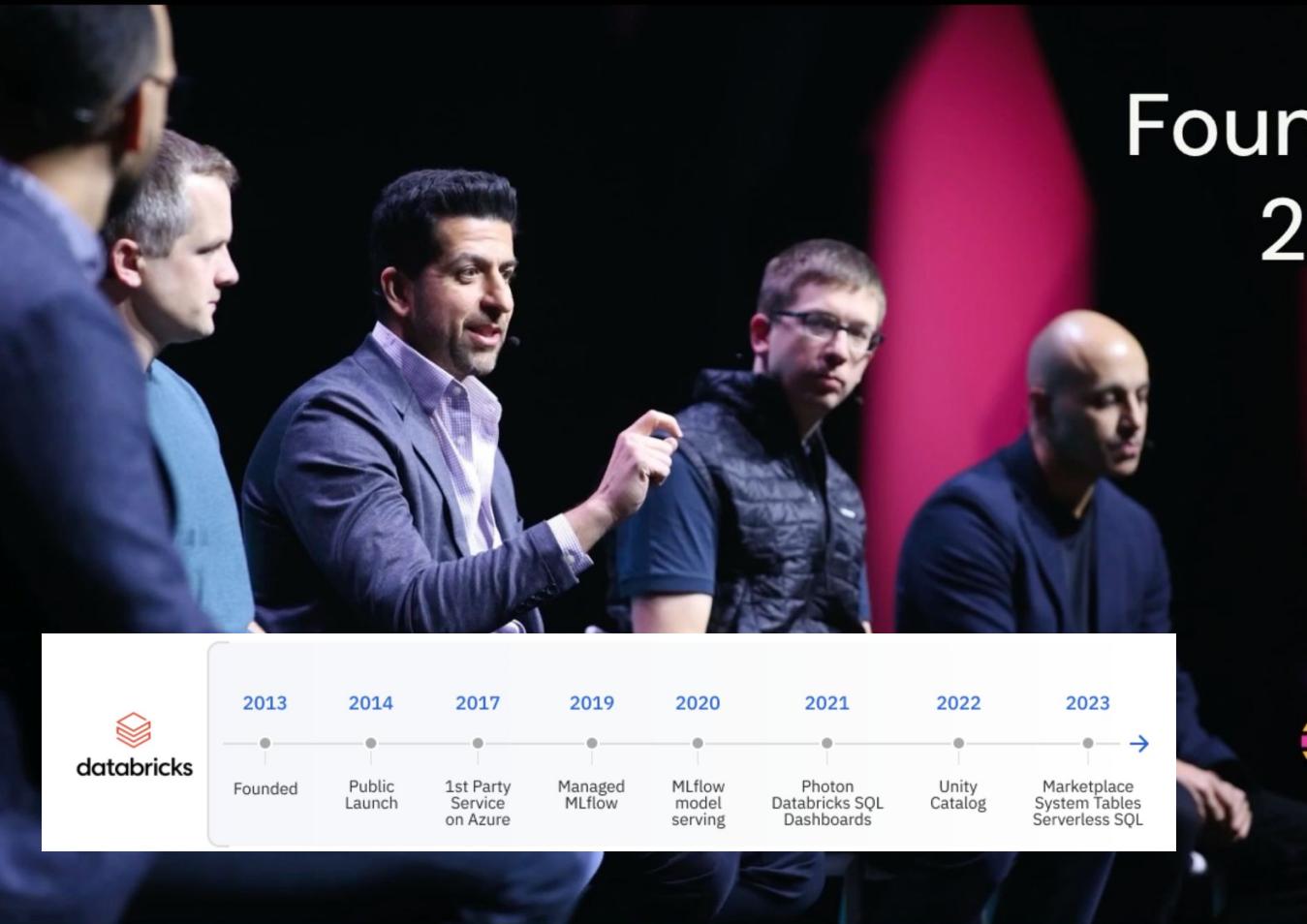




[Databricks GCP Ref Arch](#)

\* features, models, functions  
# available in selected regions

Key  
Domain  
Key capability



Founded in  
2013



**mlflow**<sup>™</sup>

**APACHE Spark**<sup>™</sup>

The logo for Unity Catalog, featuring a purple hexagon icon with white dots and the text "Unity Catalog" in a bold, sans-serif font.

D B R X



## Strengths

- DA Centric
- Open Source Flavour
  - Delta (parquet+metadata)  
DLT are proprietary
  - MLFlow
- Developer Experience
- Single Copy of Data
- Multi-Cloud Support
- Integrated Platform

## Main Competitors

- Snowflake
- Cloud Hyperscalers (Az/AWS)

## Cons

- Lakehouse: don't compete with DW
- Unity Catalog: is a must for many functionalities, management overhead, lockin
- DatabricksSQL: expensive and/or poor performant and/or hard to scale, complex to maintain and debug
- DLT is proprietary, data cannot be accessed outside o Dbx
- Migration between clouds is like between DataProc and Dbx
- ...

# Introducing DBRX: A New State-of-the-Art Open LLM

by [The Mosaic Research Team](#)

March 27, 2024 in [Mosaic AI Research](#)

Today, we are excited to introduce DBRX, an open, general-purpose LLM created by Databricks. Across a range of standard benchmarks, DBRX sets a new state-of-the-art for established open LLMs. Moreover, it provides the open community and enterprises building their own LLMs with capabilities that were previously limited to closed model APIs; according to our measurements, it surpasses GPT-3.5, and it is competitive with Gemini 1.0 Pro. It is an especially capable code model, surpassing specialized models like CodeLLaMA-70B on programming, in addition to its strength as a general-purpose LLM.

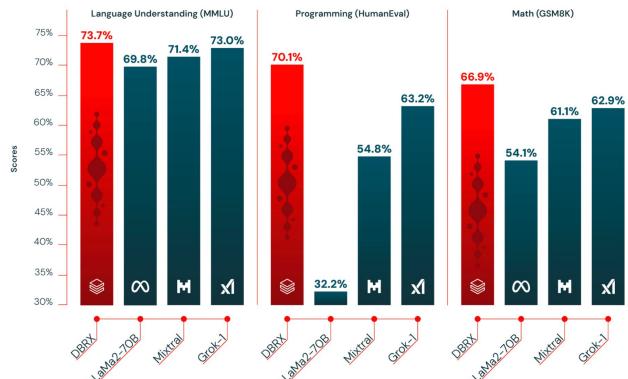


Figure 1: DBRX outperforms established open source models on language understanding (MMLU), Programming (HumanEval), and Math (GSM8K).

# Lakehouse: A New Generation of Open Platforms that Unify Data Warehousing and Advanced Analytics

Michael Armbrust<sup>1</sup>, Ali Ghodsi<sup>1,2</sup>, Reynold Xin<sup>1</sup>, Matei Zaharia<sup>1,3</sup>

<sup>1</sup>Databricks, <sup>2</sup>UC Berkeley, <sup>3</sup>Stanford University

## Abstract

This paper argues that the data warehouse architecture as we know it today will wither in the coming years and be replaced by a new architectural pattern, the Lakehouse, which will (i) be based on open direct-access data formats, such as Apache Parquet, (ii) have first-

quality and governance downstream. In this architecture, a small subset of data in the lake would later be ETLed to a downstream data warehouse (such as Teradata) for the most important decision support and BI applications. The use of open formats also made data lake data directly accessible to a wide range of other analytics engines, such as machine learning systems [30, 37, 42].

From 2015 onwards, cloud data lakes, such as S3, ADLS and GCS,

CIDR '21, Jan. 2021, Online

Michael Armbrust, Ali Ghodsi, Reynold Xin, and Matei Zaharia

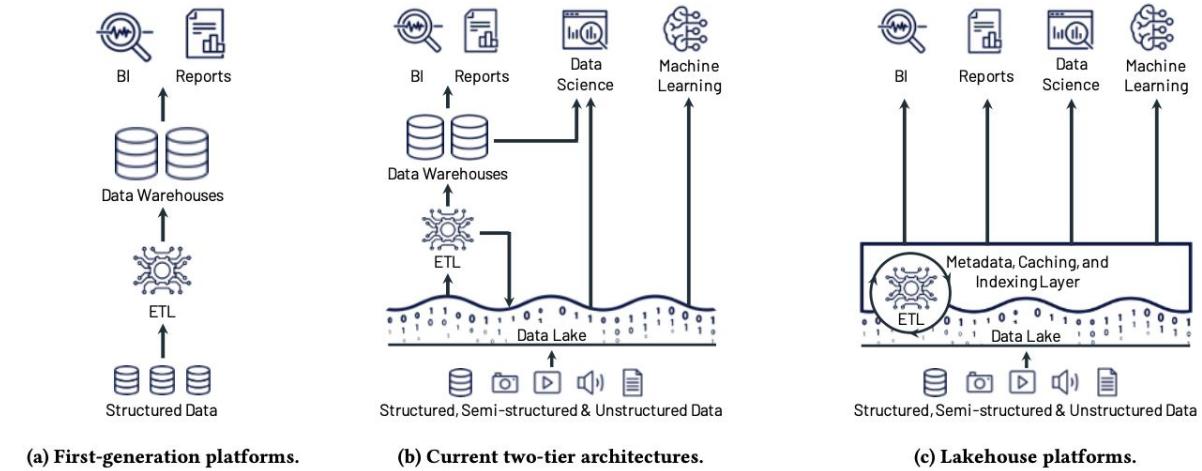


Figure 1: Evolution of data platform architectures to today's two-tier model (a-b) and the new Lakehouse model (c).

# Data Warehouses

## Strengths

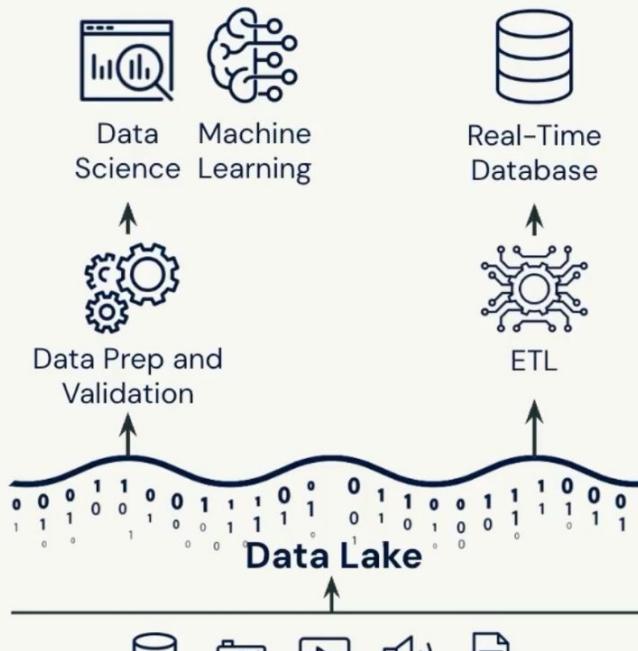
- Purpose-built for BI and reporting
- Meant to unify disparate systems
- Houses structured, clean data with a standardized schema

## Downsides

- No support for semi or unstructured data
- Poor support for data science, AI and streaming use cases
- Uses closed and proprietary formats
- Expensive to scale



# Data Lakes



## Strengths

- Store any kind of data
- Inexpensive storage
- Good starting point
- Support for GenAI and streaming use cases

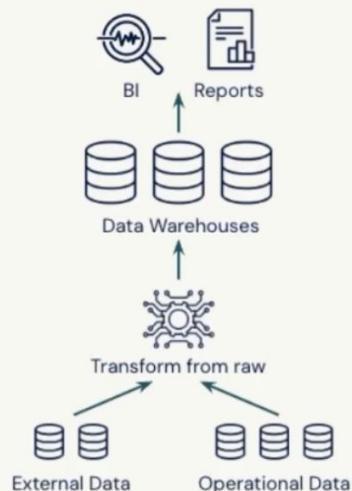
## Downsides

- Complex to set up
- Poor BI performance
- Can become unreliable data swamps
- Governance concerns
- Warehouses still needed

# Data Lakehouse

One platform to unify all your data, analytics, and AI workloads

## Data Warehouses



## Data Lakes



# Databricks Data Intelligence Platform



Use generative AI to understand the semantics of your data

## DatabricksIQ

### Unity Catalog

Unified security, governance, and cataloging

Securely get insights in natural language

### Delta Lake + Uniform

Unified data storage for reliability and sharing

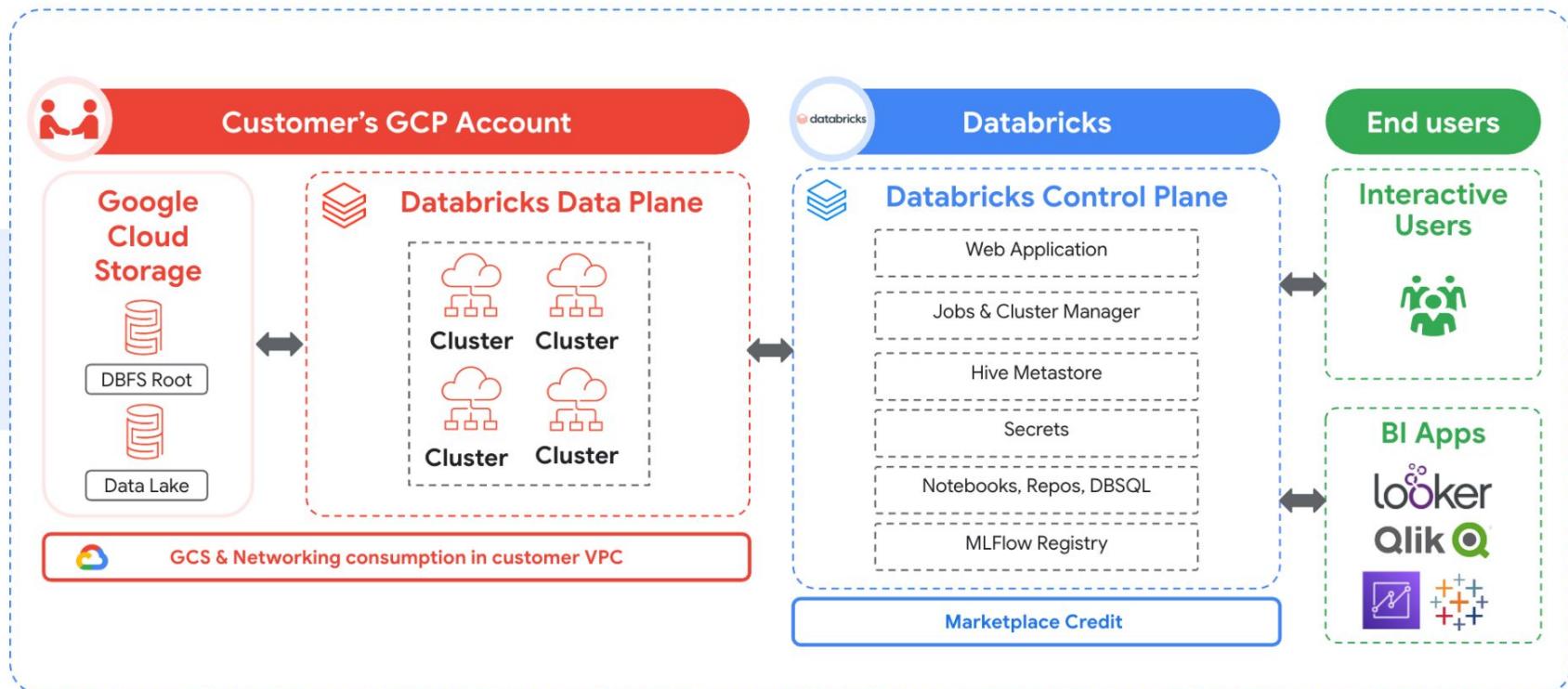
Data layout is automatically optimized based on usage patterns

## Open Data Lake

All Raw Data  
(Logs, Texts, Audio, Video, Images)

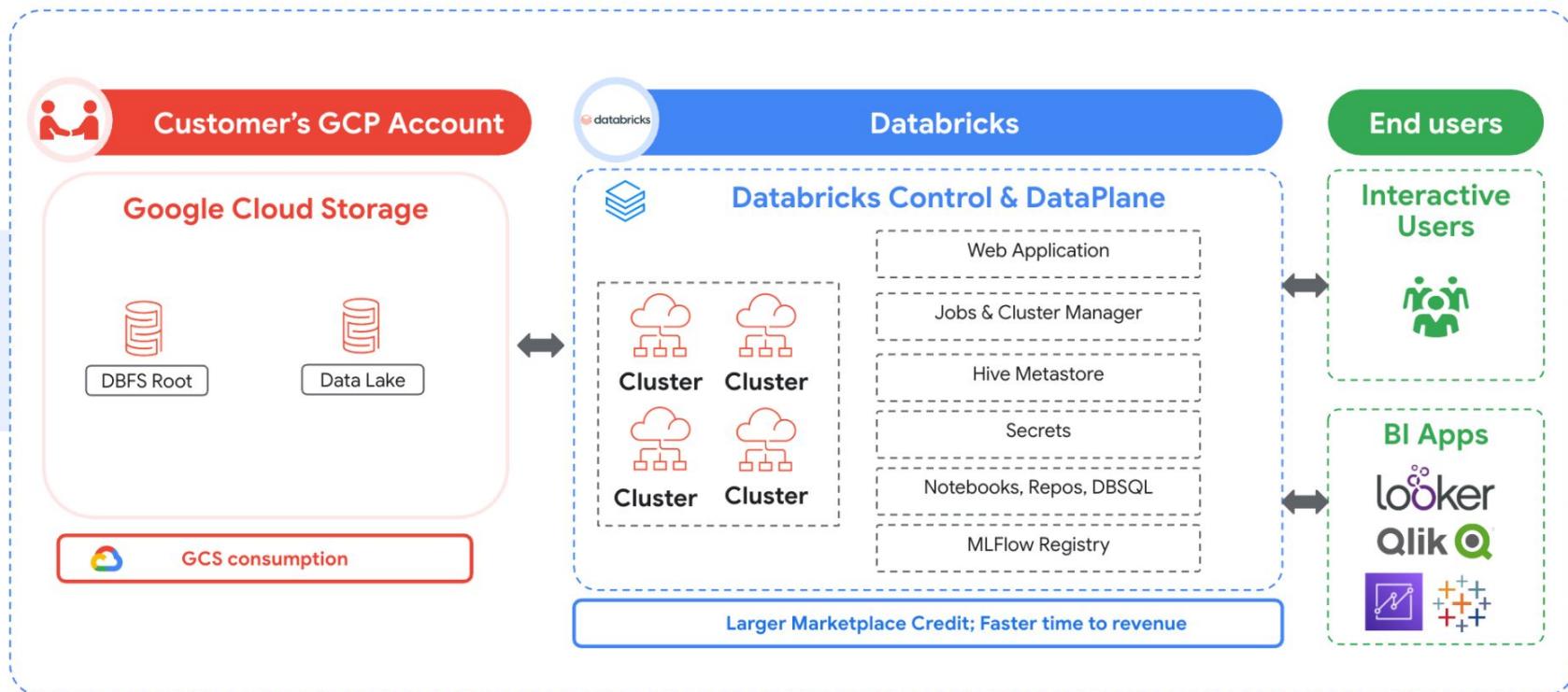
# Databricks Deployment Model (Hybrid SaaS)

Databricks uses GCP compute and storage in Customer GCP projects



# Databricks Deployment Model (Serverless)

Databricks uses storage in Customer GCP projects



# Databricks Compute

## Runtimes

### Standard

Apache Spark and many other components and updates to provide an optimized big data analytics experiences.

### Machine Learning

Adds popular machine learning libraries like TensorFlow, Keras, PyTorch, and XGBoost.

## Specialized Compute

### SQL Warehouses

Specifically designed for the optimization of SQL BI workloads with built in optimization for best price/performance.



**Enhanced with  
Databricks Photon**

# Delta Lake Overview

## Key Features



ACID Transactions



DML Operations



Time Travel



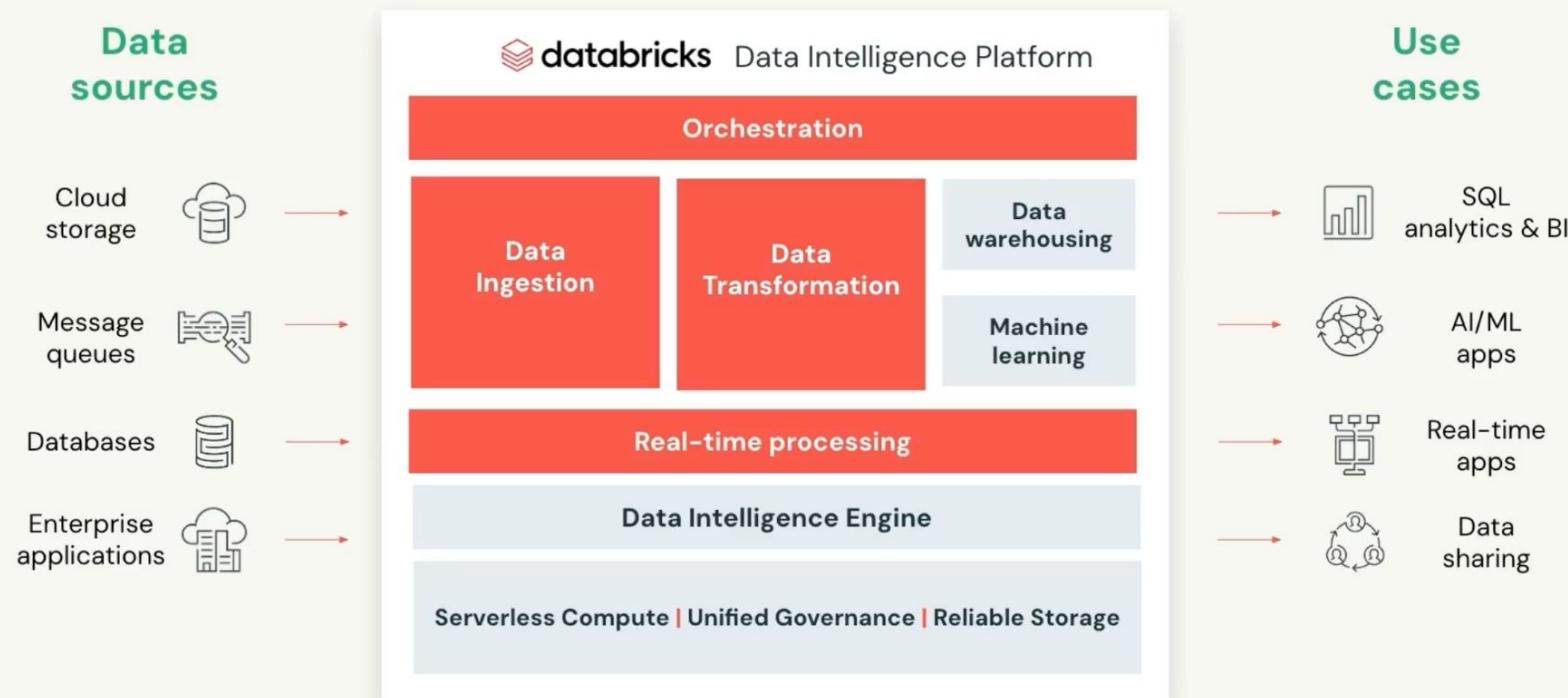
Schema Evolution and  
Enforcement



Many more!

- Unified Batch and Streaming
- Performance
- Scalable Metadata
- Optimization
- **Delta Lake is Open Source!**

# Data Engineering on Databricks



# Delta Live Tables

The best way to do ETL on the lakehouse

```
CREATE STREAMING TABLE raw_data  
AS SELECT *  
FROM cloud_files ("/raw_data",  
"json")
```



## Accelerate ETL development

Declare **SQL or Python** and DLT automatically orchestrates the DAG, handles retries, changing data



## Automatically manage your infrastructure

Automates complex tedious activities like **recovery, auto-scaling, and performance optimization**



## Ensure high data quality

Deliver reliable data with built-in **quality controls, testing, monitoring, and enforcement**



## Unify batch and streaming

Get the simplicity of SQL with freshness of streaming with one **unified API**

# AI/BI Genie Spaces

Public Preview

Ask questions of your data in natural language using your terms

The screenshot shows a Databricks AI/BI Genie Spaces interface. At the top left is the Databricks logo and the text "Genie Spaces". Below it is a breadcrumb navigation "Frontier Sports Sales Data". A descriptive text block states: "Gold data set for sales performance by customer, product, and employee. Or the TotalRevenue column for each sales representative and rounds the result to 2 decimal places. The results are grouped by Employee\_FullName and ordered in descending order by TotalRevenue. The query returns the top 20 sales representatives based on their total revenue." To the right is a "Share" button and a small circular icon with the letter "R". On the far left is a vertical sidebar with icons for message, file, list, and settings. The main area displays a table titled "Employee\_FullName" with columns "Employee\_FullName" and "TotalRevenue". The table contains 17 rows of data. At the bottom is a code input field with "break t" typed in, and a "Run" button.

|    | Employee_FullName        | TotalRevenue |
|----|--------------------------|--------------|
| 1  | Jae Pak                  | 46027844.66  |
| 2  | Linda Mitchell           | 45230622.88  |
| 3  | José Saraiva             | 33972862.23  |
| 4  | Ranjit Varkey Chudukatil | 30890245.01  |
| 5  | Jillian Carson           | 30406205.34  |
| 6  | Michael Blythe           | 28625254.28  |
| 7  | Shu Ito                  | 22293354.14  |
| 8  | Rachel Valdez            | 21539850.68  |
| 9  | Tsvi Reiter              | 20901627.45  |
| 10 | Pamela Anzman-Wolfe      | 16719218.7   |
| 11 | David Campbell           | 14929718.3   |
| 12 | John Doe                 | 13545678.9   |
| 13 | Sarah Johnson            | 12876543.2   |
| 14 | Mike Williams            | 12534567.8   |
| 15 | Alice Green              | 12210987.5   |
| 16 | Brian Lee                | 11987654.3   |
| 17 | Craig Smith              | 11764532.1   |

17 rows

break t

- Supports non technical users with a no-code solution to gaining data insights using natural language.
- Provides a self-service access point to the data for follow-up questions without needing to involve a data analytics specialist.



# Data Science, Machine Learning, and Gen AI

End-to-End Support for Generative and Traditional AI Application Development

**Mosaic AI**



MLflow  
AutoML  
Feature Store  
Model Serving

Vector Search  
AI Playground  
Agent Framework  
Model Training

And much more!



# Databricks on Google Cloud | Lakehouse Platform

## Customers

Databricks on Google Cloud launched in 2021; 12,000+ global customers; 500+ 1M+; 1,000+ subscribers on GCP

## Integrations

Built on Google Compute Engine; Integrated offering with Google BigQuery, Pub/Sub, Looker, Vertex AI, Cloud Storage

**Deal Size:** Small \$200K, Medium \$400K, Large \$1M+, Xtra-Large \$3M+ TCV driving 1.5X GCP consumption in end customer tenant

### Business Challenge

- Data silos drive high operational costs; Inconsistent policies reduce trust in the data
- Need for Insights and AI to support decisions about customer & products (Relevance & personalization, Data monetization, Real-time data for decision making)
- Disparate tools slow down cross-team productivity



### Sales Play

- **Migration:** Databricks Migration from Azure/AWS to Google Cloud; Migrating on-premise data (Hadoop) to build lakehouse
- **Multi-Cloud:** Drive Google Cloud consumption in existing Databricks Multi-Cloud customer
- **Data Engineering:** Data Lake, Delta Lake, Lakehouse (Bronze->Silver -> Gold)
- **Data Science:** Data Prep, Exploration, Model Development & Training, Model Serving

### Why Databricks?

- Data Forward: Insights and AI support decisions about customers and products
- Data, analytics, and AI working in concert (Customer lifecycle, Product development, Workforce productivity, Business Operations)
- Data powers every aspect of the business

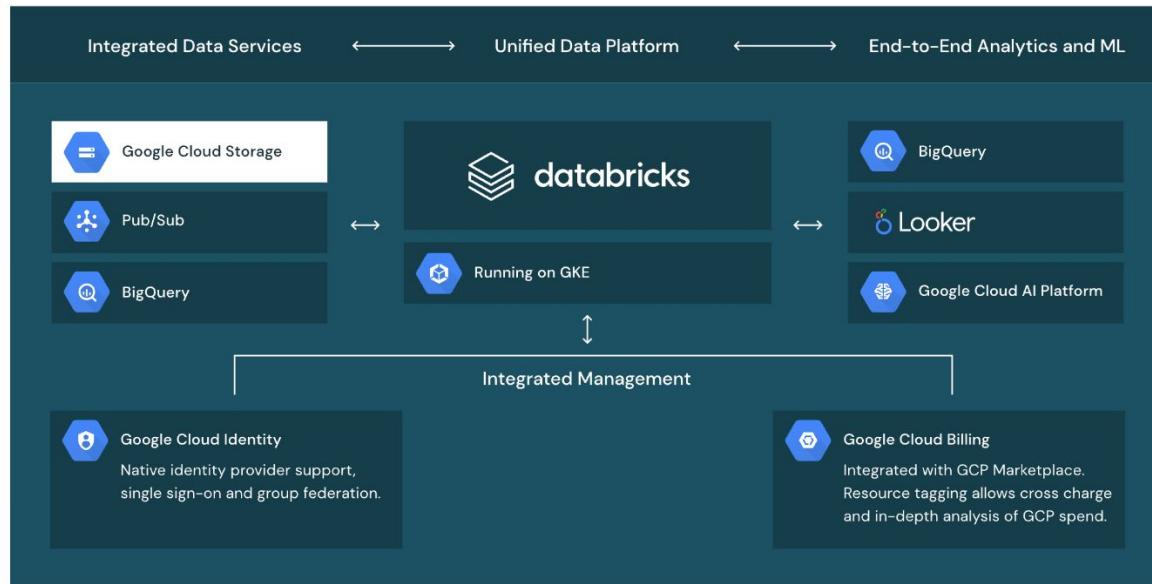
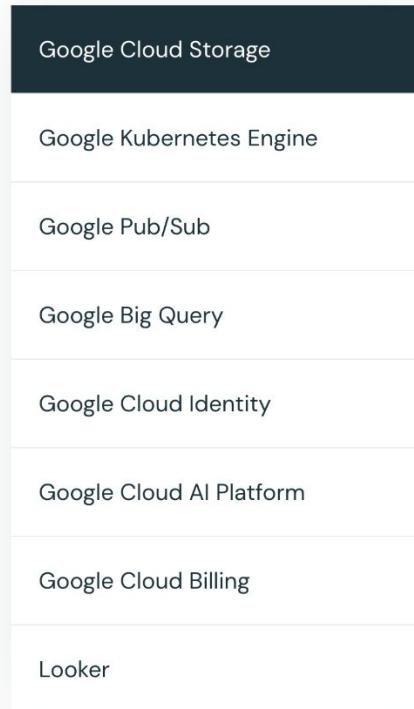


### Google + Databricks (Better Together)

**Platform:** Unified platform for all data use cases leveraging Google Cloud's superior performance, scale, and security from Google Cloud Identity, Google Cloud Storage and dedicated global network

**Interoperability:** Seamless interoperability across GCP Data & Analytics portfolio giving customers choice & flexibility

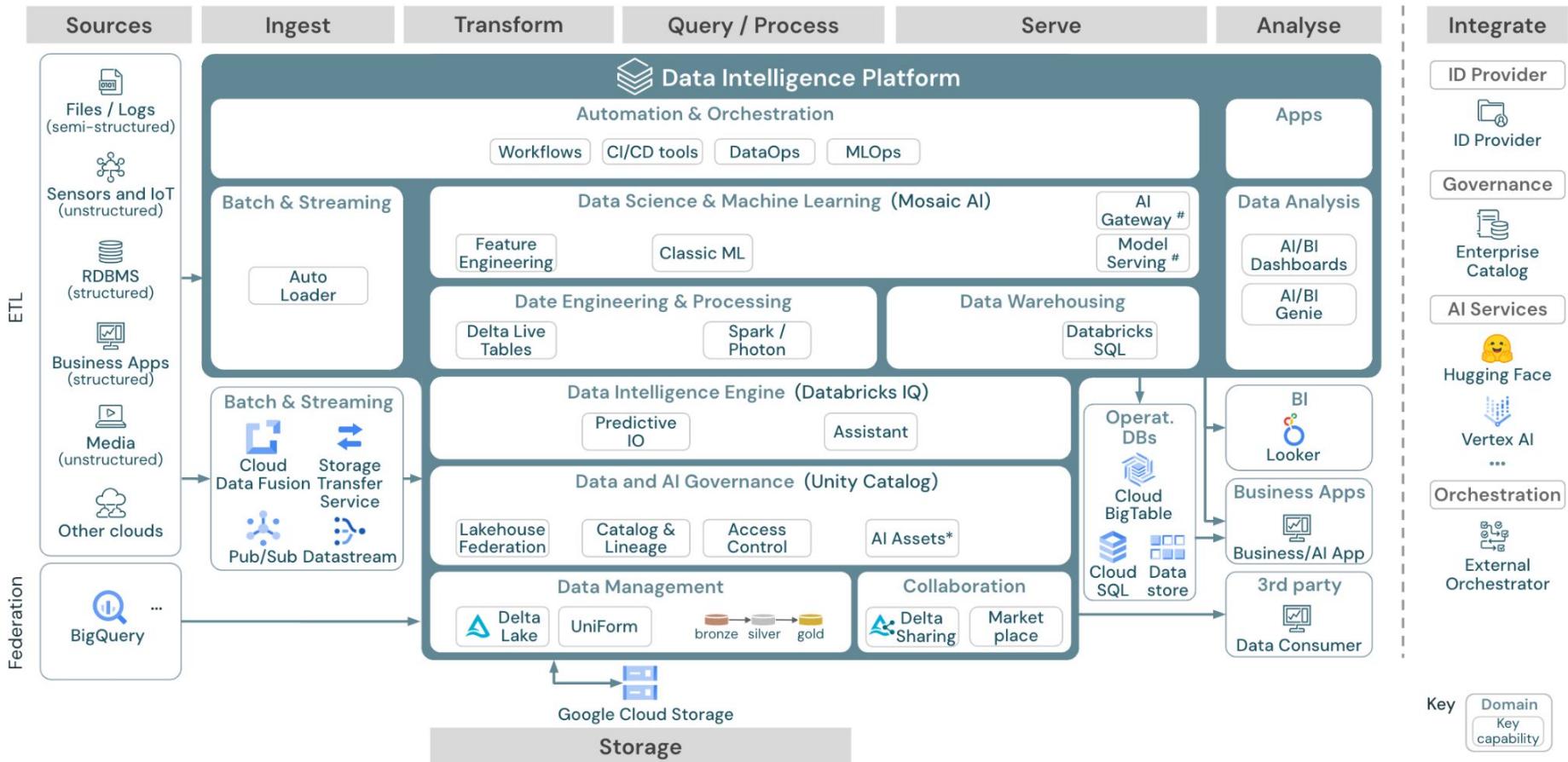
# Streamlined integration with Google Cloud



## Google Cloud Storage

Enable seamless read/write access for data in Google Cloud Storage (GCS) and leverage the Delta Lake open format to add powerful reliability and performance capabilities within Databricks.

[Databricks Site](#)



[Databricks GCP Ref Arch](#)

\* features, models, functions  
 # available in selected regions

# Business Drivers for Databricks on Google Cloud



## Infrastructure Cost Savings

Databricks on GCP processes workloads with **improved performance** reducing infrastructure costs

Use of GCS's **multiple storage tiers** and **auto-management options** to optimize cost efficiency.



## Performance & Scalability

Databricks benefits from GCS's **gRPC API**, which offers **fast and reliable data routing**

**Effortlessly transfer petabytes of data with event-driven transfers** for GCS, supporting **scalable and efficient** data movement.



## Streamlined GCP Integrations & Security

Built-in connectors for Cloud Storage, BigQuery, Pub/Sub, Dataflow, Looker

GCS provides **flexible location options** and **Managed Folders** for detailed access control, along with features like **Object Lock** and **Soft Delete** for robust data protection.



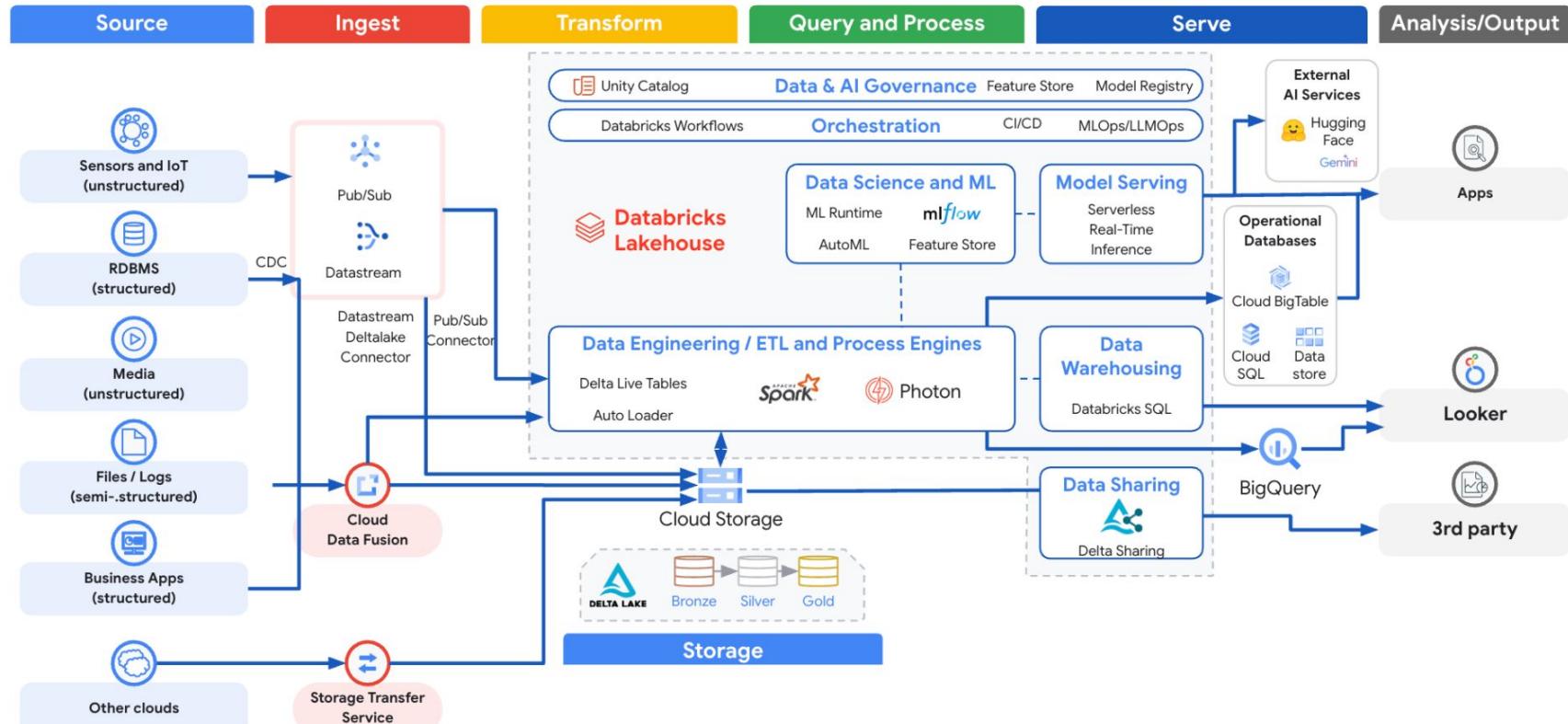
## Access Google Cloud AI / Gen AI capabilities

Access to Google's **state-of-the-art Vertex AI suite** with access to multi-modal GenAI capabilities plus **150+** models accelerating business outcomes

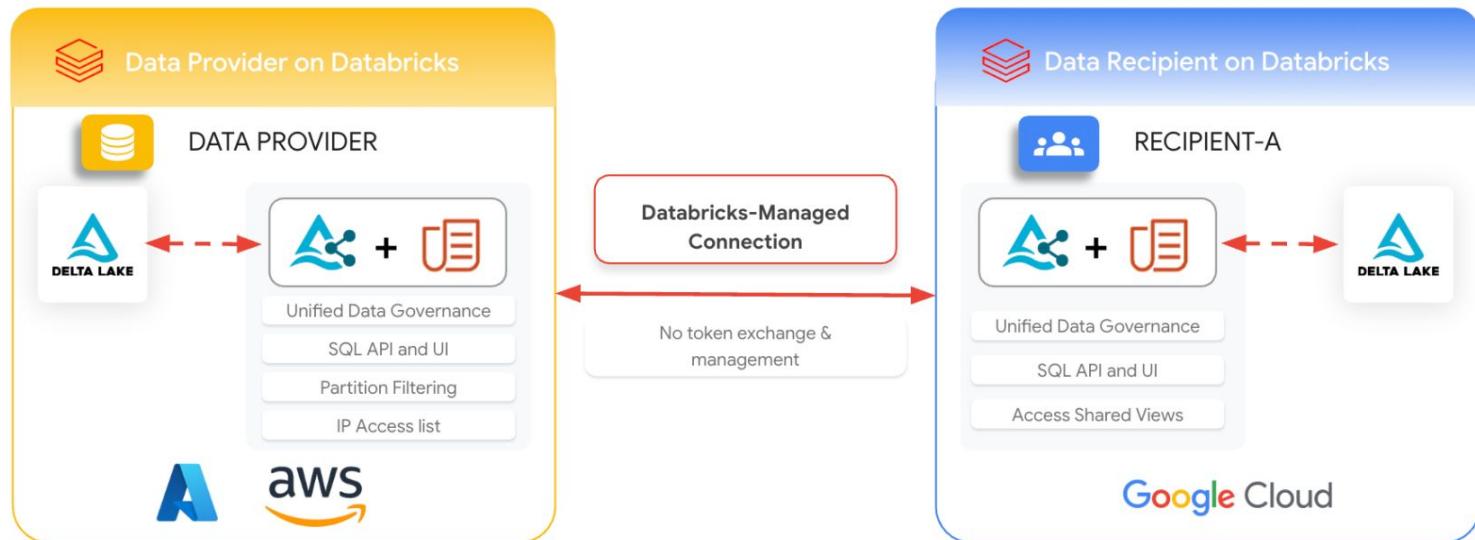
**Seamless AI at Scale** with **End-to-End MLOps** and **scalable** AI infrastructure to successfully deploy AI



# Databricks on Google Cloud Reference Architecture



# Databricks-to-Databricks Sharing



# BigQuery adds native support for Delta Lake



- **Seamless Integration:** Query Delta Lake tables directly in BigQuery from Google Cloud Storage and AWS S3 without data movement.
- **Unified Data Governance:** Advanced security with row and column-level controls, plus dynamic data masking on Delta Lake tables.
- **End-to-End Analytics:** Easily leverage Delta Lake data for ML, BI, and reporting with BigQuery ML and Vertex AI.
- **Streamlined Setup:** Create and query Delta Lake tables using familiar GoogleSQL syntax, similar to other BigQuery tables.

```
CREATE EXTERNAL TABLE `PROJECT_ID.DATASET.DELTALAKE_TABLE_NAME`  
WITH CONNECTION `PROJECT_ID.REGION.CONNECTION_ID`  
OPTIONS (  
    format = "DELTA_LAKE",  
    uris=['DELTA_TABLE_GCS_BASE_PATH']);
```





## TPC Benchmarks Overview

|  |      | TPC Enterprise Benchmark Standards |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
|--|------|------------------------------------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|
| Year   | 1988 | 1989                               | 1990 | 1991 | 1992 | 1993 | 1994 | 1995 | 1996 | 1997 | 1998 | 1999 | 2000 | 2001 | 2002 | 2003 | 2004 | 2005 | 2006 | 2007 | 2008 | 2009 | 2010 | 2011 | 2012 | 2013 | 2014 | 2015 | 2016 | 2017 | 2018 | 2019 | 2020 | 2021 | 2022 | 2023 | 2024 |
| TPC-A  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-App  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-B  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-C  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-D  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-DI   |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-DS   |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-E  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-H  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-R  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-VMS  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC-W  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC Express Benchmark Standards                  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPCx-AI  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPCx-BB  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPCx-HCI   |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPCx-HS  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPCx-HoT   |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPCx-V   |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| TPC Common Specifications                        |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| Pricing  |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| Energy   |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| - active benchmark                               |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| obsolete benchmark                               |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
| Benchmarks published since 2010 as of 12/31/2024 |      |                                    |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |      |
|  | 34   | 36                                 | 16   | 15   | 22   | 24   | 26   | 13   | 20   | 48   | 20   | 7    | 23   | 27   | 28   | 29   | 30   | 31   | 32   | 33   | 34   | 35   | 36   | 37   | 38   | 39   | 40   | 41   | 42   | 43   | 44   | 45   |      |      |      |      |      |

TPC-DS

TPC-DS is the de-facto industry standard benchmark for measuring the performance of decision support solutions including, but not limited to, Big Data systems. The current version is v2. It models several generally applicable aspects of a decision support system, including queries and data maintenance. Although the underlying business model of TPC-DS is a retail product supplier, the database schema, data population, queries, data maintenance model and implementation rules have been designed to be broadly representative of modern decision support systems.

- Examine large volumes of data
  - Give answers to real-world business questions
  - Execute queries of various operational requirements and complexities (e.g., ad-hoc, reporting, iterative OLAP, data mining)
  - Are characterized by high CPU and IO load
  - Are periodically synchronized with source OLTP databases through database maintenance functions
  - Run on “Big Data” solutions, such as RDBMS as well as Hadoop/Spark based systems

# **DEMO**

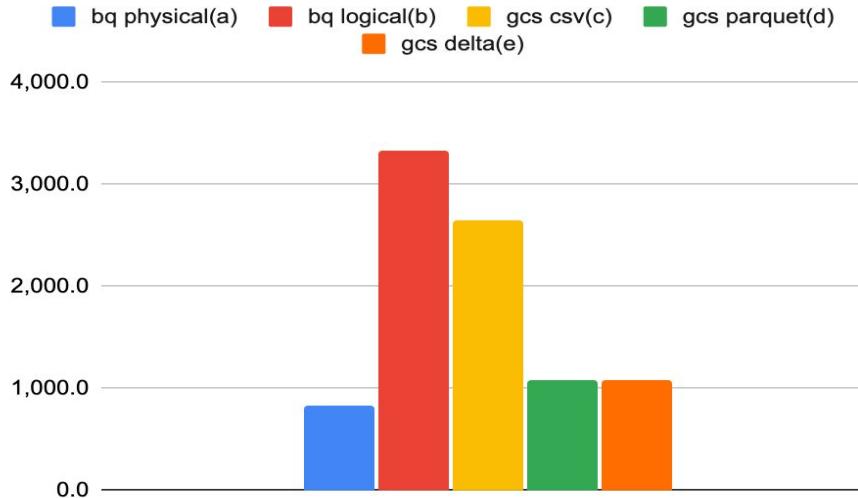
# why GCP - Storage

## Using Industry tests for DW: TPC-DS

### Storage - Comparing compression ratio

- for TPC-DS, default write in spark to parquet  
bq physical size is 76% of parquet size

| TPC-DS 2 TB tables (in MB) |                          |                      |             |            |                         |              |             |         |
|----------------------------|--------------------------|----------------------|-------------|------------|-------------------------|--------------|-------------|---------|
| table                      | bq physical(a) vs bq log | bq logical(b) vs csv | gcs csv(c)  | vs parquet | gcs parquet(d) vs delta | gcs delta(e) |             |         |
| Total                      | 831,009.8                | 25.00%               | 3,323,802.4 | 31.44%     | 2,643,298.0             | 76.79%       | 1,082,204.5 | 76.79%  |
| store_sales                | 342,067.5                | 23.21%               | 1,473,547.6 | 29.50%     | 1,159,662.81            | 72.9%        | 469,288.0   | 72.89%  |
| catalog_sales              | 282,048.9                | 26.29%               | 1,072,712.5 | 32.72%     | 862,064.06              | 81.2%        | 347,405.9   | 81.19%  |
| web_sales                  | 141,838.3                | 26.34%               | 538,472.1   | 32.89%     | 431,192.21              | 84.1%        | 168,686.3   | 84.08%  |
| store_returns              | 33,023.8                 | 26.77%               | 123,348.4   | 33.25%     | 99,316.53               | 65.3%        | 50,542.9    | 65.34%  |
| catalog_returns            | 21,302.9                 | 27.40%               | 77,736.3    | 35.38%     | 60,215.12               | 69.3%        | 30,747.2    | 69.28%  |
| web_returns                | 10,043.4                 | 28.91%               | 34,740.4    | 35.43%     | 28,348.78               | 67.8%        | 14,816.0    | 67.79%  |
| customer                   | 448.7                    | 32.75%               | 1,370.3     | 38.28%     | 1,172.24                | 90.8%        | 493.9       | 90.84%  |
| inventory                  | 144.4                    | 12.75%               | 1,132.6     | 20.64%     | 699.48                  | 121.8%       | 118.5       | 121.75% |
| customer_address           | 76.0                     | 13.05%               | 582.5       | 14.57%     | 521.74                  | 90.0%        | 84.4        | 90.03%  |
| customer_demographic       | 9.0                      | 7.38%                | 122.0       | 11.99%     | 75.09                   | 76.8%        | 11.7        | 76.77%  |
| item                       | 3.6                      | 23.66%               | 15.2        | 26.61%     | 13.53                   | 66.2%        | 5.4         | 66.08%  |
| date_dim                   | 1.7                      | 12.69%               | 13.6        | 17.65%     | 9.77                    | 82.6%        | 2.1         | 82.11%  |
| time_dim                   | 0.9                      | 13.49%               | 6.8         | 19.01%     | 4.79                    | 79.3%        | 1.1         | 78.82%  |
| call_center                | 0.0                      | 0.00%                | 0.0         | 0.00%      | 0.00                    | 70.10%       | 0.0         | 66.04%  |



# why GCP - Compute



| TPC-DS 99q 2TB               | minutes | tot      | m/\$ | Pricing  |     |      |         |               |      |         |        |         |      |          |     |          |           |             |         |         |         |          |                    |  |
|------------------------------|---------|----------|------|----------|-----|------|---------|---------------|------|---------|--------|---------|------|----------|-----|----------|-----------|-------------|---------|---------|---------|----------|--------------------|--|
|                              |         |          |      | Data     | BQ  | be   | BQ      | max           | BQ\$ | compute | vcpu/w | workers | cpu  | ram      | DBU | DataProc | avg/slot  | GB Processe | OperA   | OperB   | BQ Com  | DBK      | Compute Operations |  |
| BigQuery On-Demand           | 19.4    | \$110.97 | 0.2  | raw      | onD | -    | -       | -             | -    | -       | -      | -       | -    | -        | -   | 1,043.8  | 22,726.0  | -           | -       | 110.967 | -       | -        | -                  |  |
| BigQuery On-Demand           | 14.3    | \$104.22 | 0.1  | (a)      | onD | -    | -       | -             | -    | -       | -      | -       | -    | -        | -   | 766.8    | 21,345.2  | -           | -       | 104.224 | -       | -        | -                  |  |
| BigQuery Standard            | 95.9    | \$6.73   | 14.2 | (a)      | std | 100  | \$0.040 | -             | -    | -       | -      | -       | -    | -        | -   | 105.4    | 21,345.1  | -           | -       | 6.734   | -       | -        | -                  |  |
| BigQuery Standard            | 18.7    | \$6.31   | 3.0  | (a)      | std | 800  | \$0.040 | -             | -    | -       | -      | -       | -    | -        | -   | 506.8    | 21,345.1  | -           | -       | 6.306   | -       | -        | -                  |  |
| BigQuery Standard            | 13.1    | \$6.03   | 2.2  | (a)      | std | 1600 | \$0.040 | -             | -    | -       | -      | -       | -    | -        | -   | 690.2    | 21,345.1  | -           | -       | 6.033   | -       | -        | -                  |  |
| BigQuery Entp                | 16.8    | \$7.81   | 2.2  | (a)      | ent | 1600 | \$0.036 | -             | -    | -       | -      | -       | -    | -        | -   | 774.3    | 21,345.1  | -           | -       | 7.808   | -       | -        | -                  |  |
| BigQuery Entp                | 21.5    | \$12.92  | 1.7  | external | std | 1600 | \$0.036 | -             | -    | -       | -      | -       | -    | -        | -   | 999.9    | 33,617.9  | 302         | 412,629 | 12.919  | -       | -        | \$0.1666           |  |
| Spark                        | 533.6   | \$9.28   | 57.5 | external | -   | -    | -       | n2-standard-4 | 4    | 4       | 20     | 75      | 3.60 | -        | -   | -        | 418,523   | 944,381     | -       | \$4,937 | \$1,874 | \$2,4704 |                    |  |
| Spark + photon               | 129.7   | \$6.00   | 21.6 | external | -   | -    | -       | n2-standard-4 | 4    | 4       | 20     | 75      | 9.00 | -        | -   | -        | 384,345   | 819,750     | -       | \$3,255 | \$0,494 | \$2,2496 |                    |  |
| Spark + photon + cache:false | 205.3   | \$7.77   | 26.4 | external | -   | -    | -       | n2-standard-4 | 4    | 4       | 20     | 75      | 9.00 | -        | -   | -        | 213,505   | 2,724,890   | -       | \$4,957 | \$0,654 | \$2,1575 |                    |  |
| Spark                        | 361.9   | \$10.76  | 33.6 | external | -   | -    | -       | n1-standard-4 | 4    | 8       | 36     | 270     | 6.39 | -        | -   | -        | 471,532   | 920,119     | -       | \$6,021 | \$2,015 | \$2,7257 |                    |  |
| Spark                        | 505.6   | \$14.45  | 35.0 | external | -   | -    | -       | n1-standard-8 | 8    | 4       | 20     | 75      | 7.05 | -        | -   | -        | 373,415   | 789,032     | -       | \$9,175 | \$3,093 | \$2,1827 |                    |  |
| Spark Dataproc               | 315.9   | \$11.48  | 27.5 | external | -   | -    | -       | n1-standard-8 | 8    | 8       | 72     | 270     | -    | \$3,7908 | -   | -        | 1,234,524 | 4,819,254   | -       | -       | \$3,378 | \$8,1003 |                    |  |
| Spark Dataproc               | 301.5   | \$11.19  | 27.0 | external | -   | -    | -       | n1-standard-4 | 4    | 16      | 68     | 255     | -    | \$3,4170 | -   | -        | 1,217,734 | 4,752,216   | -       | -       | \$3,197 | \$7,9896 |                    |  |

- (a few tables clustered + small tables w/partition (<1MB)

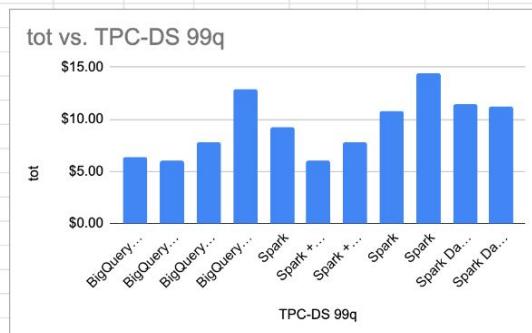
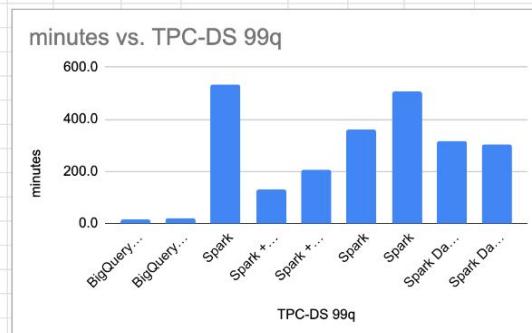
- external = parquet dump from bq

- no tuning

- only a few corrections on queries to run with SparkSQL

pvp

|                        |          |                       |
|------------------------|----------|-----------------------|
| Databricks DBU Premium | \$0.1500 | h                     |
| BQ on Demand           | \$5.0000 | TB                    |
| BQ Editions Std        | \$0.0400 | h                     |
| BQ Editions Entp 3y    | \$0.0360 | h                     |
|                        | \$0.06   | (-40% 3y reservation) |
| n2-standard-4          | \$0.2137 | h                     |
|                        | \$0.041  | (spot)                |
| n1-standard-4          | \$0.1562 | h                     |
|                        | \$0.036  | (spot)                |
| n1-standard-8          | \$0.4184 | h                     |
|                        | \$0.071  | (spot)                |
| OperA                  | \$0.0050 | p/1000                |
|                        | list     |                       |



# **Thank You!**

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# How is Google Cloud Platform Different

**Thank You!**



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