

Distributed Data Management



Introduction to data replication and the CAP theorem

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Replication

Same data in different places



Replication

•Same data

portions of the whole dataset (chunks)

•in different places

local and/or remote servers, clusters, data centers

•Goals

• Redundancy helps surviving failures (availability)

Better performance

Approaches

Master-Slave replication

A-Synchronous replication

Master-Slave replication

Master-Slave

- A master server takes all the writes, updates, inserts
- One or more Slave servers take all the reads (they can't write)
- Only read scalability
- The master is a single point of **failure**
- •Some NoSQLs (e.g., CouchDB) support Master-Master replica



Synchronous replication

o Before committing a transaction, the Master waits for (all) the Slaves to commit

- Similar in concept to the **2-Phase Commit** in relational databases
- Performance killer, in particular for replication in the cloud
- Trade-off: wait for a subset of Slaves to commit, e.g., the **majority** of them



Asynchronous replication

• The Master commits **locally**, it does not wait for any Slave

Each Slave independently fetches updates from Master, which may fail...

- IF no Slave has replicated, then you've lost the data committed to the Master
- IF some Slaves have replicated and some haven't, then you have to reconcile

Faster and unreliable



Distributed databases

Different autonomous machines, working together to manage the same dataset



Key features of distributed databases

- •There are 3 typical problems in distributed databases: •**Consistency**
 - All the distributed databases provide the same data to the application
 - Availability
 - Database failures (e.g., master node) do not prevent survivors from continuing to operate
 - Partition tolerance
 - The system continues to operate despite arbitrary message loss, when connectivity failures cause network partitions

CAP Theorem

•The CAP theorem, also known as Brewer's theorem, states that it is **impossible** for a distributed system to **simultaneously** provide **all three** of the previous guarantees

•The theorem began as a **conjecture** made by University of California in 1999-2000

 Armando Fox and Eric Brewer, "Harvest, Yield and Scalable Tolerant Systems", Proc. 7th Workshop Hot Topics in Operating Systems (HotOS 99), IEEE CS, 1999, pg. 174-178.

•In 2002 a formal proof was published, establishing it as a **theorem**

 Seth Gilbert and Nancy Lynch, "Brewer's conjecture and the feasibility of consistent, available, partition-tolerant web services", ACM SIGACT News, Volume 33 Issue 2 (2002), pg. 51-59

•In 2012, a follow-up by Eric Brewer, "CAP twelve years later: How the "rules" have changed"

o IEEE Explore, Volume 45, Issue 2 (2012), pg. 23-29.



http://guide.couchdb.org/editions/1/en/consistency.html#figure/1

CAP Theorem

•The easiest way to understand CAP is to think of **two nodes** on opposite sides of a **partition**.

•Allowing at least one node to update state will cause the nodes to become **inconsistent**, thus forfeiting C.

•If the choice is to preserve consistency, one side of the partition must act as if it is **unavailable**, thus forfeiting A.

•Only when no network **partition** exists, is it possible to preserve both consistency and availability, thereby forfeiting P.

•The general belief is that for wide-area systems, **designers cannot forfeit P** and therefore have a difficult choice between C and A.



http://www.infoq.com/articles/cap-twelve-years-later-how-the-rules-have-changed

CAPTheorem



http://blog.flux7.com/blogs/nosql/cap-theorem-why-does-it-matter

CA without P (local consistency)

• **Partitioning** (communication breakdown) causes a failure.

•We can still have **Consistency** and **Availability** of the data shared by agents **within each Partition**, by ignoring other partitions.

Local rather than global consistency / availability

•Local consistency for a partial system, 100% availability for the partial system, and no partitioning does not exclude several partitions from existing with their own "internal" CA.

•So partitioning means having **multiple independent systems** with 100% CA that do not need to interact.

CP without A (transaction locking)

•A system is allowed to *not* answer requests at all (turn off "A").

- •We claim to tolerate **partitioning/faults**, because we simply block all responses if a partition occurs, assuming that we cannot continue to function correctly without the data on the other side of a partition.
- •Once the partition is healed and **consistency** can once again be verified, we can restore availability and leave this mode.
- •In this configuration there are global consistency, and global correct behaviour in partitioning is to **block access to replica sets** that are not in synch.
- •In order to tolerate P at any time, we must sacrifice A at any time for **global consistency**.
- •This is basically the **transaction lock**.

AP without C (best effort)

•If we don't care about **global consistency** (i.e. simultaneity), then every part of the system can make available what it knows.

- •Each part might be able to answer someone, even though the system as a whole has been broken up into incommunicable regions (**partitions**).
- •In this configuration "without consistency" means without the assurance of **global** consistency **at all times**.

"Each node in a system should be able to make decisions purely based on **local state**. If you need to do something under high load with **failures** occurring and you need to reach agreement, you're lost. If you're concerned about **scalability**, any algorithm that forces you to run agreement will eventually become your **bottleneck**. Take that as a given."

Werner Vogels, Amazon CTO and Vice President

Beyond CAP

•The "2 of 3" view is misleading on several fronts.

- •First, because **partitions** are rare, there is little reason to forfeit C or A when the system is not partitioned.
- •Second, the **choice between C and A** can occur many times within the same system at very fine granularity; not only can subsystems make different choices, but the choice can change according to the operation or even the specific data or user involved.

• Finally, all three properties are more continuous than binary.

- Availability is obviously continuous from 0 to 100 percent
- There are also many levels of consistency
- Even partitions have nuances, including disagreement within the system about whether a partition exists

How the rules have changed

- Any networked shared-data system can have only 2 of 3 desirable properties at the same time
- •Explicitly handling partitions, designers can optimize consistency and availability, thereby achieving some **trade-off of all three**
- •CAP prohibits only a tiny part of the design space:
 - perfect availability (A) and consistency (C)
 - o in the presence of partitions (P), which are **rare**
- •Although designers need to choose between consistency and availability when partitions are present, there is an incredible range of **flexibility for handling partitions** and recovering from them
- •Modern CAP goal should be to maximize combinations of consistency (C) and availability (A) that make sense for the **specific application**

ACID

•The four ACID properties are:

- Atomicity (A) All systems benefit from atomic operations, the database transaction must completely succeed or fail, partial success is not allowed
- Consistency (C) During the database transaction, the database progresses from a valid state to another. In ACID, the C means that a transaction pre-serves all the database rules, such as unique keys. In contrast, the C in CAP refers only to single copy consistency.
- Isolation (I) Isolation is at the core of the CAP theorem: if the system requires ACID isolation, it can operate on at most one side during a partition, because a client's transaction must be isolated from other client's transaction
- **Durability (D)** The results of applying a transaction are permanent, it must persist after the transaction completes, even in the presence of failures.

BASE

- •Basically Available: the system provides availability, in terms of the CAP theorem
- •**Soft state:** indicates that the state of the system may change over time, even without input, because of the eventual consistency model.
- •Eventual consistency: indicates that the system will become consistent over time, given that the system doesn't receive input during that time
- •Example: DNS Domain Name Servers
 - DNS is not multi-master

ACID versus BASE

- •ACID and BASE represent two design philosophies at opposite ends of the consistency-availability spectrum
- •ACID properties focus on **consistency** and are the traditional approach of databases
- •BASE properties focus on high **availability** and to make explicit both the choice and the spectrum
- •BASE: Basically Available, Soft state, Eventually consistent, work well in the presence of **partitions** and thus promote **availability**

Conflict detection and resolution

An example from a notable NoSQL database



- •There are two customers, **A** and **B**
- •A books a hotel room, the last available room
- •B does the same, on a different node of the system, which was **not consistent**



- •The hotel room document is affected by two **conflicting updates**
- •Applications should solve the conflict with custom logic (it's a business decision)
- •The database can
 - **Detect** the conflict
 - Provide a local solution, e.g., latest version is saved as the winning version



Conflict

 CouchDB guarantees that each instance that sees the same conflict comes up with the same winning and losing revisions.

- •It does so by running a **deterministic algorithm** to pick the winner.
 - The revision with the longest revision history list becomes the winning revision.
 - If they are the same, the _rev values are compared in ASCII sort order, and the highest wins.