VICTORIA UNIVERSITY OF WELLINGTON

Te Whare Wananga o te Upoko o te Ika a Maui



Trade-offs in Cloud Databases

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SWEN 432
Advanced Database Design and
Implementation

Plan for Trade-offs in Cloud Databases

- Features of relational SQL databases
- ACID properties
- NoSQL databases
- CAP Theorem
- BASE Properties
 - More about consistency/availability trade-offs

Readings: Have a look at Useful Links at the Course Home Page

Traditional Databases

- Most traditional databases are relational
- Relational data model has been developed to serve applications that insist on data consistency and reliability
- RDBMSs are OLTP DBMSs
- Data consistency and reliability are guaranteed by ACID properties
- ACID stands for Atomicity, Consistency, Isolation, and Durability

ACID Database Properties

Atomicity

- Transactions must act as a whole, or not at all
- No changes within a transaction can persist if any change within the transaction fails

Consistency

 Changes made by a transaction respect all database integrity constraints and the database remains in a consistent state at the end of a transaction

Isolation

- Transactions cannot see changes made by other concurrent transactions ,
- Concurrent transactions leave the database in the same state as if the transactions were executed serially

Durability

- Database data are persistent,
- Changes made by a transaction that completed successfully are stored permanently

Techniques to Achieve ACID Properties

Relational databases:

- Have a schema that provides a strict structure and integrity constraints for database data,
- Schemes are normalized to avoid data redundancy and update anomalies,
- Normalization is achieved by a vertical splitting of database tables,
- Queries frequently ask for joins.
- Integrity constraints (checked by writes) provide for consistency
- Locking (to avoid transaction anomalies, e.g. lost update) provides for isolation
- Logging provides for durability
 - But changes made by each transaction have to be written on disk two times

ACID Properties and Cloud Databases

- Many cloud applications (like social networks) require databases that:
 - Are highly scalable and available,
 - Have high throughput, and
 - Be network partition tolerant
- These requirements are in a direct collision with ACID properties, particularly with integrity constraint checking, locking, logging, and performing joins
- These applications also consider strict data structuring not being suitable
- Also, consistency is considered to be of some lesser importance

NoSQL Database Systems

- Many cloud databases are not relational
- They are termed NoSQL databases (Not only SQL)
 - NoSQL does not mean "NO to SQL", but NO to "one size fits all",
 - Different applications need different database paradigms
- Generally, NoSQL database systems:
 - Adopt the key-value data model for storing semi structured or unstructured data
 - Are either scheme less, or their schemes lack rigidity and integrity constraints of relational schemes
 - Use data partitioning, replication and distribution to several independent network nodes made of commodity machines for:
 - · Scalability,
 - Availability, and
 - Network partition tolerance
 - Are not normalized, and do not support joins but store all data needed to answer a query in the same partition to achieve a high throughput

CAP Theorem

- NoSQL database systems do not support ACID properties
 - Locking and logging would severely impair scalability, availability, and throughput
- Still, Consistency, Availability, and Network Partition Tolerance (CAP) represent their desirable properties
- But, according the CAP Theorem, also known as Brewer's Conjecture, any networked, shared data system can have at most two of these three desirable properties

CAP Theorem

*(*2*)*

- For a distributed database system as a service, the components of the CAP acronym stand for:
 - Consistency A service is considered to be consistent if after an update operation of some writer all readers see the update in some shared data source (all nodes containing replicas of a datum have the same value),
 - Availability A service is considered to have a high availability if it allows read and write operations a high proportion of time (even in the case of a node crush or some hardware or software parts are down due to upgrades)
 - Network partition tolerance is the ability of a service to perform expected operations in the presence of a network partition, unless the whole network crushes
 - A network partition occurs if two or more "islands" of network nodes cannot connect to each other
 - Dynamic addition and removal of nodes is also considered to be a network partition

Comments on CAP Properties

- Contrary to traditional databases, many cloud database are not expected to satisfy any integrity constraints
- High availability of a service is often characterized by small latency
 - Amazon claims that raise of 0.1 sec in their response time will cost them 1% in sales,
 - Google claims that just .5 sec in latency caused traffic to drop by 20%
- It is hard to draw a clear border between availability and network partitions
 - Network partitions may induce delays in operations of a service, and hence a greater latency

An Informal Proof of the CAP Theorem (1)

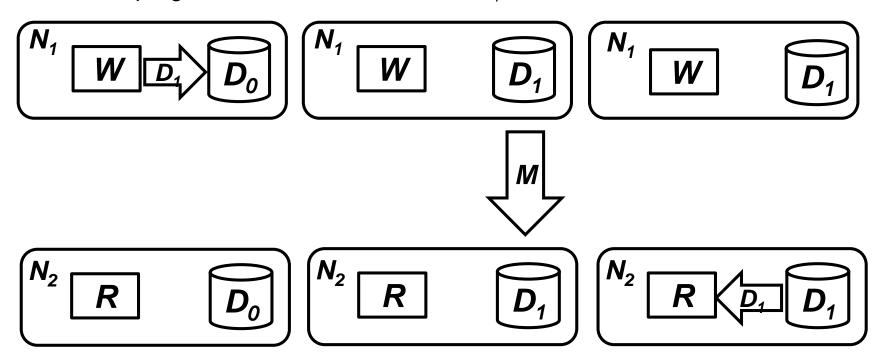
- Eric Brewer presented his conjecture in 2000, Gilbert&Lynch proved it formally in 2002
- Here is an informal presentation of their proof showing that a highly available and partition tolerant service may be inconsistent
- Let us consider two nodes N_1 and N_2
 - Both nodes contain the same copy of data D having a value D_0 ,
 - A program W (write) runs on N_1
 - A program R (read) runs on N₂





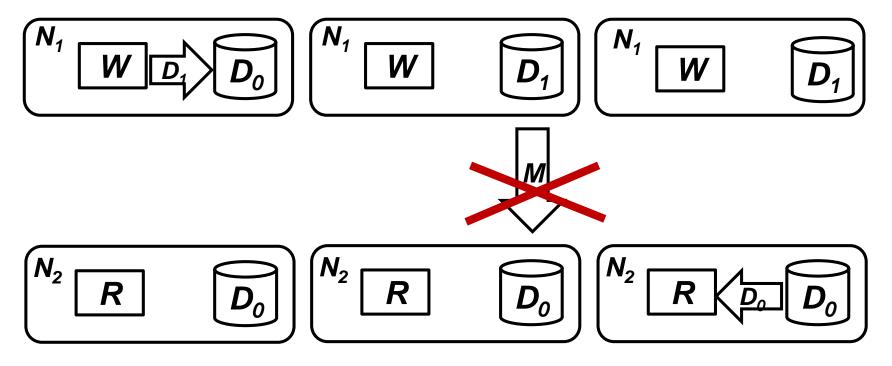
An Informal Proof of the CAP Theorem (2)

- In a regular situation:
 - The program W writes a new value of D, say D_1 on N_1
 - Then, N_1 sends a message M to N_2 to update the copy of D to D_1 ,
 - The program R reads D and returns D₁



An Informal Proof of the CAP Theorem (3)

 Assume now the network partitions and the message from N₁ to N₂ is not delivered



The program R reads D and returns D₀

Comments on the Informal Proof

(1)

- The service is highly available, since it performed both W and R without delay.
- The service is partition tolerant since it performed its tasks in spite the network partition.
- The service is inconsistent, since it delivered stale data.

Comments on the CAP Theorem

(2)

- If an application in a distributed and replicated network have to be:
 - Highly available (i.e. working with minimal latency), and
 - Tolerant to network partitions (performs expected user initiated operations), then
 - It has to perform an update even if not all nodes are able to acknowledge it
 - Hence the (strict) consistency can't be achieved
 - So, it may happen that some of N_i (i = 1, 2, ...) nodes have stale data, or even are unavailable
- The decision on which of {CA, CP, AP} property pairs to satisfy is made in early stages of a cloud application design, according to requirements and expectations defined for the application
- Most NoSQL database systems fall either in the AP class, or in the CP class

BASE Properties of Cloud DBs

(1)

- BASE (as defined by Eric Brewer) stands for:
 - BA (Basically Available) meaning that the system is always available, although there may be intervals of time when some of its components are not available
 - A basically available system answers each request in such a way that it appears that system works correctly and is always available and responsive
 - S (Soft State) means that the system has not to be in a consistent state after each transaction
 - The database state may change without user's intervention (due to an automatic replica convergence)
 - E (Eventually Consistent) guarantees that:
 - If no new updates are made to a given data item, accesses to each replica of that item will eventually return the last updated value
 - A system that has eventually achieved consistency is often said to have converged, or achieved replica convergence

BASE Properties of Cloud DBs

(2)

- The reconciliation of differences between multiple replicas requires exchanging and comparing versions of data between nodes
 - Usually "last writer wins"
 - The reconciliation is performed as a:
 - · read,
 - write, or
 - · asynchronous repair
 - Asynchronous repairs are performed in a background non transactional process
- Stale and approximate data in answers are tolerated

Consistency – Availability Trade-offs (1)

Assume:

- A cloud database is divided into p partitions and each partition is replicated on $n \ge 3$ servers (replica nodes),
- The database is unavailable if any of its parts is unavailable,
- Each node stores only one replica (to simplify analysis), and
- Each reader may contact any of n replica nodes in a partition

Consistency – Availability Trade-offs (2)

- Strict consistency: any read of a shared data item X returns the value stored by the most recent write on X
 - Requires n servers to be available in each partition
 - This is hard to achieve with commodity hardware,
 - If only one node fails, the database is unavailable
- Strong consistency requires w + r > n for each partition
 - w and r are numbers of nodes acknowledging a write and returning a read, respectively
 - At least 1 node in each partition can fail and the database is still available
 - A special case: **quorum** consistency, requires al least $q = \lfloor n/2 \rfloor + 1$ replica nodes to be available in each partition
 - For w = n and r = 1, strong consistency turns into the strict one
- **Eventual** consistency requires at least 1 node to be available in each partition in every moment
 - At least one node has to acknowledge a write,
 - At least one node has to return a read,
 - These two nodes do not have to be the same

Example: Consistency vs. Availability (1)

- Let the number of partitions p = 10 and the replication factor n = 3
 - So, the database is stored on 30 nodes
- Assume a client has requested the service to has strong consistency under quorum
- Question
 - How many nodes can fail and still the whole database to be available?
- Answer:
 - Since n = 3, quorum q = 2,
 - In the worst case, all unavailable nodes belong to the same partition
 - Then, to have the whole database available under strong consistency, only 1 node in total is allowed to be unavailable for both writing and reading
 - In the best case, all partitions
 - Then, to have the visit of a available under strong consistency, one node in each consistency to be unavailable for both writing and reading, allowing at most 10 hodes being simultaneously unavailable in total

Example: Consistency vs. Availability (2)

- Let the number of partitions p = 10 and the replication factor n = 3
 - So, the database is stored on 30 nodes
- Assume a client has requested the service to has eventual consistency
- Question
 - How many nodes can fail and still the whole database to be available?
- Answer:
 - In the worst case, all unavailable nodes belong to the same partition
 - Then, to have the whole database available under eventual consistency, only 2 nodes in total are allowed to be unavailable
 - In the best case, a pavailable nodes are evenly distributed over all partitions
 - Then, to have the consistency, to the consistency, the consistency of the consistency o

Summary

(1)

- NoSQL cloud databases are scalable, highly available and tolerate network partitions
- Cloud database system can have only two out of Consistency, Availability, and Network Partition Tolerance
- BASE represents a "best effort" approach to ensuring database reliability
 - BASE forfeits ACID properties, also by its name (base versus acid, as found in chemistry)
 - Basically Available,
 - Soft state,
 - Eventually consistent

Summary

(2)

- Eventual Consistency is a model used in distributed systems that guarantees that all accesses to replicas of an updated item will eventually return the last updated value
- Consistency levels:
 - Strict consistency,
 - Strong consistency,
 - Eventual consistency
- Consistency and availability mutually restrict each other:
 - As greater the consistency requirement as lower the database availability