Spanner: Google's Globally-Distributed Database

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Outline

- Motivation
- 2 Implementation
- TrueTime
- 4 Concurrency Control
- 5 Evaluation
- 6 Conclusions & Future Work

Motivation

- Distributed data at global scale
- Externally-consistent distributed transactions
- High availability
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- Why not use Megastore
 - Poor write throughput

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- Versioned data is stored in schematized semi-relational tables
- Each version is automatically timestamped with its commit time
- Garbage collection & reads at old timestamps
- General-purpose transactions
- SQL-like query language

Fundamentals

- Transaction i is denoted T_i
- Externally-consistent:

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- Two-Phase Locking (2PL): Guarantees serializability
 - 1 Expanding phase: locks are aquired, no locks are released
 - 2 Shrinking phase: locks are released, no locks are aquired
- Two-Phase Commit (2PC):

Coordinates processes participating in atomic distributed transaction

- Ommit-Request phase: Request "Yes" (Commit) or "No" (Abort) from every transaction process
- ② Commit phase: Commit transaction if all voted "Yes", otherwise abort

Features

As a globally-distributed database, Spanner provides interesting features:

- Dynamically controlled replication configurations at fine grain
- Externally consistent reads & writes
- globally-consistent reads at a timestamp
- ⇒ enables atomic schema changes in presence of ongoing transactions

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- Externally consistent reads & writes
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- ⇒ enables atomic schema changes in presence of ongoing transactions
 - Timestamps reflect serialization order (linearizability)
 - Key enabler: novel TrueTime API exposing clock uncertainty

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- universe master: console to display status information about zones
- placement driver:
 - handles automated data movement across zones
 - moves data due to updated replication constraints or load balancing

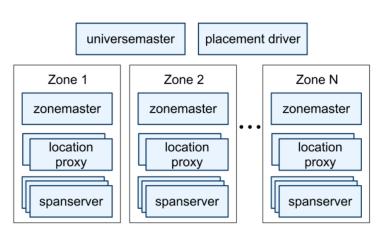


Figure 1: Spanner server organization

Spanserver Software Stack

- Responsible for 100 1000 tablets
- A tablet is a bag of key-value mappings
 (key:string, timestamp:int64) → string
- Timestamps are assigned to data ⇒ multi-version database
- Tablet states are stored on Colossus
- Single Paxos state machine on top of each tablet \Rightarrow **Replication**
- Set of replicas is called Paxos group

Spanserver Software Stack

- Every replica which is a leader implements
 - ullet a **lock table** for concurrency control: (key range) ightarrow lock state
 - a transaction manager (TM) for distributed transactions
- Transaction manager used to implement participant leader
- If multiple Paxos groups are involved: Two-Phase Commit
 - One participant group is chosen as coordinator
 - Participant leader of this group: coordinator leader
- State of each transaction manager is stored in Paxos group

Spanserver Software Stack

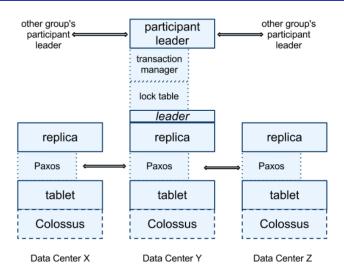


Figure 2 : Spanserver software stack

Directories:

- Bucketing abstraction on top of key-value mappings
- Contain set of contiguous keys sharing common prefix
- Control locality of data
- Unit of data placement (same replication configuration)
- Smallest unit to specify geographic-replication properties (placement)

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 co-locate directories frequently accessed together
- Placements can be controlled in two dimensions:
 - Number & types of replicas
 - @ Geographic placement of replicas
- Movedir: background task to move directories

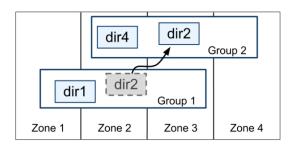


Figure 3: Directories (the unit of data movement)

Data Model

- Features:
 - Schematized semi-relational tables
 - SQL-like query language
 - General-purpose transactions
- Layered on top of directory-bucketed key-value mappings
- Applications create databases in universe
- Database may contain unlimited number of schematized tables
- Table contains rows (must have names), columns & versioned values

Data Model

- Databases must be partitioned in one or more hierarchies
- Directory table: table at the top
- Directory contains ordered rows starting at a directory table key
- Interleaving tables to form directories is significant
- Locality relationships between multiple tables ⇒ good performance

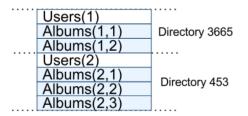


Figure 4: Interleaving example

TrueTime

Method	Returns
TT.now()	TTinterval: [earliest, latest]
TT.after(t)	true if t has definitely passed
TT.before(t)	true if t has definitely not arrived

Table 1 : TrueTime API (t is of type TTstamp)

Time is represented as a TTinterval (interval with bounded uncertainty)

Let e be an event and $t_{abs}(e)$ denote the absolute time of e. For tt = TT.now(), the following property holds:

$$tt.earliest \leq t_{abs}(e) \leq tt.latest$$

TrueTime

- Time references: GPS and atomic clocks
- Set of time master machines per datacenter
- Armageddon masters: masters using atomic clocks
- Timeslave daemon on each machine
- Masters compare their time references regularly
- Daemons poll a variety of masters (liar detection)

TrueTime



Figure 5 : Visualization of TT.now() and ϵ

Concurrency Control

- TrueTime is used to guarantee correctness properties
- Properties are used to implement
 - Externally-consistent transactions
 - Lock-free read-only transactions
 - Non-blocking reads in the past

Timestamp Management

Operation	CC	Replica required
Read-Write transaction	Pessimistic	leader
Read-Only transaction	Lock-free	leader; any
Snapshot Read, client-provided timestamp	Lock-free	any
Snapshot Read, client-provided bound	Lock-free	any

Table 1 : Supported types of reads & writes

- Standalone write ⇒ RW transaction
- Non-snapshot standalone read ⇒ RO transaction
- RO transactions
 - must predeclare to not include writes
 - can proceed on any sufficiently up-to-date replica (also snapshot reads)

Paxos Leader Leases

- Timed leases for long-lived leadership (10s by default)
- Quorum of lease votes ⇒ leader has a lease
- Successful writes extend lease vote on replica
- Disjointness invariant:

For each Paxos group, each Paxos leader's *lease interval* is disjoint from every other leader's

• Leaders must not abdicate before $TT.after(s_{max})$, where s_{max} denotes the maximum timestamp used by a leader

Timestamps in RW Transactions

- Timestamp assignment only in between the two phases (2PL)
- Monotonicity variant:
 Within each Paxos group, Spanner assigns timestamps to Paxos writes in monotonically increasing order, even across leaders
- Enforced across leaders by disjointness invariant:
 Leader must only assign timestamps within the interval of leader lease
- Timestamp s assigned $\Rightarrow s_{max} = s$
- External consistency invariant:

$$t_{abs}\left(e_{1}^{commit}
ight) < t_{abs}\left(e_{2}^{start}
ight) \Rightarrow s_{1} < s_{2}$$



Timestamps in RW Transactions

- Let e_i^{server} denote the arrival event of commit request for T_i
- Two rules guarantee the external consistency invariant:
 - Start

 Coordinator leader assigns commit timestamp $s_i \geq TT.now().latest$ after e_i^{server}
 - Commit Wait Coordinator leader ensures clients cannot see data commited by T_i until TT.after(s_i) is true

Serving Reads at a Timestamp

- Monotonicity invariant allows to determine if a replica's state is sufficiently up-to-date to satisfy a read
- Safe time t_{safe} at each replica
- Replica satisfies read if $t \le t_{safe}$
- $t_{safe} = min(t_{safe}^{Paxos}, t_{safe}^{TM})$

Timestamps in RO Transactions

- Two phases:
 - Assign timestamp s_{read}
 - ② Execute transaction's reads as snapshot reads at s_{read}
- Spanner assigns the oldest timestamp that preserves external consistency to reduce the changes of blocking

Read-Write Transactions

- RW transactions buffered at client until Commit
- Reads do not see effects of transaction's writes
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 - 2 Logs prepare record through Paxos
 - Notifies coordinator of its prepare timestamp
- Coordinator leader
 - Chooses timestamp for whole transaction
 - 2 Commit timestamp s must be
 - \geq all prepare timestamps
 - > TT.now().latest at the time the Commits were received
 - > any previously assigned timestamp
 - Some Logs Commit record through Paxos



Acquire all locks

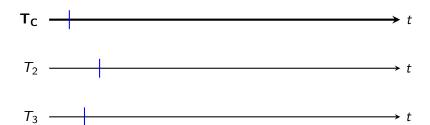


Figure 6: RW transaction illustration

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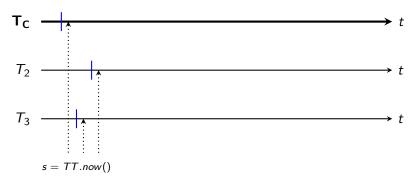


Figure 6: RW transaction illustration

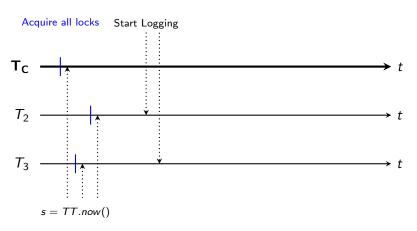


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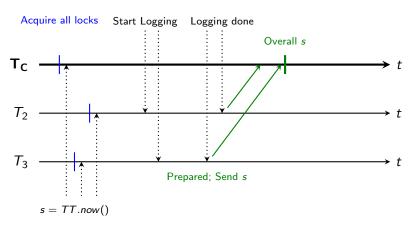


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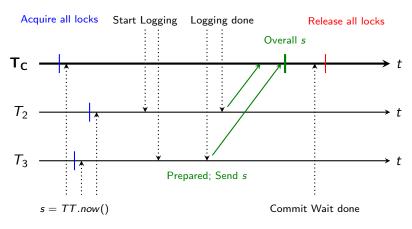


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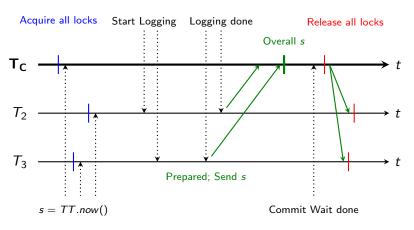


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Read-Only Transactions

- scope expression summarizing the keys that will be read
- Served by single Paxos group
 - ⇒ client issues transaction to group leader
- Served by multiple Paxos groups
 - \Rightarrow client executes reads at $s_{read} = TT.now().latest$
- All reads can be sent to replicas that are sufficiently up-to-date

Schema-Change Transactions

- TrueTime enables this feature
- Use of standard transaction infeasible
- Non-blocking variant of standard transaction
 - Explicit assignment of a future timestamp (Prepare phase)
 - Reads & writes synchronize with any registered schema-change timestamp at time t
- Without TrueTime \Rightarrow schema change at t would be meaningless

Microbenchmarks

- Setup:
 - 4GB RAM scheduling units per spanserver
 - 4 cores (AMD Barcelona 2200MHz) per spanserver
 - One spanserver per zone
 - Clients & zones in set of datacenters with network distance < 1ms
 - Database created with 50 Paxos groups with 2500 directories
- Operations: standalone reads & writes of 4KB
- One unmeasured round of reads to warm any location caches

Microbenchmarks

Latency [ms] (less is better)

Replicas	Write	RO Transaction	Snapshot Read
1D	9.4 ± 0.6	-	-
1	14.4 ± 1.0	1.4 ± 0.1	1.3 ± 0.1
3	13.9 ± 0.6	1.3 ± 0.1	1.2 ± 0.1
5	14.4 ± 0.4	1.4 ± 0.05	1.3 ± 0.04

Table 2: Latency experiments

- Sufficiently few operations to avoid queuing
- Increasing number of replicas:
 - Latency stays roughly constant with less standard deviation
 - Latency to achieve quorum less sensitive to stragglers

Microbenchmarks

Throughput [Kops/sec] (more is better)

Replicas	Write	RO Transaction	Snapshot Read
1D	4.0 ± 0.3	-	-
1	4.1 ± 0.05	10.9 ± 0.4	13.5 ± 0.1
3	2.2 ± 0.5	13.8 ± 3.2	38.5 ± 0.3
5	2.8 ± 0.3	25.3 ± 5.2	50.0 ± 1.1

Table 2 : Throughput experiments

- Sufficiently many operations to saturate servers' CPUs
- Increasing number of replicas:
 - Snapshot read throughput increases almost linear

Availability

• Setup:

- Universe with 5 zones Z_i , each of which has 25 spanservers
- Database sharded into 1250 Paxos groups
- 100 clients constantly issued snapshot reads (50K reads/sec)
- All leaders explicitly place in Z_1
- 5s into each test, all servers in one zone were killed

Availability

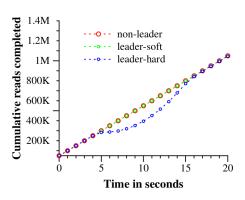


Figure 7: Effect of killing servers on throughput

• Availability benefits of running Spanner in multiple datacenters

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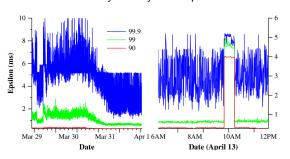


Figure 8 : Distribution of TrueTime ϵ values

Conclusions & Future Work

- 5 years from Spanner's inception to iterate to the current design
- Combines & extends on ideas from database & systems community
- Linchpin: TrueTime
- Build distributed systems with much stronger time semantics
- No longer depend on loosely synchronized clocks & weak time APIs

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- Linchpin: TrueTime
- Build distributed systems with much stronger time semantics
- No longer depend on loosely synchronized clocks & weak time APIs
- Future Work:
 - Implement optimistical reads in parallel
 - Support direct changes of Paxos configurations
 - Improve single-node performance
 - Support movement of processes between datacenters

Thank you for your attention!

Questions?