## BigQuery Architecture Fundamentals





## **Topics**

01	Introduction to BigQuery
02	BigQuery Core Infrastructure
03	BigQuery Storage
04	BigQuery Query Processing
05	BigQuery Data Shuffling

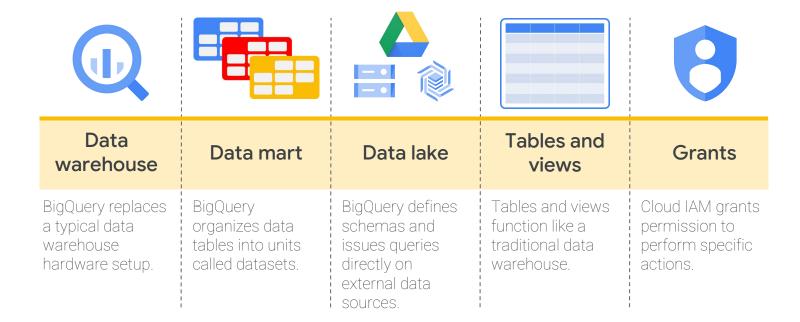




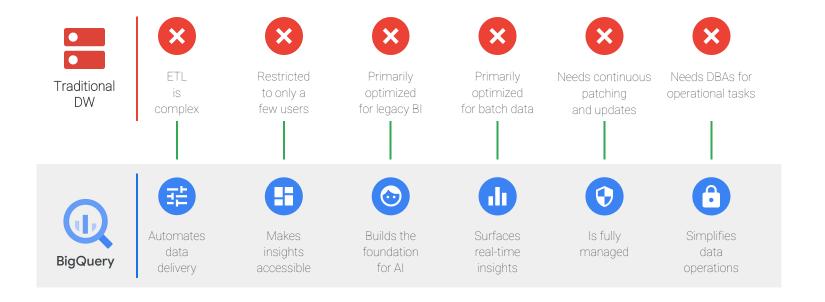


## Introduction to BigQuery

## BigQuery is Google's data warehouse solution



# BigQuery is a modern data warehouse that changes the conventional mode of data warehousing



## BigQuery has many capabilities that make it an ideal data warehouse

- Interactive SQL queries over large datasets (petabytes) in seconds
- Serverless and no-ops, including ad hoc queries
- Ecosystem of visualization and reporting tools
- Ecosystem of ETL and data processing tools
- Up-to-the-minute data
- Machine learning
- Security and collaboration



## BigQuery is a serverless, fully managed service





Storage management

Hardware

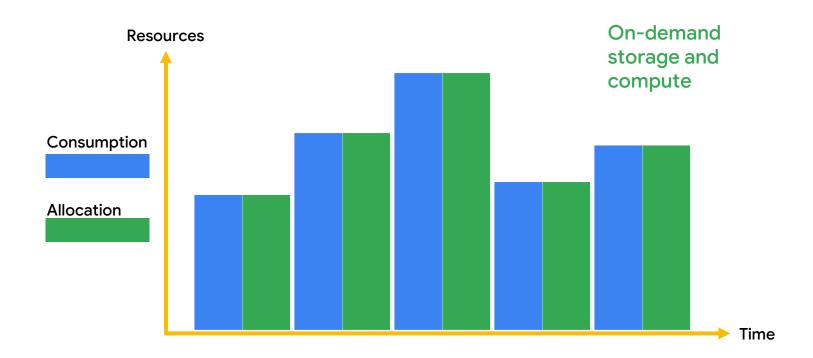
X Fault recovery



Free up real people-hours by not having to manage common tasks.



## You don't need to provision resources before using BigQuery



## Rule #1

#### Don't optimize prematurely:

- Ignore all other best practices.
- Try out your query.
- If your query runs fast/inexpensive enough, leave it alone.

### Fun with numbers

#### Analyze highly complex data at any scale



Q



350 PB of data

100,000,000,000,000 rows (one hundred trillion) 10,000 concurrent queries

Stored by one customer

Queried by multiple customers

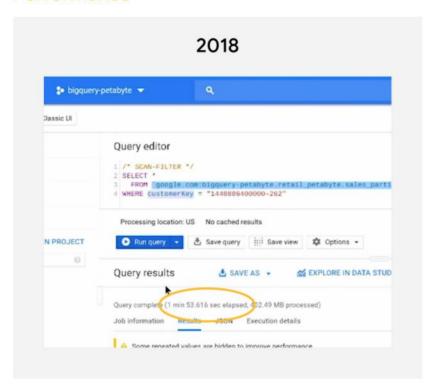
Run by another customer

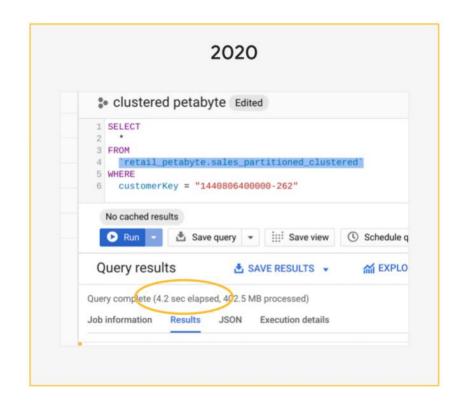
## Rule #2

BigQuery is always getting better/faster. Read Rule #1.

## Continuous performance improvement

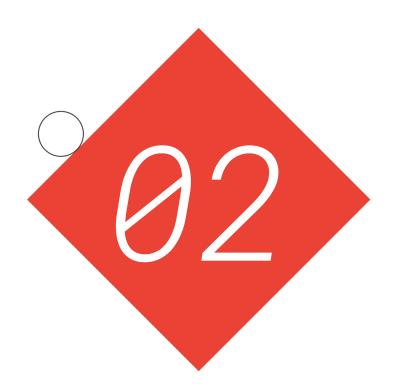
#### Performance





### Rule #3

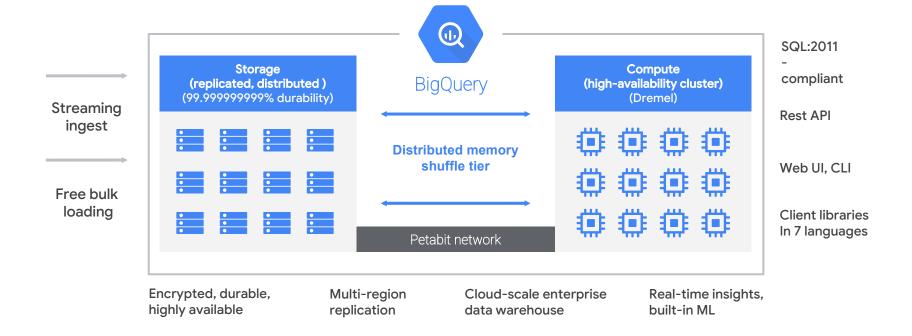
You're taking this course because Rules #1 and #2 are occasionally not enough.



## BigQuery Core Infrastructure

### Fully managed and serverless

#### Decoupled storage and compute for maximum flexibility



## Key architecture design principles

#### Storage and compute separation

- Petabyte-scale
- High availability
- Serverless and multi-tenant

#### **Colocation and caching**

High performance at low cost

#### **Integrated hardware/software stack**

- Benefit from hardware primitives
- High performance at low cost

#### **Integration with Google Cloud**

- Common security and privacy policies across products
- Seamless Google Cloud experience

## BigQuery service locations

#### BigQuery is a regional service.

- Regions (us-east4, europe-west4)
  - O Multiple zones, one or more campuses, single metropolitan area, single jurisdiction
  - O Data residence and colocation guarantees
- Multi-Regions US, EU
  - O Multiple zones, multiple campuses, multiple metropolitan areas
  - Flexible capacity planning
  - O Generally less expensive
  - O Improved durability due to off-region backups

Google Cloud zone: A zone is a deployment area for Cloud Platform resources within a region. Zones should be considered a single failure domain within a region.

## BigQuery service deployment

#### Global layer

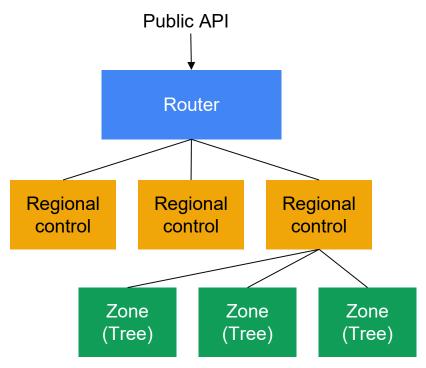
- Components needed to route to the right region
- Spread across zones globally

#### Regional layer

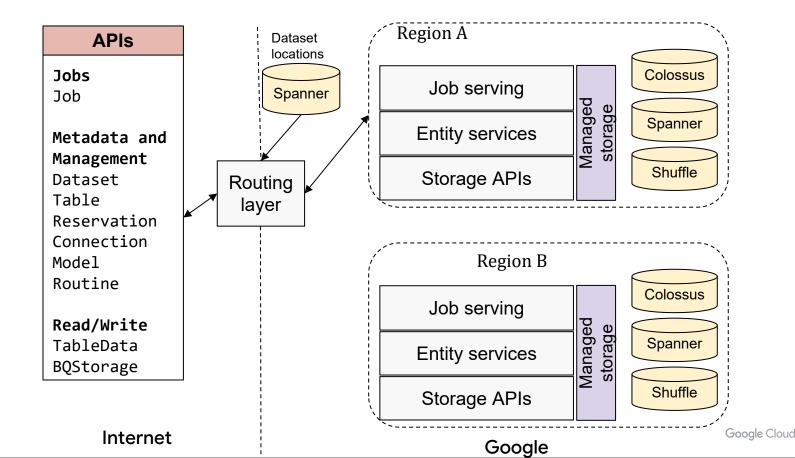
- Manage capacity, data, and metadata redundancy
- Spread across zones within the region

#### Zonal layer

Compute and storage backend



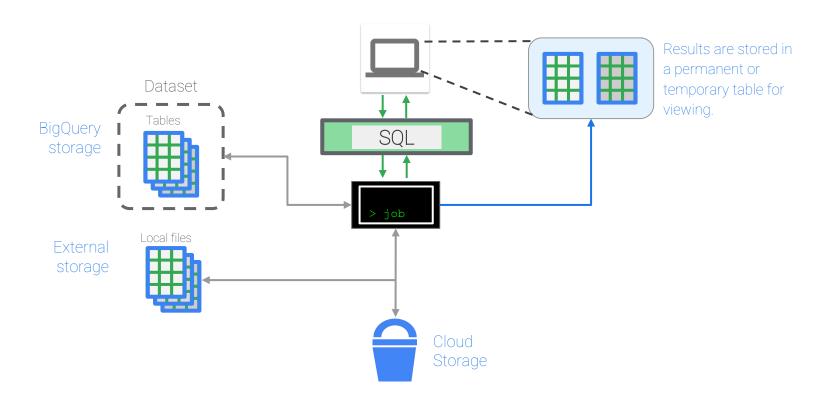
## Service layers and components



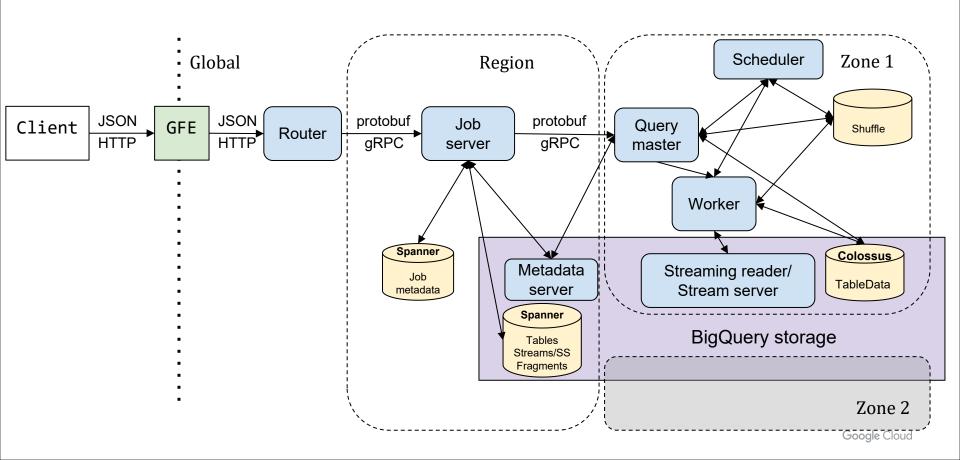
#### **APIs**

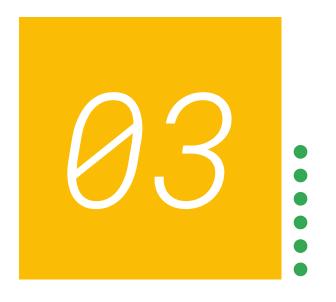
- Job/Query Run a single SQL query or a script. Load or export data.
  - O UI, ODBC/JDBC, Command line client, Looker
  - O Example: jobs.insert(), jobs.query(), jobs.getQueryResults()
- Storage Read from and write to BigQuery tables.
  - O UI, Dataflow and Dataproc, custom code
  - Example: tabledata.insertAll().
- Metadata Create dataset. Add a routine or script.
  - O UI, SQL, CLI
  - O Example: datasets.insert(), tables.list()
- Management Create and modify reservations.
  - O UI, CLI

## The life of a BigQuery SQL query



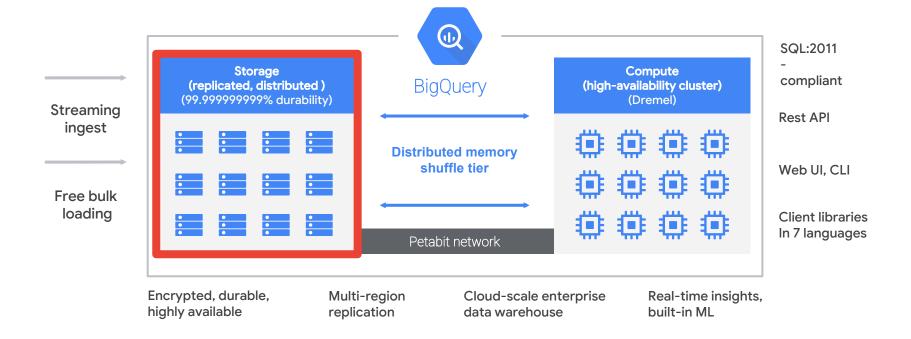
## The internal life of a BigQuery SQL query



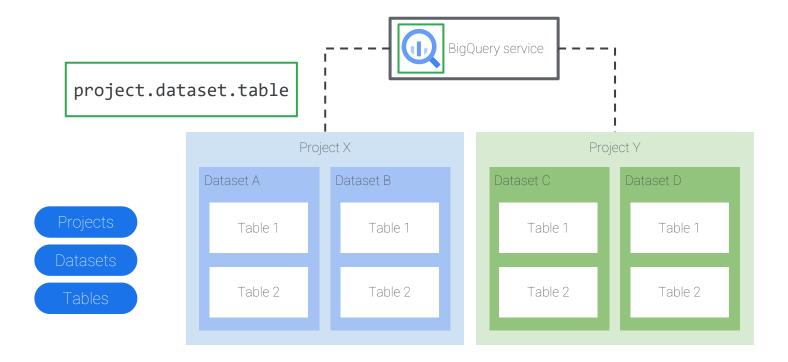


## **BigQuery Storage**

## BigQuery architecture



## BigQuery organizes data tables into units called datasets



## What makes BigQuery fast?



## Column-oriented versus row-oriented storage

- Read less data faster.
- Skip unused columns.
- Column compression versus row compression.
- Supports vectorized columnar processing.

## BigQuery divides all tables into smaller shards of data to enable massively parallel reads and operations

#### One large table

Company ID	Company Name
161218560	NY Association Inc.
•••	•••
	•••

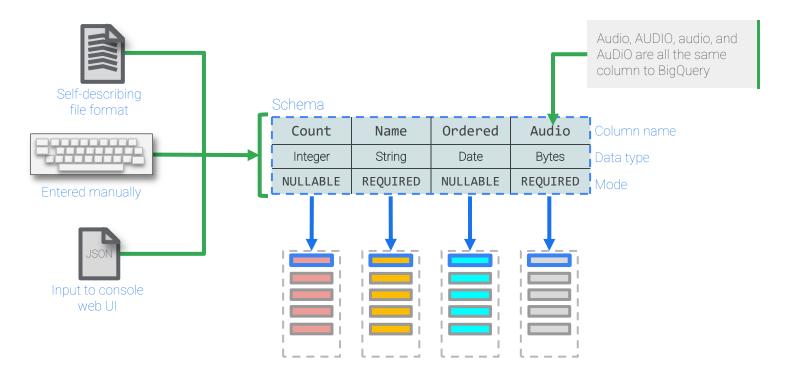
One table with 10 billion rows

BigQuery divides the table (not seen to you) into many small shards.



10 billion rows spread across many shards

### The table schema provides structure to the data



## Physical layout

#### Capacitor: Our proprietary columnar format

- Maintain the optimum sharding structure.
- Implement the logical metadata hints: partitioning/clustering.

#### Why create a new proprietary format?

- Google can apply what it learned over the last 10 years.
- The format is deeply tied to the execution engine.
- Google can improve it easily.

## Capacitor features

- Dictionary encoding (low cardinality)
- Constraints and Bloom filters (high cardinality)
- Run-length encoding
- Compression
- Row reordering

## Physical metadata

#### Critical part of BigQuery storage that is designed to support:

- Streaming
- ACID commits
- Time travel (7 days)
- Backups
- Active storage management
- Storage optimization
- Partitioning and clustering
- DML

```
CREATE OR REPLACE TABLE restored_table AS

SELECT

*

FROM current_table

FOR SYSTEM_TIME AS OF

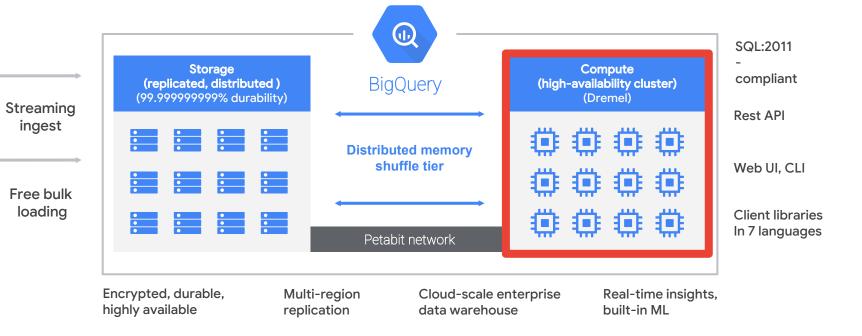
TIMESTAMP_SUB(CURRENT_TIMESTAMP(),

INTERVAL 24 HOUR)
```

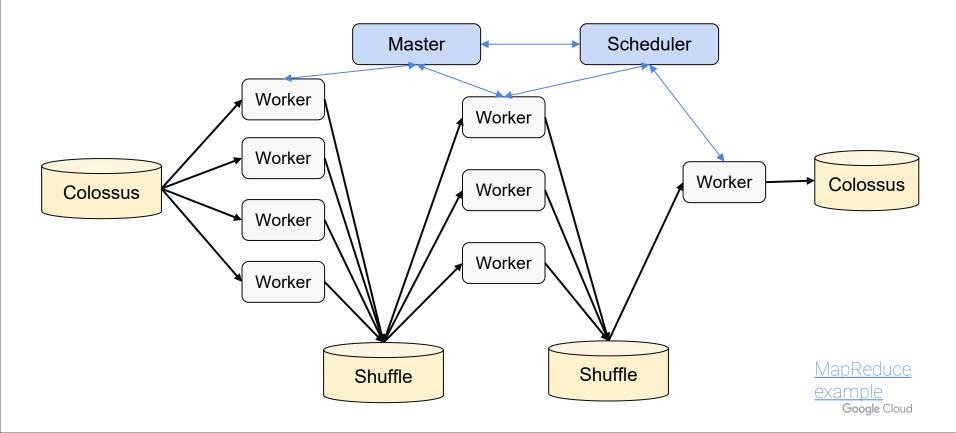


# BigQuery Query Processing

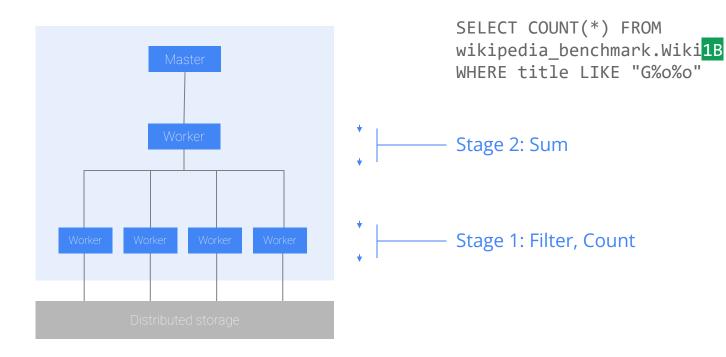
## BigQuery architecture



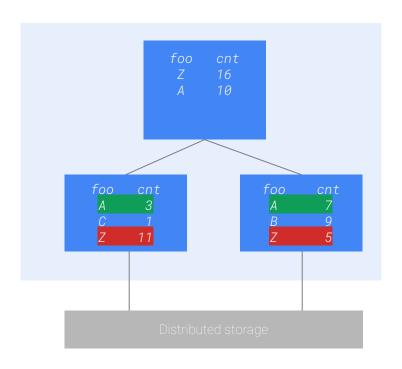
## BigQuery processes data in a distributed way



## Simple query execution



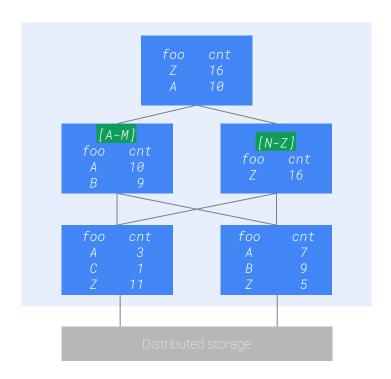
### Aggregation with high cardinality



```
SELECT foo, COUNT(*) as cnt FROM `...`
GROUP BY 1
ORDER BY 2 DESC
LIMIT 2
```

- Can't discard 'B' or 'C' until after all previous stages are complete.
- High cardinality 'foo' will overwhelm the root node.

### Aggregation with shuffle



```
SELECT foo, COUNT(*) as cnt FROM `...`
GROUP BY 1
ORDER BY 2 DESC
LIMIT 2
```

- Shuffle puts like values in the same node.
- This is scalable because you never have to return more than the LIMIT value from each node in the middle tier.

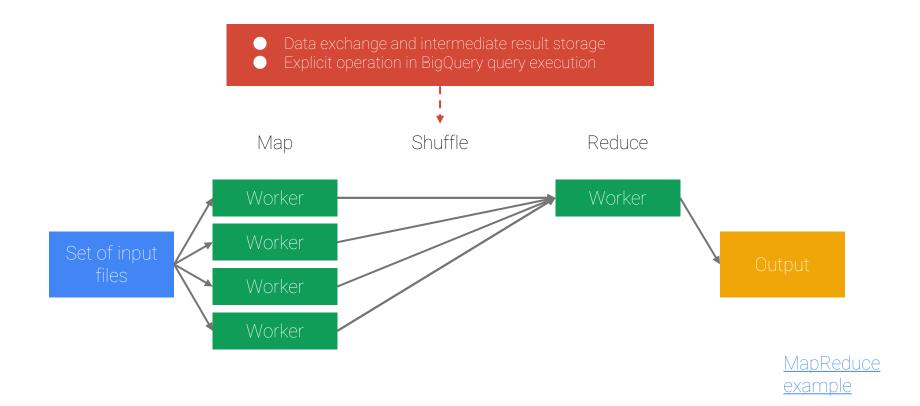


# BigQuery Data Shuffling

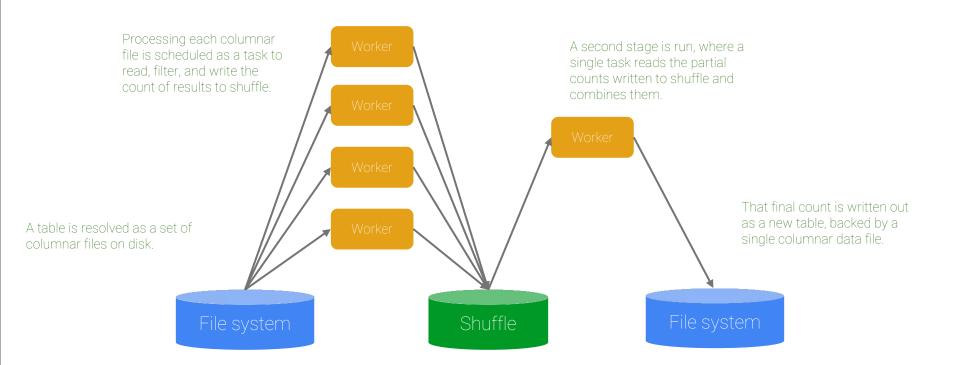
### BigQuery architecture



## Parallel to Hadoop/MapReduce



### With more detail

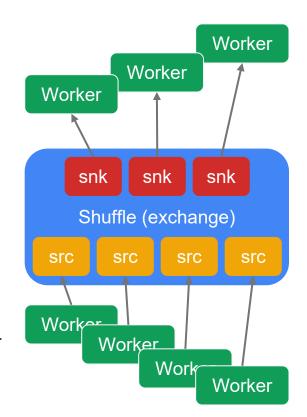


### Parallel to traditional parallel query execution

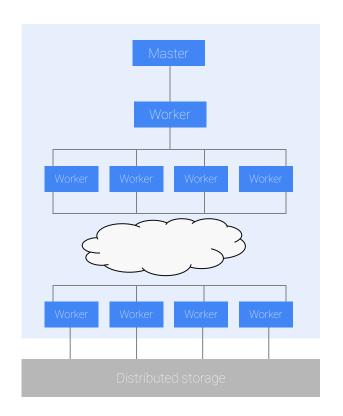
#### Shuffle is similar to Exchange operator

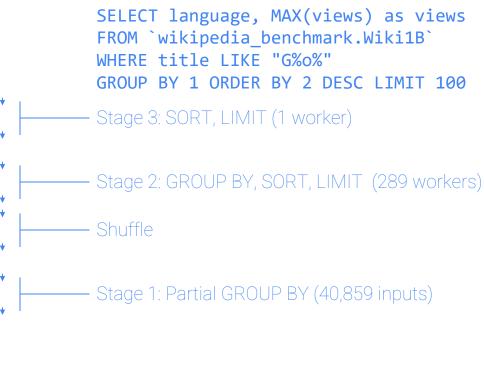
Pluggable way of changing degree of parallelism in query execution

- M inputs, N (disjoint) outputs
  - Read data from multiple (M) inputs.
  - Determine receiving output through some partitioning scheme (e.g., hash, range).
  - Write data to N outputs.
- The shuffle is the BigQuery-specific implementation of an exchange.
  - Sources model the exchange inputs.
  - Sinks model the exchange outputs.
  - Data reads and writes are orchestrated through Mindmeld, a distributed inmemory file system.



## Shuffle aggregation execution





# Shuffle aggregation execution

#### Stage 1

Row	language	views	
1	hr	131	
2	tl	160	
3	id	257	max
4	id	114	
5	da	101	
6	da	124	max
7	meta	166	

#### Stage 2

Row	language	views	
1	en	89322	max
2	en	47900	
3	en	43067	
4	en	38988	
5	en	37628	
6	en	37089	
7	en	35870	

#### Stage 3

Row	language	views
1	en	89322
2	it	38611
3	de	27715
4	pt	22974
5	tr	13552
6	fr	12447
7	meta	11117

### Query execution design choices

#### Shuffle is the data transfer mechanism between workers:

- Allows flexible query planning and execution.
- Can act as a staging area or partitioning mechanism.

### Query optimization using dynamic query execution:

- Observe execution and quickly react.
- Is more robust than static (cost-based) query optimization.

### Decouple **scheduling** from query **planning**.

# **Questions?**





### Lab Intro

Using BigQuery to Do Analysis

#### Objectives

- Execute interactive queries in the BigQuery console.
- Combine and run analytics on multiple datasets.

