# Big Data - Latest Technologies

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#### Overview

- Cloudera Record Service
- Cloudera Kudu
  - ☐ Both announced on 09/28/2015
- Apache Arrow
  - □ Apache 'Top Level Project' from 02/17/2016

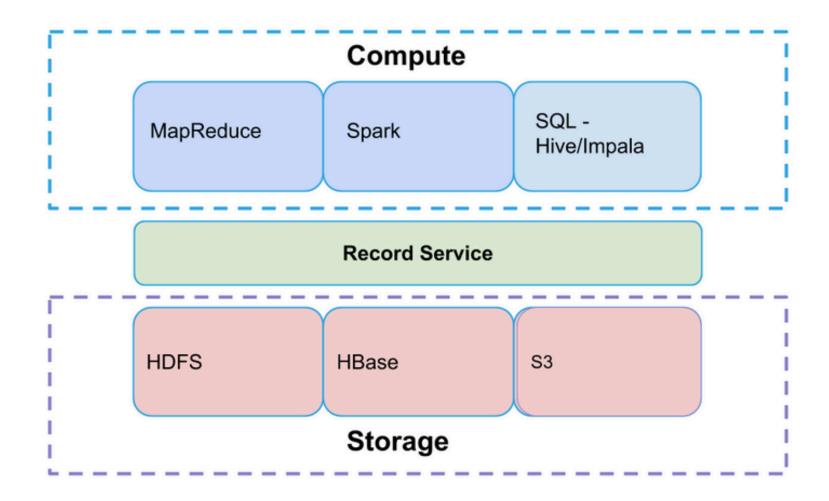


### Cloudera Record Service

#### Motivation

- Decoupling storage managers (HDFS, HBase) and compute frameworks (MapReduce, Impala, Apache Spark)
  - Greater flexibility to pick the framework that best solves your problem
  - Leads to more complexity to ensure everything works together seamlessly
- □ A new core security layer for Hadoop
  - Sits between the storage managers and compute frameworks
  - Provide a unified data access path.

## **Unified Data Access Path**





# Design

- RecordServicePlanner Generates tasks, performs authorization checks, handles metadata access.
  - □ Each task describes a work unit and the preferred locality
  - □ Called during input split generation.
  - □ Provides a layer of abstraction over the metadata services (NameNode, Hive Metastore, Sentry server)
- RecordServiceWorker Executes tasks, reads and writes to the storage layer.
  - □ Builds on Impala's highly optimized I/O scheduler and file parsers.
  - □ Returns reconstructed, filtered records in a canonical wire format.
  - □ Provides a layer of abstraction of the data stores (DataNode)

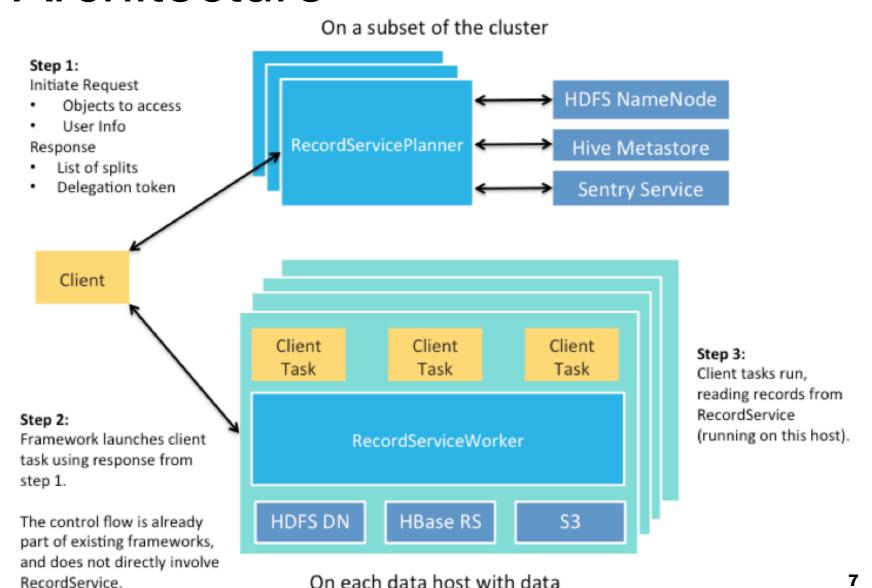


# Design

- Thrift APIs.
- Client Integration Libraries Allow easy migration to RecordService.
- Planner and Worker share only minimal state via Apache ZooKeeper
  - □ Scalability
  - □ Fault tolerance

### Architecture

RecordService.





# Review: Hadoop Execution

- Simple job execution flow
  - □ Client contacts the NameNode
    - get the block locations, represented as InputSplits
  - □ The compute framework (MapReduce/Spark) will launch the map task on the slave nodes.
  - □ Each slave node read the data (idealy locally) and perform the computation



# **Hadoop Integration**

- Clients talk to the RecordService planner
  - □ Planners run on a few nodes (default 3)
- Tasks read data from Worker (instead of directly going to the data stores)
  - Workers run on all nodes with data
    - optimize for read locality
- RecordService provides client libraries that implement the common Hadoop InputFormats



### Benefits

- Fine-grained data permissions and enforcement across Hadoop ecosystem
  - column-level permissions (projections), row-level permissions (filtering), and data masking
- Performance
  - Sits on the main data access path
  - Scale horizontally
  - □ Impala IO layer provides low-level optimizations
- Simplicity
  - □ Provides a higher level, logical abstraction for data
    - Datasets can be specified as logical names (i.e. tables or views)
    - Applications don't need to worry about differences in file formats etc.



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### Motivation

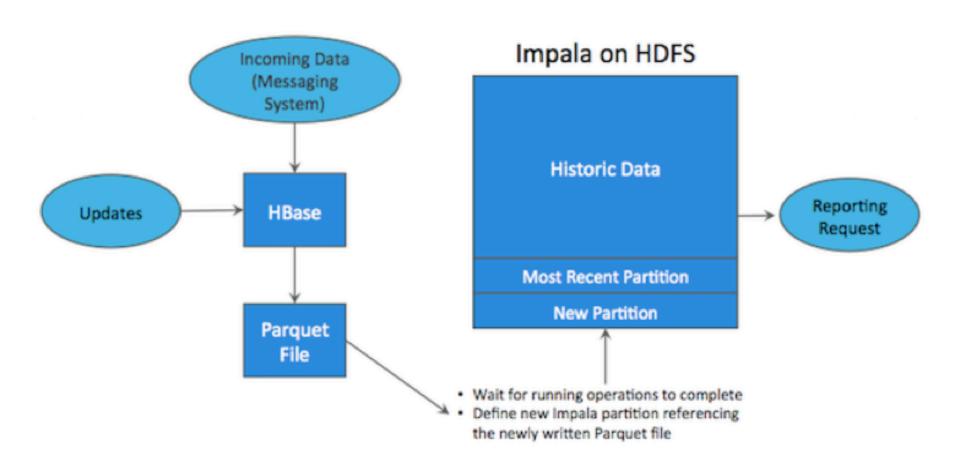
- No update on HDFS data
- In practice, may require constant change of fact or dimension tables
- HBase or Cassandra solutions
  - □ Periodic roll-over to HDFS
  - □ Complicated workflows
  - □ Problem with latency and data freshness



# Gap in Hadoop Capabilities

- Data storage and processing technologies that define the Apache Hadoop ecosystem are expansive and ever-improving
- Hybrid architectures
  - Enterprise use cases require the simultaneous availability of capabilities
    - Stitch multiple tools together
      - □ ingest and update data in one storage system
      - reorganize this data to optimize for an analytical reporting use-case served from another

# A complex hybrid architecture designed to cover gaps in storage system capabilities





# Kudu Properties

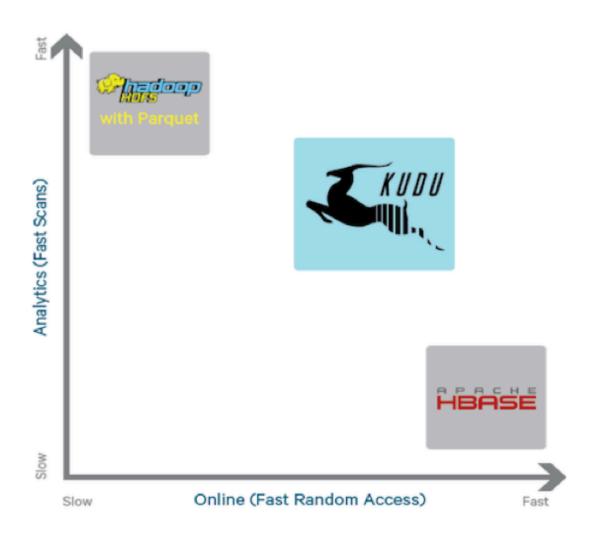
- Provide fast analytical and real-time capabilities
- Efficient utilization of modern CPU and I/O resources
- Do updates in place
- Simple and evolvable data model



### Benefits

- Fast Data access
  - □ Nearly as fast as raw HDFS for scans
  - □ Nearly as fast as HBase for random access
- High CPU performance
  - □ Single-column scan rate 10-100x faster than HBase
  - □ Needed to take advantage of RAM and Flash
- High IO efficiency
  - True column store with type-specific encodings
  - ☐ Allow efficient analytics when only certain columns are accessed

## **Fast Data Access**





# **Basic Design**

- A storage system for tables of structured data in user view
  - □ well-defined schema
  - predefined number of typed columns
  - primary key composed of one or more of its columns
- Tables are composed of tablets/partitions
  - □ Replicated
  - □ Tablets are typically tens of gigabytes, and an individual node typically holds 10-100 Tablets.



# Basic Design (cont'd)

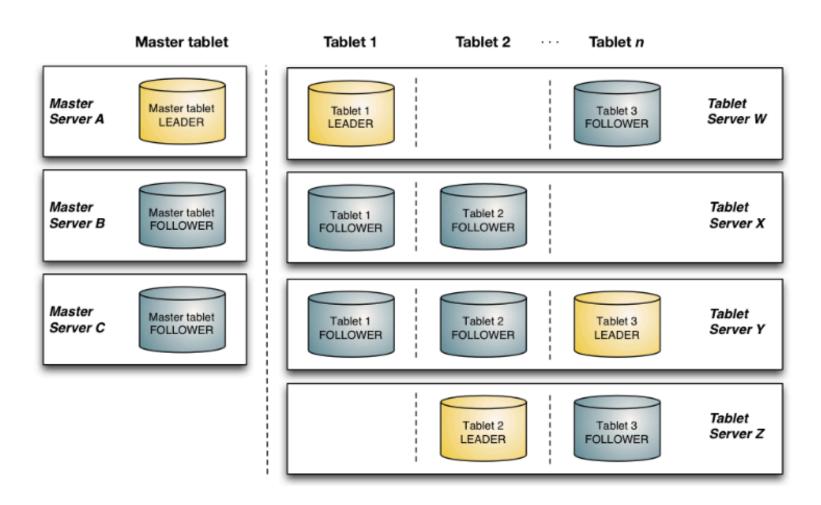
- Master process responsible for managing the metadata
  - Describes the logical structure of the data stored in Tablet Servers
  - Acts as a coordinator when recovering from hardware failure
  - Keeps track of which tablet servers are responsible for hosting replicas of each Tablet.
  - Multiple standby master servers can be defined to provide high availability.



# Basic Design (cont'd)

- Log-structured storage
  - □ Updates, inserts, and deletes are temporarily buffered in memory before being merged into persistent columnar storage
  - Constantly performing small maintenance operations such as compactions
    - Protect against spikes in query latency
- Direct APIs in C++/Java
  - allow for point and batch retrieval of rows
  - □ writes, deletes, schema changes

### **Netowork Architecture**





#### **Architecture**

- Cluster roles
  - Single master server: metadata
    - Replicated for fault tolerance
  - Multiple tablet server: data
- Partitioning
  - Tables are horizontally partitioned into tablets
  - Partition schema is made up of zero or more hash-partitioning rules followed by an optional range-partitioning rule
    - Trade off between query parallelism and query concurrency
- Replication
  - □ Raft Consensus algorithm
    - Leader replica handles replication details



#### Kudu Master

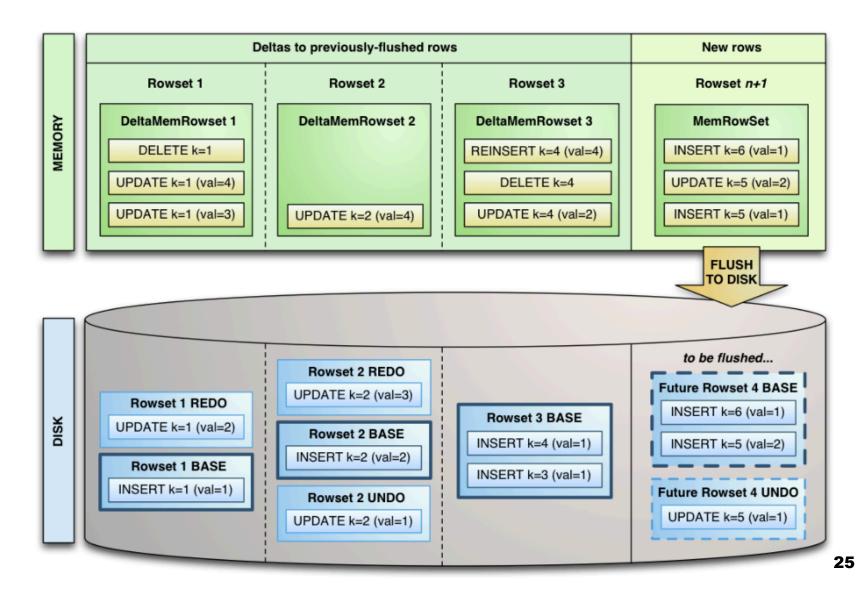
- Catalog Manager
  - □ Single tablet table for catalog information
    - Current version of the table schema
    - State of the table (creating, running, deleting etc.)
    - Set of tablets for this table
  - □ A full write-through cache of the catalog is kept in memory



### Cluster Co-ordination

- Each tablet server is configured with a list of host names
- On startup, tablet server register with Master and sends report
  - □ Future reports are incremental
- Tablet servers are responsible for all replica configuration
  - Raft Leader replica decides replication
  - Master is only observer
    - Suggests configuration change to lead replica of the tablet, on replica count change
- Clients query the Master for tablet location information

# Data Model: single-tablet view





# **Tablet Storage**

- Tablets are subdivided into RowSets
  - MemRowSets
    - Stores all recently inserted rows
    - flushed to disk periodically
  - DiskRowSets
    - base data
      - Column representation of rows
      - Immutable once flushed
    - delta stores
      - Updates and deletes
      - □ In-memory *DeltaMemStores*, or on-disk *DeltaFiles*
    - delta flush
      - □ Flushing a DeltaMemStore, a new empty store is swapped in while the existing one is written to disk and becomes a DeltaFile.



#### Data access

#### Read

- □ Seeks the target column to the correct row offset
- Copies cells from the source column into output row batch
- Consult the delta stores to see if any later updates have replaced cells with newer versions

#### Insert

- □ Need to consult all DiskRowSets before save to MemRowSets, to enforce primary key uniqueness constraints
  - Unlike many other NoSQL stores, where INSERT = UPSERT
  - Achieved by bloom filter on keys, and key range partition for each DiskRowSet

#### Lazy materialization

□ Read columns which have associated range predicates before reading any other columns.



### Maintenance

- Delta compaction
  - Where there're large number of deltas, merge back into the base data columns
- RowSet compaction
  - Merge two or more DiskRowSets into a sorted stream of output rows, written back to a new DiskRowSets
    - Remove deleted
    - Reduce key range overlap
- Scheduling maintenance
  - A pool of maintenance threads within the tablet server, running all the time
  - Run small units of flush/compaction work, adapt to workload



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# Background

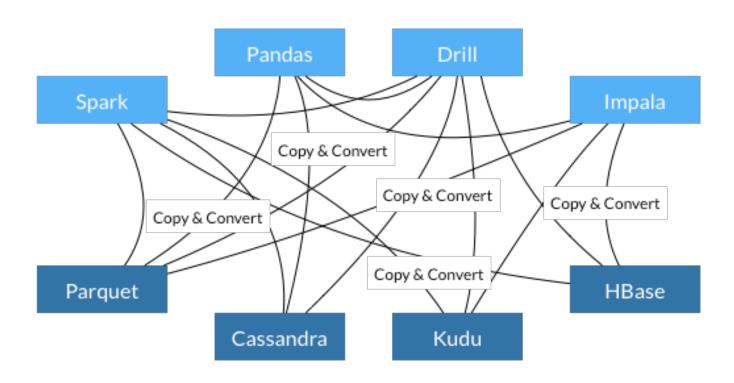
- In-memory data systems
- Columnar storage
- Common approach, varied implementation

# **Key Benefits**

- Columnar memory-layout, O(1) random access
- Standard data interchange format
  - □ Sharing of data in a seamless and frictionless fashion between systems and processes.
  - Reduce the large amount of CPU cycles serializing and de-serializing data, to convert data between various formats
  - customers are able to deploy these systems in conjunction with each other without incurring any overhead
- A flexible structured data model supporting complex types like flat tables, json etc

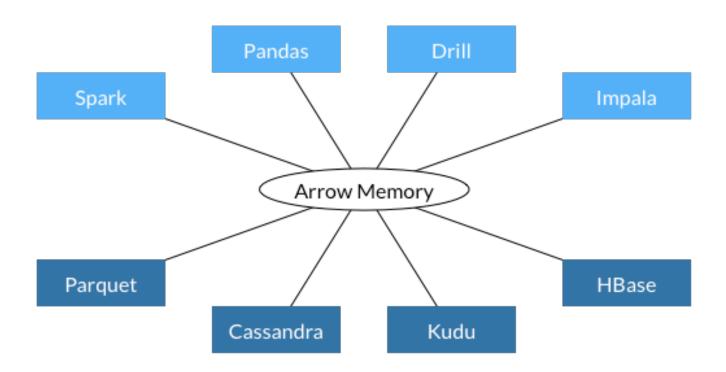


## Without Arrow





## With Arrow

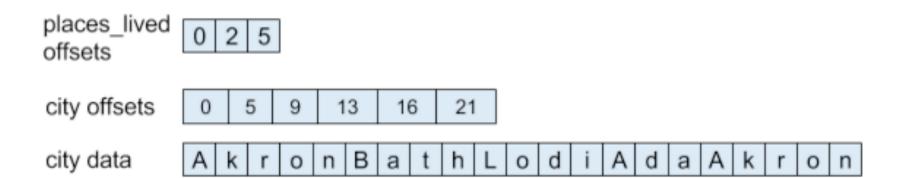




# What Arrow array looks like

```
people = [
  name: 'mary', age: 30,
  places_lived: [
    {city: 'Akron', state: 'OH'},
    {city: 'Bath', state: OH'}
  name: 'mark', age: 33,
  places_lived: [
    {city: 'Lodi', state: 'OH'},
    {city: 'Ada', state: 'OH'},
    {city: 'Akron', state: 'OH}
```

Look at the people.places\_lived.city values:



# Thank you!