Overview

Introduce Hive as a data warehouse for analytical processing

Understand how Hive deals with queries on huge amounts of data

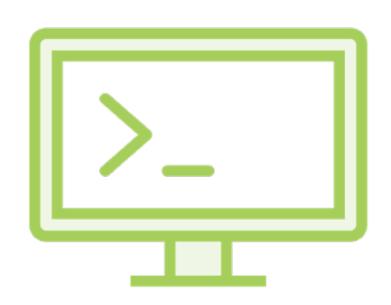
Optimizing Queries on Big Data



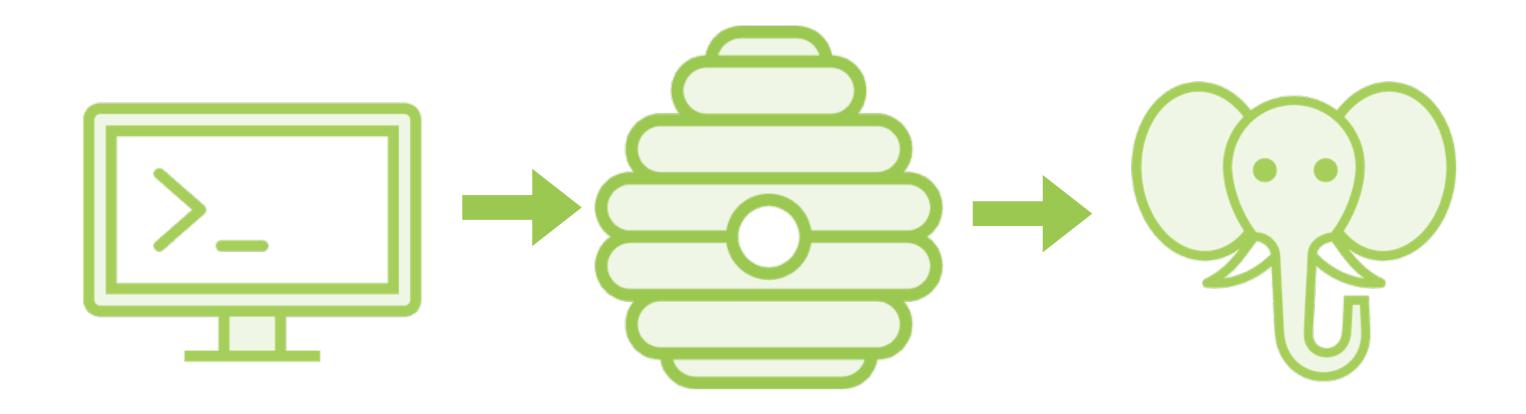
Consider an e-commerce site with transactions from all over the United States



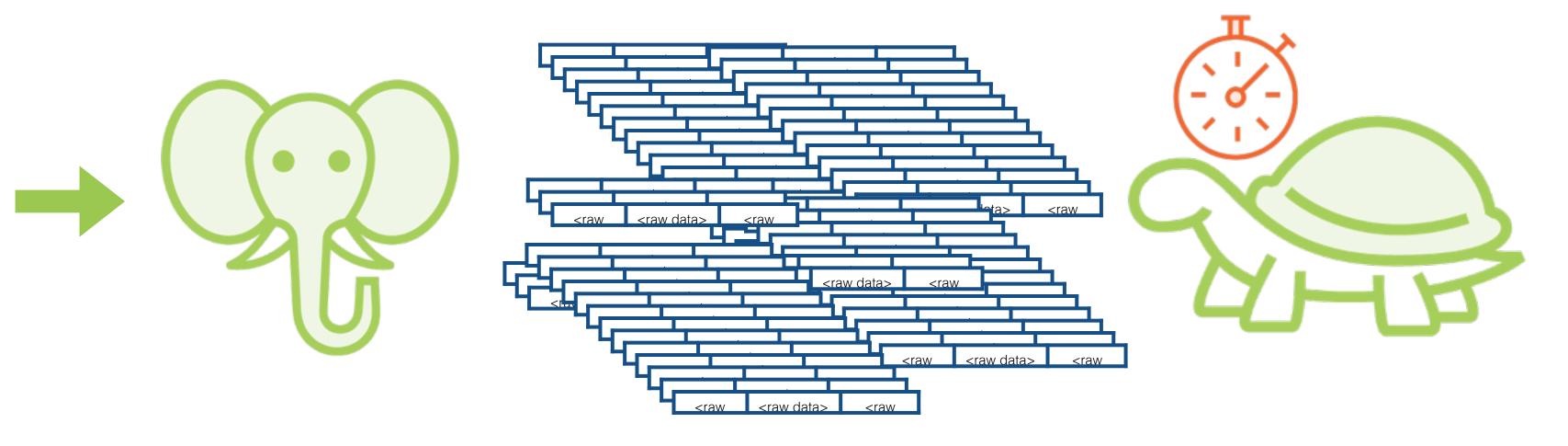
Potentially billions of records over many years



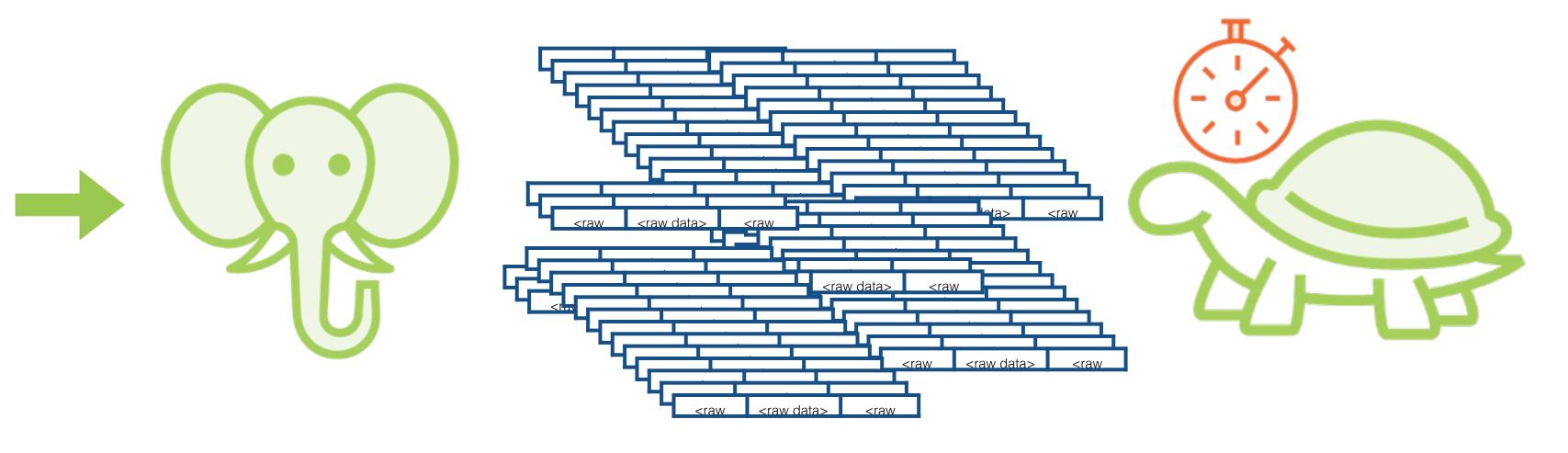
select * from orders where state = "WA"



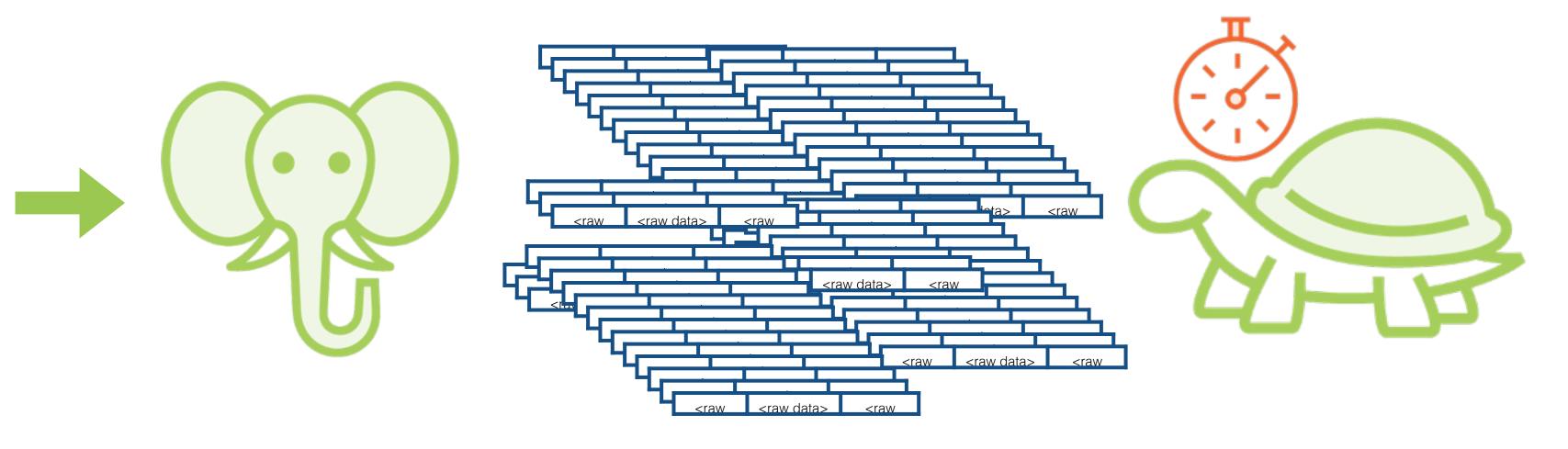
select * from orders where state = "WA"



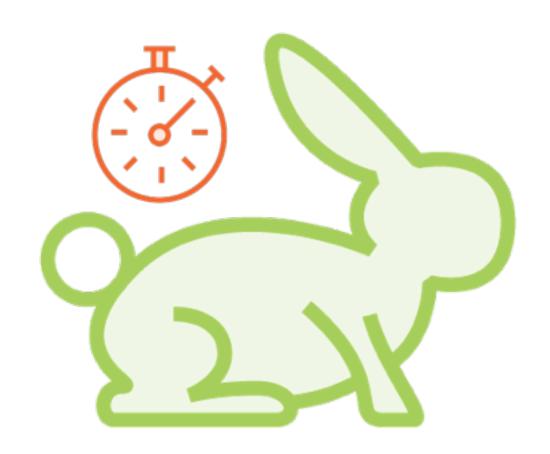
select * from orders where state = "WA"



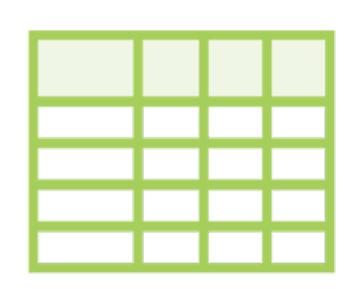
Running MapReduce jobs on huge datasets is slow...



Queries run on it are very complicated, hard to debug and maintain



Hive tries to simplify query writing and improve query performance in many ways



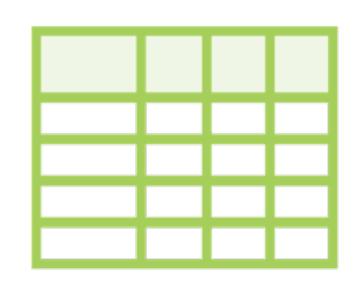




Structure queries so they run faster



Simplify query expressions so they're easy to maintain



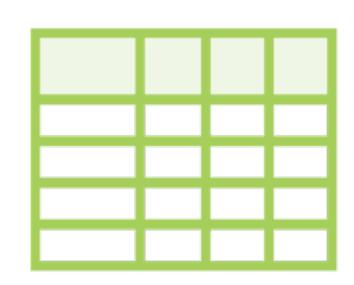




Partitioning and Bucketing of Tables

Structure queries so they run faster

Simplify query expressions so they're easy to maintain



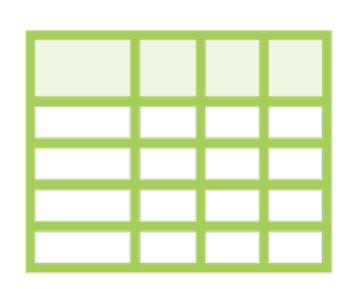




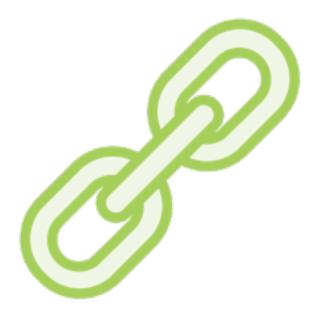
Partitioning and Bucketing of Tables

Join Optimizations

Simplify query expressions so they're easy to maintain



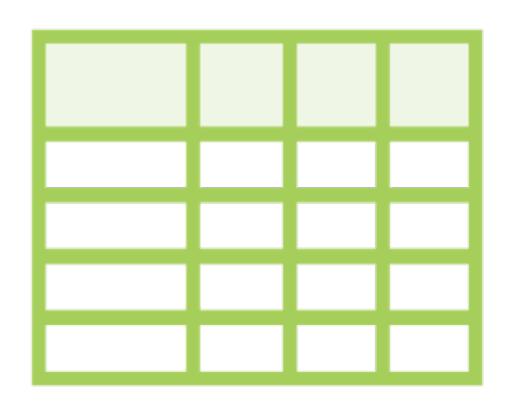




Partitioning and Bucketing of Tables

Join Optimizations

Window Functions



Design tables so queries run only on subsets of the dataset

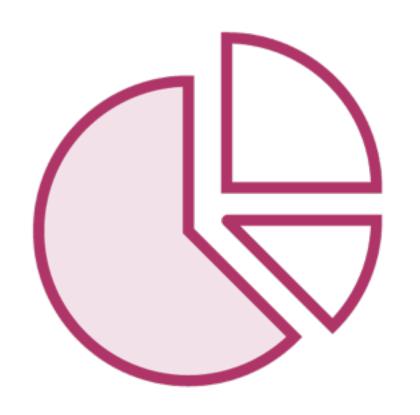






Bucketing

Splits data into smaller, manageable parts







Bucketing

Enables performance optimizations

Data may be naturally split into logical units

Customers in the US

Each of these units will be stored in a different directory

WA CT

OR

NY

CA

GA

State specific queries will run only on data in one directory

OR NY GA

Splits may not of the same size

OR NY GA



Partitioning



Bucketing

Size of each split should be the same

Customers in the US

Hash of a column value - address, name, timestamp anything

Customers in the US

Each bucket is a separate file

Bucket 1

Bucket 2

Bucket 3

Bucket 4

Makes sampling and joining data more efficient

Bucket 1

Bucket 2

Bucket 3

Bucket 4



Partitioning



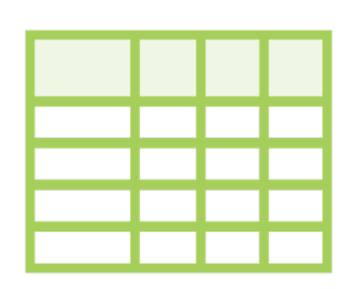
Bucketing



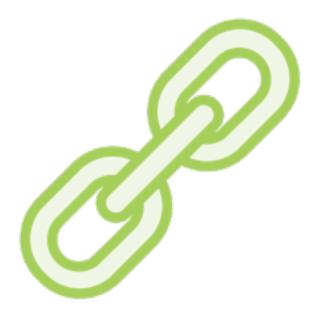
Rather than running queries on such tables



Run queries on subsets that are manageable





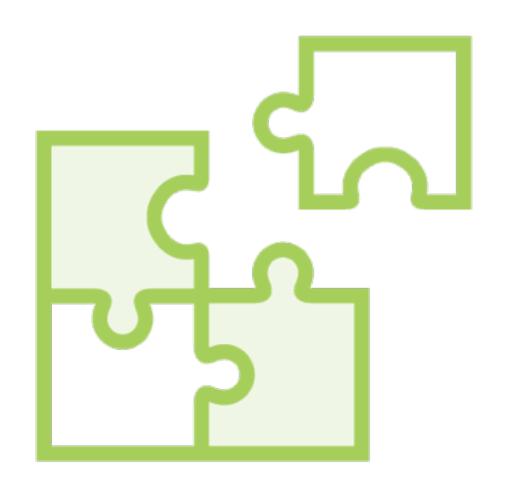


Partitioning and Bucketing of Tables

Join Optimizations

Window Functions

Join Optimizations



Join operations are MapReduce jobs under the hood

Join Optimizations

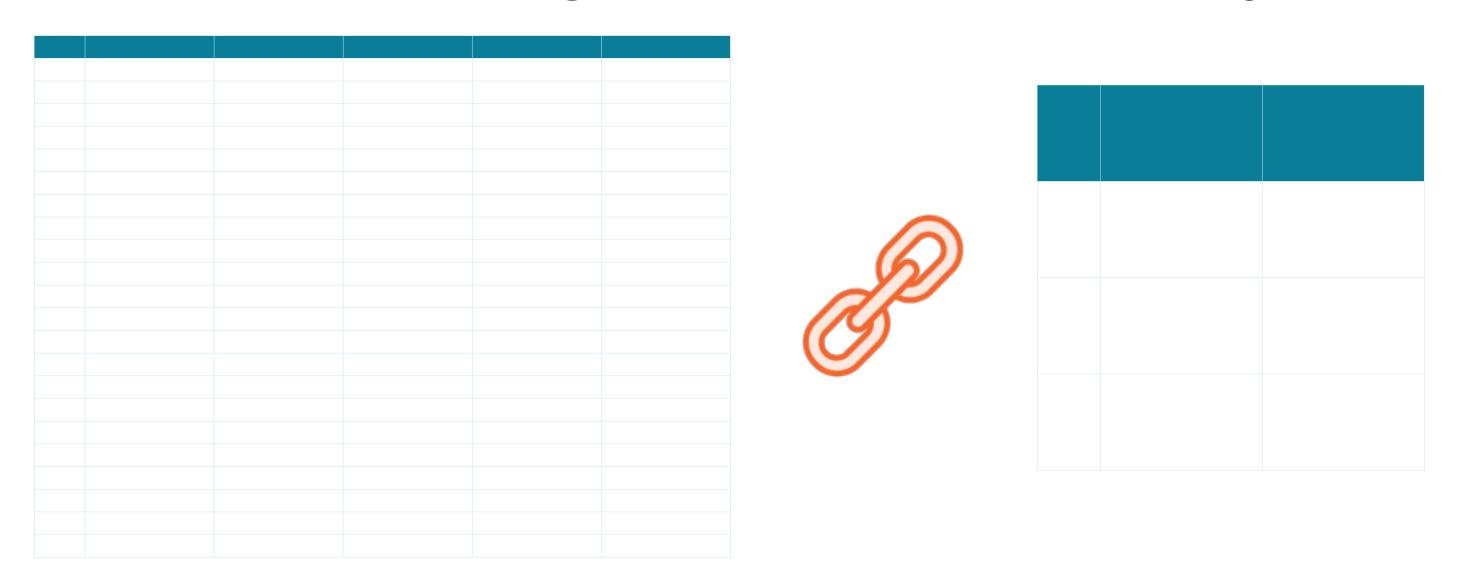


Optimize joins by reducing the amount of data held in memory

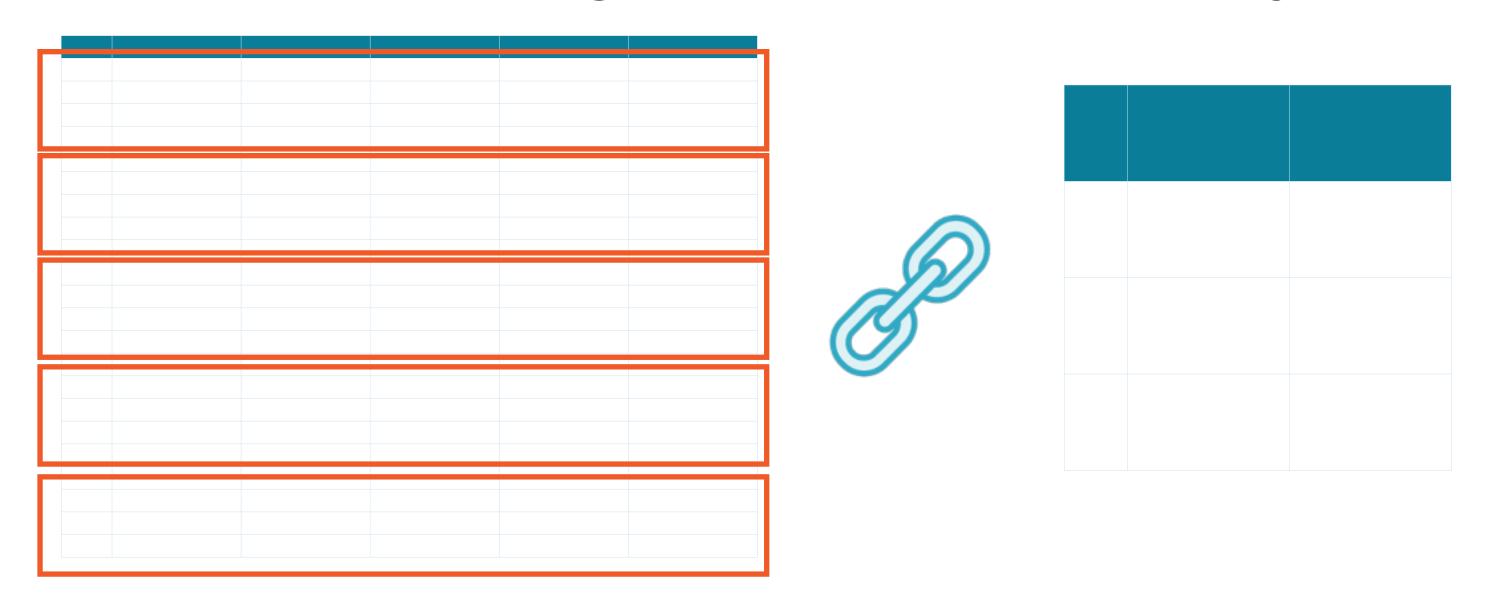
Join Optimizations



Or by structuring joins as a map-only operation



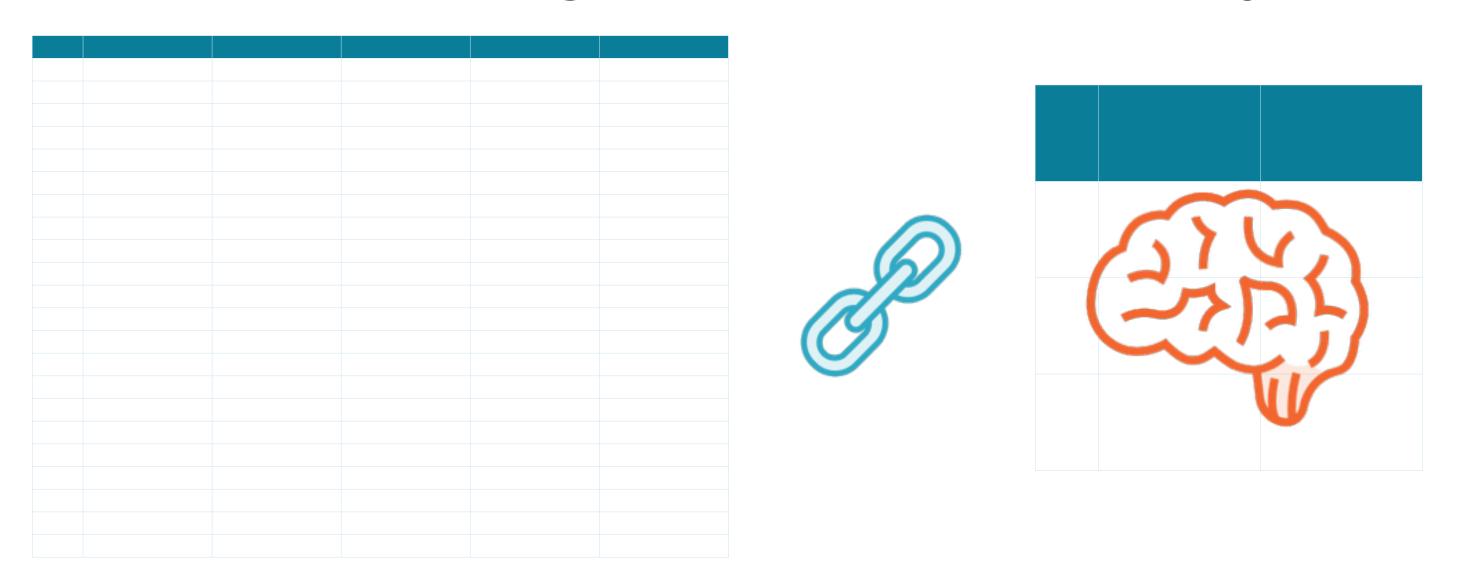
A 500GB table joined with a 5MB table



The large table will be split across multiple machines in the cluster



One table is held in memory while the other is read from disk



For better performance the smaller table should be held in memory

Joins as Map-only Operations



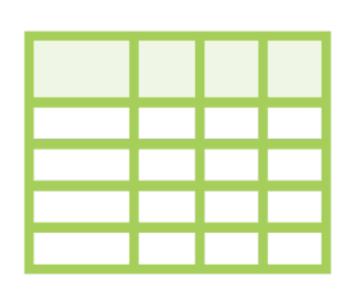
MapReduce operations have 2 phases of processing

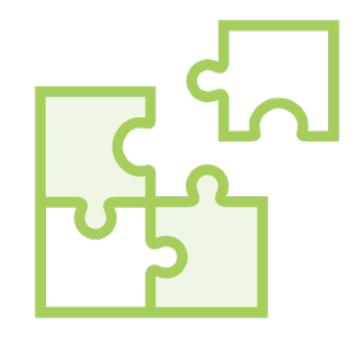


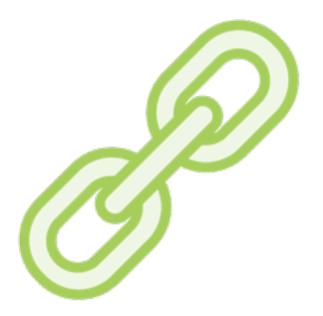
Joins as Map-only Operations



Certain queries can be structured to have no reduce phase







Partitioning and Bucketing of Tables

Join Optimizations

Window Functions

Summary

A brief overview of Hive and its place in the Hadoop eco-system

Introduced some Hive features that allows us to work with large datasets

- Partitioning and Bucketing
- Join Optimizations
- Window Functions