Advanced Data Management (CSCI 490/680)

Scalable Databases

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Parallel DB Architecture: Shared Nothing



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[Hellerstein et al., Architecture of a Database System]









Sharding













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 handly 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







Problems with Relational Databases

	1
ID: 100	

Customer: Ann

Line Items:

0321293533	2	\$48	\$96
0321601912	1	\$39	\$39
0131495054	1	\$51	\$51

Payment Details:

Card: Amex **CC Number:** 12345 Expiry: 04/2001











Horizontal Partitioning vs. Vertical Partitioning

Vertical Partitions

VP1

VP2

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

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

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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









CAP Theorem



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7

Cassandra and CAP









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









NoSQL: Column Stores

- Instead of having rows grouped/sharded, we group columns
- ... or families of columns
- Put similar columns together
- Examples: Cassandra, HBase











Relational Databases vs. Cassandra

Relational Database

Handles moderate incoming data velocity

Data arriving from one/few locations

Manages primarily structured data

Supports complex/nested transactions

Single points of failure with failover

Supports moderate data volumes

Centralized deployments

Data written in mostly one location

Supports read scalability (with consistency sacrifices)

Deployed in vertical scale up fashion

Cassandra
Handles high incoming data velocity
Data arriving from many locations
Manages all types of data
Supports simple transactions
No single points of failure; constant uptime
Supports very high data volumes
Decentralized deployments
Data written in many locations
Supports read and write scalability
Deployed in horizontal scale out fashion
[Data





Cassandra: Replication











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)









Multiple Data Center Replication



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14

Reading Response

• <u>Spanner: Google's Globally-Distributed Database</u>





<u>Assignment 4</u>

- World Education Data
- Collected/collated by UNESCO, World Bank, and OECD
- Transform World Bank Data
- Impute missing year data
- Integrate teacher and student numbers
- Fuse three datasets
- Think about how to integrate based on country





<u>NewSQL</u>

A. Pavlo





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



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- 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





20

Spanner Server Organization



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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;









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	
TT.now()	
TT.after(t)	true
TT.before(t)	true if

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Returns

Tinterval: [*earliest*, *latest*]

e if t has definitely passed

f t has definitely not arrived









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 $\S 4.1.3$
Snapshot Read, client-provided timestamp		lock-free	any, subject to $\S 4.1.3$
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	

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[Corbett et al., 2012]











Distribution of TrueTime Epsilons



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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 (CustomerId)
- Campaign (child): PK (CustomerId, CampaignId)

Rows and PKs



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```
• AdGroup (child): PK (CustomerId, CampaignId, AdGroupId)
```

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)











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



Rows and PKs

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Very efficient joins inside clusters (can merge with no sorting)

Storage Layout





[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?

nt schemas mistic, Optimistic caching







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Discussion







Google Cloud Spanner

- <u>https://cloud.google.com/spanner/</u>
- 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	X No
SQL	Yes	Yes	X No
Consistency	Strong	Strong	× Eventual
Availability	High	× Failover	High
Scalability	Horizontal	× Vertical	Horizontal
Replication	Automatic	🗘 Configurable	🗘 Configurable

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

Rely on Strong Consistency, Scale, and Performalities University









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









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









More Recent Tests: Spanner vs. MySQL

	Frequency	Query
1	0.30%	INSERT INTO `terms` (`term`, `rank`,
2	0.25%	INSERT INTO `terms` (`term`, `rank`,
3	4.22%	INSERT INTO `terms` (`term`,`rank`,`
4	1.88%	INSERT INTO `terms` (`term`,`rank`,`
5	3.28%	SELECT * FROM `terms` WHERE (`i
6	14.13%	SELECT `set_id`, COUNT(*) FROM `
7	12.56%	SELECT * FROM `terms` WHERE (`i
8	0.49%	SELECT * FROM `terms` WHERE (`i
9	4.11%	SELECT `id`, `set_id` FROM `terms`
10	0.43%	SELECT `id`, `set_id` FROM `terms`
11	0.59%	SELECT * FROM `terms` WHERE (`i
12	36.76%	SELECT * FROM `terms` WHERE (`s
13	0.61%	SELECT * FROM `terms` WHERE (`s
14	6.10%	UPDATE `terms` SET `definition`=?, `
15	0.33%	UPDATE `terms` SET `is_deleted`=?,
16	12.56%	UPDATE `terms` SET `rank`=?, `last_
17	1.06%	UPDATE `terms` SET `word`=?, `last
18	0.32%	UPDATE `terms` SET `definition`=?, `

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, `set_id`, `last_modified`) VALUES (?,?,?,?),(?,?,?)

, `set_id`, `last_modified`, `definition`) VALUES (?,?,?,?,?),(?,?,?,?),(?,?,?,?),...

`set_id`,`last_modified`) VALUES (?,?,?,?)

`set_id`,`last_modified`,`definition`) VALUES (?,?,?,?,?)

is_deleted` = 0) AND (`set_id` IN (??)) AND (`rank` IN (0,1,2,3)) AND (`term` != ")

`terms` WHERE (`is_deleted` = 0) AND (`set_id` = ?) GROUP BY `set_id`

`id` = ?)

`id` IN (??) AND `set_id` IN (??))

WHERE (`set_id` = ?) LIMIT 20000

WHERE (`set_id` IN (??)) LIMIT 20000

`id` IN (??))

`set_id` = ?)

`set_id` IN (??))

`last_modified`=? WHERE `id`=? AND `set_id`=?

, `last_modified`=? WHERE `id` IN (??) AND `set_id`=??

_modified`=? WHERE `id`=? AND `set_id`=?

_modified`=? WHERE `id`=? AND `set_id`=?

`word`=?, `last_modified`=? WHERE `id`=? AND `set_id`=?

[P. Bakkum and D. Cepeda, 2017]



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Latency: Spanner vs. MySQL



Latency at 3,000 Queries per Second

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Median Latency (ms)



Query







40

Latency: Spanner vs. MySQL





Median Latency (ms)

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Query





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41

Throughput: Spanner vs. MySQL











Max Throughput vs. Nodes

Max Throughput vs Nodes



Max Throughput (Queries per Second)

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Spanner: Latency vs. Nodes

Latency at 3000 QPS vs Nodes



Latencies @ 3000 QPS

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44