Big Telco, Bigger real-time demands: Moving towards Real-time analytics

Jung Ryong Lee

SK telecom

- IT Manager of SK Telecom, South Korea's largest wireless communications provider

Real-time analytics

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SK telecom

- IT Manager of SK Telecom, South Korea's largest wireless communications provider
- Work on commercial products (~ '12)
  - Has worked with email archived solutions
  - Has worked with IDS using In-stream processing engine
- Open source activity ('14 ~)
  - Contributor of REEF

Overview

- Background
- Real-time analytics in Telco

- Has worked with IDS using In-stream processing engine

- Open source activity ('14 ~)
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Overview

- Background
- Real-time analytics in Telco
- Project 1 - High speed data processing
  - Issues & solutions
  - Performance
- Project 2 - In-stream processing
- Lessons Learned

Background

- Telco data characteristics
  - Huge amount of data daily

- Performance
- Project 2 - In-stream processing
- Lessons Learned

Background

- Telco data characteristics
  - Huge amount of data daily
    - 40 TB/day
    - 15 PB (estimated by the end of 2014)
- Active user of Hadoop
  - Involved with 10 + Hadoop clusters
    - The largest one has 500 + nodes and a total of 900 + nodes altogether.
- Uses various commercial MPP databases for analytics

10 + Clusters

Real-time analytics in Telco

Introduce 2 projects using Spark

- Involved with 10 + Hadoop clusters
  - The largest one has 500 + nodes and a total of 900 + nodes altogether.
- Uses various commercial MPP databases for analytics

Real-time analytics in Telco

Introduce 2 projects using Spark

1. The first being high speed data processing
   - Replacement of MPP database
2. The second being In-Stream data processing
   - Replacement of Hive batch job

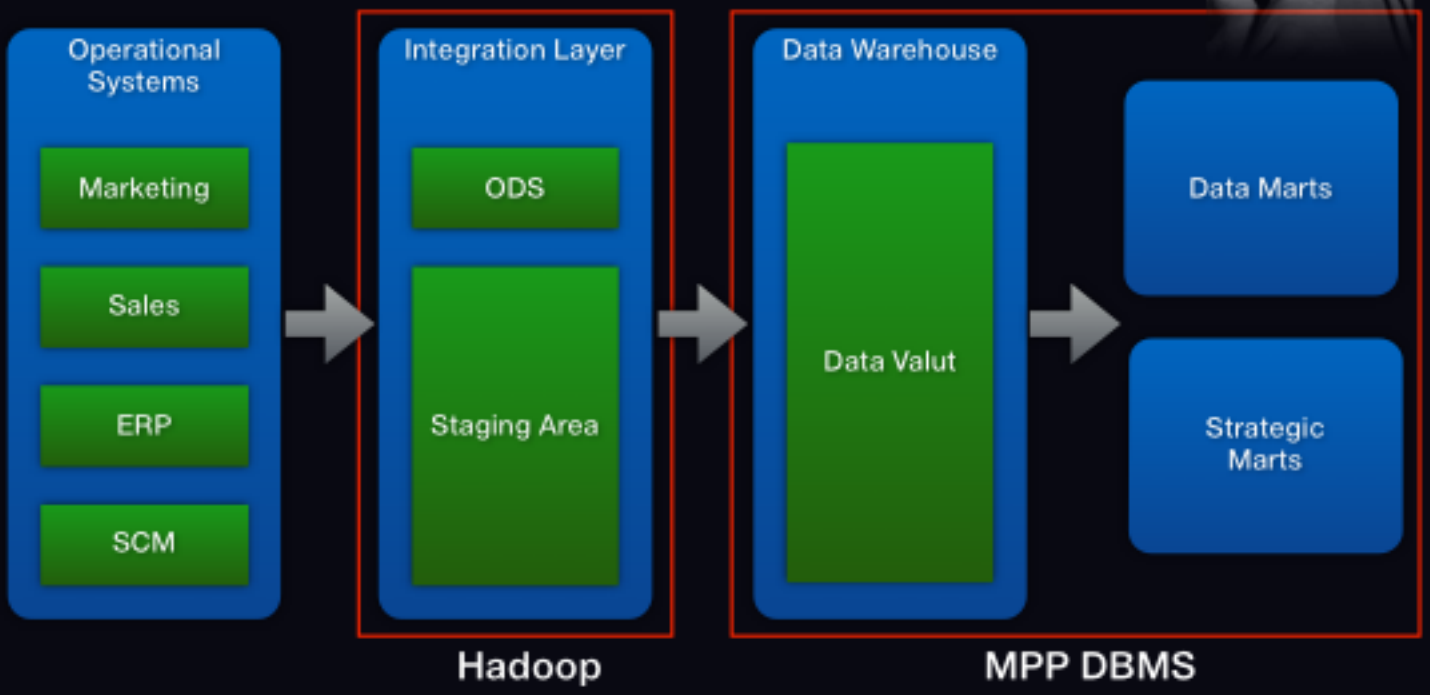
Previous approach

Working, But…

2. The second being In-Stream data processing
   - Replacement of Hive batch job

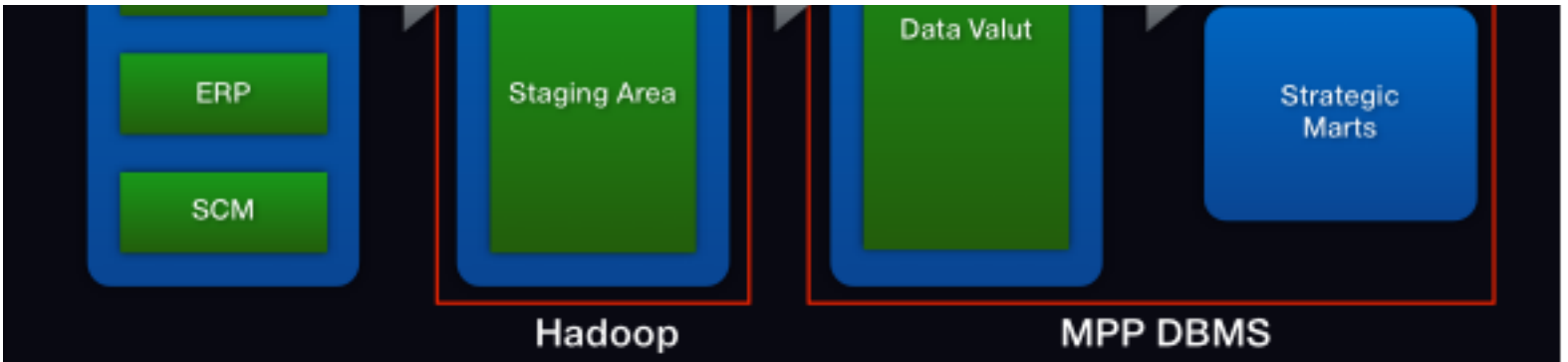
Previous approach

Working, But…



Previous approach

have to load too much data into MPP DBMS



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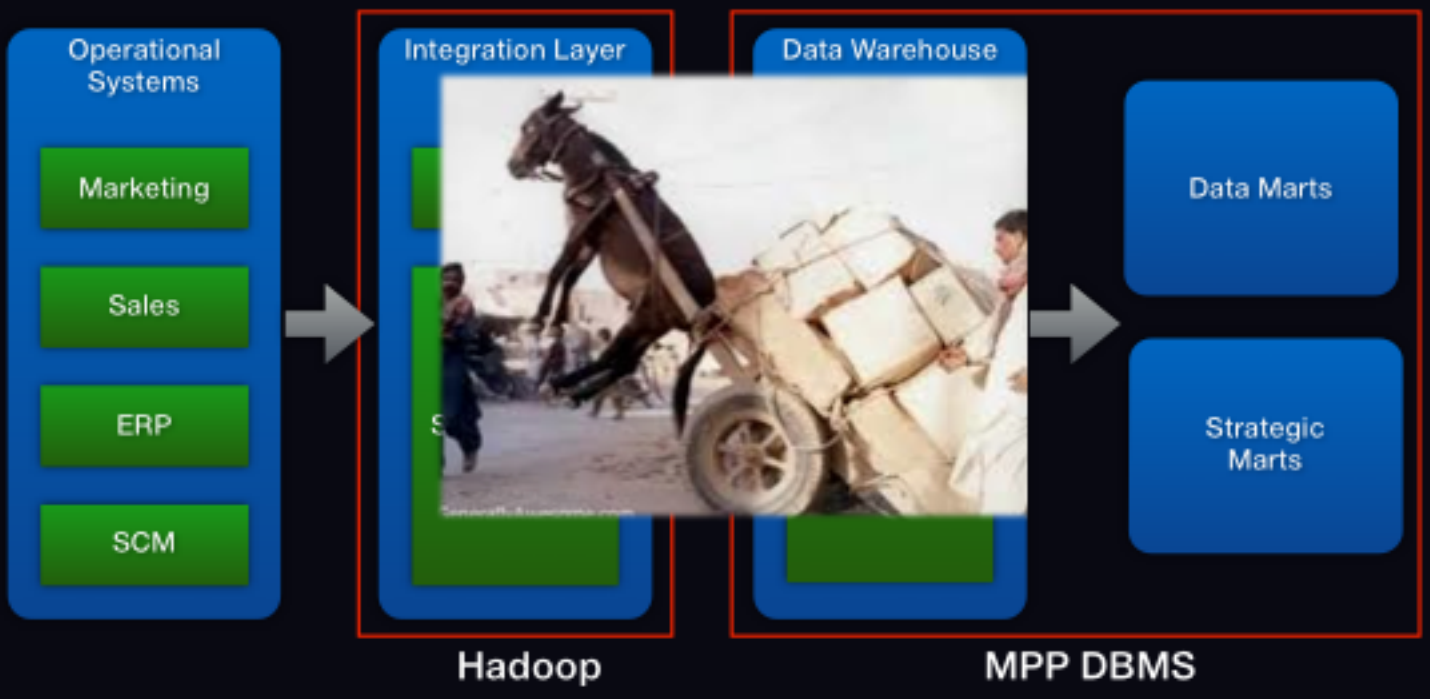
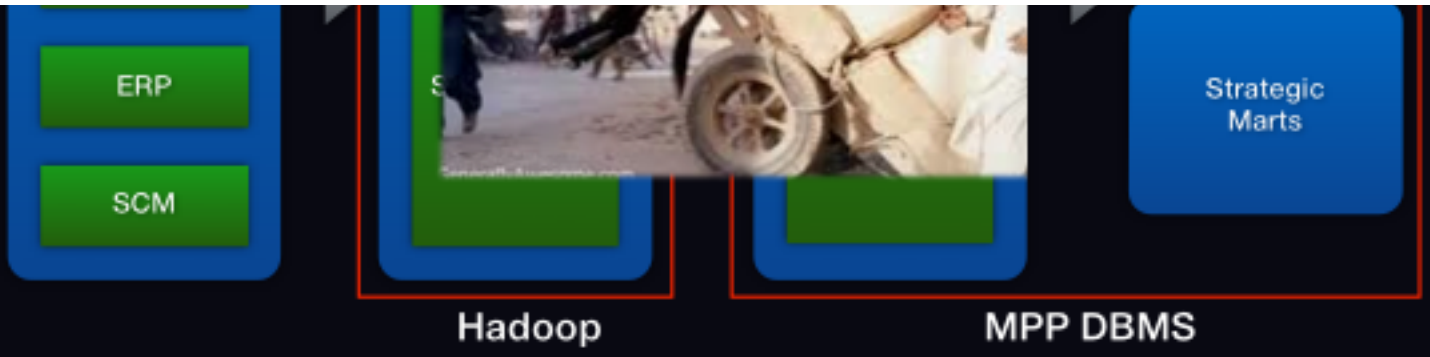


New approach

Support High speed ETL data processing

Real-time query processing for web client users



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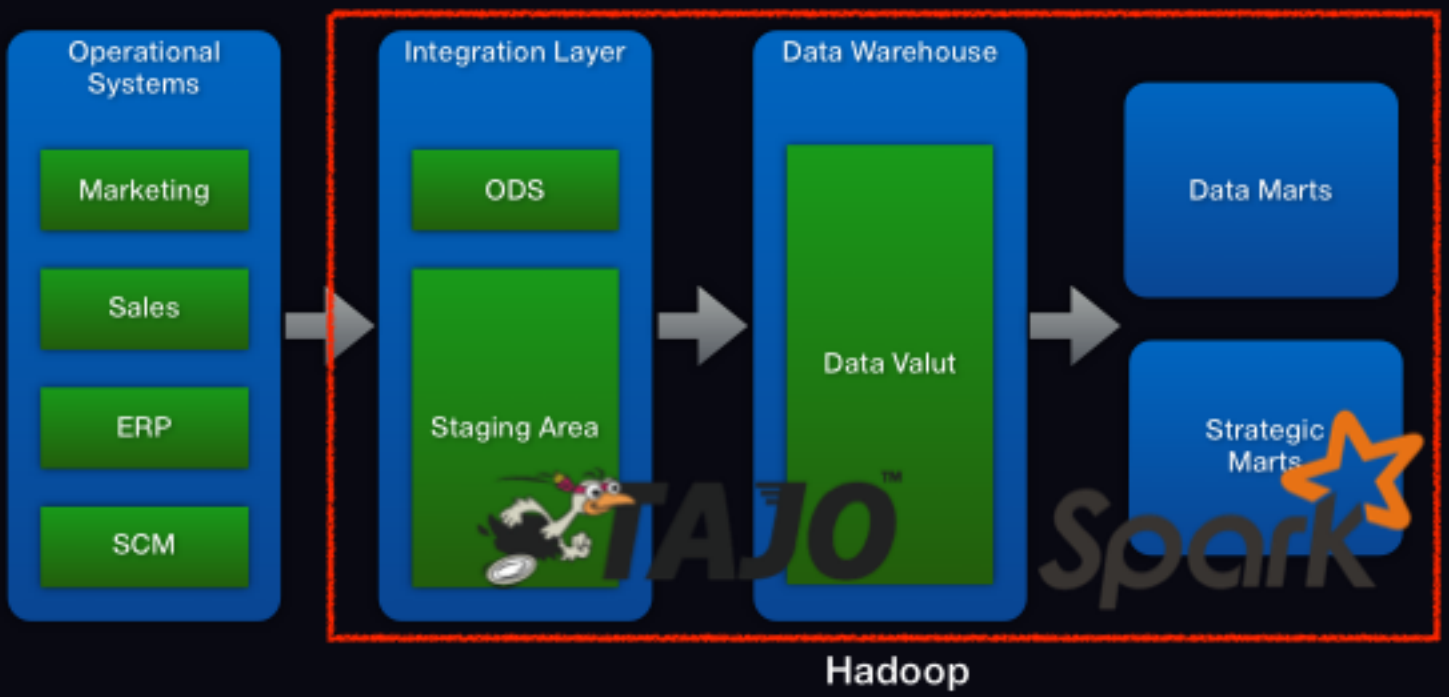

System Requirements

- Low latency ad-hoc query(< 2secs)

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- ANSI SQL support
  (no need for Insert/Update/Delete)
- JDBC support
- Support concurrent users(10 users per sec)
- High availability

Shark on Spark

- It can replace RDBMS

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- It can replace RDBMS
  - Low latency for ad-hoc query with caching tables
  - HiveQL support
  - JDBC support
- Support concurrent users(10users per sec)
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Shark on Spark

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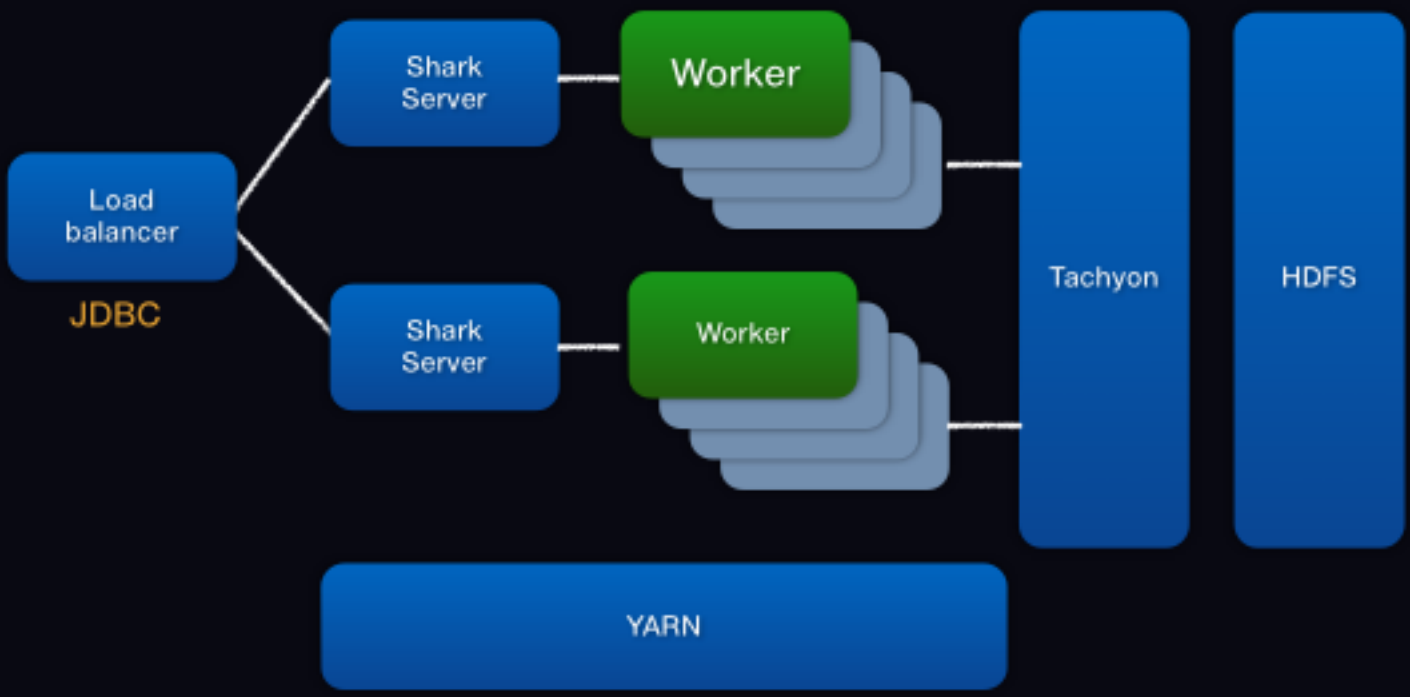
Shark on Spark



Function & Performance

- All queries work good with some modifications of queries

URI	Samples	Samples diff	Average (ms) ↑
test430	1	0	21635

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URI	Samples	Samples diff	Average (ms) ↑
test430	1	0	21635
test431	1	0	21584
test428	1	0	21360
test429	1	0	21297
test432	1	0	21171
test228	1	0	17648
test230	1	0	14802
test229	1	0	14738
test231	1	0	14603
test232	1	0	14354
test309	1	0	5336
test321	1	0	5325
test323	1	0	5249
test308	1	0	5246
test310	1	0	5231
test315	1	0	5231
test316	1	0	5220
test320	1	0	5193
test313	1	0	5191
test337	1	0	5176
test318	1	0	5175
test312	1	0	5150

Improvement

- Improvement and Bug fix
  - Bug fix of Hive 0.11 / Support Non ASCII table name
  - Implement UDF(Oracle like functions)

test308	1	0	5246
test310	1	0	5231
test315	1	0	5231
test316	1	0	5220
test320	1	0	5193
test313	1	0	5191
test337	1	0	5176
test318	1	0	5175
test312	1	0	5150

Improvement

- Improvement and Bug fix
  - Bug fix of Hive 0.11 / Support Non ASCII table name
  - Implement UDF(Oracle like functions)
  - Change Correlation Sub Query to Left Outer Join
- Performance Improvement
  - Control reducer count
  - Material view can reduce query time
  - Caching query result

Improvement - Example

- Can reduce query time with a simple understanding of Shark and Hive

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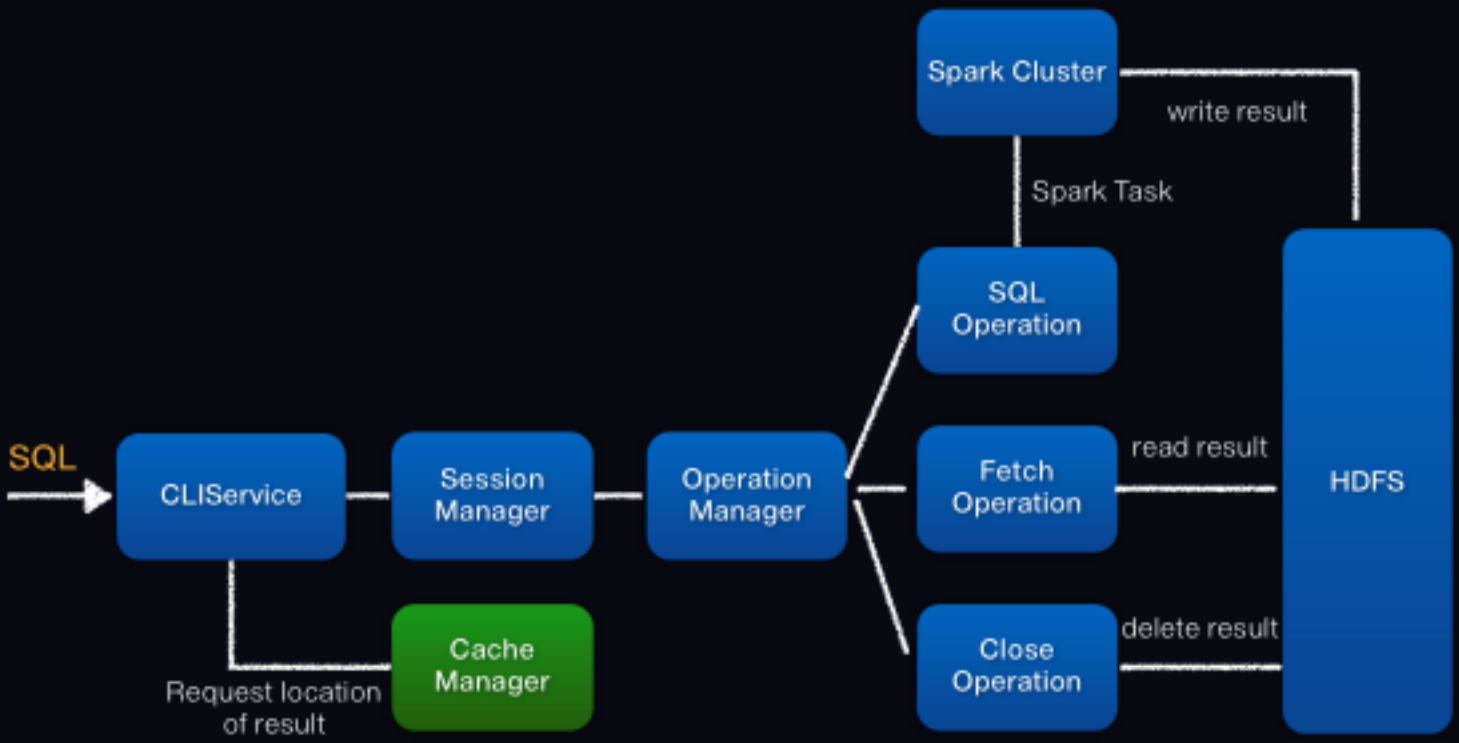
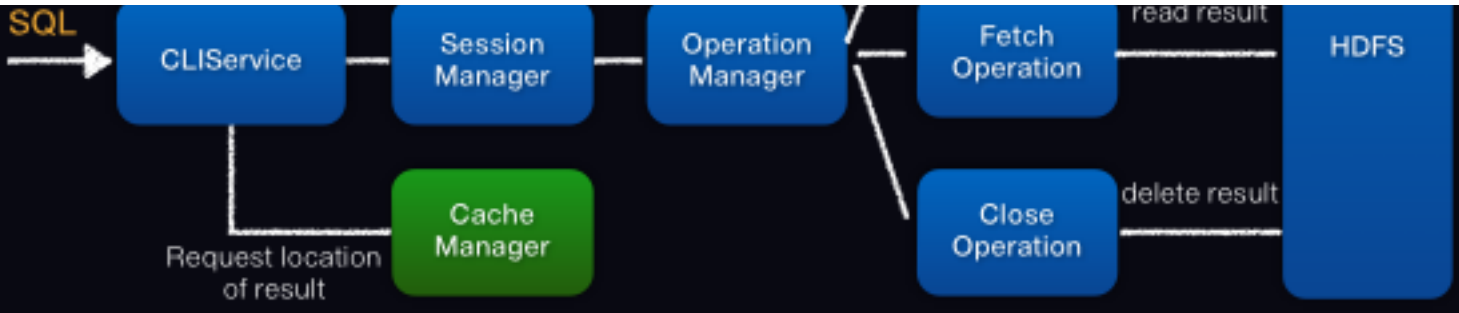


Result of Caching query

- Most of queries complete in 0.3 second.

URI	Samples	Samples diff	Average (ms)
test147	1	0	603
test052	1	0	337



Result of Caching query

- Most of queries complete in 0.3 second.

URI	Samples	Samples diff	Average (ms)
test147	1	0	603
test052	1	0	337
test244	1	0	332
test323	1	0	332
test076	1	0	331
test308	1	0	330
test068	1	0	328
test315	1	0	328
test245	1	0	326
test015	1	0	326
test253	1	0	325
test160	1	0	323
test022	1	0	323
test066	1	0	322
test150	1	0	321
test261	1	0	319
test073	1	0	319
test317	1	0	318
test030	1	0	318
test318	1	0	318
test048	1	0	317
test314	1	0	317
test049	1	0	317
test170	1	0	317
All URIs	524	0	275

In-Stream data processing

- Replace Hive batch job with Spark
  - One hour batch job -> 5 sec batch job in 1 minute of window time
- Calculate top 100 keywords and applications

In-Stream data processing

- Replace Hive batch job with Spark
  - One hour batch job -> 5 sec batch job in 1 minute of window time
- Calculate top 100 keywords and applications
- Processing data with 530MB/s
  -> 1 mil records / sec
- Must be implemented within one month

The answer is Spark Streaming!!

- Very similar to Spark
  - DStream is very similar to RDD

- Must be implemented within one month

The answer is Spark Streaming!!

- Very similar to Spark
  - DStream is very similar to RDD
  - Can use most functions that RDD provides(groupByKey, sortByKey)
  - Easily change batch application to in-stream processing application
- Support local environment
  - Can evaluate the application with laptop

Implementation process

- It only takes one week to complete the process

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  - Can evaluate the application with laptop

Implementation process

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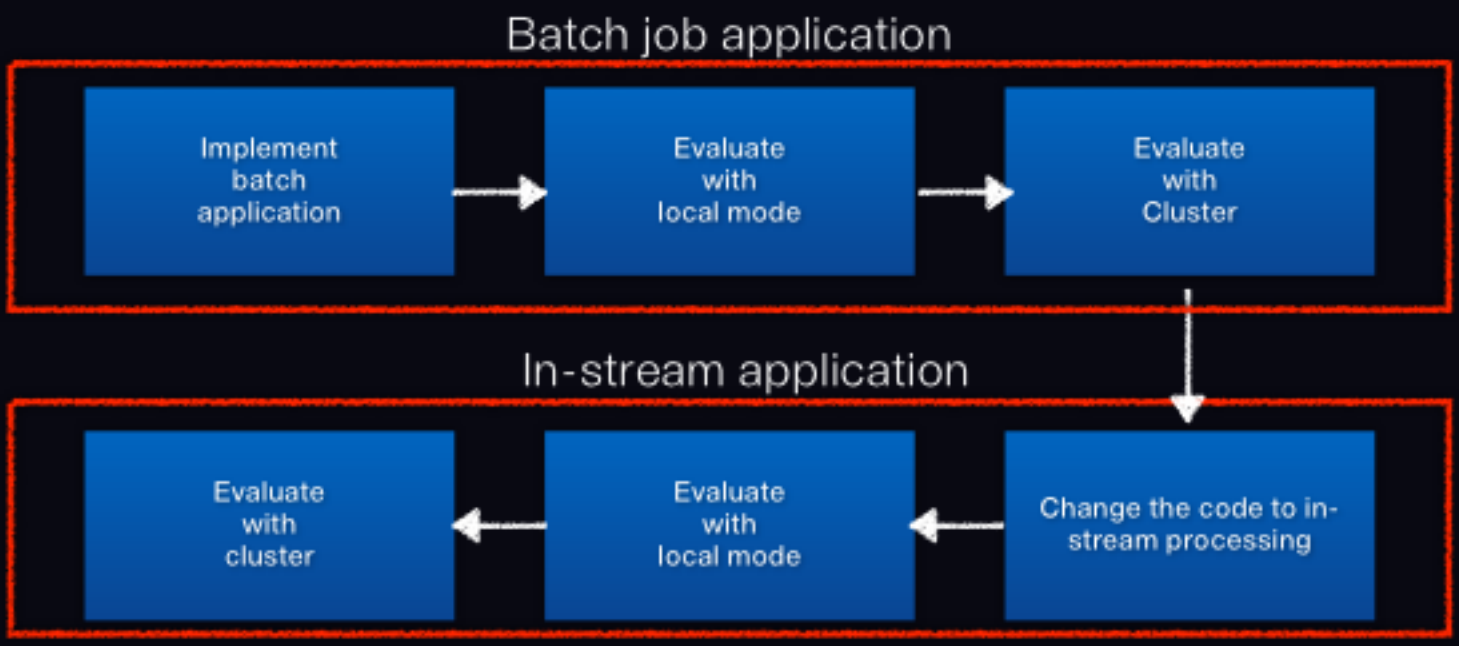
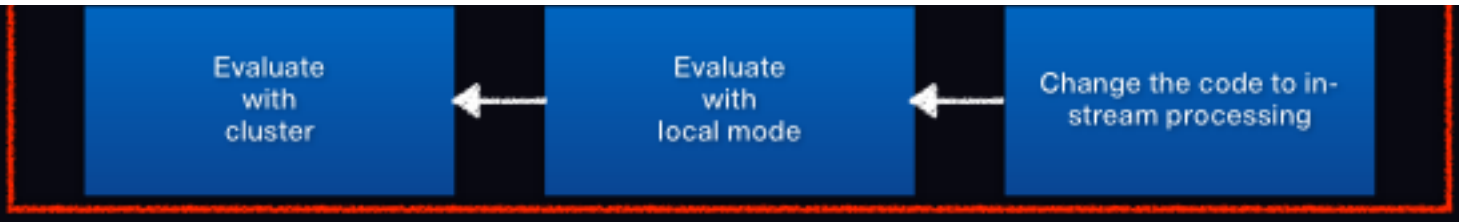


Data processing architecture

- Only “84 **lines of code**” are needed!



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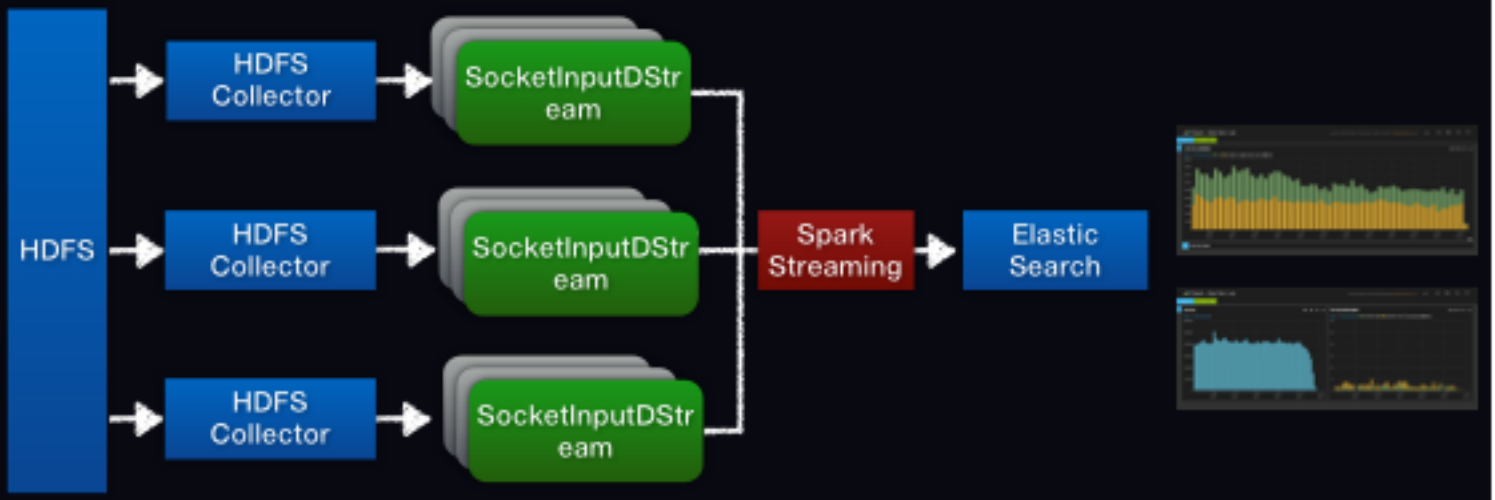
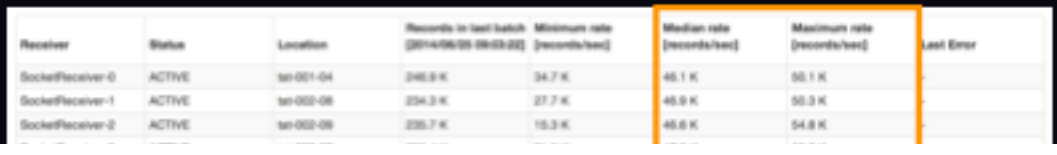
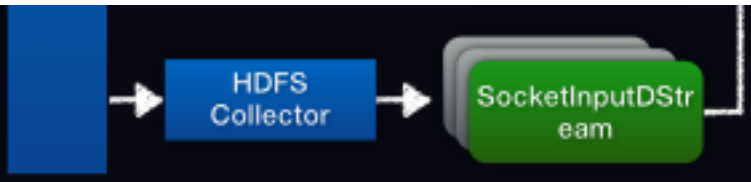
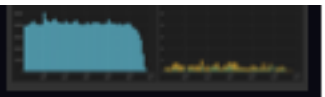


Performance

- About 3 times faster than other In-streaming processing engines.







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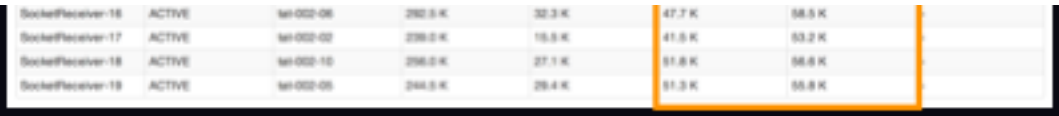
Receiver	Status	Location	Records in last batch [2014/06/05 09:03:22]	Minimum rate [records/sec]	Median rate [records/sec]	Maximum rate [records/sec]	Last Error
SocketReceiver-0	ACTIVE	tel-001-04	246.9 K	34.7 K	46.1 K	50.1 K	
SocketReceiver-1	ACTIVE	tel-002-08	234.3 K	27.7 K	46.9 K	50.3 K	
SocketReceiver-2	ACTIVE	tel-002-09	235.7 K	15.3 K	46.6 K	54.8 K	
SocketReceiver-3	ACTIVE	tel-002-07	256.4 K	21.8 K	47.9 K	56.8 K	
SocketReceiver-4	ACTIVE	tel-002-04	177.6 K	31.0 K	44.5 K	55.0 K	
SocketReceiver-5	ACTIVE	tel-001-11	235.6 K	34.9 K	47.1 K	55.9 K	
SocketReceiver-6	ACTIVE	tel-001-09	265.0 K	26.1 K	48.0 K	55.0 K	
SocketReceiver-7	ACTIVE	tel-001-03	239.5 K	22.1 K	44.2 K	54.8 K	
SocketReceiver-8	ACTIVE	tel-001-08	203.4 K	26.2 K	42.7 K	56.1 K	
SocketReceiver-9	ACTIVE	tel-001-10	214.3 K	32.3 K	46.5 K	58.0 K	
SocketReceiver-10	ACTIVE	tel-001-05	228.0 K	30.0 K	47.4 K	56.8 K	
SocketReceiver-11	ACTIVE	tel-001-02	202.8 K	27.7 K	47.9 K	57.0 K	
SocketReceiver-12	ACTIVE	tel-002-11	244.5 K	21.7 K	48.9 K	57.6 K	
SocketReceiver-13	ACTIVE	tel-001-06	210.8 K	28.8 K	46.6 K	56.9 K	
SocketReceiver-14	ACTIVE	tel-002-03	219.3 K	27.6 K	49.6 K	54.8 K	
SocketReceiver-15	ACTIVE	tel-001-07	241.6 K	41.7 K	51.8 K	56.2 K	
SocketReceiver-16	ACTIVE	tel-002-06	282.5 K	32.3 K	47.7 K	58.5 K	
SocketReceiver-17	ACTIVE	tel-002-02	239.0 K	15.5 K	41.5 K	53.2 K	
SocketReceiver-18	ACTIVE	tel-002-10	256.0 K	27.1 K	51.8 K	58.6 K	
SocketReceiver-19	ACTIVE	tel-002-05	244.5 K	29.4 K	51.3 K	55.8 K	

Batch Processing Statistics

Metric	Last batch	Minimum	25th percentile	Median	75th percentile	Maximum
Processing Time	94 ms	73 ms	86 ms	92 ms	99 ms	303 ms
Scheduling Delay	0 ms	0 ms	0 ms	0 ms	0 ms	1 ms
Total Delay	94 ms	73 ms	87 ms	93 ms	99 ms	303 ms

Lessons learned

- We can implement **OLTP** style systems and **In-stream data processing** with Spark and Shark.
- Win-win between community and company



SocketReceiver-16	ACTIVE	tel-002-06	282.0 K	32.3 K	47.7 K	58.5 K
SocketReceiver-17	ACTIVE	tel-002-02	239.0 K	15.5 K	41.5 K	53.2 K
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- We can implement **OLTP** style systems and **In-stream data processing** with Spark and Shark.
- Win-win between community and company
  - Test in real working cluster
    - Finding some bugs and function requirement, as well.
    - mainly focusing on low latency query.


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