CSE 444: Database Internals

Lectures 26 NoSQL: Key Value Stores

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References

- Scalable SQL and NoSQL Data Stores, Rick Cattell, SIGMOD Record, December 2010 (Vol. 39, No. 4)
- Dynamo: Amazon's Highly Available Key-value Store. By Giuseppe DeCandia et. al. SOSP 2007.
- Online documentation: Amazon DynamoDB.

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NoSQL Motivation

- Originally motivated by Web 2.0 applications
- Goal is to scale simple OLTP-style workloads to thousands or millions of users
- · Users are doing both updates and reads

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Why NoSQL as the Solution?

- Hard to scale transactions
 - Need to partition the database across multiple machines
 - If a transaction touches one machine, life is good
 - If a transaction touches multiple machines, ACID becomes extremely expensive! Need two-phase commit
- Replication
 - Replication can help to increase throughput and lower latency
 - Create multiple copies of each database partition
 - Spread queries across these replicas
 - Easy for reads but writes, once again, become expensive!

NoSQL Key Feature Decisions

- Want a data management system that is
 - Elastic and highly scalable
 - Flexible (different records have different schemas)
- To achieve above goals, willing to give up
 Complex queries: e.g., give up on joins
 - Complex queries: e.g., give up on jo
 - Multi-object transactions
 - ACID guarantees: e.g., eventual consistency is OK
 - Eventual consistency: If updates stop, all replicas will converge to the same state and all reads will return the same value

All updates

eventually reach

- BASE (Basically Available, Soft state, Eventually consistent)
- Not all NoSQL systems give up all these properties

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Cattell, SIGMOD Record 2010

NoSQL

"Not Only SQL" or "Not Relational". Six key features:

- 1. Scale horizontally "simple operations"
- 2. Replicate/distribute data over many servers
- 3. Simple call level interface (contrast w/ SQL)
- 4. Weaker concurrency model than ACID
- 5. Efficient use of distributed indexes and RAM
- 6. Flexible schema

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Data Models

- Tuple = row in a relational db
- Key-value = records identified with keys have values that are opaque blobs
- Extensible record = families of attributes have a schema, but new attributes may be added
- Document = nested values, extensible records (XML, JSON, protobuf, attribute-value pairs)

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Cattell, SIGMOD Record 2010

Different Types of NoSQL

Taxonomy based on data models:



- · Key-value stores
 - e.g., Project Voldemort, Memcached, Redis
- Extensible Record Stores
 - e.g., HBase, Cassandra, PNUTS
- · Document stores
 - e.g., SimpleDB, CouchDB, MongoDB, Couchbase
- · Most recently: Graph databases
- New types of RDBMSs.. not really NoSQL
 - Next lecture

Key-Value Store: Dynamo

- Dynamo: Amazon's Highly Available Keyvalue Store. By Giuseppe DeCandia et. al. SOSP 2007.
- · Main observation:
 - "There are many services on Amazon's platform that only need **primary-key access** to a data store."
 - Best seller lists, shopping carts, customer preferences, session management, sales rank, product catalog

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Basic Features

- · Data model: (key,value) pairs
 - Values are binary objects (blobs)
 - No further schema
- Operations
 - Insert/delete/lookup by key
 - No operations across multiple data items
- Consistency
 - Replication with eventual consistency
 - Goal to NEVER reject any writes (bad for business)
 - Multiple versions with conflict resolution during reads

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Operations

- get(key)
 - Locates object replicas associated with key
 - Returns a single object
 - Or a list of objects with conflicting versions
 - Also returns a context
 - Context holds metadata including version
 - · Context is opaque to caller
- put(key, context, object)
 - Determines where replicas of object should be placed

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- Location depends on key value
- Data stored persistently including context

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Storage: Distributed Hash Table

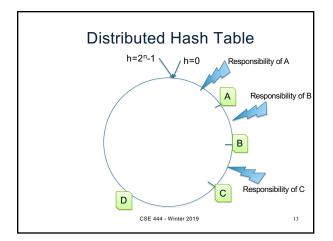
Implements a distributed storage

- Each key-value pair (k,v) is stored at some server h(k)
- API: write(k,v); read(k)

Use standard hash function: service key k by server h(k)

- Problem 1: a client knows only one server, doesn't know how to access h(k)
- Problem 2. if new server joins, then N → N+1, and the entire hash table needs to be reorganized
- Problem 3: we want replication, i.e. store the object at more than one server

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Distributed Hash Table Details

- This type of hashing called "consistent hashing"
- · Basic approach leads to load imbalance
 - Solution?

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Distributed Hash Table Details

- · This type of hashing called "consistent hashing"
- · Basic approach leads to load imbalance
 - Solution: Use V virtual nodes for each physical node
 - Virtual nodes provide better load balance
 - Nb of virtual nodes can vary based on capacity

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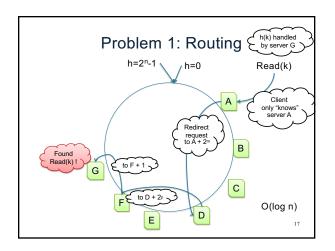
Problem 1: Routing

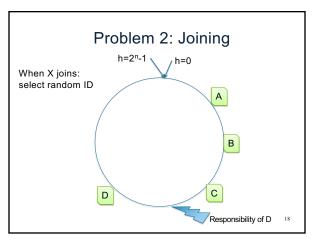
A client doesn't know server h(k), but some other server

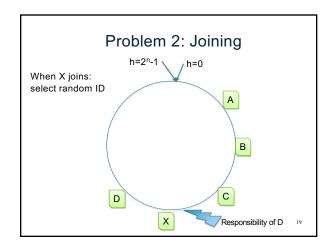
- Naive routing algorithm:
 - Each node knows its neighbors
 - Send message to nearest neighbor
 Hop-by-hop from there
 Obviously this is O(n), so no good
- · Better algorithm: "finger table"
 - Memorize locations of other nodes in the ring
 - a, a + 2, a + 4, a + 8, a + 16, ... a + 2ⁿ 1
 Send message to closest node to destination

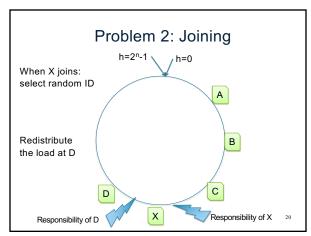
 - Hop-by-hop again: this is log(n)

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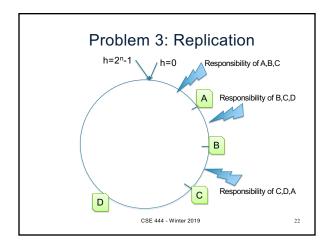




Problem 3: Replication

- Need to have some degree of replication to cope with node failures
- Let N=degree of replication
- Assign key k to h(k), h(k)+1, ..., h(k)+N-1

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Additional Dynamo Details

- Each key assigned to a coordinator
- · Coordinator responsible for replication
 - Replication skips virtual nodes that are not distinct physical nodes
- · Set of replicas for a key is its preference list
- · One-hope routing:
 - Each node knows preference list of each key
- "Sloppy quorum" replication
 - Each update creates a new version of an object
 - Vector clocks track causality between versions

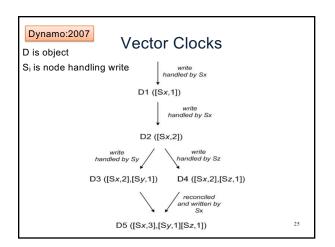
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Vector Clocks

- An extension of Multiversion Concurrency Control (MVCC) to multiple servers
- Standard MVCC: each data item X has a timestamp t: X₄, X₉, X₁₀, X₁₄, ..., X_t
- Vector Clocks: X has set of [server, timestamp] pairs X([s1,t1], [s2,t2],...)

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Vector Clocks: Example

- A client writes D1 at server SX:
 - D1 ([SX,1])
- Another client reads D1, writes back D2; also handled by server SX:

D2 ([SX,2]) (D1 garbage collected)

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Vector Clocks: Example

- A client writes D1 at server SX: D1 ([SX,1])
- Another client reads D1, writes back D2; also handled by server SX:

 $D2 \; ([SX,2]) \; \; (D1 \; garbage \; collected)$

 Another client reads D2, writes back D3; handled by server SY:

D3 ([SX,2], [SY,1])

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Vector Clocks: Example

- A client writes D1 at server SX:
 - D1 ([SX,1])
- Another client reads D1, writes back D2; also handled by server SX:

D2 ([SX,2]) (D1 garbage collected)

Another client reads D2, writes back D3; handled by server SY:

D3 ([SX,2], [SY,1])

 Another client reads D2, writes back D4; handled by server SZ:

D4 ([SX,2], [SZ,1])

Vector Clocks: Example

- A client writes D1 at server SX: D1 ([SX,1])
- Another client reads D1, writes back D2; also handled by server SX:

D2 ([SX,2]) (D1 garbage collected)

 Another client reads D2, writes back D3; handled by server SY:

D3 ([SX,2], [SY,1])

 Another client reads D2, writes back D4; handled by server SZ:

D4 ([SX,2], [SZ,1])

• Another client reads D3 and D4: CONFLICT!

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Vector Clocks: Meaning

- A data item D[(S1,v1),(S2,v2),...] means a value that represents version v1 for S1, version v2 for S2, etc.
- · If server Si updates D, then:
 - It must increment vi, if (Si, vi) exists
 - Otherwise, it must create a new entry (Si,1)

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Vector Clocks: Conflicts

- A data item D is an ancestor of D' if for all (S,v)∈D there exists (S,v')∈D' s.t. v ≤ v'
- Otherwise, D and D' are on parallel branches, and it means that they have a conflict that needs to be reconciled semantically

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(Sloppy) Quorum Read/Write

- · Parameters:
 - N = number of copies (replicas) of each object
 - R = minimum number of nodes that must participate in a successful read
 - W = minimum number of nodes that must participate in a successful write
- Quorum: R+W > N
- Sloppy Quorum (Dynamo): allow R+W ≤ N
 - Allow fewer than N to get better latency

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Operation Execution

- · Write operations
 - Initial request sent to coordinator
 - Coordinator generates vector clock & stores locally
 - Coordinator forwards new version to all N replicas
 - If at least W-1 < N-1 nodes respond then success!</p>
- Read operations
 - Initial request sent to coordinator
 - Coordinator requests data from all N replicas
 - Once gets R responses, returns data
- Sloppy quorum: Involve first N healthy nodes

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Amazon DynamoDB

Additional functionality:

- Both document and key-value store models
- Offers secondary indexes to enable queries over non-key attributes
- So can support selection and projection queries
- · Offers choice of eventual consistent vs strongly consistent read

Try Amazon DynamoDB

http://aws.amazon.com/dynamodb/

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Amazon DynamoDB Data Model

- · Tables containing Items
 - Items are described with attributes
 - One attribute must be the primary key
 - · Primary key can be a single partition key attribute
 - Or a pair of (partition key k1, sort key k2)
 - Items partitioned across nodes on k1
 - Sorted within the node on k2

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Amazon DynamoDB Querying

- · Selection and projection queries
 - Equality predicates on primary key
 - Must create secondary indexes to query other attributes. Also equality predicates
 - Can specify attributes to return (projection)
 - Can specify path notation for document attributes

Amazon DynamoDB Consistency

- · Eventually consistent read
 - "When you read data from a DynamoDB table, the response might not reflect the results of a recently completed write operation. The response might include some stale data. However, if you repeat your read request after a short time, the response should return the latest data."
- · Strongly consistent read
 - "When you request a strongly consistent read, DynamoDB returns a response with the most up-to-date data, reflecting the updates from all prior write operations that were successful. Note that a strongly consistent read might not be available in the case of a network delay or outage."

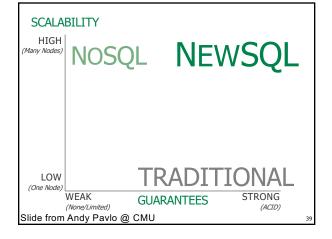
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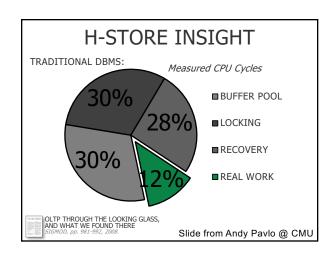
Lecture 27 NewSQL

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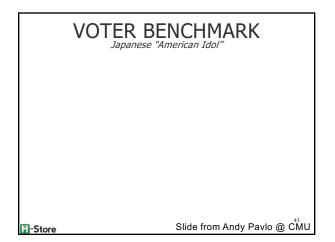
Some Popular NewSQL Systems

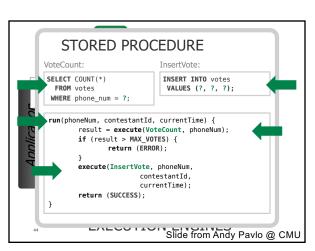
- H-Store
 - Research system from Brown U., MIT, CMU, and Yale
 - Commercialized as VoltDB
- Hekaton
 - Microsoft
- Fully integrated into SQL Server
- Spanner
 - Google

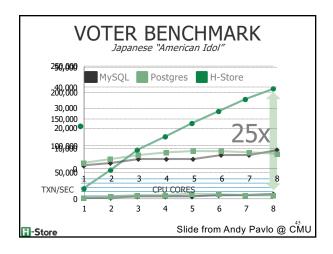


H-Store Key Ideas

- · Main-memory storage
 - Avoids disk IO costs / buffer pool costs
 - Durability through snapshots + cmd log
 - Replication
- · Serial execution
 - One database partition per thread on one core
 - Avoid overheads related to locking
- · All transactions are stored procedures
 - Command logging avoids heavy recovery overheads
- · Avoid distributed transactions
 - But when needed, run 2PC







Hekaton

- Focus: DBMS with large main memories and many core CPUs
- · Integrated with SQL Server
- · Key user-visible features
 - Simply declare a table "memory resident"
 - Hekaton tables are fully durable and transactional, though non-durable tables are also supported
 - Query can touch both Hekaton and regular tables

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Hekaton Key Details

- Idea: To increase transaction throughput must decrease number of instructions / transaction
- Main-memory DBMS
 - Optimize indexes for memory-resident data
 - Durability by logging and checkpointing records to external storage
- No partitioning
 - Any thread can touch any row of any table
- No locking
 - Uses a new MVCC method for isolation

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Hekaton More Details

- · Optimized stored procedures
 - Compile statements and stored procedures into customized, highly efficient machine code

Conclusion

- Many innovations recently in
 - Big data analytics
 - Transaction processing at very large scale
- Many more problems remain open
- This course teaches foundations
- Innovate with an open mind!