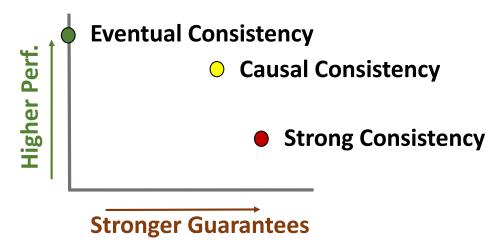
I Can't Believe It's Not Causal! Scalable Causal Consistency with No Slowdown Cascades

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¹UT Austin, ²Facebook, ³USC, ⁴Cornell University

Causal Consistency: Great In Theory



- Lots of exciting research building scalable causal data-stores, e.g.,
 - > COPS [SOSP 11]
- Eiger [NSDI 13]
 Cure [ICDCS 16]

- Bolt-On [SIGMOD 13] > Orbe [SOCC 13] > TARDIS [SIGMOD 16]
- Chain Reaction [EuroSys 13] ➤ GentleRain [SOCC 14]

Causal Consistency: But In Practice ...

The middle child of consistency models

Reality: Largest web apps use eventual consistency, e.g.,

Espresso



TAO



Manhattan



Key Hurdle: Slowdown Cascades







Implicit Assumption of Current Causal Systems

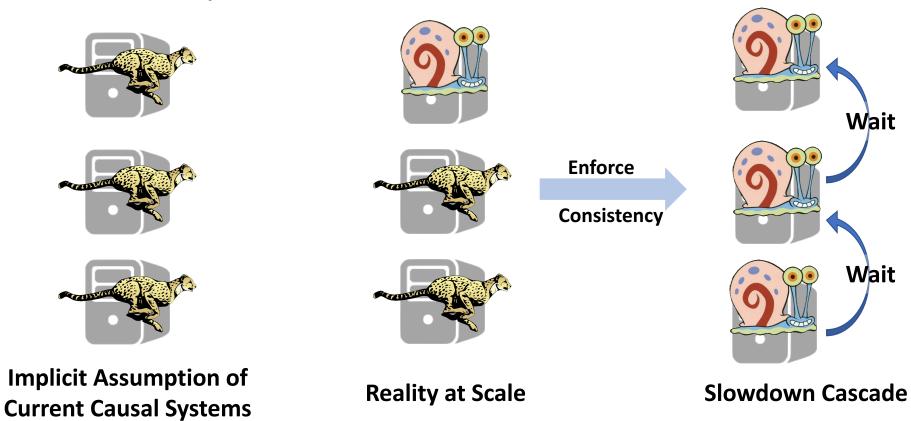






Reality at Scale

Key Hurdle: Slowdown Cascades





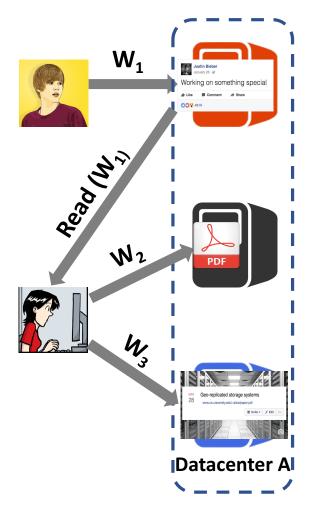








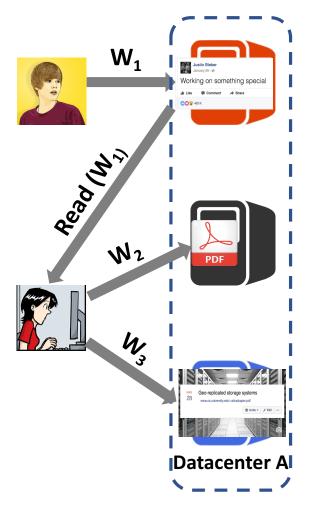
Replicated and sharded storage for a social network

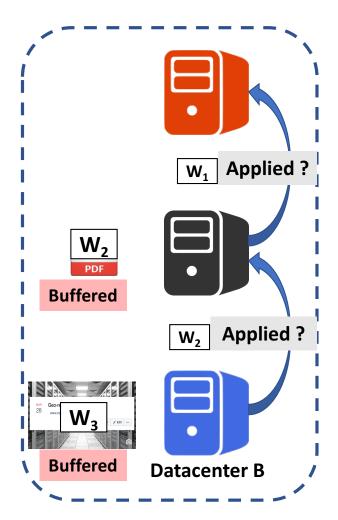






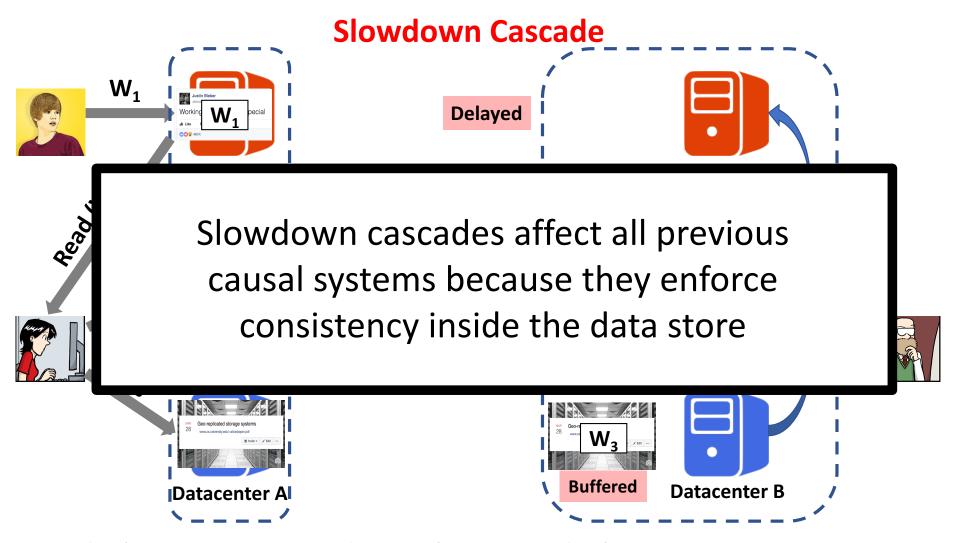
Writes causally ordered as $W_1 \rightarrow W_2 \rightarrow W_3$





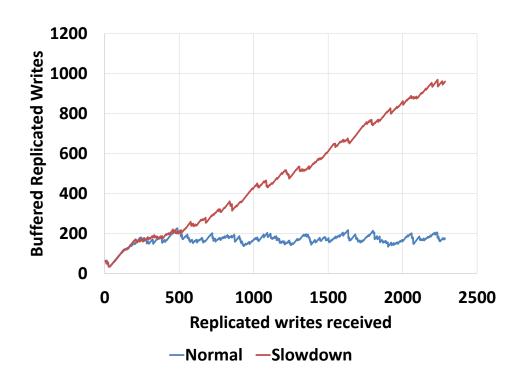


Current causal systems enforce consistency as a datastore invariant



Alice's advisor unnecessarily waits for Justin Bieber's update despite not reading it

Slowdown Cascades in Eiger (NSDI '13)



Replicated write buffers grow arbitrarily because Eiger enforces consistency inside the datastore

OCCULT

Observable Causal Consistency Using Lossy Timestamps

Observable Causal Consistency

Causal Consistency guarantees that each *client observes* a monotonically non-decreasing set of updates (including its own) in an order that respects potential causality between operations

Key Idea:

Don't implement a causally consistent data store Let clients *observe* a causally consistent data store

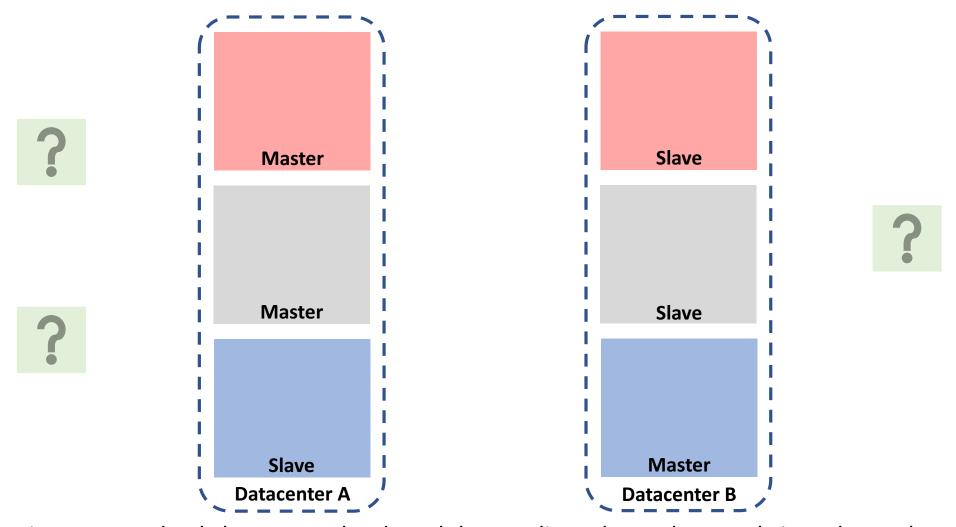


How do clients *observe* a causally consistent datastore ?

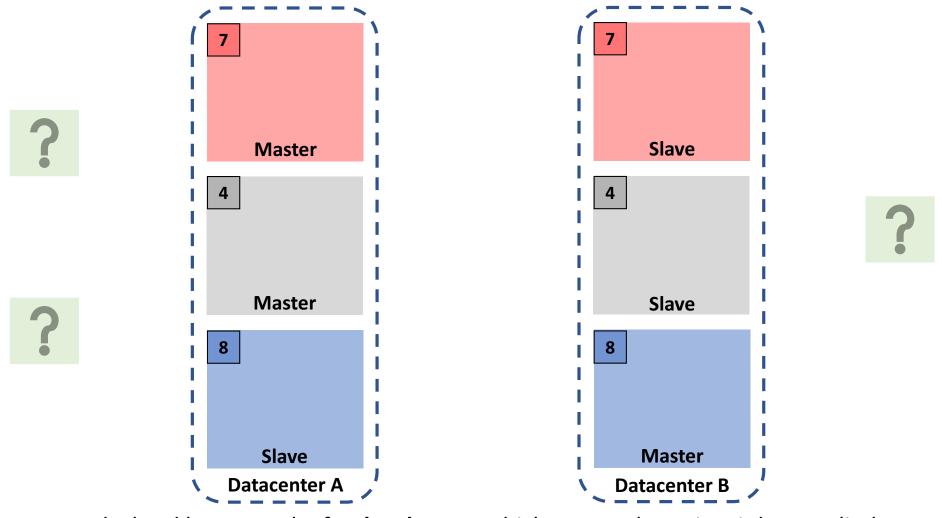




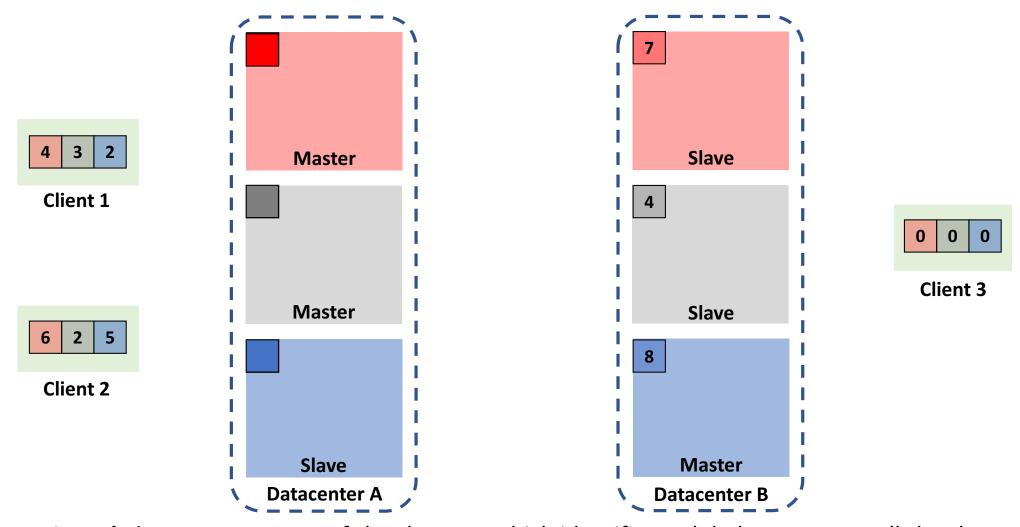
Datacenter Al



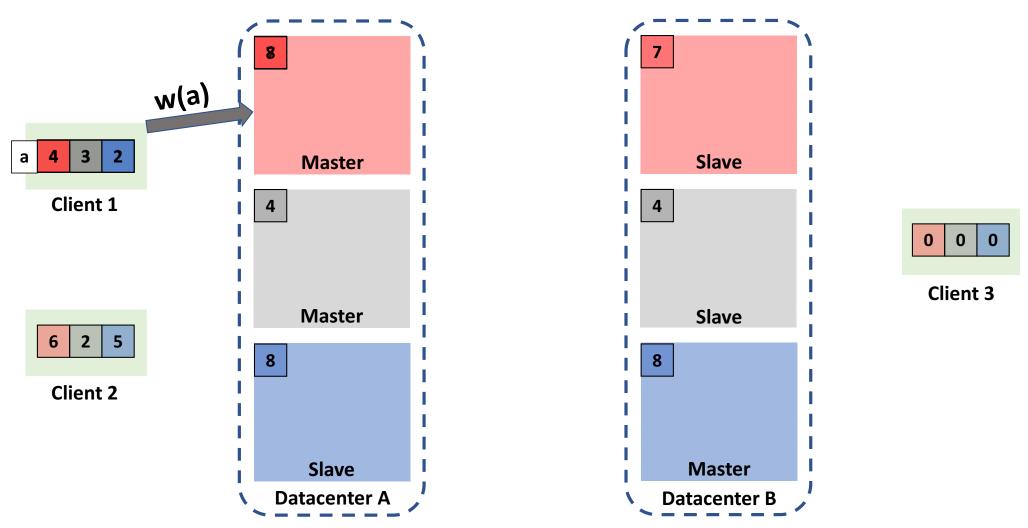
Writes accepted only by master shards and then replicated asynchronously in-order to slaves



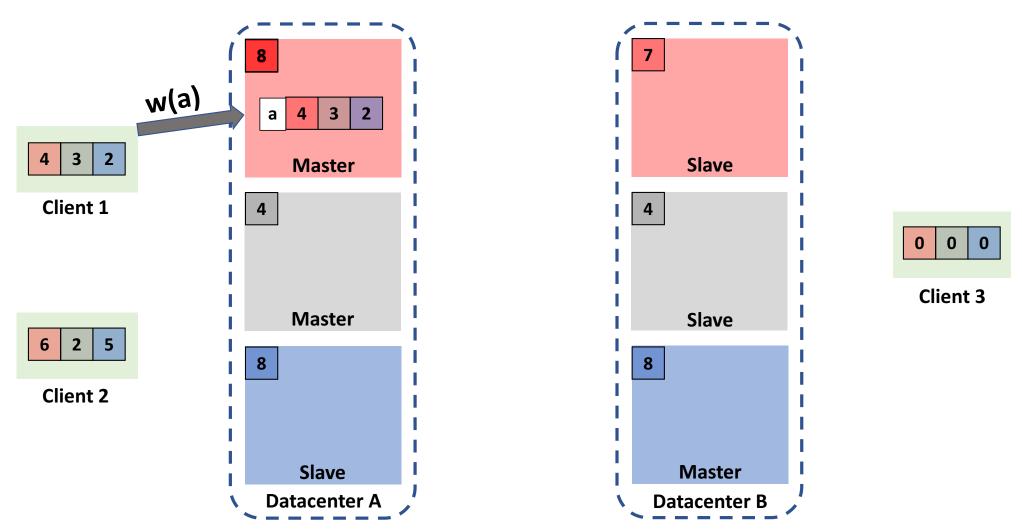
Each shard keeps track of a **shardstamp** which counts the writes it has applied



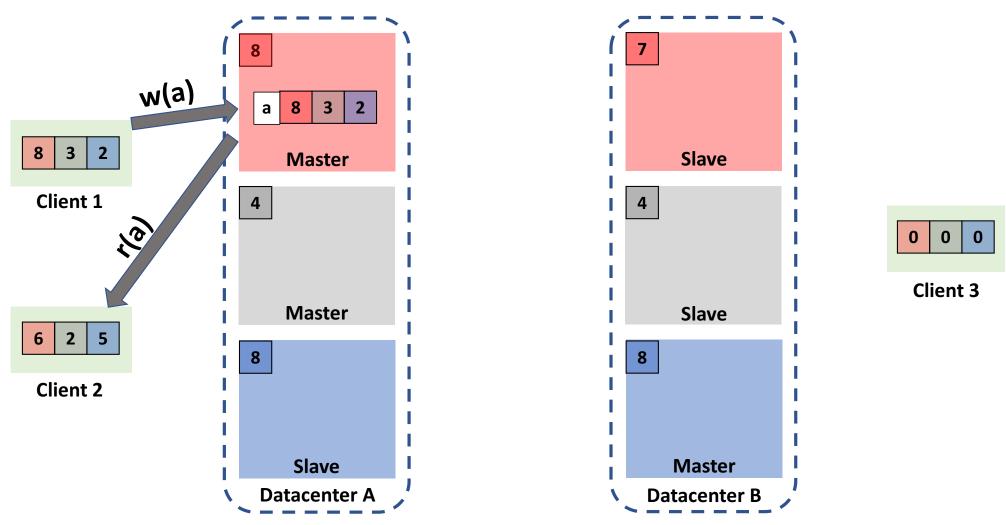
Causal Timestamp: Vector of shardstamps which identifies a global state across all shards



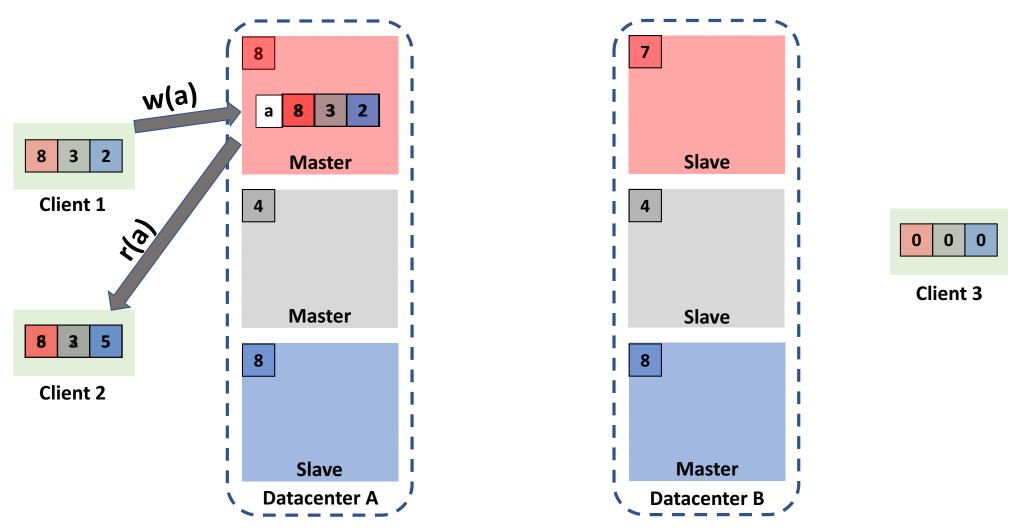
Write Protocol: Causal timestamps stored with objects to propagate dependencies



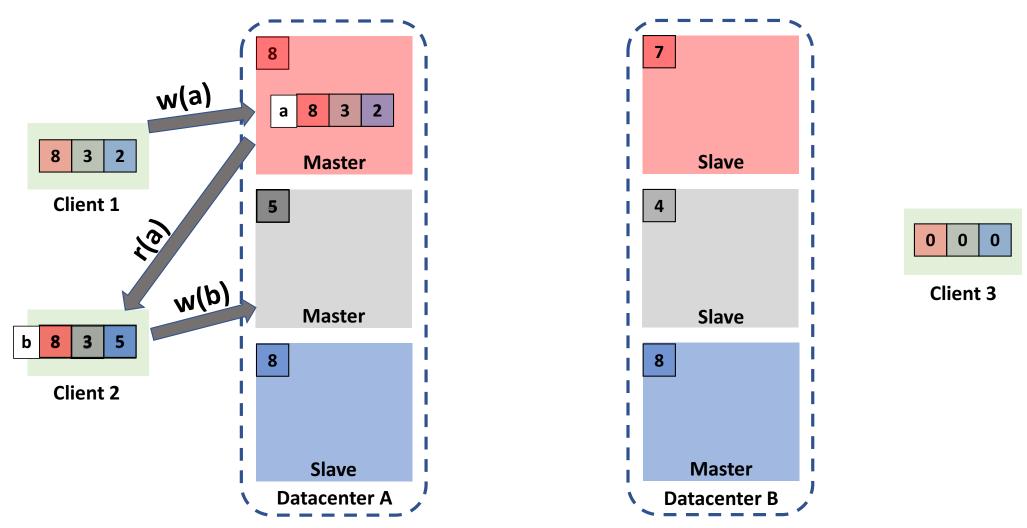
Write Protocol: Server shardstamp is incremented and merged into causal timestamps



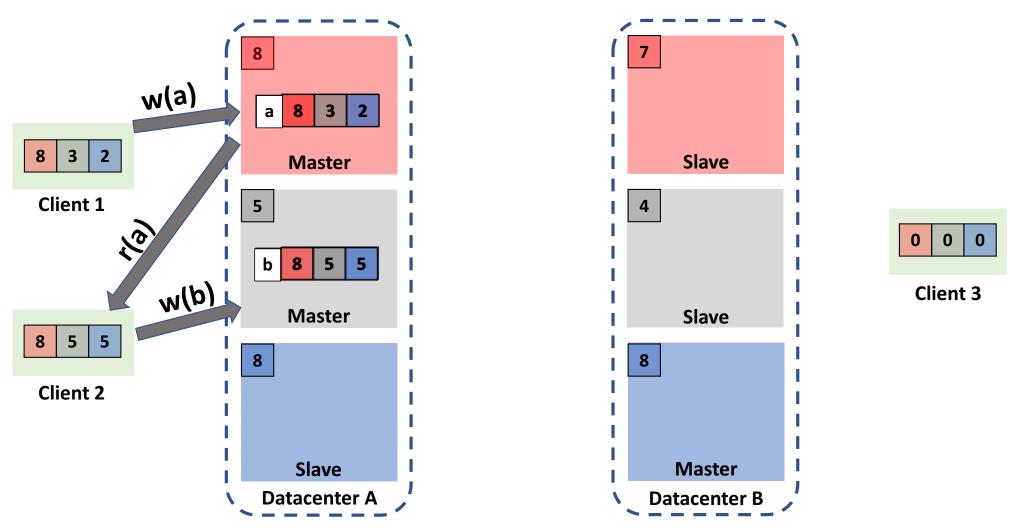
Read Protocol: Always safe to read from master



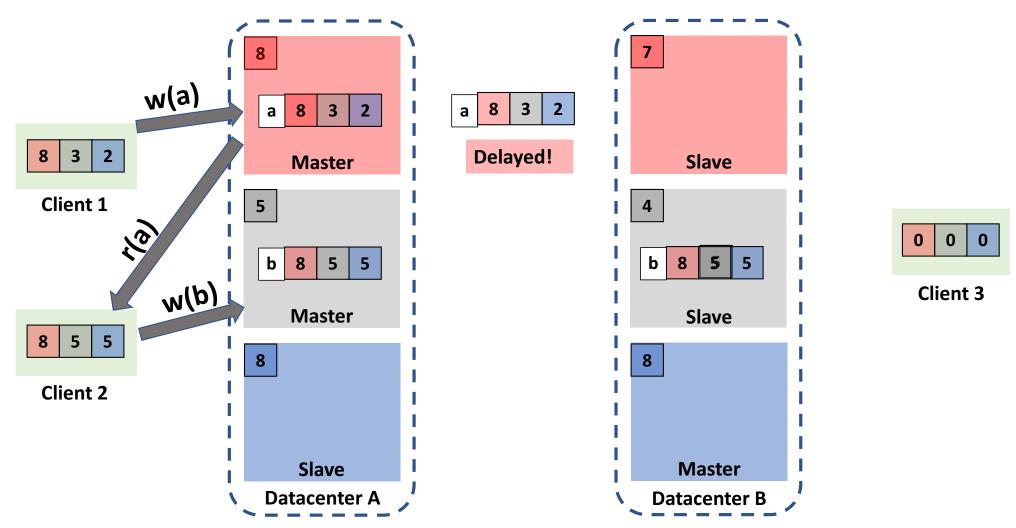
Read Protocol: Object's causal timestamp merged into client's causal timestamp



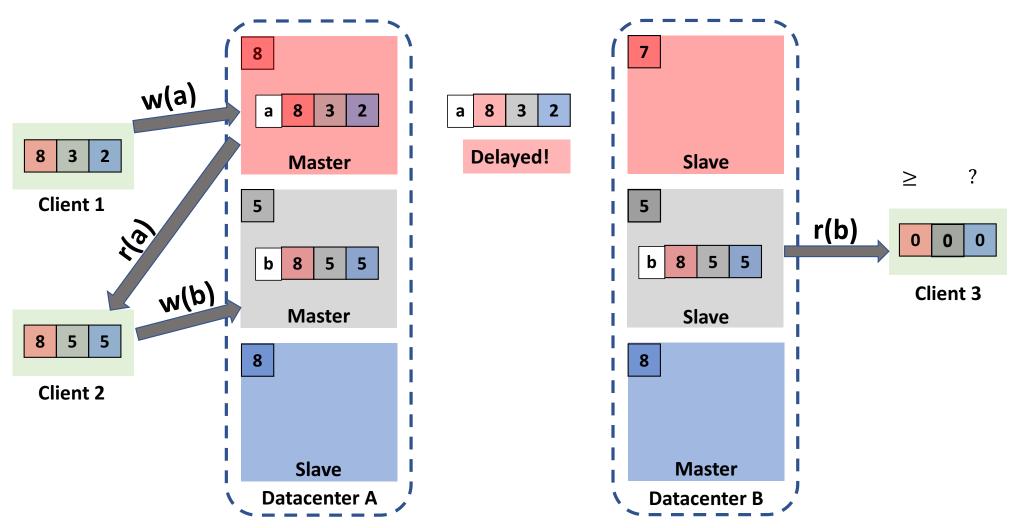
Read Protocol: Causal timestamp merging tracks causal ordering for writes following reads



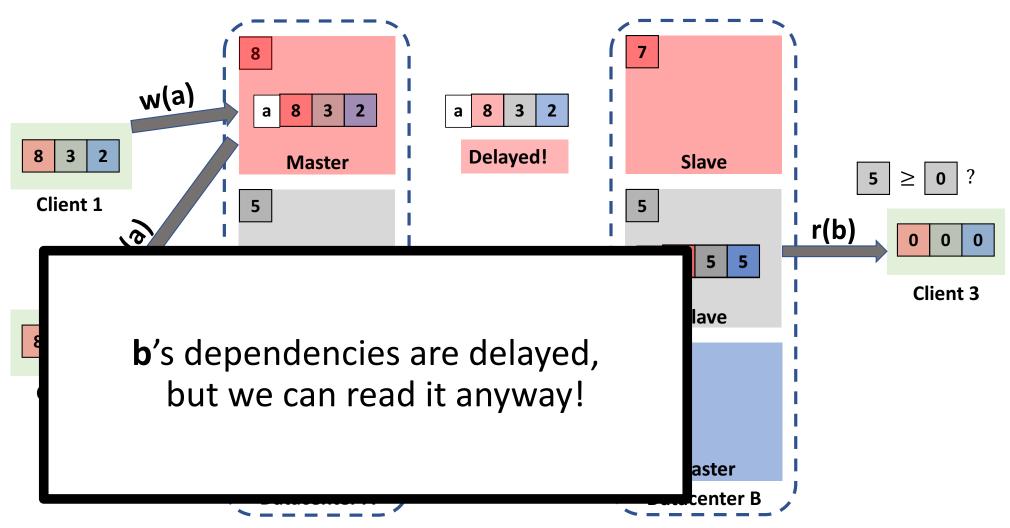
Replication: Like eventual consistency; asynchronous, unordered, writes applied immediately



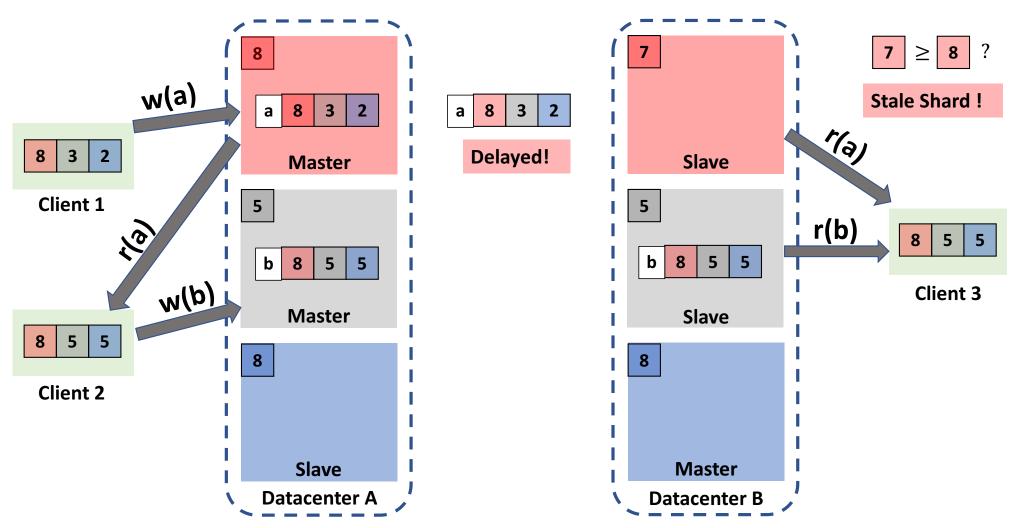
Replication: Slaves increment their shardstamps using causal timestamp of a replicated write



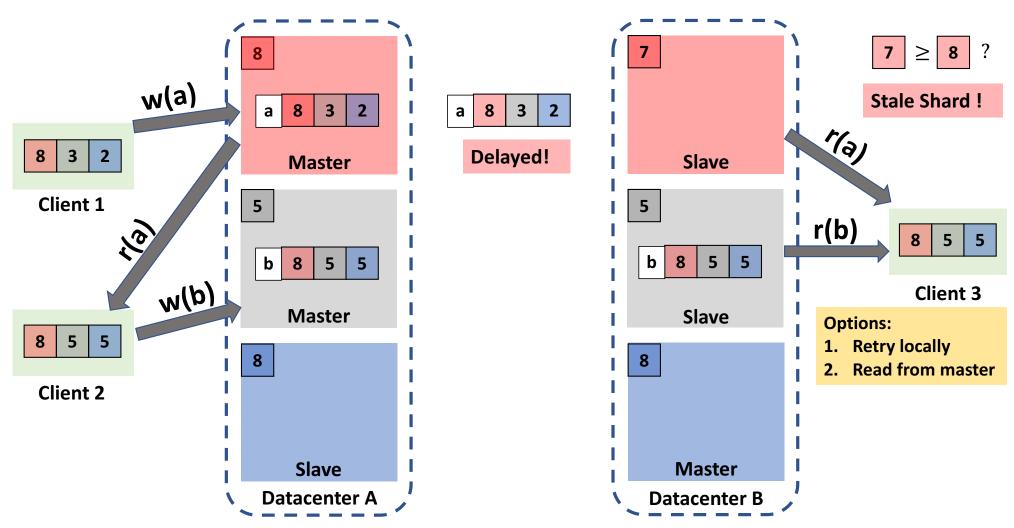
Read Protocol: Clients do consistency check when reading from slaves



Read Protocol: Clients do consistency check when reading from slaves



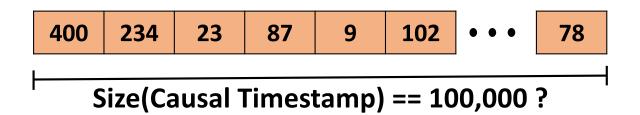
Read Protocol: Clients do consistency check when reading from slaves



Read Protocol: Resolving stale reads

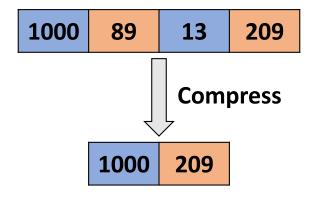
Causal Timestamp Compression

• What happens at scale when number of shards is (say) 100,000?



Causal Timestamp Compression: Strawman

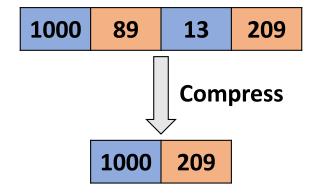
• To compress down to *n*, conflate shardstamps with same ids modulo *n*



• Problem: False Dependencies

Causal Timestamp Compression: Strawman

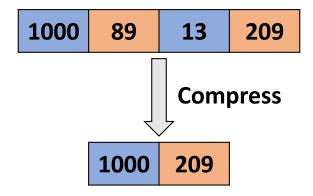
To compress down to n, conflate shardstamps with same ids modulo n



- Problem: False Dependencies
- Solution:
 - Use system clock as the next value of shardstamp on a write
 - Decouples shardstamp value from number of writes on each shard

Causal Timestamp Compression: Strawman

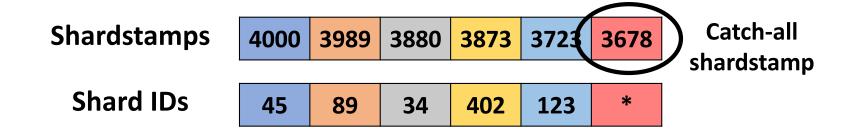
• To compress from **N** to **n**, conflate shardstamps with same ids modulo **n**



• Problem: Modulo arithmetic still conflates unrelated shardstamps

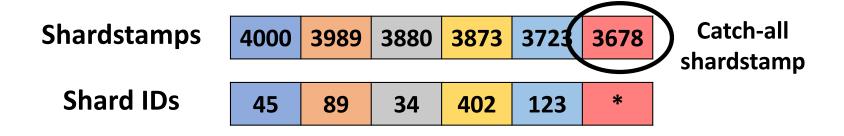
Causal Timestamp Compression

- Insight: Recent shardstamps more likely to create false dependencies
- Use high resolution for recent shardstamps and conflate the rest



Causal Timestamp Compression

- Insight: Recent shardstamps more likely to create false dependencies
- Use high resolution for recent shardstamps and conflate the rest



• 0.01 % false dependencies with just 4 shardstamps and 16K logical shards

Transactions in OCCULT

Scalable causally consistent general purpose transactions

Properties of Transactions

- A. Atomicity
- B. Read from a causally consistent snapshot
- C. No concurrent conflicting writes

Properties of Transactions

- A. Observable atomicity
- **B.** Observably read from a causally consistent snapshot
- C. No concurrent conflicting writes

- A. Observable Atomicity
- B. Observably read from a causally consistent snapshot
- C. No concurrent conflicting writes

Properties of Protocol

- 1. No centralized timestamp authorities (e.g. per-datacenter)
 - Transactions ordered using causal timestamps

- A. Observable Atomicity
- B. Observably read from a causally consistent snapshot
- C. No concurrent conflicting writes

Properties of Protocol

- 1. No centralized timestamp authority (e.g. per-datacenter)
 - Transactions ordered using causal timestamps
- 2. Transaction commit latency is independent of number of replicas

- A. Observable Atomicity
- B. Observably read from causally consistent snapshot
- C. No concurrent conflicting writes

Three Phase Protocol

1. Read Phase

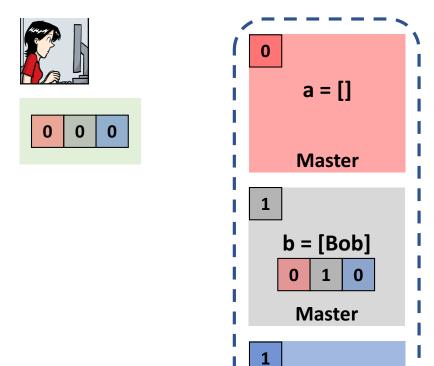
Buffer writes at client

2. Validation Phase

Client validates A, B and C using causal timestamps

3. Commit Phase

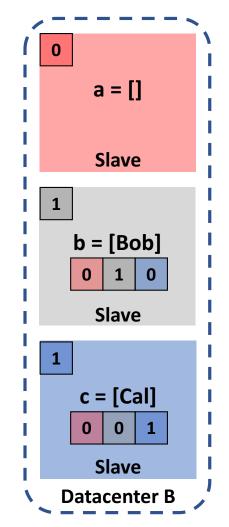
Buffered writes committed in an observably atomic way



c = [Cal]

Master

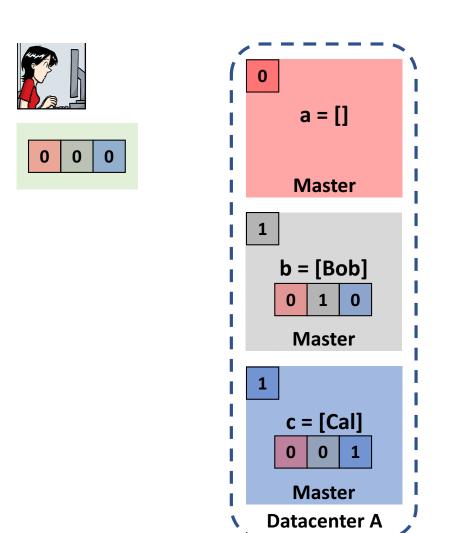
Datacenter A

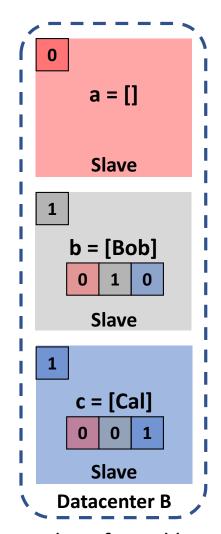


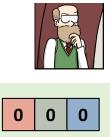


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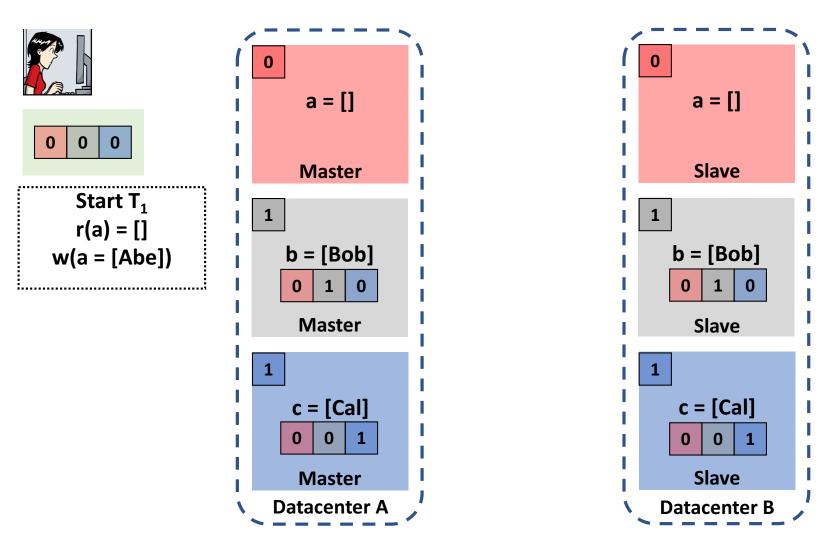
Alice and her advisor are managing lists of students for three courses



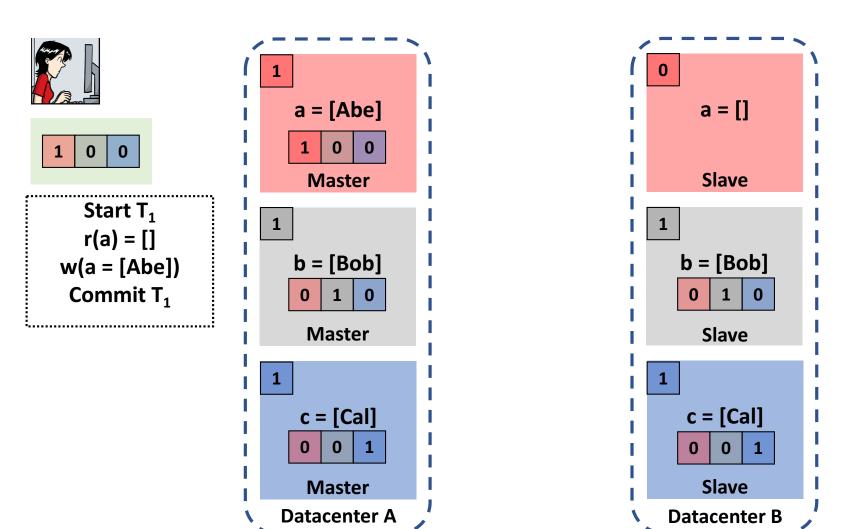




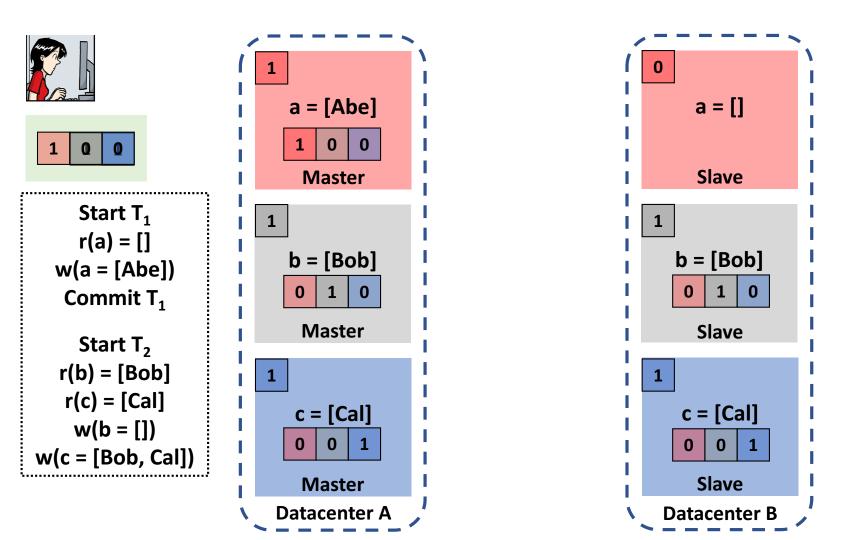
Observable atomicity and causally consistent snapshot reads enforced by same mechanism



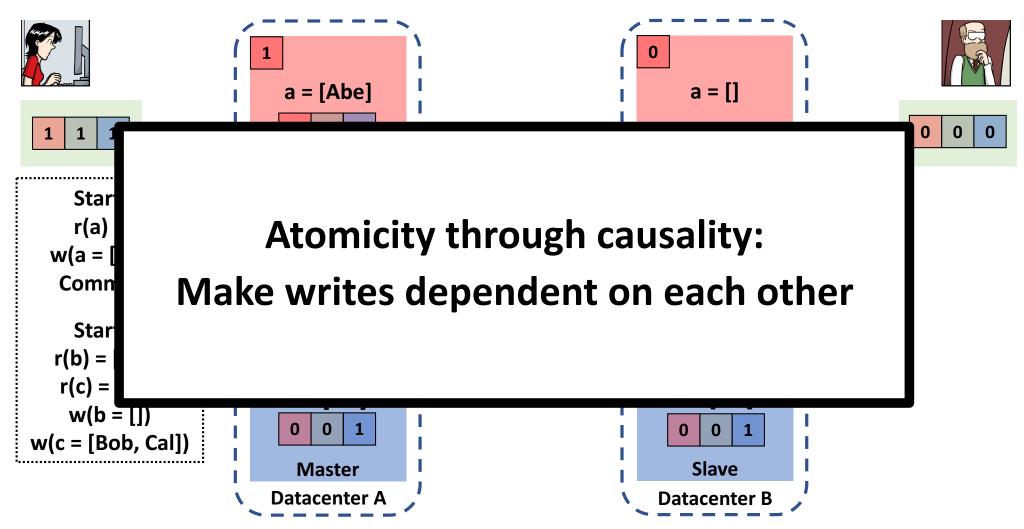
Transaction T₁: Alice adding Abe to course a



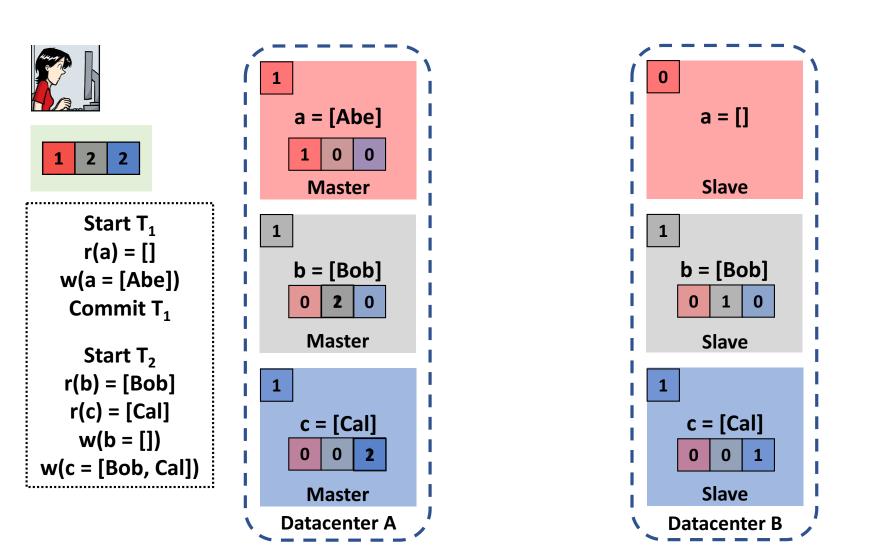
Transaction T₁: After Commit



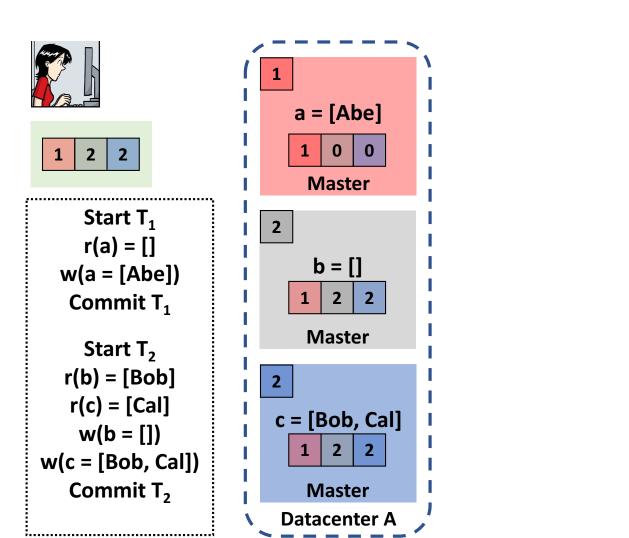
Transaction T₂: Alice moving Bob from course **b** to course **c**

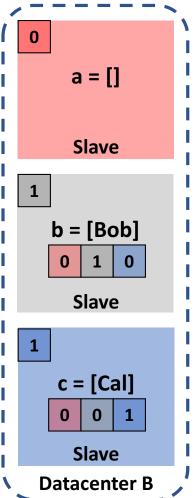


Observable Atomicity: Make writes causally dependent on each other



Observable Atomicity: Same commit timestamp makes writes causally dependent on each other







0 0 0

Observable Atomicity: Same commit timestamp makes writes causally dependent on each other



Start T₁

r(a) = []

w(a = [Abe])

Commit T₁

Start T₂

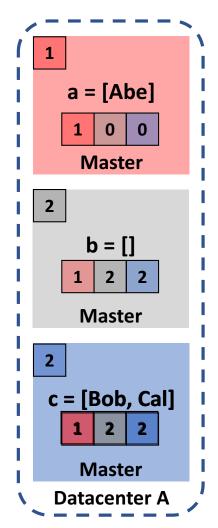
r(b) = [Bob]

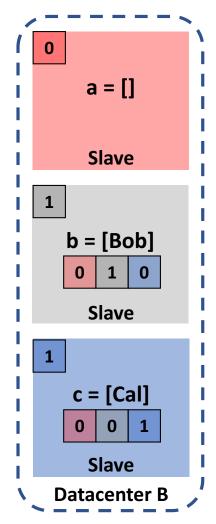
r(c) = [Cal]

w(b = [])

w(c = [Bob, Cal])

Commit T₂

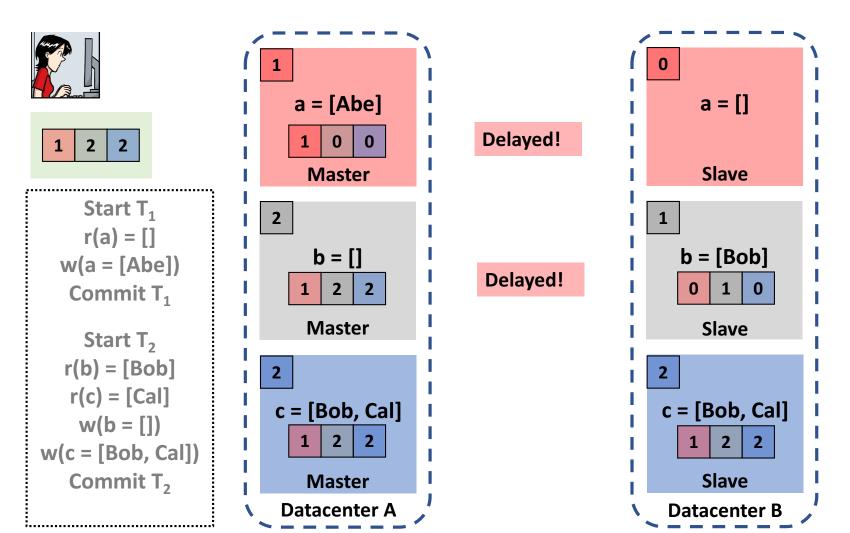




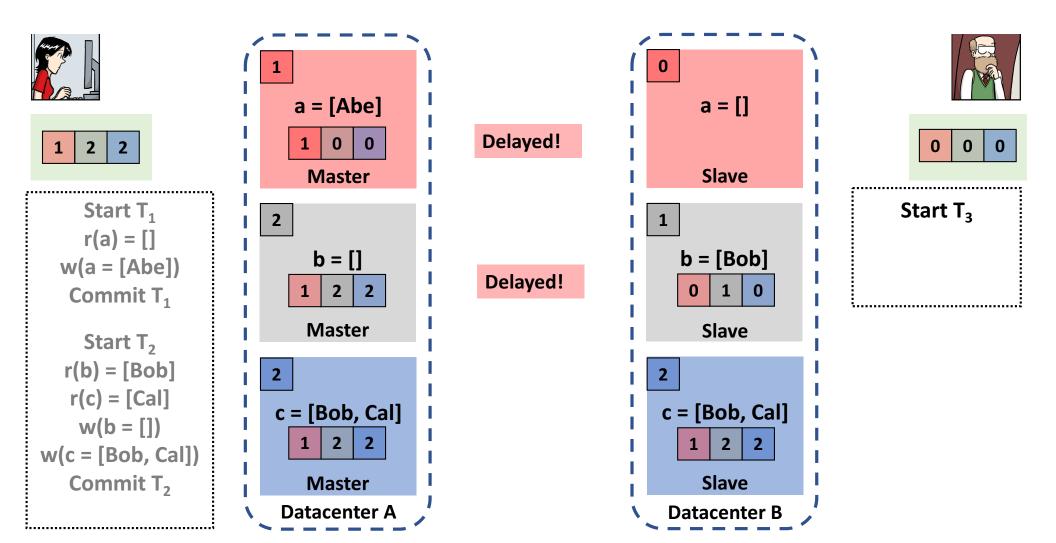


0 0 0

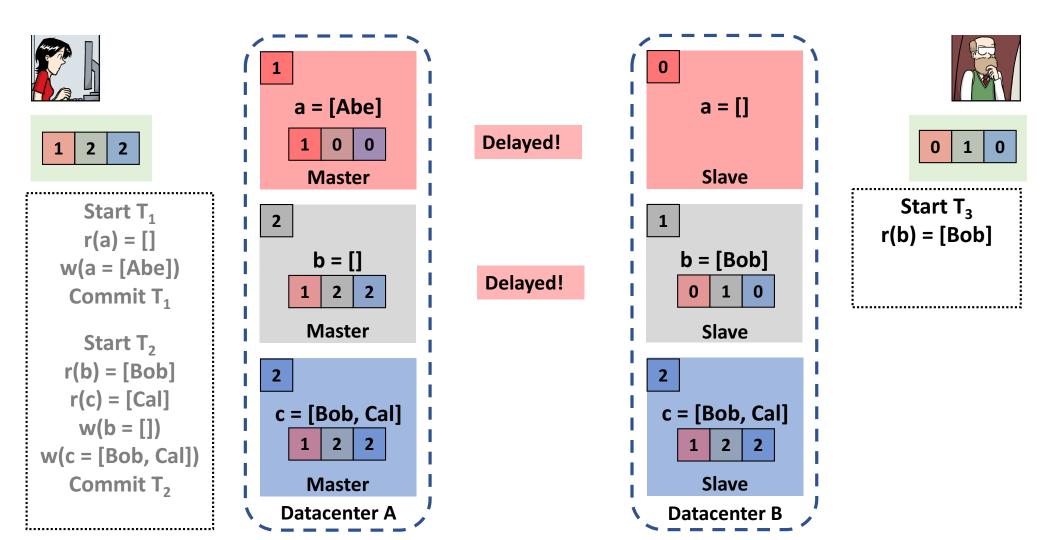
Transaction writes replicate asynchronously



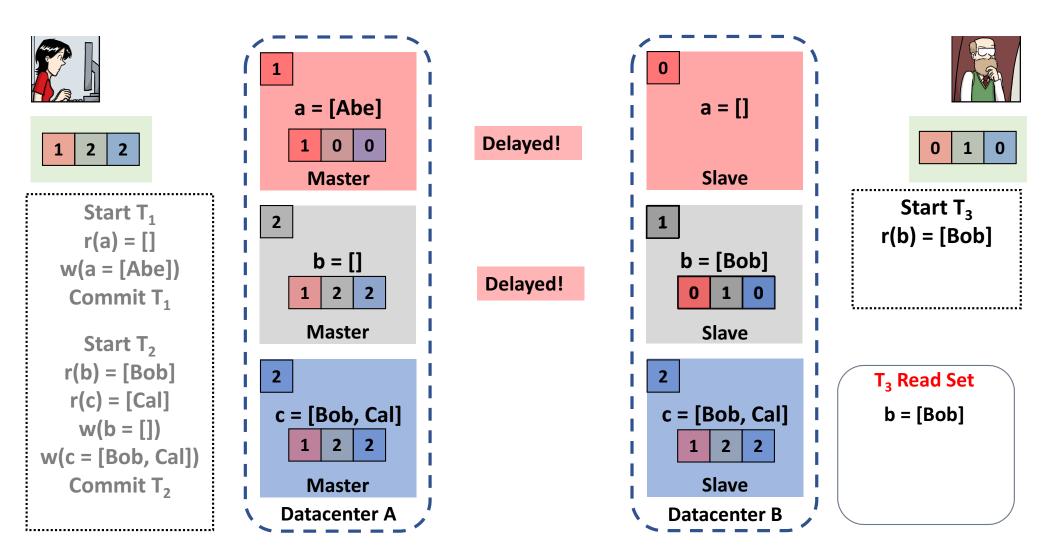
Transaction writes replicate asynchronously



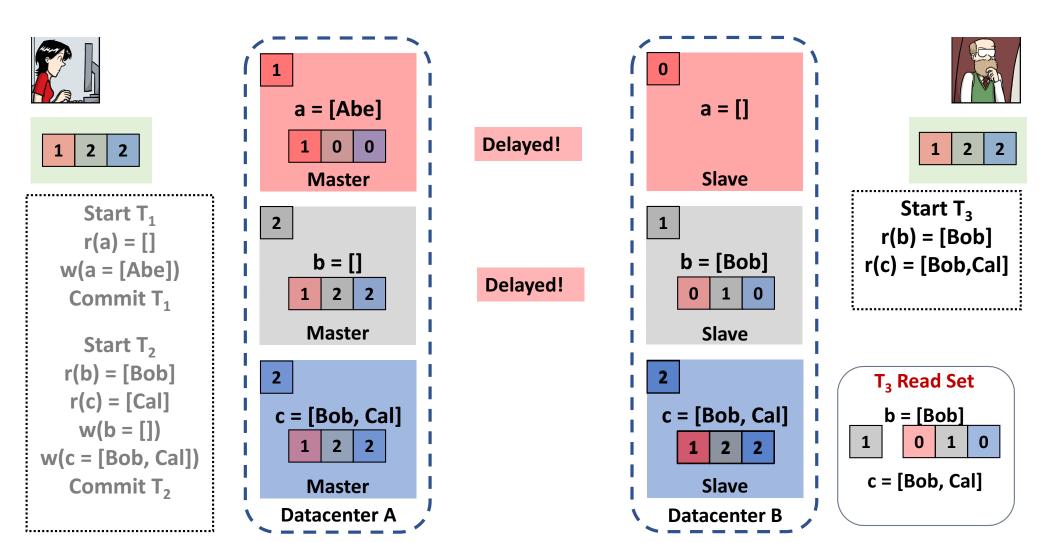
Alice's advisor reads the lists in a transaction



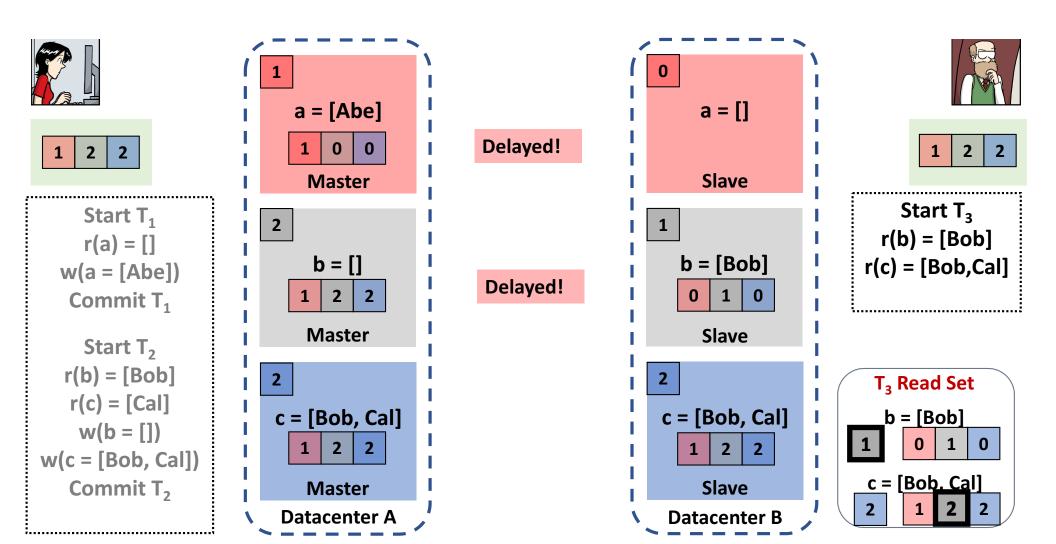
Alice's advisor reads the lists in a transaction



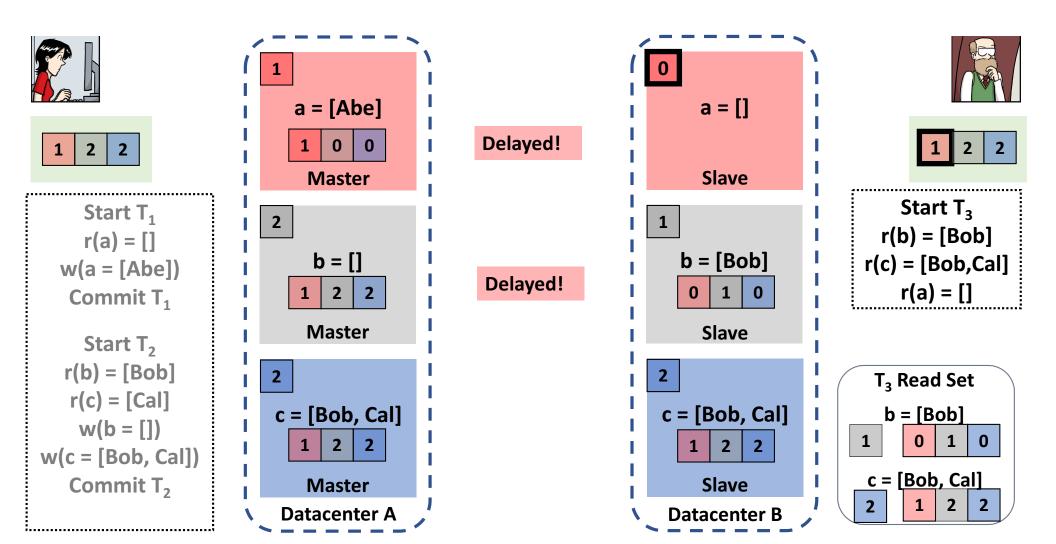
Transactions maintain a Read Set to validate atomicity and causal snapshot reads



Transactions maintain a Read Set to validate atomicity and read from causal snapshot



Validation failure: C knows more writes from grey shard than applied at the time b was read



Ordering Violation: Detected in the usual way. Red Shard is stale!

- A. Observable Atomicity
- B. Observably read from causally consistent snapshot
- C. No concurrent conflicting writes

Three Phase Protocol

1. Read Phase

Buffer writes at client

2. Validation Phase

Client validates A, B and C using causal timestamps

3. Commit Phase

Buffered writes committed in an observably atomic way

- A. Observable Atomicity
- B. Observably read from causally consistent snapshot
- C. No concurrent conflicting writes

Three Phase Protocol

1. Read Phase

Buffer writes at

2. Validation Phase

Client_validates

2. Validation Phase

- a. Validate Read Set to verify A and B
- b. Validate Overwrite Set to verify C

3. Commit Phase

Buffered writes committed in an observably atomic way

Evaluation

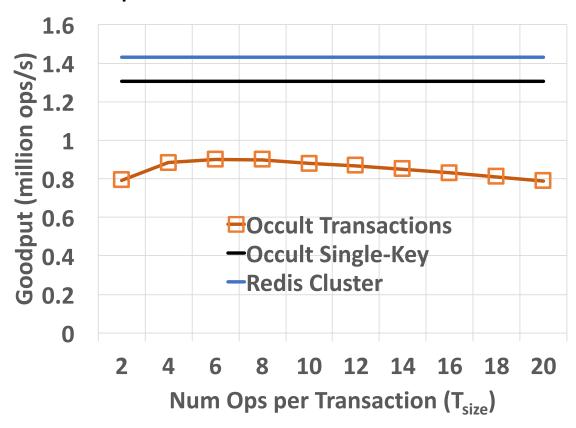
Evaluation Setup

- Occult implemented by modifying Redis Cluster (baseline)
- Evaluated on CloudLab
 - Two datacenters in WI and SC
 - 20 server machines (4 server processes per machine)
 - 16K logical shards
- YCSB used as the benchmark
 - For graphs shown here read-heavy (95% reads) workload with zipfian distribution

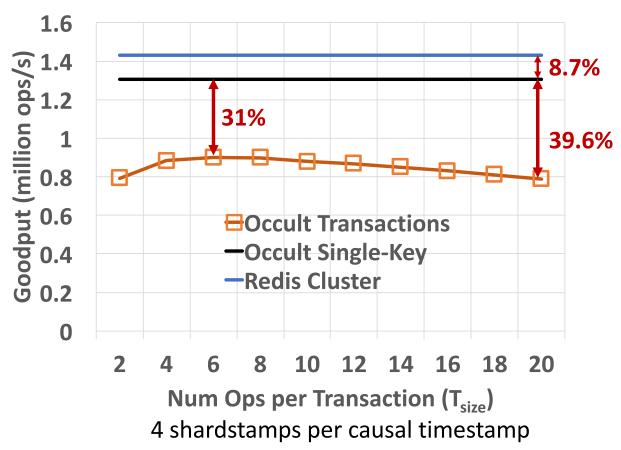
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 - For graphs shown here read-heavy (95% reads) workload with zipfian distribution
- We show cost of providing consistency guarantees

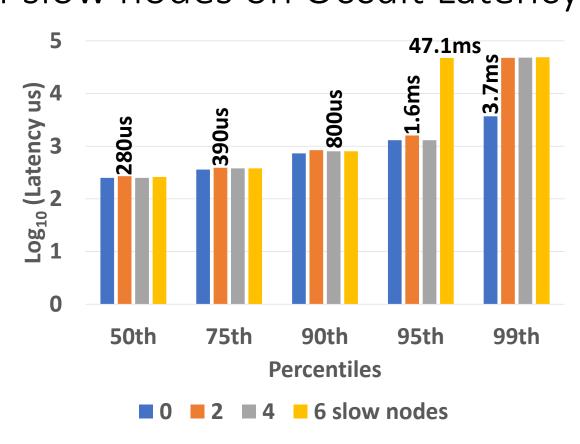
Goodput Comparison



Goodput Comparison



Effect of slow nodes on Occult Latency



Conclusions

- Enforcing causal consistency in the data store is vulnerable to slowdown cascades
- Sufficient to ensure that clients **observe** causal consistency:
 - Use lossy timestamps to provide the guarantee
 - Avoid slowdown cascades
- Observable enforcement can be extended to causally consistent transactions
 - Make writes causally dependent on each other to observe atomicity
 - Also avoids slowdown cascades