

Data Warehouse on  Cloud

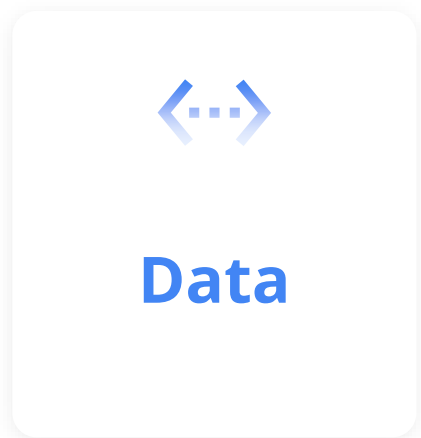
수많은 기업들이 BigQuery를 활용하는 이유는?!

Ver. 1.0
2022-07-14
update

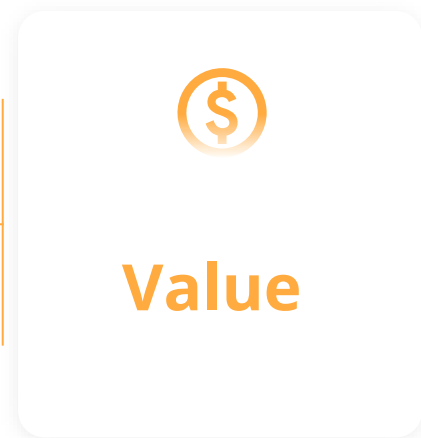
AGENDA

- 01 | **Intro**
Smart Analytics with BigQuery
- 02 | **Session 1**
Common Data Analytics Pipeline on GCP
- 03 | **Session 2**
integration BigQuery with GWS & GA
- 04 | **Summary & QnA**

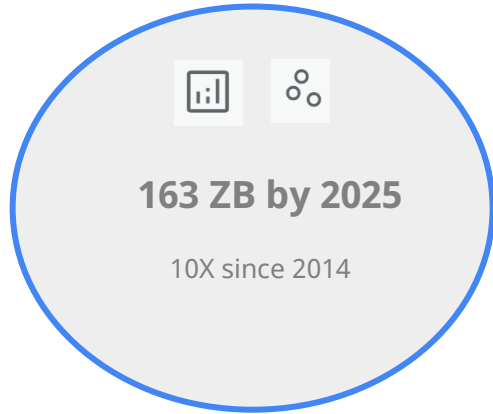
Closing the Data Value Gap



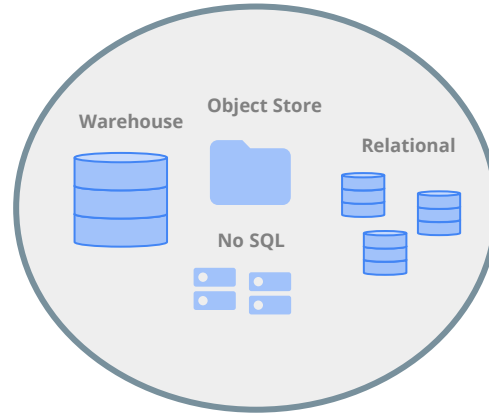
68%
of companies are
unable to realize
tangible &
measurable **Value**
from **Data**.



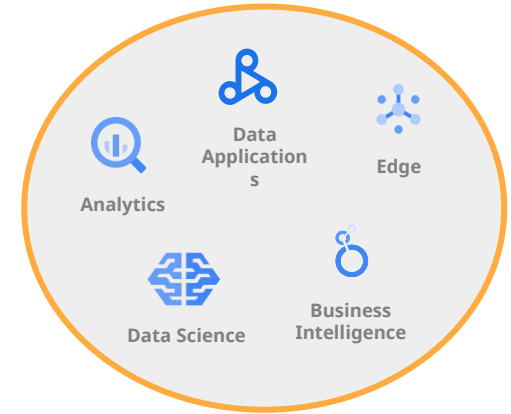
Accelerating trends are pushing the boundaries of existing data analytics stack



Data volume, velocity & variety continues to grow at rapid pace

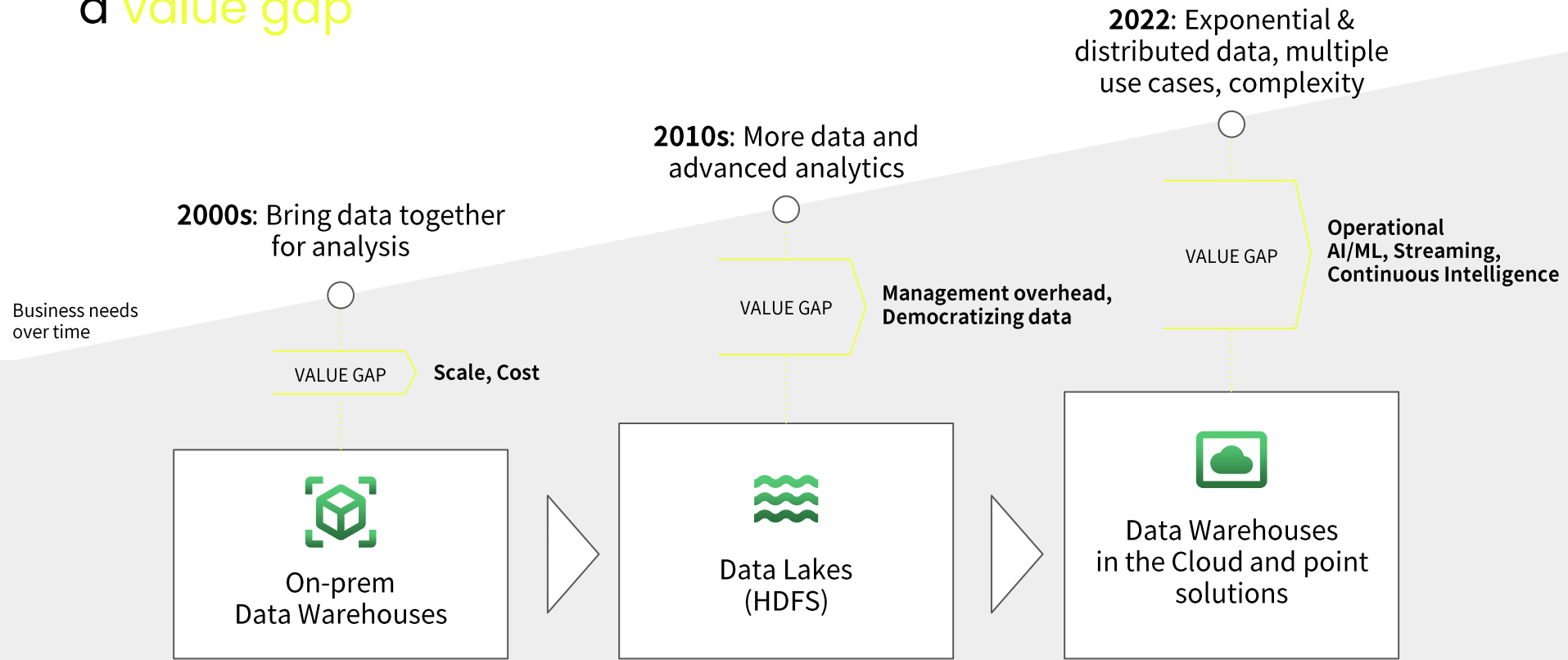


Data is increasingly distributed across storage systems, clouds, regions

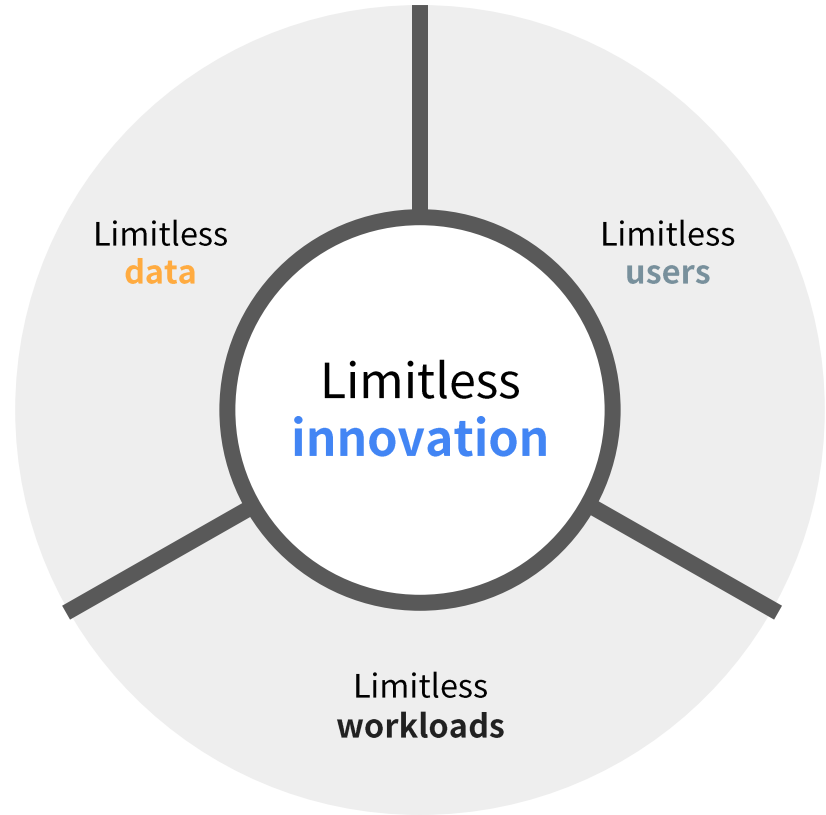


Data is serving more users & use cases than ever before

The evolution of business needs has created a **value gap**



Moving to the future
requires a **limitless**
data ecosystem
to drive **innovation**

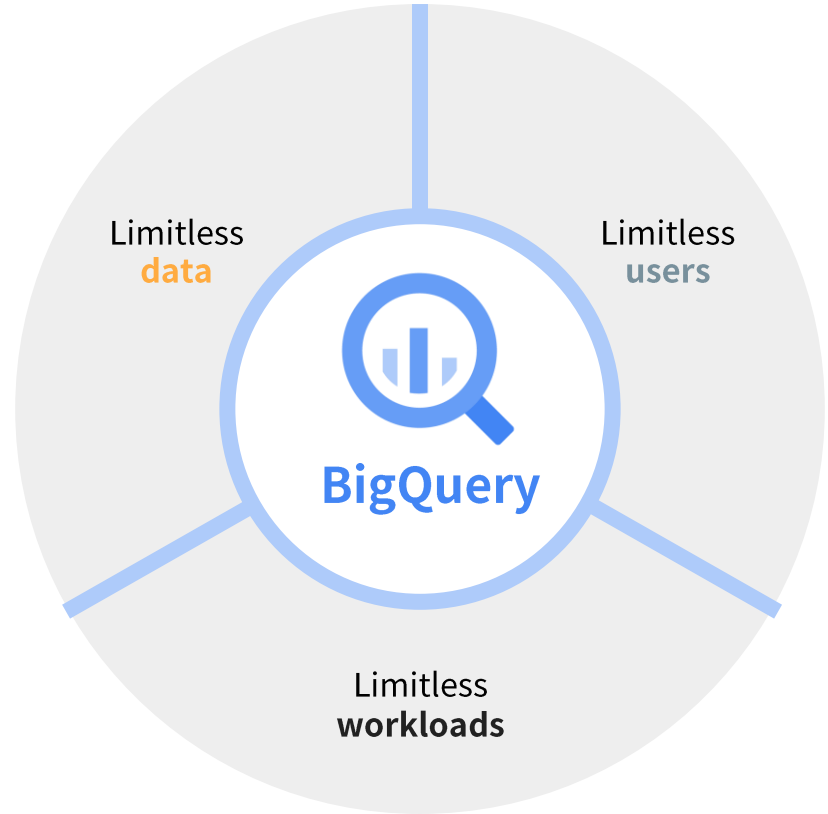


Google



Our mission is to **organize** the world's **information** and make it **universally accessible** and **useful**.

Customers unlock
innovation with
Google's **Data Cloud**



Google BigQuery

Data warehouse with customers ranging from TB to 100+ PB



Cloud-scale enterprise data warehouse



Standard SQL(ANSI 2011) with DML Support



Encrypted, durable, highly available



Serverless platform

Unique



Real-time insights

Unique



Built-in **ML**

Unique




Insights for everyone

Unique



Continually improving performance



 **Graham Polley** @polleyg Following ▼

Holy shit balls! You're weren't kidding about those #BigQuery performance improvements @thetinot.. 🚀

```
12 LANGUAGE
13 ORDER BY
14 total_views DESC;
```

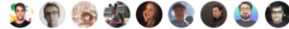
No cached results

▶ Run ⬇ Save query ⋮ Save view 🕒 Sch


Query results ⬇ SAVE RESULTS ▼


Query complete (16.7 sec elapsed, 4.1 TB processed)

11:40 PM - 7 Apr 2019

15 Retweets **76 Likes** 

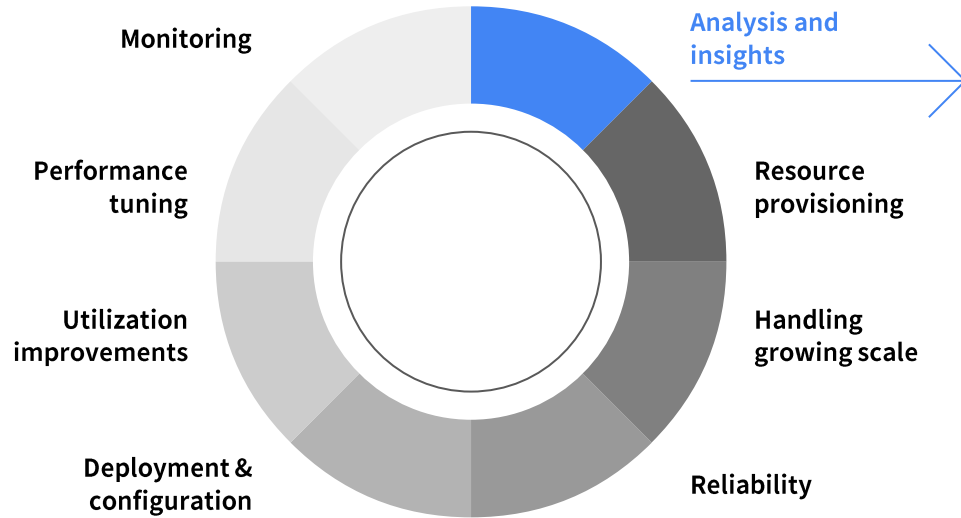
💬 7 ↻ 15 ❤️ 76 ✉



 **Graham Polley** @polleyg · Apr 7 ▼
For comparison, this query used to take 50s-70s in the past.

Serverless big data analytics

Serverless big data analytics



Serverless data analytics

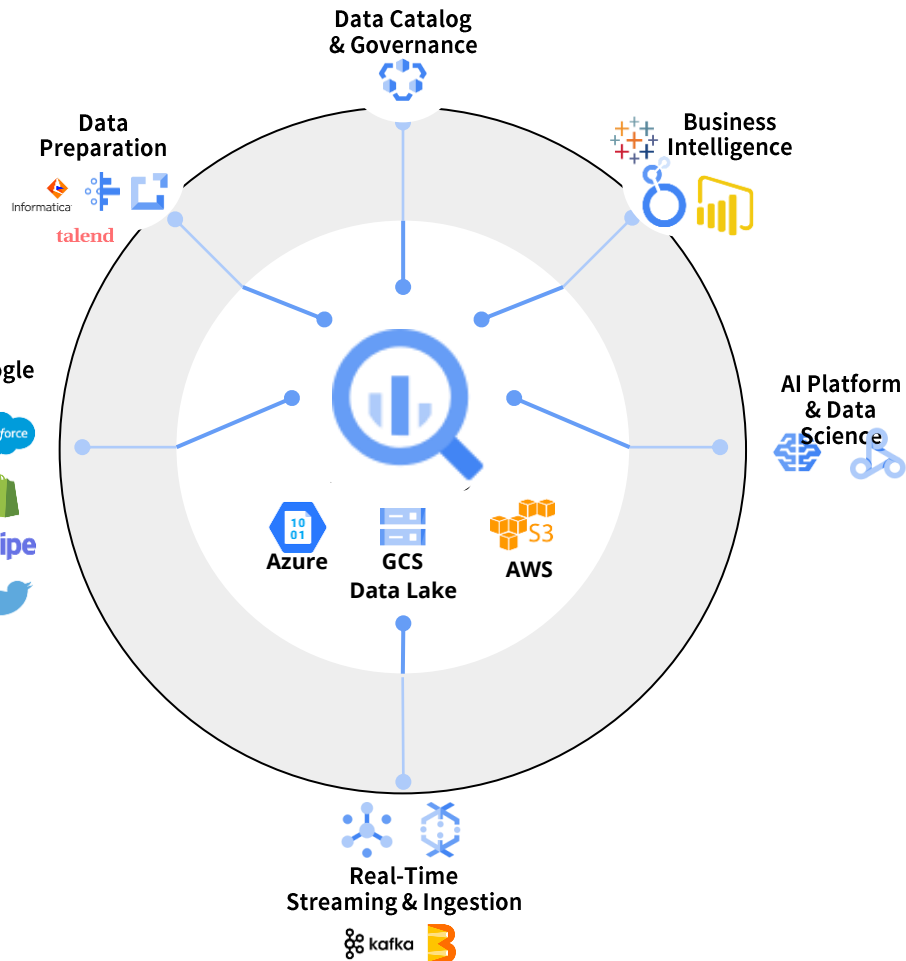


Limitless Data & Workloads

Intelligent: Self-optimizing architecture and ML “built-in” for broader adoption of advanced analytics

Simplified: Real-time and batch. Structured, unstructured, and spatial. In GCP, AWS and Azure

Reliable: Global scale at a predictable cost. Industry’s best reliability with 99.99% SLA is 10x more reliable than other clouds



BigQuery Omni: Overview

BigQuery Control Plane
(UI / API / CLI) on Google Cloud

Google Cloud region

BigQuery Compute clusters (Dremel)



Distributed Memory
Shuffle Tier

Petabit Network



BigQuery Storage



Decoupled
compute & storage
for maximum
flexibility

AWS region (Omni)

BigQuery Compute clusters (Dremel)



Distributed Memory
Shuffle Tier



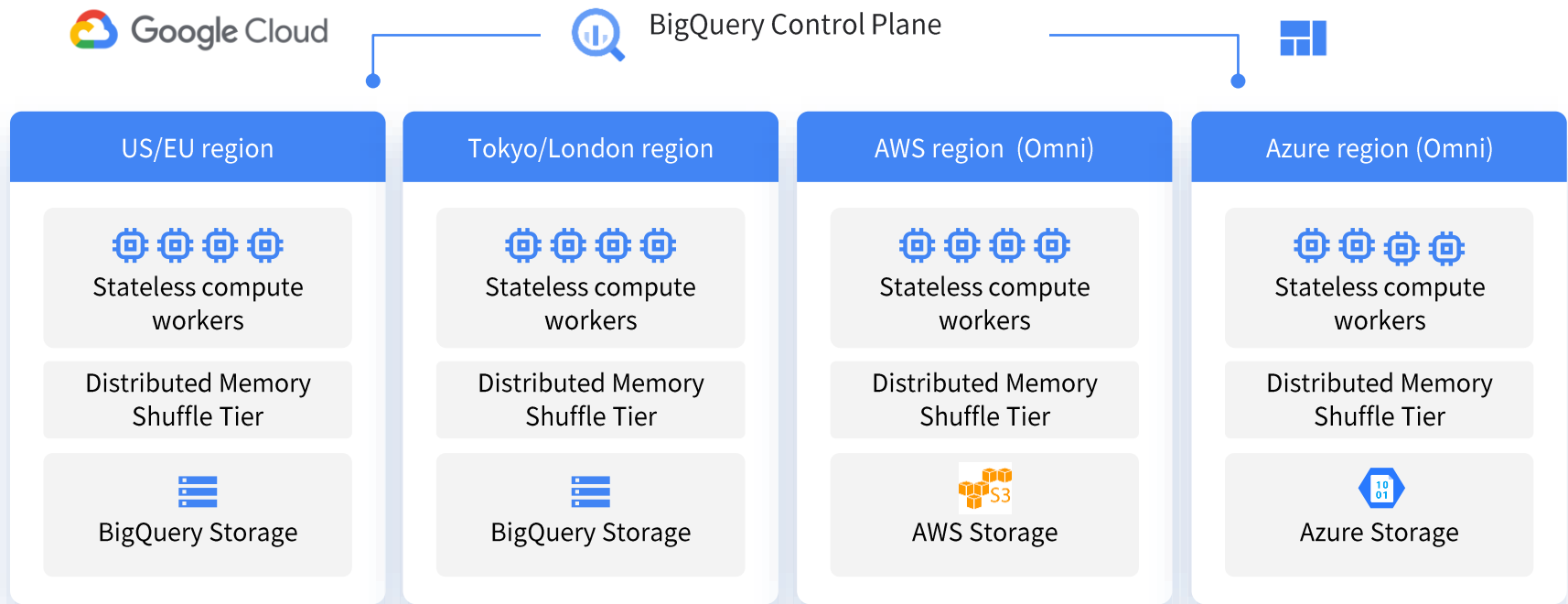
powered by Anthos
technology

AWS APIs



Customer S3 Storage

Embrace a flexible, multi cloud analytics solution with BigQuery Omni




go/BigQueryOmni to learn more!

Cross-cloud data transfer

Preview Coming Soon!

 Google Cloud

 BigQuery Control Plane

GCP Region



Stateless compute workers

Distributed Memory Shuffle Tier



BigQuery Storage

LOAD DATA FROM FILES
[Files on other cloud]



Use SQL to enable cross-cloud functionality

AWS / Azure Region



Stateless compute workers

Distributed Memory Shuffle Tier



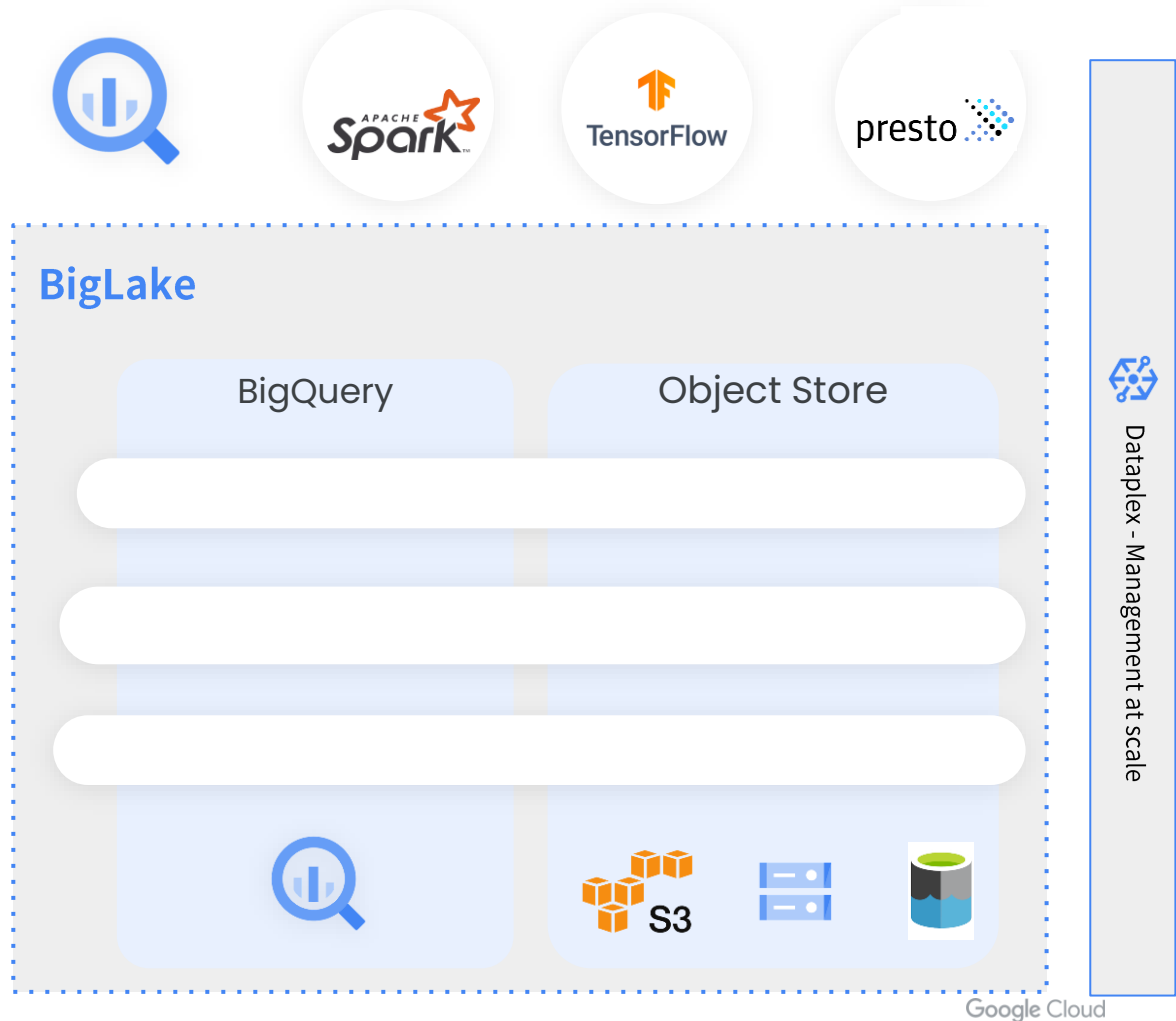
AWS / Azure Storage

BigLake

Built on years of investment in BigQuery, BigLake is a **storage engine** that unifies data warehouses and lakes, by providing uniform fine-grained access control, performance acceleration across multi-cloud storage and **open** formats

Open API interface

Provides an Open API interface for query engines spanning across Google Cloud and open source runtimes to access your distributed data with consistent security & governance controls



Dataplex - Management at scale

BigQuery Security & Governance for Data Lakes



```
CREATE EXTERNAL TABLE bq_demo  
WITH CONNECTION 'service account'  
OPTIONS (uris=["gs://mybucket"])  
  
CREATE ROW ACCESS POLICY row_access  
ON bq_demo  
GRANT TO group FILTER USING (c1='filter')
```

SQL

Authorized external tables extend the data management features of BigQuery tables to data lakes. Configured through a service account connection, authorized external tables allow you to manage end users without requiring you to grant permissions on files residing in data lakes.

Fine grain security for your data in object stores

Row and column level security for authorized external tables defined on Google Cloud Storage, Amazon S3 and Azure data lake storage.

Better interoperability with access through open APIs

Row, column level security **consistently enforced** when using BigQuery storage APIs from Spark and other Dataproc supported OSS engines.

Simplify user access management

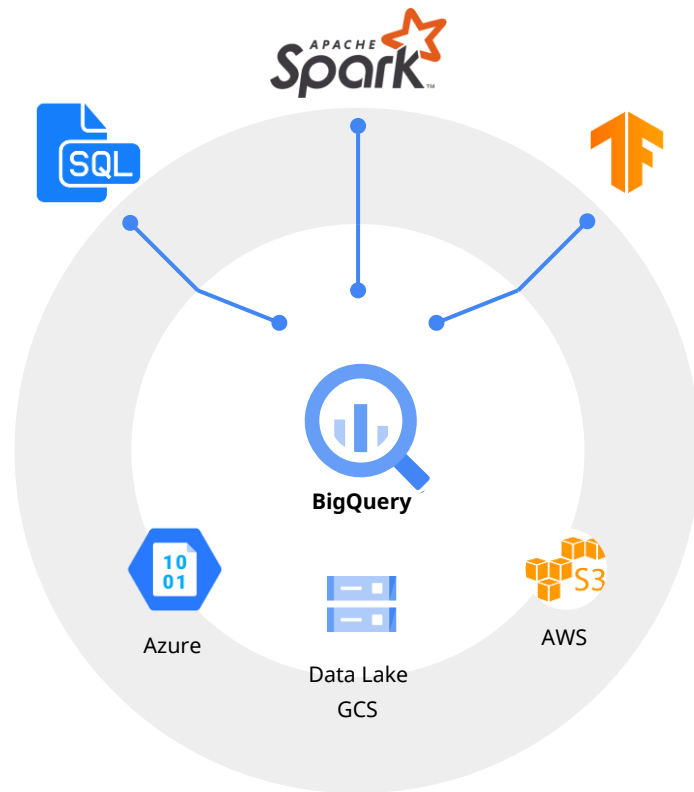
Manage access similar to BQ tables **without needing to manage file level permissions.**

Google's **open data platform** around BigQuery delivers value from data

Intelligent: Self-optimizing architecture and ML built-in for broader adoption of advanced analytics.

Simplified: Real-time and batch. Structured, unstructured, and spatial. In GCP, AWS and Azure.

Reliable: Global scale at a predictable cost. Industry's best reliability with 99.99% SLA.



Announcing



Spark on Google Cloud

Industry's first [Serverless Spark](#),
[integrated](#) with the best of Google Cloud.
Run and write spark where you need it across all
use-cases: [ETL](#), [data science](#) and [exploration](#).

Industry's First Serverless Spark for All Workloads

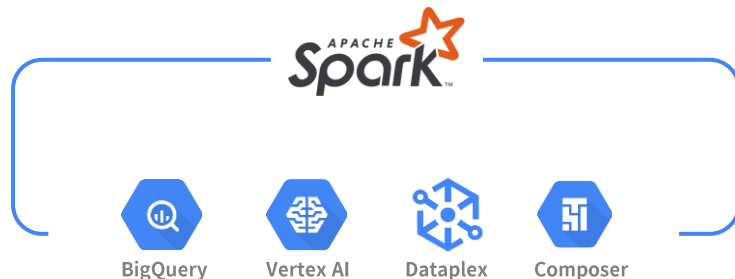
Auto-scale without any manual infrastructure provisioning or tuning for Spark. Empowers customers to shift from managing clusters to workloads.

Pervasive Spark Experience

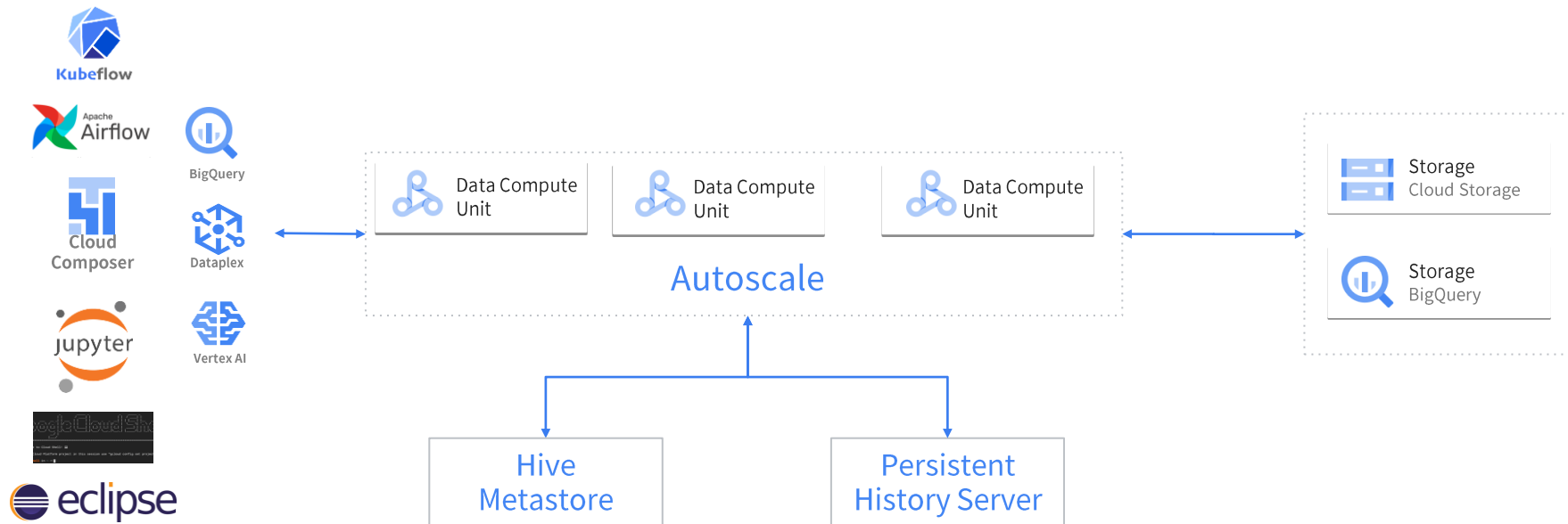
Connect, analyze and execute Spark jobs from BigQuery, Vertex AI or Dataplex in 2 clicks, without any custom integrations, using the best of Google-native and Open Source tools.

Flexibility of Consumption

One size does not fit all. Choose between Serverless, Google Kubernetes Engine (GKE), and compute clusters for your Spark applications



Serverless Spark: no clusters to create or manage



```
gcloud beta dataproc batches submit pyspark \  
  --project=${PROJECT} \  
  --region=${REGION} \  
  gs://dataproc-spark-preview/spark_terasort.py \  
  -- --gbs 100 --partitions 1000 \  
  --base_dir gs://${BUCKET}/spark_terasort_100/${RANDOM} \  
  --phases gen sort
```

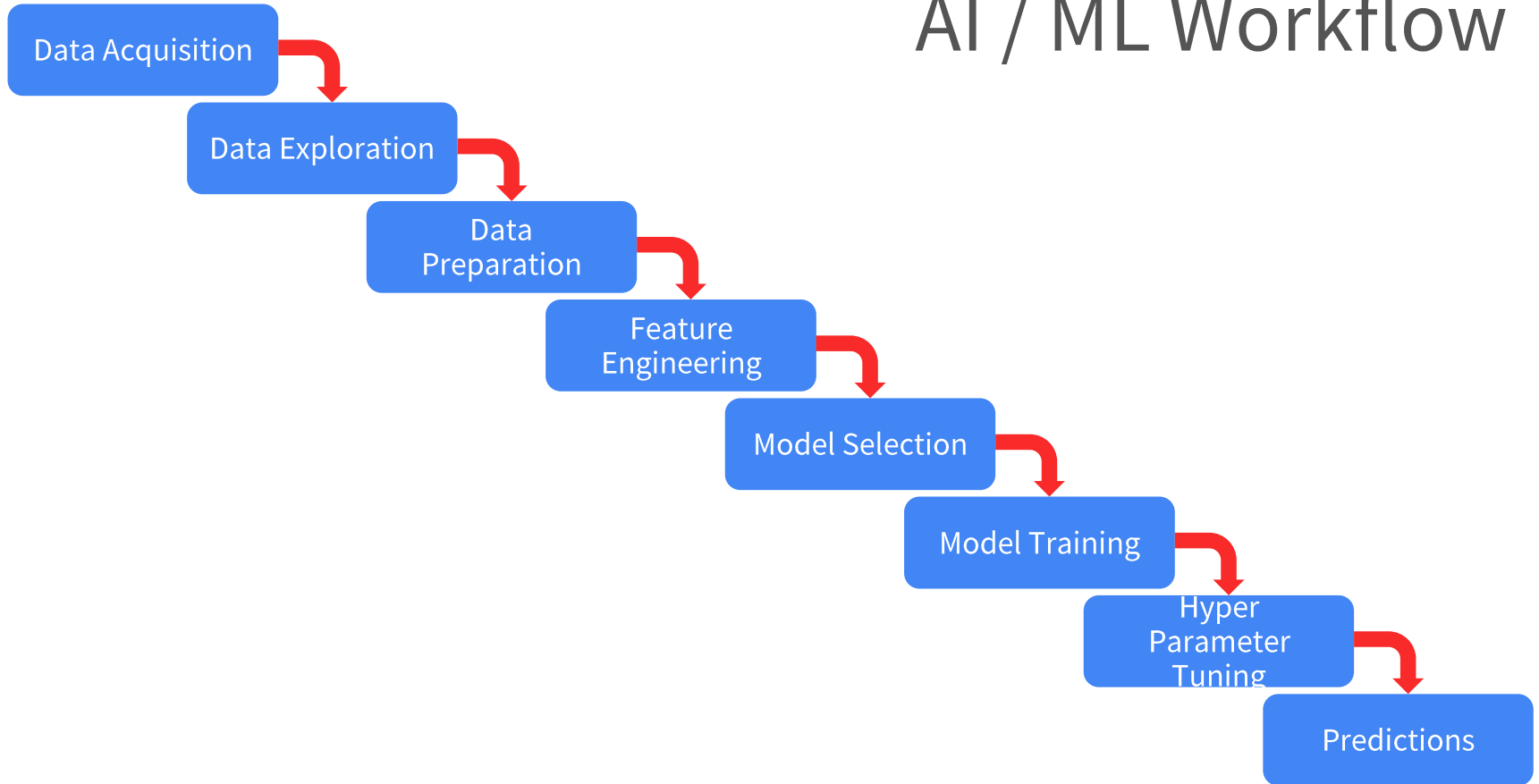


BQML

In-Database ML engine



AI / ML Workflow



Pipeline for ML workflow



Define objectives



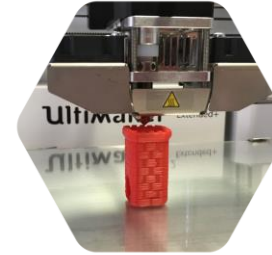
Collect data



Understand and prepare the data



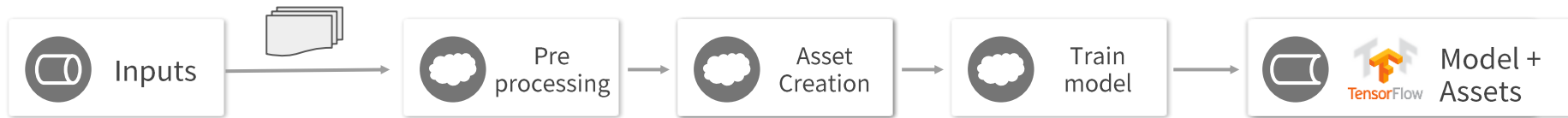
Create the model



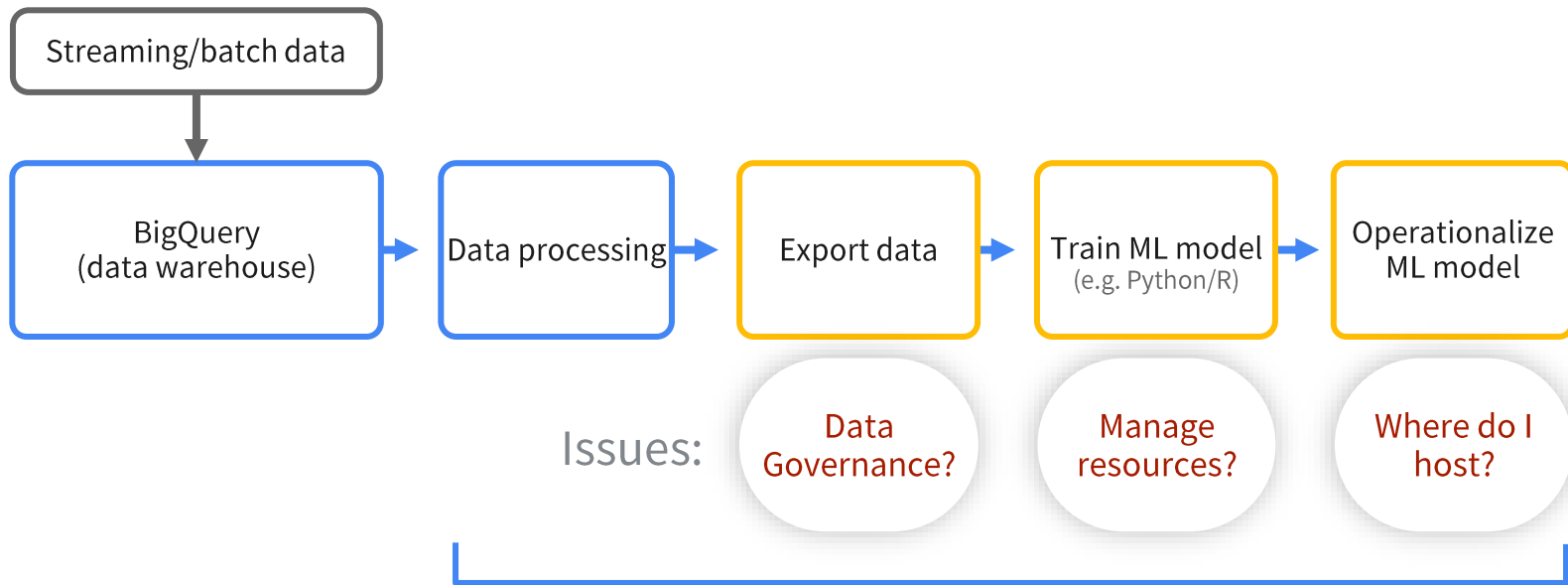
Refine the model



Serve the model



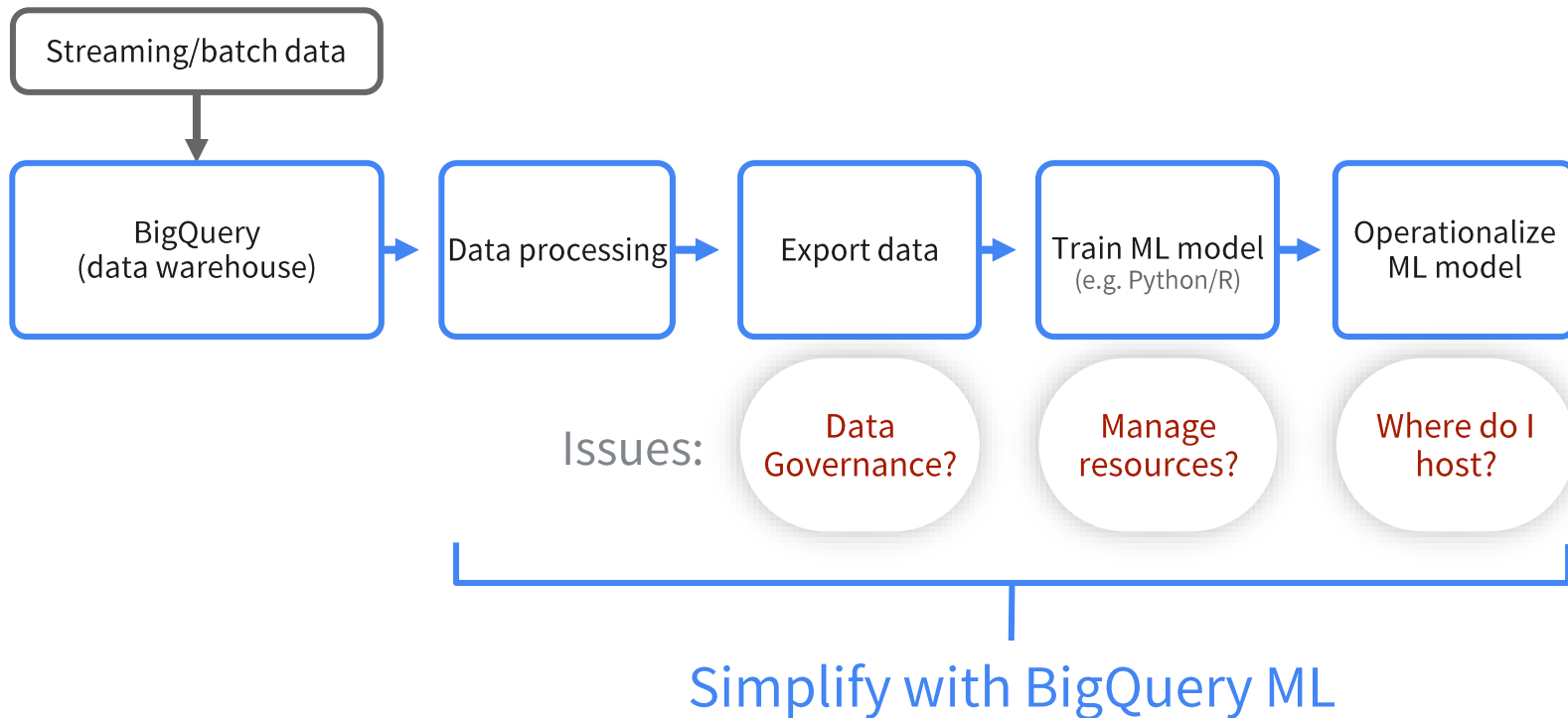
Typical ML Workflow



Multiple products & roles can lead to unnecessary complexity & costs



Typical ML Workflow

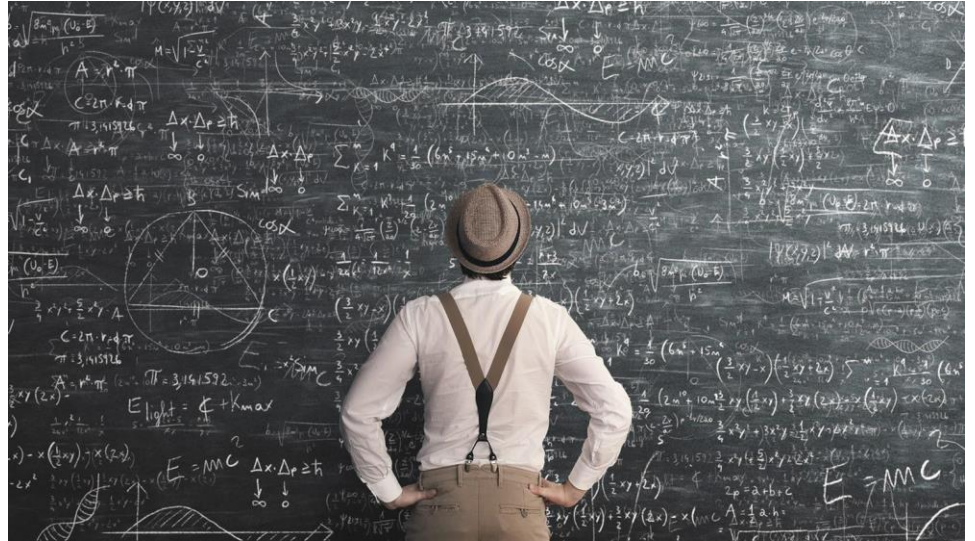


Key challenges affecting ML

Complex and time consuming to set up data pipelines and multiple tools

Data governance required across tools

Needs advanced data science and programming skill set



BigQuery ML makes machine learning super easy

The easiest way to unlock powerful AI/ML capabilities in BigQuery

- **Train and deploy** ML models in SQL
- **Execute** ML workflows without moving data from BigQuery
- **Automate** common ML tasks
- **Simplify** infrastructure management, security & compliance



Behind the scenes - BigQuery ML

Through SQL and within BigQuery

Leverage BigQuery's processing power to build a model

Auto-tuned learning rate

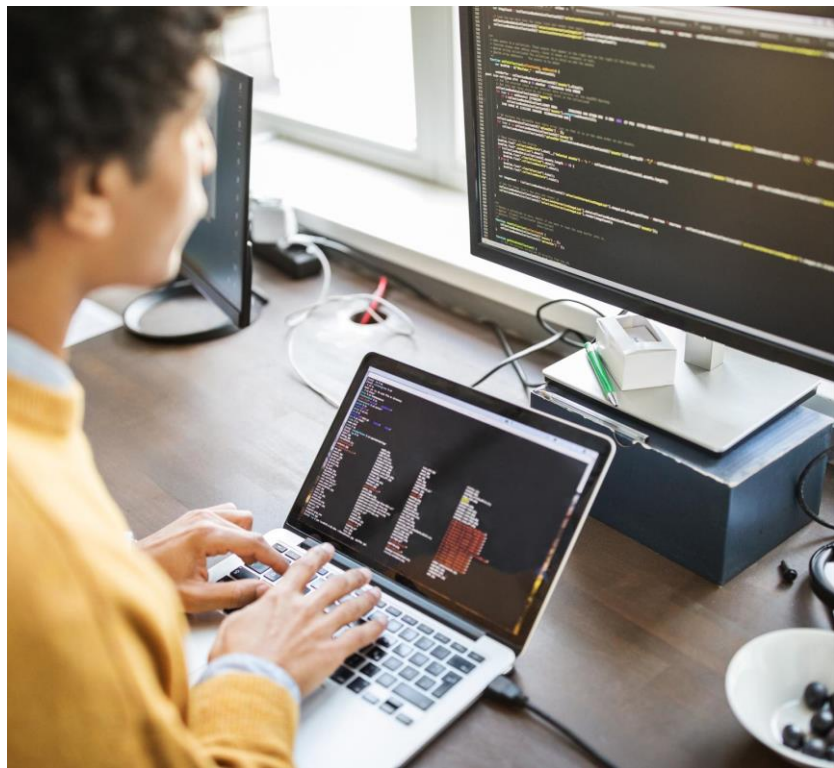
Auto-split of data into training and test

Null imputation

Standardization of numeric features

One-hot encoding of strings

 Class imbalance handling



Build and train with Create Model

Linear Regressor: given the weather, how many NYC Citibikes are in circulation?



```
CREATE OR REPLACE MODEL `sara-  
bqml.bikes.num_trips`
```

```
OPTIONS
```

```
(model_type='linear_reg',  
labels=['num_trips']) AS (
```

```
SELECT
```

```
temp,
```

```
visib,
```

```
prcp,
```

```
fog,
```

```
rain_drizzle,
```

```
snow_ice_pellets,
```

```
num_trips
```

```
FROM
```

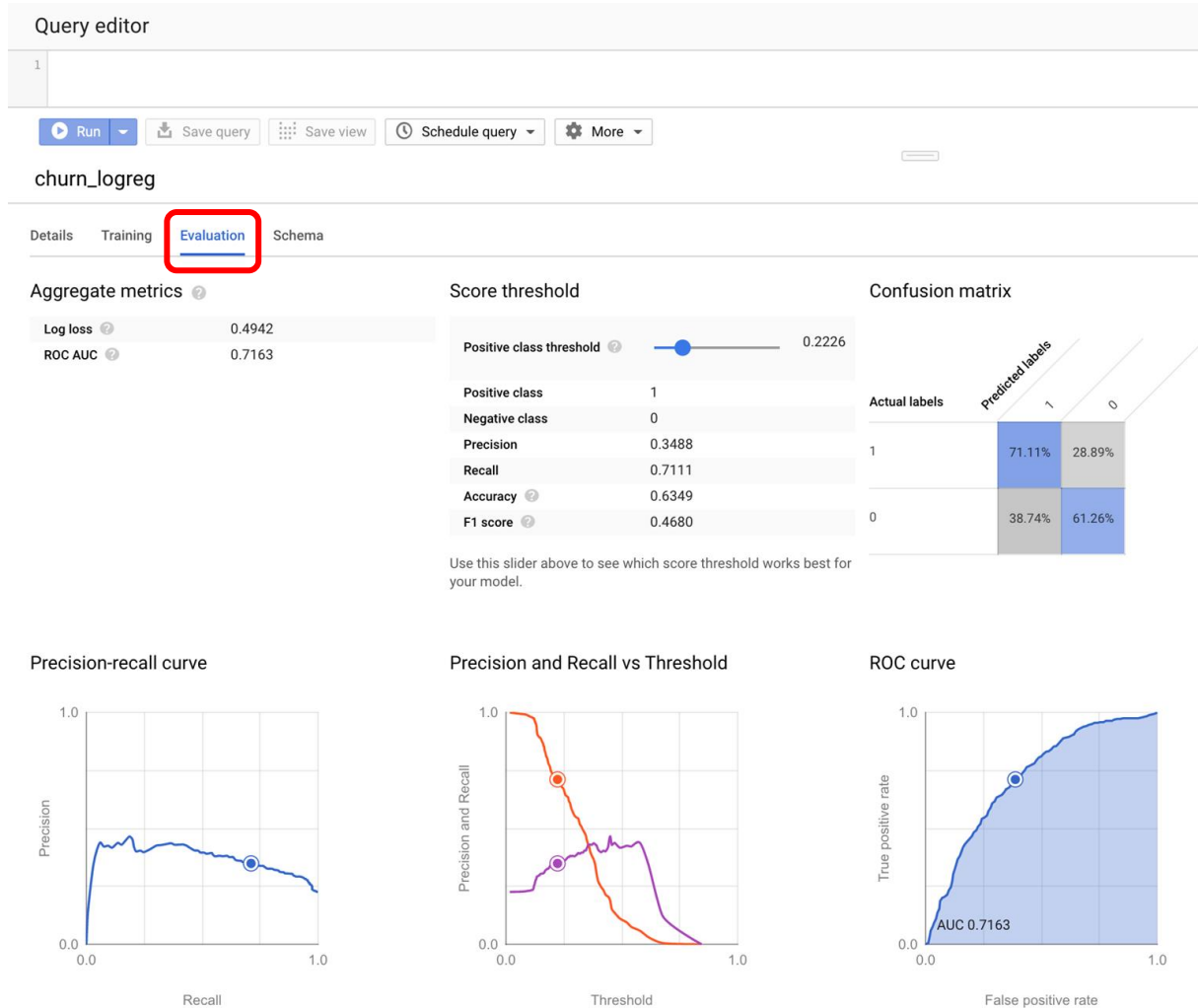
```
`sara-bqml.bikes.trips_weather_nyc18`)
```

Use the model with ML.PREDICT

```
SELECT
  predicted_num_trips,
  num_trips
FROM
  ML.PREDICT(MODEL `sara-
  bqml.bikes.num_trips_nyc`,
  (
```

```
    SELECT
      temp,
      visib,
      prcp,
      fog,
      rain_drizzle,
      snow_ice_pellets
    FROM
      `sara-bqml.bikes.2017_nycbikes_test` ))
```

Evaluate the model with MLEVALUATE

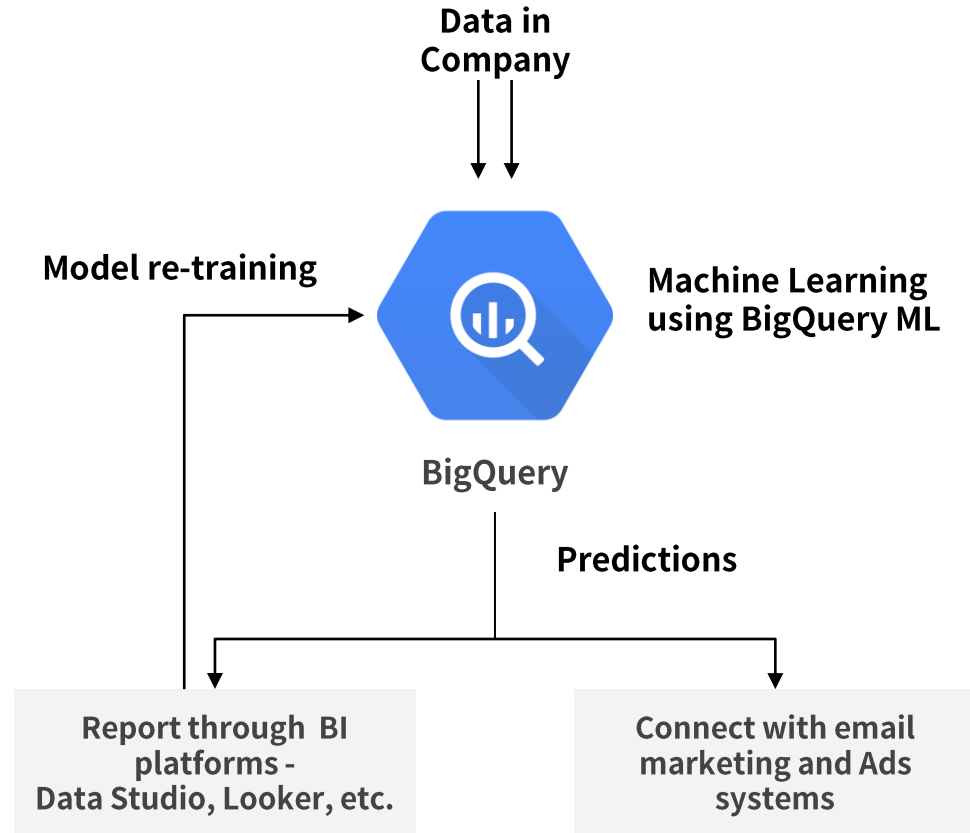


Evaluate the model with MLEVALUATE

```
SELECT
  *
FROM
  ML.EVALUATE(
    MODEL [DATASET].churn_logreg
  )
```

Row	precision	recall	accuracy	f1_score	log_loss	roc_auc
1	0.4367816091954023	0.10555555555555556	0.7672521957340025	0.17002237136465326	0.49420724043767983	0.7162997002997002

ML without data transfer



BigQuery ML supported models and features

Classification

- Logistic regression
- DNN classifier (TensorFlow)
- XGBoost
- AutoML Tables
- Wide and Deep NNs

Regression

- Linear regression
- DNN regressor (TensorFlow)
- XGBoost
- AutoML Tables
- Wide and Deep NNs

Other Capabilities

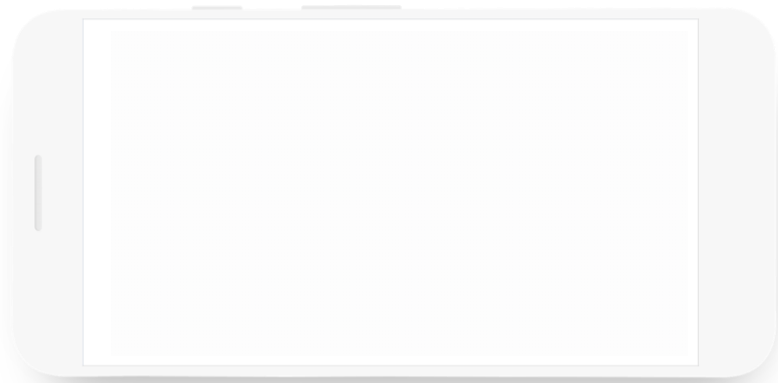
- k-means clustering
- Time series forecasting (ARIMA+)
- Recommendation: Matrix factorization
- Anomaly Detection
- Principal Component Analysis
- Hyperparameter tuning using Cloud AI Vizier
- Explainable AI

ML Ops

- Importing TensorFlow models for batch prediction
- Exporting models from BigQuery ML for online prediction



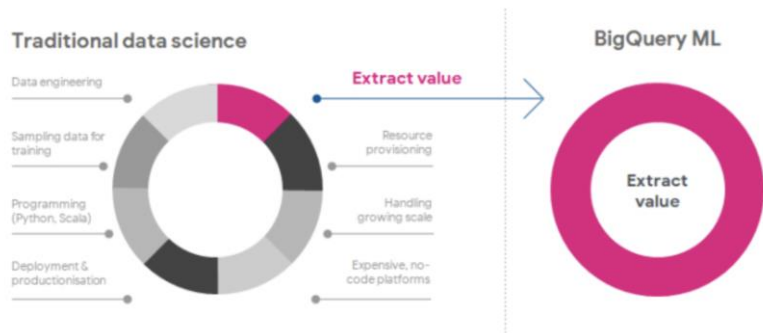
MLOps tooling for BigQuery ML



Manage production models using **Vertex AI Model Registry**

Use models in **BigQuery** or deploy behind a **REST endpoint**

Orchestrate data prep, training, and eval with **Vertex AI Pipelines**



Manage production models using **Vertex AI Model Registry**

- **Register, organize, track, and version** your trained and deployed ML models supporting AutoML, BQML, and custom workflows
- **Store model metadata** and runtime dependencies for deployability.
- **Integrate with the model evaluation and deployment** capabilities and track online and offline evaluation metrics for the models. [on roadmap]
- **Govern the model launch process:** review, approve, release, and roll back. [on roadmap]

Filter models

Name	Deployment status	Description	Default version	Type	Source	Updated ↑
▶ music-ranking	✔ Deployed on Vertex AI	Lorem ipsum dolor sit amet, consectetur adipiscing elit...	2	Tabular	BigQuery ML	2:53 AM, Jun 27, 2020
▶ model-name-1	✔ Deployed on Vertex AI	Lorem ipsum dolor sit amet, consectetur adipiscing elit...	3	Custom training	Imported	2:53 AM, Jun 27, 2020
▶ model-name-2	✔ Deployed on Vertex AI	Lorem ipsum dolor sit amet, consectetur adipiscing elit...	3	Custom training	Imported	2:53 AM, Jun 27, 2020
▶ model-name-3	✔ Deployed on Vertex AI	Lorem ipsum dolor sit amet, consectetur adipiscing elit...	3	Custom training	Imported	2:53 AM, Jun 27, 2020
▶ model-name-4	✔ Deployed on Vertex AI	Lorem ipsum dolor sit amet, consectetur adipiscing elit...	3	Image classification	AutoML training	2:53 AM, Jun 27, 2020
▶ model-name-5	–	Lorem ipsum dolor sit amet, consectetur adipiscing elit...	3	Image classification	Custom training	2:53 AM, Jun 27, 2020

Rows per page: 10 1-1 of 1

theon-waw project

Search products and resources

SQL QUERY x 1_MODEL_CREATE_LINEAR_4 x

1_model_create_linear QUERY MODEL **DEPLOY TO VERTEX AI** DELETE MODEL EXPORT MODEL

DETAILS TRAINING EVALUATION SCHEMA

Model type Linear regression	Data location EU	Model size 23.2 MB
---------------------------------	---------------------	-----------------------

Use models in **BigQuery** or deploy behind a **REST endpoint**

Batch Predictions inside BigQuery

Model prediction inside of BigQuery queries

Good for data analysis, especially over large batches of data

Example

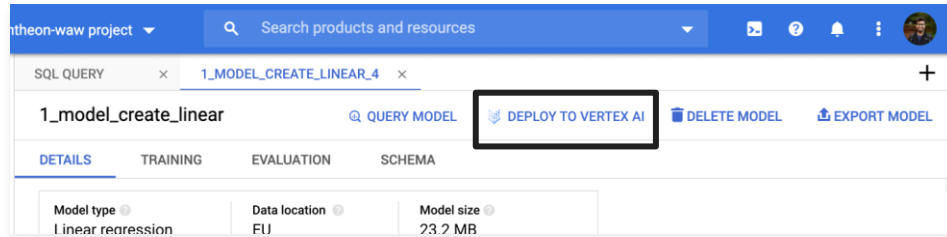
```
#standardSQL
SELECT
  *
FROM
  ML.PREDICT(MODEL `bqml_tutorial.penguins_model`,
  (
    SELECT * FROM
      `bigquery-public-data.ml_datasets.penguins`
    WHERE
      body_mass_g IS NOT NULL
      AND island = "Biscoe"))
```

Vertex Online prediction

Get model predictions for single examples over a REST API

Good for integration with applications, especially requiring low-latency

Leverage Vertex Model Monitoring



Orchestrate data prep, training, and eval with **Vertex AI Pipelines**



Easy to use Python SDKs

Build pipelines using Data Scientist friendly SDKs like TensorFlow Extended and Kubeflow Pipelines.



Automated, Scalable and Cost Effective

Leverage GCPs managed services to build scalable pipelines. Pay for only the resource you use.

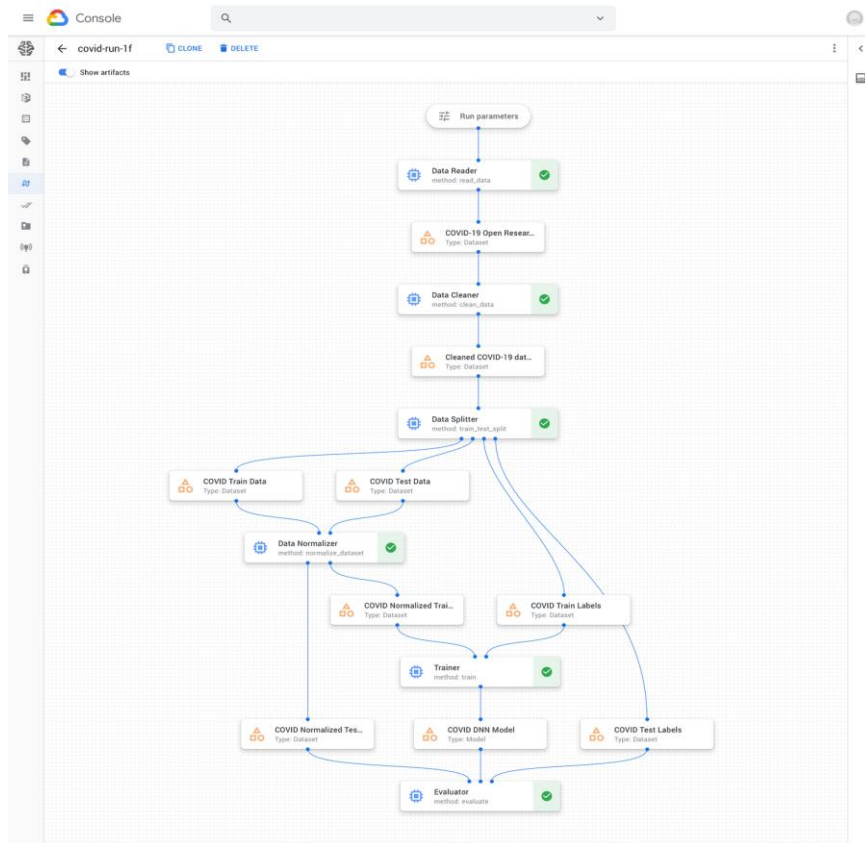


Streamlined MLOps

Automatic metadata tracking

BigQuery Model operators

- BigqueryQueryJobOp
- BigqueryCreateModelJobOp
- BigqueryExportModelJobOp
- BigqueryPredictModelJobOp
- BigqueryEvaluateModelJobOp



Limitless Users

By delivering data experiences for all data people

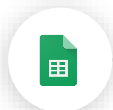
Deliver the right tools to the right people.

Google's varied data experiences make it easy for any team to drive analysis from its leading DBMS platform.

Self-Service BI
Data Studio



EDW-Powered Spreadsheets
Google Sheets



Governed BI with
Semantic Model
Looker



Natural Language
Data QnA



Partner Ecosystem



Introducing Connected Sheets

Combining the best of Big Query and the familiarity of Sheets to empower workforces and assist with:

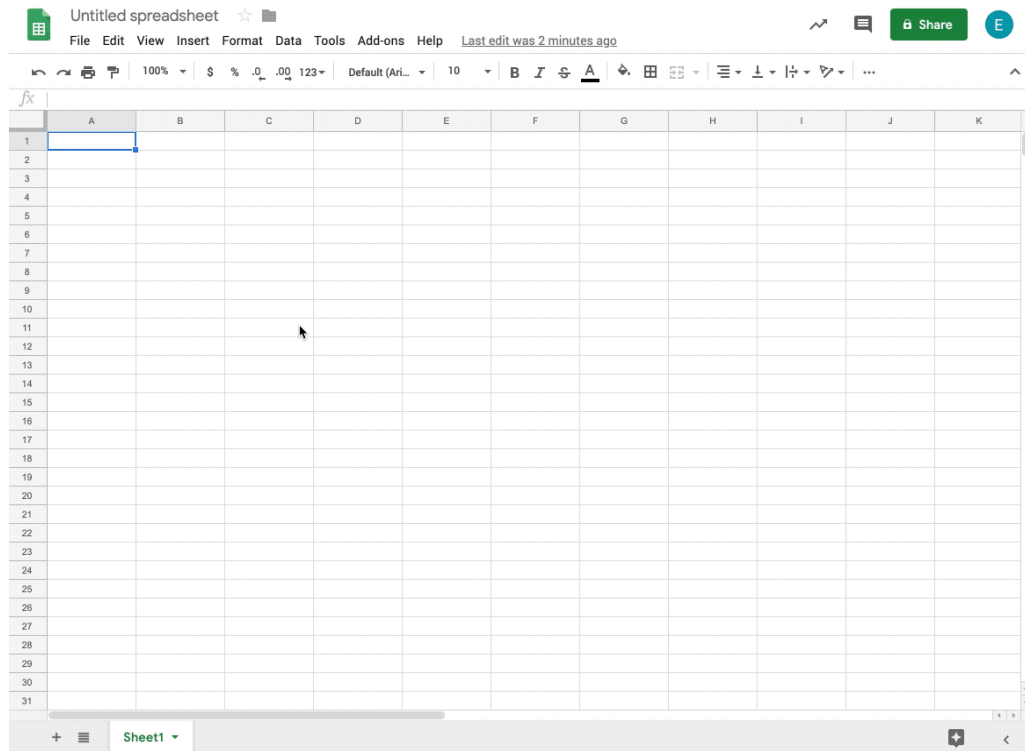
- Unlocking big data insights
- Accelerating data workflows
- Improving cost-efficiency
- Strengthening data security



1

Single source of truth

Access billion rows of BigQuery data directly in Sheets without compromising security and performance.



2

Do more with familiar tools

Unlock insights from your data using features you're already familiar with in Sheets, such as pivot tables, charts, and formulas.

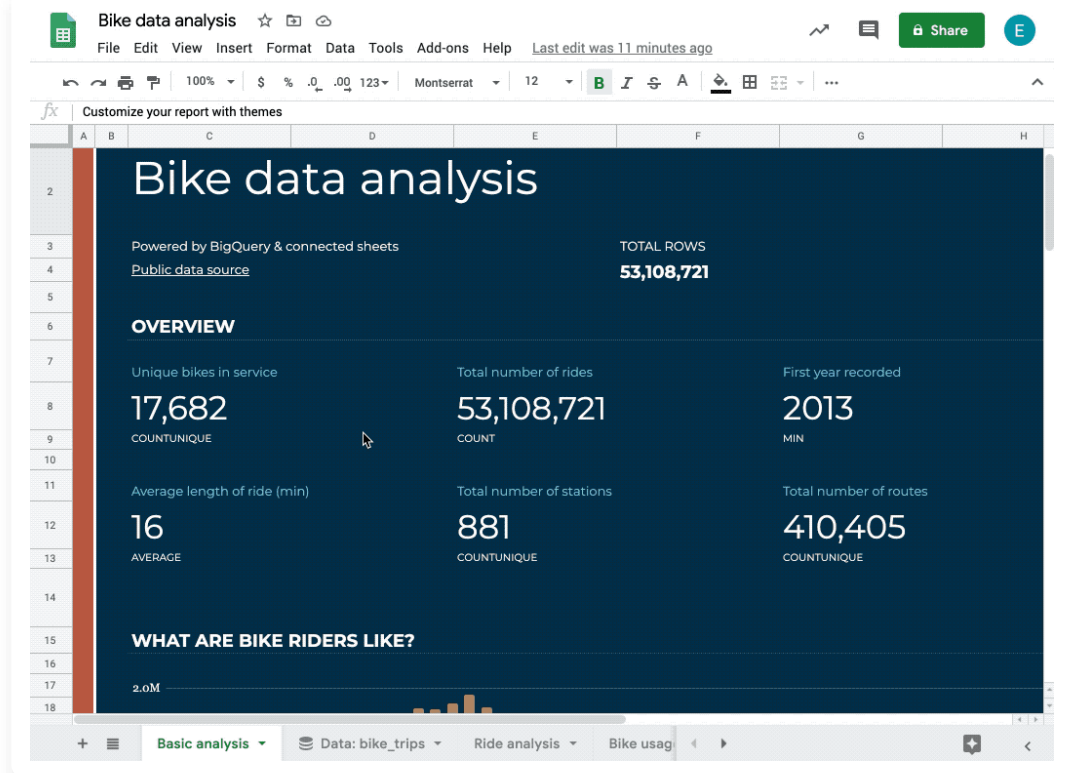
The screenshot displays the Google Sheets interface for a spreadsheet titled "Bike data analysis". The menu bar includes File, Edit, View, Insert, Format, Data, Tools, Add-ons, and Help. The spreadsheet shows a table with columns for trip duration, start and stop times, and start station details. A toolbar above the table offers options for Chart, Pivot table, Function, Extract, and Calculated column. A preview of the full data set is visible at the bottom of the table.

tripduration	starttime	stoptime	start_station_id	start_station_name	start_station_latitude	start_station_longitude
1582	5/30/2018 16:41:12	5/30/2018 17:07:35	3536	W 116 St & Broadway	40.8082	-73
500	3/26/2018 17:04:10	3/26/2018 17:12:31	3398	Smith St & 9 St	40.6746957	-73
85	5/23/2018 15:59:58	5/23/2018 16:01:24	3674	Jay St & York St	40.70140317	-73
802	5/11/2018 16:15:25	5/11/2018 16:28:47	3305	E 91 St & 2 Ave	40.7811223	-73
240	5/24/2018 18:55:30	5/24/2018 18:59:31	3674	Jay St & York St	40.70140317	-73
964	2/24/2018 12:39:21	2/24/2018 12:55:26	3644	Van Dyke St & Van Brunt St	40.67581623	-74
240	5/3/2018 14:59:13	5/3/2018 15:03:14	150	E 2 St & Avenue C	40.7208736	-73
284	5/26/2018 19:34:26	5/26/2018 19:39:10	2000	Front St & Washington St	40.70255088	-73
306	5/14/2018 15:54:34	5/14/2018 15:59:40	433	E 13 St & Avenue A	40.72955361	-73
426	4/9/2018 16:18:19	4/9/2018 16:25:25	3539	W 116 St & Amsterdam Ave	40.8067581	-73
339	5/3/2018 11:29:30	5/3/2018 11:35:09	3552	W 113 St & Broadway	40.805973	-73
713	2/22/2018 18:40:15	2/22/2018 18:52:09	433	E 13 St & Avenue A	40.72955361	-73
202	5/16/2018 9:52:41	5/16/2018 9:56:03	3542	Amsterdam Ave & W 119 St	40.8086249	-73
196	5/21/2018 8:12:38	5/21/2018 8:15:55	3632	E 12 St & Avenue B	40.72804857	-73
816	4/7/2018 11:32:44	4/7/2018 11:46:21	3301	Columbus Ave & W 95 St	40.7919557	-73
252	5/14/2018 8:10:08	5/14/2018 8:14:21	3632	E 12 St & Avenue B	40.72804857	-73
			3599	Franklin Ave & Empire Blvd	40.66314	-73
			3552	W 113 St & Broadway	40.805973	-73

3

Fresh insights

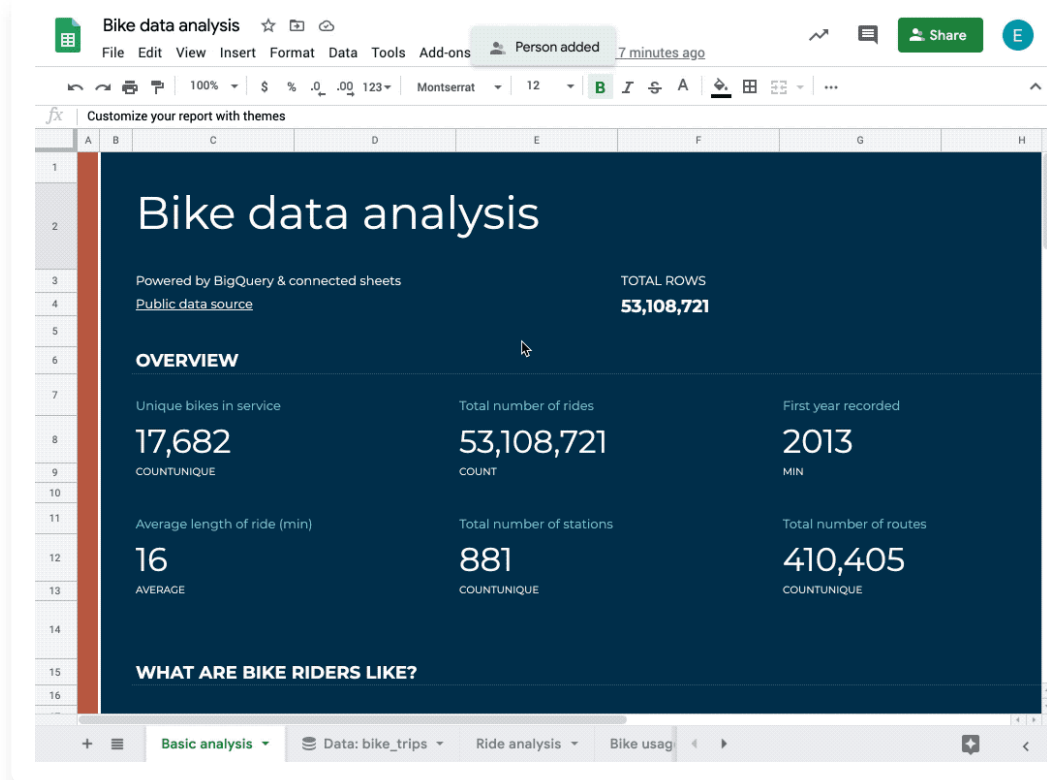
Help ensure insights are based on the most up-to-date BigQuery data by setting up automatic data refreshes in Sheets.



4

Help secure collaboration

Share insights and collaborate securely with colleagues, analysts, or other stakeholders in a familiar spreadsheet interface.



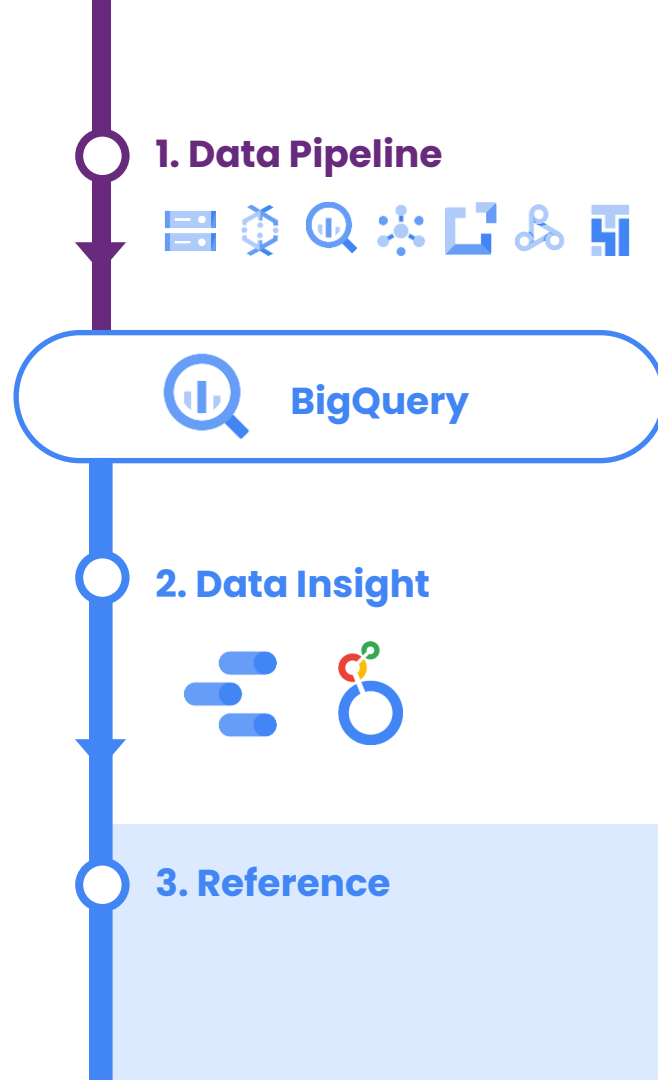
01 Intro



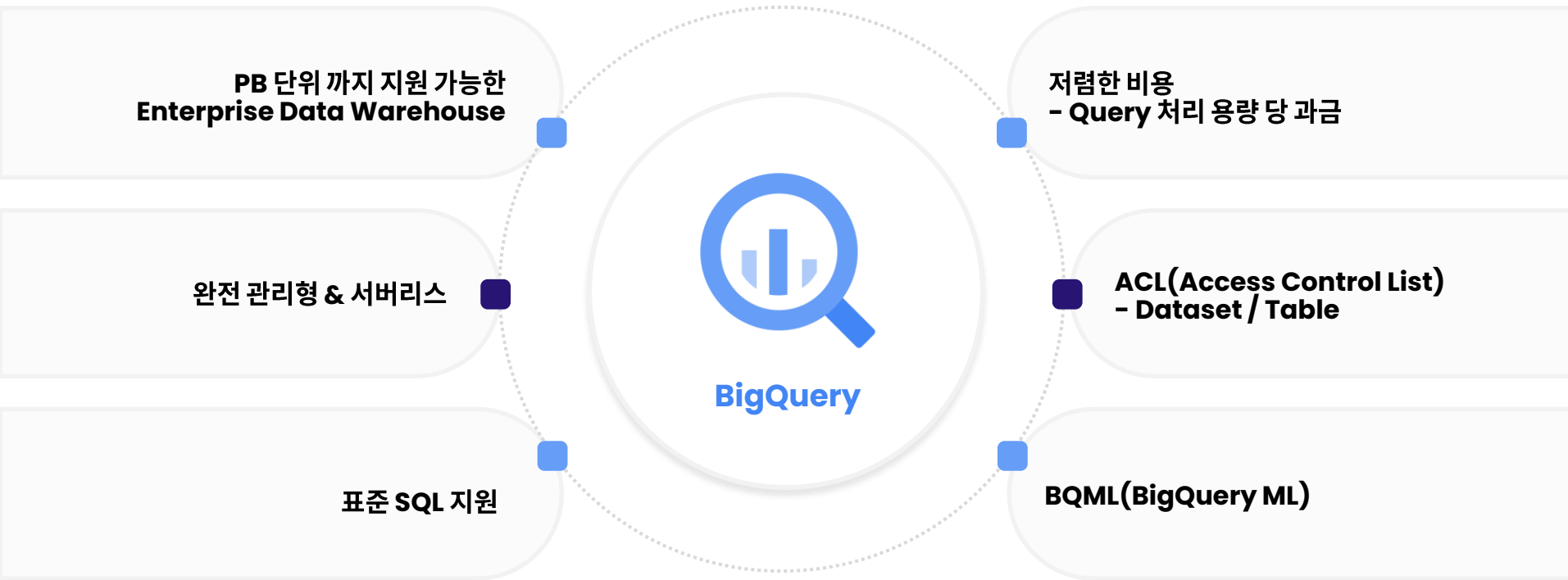
02 Session 1

Common Data Analytics Pipeline on GCP

- 1. Data Pipeline on GCP
- 2. Data Insight on GCP
- 3. Reference



Enterprise Data Warehouse인 BigQuery의 장점을 소개합니다.



2-1 Data Pipeline on GCP

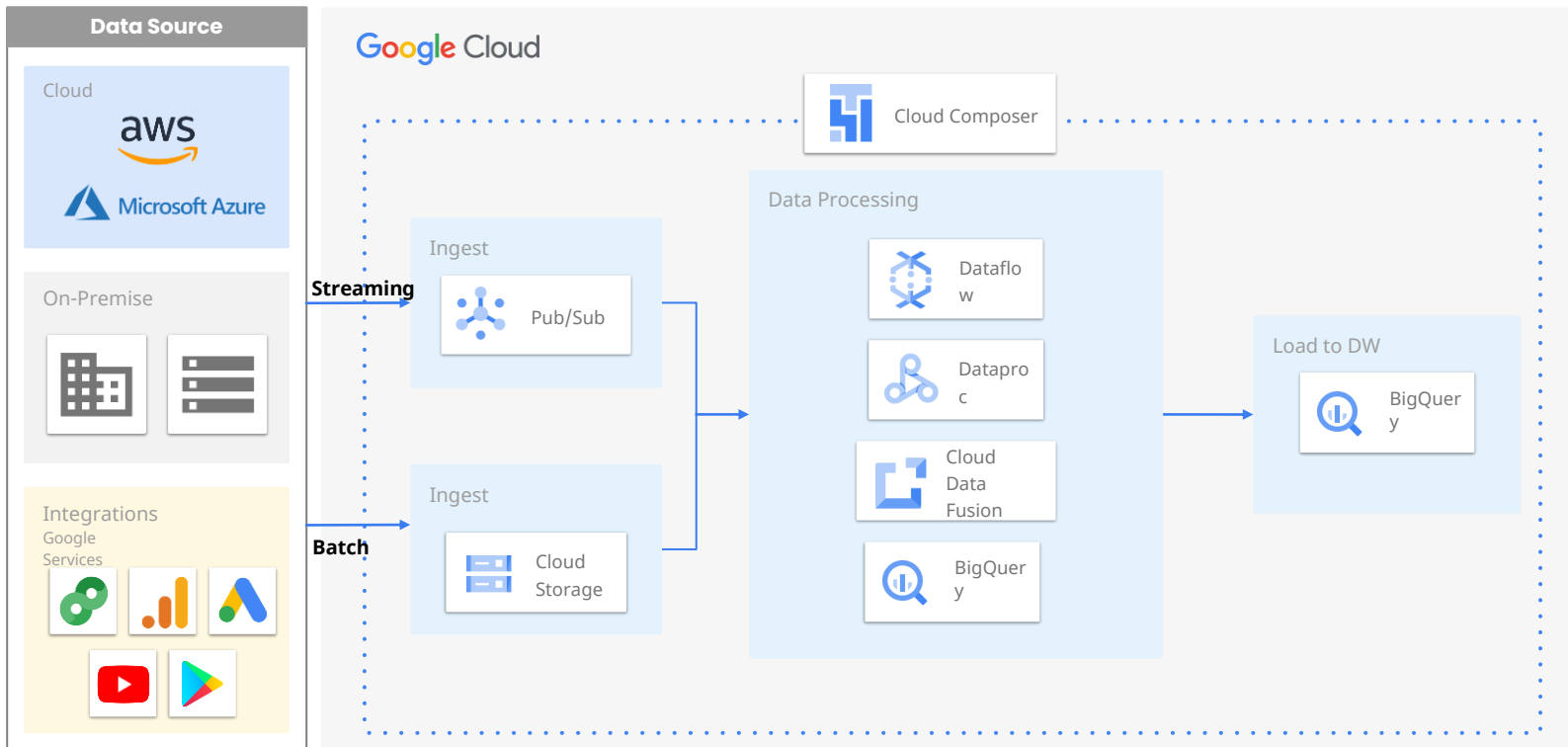


BigQuery

2-2 Data Insight on GCP

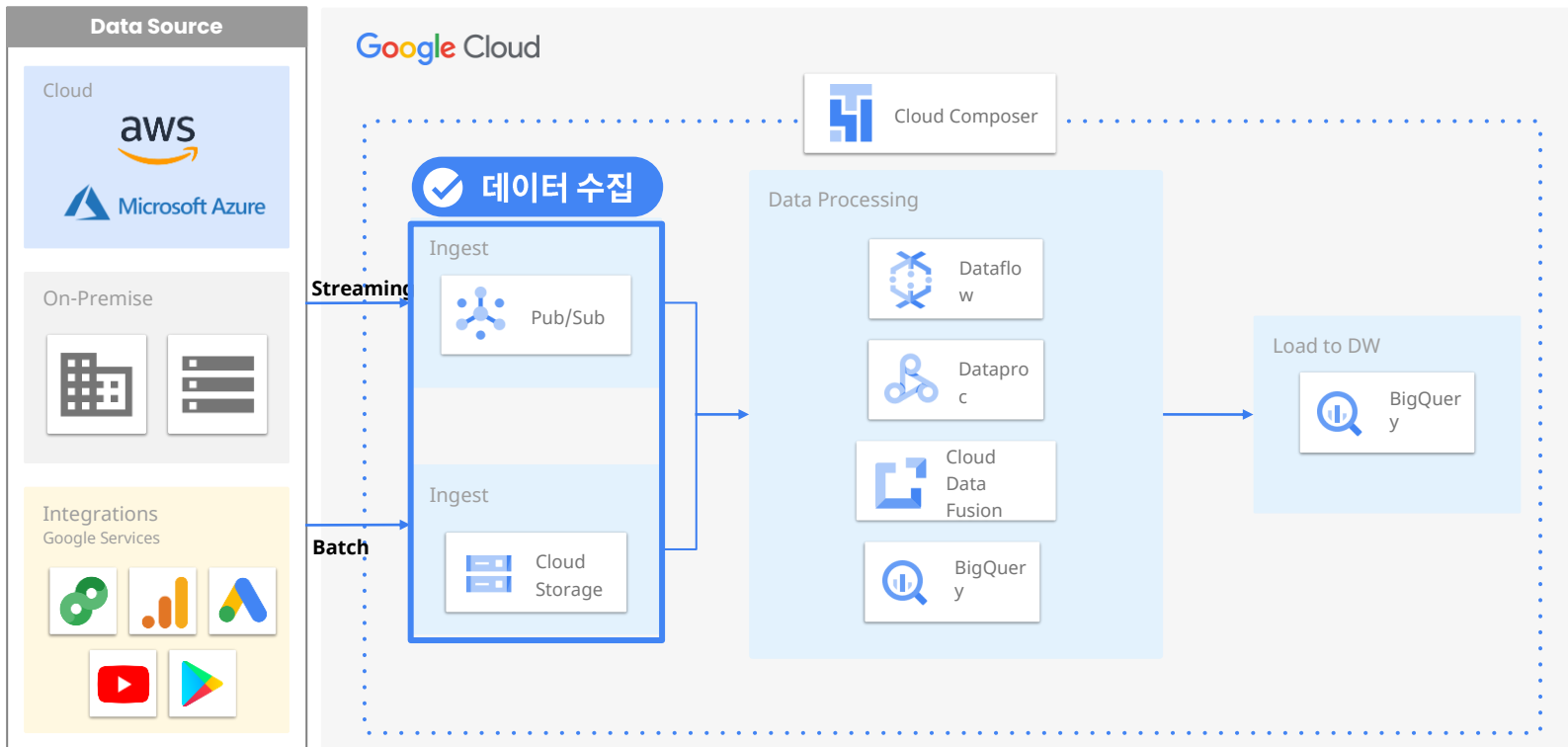
Google Cloud 기반의 일반적인 Data Pipeline 예시입니다.

Architecture: Streaming & Batch Data Analytics Pipeline



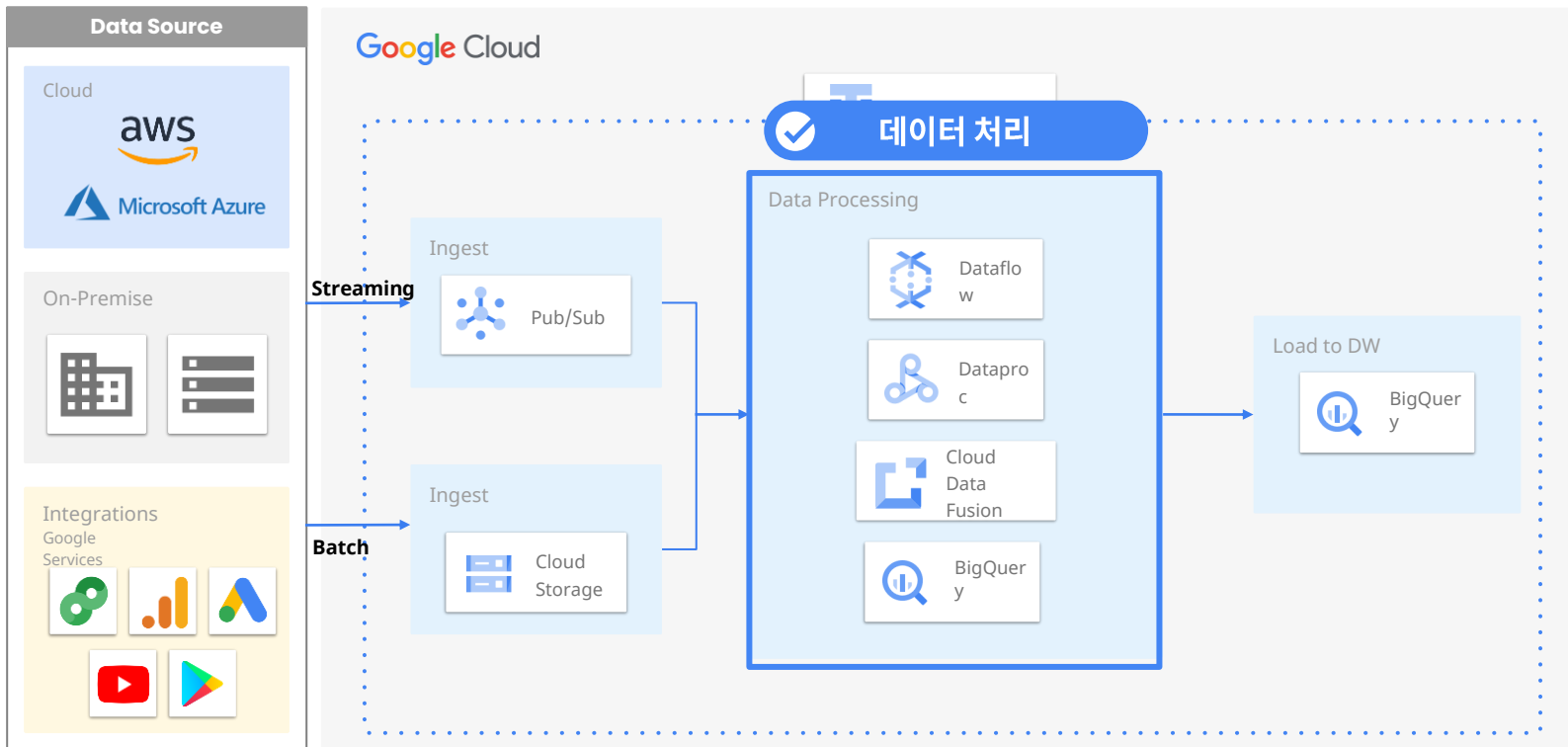
Google Cloud 기반의 일반적인 Data Pipeline 예시입니다.

Architecture: Streaming & Batch Data Analytics Pipeline

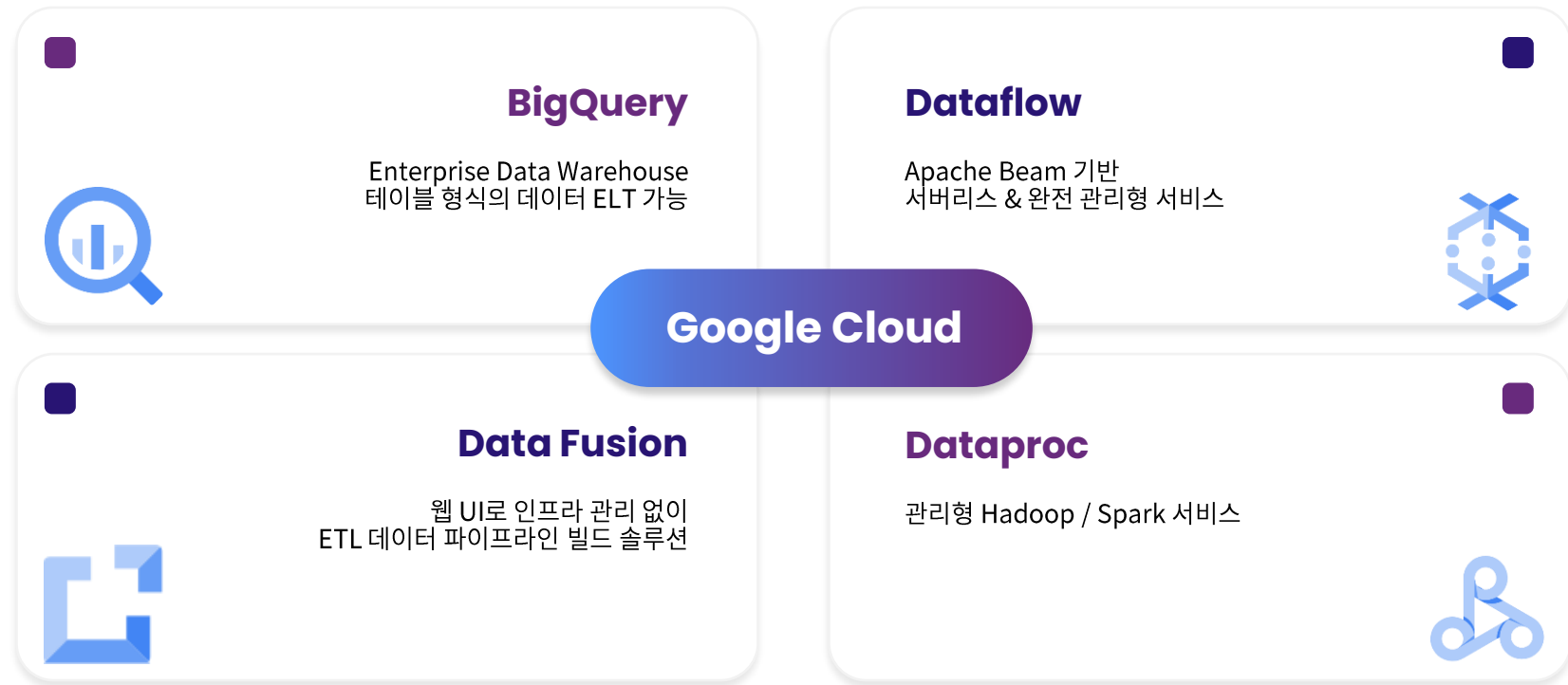


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Architecture: Streaming & Batch Data Analytics Pipeline

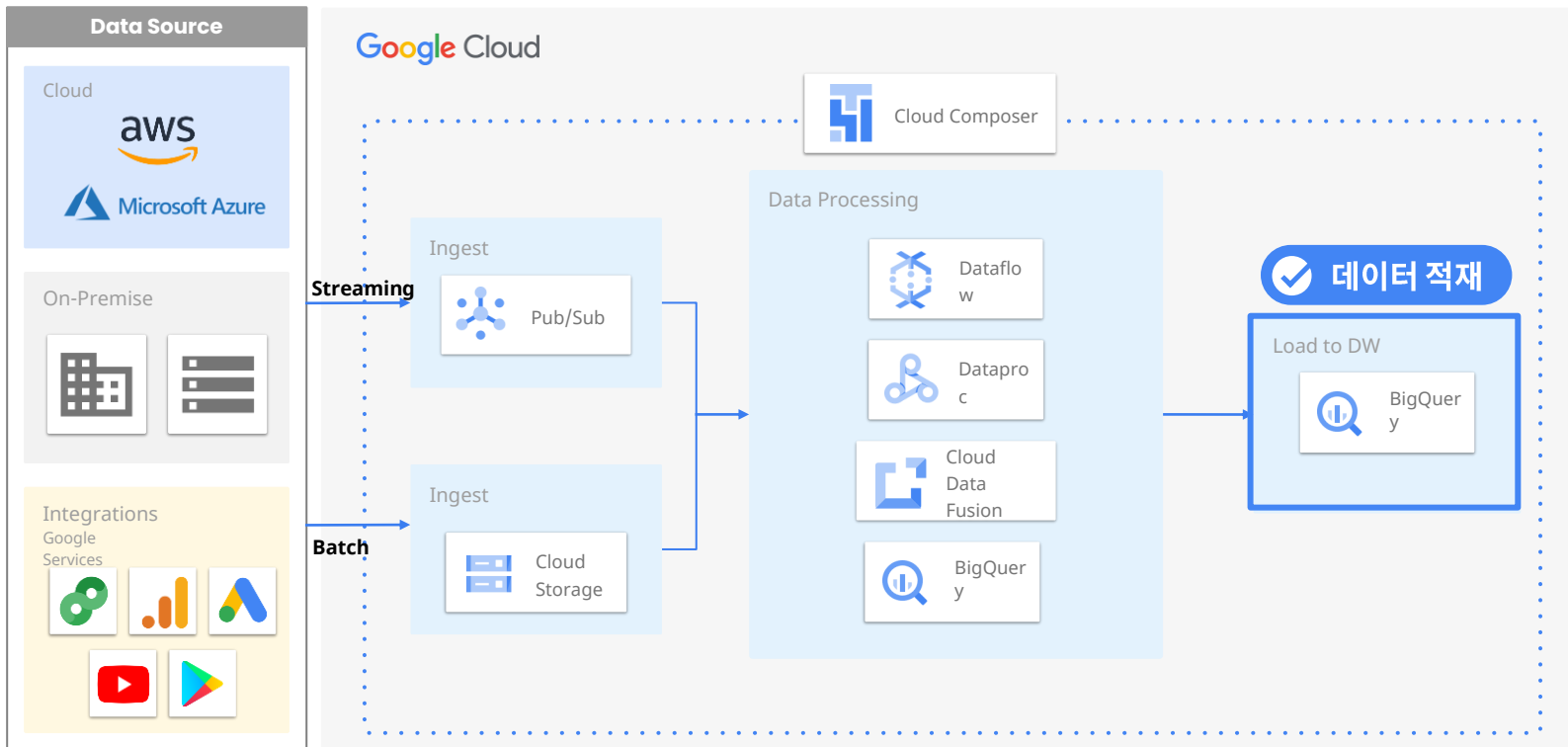


Google Cloud에서 데이터를 처리할 때 사용하는 서비스를 소개합니다.



Google Cloud 기반의 일반적인 Data Pipeline 예시입니다.

Architecture: Streaming & Batch Data Analytics Pipeline





BigQuery



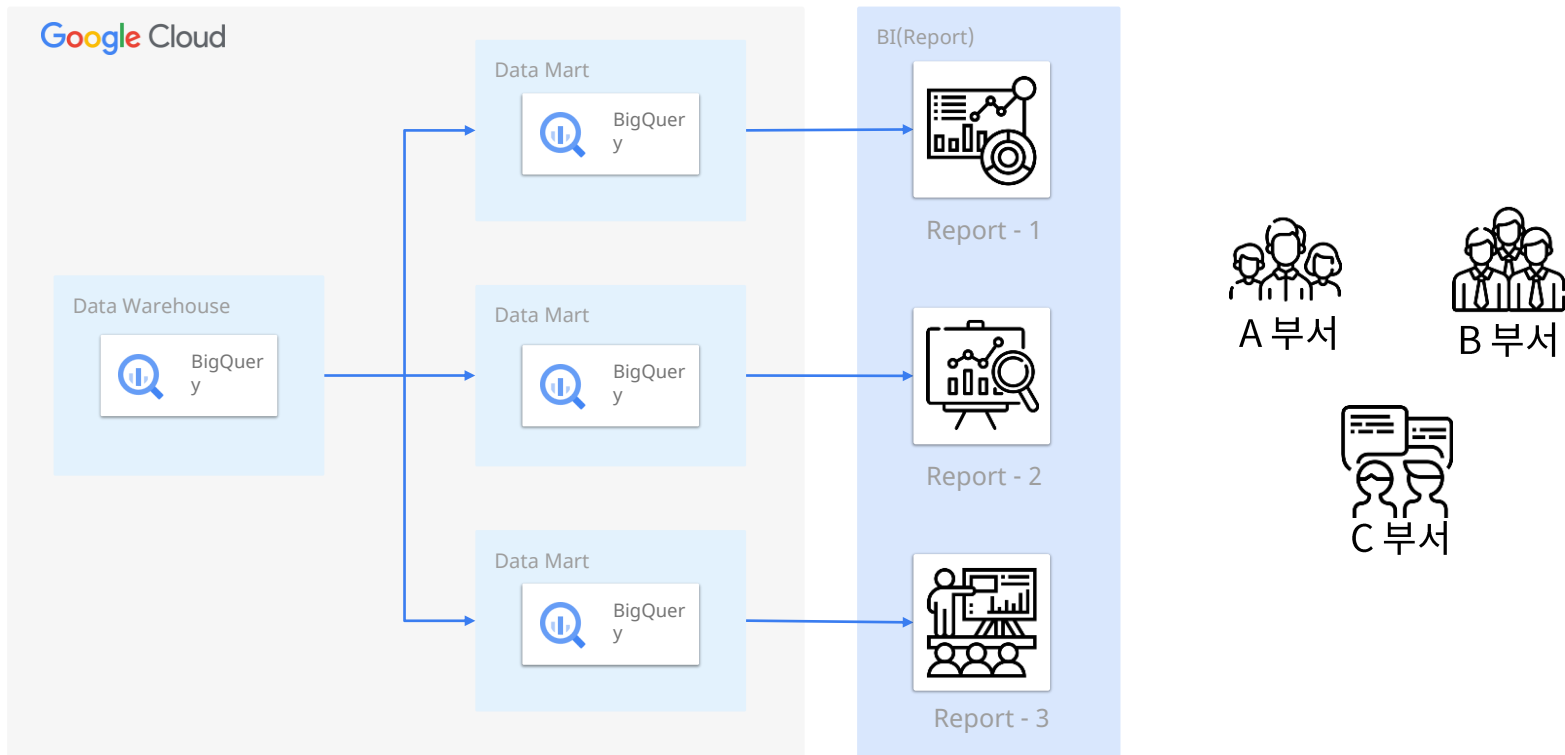
2-2 Data Insight with GCP



2-3 Reference

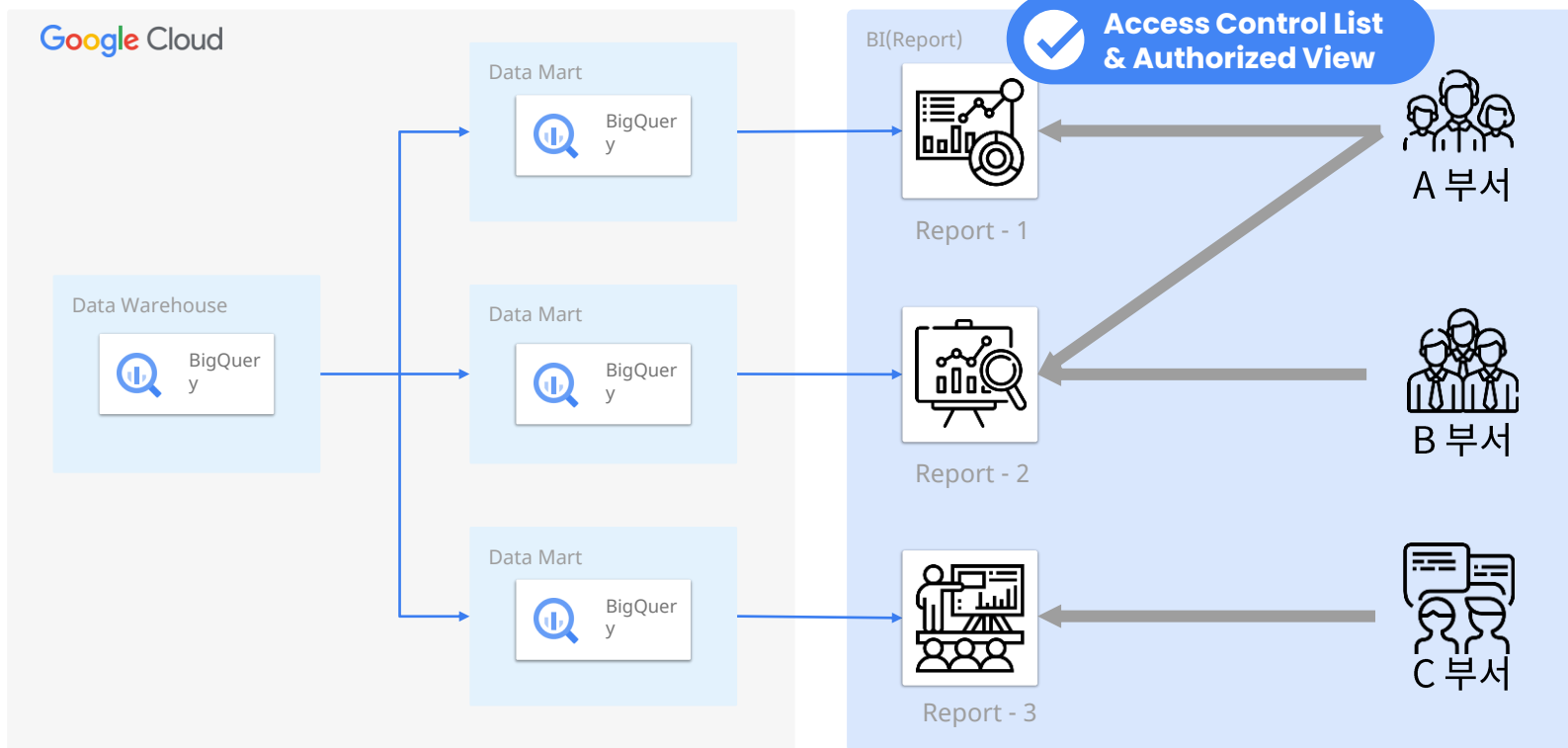
BigQuery ACL(Access Control List)과 View 기능을 활용해 Data Mart로서 BigQuery를 활용합니다.

Using BigQuery as a Data Mart

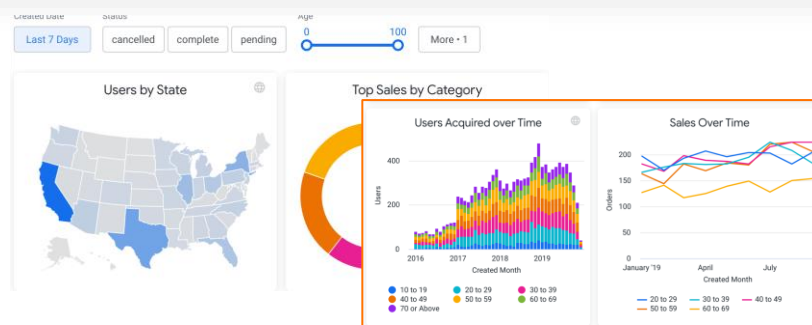
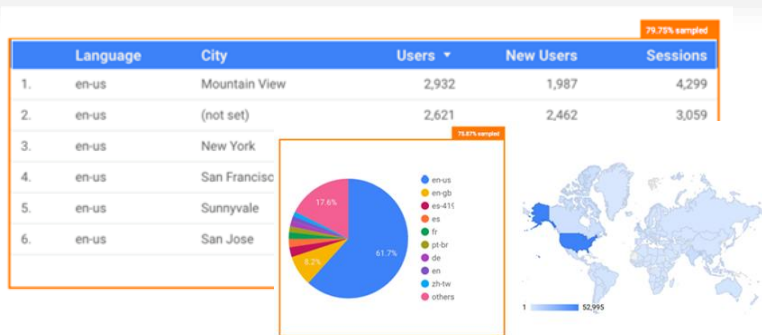


BigQuery ACL(Access Control List)과 View 기능을 활용해 Data Mart로서 BigQuery를 활용합니다.

Using BigQuery as a Data Mart



Google Cloud에서 데이터를 시각화할 때 주로 사용되는 두 가지 서비스를 소개합니다.



Data Studio



- 무료
- 다양한 데이터소스 지원
- Code/Query 없이 대시보드 생성

Looker



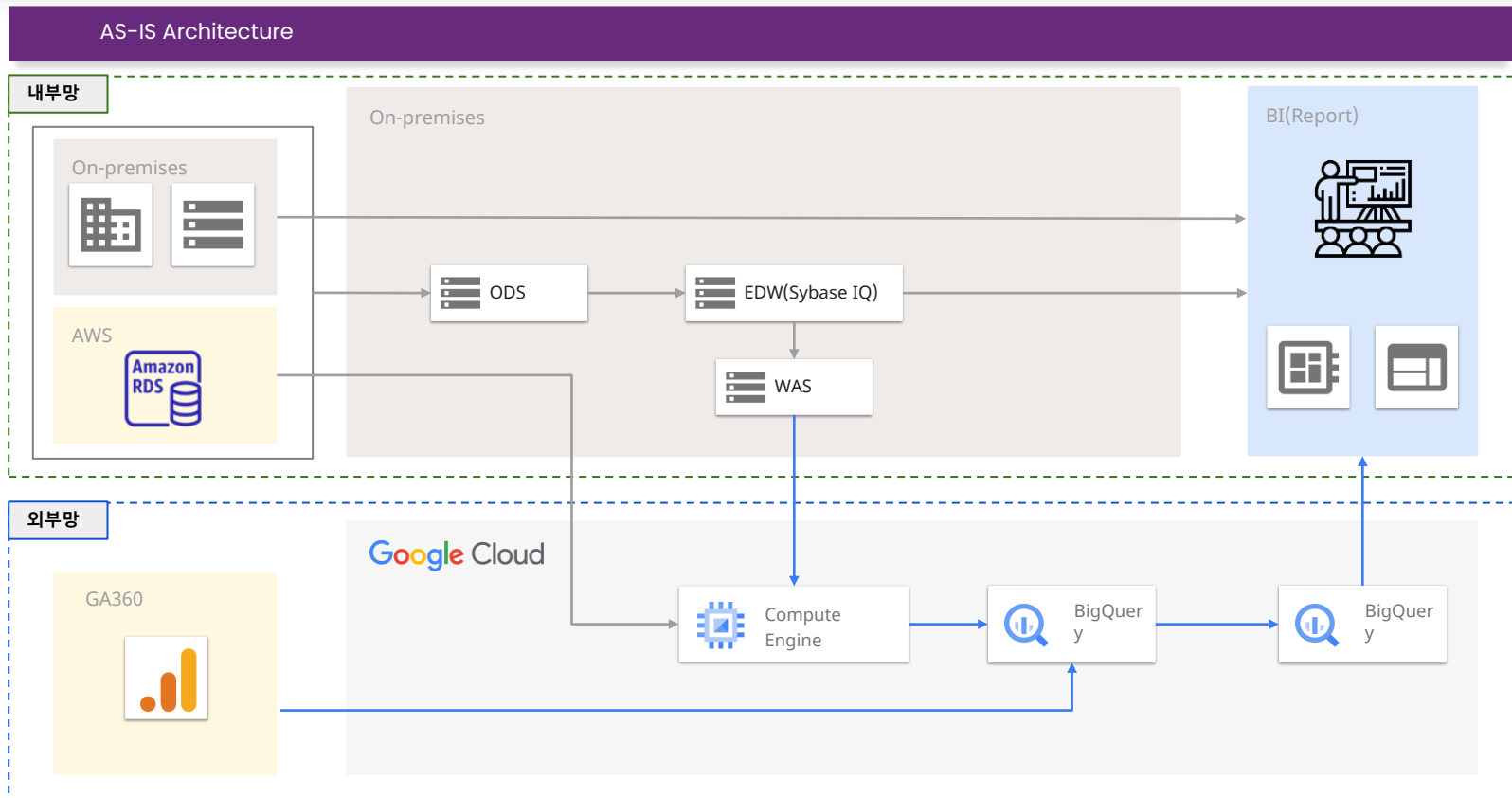
- 유료
- 접근제어-IPAllowList&ACL
- LookML로 Custom Logic 설정
- 개발/배포 환경 분리
- 연결한 DataBase에 직접 Query



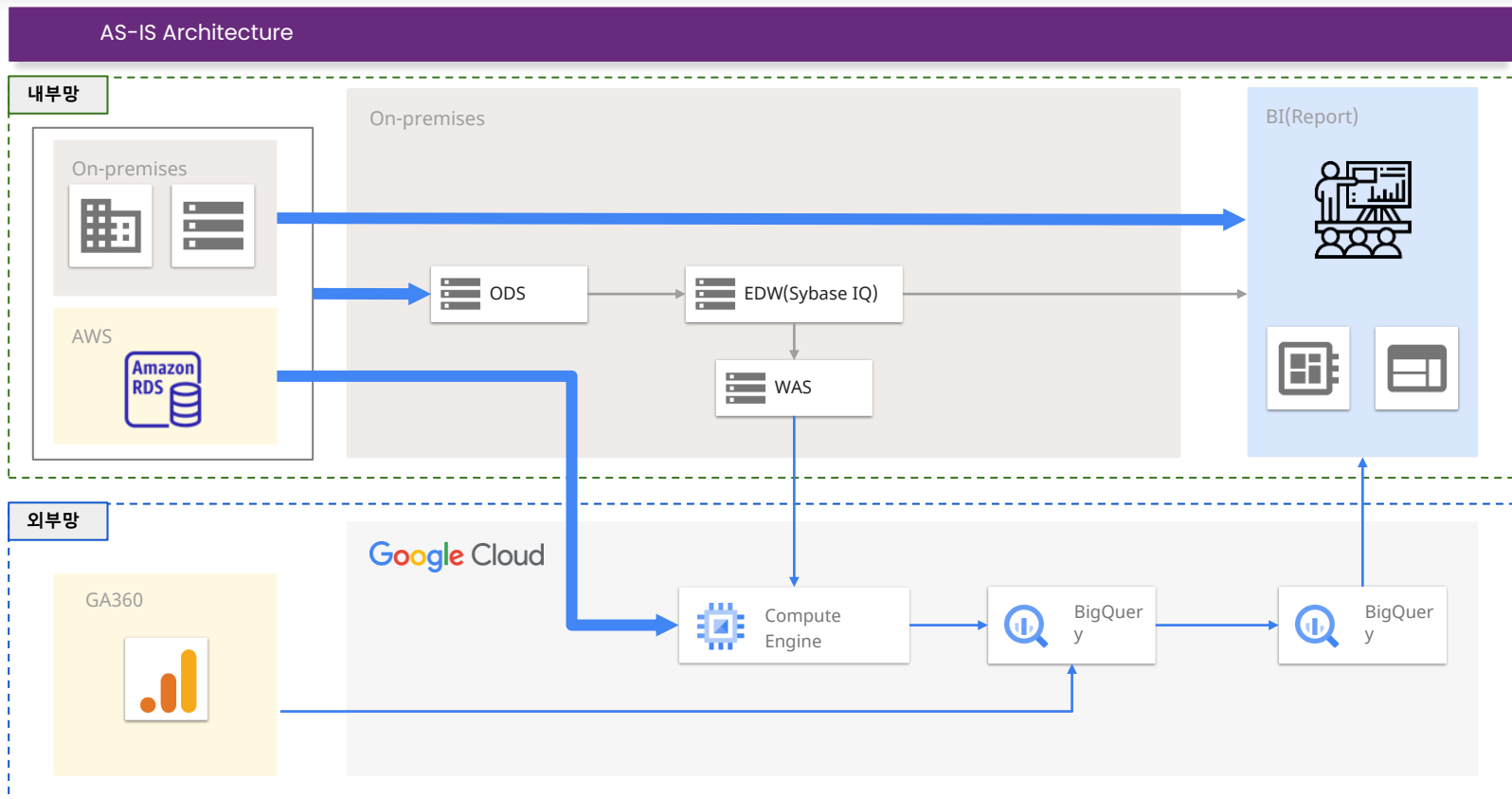
2-3 Reference

유명 패션업계 제조업체 K사

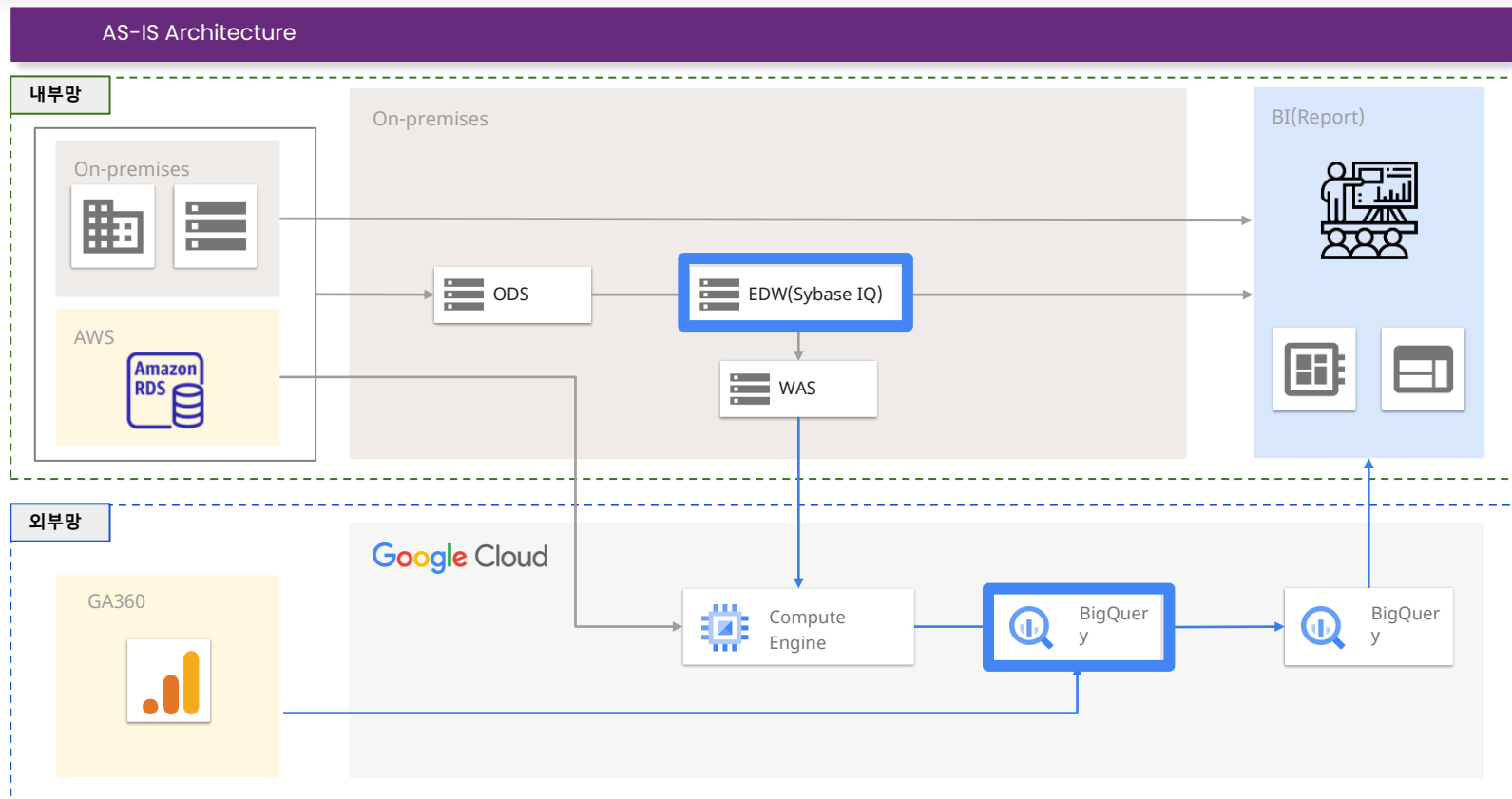
본 사례의 기존 분석계 시스템 아키텍처입니다.



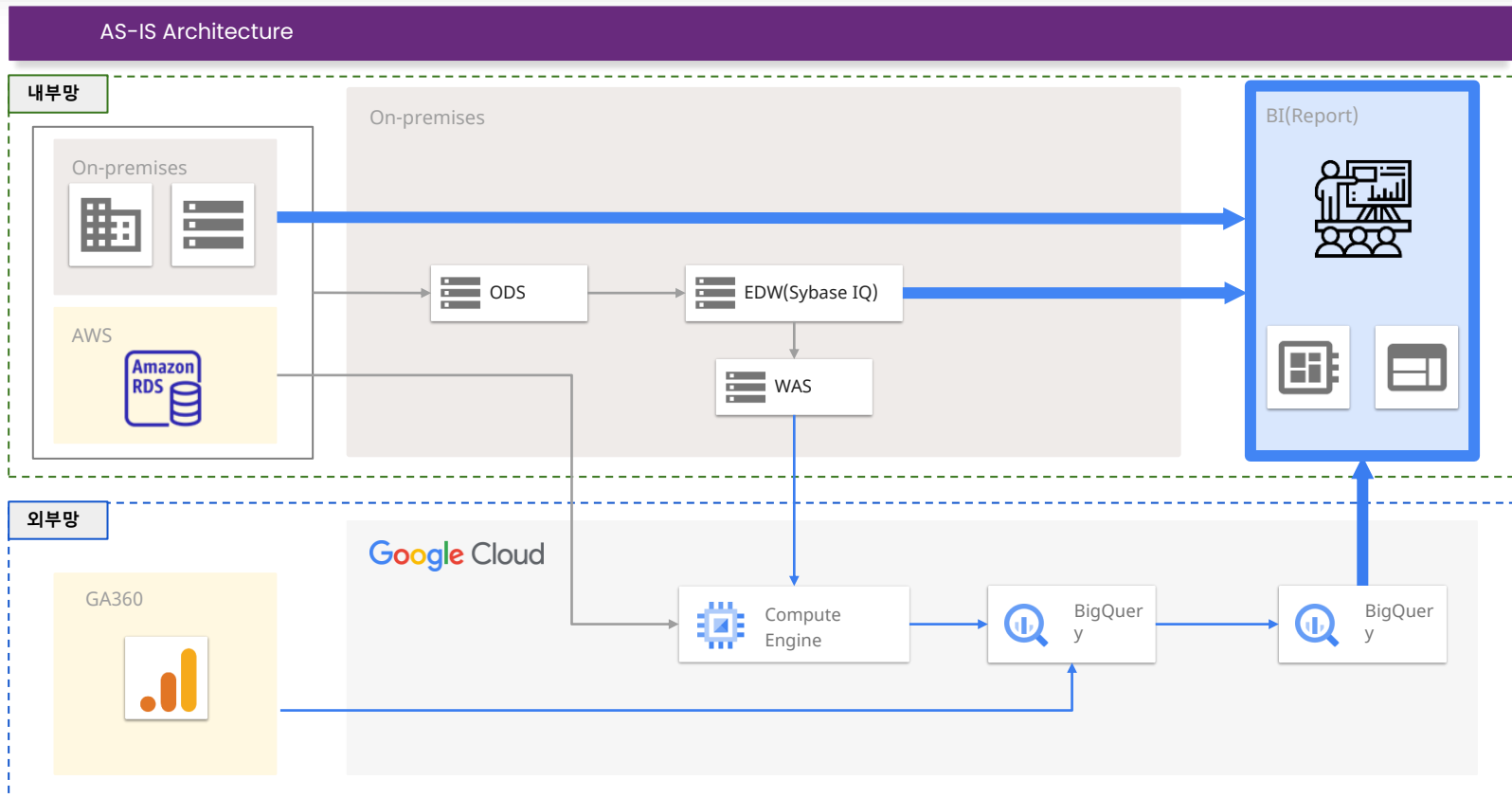
본 사례의 기존 분석계 시스템 아키텍처 입니다.



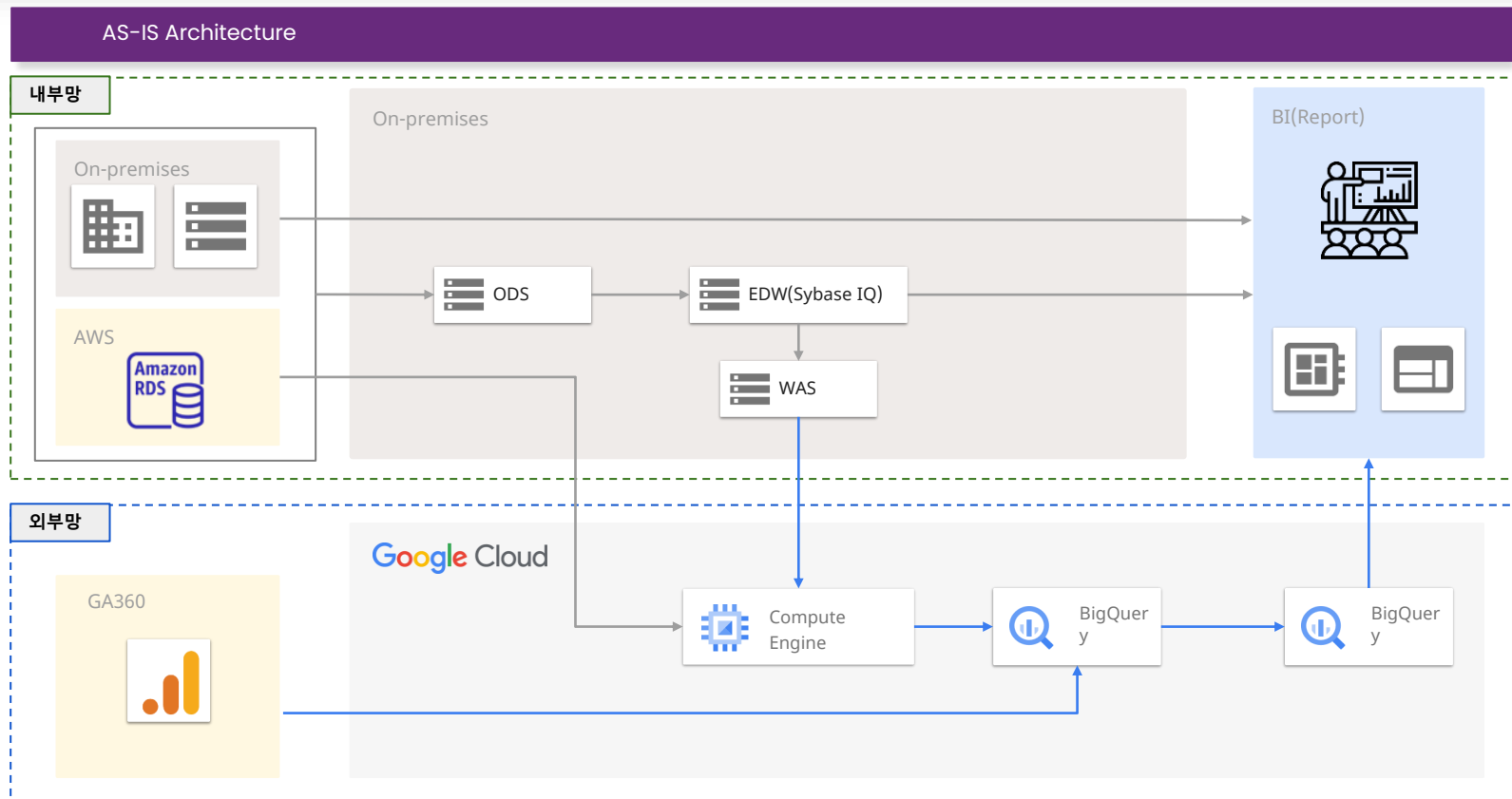
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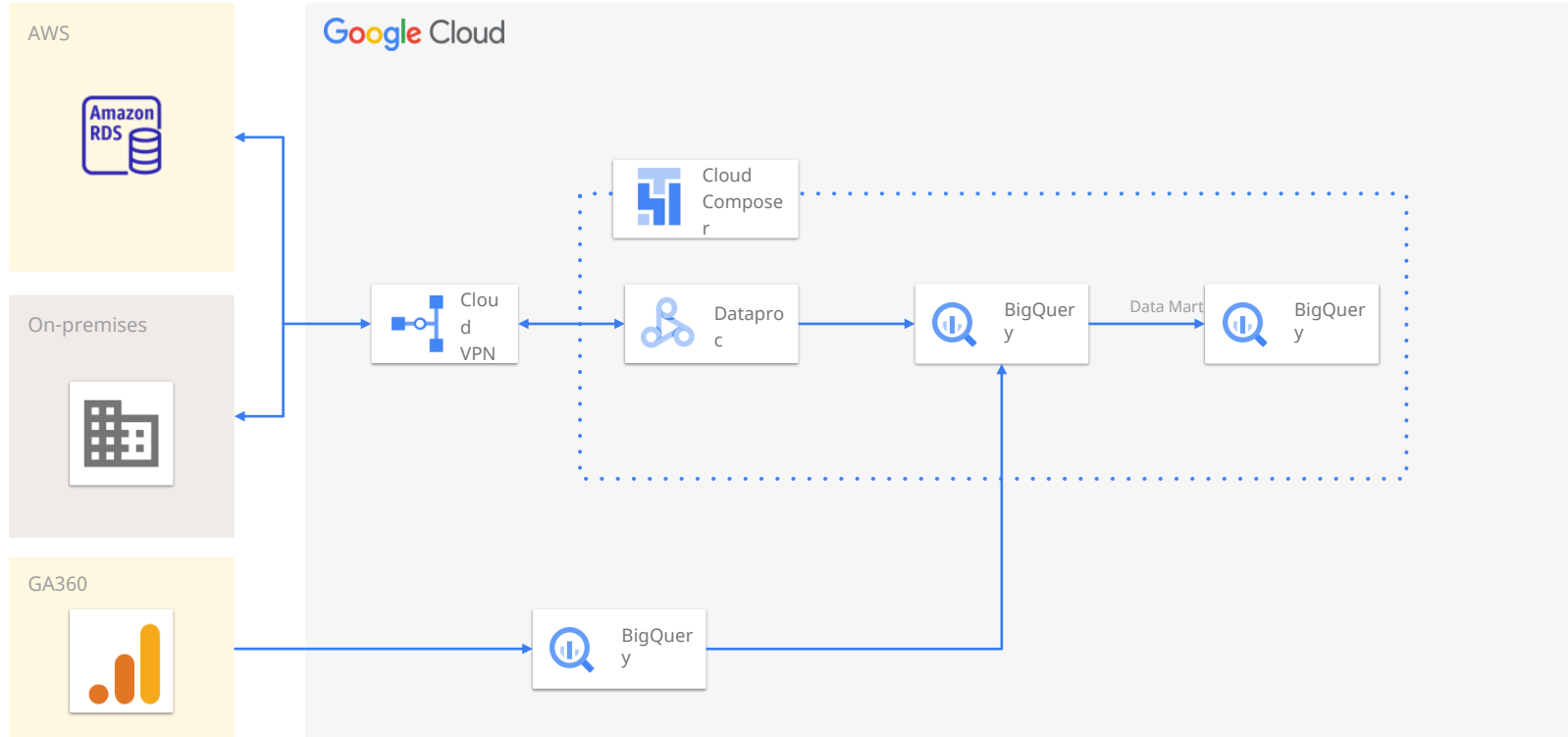


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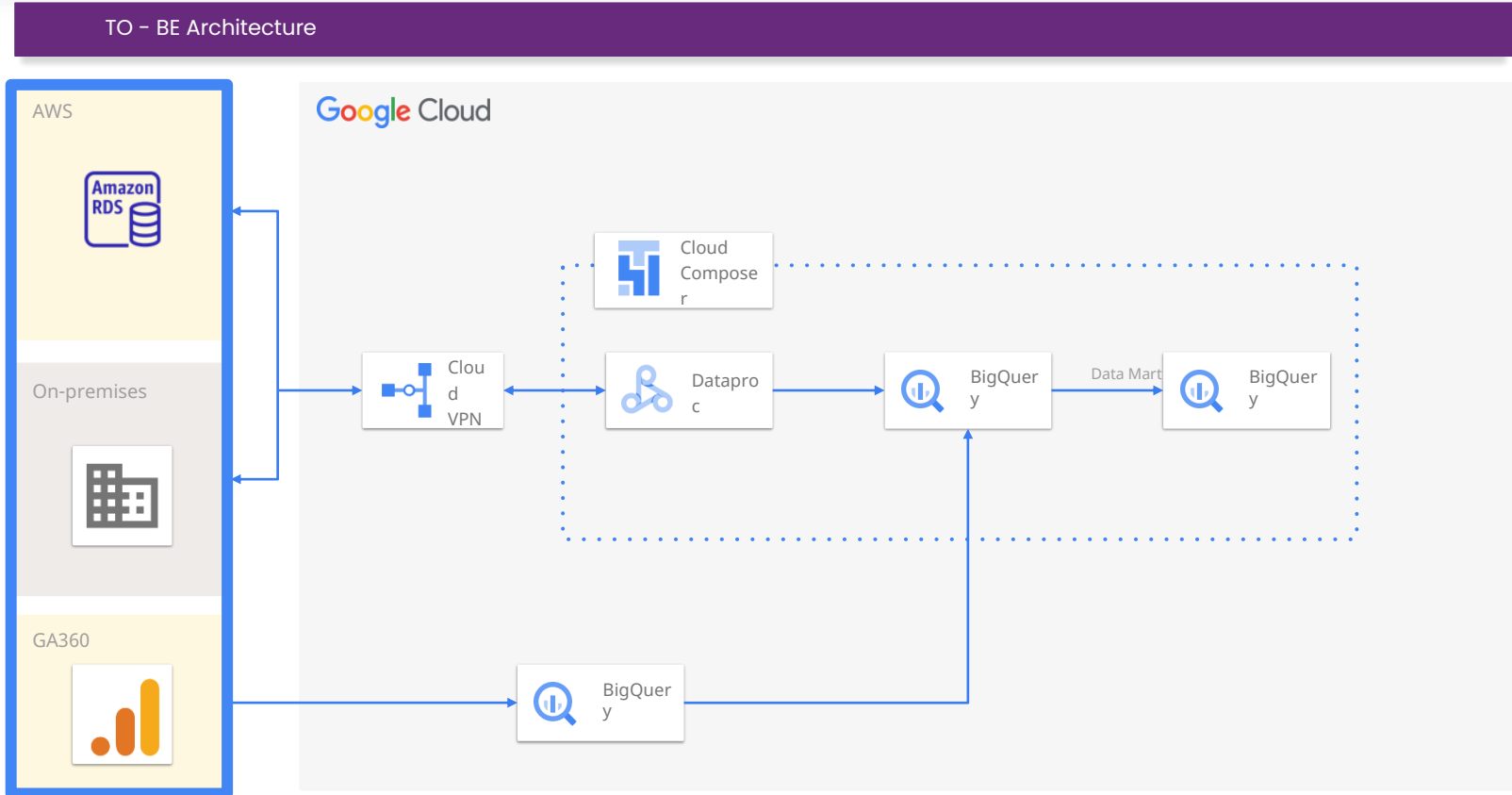


Google Cloud를 기반으로 통합된 분석계 환경을 구축한 사례입니다.

TO - BE Architecture

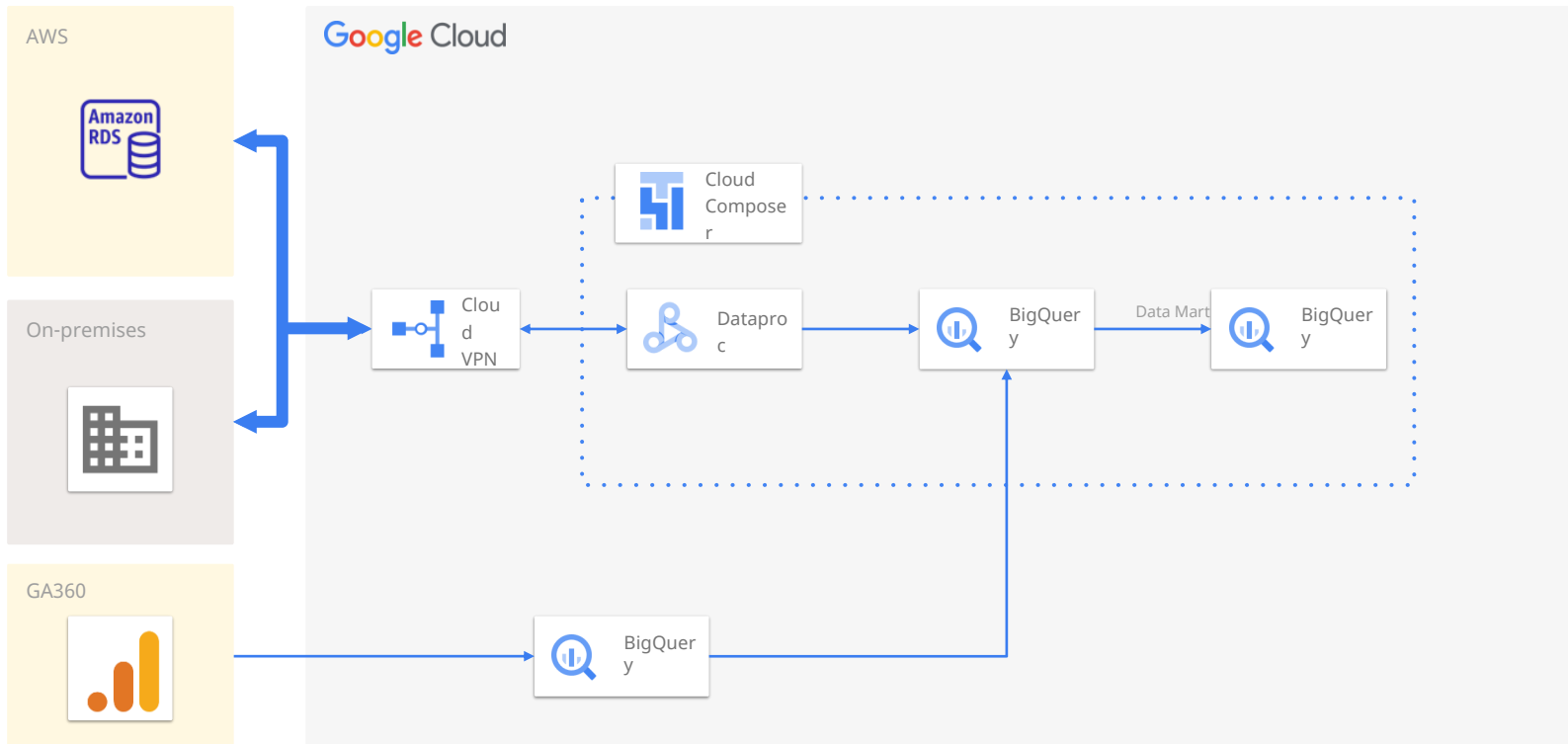


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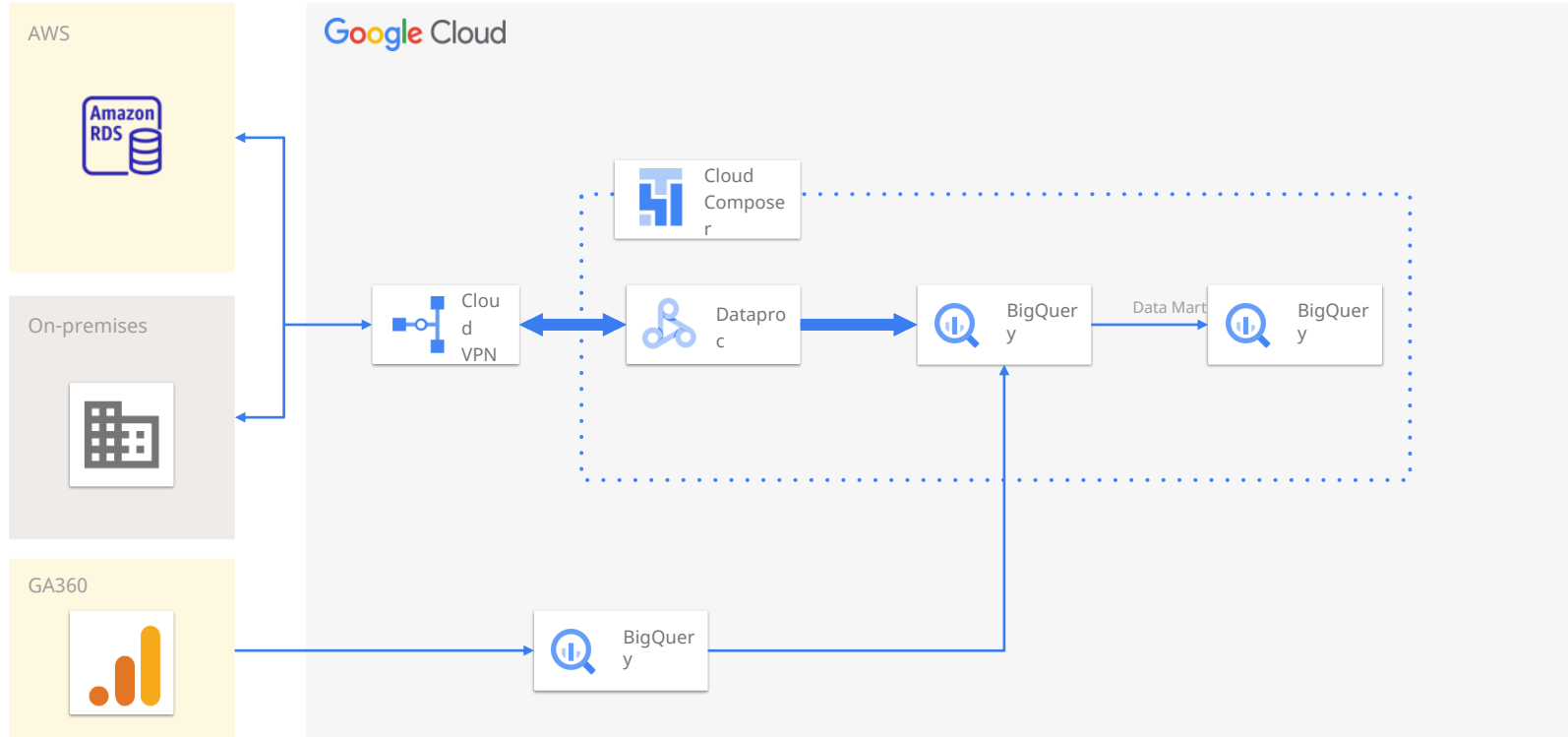
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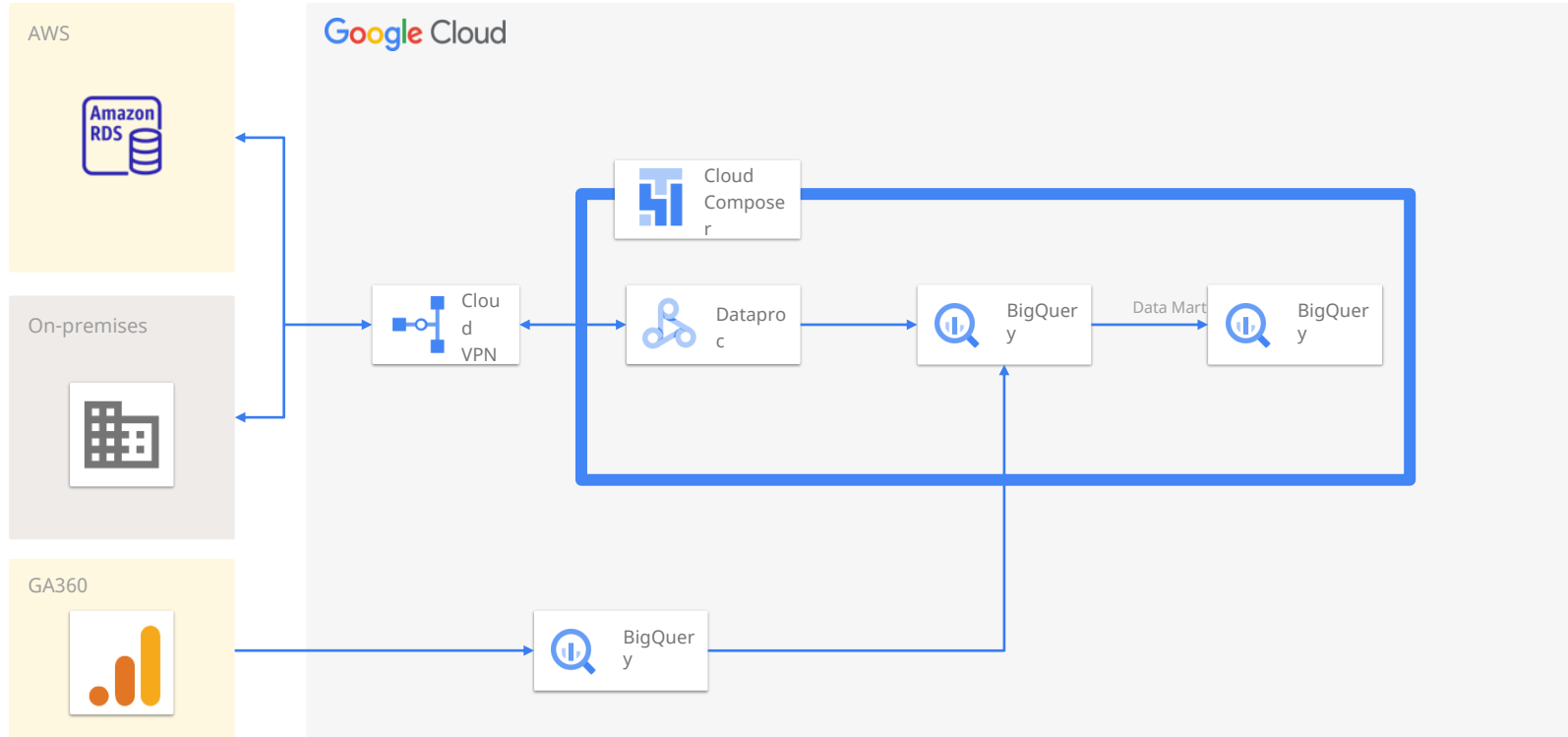
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TO - BE Architecture



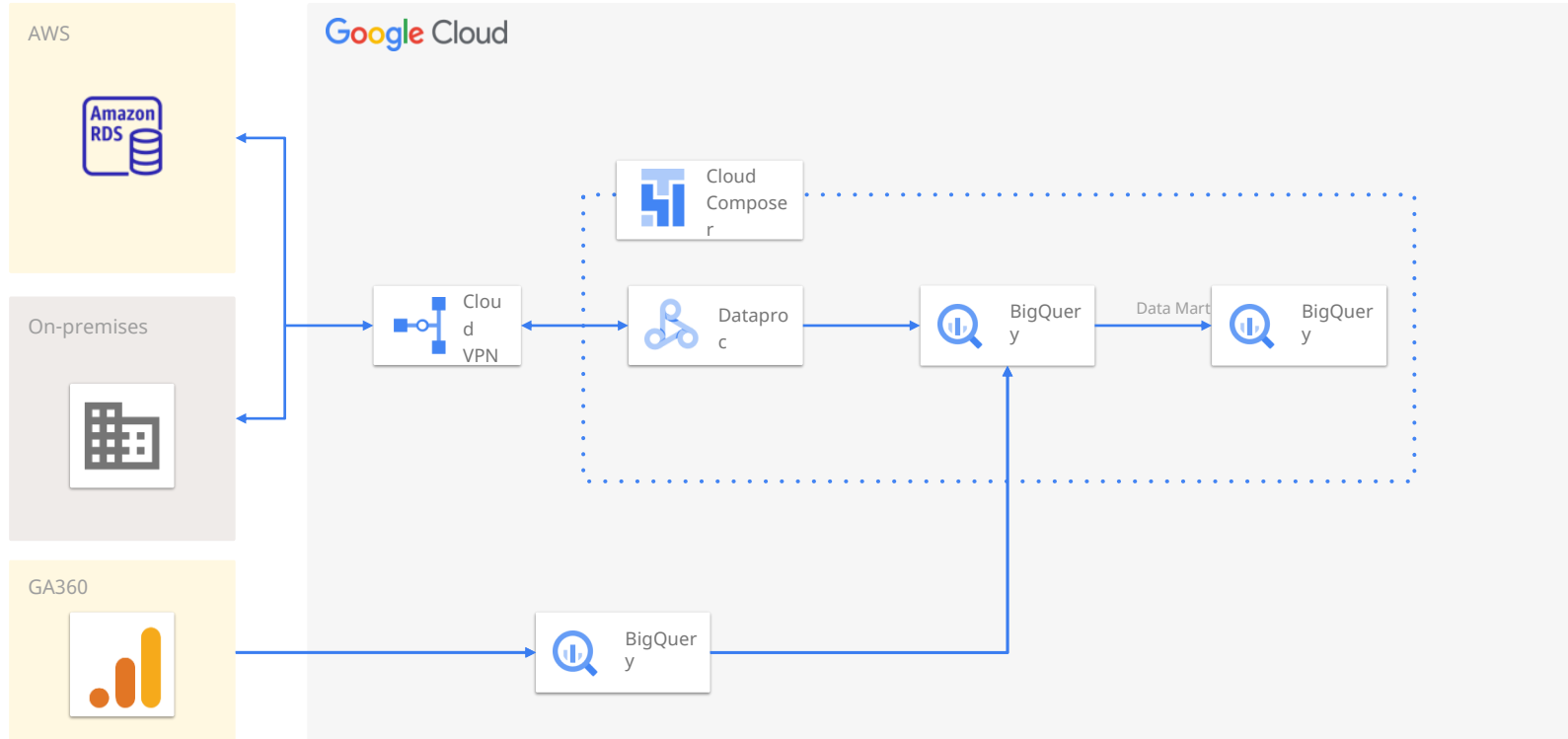
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TO - BE Architecture



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TO - BE Architecture



본 사례의 도입 효과입니다.



본 사례의 도입 효과입니다.

GCP 기반 통합 분석 환경
(BigQuery)



운영 및 분석
비용 절감

도입효과

Private 환경 구축

파이프라인
간소화

본 사례의 도입 효과입니다.



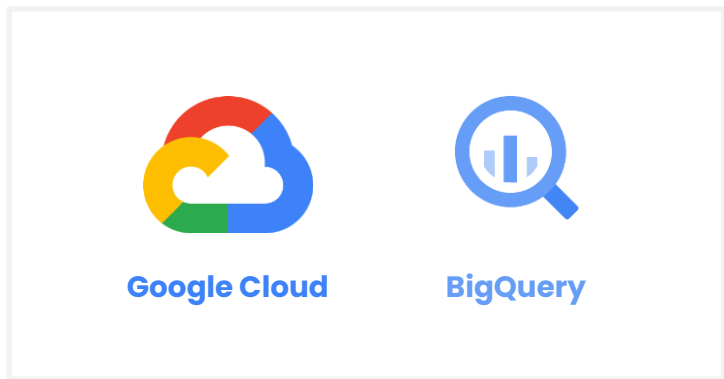
본 사례의 도입 효과입니다.



본 사례의 도입 효과입니다.



본 프로젝트 진행 이후 단계에서의 활용 및 기대 효과입니다.



Next Step with BigQuery ML

BQML을 활용한 이점

- BigQuery에 저장되어 있는 데이터를 별도 이동없이 그대로 활용해 ML 모델 빌드
- SQL이 익숙한 분석 실무자들이 직접 ML 모델 빌드 및 실행

1

매출 증가

- 추천 시스템 도입
-> 매출 증대



2

재고 관리

- 수요 예측
-> 재고 관리 효율 증가



3

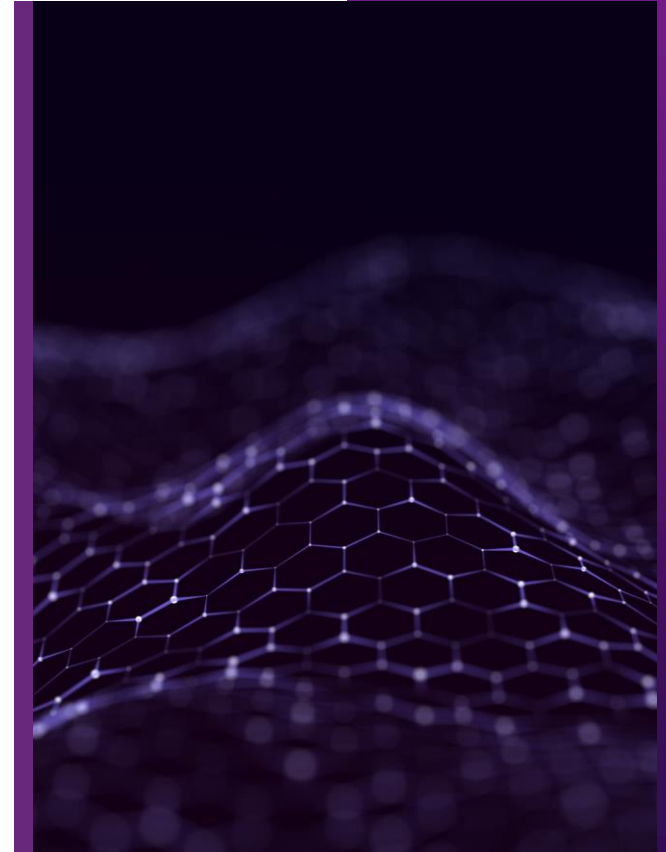
우수 고객 확보

- 고객 Segmentation
-> 타겟 마케팅, 우수 고객 확보



03 Session 2

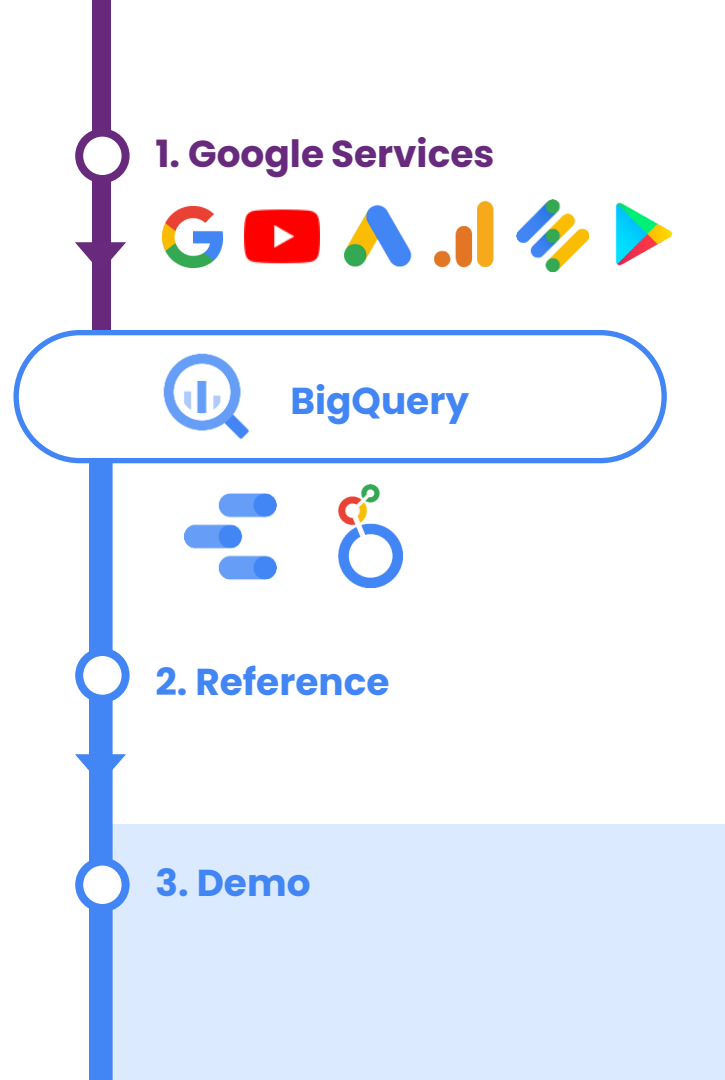
Integration BigQuery with GWS & GA



03 Session 2

Integration BigQuery with GWS & GA

- 3-1. BigQuery with Various Data
- 3-2. Reference
- 3-3. Demo



3-1 BigQuery with Various Data



BigQuery



3-2 Reference

BigQuery는 구글 내 서비스부터 3rd Party 제품까지 다양한 데이터 소스를 지원합니다.

Google Workspace

- GWS 리소스 활동 감사
- GWS 서비스 사용량 수집
- 조직 자산의 보안관리 지원



Google Analytics

- 사용자의 접속정보 및 활동기록 수집
- Web & Application 모두 지원



BigQuery

Additional Resources

- Youtube 채널 사용자 활동 로그,
Play Store 예상매출 및 순익 등
사용자의 콘텐츠 활동기록

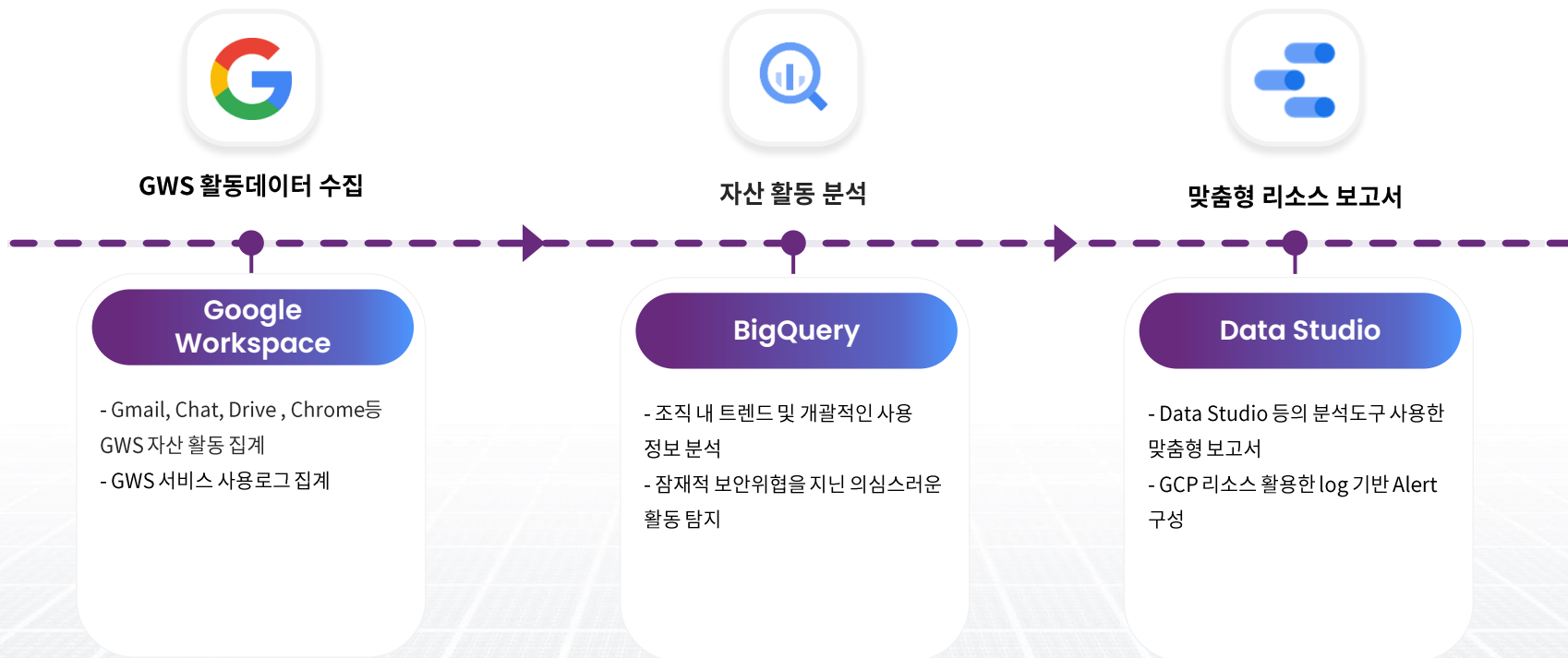


Google Spreadsheet

- 데이터 분석을 위한 단일소스 저장소
- 익숙한 인터페이스를 통한 데이터 관리
- 파일 링크만으로 BigQuery에서 쿼리 수행 가능



BigQuery에 조직 내 리소스 사용에 대한 감사로그를 저장하고, GWS 로그를 분석 할 수 있습니다.



지원버전: Enterprise, Education Standard, Education Plus

Google Analytics를 통해 수집한 고객의 행동데이터를 BigQuery와 연결하여 고객에 대한 이해도를 높일 수 있습니다.

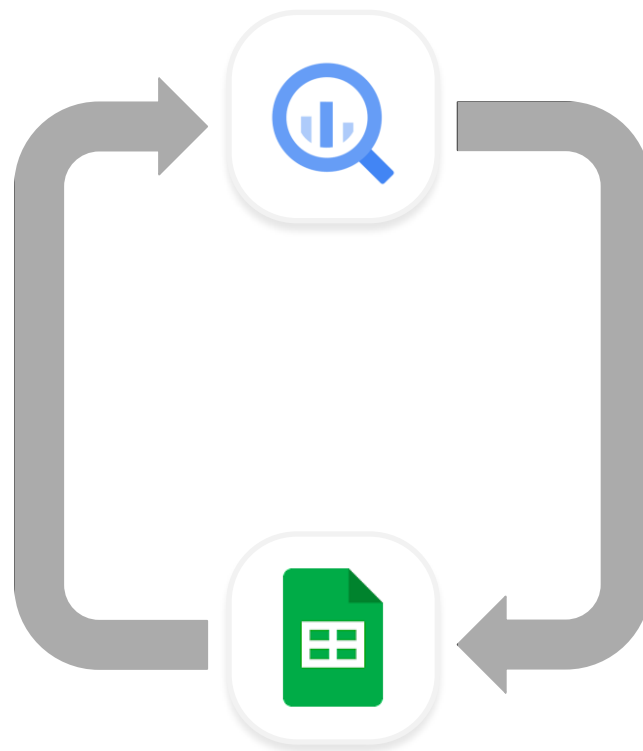


GA Standard의 일일 BigQuery Export 한도: 이벤트 100만 개

BigQuery Data Transfer Service를 통해 다양한 Google 제품과 3rd party 제품을 분석할 수 있습니다.



정형데이터 포맷인 Spreadsheet는 다양한 방법으로 BigQuery와 함께 사용할 수 있습니다.



External Table

- 원본데이터유지
- 쿼리대상 범위 지정하여 사용가능
- 쿼리대상 데이터가 BigQuery 외부에 있어 속도저하
⇒ 일회성 프로세스에 유용

Connected Sheet

- 익숙한 인터페이스에서 다른 이해관계자와 공동작업수행
- 별도 내보내기 작업 없이 데이터 출처 확인 가능
- 보고 및 대시보드 Workflow 간소화



BigQuery

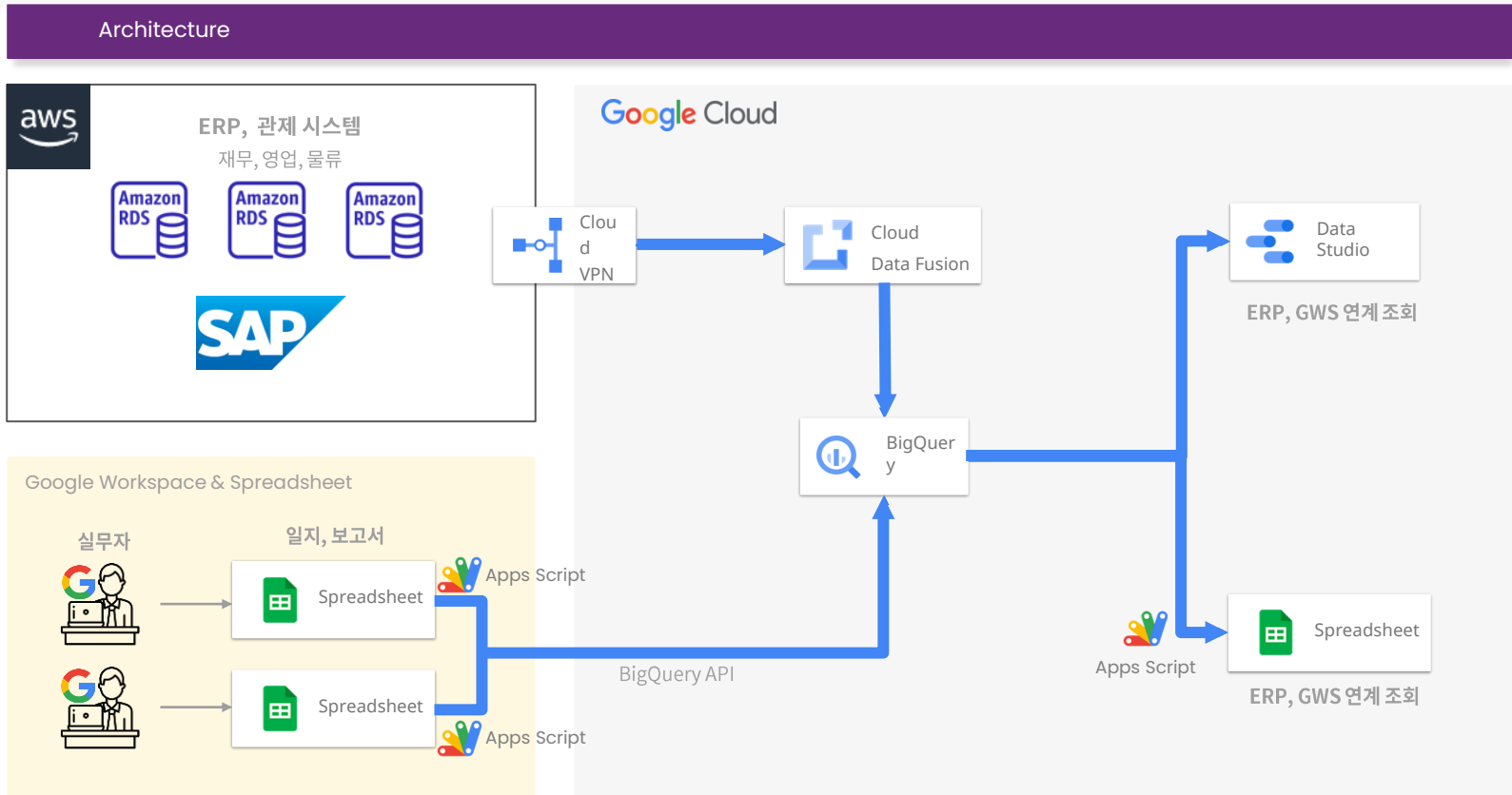
3-2 Reference

국내 환경 제조업체 E사

3-3 Demo

Reference

Spreadsheet와 BigQuery를 결합하여 데이터를 수집, ETL, 저장 및 분석하는 환경을 구축한 사례입니다.



Reference) 도입효과

본 사례의 도입 효과입니다.

GWS로
사용자를 인증하는
Private 환경



운영 및 분석비용
절감

BigQuery에 통합된
실시간 분석 환경 구축

시각화 된 대시보드로
주요 KPI 분석 가능



3-3 Demo

04 Summary



BigQuery를 사용하는 이유



Thank you!



Cloócus