

# **Serving Large-scale Batch Computed Data with Project Voldemort**

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#### **Challenges of Serving Data-Derived Features**

lata warehouse



- Requires an almost complete refresh of the data
- **Problem 1:** Serving latency degradation during bulk load • Due to updates on serving index
- **Problem 2:** Potential long time in error state



• Error in algorithm = Bulk load bad data = Bad state till next load

#### Motivation

Serving system with fast bulk loads and minimal read latency penalty

# Voldemort

- Distributed key/value system
- Pluggable storage layer
- A cluster has multiple *stores*  $(\sim tables)$



## **Our solution**

- Custom storage engine leveraging Hadoop
- Building store offline solves online performance problem
- Store kept on HDFS in directories per Voldemort node
- Voldemort nodes pulls store data in parallel

## Hadoop / HDFS

- Suitable for batch algorithms
- De-facto for storing large logs

- Every store saved into multiple versioned directories
- One version of every store is serving, rest for rollback
- Swap = close old store files, open new ones
- Rollback = close new store files, open old ones

# Chunk set







- A store consists of multiple chunk sets (data + index file)
- Increase parallelism by increasing reducers
- Swap = memory map index files of all chunk sets
- Key search = binary search in index + jump in data file
- Subset of key of key in index = Increase cache locality

100 200 300 400 500 600 700 throughput (qps)

Single node (24 GB RAM) latency for 100 GB data

• Scales to twice the throughput of MySQL while maintaining latency

50 output size (TB)

32 node latency for varying data size

- Latency grows linearly with  $\bullet$ data size
- Running at LinkedIn for the past 2 years
- Pushing ~ 4 TB of new data to production daily
- Open source (Apache License) http://project-voldemort.com