HadoopDB: An Architectural Hybrid of MapReduce and DBMS Technologies for Analytical Workloads

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Outline

- Context: Analytical DBMS Systems
- Background: Parallel Databases and Query Processing
- Key Properties for Very Large Scale Data Analytics
- Architecture of HadoopDB
- Performance and Scalability Results
Context: Analytical DBMS Systems

- Multi-dimensional structured data
  - Star schema: Fact tables and dimension tables

- Types of queries
  - TableScan, Joins, multi-dimensional aggregation (CUBE), Pattern Mining, Top-K and ranking

- Data explosion in terabytes and petabytes
Background: Parallel Databases

- DBMSs deployed on a shared nothing architecture
- Query execution is divided equally among all machines
- Results are computed on different machines and transferred over the network

Important tasks:
- Partitioning the tables on to several machines
- Parallel evaluation of relational query operators
Background: Query Processing

- SELECT *
  FROM R CROSS JOIN S
  WHERE R.a > 100 AND S.b < 1000

- Pipelining: Transfer intermediate results of one operator to another operator on the fly
Key properties for very large scale data analytics

- Performance: Computing the results of a query faster
- Fault Tolerance: Rescheduling parts of query execution in the case of node failures
- Adapt to heterogeneous distributed environment: Getting the same performance from all the machines is difficult
- Flexible Query interface: Should support ODBC/JDBC and user defined functions
Architecture of HadoopDB
Data Loader

- All data initially resides on the HDFS; table data is stored as raw files.
- Tables are partitioned (on-demand) and partitions are loaded onto the nodes’ file systems.
- Data that comes at each node is re-partitioned into small chunks.
- From there it is bulk-loaded into the DBMS and indexed if required.
- **Hash Partitioning:**
  - **Global Hasher:** Partition the tables which are stored as raw files on HDFS and distribute them.
  - **Local Hasher:** Partition the single-node data into file chunks and store them in disk blocks for efficient processing.
Catalog

- Metadata about tables and their partitions:
  - Attribute on which partition of a table exists in the cluster
  - Size and location of the blocks of a partition on a particular node
  - Replicas, if replicas exist for the partitions

- For each node store the DBMS connection details
  - IP Address, Driver class, username and password, database name, etc.

- MetaStore: Table schema information on the DBMSs in the nodes. Used by SMS Planner for query plan generation
SMS Planner

- Extends Hive, an SQL query processor built on top of Hadoop

- Parses the SQL Query, and transforms it into an operator DAG or the logical plan

- Generates an optimal query plan after doing any transformations

- It breaks up the plan into a batch of map and reduce functions

- Checks if a partitioning of a table exists on the join or group-by attributes and decides on map and reduce functions
SMS Planner on an example query

- SELECT YEAR(saleDate), SUM(revenue) FROM sales GROUP BY YEAR(saleDate);
SMS Planner and Hadoop Jobs

- SMS Planner generates map or reduce functions that encapsulate code about database connection and SQL query to execute.

- A `DatabaseConnector` object is created by a Map function to connect to the database using JDBC and execute SQL query.

- Assuming tables are loaded in the database, an execution of a map function triggers a database connection, query execution and transforming the `ResultSet` into key value pairs.

- Reduce function simply aggregates over the repartitioned tuples and produces output to the files.
Salient Features of HadoopDB

- Hadoop is used:
  - To store the data using the HDFS file system
  - For task scheduling, Hadoop’s JobTracker is used to schedule Map and Reduce tasks on the nodes
  - As network communication layer to transfer the intermediate results of SQL query computations between nodes
- An SQL Query is initially broken down into a batch of MapReduce jobs and then scheduled using Hadoop
- Ultimately, execution of relational query operators happens in a single node DBMS
- Queries are embedded in map and reduce functions and executed
- Results are returned as key value pairs after query execution
Performance and Scalability Benchmark

• Architectures compared:
  ◦ Hadoop
  ◦ HadoopDB
  ◦ Vertica
  ◦ DBMS-X

• Tasks evaluated in the benchmark:
  ◦ Grep
  ◦ Selection (Filtering)
  ◦ Aggregation
  ◦ Join
  ◦ UDF Aggregation
Grep Task

- Data consists of 5.6 million 100-byte records per node
- For Hadoop, a map function that performs a simple string match over records stored in a file, one per line
- Vertica, DBMS-X, HadoopDB execute the query:
  - SELECT * FROM Data WHERE field LIKE ‘%XYZ%’;
- HadoopDB performs better than Hadoop because it saves on I/O
Selection Query

- SELECT pageURL, pageRank
  FROM Rankings WHERE
  pageRank > 10;

- Hadoop as usual parses the data files and filters records

- HadoopDB pushes the execution of selection and projection operators into the PostgreSQL

- Using clustered indices boosts the performance of parallel databases and HadoopDB over Hadoop
Aggregation Query

- SELECT sourceIP, SUM(adRevenue) FROM UserVisits GROUP BY sourceIP;
- There is a map and a reduce phase in these queries
- HadoopDB pushes the SQL operators’ execution in to the PostGreSQL
- Using Hive’s query optimizer helps in choosing either sorting or hashing method to perform aggregation
Join Queries

- Hadoop supports a sort-merge kind of algorithm but incurs sorting overhead

- HadoopDB assumes a collocation of tables partitioned on the join attributes
UDF Aggregation Task

- HTML Documents are processed for counting number of out-links
- In parallel DBMS a user defined function accesses chunks of HTML documents and parses them in memory
- Outputs results of chunks on to a temporary table which are later aggregated
- Hadoop and HadoopDB executes the same and Map and Reduce code
Fault Tolerance and Heterogeneity
Conclusions

- **HADOOPDB**
  - Fault Tolerance: In the presence of node failures, Hadoop reschedules the tasks and completes the query.
  - Hadoop redundantly executes tasks of straggler nodes thus reducing effect of slow nodes on query time.
  - PostgreSQL is not a column-store and hence a drawback for HadoopDB.
  - In the event of data explosion and using several hundreds of nodes scalability comes in to picture.

- **PARALLEL DATABASES**
  - In case of node failures unfinished queries are aborted and query processing is restarted.
  - There is no way to counter the slow node's effect on overall query time.
  - Parallel databases like Vertica achieve much better performance due to column store and data compression.
  - Parallel databases are not scalable.