Implementing Linearizability at Large Scale and Low Latency

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## **Overview**

- **Goal:** take back consistency in large-scale systems
- **Approach:** distinct layer for linearizability
- Reusable Infrastructure for Linearizability (RIFL)
  - At-least-once RPC → exactly-once RPC
  - Records RPC results durably
  - To handle reconfiguration, associates metadata with object

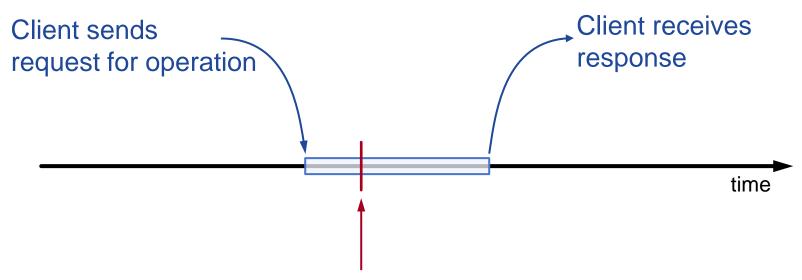
#### • Implemented on distributed KV store, RAMCloud

- Low latency: < 5% (500ns) latency overhead</p>
- Scalable: supports states for 1M clients

#### • **RIFL simplified implementing transactions**

- Simple distributed transaction commits in ~22 μs
- Outperforms H-Store

## What is Linearizability?

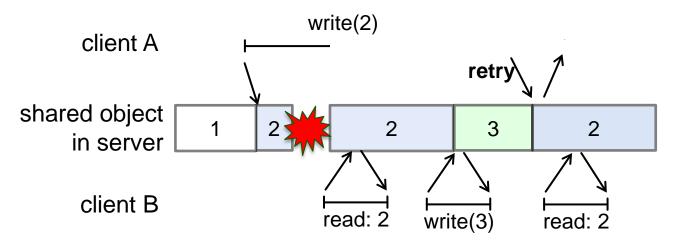


Behaves as if executing exactly once and instantaneously

Strongest form of consistency for concurrent systems

#### What is Missing from Existing Systems?

- Most systems: at-least-once semantics, not exactly-once
  - Retry (possibly) failed operations after crashes
  - Idempotent semantics: repeated executions O.K.?



- At-least-once + idempotency ≠ linearizability
- Need exactly-once semantics!

# **Architecture of RIFL**

#### • **RIFL saves results of RPCs**

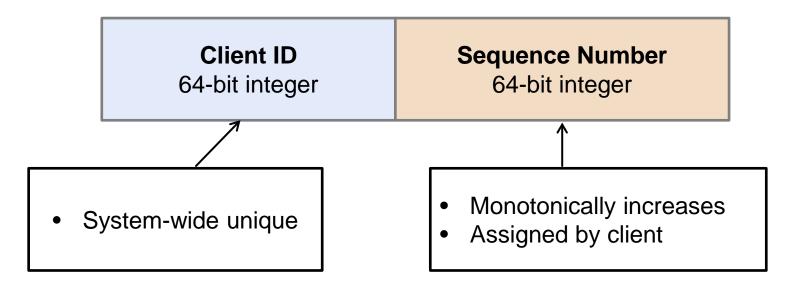
- If client retries,
  - Don't re-execute
  - Return saved result

#### Key problems

- Unique identification for each RPC
- Durable completion record
- Retry rendezvous
- Garbage collection

# 1) RPC Identification

- Each RPC must have a unique ID
- Retries use the same ID
- Client assigns RPC IDs



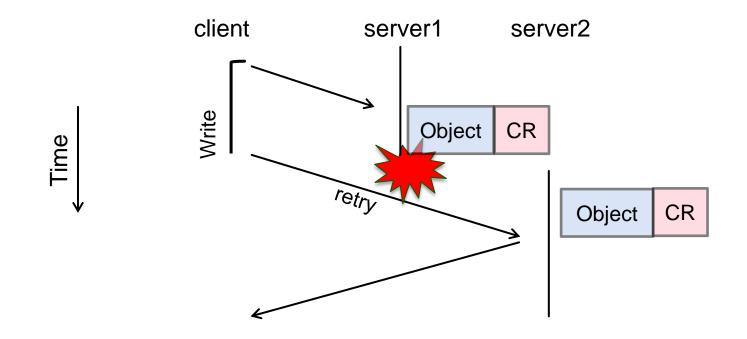
# 2) Durable Completion Record

- Written when an operation completes
- Same durability as object(s) being mutated
- Atomically created with object mutation

| <b>Completion Record</b> |                    |
|--------------------------|--------------------|
| Client ID                | Sequence<br>Number |
| RPC result for client    |                    |

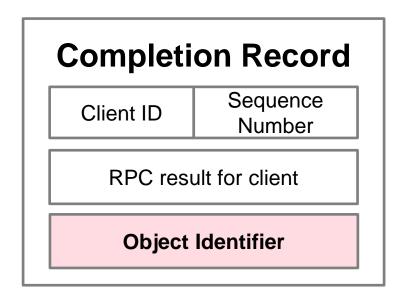
# 3) Retry Rendezvous

- Data migration is popular in large systems (eg. crash recovery)
- Retries must find completion record



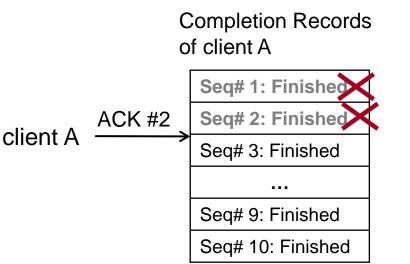
# 3) Retry Rendezvous (cont.)

- Associate each RPC with a specific object
- Completion record follows the object during migration



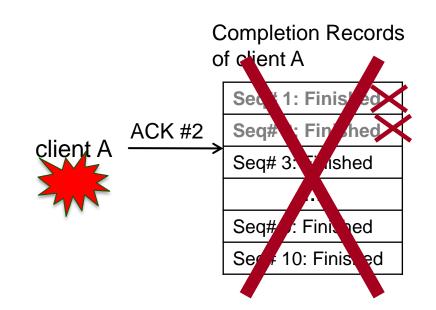
# 4) Garbage Collection

- Lifetime of completion record ≠ lifetime of object value
- Can't GC if client may retry later
- Server knows a client will never retry if
- 1. Client acknowledges receipt of RPC result

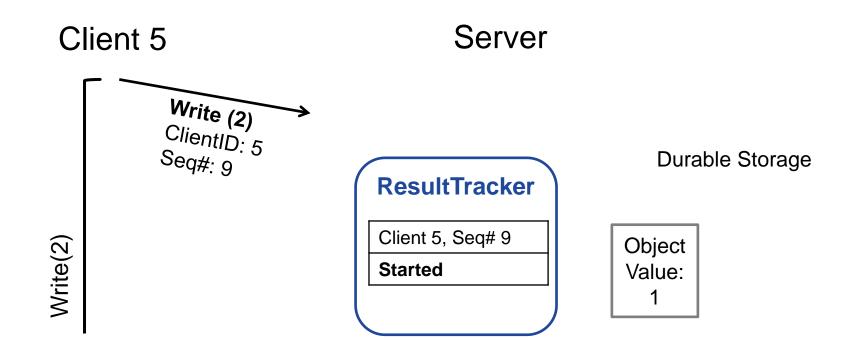


# 4) Garbage Collection

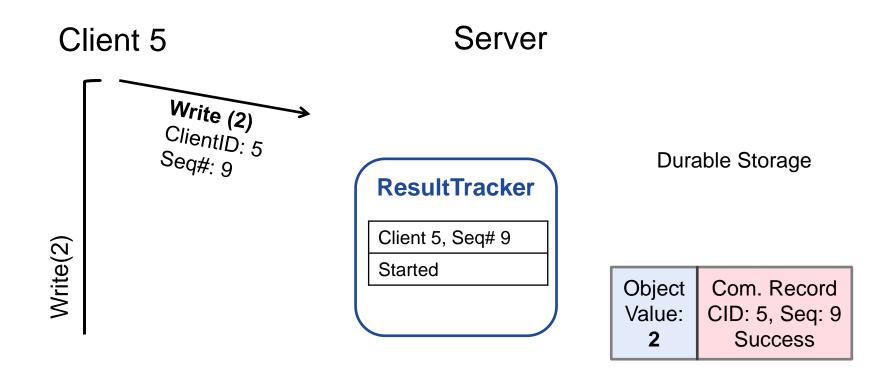
- Lifetime of completion record
   ≠ lifetime of object value
- Can't GC if client may retry later
- Server knows a client will never retry if
- 1. Client acknowledges receipt of RPC result
- 2. Detect client crashes with lease.



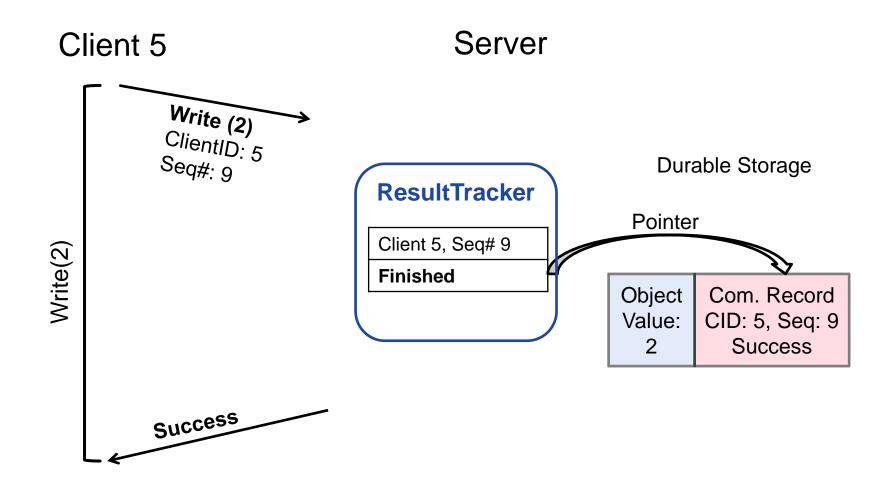
### **Linearizable RPC in Action**



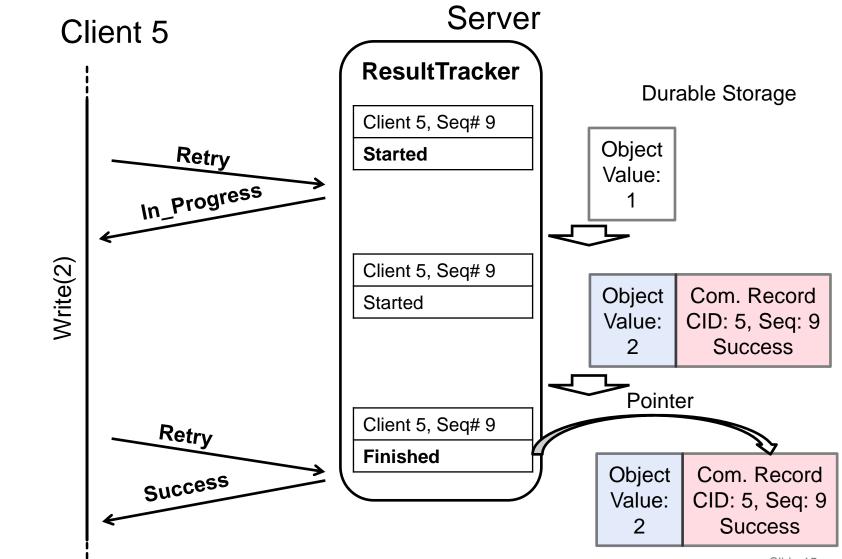
### **Linearizable RPC in Action**



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# **Handling Retries**



Time

# Performance of RIFL in RAMCloud

- Achieved linearizability without hurting performance of RAMCloud
  - Minimal overhead on latency(< 5%) and throughput (~0%)
  - Supports state for 1M clients

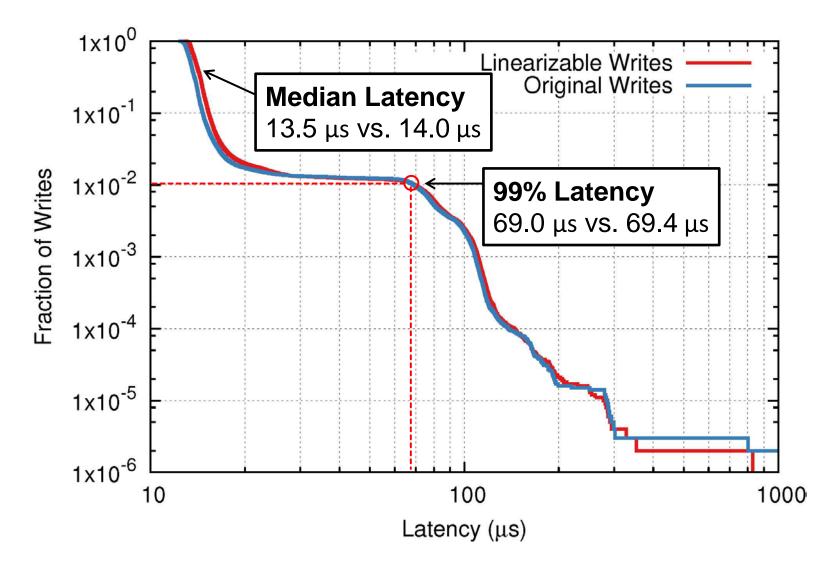
## Why RAMCloud?

- General-purpose distributed in-memory key-value storage system
- **Durability:** 3-way replication
- Fast recovery: 1~2 sec for server crash
- Large scale: 1000+ servers, 100+ TB
- Low latency: 4.7 µs read, 13.5 µs write (100B object)
- RIFL is implemented on top of RAMCloud
  - Core: 1200 lines of C++ for infrastructure
  - Per operation: 17 lines of C++ to make an operation linearizable

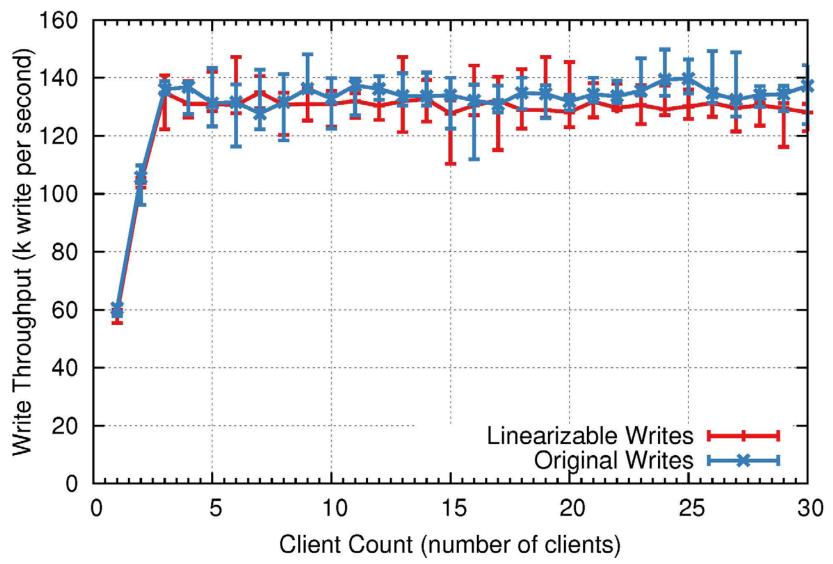
## **Experimental Setup**

- Server: Xeon 4 cores at 3 GHz
- Fast Network
  - Infiniband (24 Gbps)
  - Kernel-bypassing transport (RAMCloud default)

### **Impact on Latency**

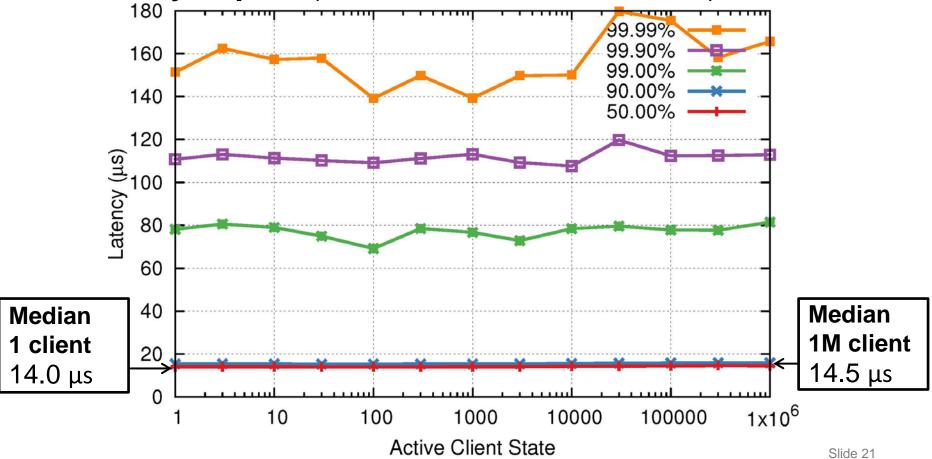


### **Impact on Throughput**



## **Impact of Many Client States**

- **Storage impact:** 116B per client → 116MB for 1M clients
- Latency impact (linearizable write, unloaded):



# Case study: Distributed Transactions with RIFL

- Extended use of RIFL for more complex operations
- > Two-phase commit protocol based on Sinfonia
- RIFL reduced mechanisms of Sinfonia significantly
- Lowest possible round-trip for distributed transactions

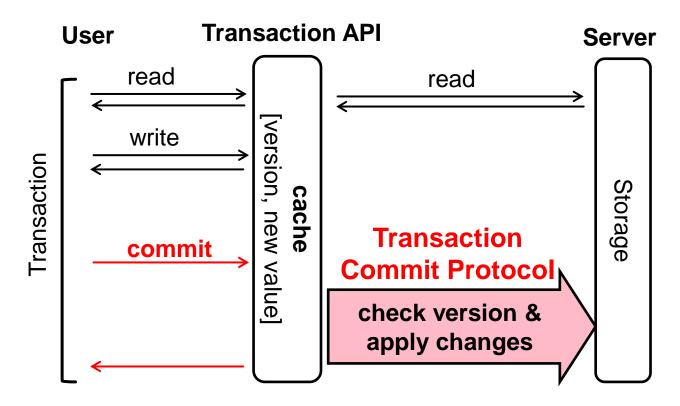
```
class Transaction {
    read(tableId, key) => blob
    write(tableId, key, blob)
    delete(tableId, key)
    commit() => COMMIT or ABORT
}
```

- Optimistic concurrency control
- Mutations are "cached" in client until commit

## **Transaction Commit Semantics**

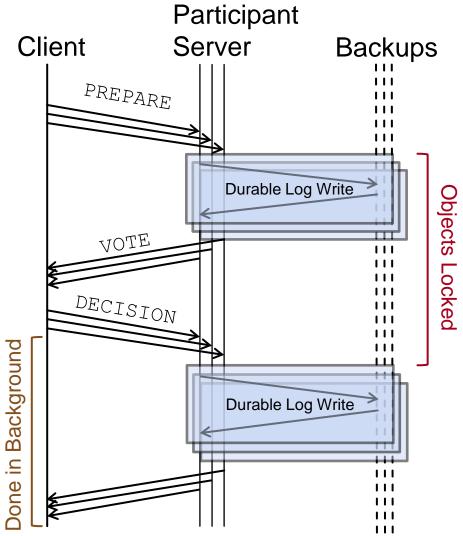
#### • commit(): atomic multi-object operation

- Operations in client's cache are transmitted to servers
- Conditioned on a version (same version = object didn't change)



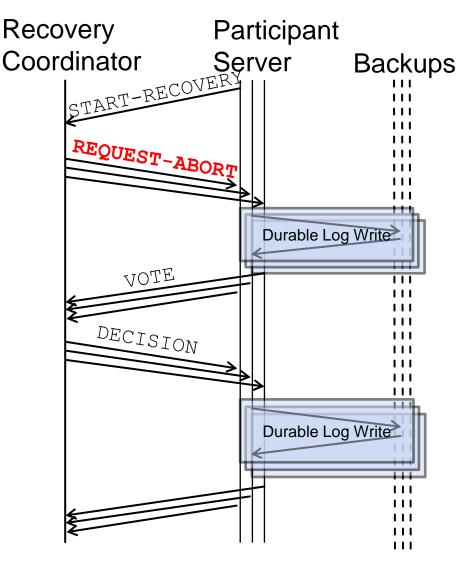
## **Transaction Commit Protocol**

- Client-driven 2PC
- RPCs:
  - PREPARE() => VOTE
  - DECISION()
- Fate of TX is determined after 1<sup>st</sup> phase
- Client blocked for 1 RTT + 1 log write
- Decisions processed in background



## **Transaction Recovery on Client-crash**

- Server-driven 2PC
- Initiated by "worried" server
- RPCs:
  - START-RECOVERY()
  - REQUEST-ABORT() => VOTE
  - DECISION()



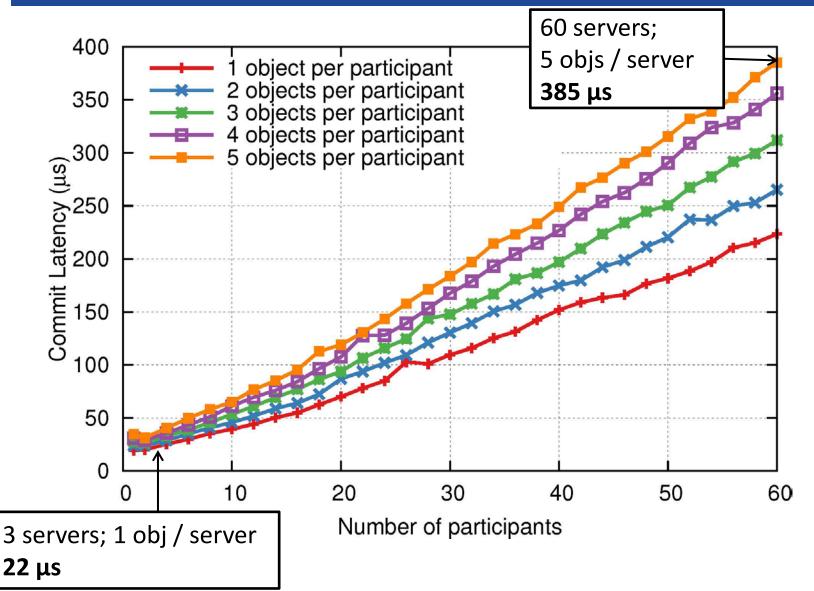
## **RIFL Simplifies TX Recovery**

- PREPARE() => VOTE is linearizable RPC
- Server crash: client retries PREPARE
- Client crash: recovery coordinator sends fake PREPARE (REQUEST-ABORT)
  - Query ResultTracker with same RPC ID from client
  - Writes completion record of PREPARE / REQUEST-ABORT
- Race between client's 2PC and recovery coordinator's 2PC is safe

# Performance of Transactions in RAMCloud

- **>** Simple distributed transactions commits in 22 μs
- > TPC-C benchmark shows RIFL-based transactions outperform H-Store

## **Latency of Transactions**



## **TPC-C Benchmark**

#### • TPC-C simulates order fulfillment systems

- New-Order transaction: 23 reads, 23 writes
- Used full-mix of TPC-C (~10% distributed)
- Latency is measured from end to end
- Modified TPC-C for benchmark to increase server load

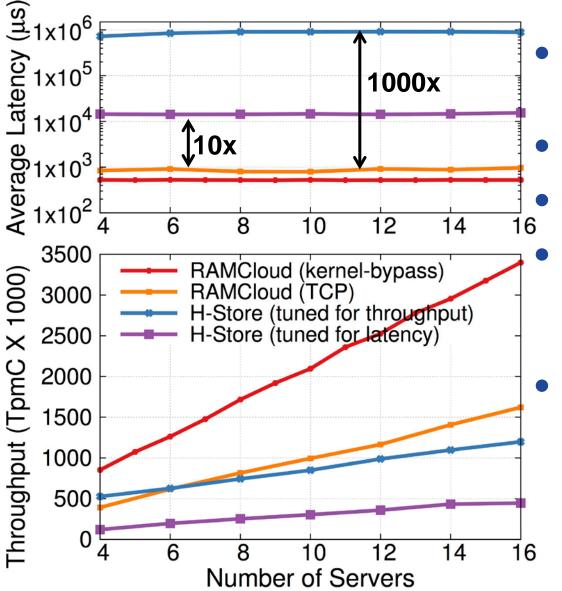
#### Compared with H-Store

Main-memory DBMS for OLTP

#### • Two RAMCloud configurations

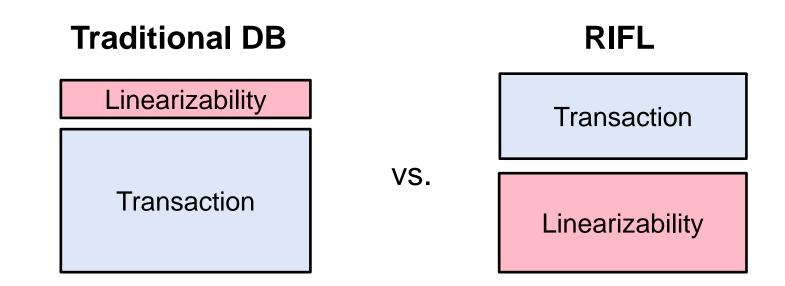
- Kernel-bypass for maximum performance
- Kernel TCP for fair comparison

## **TPC-C** Performance



- TpmC: NewOrder committed per minute
- 210,000 TpmC / server
  - Latency: ~500µs
  - **RAMCloud faster than H-Store** (even over TCP)
- Limited by serial logreplication (need batching)

#### Should linearizability be a foundation?



#### • Faster simple operations (eg. atomic increment)

- lower latency
- higher throughput
- Can be added on top of existing systems
- Better modular decomposition
  - easier to implement transaction

## Conclusion

- Distinct layer for linearizability → take back consistency in large-scale systems
- RIFL saves results of RPCs; If client retries, returns saved result without re-executing
  - 0.5us (< 5%) latency overhead, almost no throughput overhead.
- **RIFL** makes **transactions** easier
  - RIFL-based RAMCloud transaction: ~20 μs for commit
  - Outperform H-Store for TPC-C benchmark

#### Questions