On Brewing Fresh Espresso: LinkedIn's Distributed Data Serving Platform

Motivation

- Better performance and horizontal scalability than traditional RDBMS.
- Better consistency, transactions, and schema support than NoSQL.
- Integration into LinkedIn's data ecosystem.

Data Model

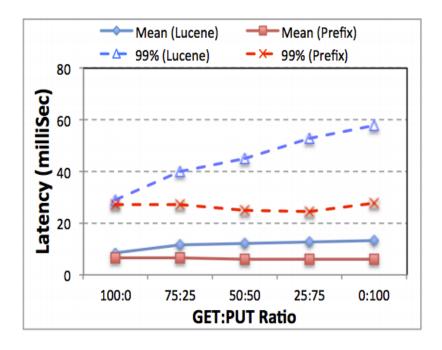
- Nested entities and independent entities.
- Relational
 - Documents the equivalent of rows
- Hierarchical
 - Document groups share same partitioning key, span tables, largest unit of transactions

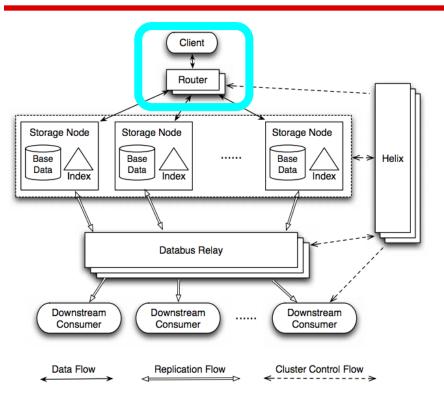
Secondary Indexes

- Allow for efficient lookup based on values other than the primary key.
- Local secondary indexes apply to one document group.
- Global secondary indexes apply across doc groups, implemented as derived tables.

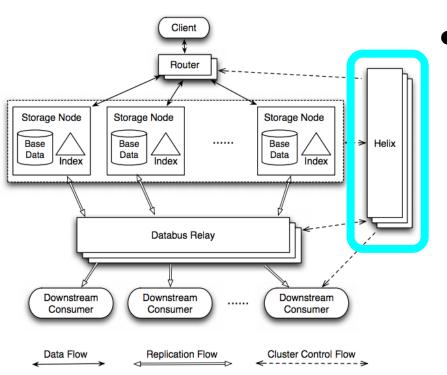
Secondary Indexes

- Lucene
 - Inverted index.
 - Log structured.
- Prefix
 - Inverted index, prefixed by the partition key.



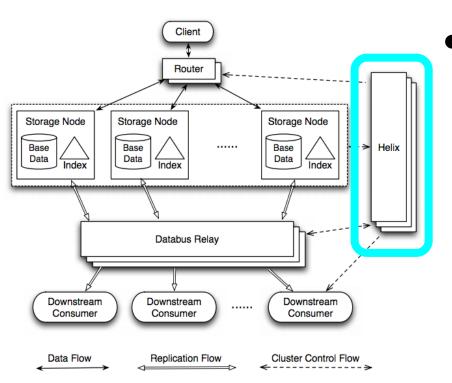


- Client submit requests via REST API.
- Router send request to appropriate node based on partitioning protocol.



Helix

- Cluster management system
- Assigns partitions

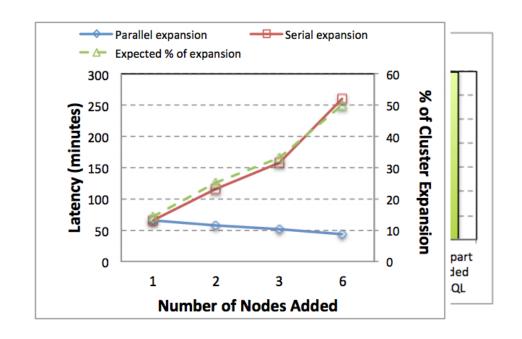


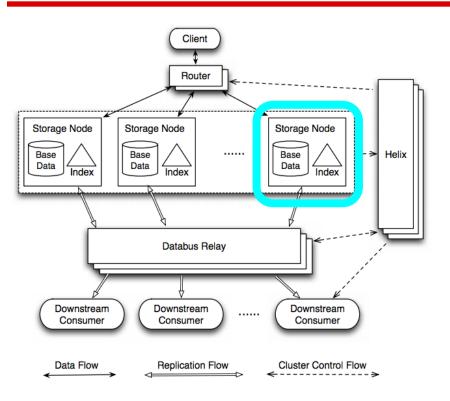
Fault tolerance

- When a master
 partition fails, a slave
 is promoted by Helix.
- Zookeeper heartbeat and performance metrics determine failure.

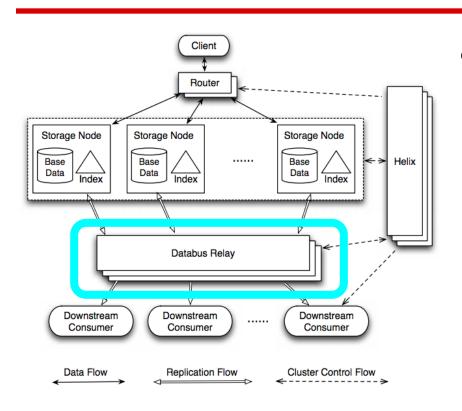
Overpartitioning

- Shard data into many more partitions than there are nodes.
- Eases failover/cluster expansion.





- Storage node
 - Stores partitions.
 - o Performs queries.
 - Maintains log.
 - Performs background tasks.



Databus

- Achieves replication via pub/sub
- Ensures timeline consistency
- Replicated for fault tolerance

Future Work

- Transactions across document groups.
- OLAP workloads.
- Multiple data center deployment.

Conclusion

- Espresso attempts to find a nice medium between traditional RDBMS and NoSQL.
- LinkedIn particularly emphasized operability
 ease of schema changes, horizontal scalability, etc.