Apache Kylin

OLAP on Hadoop





TOUB





Agenda

What's Apache Kylin?

- Tech Highlights
- Performance
- Roadmap
- Q & A





kylin / 'kiː'lın / 麒麟

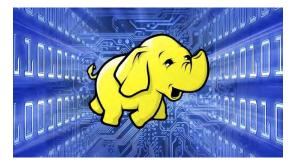
--n. (in Chinese art) a mythical animal of composite form

Extreme OLAP Engine for Big Data

Kylin is an open source Distributed Analytics Engine from eBay that provides SQL interface and multi-dimensional analysis (OLAP) on Hadoop supporting extremely large datasets

- Open Sourced on Oct 1st, 2014
- Be accepted as Apache Incubator Project on Nov 25th, 2014

Big Data Era



- More and more data becoming available on Hadoop
- Limitations in existing Business Intelligence (BI) Tools
 - Limited support for Hadoop
 - Data size growing exponentially
 - High latency of interactive queries
 - Scale-Up architecture
- Challenges to adopt Hadoop as interactive analysis system
 - Majority of analyst groups are SQL savvy
 - No mature SQL interface on Hadoop
 - OLAP capability on Hadoop ecosystem not ready yet

Why not Build an engine from scratch?

Features Highlights

Extreme Scale OLAP Engine

Kylin is designed to query 10+ billions of rows on Hadoop

ANSI SQL Interface on Hadoop

Kylin offers ANSI SQL on Hadoop and supports most ANSI SQL query functions

Seamless Integration with BI Tools

Kylin currently offers integration capability with BI Tools like Tableau.

Interactive Query Capability

Users can interact with Hive tables at sub-second latency

MOLAP Cube

Define a data model from Hive tables and pre-build in Kylin

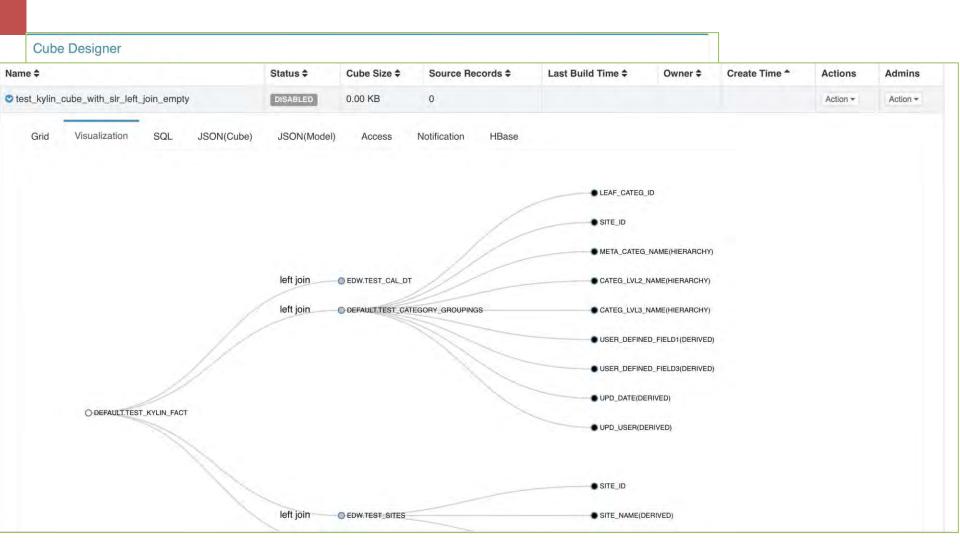
Scale Out Architecture

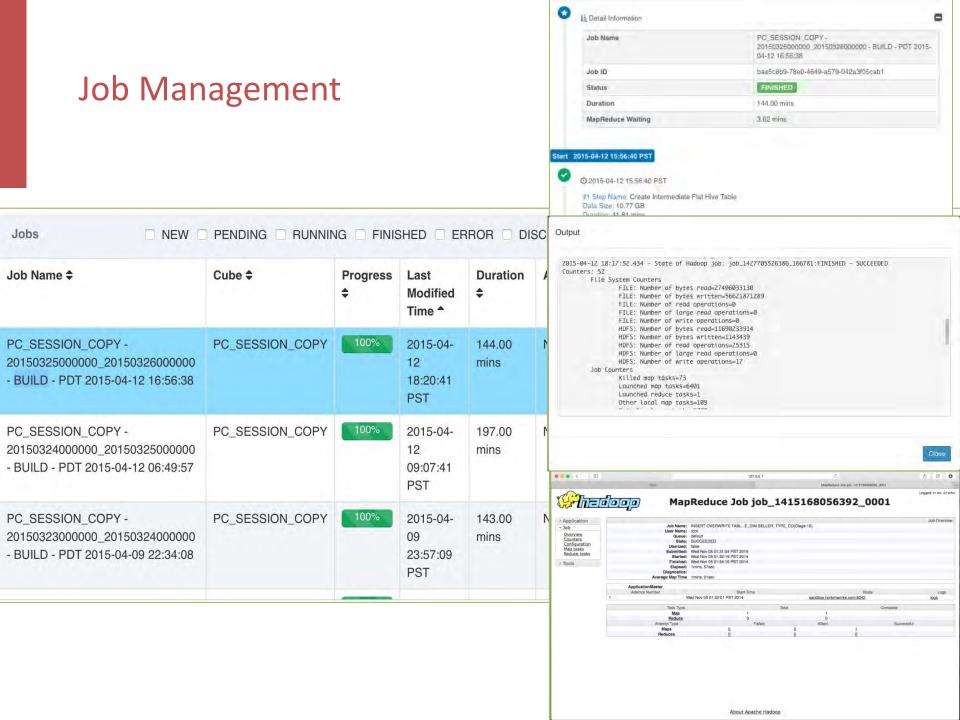
Query server cluster supports thousands concurrent users and provide high availability

Features Highlights...

- Compression and Encoding Support
- Incremental Refresh of Cubes
- Approximate Query Capability for distinct count (HyperLogLog)
- Leverage HBase Coprocessor for query latency
- Job Management and Monitoring
- Easy Web interface to manage, build, monitor and query cubes
- Security capability to set ACL at Cube/Project Level
- Support LDAP Integration

Cube Designer





Query and Visualization

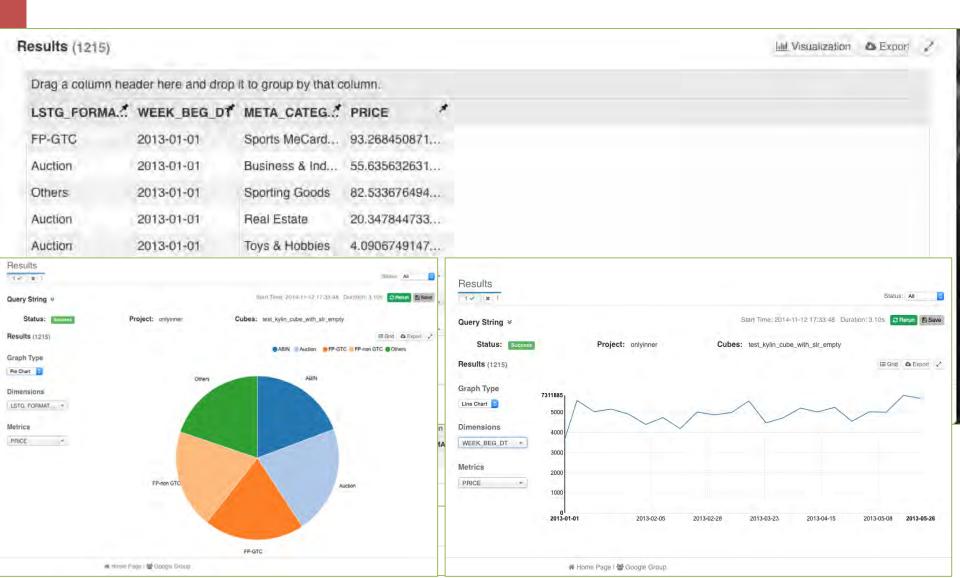
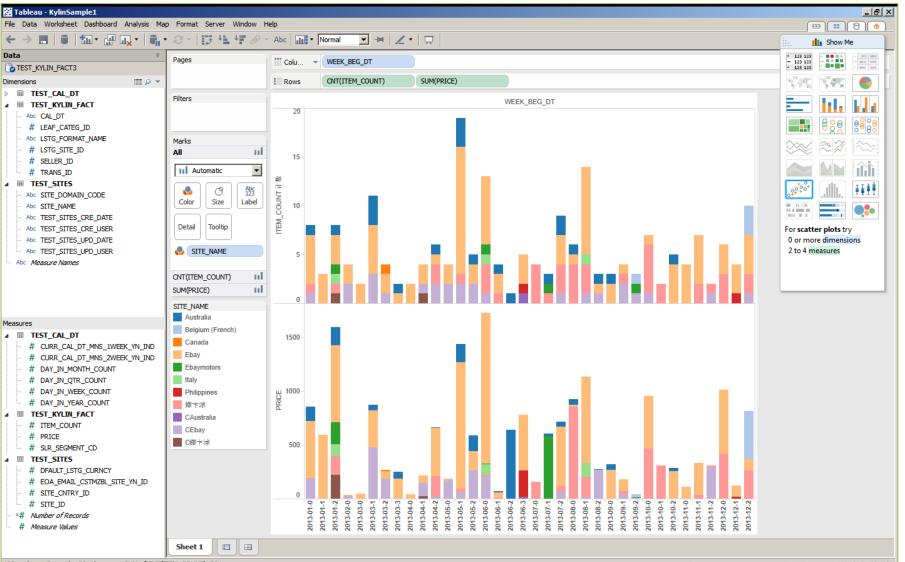


Tableau Integration



Who are using Kylin

eBay

90% query < 5 seconds</p>

Case	Cube Size	Raw Records
User Session Analysis	26 TB	28+ billion rows
Classified Traffic Analysis	21 TB	20+ billion rows
GeoX Behavior Analysis	560 GB	1.2+ billion rows

- Baidu
 - Baidu Map internal analysis
- Many other Proof of Concepts
 - Bloomberg Law, British GAS, JD, Microsoft, StubHub, Tableau ...

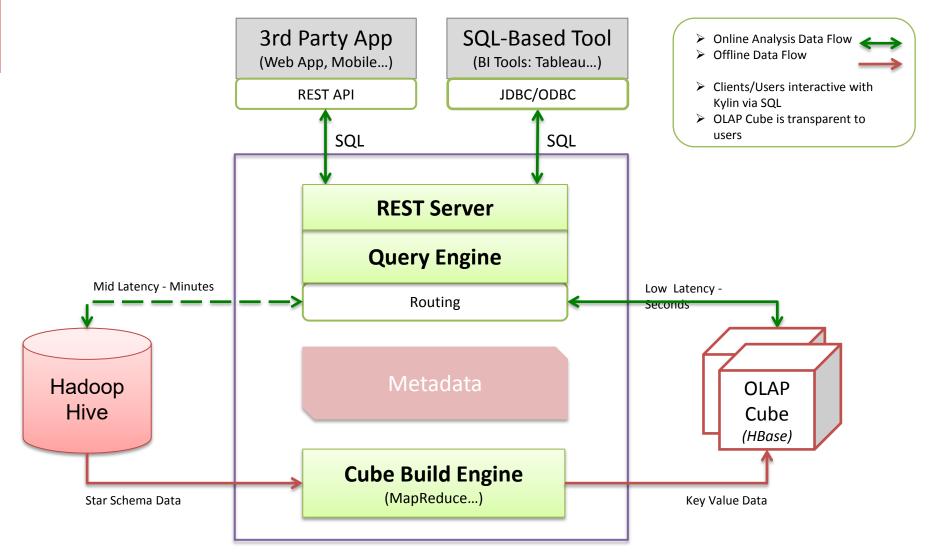


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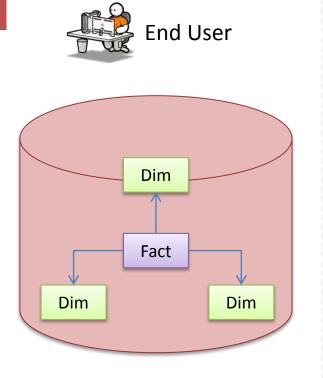
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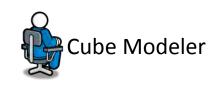
Kylin Architecture Overview



Data Modeling

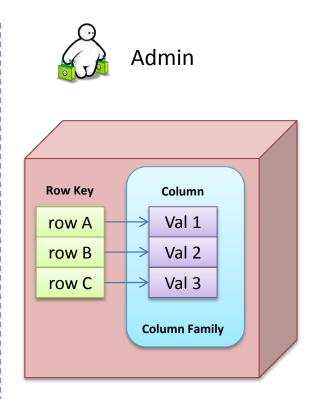


<u>Source</u> Star Schema



Cube: ... Fact Table: ... Dimensions: ... Measures: ... Storage(HBase): ...

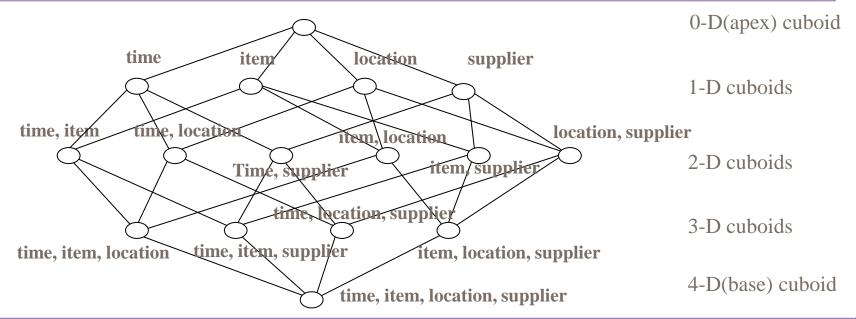
Mapping Cube Metadata



<u>Target</u> HBase Storage

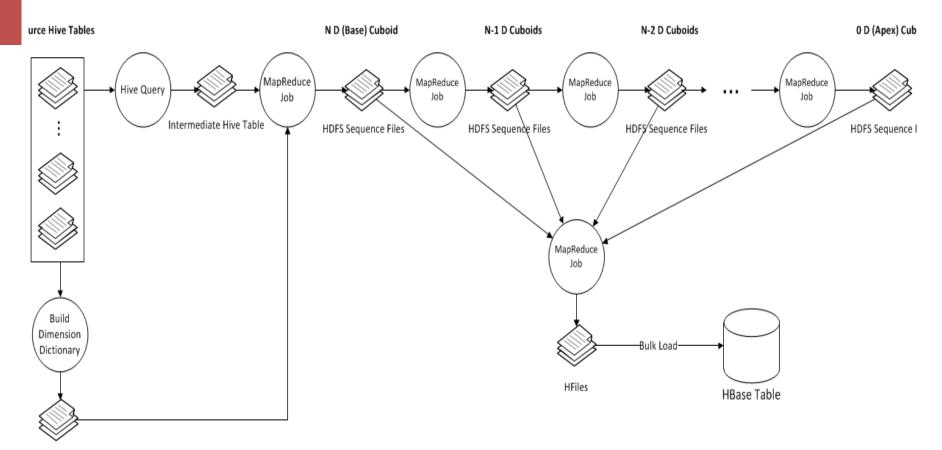
OLAP Cube – Balance between Space and Time

- Cuboid = one combination of dimensions
- Cube = all combination of dimensions (all cuboids)



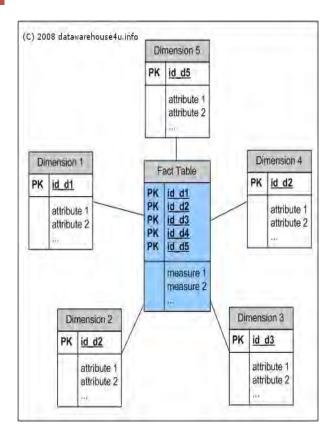
- Base vs. aggregate cells; ancestor vs. descendant cells; parent vs. child cells
 - 1. (9/15, milk, Urbana, Dairy_land) <time, item, location, supplier>
 - 2. (9/15, milk, Urbana, *) <time, item, location>
 - 3. (*, milk, Urbana, *) <item, location>
 - 4. (*, milk, Chicago, *) <item, location>
 - 5. (*, milk, *, *) **<item>**

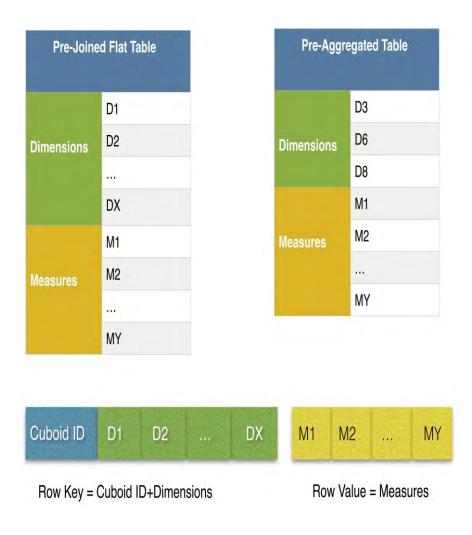
Cube Build Job Flow





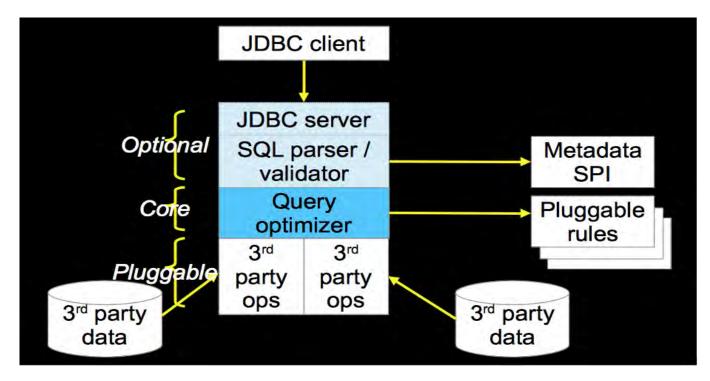
How To Store Cube? – HBase Schema





Query Engine – Calcite

- Dynamic data management framework.
- Formerly known as Optiq, Calcite is an Apache incubator project, used by Apache Drill and Apache Hive, among others.
- <u>http://optiq.incubator.apache.org</u>



Kylin Extensions on Calcite

Metadata SPI

- Provide table schema from Kylin metadata
- Optimize Rule
 - Translate the logic operator into Kylin operator
- Relational Operator
 - Find right cube
 - Translate SQL into storage engine API call
 - Generate physical execute plan by linq4j java implementation
- Result Enumerator
 - Translate storage engine result into java implementation result.
- SQL Function
 - Add HyperLogLog for distinct count
 - Implement date time related functions (i.e. Quarter)

Query Engine – Kylin Explain Plan

SELECT test_cal_dt.week_beg_dt, test_category.category_name, test_category.lvl2_name, test_category.lvl3_name, test_kylin_fact.lstg_format_name, test_sites.site_name, SUM(test_kylin_fact.price) AS GMV, COUNT(*) AS TRANS_CNT FROM test_kylin_fact

LEFT JOIN test_cal_dt ON test_kylin_fact.cal_dt = test_cal_dt.cal_dt

LEFT JOIN test_category ON test_kylin_fact.leaf_categ_id = test_category.leaf_categ_id AND test_kylin_fact.lstg_site_id = test_category.site_id

LEFT JOIN test_sites ON test_kylin_fact.lstg_site_id = test_sites.site_id

WHERE test_kylin_fact.seller_id = 123456OR test_kylin_fact.lstg_format_name = 'New'

GROUP BY test_cal_dt.week_beg_dt, test_category.category_name, test_category.lvl2_name, test_category.lvl3_name, test_kylin_fact.lstg_format_name,test_sites.site_name

OLAPToEnumerableConverter

OLAPProjectRel(WEEK_BEG_DT=[\$0], category_name=[\$1], CATEG_LVL2_NAME=[\$2], CATEG_LVL3_NAME=[\$3], LSTG_FORMAT_NAME=[\$4], SITE_NAME=[\$5], GMV=[CASE(=(\$7, 0), null, \$6)], TRANS_CNT=[\$8]) OLAPAggregateRel(group=[{0, 1, 2, 3, 4, 5}], agg#0=[\$SUM0(\$6)], agg#1=[COUNT(\$6)], TRANS_CNT=[COUNT()]) OLAPProjectRel(WEEK_BEG_DT=[\$13], category_name=[\$21], CATEG_LVL2_NAME=[\$15], CATEG_LVL3_NAME=[\$14], LSTG_FORMAT_NAME=[\$5], SITE_NAME=[\$23], PRICE=[\$0]) OLAPFilterRel(condition=[OR(=(\$3, 123456), =(\$5, 'New'))]) OLAPJoinRel(condition=[OR(=(\$3, 123456), =(\$5, 'New'))]) OLAPJoinRel(condition=[AND(=(\$6, \$22), =(\$2, \$17))], joinType=[left]) OLAPJoinRel(condition=[AND(=(\$6, \$22), =(\$2, \$17))], joinType=[left]) OLAPJoinRel(condition=[(DEFAULT, TEST_KYLIN_FACT]], fields=[[0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11]]) OLAPTableScan(table=[[DEFAULT, TEST_CAL_DT]], fields=[[0, 1, 2, 3, 4, 5, 6, 7, 8]]) OLAPTableScan(table=[[DEFAULT, TEST_SITES]], fields=[[0, 1, 2, 3, 4, 5, 6, 7, 8]])

Storage Engine

Plugin-able storage engine

- Common iterator interface for storage engine
- Isolate query engine from underline storage

Translate cube query into HBase table scan

- Columns, Groups \rightarrow Cuboid ID
- Filters -> Scan Range (Row Key)
- Aggregations -> Measure Columns (Row Values)

Scan HBase table and translate HBase result into cube result

HBase Result (key + value) -> Cube Result (dimensions + measures)

Cube Optimization

- Curse of dimensionality: N dimension cube has 2^N cuboid
 - Full Cube vs. Partial Cube
- Hugh data volume
 - Dictionary Encoding
 - Incremental Building

Full Cube vs. Partial Cube

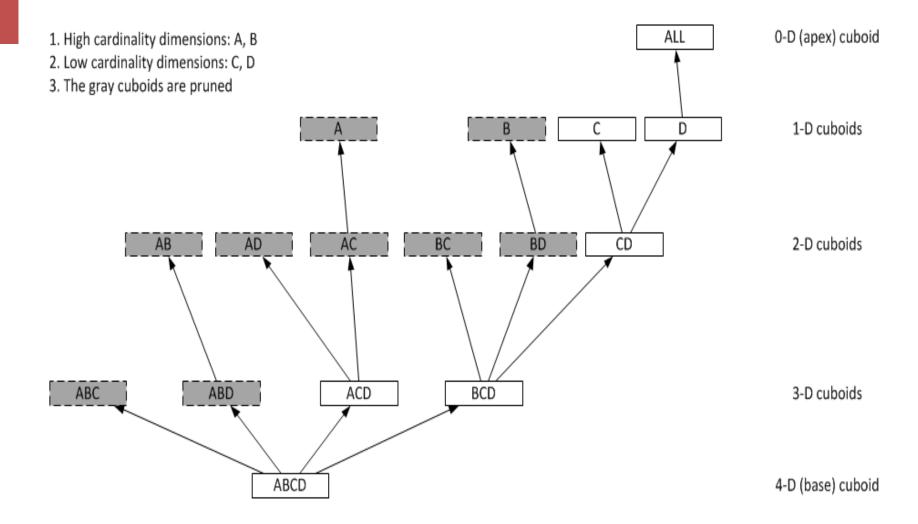
Full Cube

- Pre-aggregate all dimension combinations
- "Curse of dimensionality": N dimension cube has 2^N cuboid.

Partial Cube

- To avoid dimension explosion, we divide the dimensions into different aggregation groups
 - $2^{N+M+L} \rightarrow 2^{N} + 2^{M} + 2^{L}$
- For cube with 30 dimensions, if we divide these dimensions into 3 group, the cuboid number will reduce from 1 Billion to 3 Thousands
 2³⁰ → 2¹⁰ + 2¹⁰ + 2¹⁰
- Tradeoff between online aggregation and offline pre-aggregation

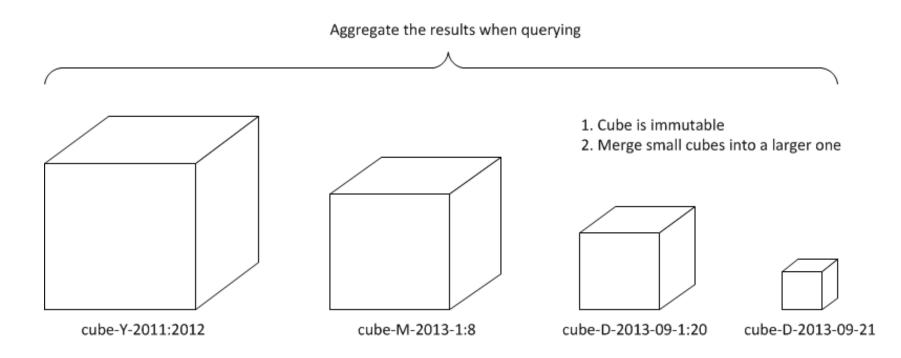
Partial Cube



Dictionary Encoding

- Data cube has lost of duplicated dimension values
- Dictionary maps dimension values into IDs that will reduce the memory and storage footprint.
- Dictionary is based on Trie

Incremental Build



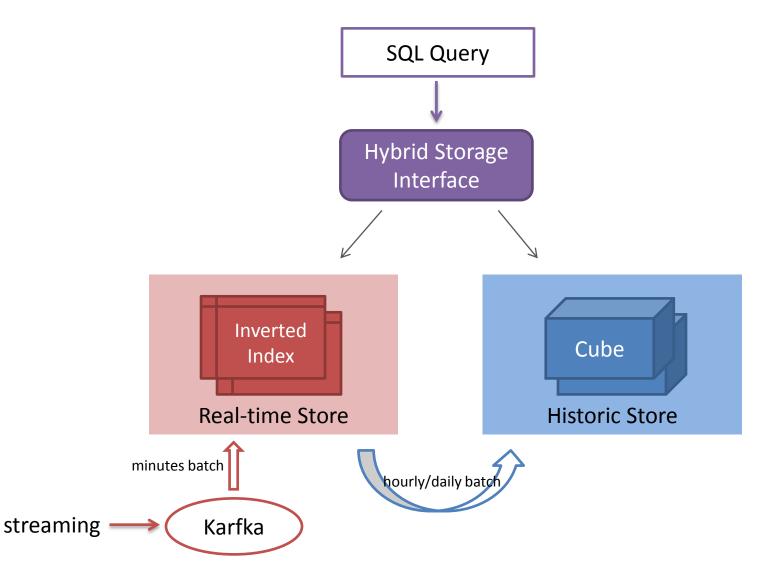
Streaming, ongoing effort

• Cube is great, but...

- Sometimes we want to drill down to row level information
- Cube takes time to build, how about real-time analysis?

Streaming with inverted index					
	Cube Comp	Inverted Index			
Storage format	Pre-aggregated cuboids	Sharding, columnar storage, with inverted index on row blocks			
Query method	Cuboid scanning	Massive parallel processing			
Strength	Pre-aggregate huge <u>historic</u> data to small summaries	Swift response to <u>real-time</u> data			
Weakness	Take time to build	Slow at scanning large data volume			

Kylin 0.8, Lambda Architecture

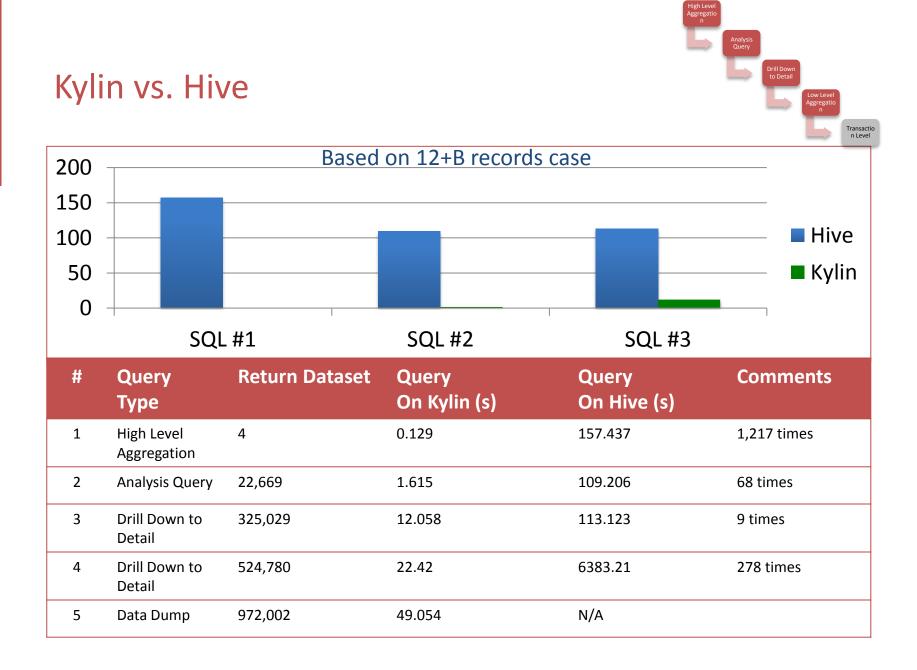




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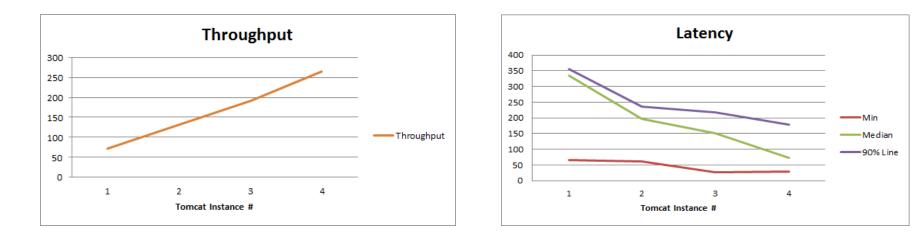




Performance -- Concurrency

Single Tomcat Instance on a Single Machine

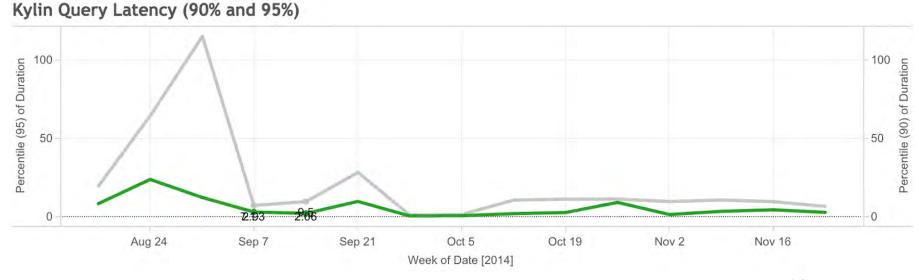
	Parallel Thread #	Data			Latency (ms)				Throughput
		Raw Recors	HBase Scan	Return	Min	Мах	Median	90% Line	
High Level Aggregation Query	30	1,940,304,293	5	5	67	1809	334	355	72.5/sec
Detail Level Query (with Seller ID)	30	13,683,834,542	43934	7283	1758	4534	2182	3171	9.7/sec



Linear scale out with more nodes

Performance - Query Latency

90% queries <5s



Green Line: 90%tile queries Gray Line: 95%tile queries

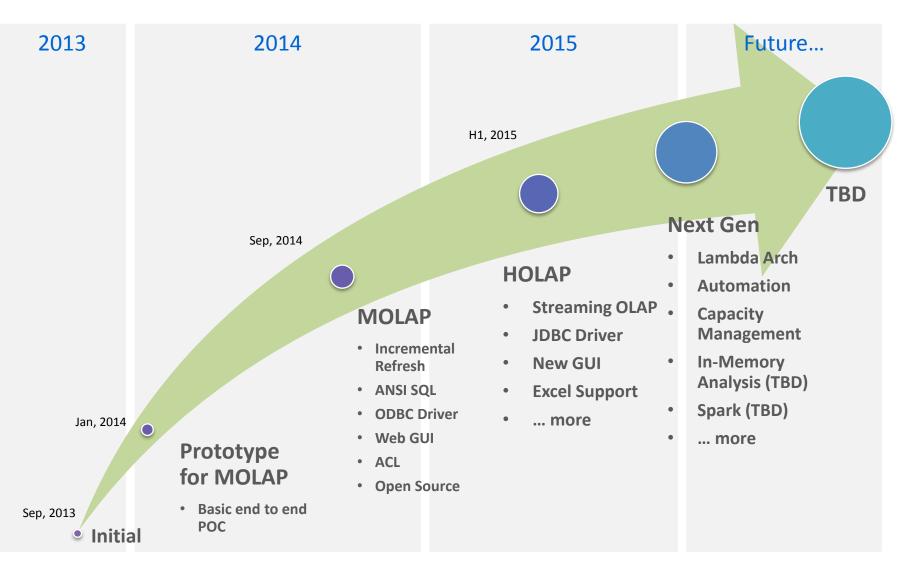


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Kylin Evolution Roadmap



Kylin Ecosystem

Kylin Core

 Fundamental framework of Kylin OLAP Engine

Extension

 Plugins to support for additional functions and features

Integration

 Lifecycle Management Support to integrate with other applications

Interface

 Allows for third party users to build more features via userinterface atop Kylin core

Driver

ODBC and JDBC Drivers

Integration

- \rightarrow ODBC Driver
- \rightarrow ETL
- \rightarrow Drill
- → SparkSQL

Kylin OLAP Core

Extension

- \rightarrow Security
- \rightarrow Redis Storage
- \rightarrow Spark Engine
- \rightarrow Docker

Interface

- \rightarrow Web Console
- \rightarrow Customized BI
- → Ambari/Hue Plugin

Apache Kylin

- Kylin Site:
 <u>http://kylin.io</u>
- Twitter:
 - @ApacheKylin
- 微信
 - ApacheKylin



