

# Module 13

## Optimizing for Performance

*In this module we will:*

- **Avoid BigQuery Performance Pitfalls**
- Prevent Hotspots in your Data
- Diagnose Performance Issues with the Query Explanation map

# Four Key Elements of Work

- **I/O** – How many bytes did you read?
- **Shuffle** – How many bytes did you pass to the next stage?
  - Grouping – How many bytes do you pass to each group?
- **Materialization** – How many bytes did you write to storage?
- **CPU work** – User-defined functions (UDFs), functions



# Avoid Input / Output Wastefulness

- Do not SELECT \*, use only the columns you need
- Filter using WHERE as early as possible in your queries
- Do not use ORDER BY without a LIMIT



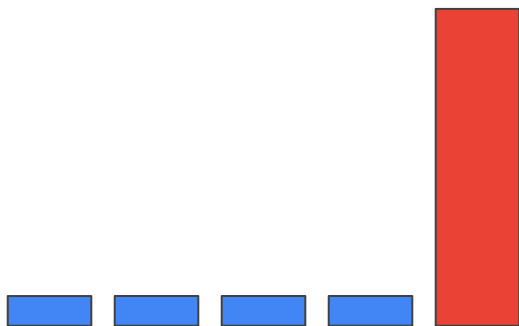
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# Shuffle Wisely: Be Aware of Data Skew in your Dataset



Skewed Data creates an imbalance between BigQuery worker slots (uneven data partition sizes)

- **Filter your dataset** as early as possible (this avoids overloading workers on JOINS)
- Hint: Use the Query Explanation map and compare the Max vs the Avg times to highlight skew
- BigQuery will automatically attempt to reshuffle workers that are overloaded with data

# Careful use of GROUP BY

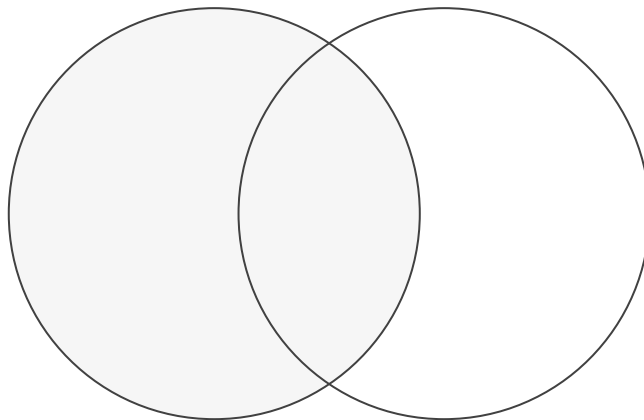
- Best when the number of distinct groups is small (fewer shuffles of data).
- Grouping by a high-cardinality unique ID is a bad idea.

Row	contributor_id	LogEdits
1	2221364	4
2	104574	4
3	73576	4
4	311307	4
5	291919	4
6	140178	4
7	181636	4
8	3661553	4
9	3600820	4
10	4737290	4
11	938404	4
12	295955	4
13	183812	4
14	1811786	4
15	8918196	4
16	561624	4
17	5338406	4

← Do not group on an ID

# Joins and Unions

- Know your join keys and if they're unique -- no accidental cross joins
- LIMIT Wildcard UNIONS with `_TABLE_SUFFIX` filter
- Do not use self-joins (consider window functions instead)



# Limit UDFs to Reduce Computational Load

- Use native SQL functions whenever possible
- Concurrent rate limits:
  - for non-UDF queries: 50
  - for UDF-queries: **6**

```
CREATE TEMP FUNCTION SumFieldsNamedFoo(json_row STRING)
  RETURNS FLOAT64
  LANGUAGE js AS """
function SumFoo(obj) {
  var sum = 0;
  for (var field in obj) {
    if (obj.hasOwnProperty(field) && obj[field] != null) {
      if (typeof obj[field] == "object") {
        sum += SumFoo(obj[field]);
      } else if (field == "foo") {
        sum += obj[field];
      }
    }
  }
  return sum;
}
var row = JSON.parse(json_row);
return SumFoo(row);
""";
```



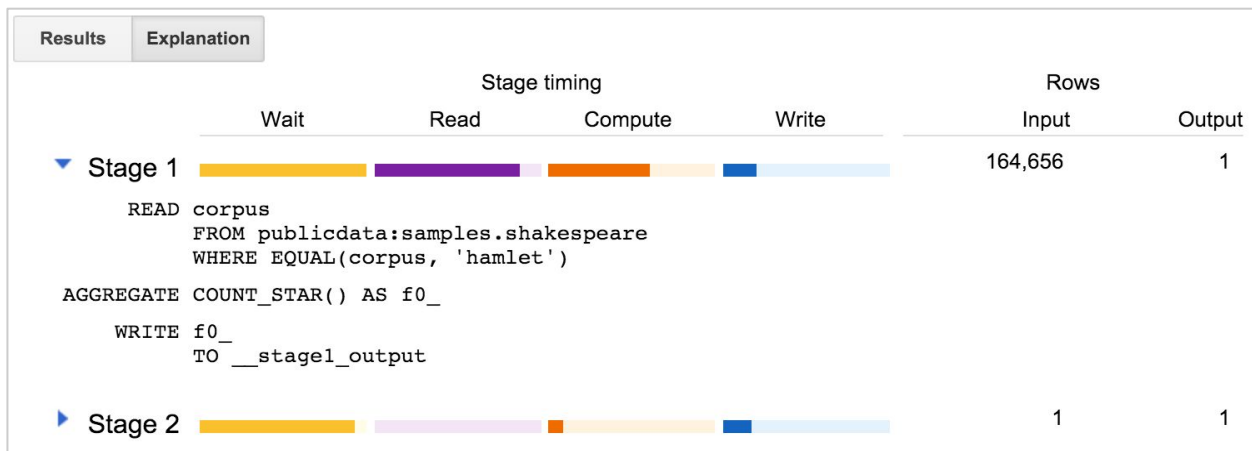
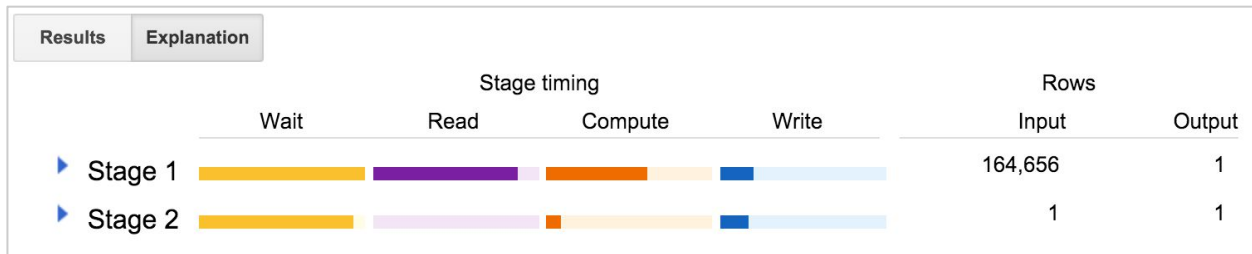
# Module 13

## Optimizing for Performance

*In this module we will:*








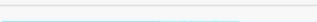
- Avoid BigQuery Performance Pitfalls
- Prevent Hotspots in your Data
- **Diagnose Performance Issues with the Query Explanation map**

# Diagnose Performance Issues with the Query Explanation Map



# Diagnose Performance Issues with the Query Explanation Map

The following ratios are also available for each stage in the query plan.

API JSON Name	Web UI*	Ratio Numerator **
waitRatioAvg		Time the average worker spent waiting to be scheduled.
waitRatioMax		Time the slowest worker spent waiting to be scheduled.
readRatioAvg		Time the average worker spent reading input data.
readRatioMax		Time the slowest worker spent reading input data.
computeRatioAvg		Time the average worker spent CPU-bound.
computeRatioMax		Time the slowest worker spent CPU-bound.
writeRatioAvg		Time the average worker spent writing output data.
writeRatioMax		Time the slowest worker spent writing output data.

\* The labels 'AVG' and 'MAX' are for illustration only and do not appear in the web UI.

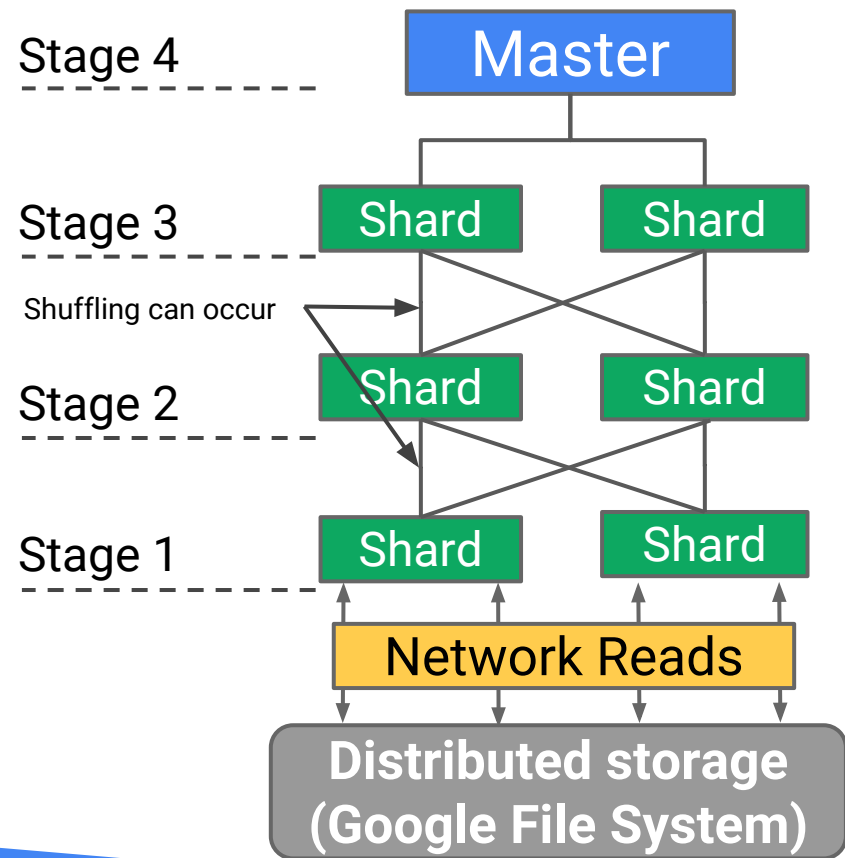
\*\* All of the ratios share a common denominator that represents the longest time spent by any worker in any segment.

# Example: Large GROUP BY

Large GROUP BY query

```
SELECT
  LogEdits, COUNT(contributor_id) AS Contributors
FROM (
  SELECT
    contributor_id,
    CAST(LOG10(COUNT(*)) AS INT64) LogEdits # Buckets user edits
  FROM `publicdata.samples.wikipedia`
  GROUP BY contributor_id)
GROUP BY LogEdits
ORDER BY LogEdits DESC
```

# Large GROUP BY Means Many Forced Shuffles



```
READ Contributors, LogEdits FROM stage3 output AS  
publicdata:samples.wikipedia  
SORT LogEdits DESC  
WRITE Contributors, LogEdits to output
```

```
READ Contributors, LogEdits from stage2 output AS  
publicdata:samples.wikipedia  
AGGREGATE SUM_OF_COUNTS(Contributors) AS Contributors  
GROUP BY LogEdits  
WRITE Contributors, LogEdits to stage3 output
```

```
READ contributor_id, f0_ FROM stage1 output AS  
publicdata:samples.wikipedia  
AGGREGATE SUM_OF_COUNTS(f0_) AS f0_ GROUP BY contributor_id  
COMPUTE INTEGER(LOG10(f0_))  
AGGREGATE COUNT(contributor_id) AS Contributors GROUP BY  
LogEdits  
WRITE Contributors, LogEdits to stage2_output BY  
HASH(LogEdits)
```

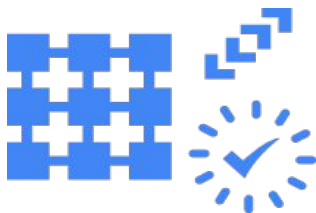
```
READ contributor_id FROM publicdata:samples.wikipedia  
AGGREGATE COUNT_STAR() AS f0_ GROUP BY contributor_id  
WRITE contributor_id, f0_ to stage1 output BY  
HASH(contributor_id)
```

# Table Sharding - Then and Now



Traditional databases get performance boost by partitioning very large tables

Usually requires an administrator to pre-allocate space, define partitions, and maintain them



Manual **Table sharding** divides big table into smaller tables with new suffix of YYYYMMDD

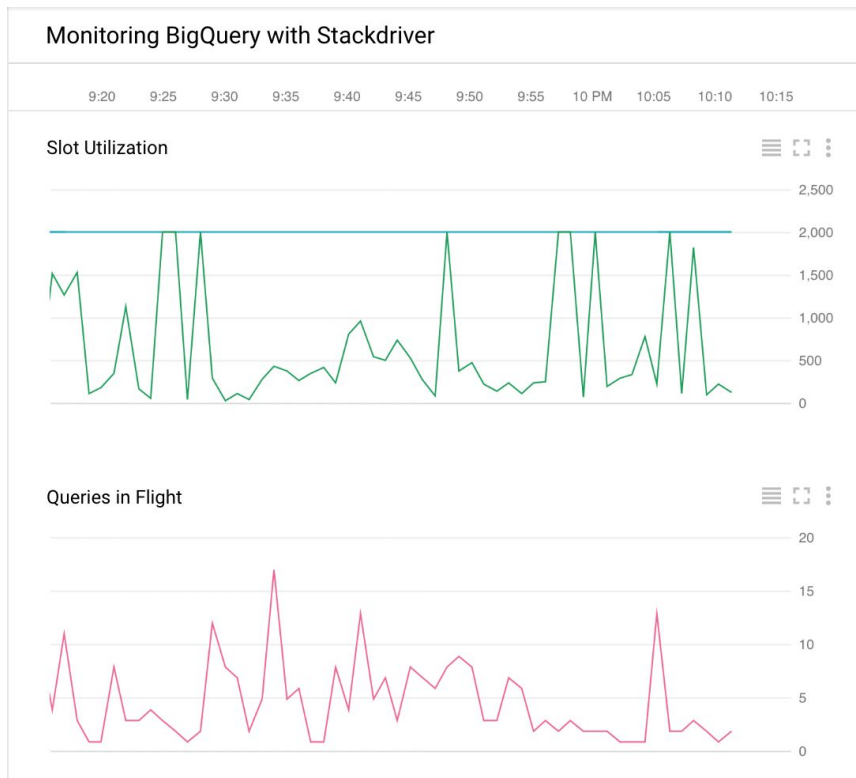
Queries use Table Wildcard functions



**Date Partition** a single table based on specified DAY or a Date Column

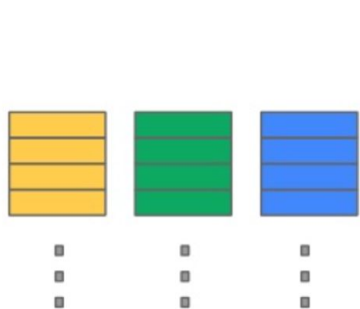
[Creating Partitioned Tables](#)

# Monitor Performance with Stackdriver

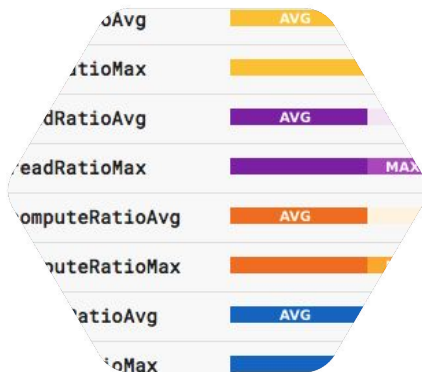


- Available for all BigQuery customers
- Fully interactive GUI. Customers can create custom dashboards displaying up to 13 BigQuery metrics, including:
  - Slots Utilization
  - Queries in Flight
  - Uploaded Bytes (not shown)
  - Stored Bytes (not shown)

# Summary: Query and data model design has a significant impact on performance



Query only the rows and columns you need to reduce bytes processed



Investigate the query explanation map to see if data skew is bottlenecking your query



Avoid SQL anti-patterns like ORDER BY without a LIMIT or a GROUP BY on high-cardinality fields



Use table partitioning to reduce the volume of data scanned



# Lab 11

## Optimizing and Troubleshooting Query Performance

# Optimizing and Troubleshooting Query Performance

In this lab, you will fix and troubleshoot SQL queries for performance improvements.

