

Having Your Cake, And Eating It Too

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An intro to the history of OLTP DBs, NoSQL databases, and an overview of a new breed of DBs that promise the best of both worlds ¹

¹ Mostly. Except when it doesn't. There is no free lunc.. err .. cake. Free lunch is a lie.

History of (OLTP) DBs

- Online Transaction Processing
- Mostly single machine
- Scaled vertically
- Traditionally RDBMs
- Typically fulfill ACID properties

ACID?



Atomicity

Transactions succeed completely or fail completely

Consistency

Operations must bring the DB to a *valid* state respecting DB constraints, and future reads

Isolation

No dirty reads, non-repeatable reads, or phantom reads

Durability

Commits stay committed, through network/hardware failure

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- modern apps have missive writes
- many apps have more relaxed consistency needs

How do we solve these problems?

Put another way, what tradeoffs can we make?

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 - function with network failures or delays between nodes

Clarifiying a common misconception

CAP says during a partition, you must choose between availability and consistency.

However, if there is no networking failure, you can have both consistency and availability.

NoSQL

- We always need parition tolerance in a distributed system
- In order to scale, pick Availability over Consistency
- Accept 'Eventual Consistency'
- Many types, but of the BigTable/Dynamo family:
 - BigTable
 - DynamoDB
 - Cassandra
 - Riak

Problems with SQL/NoSQL systems

- RDBMs give us strong consistency, but poor performance
 - Due to 2PL, writes are blocked while strong reads are happening
- NoSQL DBs give us great performance, but eventual consistency
 - You can get stale reads, because all nodes may not have converged to the latest commit

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- ensure resources that locks are respected (ie, if transaction A has not yet released, stop transaction B from accessing the resource)
- make guarantees about consistency between nodes, regions, and even continents

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Linearizability

The ability to absoutely order events, across nodes, in order to guarantee **Consistency** of data across nodes. Atomic/GPS clocks enable systems to agree on time without having to consult a single source of truth.

Serializability

Transactions should behave as if they had a lock over the data they are reading/writing. Easy in a non-distributed DB, but we need atomic clocks for a distributed environemnt. Provides **Isolation**.

==> External Consistency

With these two properties, you have external consistency. You won't suffer from stale/dirty reads, can read data from one node that was committed in another, and events won't appear out of order.

Google Spanner

- CP
- C(A)P
 - A is effectively more than 9 9s
 - mostly due to config/user errors, only 7.6% due to network (ie, partition) reasons
- serializabity from lock
- external consistency (linearizability) from TrueTime

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 - Paxos groups to achieve consensus on updates

ê Questions?

Appendix

Spanner, TrueTime & The CAP Theorem

NoSQL (Wikipedia)

NewSQL (Wikipedia)

ACID (Wikipedia)>)

CAP Theorem

Cloud Spanner: TrueTime and external consistency