Advanced Data Management (CSCI 640/490)

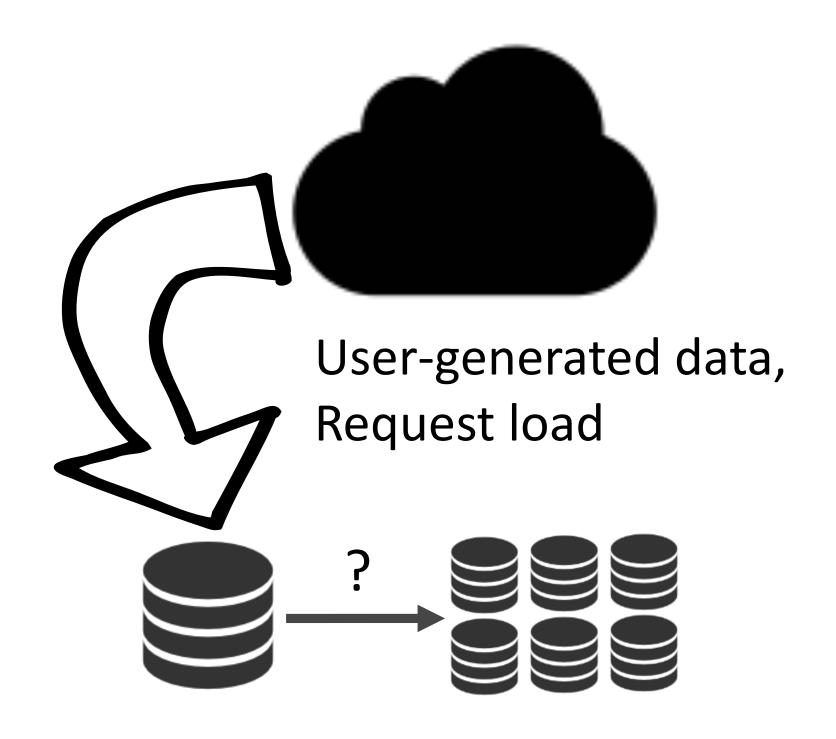
Scalable Databases

Dr. David Koop

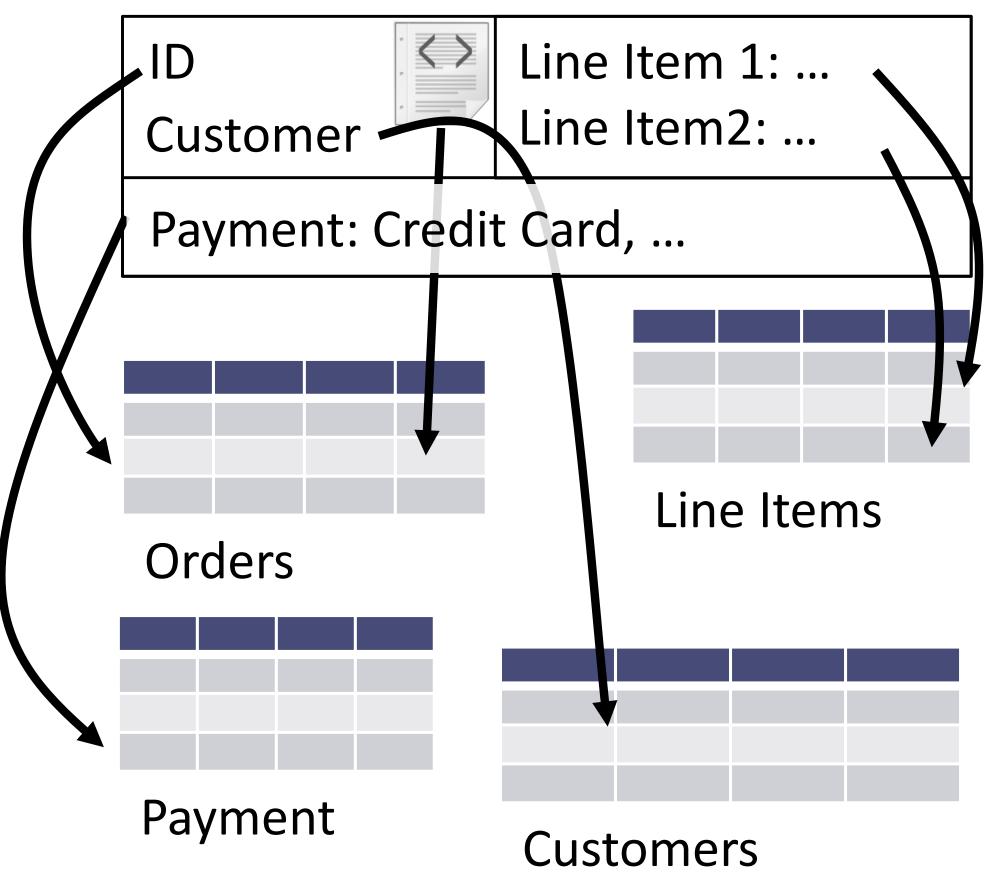


NoSQL Motivation

Scalability



Impedance Mismatch

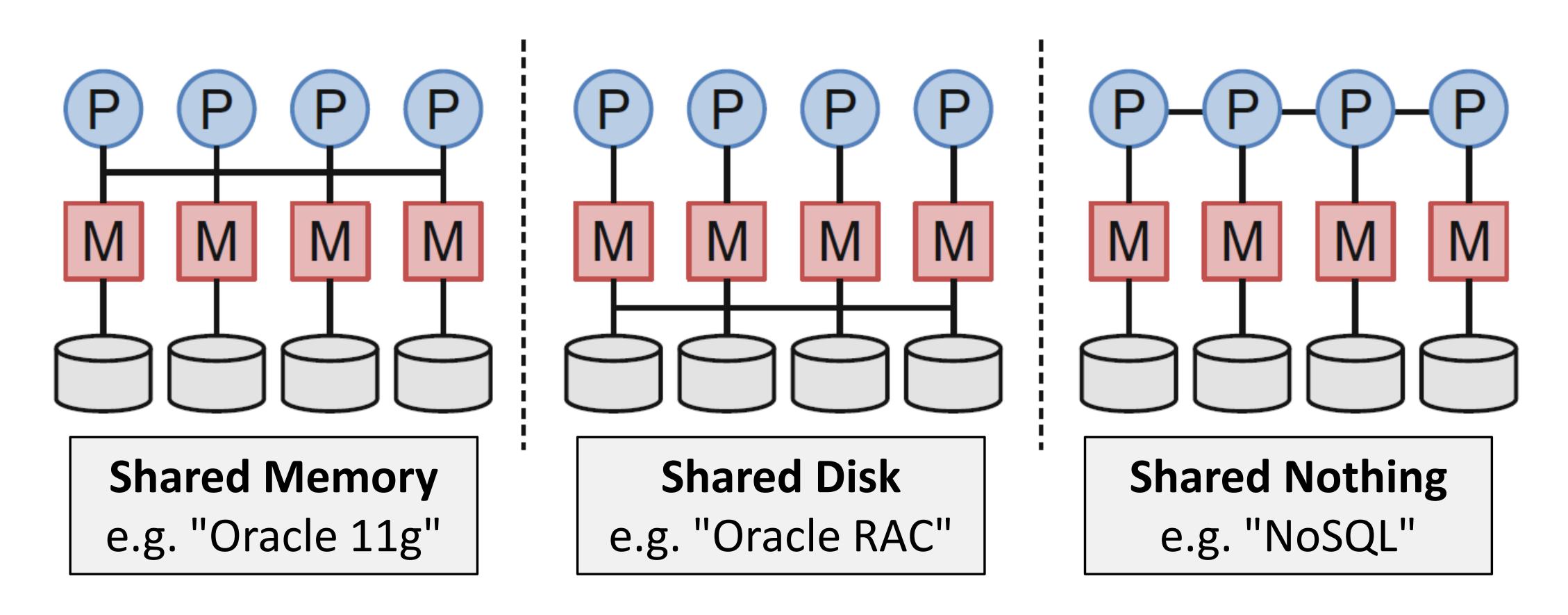


[F. Gessert et al., 2017]



Shared Nothing Architecture

Shift towards higher distribution & less coordination:



[F. Gessert et al., 2017]

Stonebraker: The End of an Architectural Era

- "RDBMSs were designed for the business data processing market, which is their sweet spot"
- "They can be beaten handily in most any other market of significant enough size to warrant the investment in a specialized engine"
- Changes in markets (science), necessary features (scalability), and technology (amount of memory)
- RDBMS Overhead: Logging, Latching, and Locking
- Relational model is not necessarily the answer
- SQL is not necessarily the answer

Horizontal Partitioning vs. Vertical Partitioning

Vertical Partitions

VP1 VP2

| CUSTOMER ID | FIRST NAME | LAST NAME |
|----------------|---------------|--------------|
| 1 | TAEKO | OHNUKI |
| 2 | O.V. | WRIGHT |
| 3 | SELDA | BAĞCAN |
| 4 | JIM | PEPPER |

| CUSTOMER ID | FAVORITE COLOR |
|----------------|-------------------|
| 1 | BLUE |
| 2 | GREEN |
| 3 | PURPLE |
| 4 | AUBERGINE |

CUSTOMER FIRST LAST FAVORITE COLOR 1 TAEKO OHNUKI BLUE 2 O.V. WRIGHT GREEN 3 SELDA BAĞCAN PURPLE 4 JIM PEPPER AUBERGINE

Horizontal Partitions

HP1

| CUSTOMER ID | FIRST NAME | LAST NAME | FAVORITE COLOR |
|----------------|---------------|--------------|-------------------|
| 1 | TAEKO | OHNUKI | BLUE |
| 2 | O.V. | WRIGHT | GREEN |

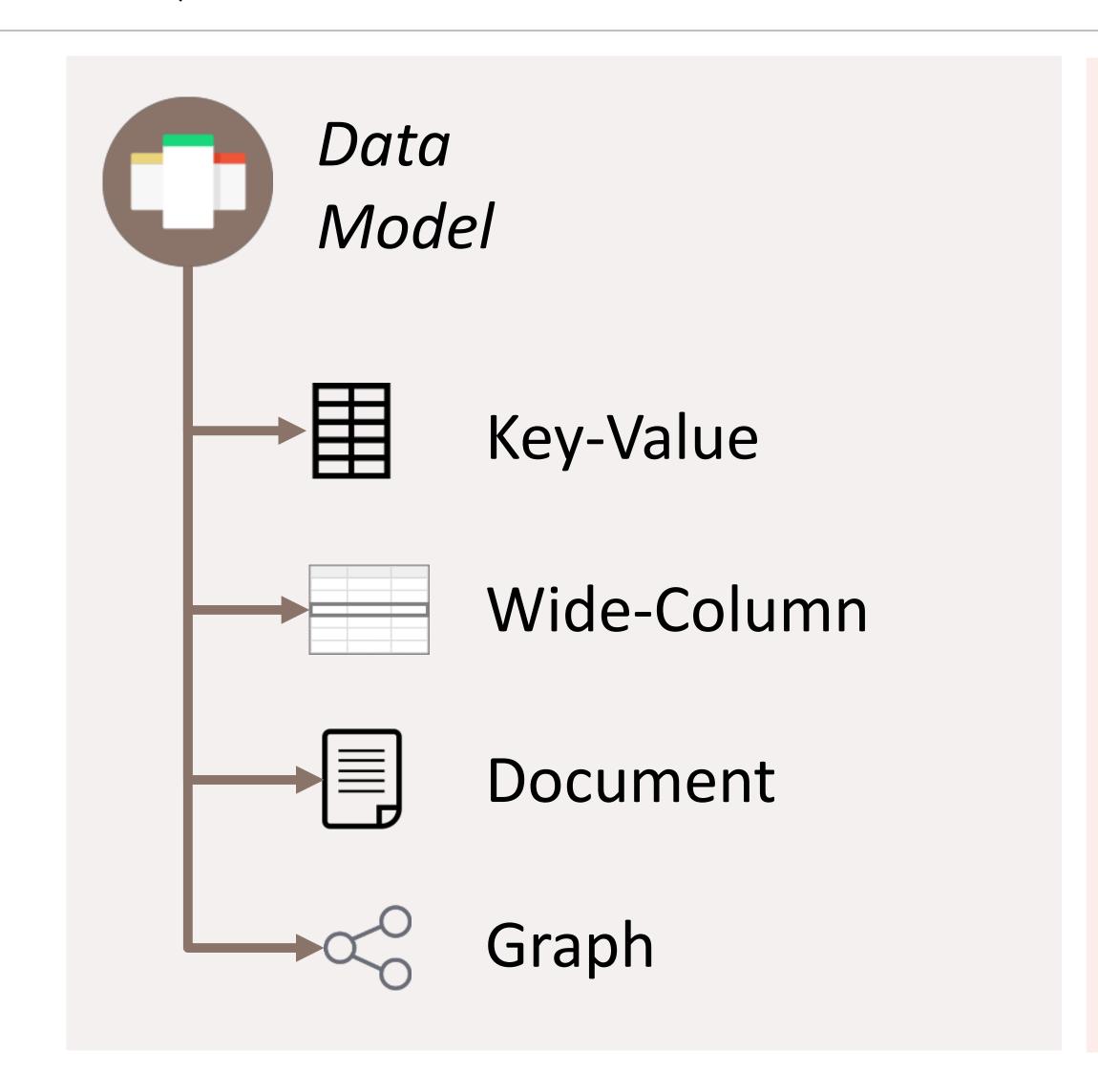
HP2

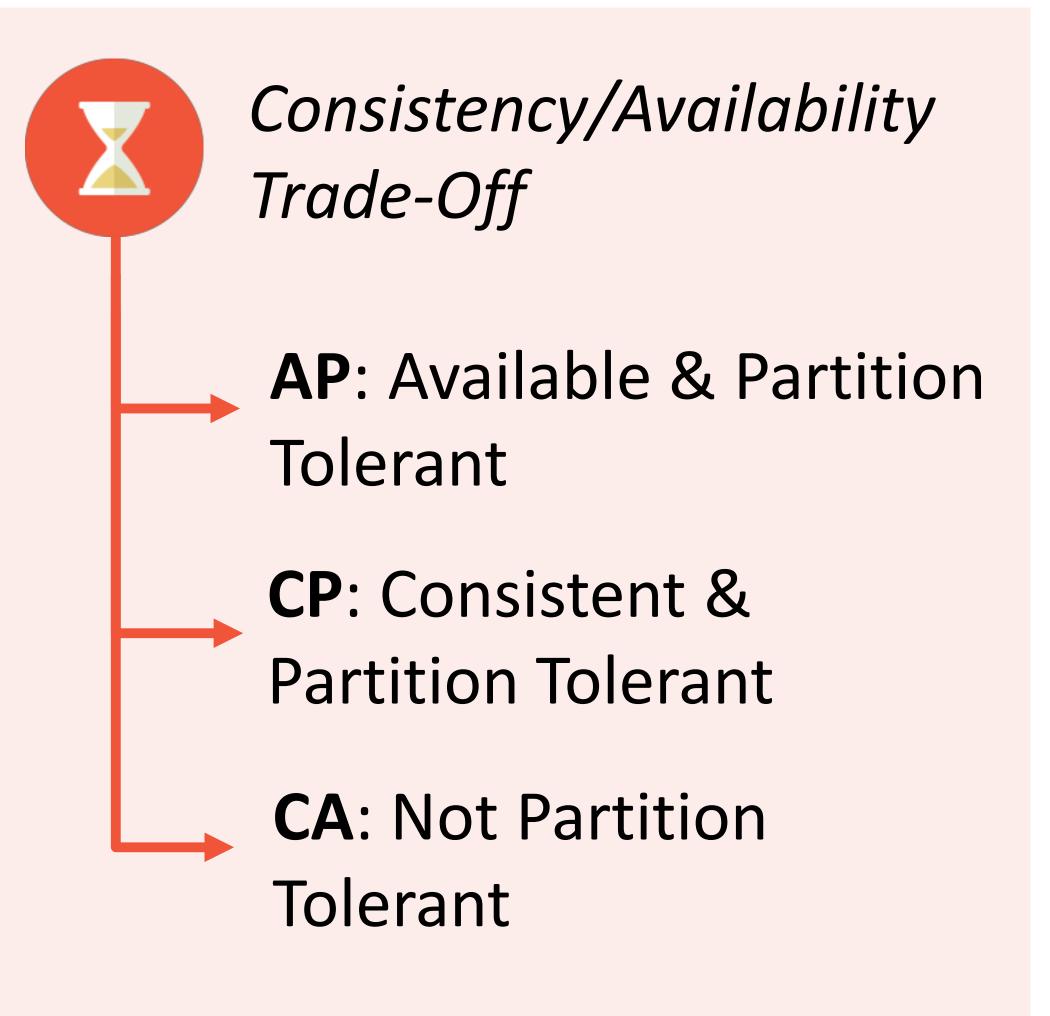
| CUSTOMER ID | FIRST NAME | LAST NAME | FAVORITE COLOR |
|----------------|---------------|--------------|-------------------|
| 3 | SELDA | BAĞCAN | PURPLE |
| 4 | JIM | PEPPER | AUBERGINE |

[M. Drake]



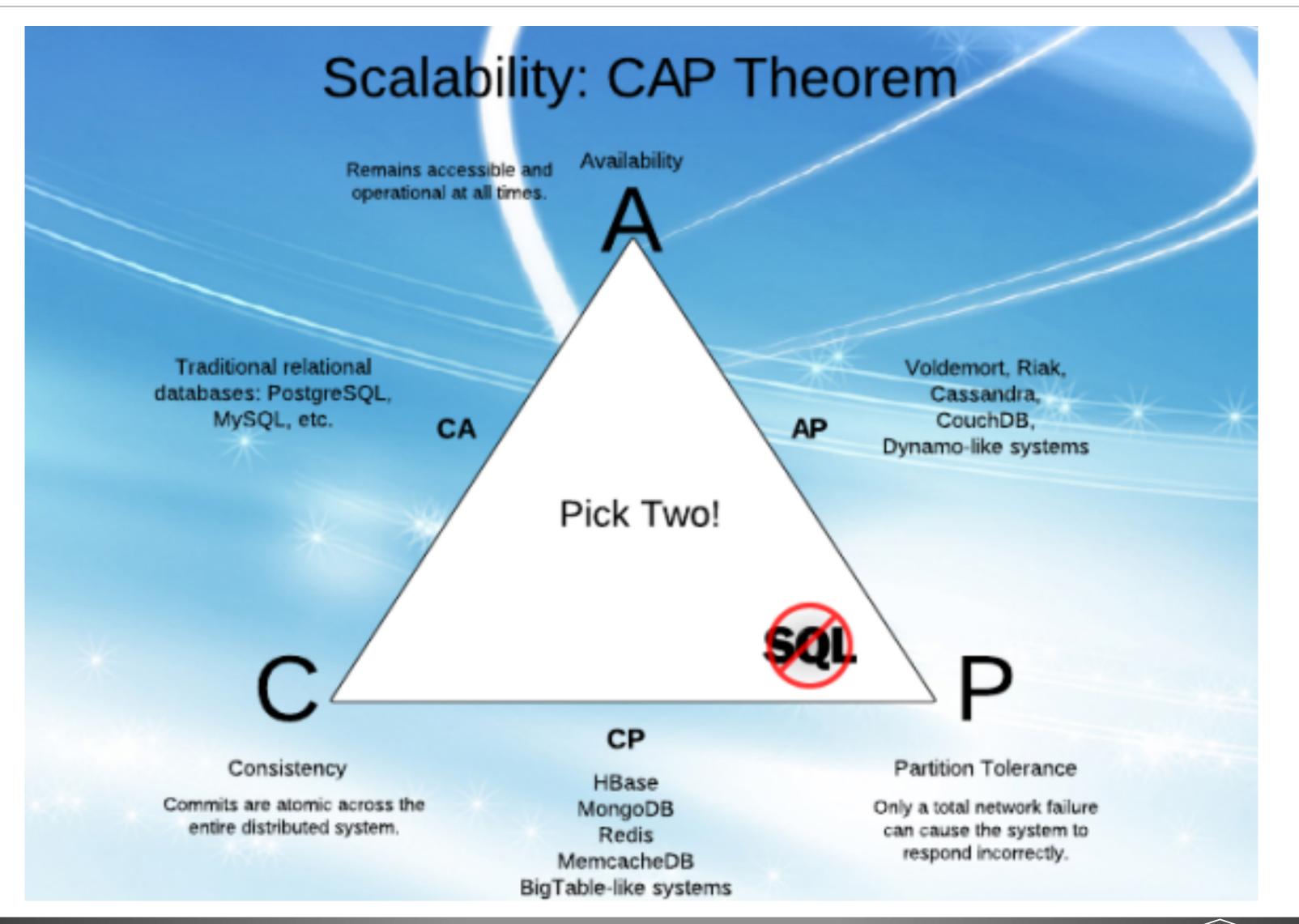
NoSQL Classification Criteria





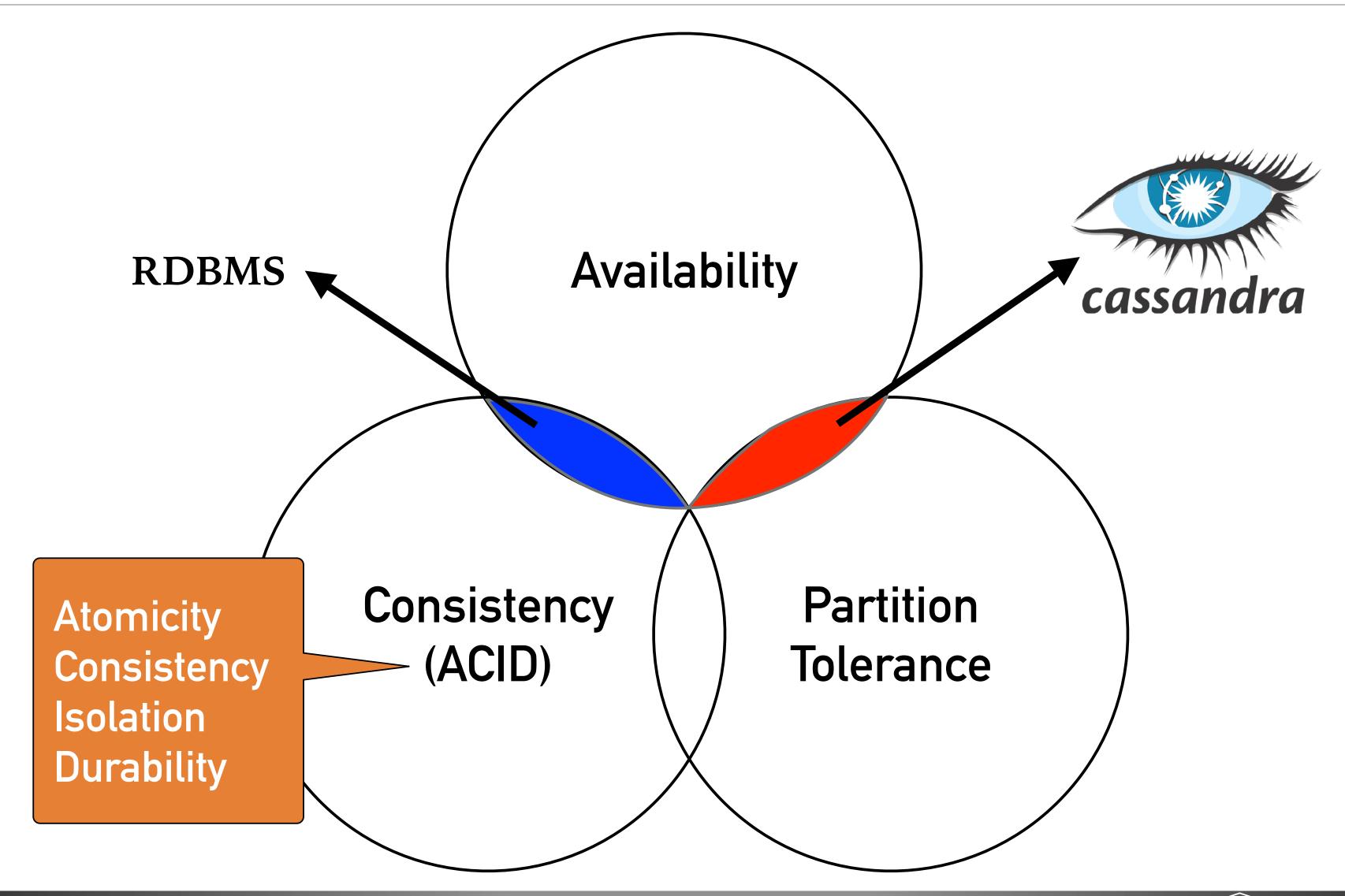
[F. Gessert et al., 2017]

CAP Theorem



[E. Brewer]

Cassandra and CAP



[G. Atil]

What is Cassandra?

- Fast Distributed (Column Family NoSQL) Database
 - High availability
 - Linear Scalability
 - High Performance
- Fault tolerant on Commodity Hardware
- Multi-Data Center Support
- Easy to operate
- Proven: CERN, Netflix, eBay, GitHub, Instagram, Reddit

[G. Atil]

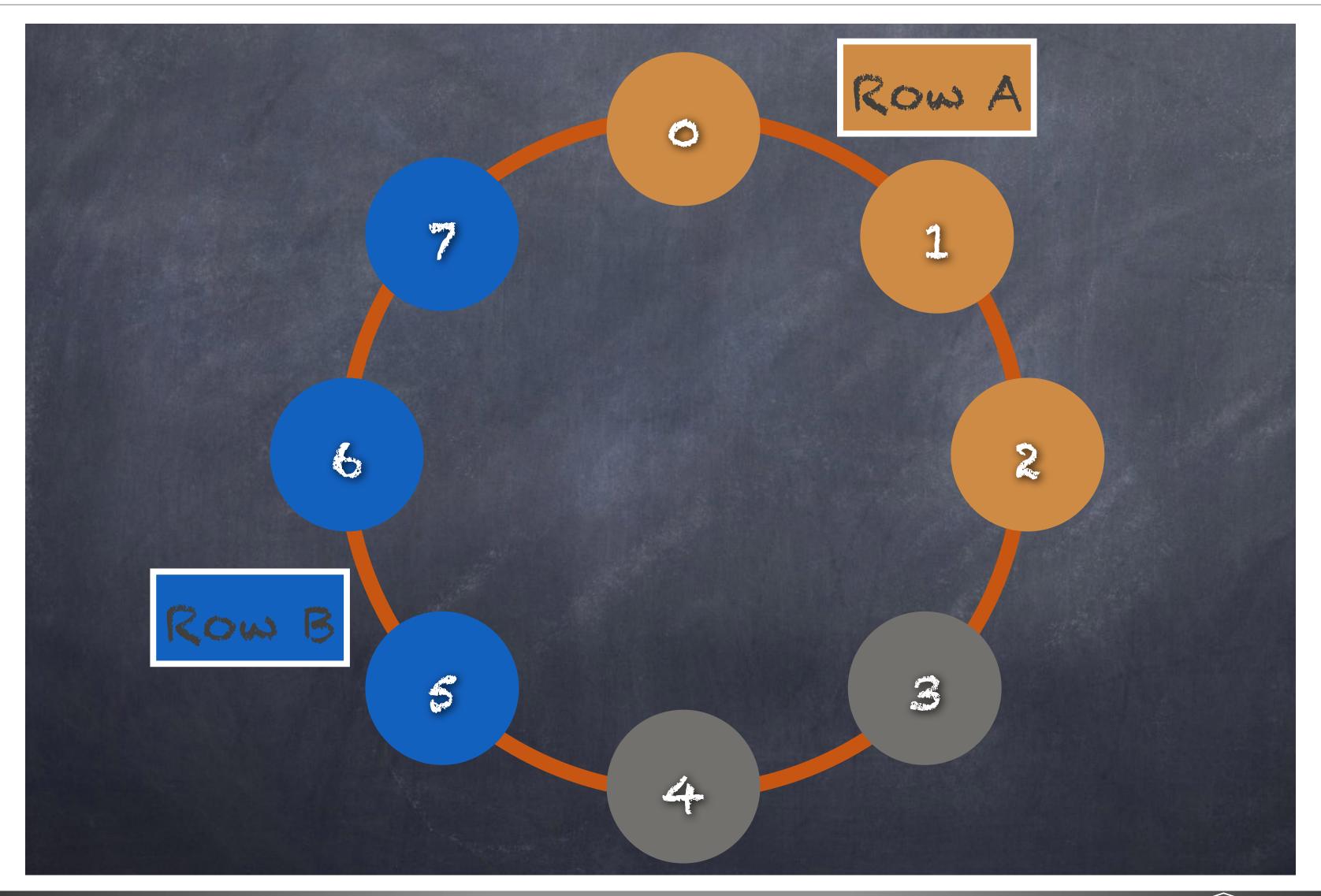
Relational Databases vs. Cassandra

| Relational Database | Cassandra |
|---|--|
| Handles moderate incoming data velocity | Handles high incoming data velocity |
| Data arriving from one/few locations | Data arriving from many locations |
| Manages primarily structured data | Manages all types of data |
| Supports complex/nested transactions | Supports simple transactions |
| Single points of failure with failover | No single points of failure; constant uptime |
| Supports moderate data volumes | Supports very high data volumes |
| Centralized deployments | Decentralized deployments |
| Data written in mostly one location | Data written in many locations |
| Supports read scalability (with consistency sacrifices) | Supports read and write scalability |
| Deployed in vertical scale up fashion | Deployed in horizontal scale out fashion |

[DataStax]



Cassandra: Replication



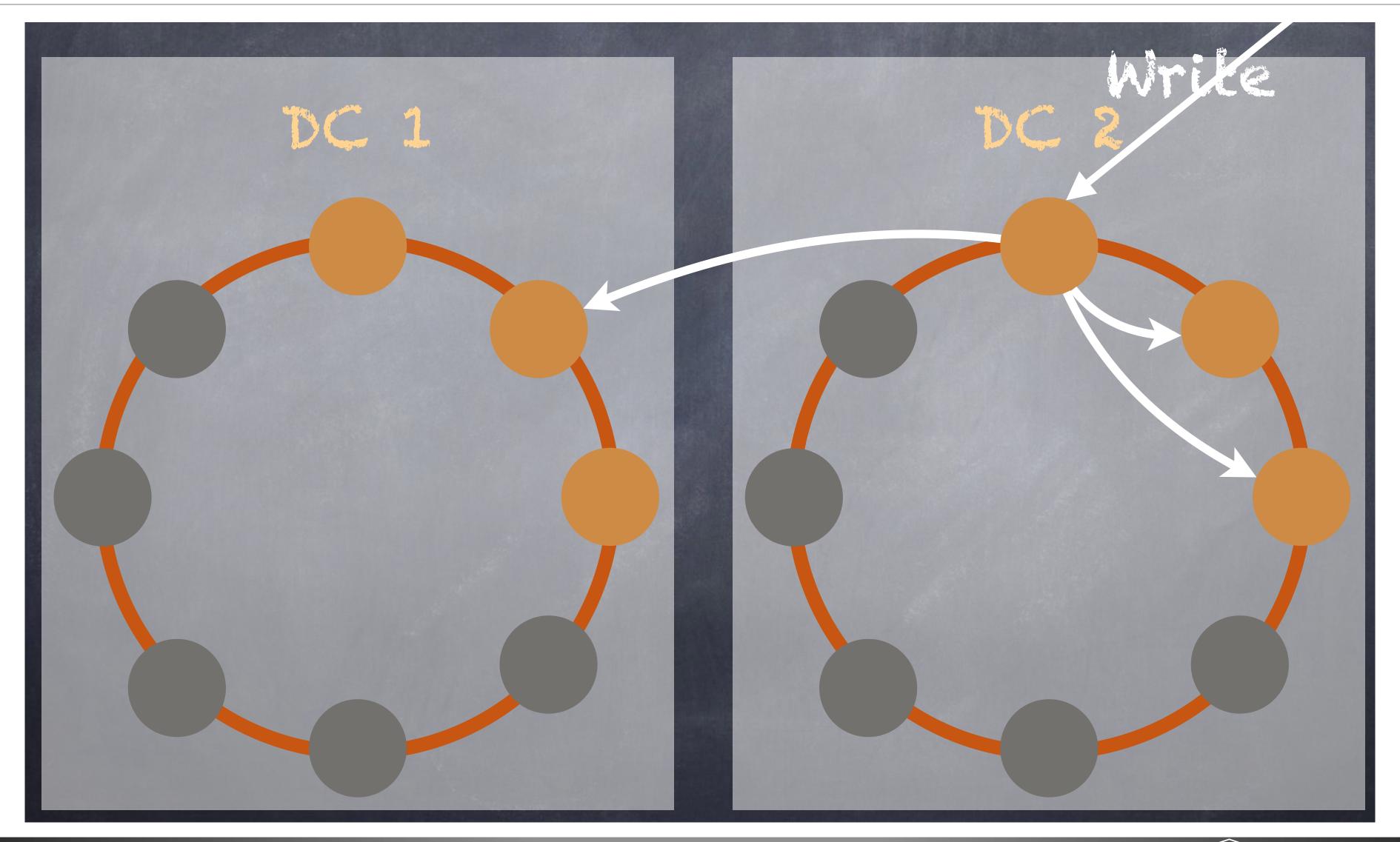
[R. Stupp]

Cassandra: Consistency Levels

- Data is always replicated according to replication factors
- Consistency Levels: ANY (only writes), ONE, LOCAL_ONE, QUORUM, LOCAL_QUORUM
- Consistency levels defines how many replicas must fulfill the request
- LOCAL_* are local to the data center, others go across data centers
- quorum = (sum-of-replication-factors / 2) + 1
 - Each data center may have its own replication factor
- ANY provides lowest consistency but highest availability
- ALL provides the highest consistency and lowest availability (not recommended)

[R. Stupp]

Multiple Data Center Replication



[R. Stupp]

Assignment 4

- Work on Data Integration and Data Fusion
- Integrate travel datasets from different institutions (UN World Tourism Office, World Bank, OECD)
 - Integrate information with population
- Record Matching:
 - Which countries are the same?
- Data Fusion:
 - The receipts/expenditures
 - Country names

<u>NewSQL</u>

A. Pavlo



Recent History in Databases

- Early 2000s: Commercial DBs dominated, Open-source DBs missing features
- Mid 2000s: MySQL adopted by web companies
- Late 2000s: NoSQL dos scale horizontally out of the box
- Early 2010s: New DBMSs that can scale across multiple machines natively and provide ACID guarantees





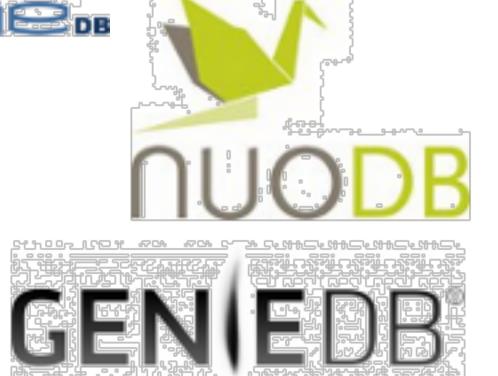






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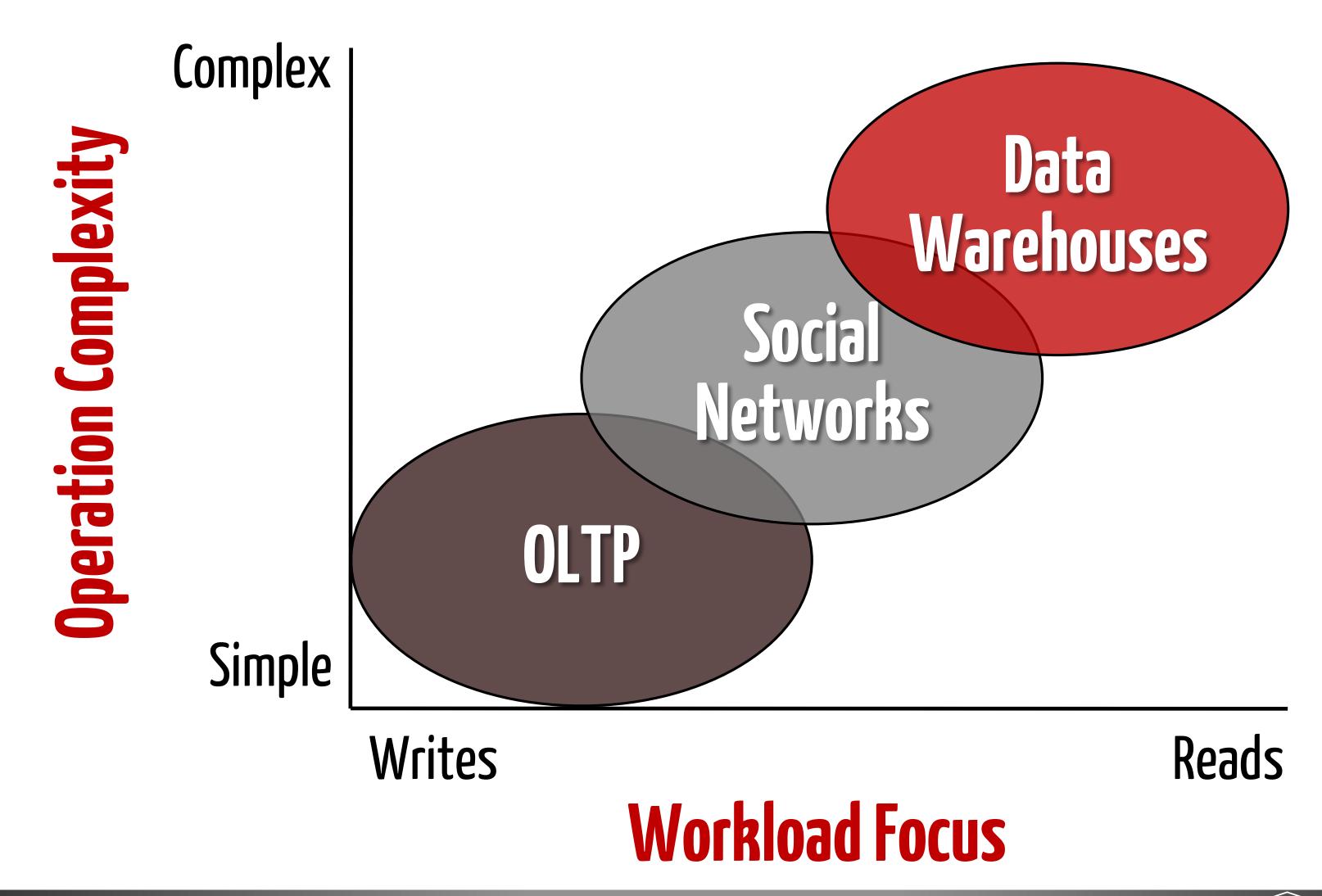


NewSQL

- 451 Group's Definition:
 - A DBMS that delivers the scalability and flexibility promised by NoSQL while retaining the support for SQL queries and/or ACID, or to improve performance for appropriate workloads.
- Stonebraker's Definition:
 - SQL as the primary interface
 - ACID support for transactions
 - Non-locking concurrency control
 - High per-node performance
 - Parallel, shared-nothing architecture

[A. Pavlo]

OLTP Workload



[A. Pavlo]

Ideal OLTP System

- Main Memory Only
- No Multi-processor Overhead
- High Scalability
- High Availability
- Autonomic Configuration

What's Really New with NewSQL?

A. Pavlo & M. Aslett



The Official Ten-Year Retrospective of NewSQL

A. Pavlo



Three Types of NewSQL Systems

- New Architectures
- Transparent Sharding Middleware
- Database-as-a-Service

What went wrong?

- Selling an OLTP Database System is hard
- Startup cost of a relational system is harder than NoSQL
- Existing DBMS Systems (MySQL, postgresql) are Good
- Cloud Disruption
 - Can't sell on-premises
 - Can't complete on cost with cloud vendors
- Lack of Open Source

Northern Illinois University

NewSQL is dead, Long live Distributed SQL

- E.g., Cockroach
- Core concepts are similar to earlier systems

Spanner: Google's Globally-Distributed Database

J. C. Corbett et al.

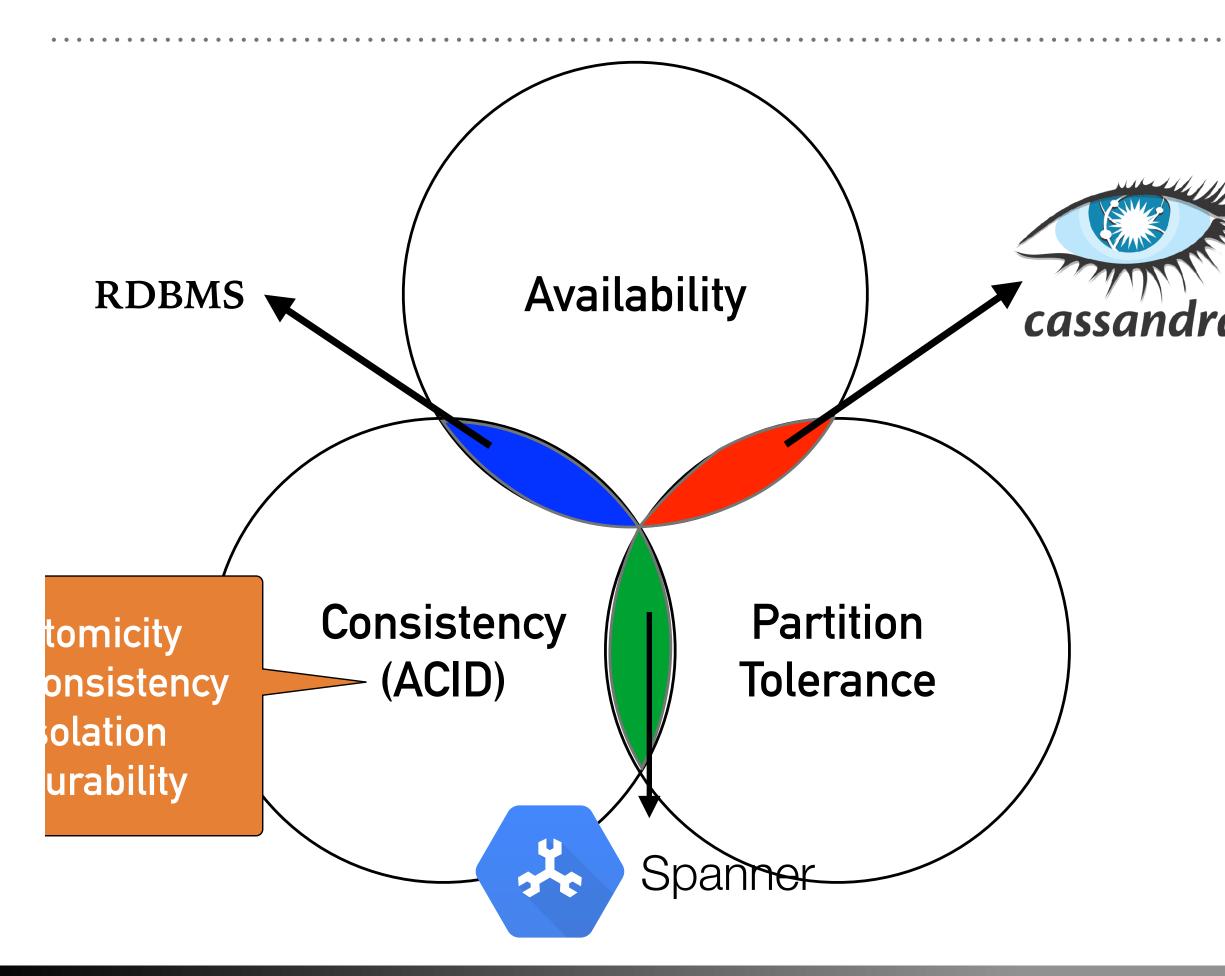


Spanner Overview

- Focus on scaling databases focused on OLTP (not OLAP)
- Since OLTP, focus is on sharding rows
- Tries to satisfy CAP (which is impossible per CAP Theorem) by not worrying about 100% availability
- External consistency using multi-version concurrency control through timestamps
- ACID is important
- Structured: universe with zones with zone masters and then spans with span masters
- SQL-like (updates allow SQL to be used with Spanner)

Spanner and the CAP Theorem

3H AVAILABILITY: CAP THEOREM AND CASSANDRA



- Which type of system is Spanner?
 - C: consistency, which implies a single value for shared data
 - A: 100% availability, for both reads and updates
 - P: tolerance to network partitions
- Which two?
 - CA: close, but not totally available
 - So actually CP

Spanner Server Organization

placement driver universemaster Zone 1 Zone 2 Zone N zonemaster zonemaster zonemaster location location location proxy proxy proxy spanserver spanserver spanserver

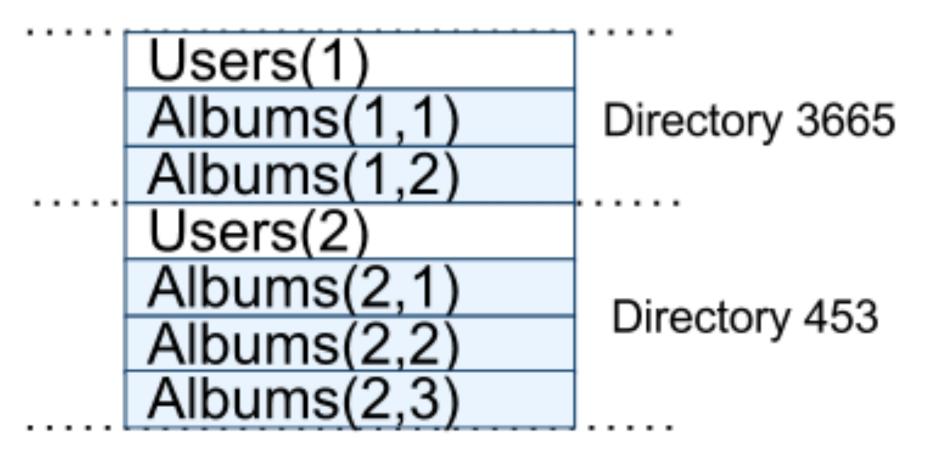
[Corbett et al., 2012]



Interleaved Schema

```
CREATE TABLE Users {
   uid INT64 NOT NULL, email STRING
} PRIMARY KEY (uid), DIRECTORY;

CREATE TABLE Albums {
   uid INT64 NOT NULL, aid INT64 NOT NULL,
   name STRING
} PRIMARY KEY (uid, aid),
   INTERLEAVE IN PARENT Users ON DELETE CASCADE;
```



[Corbett et al., 2012]

External Consistency

- Traditional DB solution: two-phase locking—no writes while client reads
- "The system behaves as if all transactions were executed sequentially, even though Spanner actually runs them across multiple servers (and possibly in multiple datacenters) for higher performance and availability" [Google]
- Semantically indistinguishable from a single-machine database
- Uses multi-version concurrency control (MVCC) using timestamps
- Spanner uses TrueTime to generate monotonically increasing timestamps across all nodes of the system

TrueTime

- API to try to keep computers on a globally-consistent clock
- Uses GPS and Atomic Clocks!
- Time masters per datacenter (usually with GPS)
- Each machine runs a timeslave daemon
- Armageddon masters have atomic clocks
- API:

| 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 | |

[Corbett et al., 2012]

Concurrency Control

- Use TrueTime to implement concurrency control
- Types of reads and writes:

| | Timestamp | Concurrency | |
|--|------------|-------------|--|
| Operation | Discussion | Control | Replica Required |
| Read-Write Transaction | § 4.1.2 | pessimistic | leader |
| Read-Only Transaction | § 4.1.4 | lock-free | leader for timestamp; any for read, subject to § 4.1.3 |
| Snapshot Read, client-provided timestamp | | lock-free | any, subject to § 4.1.3 |
| Snapshot Read, client-provided bound | § 4.1.3 | lock-free | any, subject to § 4.1.3 |

Use Two-Phase Commits (2PC)

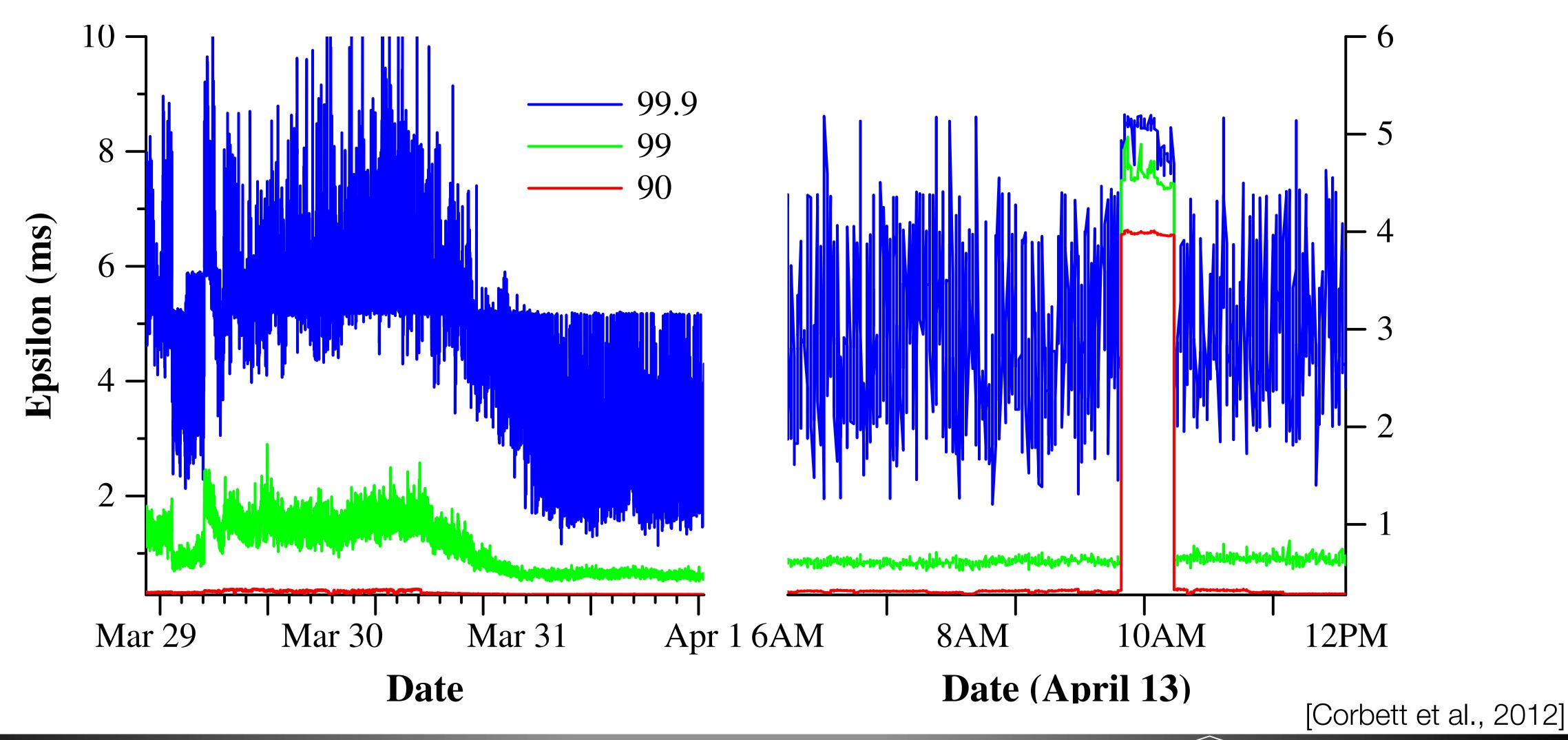
Two-Phase Commit Scalability

| | latency (ms) | | |
|--------------|------------------|------------------|--|
| participants | mean | 99th percentile | |
| 1 | 17.0 ±1.4 | 75.0 ±34.9 | |
| 2 | 24.5 ± 2.5 | 87.6 ± 35.9 | |
| 5 | 31.5 ±6.2 | 104.5 ± 52.2 | |
| 10 | 30.0 ± 3.7 | 95.6 ±25.4 | |
| 25 | 35.5 ±5.6 | 100.4 ± 42.7 | |
| 50 | 42.7 ± 4.1 | 93.7 ±22.9 | |
| 100 | 71.4 ± 7.6 | 131.2 ± 17.6 | |
| 200 | 150.5 ± 11.0 | 320.3 ± 35.1 | |

[Corbett et al., 2012]



Distribution of TrueTime Epsilons



F1: A Distributed SQL Database That Scales

- J. Shute, R. Vingralek, B. Samwel, B. Handy, C. Whipkey,
- E. Rollins, M. Oancea, K. Littlefield, D. Menestrina, S. Ellner,
- J. Cieslewicz, I. Rae, T. Stancescu, and H. Apte

F1: OLTP and OLAP Together

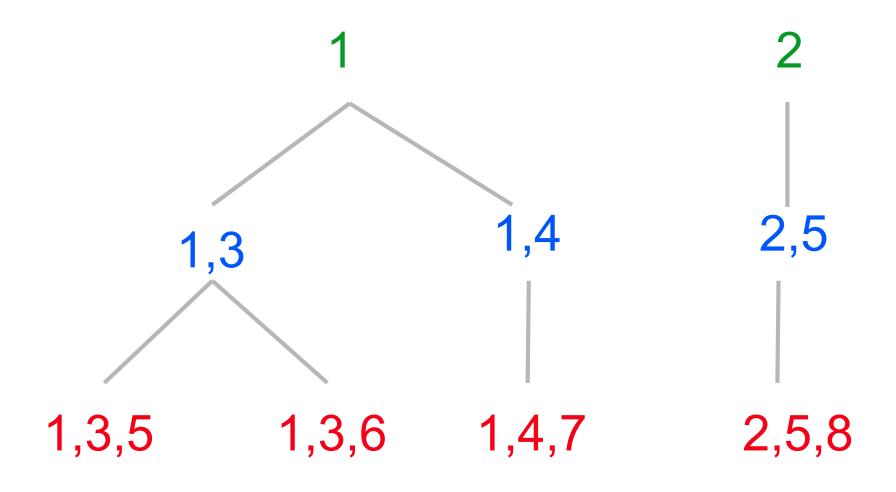
- Distributed data storage: data is not stored at one central location
- Need to keep data and schemas in sync
- Hierarchical schemas keep data that is likely to be accessed at the same time together
- Optimistic Transactions: Long reads that keep track of timestamps and don't lock the database until the write happens
- Change History: Keep track of history with database, also helps with caching
- DIY Object-Relational Mapping: don't automatically join or implicitly traverse relationships
- Protocol buffers as a way to store application data without translation + support for queries

Hierarchical Schema

Explicit table hierarchies. Example:

- Customer (root table): PK (Customerld)
- Campaign (child): PK (Customerld, CampaignId)
- AdGroup (child): PK (Customerld, CampaignId, AdGroupId)

Rows and PKs



Storage Layout

```
Customer (1)
Campaign (1,3)
AdGroup (1,3,5)
AdGroup (1,3,6)
Campaign (1,4)
AdGroup (1,4,7)
Customer (2)
Campaign (2,5)
AdGroup (2,5,8)
```

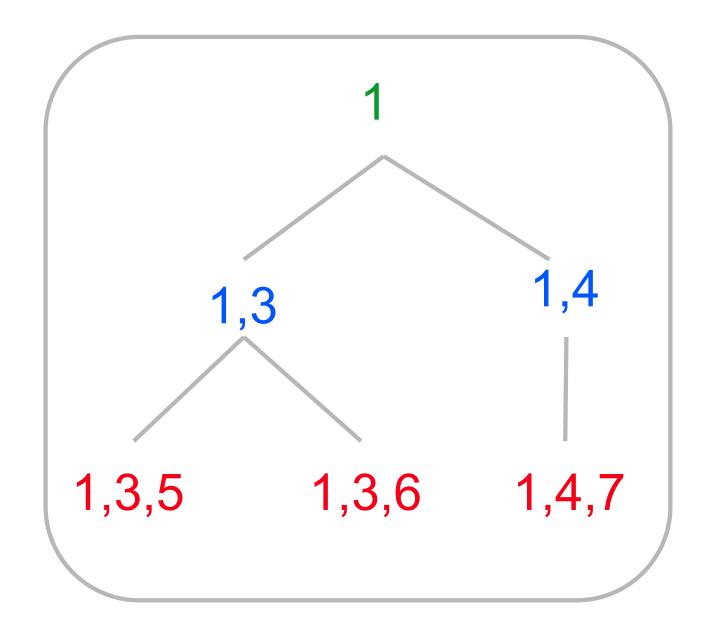
[Shute et al., 2012]

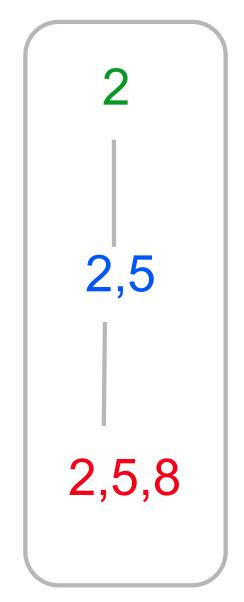


Clustered Storage

- Child rows under one root row form a cluster
- Cluster stored on one machine (unless huge)
- Transactions within one cluster are most efficient
- Very efficient joins inside clusters (can merge with no sorting)

Rows and PKs





Storage Layout

```
Customer (1)
Campaign (1,3)
AdGroup (1,3,5)
AdGroup (1,3,6)
Campaign (1,4)
AdGroup (1,4,7)
Customer (2)
Campaign (2,5)
AdGroup (2,5,8)
```

[Shute et al., 2012]



F1 Notes

- Schema changes: allow two different schemas
- Transaction types: Snapshot, Pessimistic, Optimistic
- Change History and application to caching
- Disk latency or network latency?

Discussion

Google Cloud Spanner

- https://cloud.google.com/spanner/
- Features:
 - Global Scale: thousands of nodes across regions / data centers
 - Fully Managed: replication and maintenance are automatic
 - Transactional Consistency: global transaction consistency
 - Relational Support: Schemas, ACID Transactions, SQL Queries
 - Security
 - Highly Available

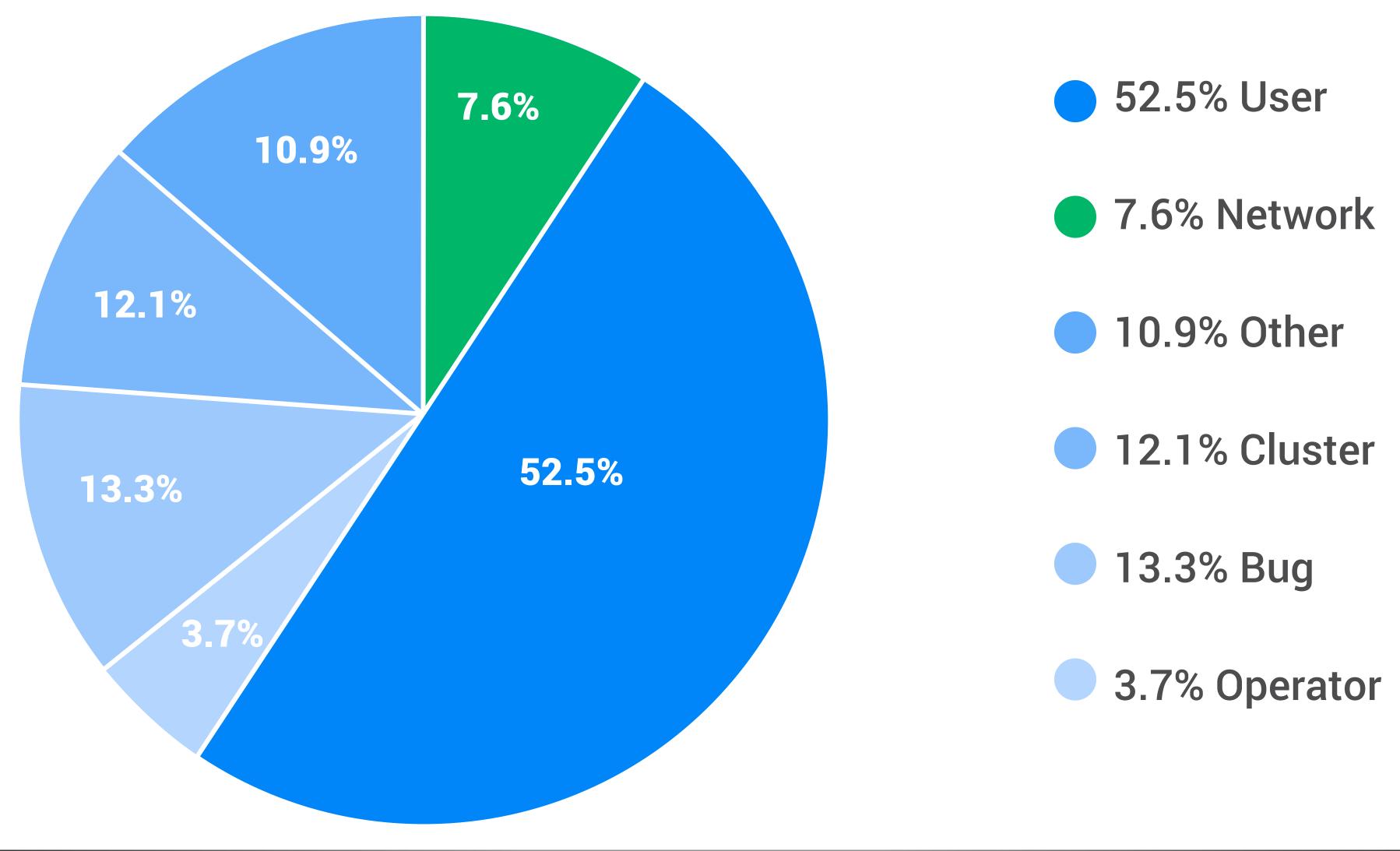
Google Cloud Spanner: NewSQL

Cloud Spanner: The best of the relational and NoSQL worlds

| | CLOUD SPANNER | TRADITIONAL RELATIONAL | TRADITIONAL NON-RELATIONAL |
|--------------|---------------|------------------------|----------------------------|
| Schema | Yes | Yes | × No |
| SQL | Yes | Yes | × No |
| Consistency | Strong | Strong | × Eventual |
| Availability | High | × Failover | High |
| Scalability | Horizontal | × Vertical | Horizontal |
| Replication | Automatic | Configurable | Configurable |

[https://cloud.google.com/spanner/]

Causes of Spanner Availability Incidents



Causes of Spanner Incidents

- User: overload or misconfiguration (specific to one user)
- Cluster: non-network problems, e.g. servers and power
- Operator: misconfiguration by people
- Bug: software error that caused some problem
- Other: most are one-offs
- Network: individual data centers/regions cut off and under-provisioned bandwidth, uni-directional traffic

[E. Brewer, 2017]

Spanner as "Effectively CA"

- Criteria for being "effectively CA"
 - 1. At a minimum it must have very high availability in practice (so that users can ignore exceptions), and
 - 2. As this is about partitions it should also have a low fraction of those outages due to partitions.
- Spanner meets both of these criteria
- Spanner relies on Google's network (private links between data centers)
- TrueTime helps create consistent snapshots, sometimes have a commit wait

[<u>E. Brewer</u>, 2017]

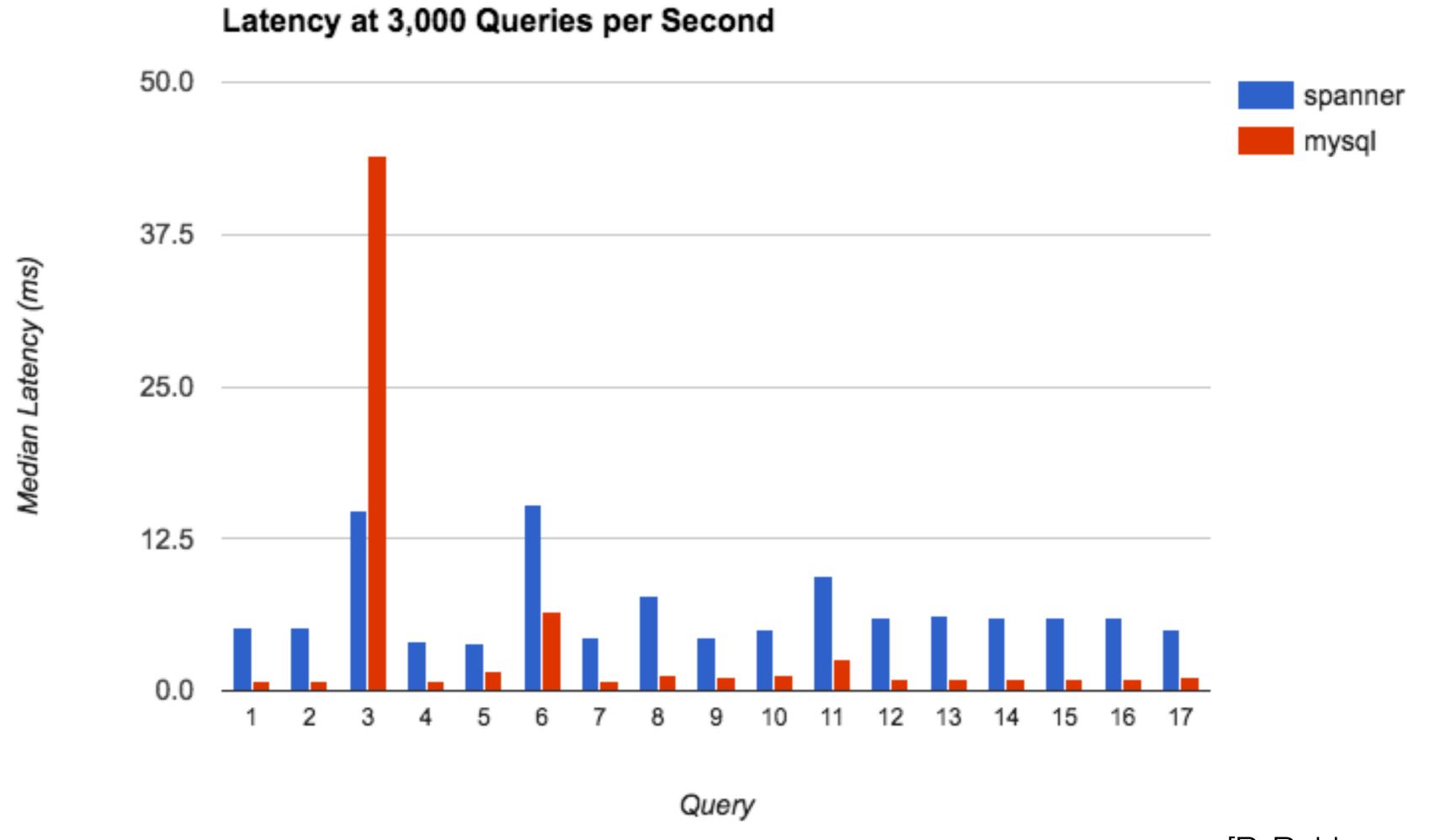
More Recent Tests: Spanner vs. MySQL

| | Frequency | Query | |
|----|-----------|---|--|
| 1 | 0.30% | INSERT INTO `terms` (`term`, `rank`, `set_id`, `last_modified`) VALUES (?,?,?,?),(?,?,?,?) | |
| 2 | 0.25% | INSERT INTO `terms` (`term`, `rank`, `set_id`, `last_modified`, `definition`) VALUES (?,?,?,?,?),(?,?,?,?,?),(?,?,?,?,?), | |
| 3 | 4.22% | INSERT INTO `terms` (`term`,`rank`,`set_id`,`last_modified`) VALUES (?,?,?,?) | |
| 4 | 1.88% | INSERT INTO `terms` (`term`,`rank`,`set_id`,`last_modified`,`definition`) VALUES (?,?,?,?,?) | |
| 5 | 3.28% | SELECT * FROM `terms` WHERE (`is_deleted` = 0) AND (`set_id` IN (??)) AND (`rank` IN (0,1,2,3)) AND (`term` != ") | |
| 6 | 14.13% | SELECT `set_id`, COUNT(*) FROM `terms` WHERE (`is_deleted` = 0) AND (`set_id` = ?) GROUP BY `set_id` | |
| 7 | 12.56% | SELECT * FROM `terms` WHERE (`id` = ?) | |
| 8 | 0.49% | SELECT * FROM `terms` WHERE (`id` IN (??) AND `set_id` IN (??)) | |
| 9 | 4.11% | SELECT `id`, `set_id` FROM `terms` WHERE (`set_id` = ?) LIMIT 20000 | |
| 10 | 0.43% | SELECT `id`, `set_id` FROM `terms` WHERE (`set_id` IN (??)) LIMIT 20000 | |
| 11 | 0.59% | SELECT * FROM `terms` WHERE (`id` IN (??)) | |
| 12 | 36.76% | SELECT * FROM `terms` WHERE (`set_id` = ?) | |
| 13 | 0.61% | SELECT * FROM `terms` WHERE (`set_id` IN (??)) | |
| 14 | 6.10% | UPDATE `terms` SET `definition`=?, `last_modified`=? WHERE `id`=? AND `set_id`=? | |
| 15 | 0.33% | UPDATE `terms` SET `is_deleted`=?, `last_modified`=? WHERE `id` IN (??) AND `set_id`=?? | |
| 16 | 12.56% | UPDATE `terms` SET `rank`=?, `last_modified`=? WHERE `id`=? AND `set_id`=? | |
| 17 | 1.06% | UPDATE `terms` SET `word`=?, `last_modified`=? WHERE `id`=? AND `set_id`=? | |
| 18 | 0.32% | UPDATE `terms` SET `definition`=?, `word`=?, `last_modified`=? WHERE `id`=? AND `set_id`=? | |

[P. Bakkum and D. Cepeda, 2017]

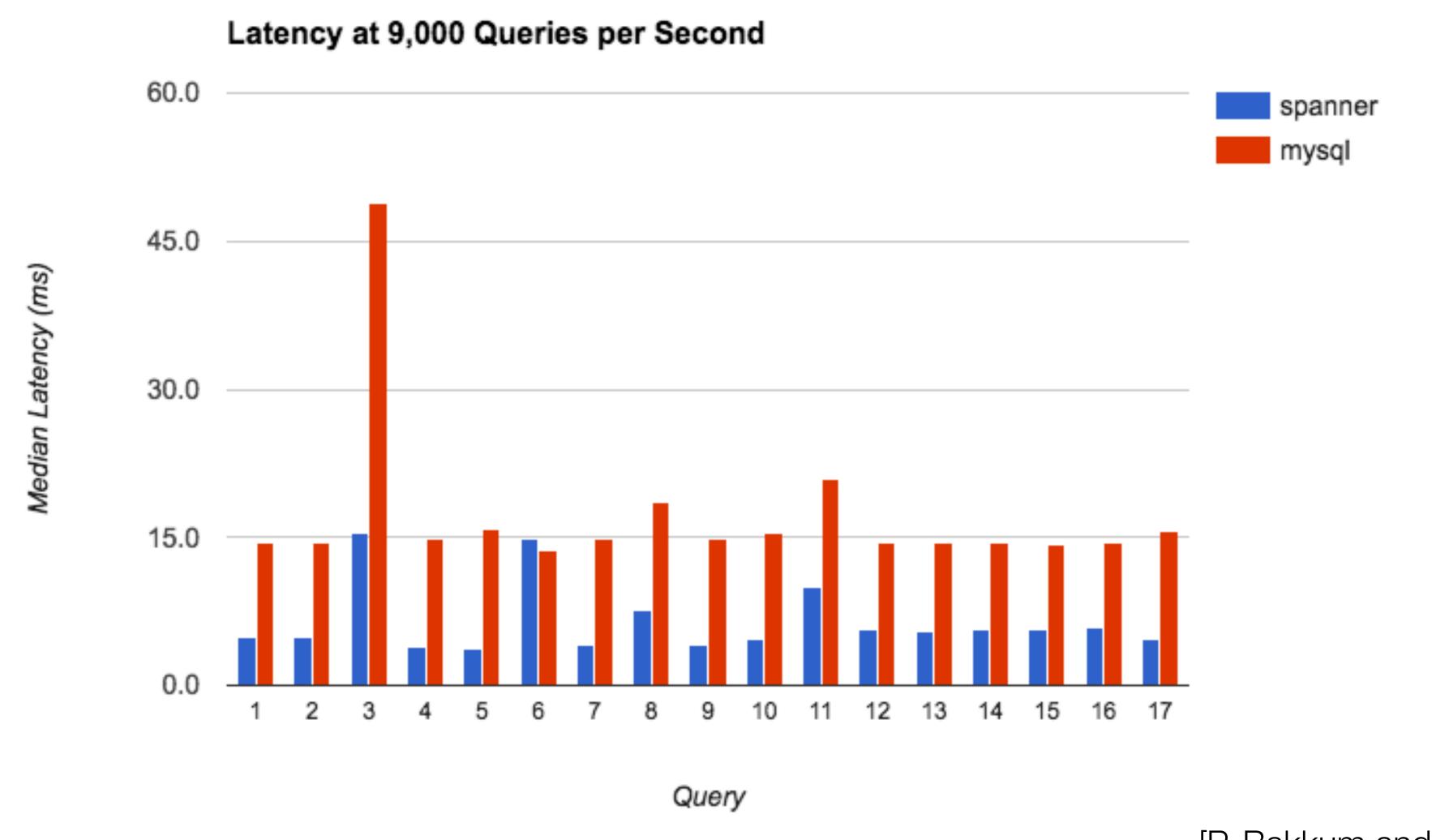


Latency: Spanner vs. MySQL





Latency: Spanner vs. MySQL



Throughput: Spanner vs. MySQL

Median Latency as Throughput Increases MySQL (median) ---- spanner 9 nodes (median) — spanner 15 nodes (median) spanner 30 nodes (median) 4000 6000 8000 10000 20000

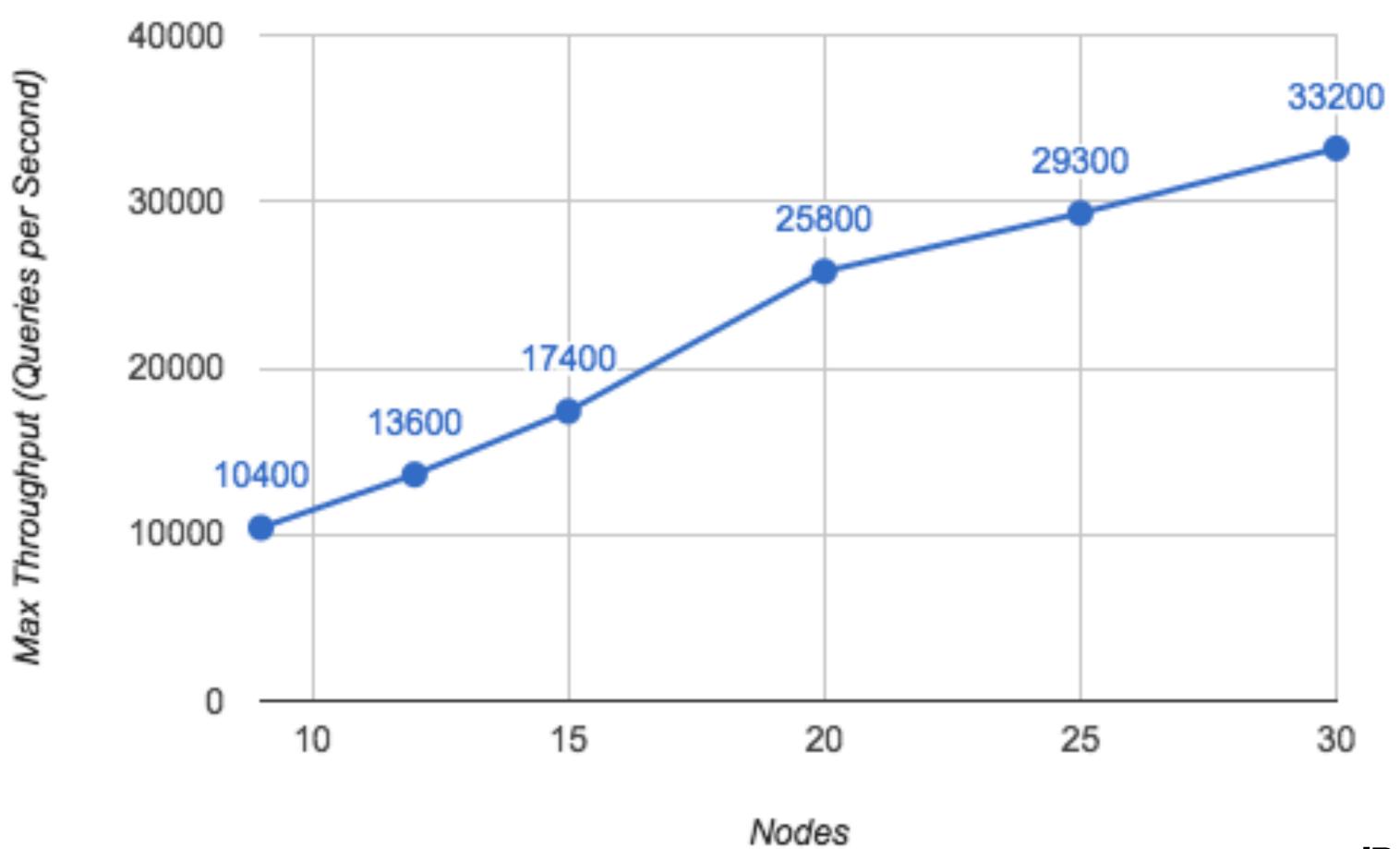
Throughput (queries per second)

[P. Bakkum and D. Cepeda, 2017]



Max Throughput vs. Nodes

Max Throughput vs Nodes



[P. Bakkum and D. Cepeda, 2017]



Spanner: Latency vs. Nodes

Latency at 3000 QPS vs Nodes

