

Data Infrastructure @ LinkedIn Sid Anand QCon London 2012

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About Me

Current Life...

LinkedIn

- Web / Software Engineering
 - Search, Network, and Analytics (SNA)
 - Distributed Data Systems (DDS)
 - > Me

Linked Inc.

In a Previous Life...

- > Netflix, Cloud Database Architect
- eBay, Web Development, Research Lab, & Search Engine

And Many Years Prior...

Studying Distributed Systems at Cornell University

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Our mission

Connect the world's professionals to make them more productive and successful





The world's largest professional network Over 60% of members are now international





Other Company Facts

• Headquartered in Mountain View, Calif., with offices around the world!

- As of December 31, 2011, LinkedIn has 2,116 full-time employees located around the world.
 - Currently around 650 people work in Engineering
 - 400 in Web/Software Engineering
 - Plan to add another 200 in 2012
 - 250 in Operations

*as of February 9, 2012 **as of September 30, 2011 ***as of December 31, 2011



Agenda

✓ Company Overview

- Architecture
 - Data Infrastructure Overview
 - Technology Spotlight
 - Oracle
 - Voldemort
 - DataBus
 - Kafka
- Q & A



Overview

- Our site runs primarily on Java, with some use of Scala for specific infrastructure
- What runs on Scala?
 - Network Graph Service
 - Kafka
- Most of our services run on Apache + Jetty



LinkedIn : Architecture





LinkedIn : Architecture



LinkedIn : Data Infrastructure Technologies

- Database Technologies
 - Oracle
 - Voldemort
 - Espresso
- Data Replication Technologies
 - Kafka
 - DataBus
- Search Technologies
 - Zoie real-time search and indexing with Lucene
 - Bobo faceted search library for Lucene

SenseiDB – fast, real-time, faceted, KV and full-text Search Engine and more



LinkedIn : Data Infrastructure Technologies

This talk will focus on a few of the key technologies below!

- Database Technologies
 - Oracle
 - Voldemort
 - Espresso A new K-V store under development
- Data Replication Technologies
 - Kafka
 - DataBus
- Search Technologies

- Zoie real-time search and indexing with Lucene
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LinkedIn Data Infrastructure Technologies

Oracle: Source of Truth for User-Provided Data



Oracle : Overview

Oracle

- All user-provided data is stored in Oracle our current source of truth
- About 50 Schemas running on tens of physical instances
- With our user base and traffic growing at an accelerating pace, so how do we scale Oracle for user-provided data?

Scaling Reads

- Oracle Slaves (c.f. DSC)
- Memcached
- Voldemort for key-value lookups

Scaling Writes

• Move to more expensive hardware **or** replace Oracle with something else



Scaling Oracle Reads using DSC

- DSC uses a token (e.g. cookie) to ensure that a reader always sees his or her own writes immediately
 - If I update my own status, it is okay if you don't see the change for a few minutes, but I have to see it immediately



Oracle : Overview – How DSC Works?

- When a user writes data to the master, the DSC token (for that data domain) is updated with a timestamp
- When the user reads data, we first attempt to read from a replica (a.k.a. slave) database
- If the data in the slave is older than the data in the DSC token, we read from the Master instead



LinkedIn Data Infrastructure Technologies

Voldemort: Highly-Available Distributed Data Store



Voldemort : Overview

- A distributed, persistent key-value store influenced by the AWS Dynamo paper
- Key Features of Dynamo
 - Highly Scalable, Available, and Performant
 - Achieves this via Tunable Consistency
 - For higher consistency, the user accepts lower availability, scalability, and performance, and vice-versa
 - Provides several self-healing mechanisms when data does become inconsistent
 - Read Repair
 - Repairs value for a key when the key is looked up/read
 - Hinted Handoff
 - > Buffers value for a key that wasn't successfully written, then writes it later
 - Anti-Entropy Repair
 - Scans the entire data set on a node and fixes it
 - Provides means to detect node failure and a means to recover from node failure
 - Failure Detection
 - Bootstrapping New Nodes

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Voldemort : Overview

API VectorClock<V> get (K key) put (K key, VectorClock<V> value) applyUpdate(UpdateAction action, int retries)

Voldemort-specific Features

- Implements a layered, pluggable architecture
- Each layer implements a common interface (c.f. API). This allows us to replace or remove implementations at any layer
 - Pluggable data storage layer
 BDB JE, Custom RO storage, etc...
 - Pluggable routing supports
 - Single or Multi-datacenter routing





Voldemort : Overview

Voldemort-specific Features

- Supports Fat client or Fat Server
 - Repair Mechanism + Failure Detector + Routing can run on server or client
- LinkedIn currently runs Fat Client, but we would like to move this to a Fat Server Model





Layered, Pluggable Architecture

Where Does LinkedIn use Voldemort?



- 2 Usage-Patterns
- Read-Write Store
 - Uses BDB JE for the storage engine
 - 50% of Voldemort Stores (aka Tables) are RW
- Read-only Store
 - Uses a custom Read-only format
 - 50% of Voldemort Stores (aka Tables) are RO
- Let's look at the RO Store



Voldemort : RO Store Usage at LinkedIn

People You May Know



Viewers of this profile also viewed

Viewers of this profile also viewed...

Igor Perisic

Anmol Bhasin

Jun Rao

Principal Engineer at LinkedIn

Director of Engineering; Search,...

Recommendations, A/B Testing and ...

Principle Software Engineer at LinkedIn

Related Searches

Related searches for hadoop		
mapreduce	java	
big data	hbase	
machine learning	lucene	
data mining	data warehouse	

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Jobs you may be interested in

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other people are ng.			
Ninth Software Craftsmanship Meeting December 19, 2011 – SAP Labs, HaTidhar 15 Ra'anana, 43665, Israel			
3rd Italian Information Retrieval Workshop (IIR 2012) January 26-27, 2012 – Dipartimento di Informatica (DIB), Università di Bari "Ald			
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Clojure/West 2012 March 16-17, 2012 – San Jose Marriott			
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GiberCodes	Senior Software Engineer CyberCoders - San Jose ,CA	×	
Pelican	Senior Software Engineer - Qualcomm Platform Pelican Imaging Corporation - San Francisco Bay Area	×	

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Voldemort : Usage Patterns @ LinkedIn

RO Store Usage Pattern

- 1. Use Hadoop to build a model
- 2. Voldemort loads the output of Hadoop
- 3. Voldemort serves fast key-value look-ups on the site
 - e.g. For key="Sid Anand", get all the people that "Sid Anand" may know!
 - e.g. For key="Sid Anand", get all the jobs that "Sid Anand" may be interested in!



Read-Only Store Build and Swap Process

How Do The Voldemort RO Stores Perform?





Voldemort : RO Store Performance : TP vs. Latency

- MySQL - Voldemort





throughput (qps)

100 GB data, 24 GB RAM

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LinkedIn Data Infrastructure Solutions

Databus : Timeline-Consistent Change Data Capture



Where Does LinkedIn use DataBus?



DataBus : Use-Cases @ LinkedIn



A user updates his profile with skills and position history. He also accepts a connection

- The write is made to an Oracle Master and DataBus replicates:
- the profile change to the Standardization service
 - E.G. the many forms of IBM are canonicalized for search-friendliness and recommendation-friendliness
- the profile change to the Search Index service
 - Recruiters can find you immediately by new keywords
- the connection change to the Graph Index service
 - > The user can now start receiving feed updates from his new connections immediately



DataBus : Architecture



DataBus consists of 2 services

- Relay Services
 - Sharded
 - Maintain an in-memory buffer per shard
 - Each shard polls Oracle and then deserializes transactions into Avro

Bootstrap Service

- Picks up online changes as they appear in the Relay
- Supports 2 types of operations from clients
 - If a client falls behind and needs records older than what the relay has, Bootstrap can send consolidated deltas!
 - If a new client comes on line, Bootstrap can send a consistent snapshot



DataBus : Architecture



Guarantees

- Transactional semantics
- In-commit-order Delivery
- At-least-once delivery
- Durability (by data source)
- High-availability and reliability
- Low latency

 Generate consistent snapshots and consolidated deltas during continuous updates with long-running queries



LinkedIn Data Infrastructure Solutions

Kafka: High-Volume Low-Latency Messaging System



Kafka : Usage at LinkedIn

Where as DataBus is used for Database change capture and replication, Kafka is used for application-level data streams

Examples:

- End-user Action Tracking (a.k.a. Web Tracking) of
 - Emails opened
 - Pages seen
 - Links followed
 - Executing Searches
- Operational Metrics
 - Network & System metrics such as
 - TCP metrics (connection resets, message resends, etc...)
 - System metrics (iops, CPU, load average, etc...)

Kafka : Overview



Features

- Pub/Sub
- Batch Send/Receive
- System Decoupling

Guarantees

- At least once delivery
- Very high throughput
- Low latency
- Durability
- Horizontally Scalable

<u>Scale</u>

- Billions of Events
- TBs per day
- Inter-colo: few seconds
- Typical retention: weeks

Key Design Choices

- When reading from a file and sending to network socket, we typically incur 4 buffer copies and 2 OS system calls
 - Kafka leverages a sendFile API to eliminate 2 of the buffer copies and 1 of the system calls
- No double-buffering of messages we rely on the OS page cache and do not store a copy of the message in the JVM
 - Less pressure on memory and GC
 - If the Kafka process is restarted on a machine, recently accessed messages are still in the page cache, so we get the benefit of a warm start
- Kafka doesn't keep track of which messages have yet to be consumed -- i.e. no book keeping overhead
 - Instead, messages have time-based SLA expiration -- after 7 days, messages are deleted



How Does Kafka Perform?





Kafka : Performance : Throughput vs. Latency

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Kafka : Performance : Linear Incremental Scalability



(10 topics, broker flush interval 100K)



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Kafka : Performance : Resilience as Messages Pile Up

(1 topic, broker flush interval 10K)





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