

Cloud Futures Workshop Berkeley, California May 8, 2012

FP7-257993

CumuloNimbo: Parallel-Distributed Transactional Processing

Ricardo Jiménez-Peris, Marta Patiño, Iván Brondino – Univ. Politecnica de Madrid José Pereira, Rui Oliveira, R. Vilaça, F. Cruz [–] Minho Univ. Bettina Kemme, Yousuf Ahamad – McGill Univ.





- CumuloNimbo aims at solving the lack of scalability of transactional applications that represent a large fraction of existing applications.
- CumuloNimbo aims at conceiving, architecting and developing a ٠ transactional, coherent, elastic and ultra scalable Platform as a Service.
- Goals:
 - Ultra scalable and dependable -- able to scale from a few users to many millions of users while at the same time providing continuous availability;
 - Support transparent migration of multi-tier applications (e.g. Java EE applications, relational DB applications, etc.) to the cloud with automatic scalability and elasticity.
 - Avoid re-programming of applications and non-transparent scalability techniques such as sharding.
 - Support transactions for new data stores such as cloud data stores, graph databases, etc.





Challenges

- Main Challenges:
 - Update ultra-scalability (million update transactions per second and as many read-only transactions as needed).
 - Strong transactional consistency.
 - Non-intrusive elasticity.
 - Inexpensive high availability.
 - Low latency.
- CumuloNimbo goes beyond the State of the Art by scaling transparently transactional applications to very large rates without sharding, the current practice in Today's cloud.





Global Architecture

FP7-257993





Framework



- Guarantees transactional coherence across all tiers: application server, object cache and database.
- No constraints on applications, transactional processing and data, no required a priori knowledge.
- Fully transparent:
 - Syntactically: no changes required in the application.
 - Semantically: equivalent behavior to a centralized system.
- Can be integrated with any other infrastructure requiring transactional support (e.g. graph databases).





Ultra-Scalable Transactional Processing: Approach



- Decomposition of transactional processing.
 - No DB or transactional manager as a single component.
- Atomicity, consistency, isolation and durability are attained separately.
 - Each component scaled independently but in a composable manner.
 - The first bottleneck is in a component able to do million update transactions per second.
- Transactions are committed in parallel.
- Based on snapshot isolation:
 - Avoids read/write conflicts providing an isolation very close to serializability.
 - Serializability can be implemented on top of it, if needed.





Ultra-Scalable Transactional Processing: Snapshot Isolation

• Serializability provides a fully atomic view of a transaction, reads and writes happen atomically at a single point in time.



 Snapshot isolation splits atomicity in two points one at the beginning of the transaction where all reads happen and one at the end of the transaction where all writes happen.







Ultra-Scalable Transactional Processing: Components and Txn Life Cycle





SEVENTH FRAMEWORI

Ultra-Scalable Transactional Processing: Components and Txn Life Cycle





SEVENTH FRAMEWORK PROGRAMME

Ultra-Scalable Transactional Processing: Components and Txn Life Cycle





Ultra-Scalable Transactional Processing: Components and Txn Life Cycle



Commit Sequencer





Ultra-Scalable Transactional Processing: Snapshot Server

- The Snapshot server keeps track of the most recent snapshot that is consistent:
 - Its TS should such that there is no previous commit TS that is not yet durable and readable or it has been discarded.
 - That is, it keeps the longest prefix of used/discarded TSs such that there are no gaps.
- In this way transactions can commit in parallel and consistency preserved.







Ultra-Scalable Transactional Processing: Loggers

- Each logger takes care of a fraction of the log records.
- Loggers log in parallel and are uncoordinated.
- Loggers can be replicated.
- If this is the case the durability can be configured as:
 - To be in the memory of a majority of logger replicas (replicated memory durability).
 - To be in a persistent storage of a logger replica (1-safe durability).
 - To be in a persistent storage of a majority of logger replicas (n-safe durability).
- The client gets the commit reply after the writeset is durable (with respect the configured durability).



Ultra-Scalable Transactional Processing: Snapshot Server

FP7-257993

Sequence of timestamps received by the Snapshot Server



Evolution of the current snapshot at the Snapshot Server







Ultra-Scalable Transactional Processing: Increasing Efficiency

- The described approach so far is the original reactive approach.
- It results in multiple messages per update transaction.
- The adopted approach is proactive:
 - The local transaction managers report periodically about the number of committed update transactions per second.
 - The commit sequencer distributes batches of commit timestamps to the local transaction managers.
 - The snapshot server gets periodically batches of timestamps (both used and discarded) from local transaction managers.
 - The snapshot server reports periodically to local transaction managers the most current consistent snapshot.





Scalable Application Server and Distributed Object Cache

- We exploit JBoss and Hibernate as application server technology.
- We rely on their reflection capabilities (interceptors and hooks respectively) to intercept:
 - Transactional processing \rightarrow Becomes ultra-scalable.
 - Second level cache \rightarrow Becomes a distributed elastic cache.
- No changes required in the application server/persistency manager.
- The cache is multi-version aware guaranteeing full cache transparency.
- Approach applicable to any transactional application server either source code or with sufficient reflection capabilities.
- Support very large caches at both object and DB level enabling in-memory databases/application servers.





- SQL processing is performed at the SQL engine tier.
- A SQL engine instance:
 - Transforms SQL code into a query plan.
 - The query plan is optimized according the collected statistics (e.g. cardinality of keys).
 - Orchestrate the query plan execution on top of the distributed data store.
 - Returns the result of the SQL execution to the client.
 - Maintains updated the statistics in the data store.
- The SQL engine has been implemented by modifying Apache Derby, changing its transactional processing by CumuloNimbo's.





Scalable SQL Processing: Data Store

- To scale the data store, we leverage a key-value data store, Apache HBase.
- Relational tables are mapped to HBase tables.
- Secondary indexes are mapped to additional HBase tables that translate secondary keys into primary keys.
- Traffic between the query engine and HBase instances is minimized by:
 - Exploiting HBase filters to implement scan operators.
 - Reduces the cost of scans.
 - Leveraging HBase co-processors to compute local statistics on each region necessary.
 - Reduces the cost of statistics necessary for query optimization.





Efficient Deployment Collocation of Instances across Tiers

- One of the main goals is throughput efficiency, i.e., to attain a particular required throughput with the minimal number of resources.
- Both elasticity and dynamic load balancing contribute towards this goal.
- But another aspect is related on how to deploy the multiple instances of the multiple tiers to minimize the distribution overhead.
- Collocation of tiers has been considered and actually performed to diminish the number of distributed hops required to process a transaction.





Efficient Deployment Collocation of Instances across Tiers

FP7-257993



Ongoing work: Elasticity

- Elasticity is controlled at each layer with customized elastic rules.
 - For instance, the object cache can provision nodes either due to lack of memory or CPU saturation.
- Elasticity is combined with dynamic load balancing to guarantee that provisioning is only triggered when needed.
- Non-intrusive reconfiguration:
 - Focusing on maintaining throughput close to the peak one during reconfiguration.

Ongoing work: Fault Tolerance

- Replication is used for high availability and not for scaling.
 - Low cost data fault tolerance
 - Pushed down to the storage layer (distributed file system)
 - Outside the transaction response time path.
 - Fault tolerance for other components with a simple approach
 - Configuration and vital data stored on a replicated data store (Zookeeper).
 - Single replicated server keeps track of configuration metadata for all tiers and instances.
 - Fault tolerance of critical components:
 - Specialized replication that maximizes throughput and minimizes latency.
 - Commit server, snapshot server, loggers.

Evaluation Setup

- HBase+HDFS deployed on 5+1 dual-core nodes (12 cores).
- Distributed cache deployed on 5 dual-core nodes (10 cores).
- Transaction manager core components deployed on 2 dual-core nodes (4 cores).
- JBoss+Hibernate+Derby+HBase client deployed on 5 and 20 quad-core nodes (20 and 80 cores).
- Configuration manager deployed on a dual-core node (2 cores).
- Total cores: 28+20 to 80 (48 to 108 cores)

Scalability Results

FP7-257993

Linear scalability with 100+ cores Currently exercising 300+ cores

Contact Information

- Ricardo Jiménez-Peris
 - Technical Coordinator.
 - Univ. Politécnica de Madrid.
 - rjimenez@fi.upm.es
- http://cumulonimbo.eu

