Introduction to NoSQL Databases

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UDBL

Evolution of DBMS technology





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Kinds of DBMS support



Simple Data Cor

Complex data



Classification of Modern Database applications

Complex Queries SQL, <mark>New QL</mark>	Business operations Business analytics Personal db	Multimedia search Custom Datatypes Data streams Web analytics
Simple Queries <mark>SQL-</mark>	E-business	Web search
No Queries <mark>NoSQL</mark>	Text, data logs Simple computations Web documents	CAD system Complex computations Web surfing



Kind of Database support

Complex Queries SQL, <mark>New QL</mark>	Relational DBMS	Object-Relational DBMS Data Stream Mgmt. Syst. Virtuoso, Neo4J, Amos II
Simple Queries SQL-	Google App Engine (GQL) Amazon SimpleDB Microsoft Azure	Google search engine Facebook HIVE
No Queries <mark>NoSQL</mark>	File systems Content Mgmt. Syst.	Object Stores Google BigTable Yahoo Hadoop MongoDB,CouchDB

What is a NoSQL Database?

- A key/value store Basic index manager, no complete query language
 - E.g. Google BigTable, Amazon Dynamo
- A web document database
 For web documents, not for small business transactions
 - E.g. MongoDB, CouchDB
- A DBMS with a limited query language Provides for high volume small business transactions
 - Sometimes called *cloud databases*
 - E.g. Google App Engine, Microsoft Azure, Amazon SimpleDB, Facebook HIVE

What is a NoSQL Database?

- A DBMS where mapreduce is used instead of queries Manual programs to iterate over entire data sets
 - E.g. Hadoop, MongoDB, CouchDB, Dynamo
- A mapreduce engine with a limited query language on top:
 - HIVE on top of Hadoop provides HIVEQL
 - Provides non-procedural data analythics (select from groupby) without detailed programming
 - Executed in batch as paralle Hadoop jobs
- A DBMS with a new query language for new applications
 - Streambase, Virtuoso, Neo4J, Amos II
- Other non-relational databases
 - Including Object Stores

NoSQL Characteristics

- Highly distributed and parallel architectures
 - Typically runs on data centers
 - This is similar to parallel databases!
- Highly scalable systems by compromised consistency
 - No 2-phase commit as in distributed databases
 - Eventual consistency
 - Or perhaps never consistency
 - Similar options available in modern DBMSs too
 - Puts burden on programmer to handle consistency!
 - Race conditions
 - Recovery
 - \Rightarrow Customized implementation of transacations
 - Mainly suitable for applications not needing consistency

NoSQL Characteristics

- New query languages for new applications
 - SQL- for cloud databases
 - Most simple web applications do not need full SQL
 - Simple SQL permits high scalability
 - Familiar model
 - Full SQL too complex for new systems
 - Graph query languages
 - SPARQL for RDF
 - RDF data model
 - For searching linked data (http://linkeddata.org/)
 - Stream query languages
 - CQL
 - Variant of SQL for streams
 - SCSQL (UU)
 - Functional parallel data stream query language

MapReduce

- Parallel *batch* processing using mapreduce
 - Many NoSQL databases uses mapreduce for parallel batch processing of data stored in data centers
 - Highly scalable implementation of
 - parallell batch processing
 - of same (e.g. Java) program
 - over large amounts of data stored in different files
 - Based on a scalable file system (e.g. HDFS)
- The mapreduce function:
 - Applies a (costly) user function *mapper* producing key/value pairs in parallel on many nodes accessing files in a cluster
 - Applies a user aggregate function on the key/value pairs produced by the mapper
 - Very similar to GROUP BY in SQL
 - Read reference article on MapReduce

Mapreduce code

function map(String name, String document):
// name: document name, i.e. HDFS file contents
// document: document contents, parsed HDFS file tokens
// Can make own parser as preprocessor

for each word w in document: emit (w, 1)

function reduce(String word, Iterator partialCounts):
// word: a word
// partialCounts: a list of aggregated partial counts

sum = 0; for each pc in partialCounts: sum += ParseInt(pc); emit (word, sum)

Mapreduce manager architecture



Mapreduce stages

- Input reader
 - System component that reads files from scalable file system (e.g. HDFS) and sends to map functions applied in parallell
- Map function
 - Applied in parallel on many different files
 - Parsers input file data from HDFS
 - Does some (expensive) computation
 - Emits key value pairs as result
 - Result stored by MapReduce system as file
- Partion function (optional)
 - Partitions output key/value pairs fro map function into groups of key/value pairs to be reduced in parallel
 - Usually hash partitioning
- Reduce function
 - Iterates over set of key/value pairs to produced a reduced set of key value pairs stored in the file system
 - C.f. aggregate functions

Mapreduce

File I/O



Hive architecture



Wordcount in Hive

FROM (MAP doctext USING 'python wc_mapper.py' AS (word, cnt) FROM docs **CLUSTER BY word** REDUCE word, cnt USING 'pythonwc_reduce.py';

Architecture

- Metastore: stores system catalog
- Driver: manages life cycle of HiveQL query, session handles and session statistics
- Query compiler: Compiles HiveQL into a distributed execution plan of map/reduce tasks
- Execution engines: The component executes the execution plan in proper dependency order; interacts with Hadoop
- HiveServer: provides JDBC/ODBC drivers and other interfaces for integrating other applications.
- Client API components: CLI, web interface, jdbc/odbc API
- Extensibility interface include SerDe (parse/print), User Defined Functions and User Defined Aggregate Functions.