

# Apache HBase Application Archetypes

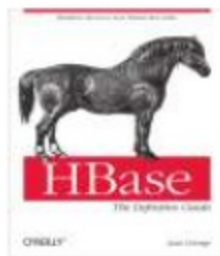
Lars George | @larsgeorge | Cloudera EMEA Chief Architect | HBase PMC  
Jonathan Hsieh | @jmhsieh | Cloudera HBase Tech lead | HBase PMC  
HBaseCon 2014  
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# About Lars and Jon

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## Lars George

- EMEA Chief Architect  
@Cloudera
  - Apache HBase PMC
  - O'Reilly Author of *HBase – The Definitive Guide*
- Contact
  - lars@cloudera.com
  - @larsgeorge



## Jon Hsieh

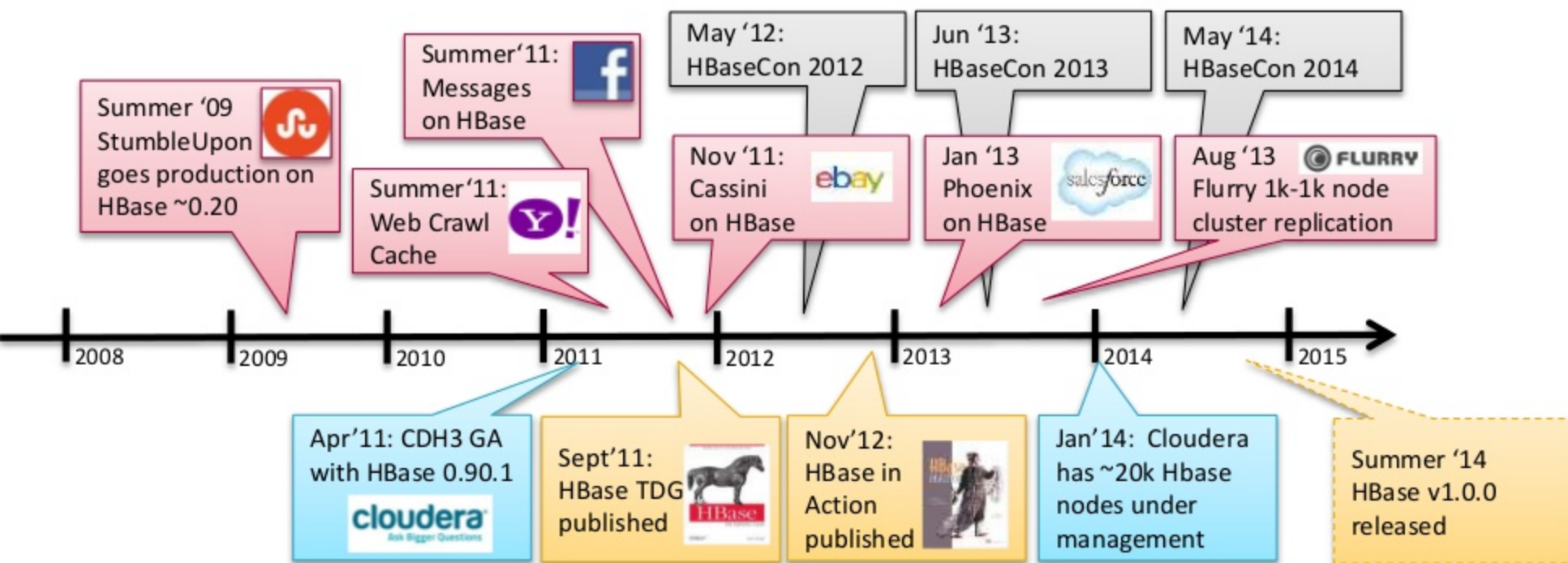
- Tech Lead HBase Team  
@Cloudera
  - Apache HBase PMC
  - Apache Flume founder
- Contact:
  - jon@cloudera.com
  - @jmhsieh

# About Supporting HBase at Cloudera

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- Supporting Customers using HBase since 2011
  - HBase Training
  - Professional Services
- Team has experience supporting and running HBase since 2009
  - 8 committers on staff
  - 2 HBase book authors
- As of Jan 2014, ~20,000 HBase nodes (in aggregate) under management
- Information in this presentation is either aggregated customer data or from public sources.

# An Apache HBase Timeline





# Apache HBase “Nascar” Slide



# Outline

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- Definitions
- Archetypes
  - The Good
  - The Bad
  - The Maybe
- Conclusion

# Definitions

A vocabulary for HBase Archetypes

# Defining HBase Archetypes

- There are a lot of HBase applications
  - Some successful, some less so
  - They have common architecture patterns
  - They have common tradeoffs
- **Archetypes** are common architecture patterns
  - Common across multiple use-cases
  - Extracted to be repeatable
- Our Goal: Define patterns à la “Gang of Four” (Gamma, Helm, Johnson, Vlissides)





# So you want to use HBase?

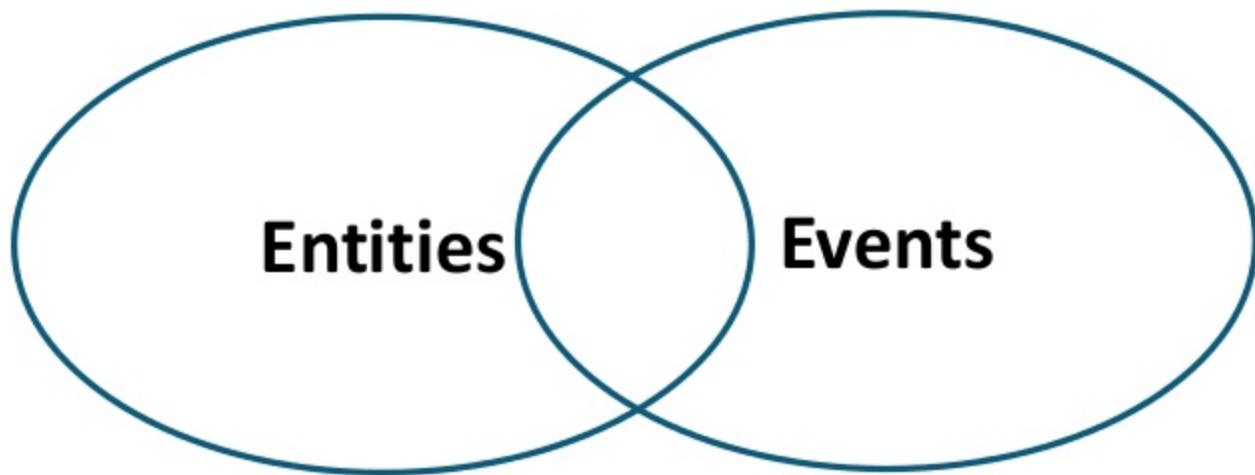
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- What data is being stored?
  - Entity data
  - Event data
- Why is the data being stored?
  - Operational use cases
  - Analytical use cases
- How does the data get in and out?
  - Real time vs. Batch
  - Random vs. Sequential

# What is being stored?

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There are primarily two kinds of big data workloads. They have different storage requirements.



# Entity Centric Data

- **Entity** data is information about **current state**
  - Generally real time reads and writes
- Examples:
  - Accounts
  - Users
  - Geolocation points
  - Click Counts and Metrics
  - Current Sensors Reading
- Scales up with # of Humans and # of Machines/Sensors
  - Billions of distinct entities



# Event Centric Data

- **Event** centric data are time-series data points recording successive points spaced over time intervals.
  - Generally real time write, some combination of real time read or batch read
- Examples:
  - Sensor data over time
  - Historical Stock Ticker data
  - Historical Metrics
  - Clicks time-series
- Scales up due to finer grained intervals, retention policies, and the passage of time





# Events about Entities

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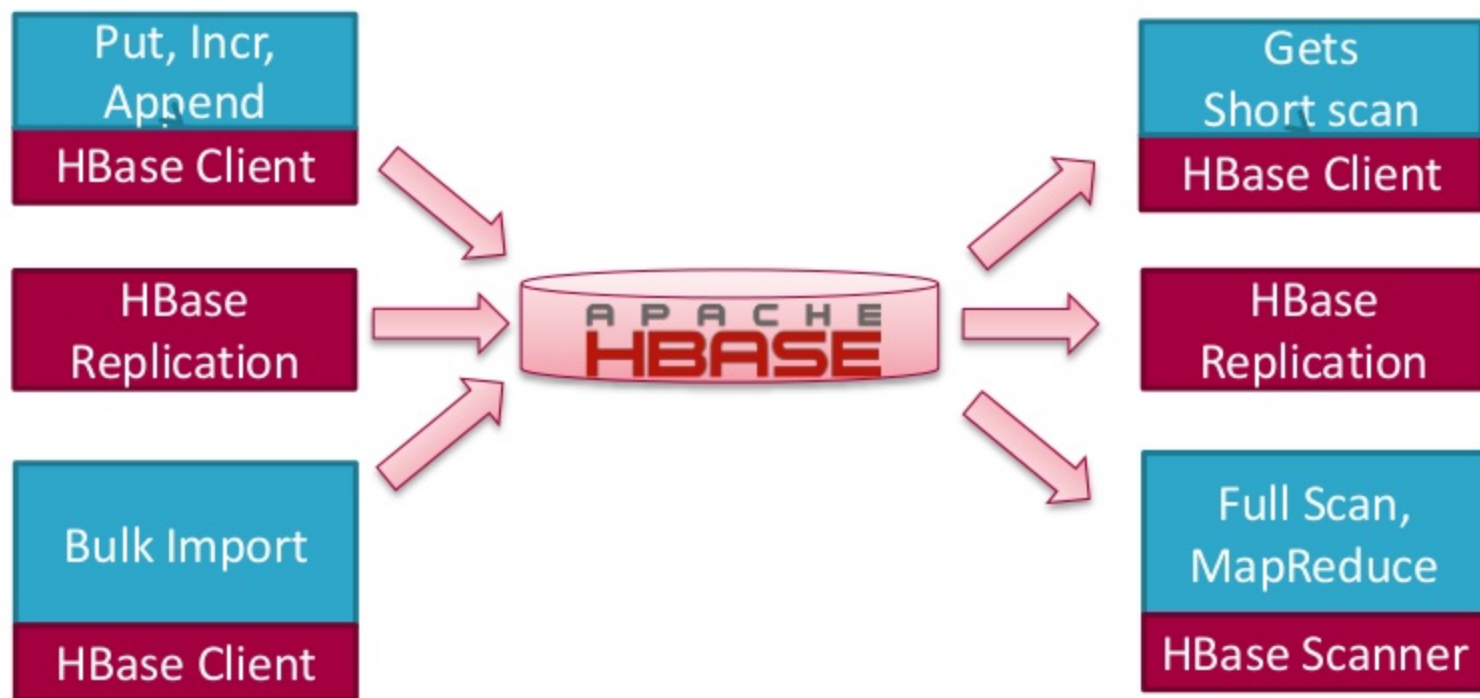
- Majority Big Data use cases are dealing with event-based data
  - **|Entities| \* |Events| = Big data**
- When you ask questions, do you hone in on entity first?
- When you ask questions, do you hone in on time ranges first?
- Your answer will help you determine where and how to store your data.

# Why are you storing the data?

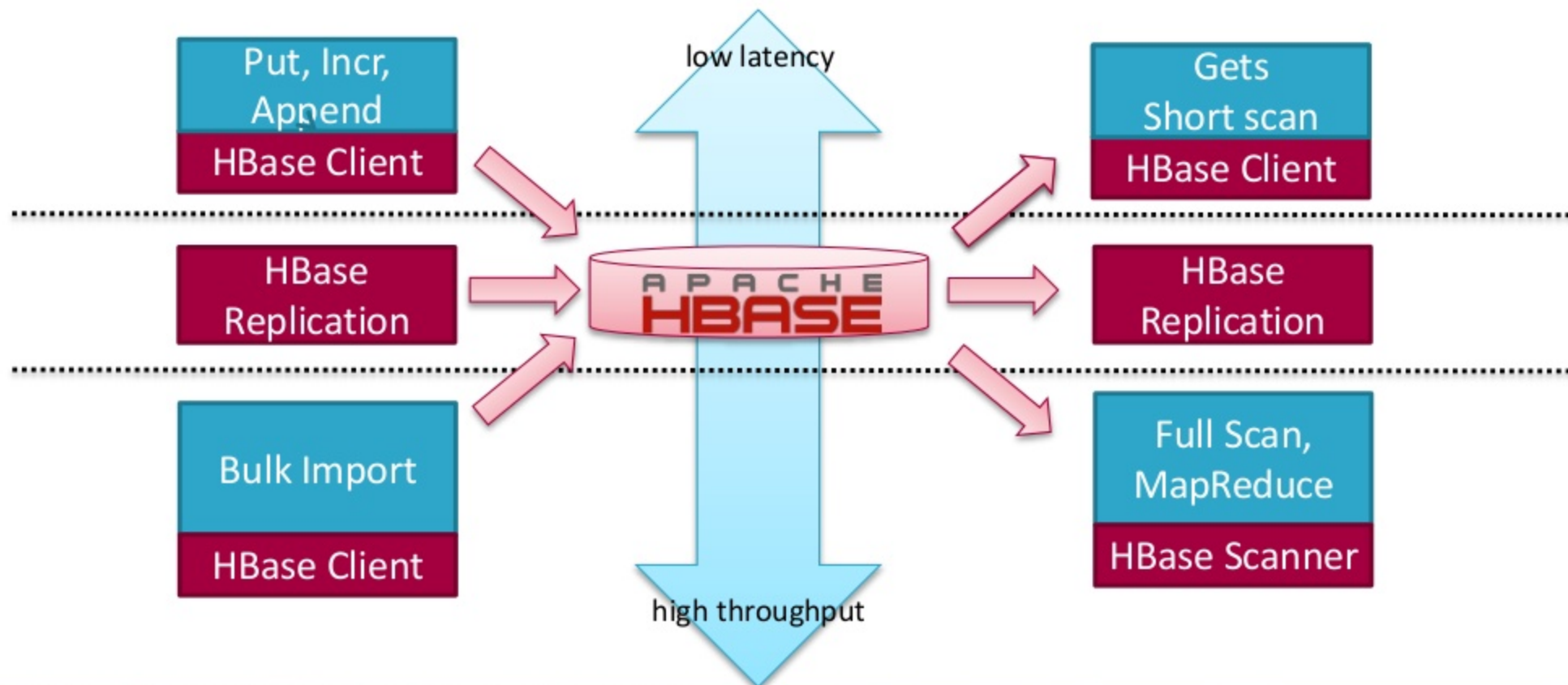
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- So what kind of questions are you asking the data?
- Entity-centric questions
  - Give me everything about entity **e**
  - Give me the most recent event **v** about entity **e**
  - Give me the **n** most recent events **V** about entity **e**
  - Give me all events **V** about **e** between time **[t1,t2]**
- Event and Time-centric questions
  - Give me an aggregates on each entity between time **[t1,t2]**
  - Give me an aggregate on each time interval for entity **e**
  - Find events **V** that match some other given criteria

# How does data get in and out of HBase?



# How does data get in and out of HBase?





# What system is most efficient?






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- It is all physics
- You have a limited I/O budget
  - Use all your I/O by parallelizing access and read/write sequentially.
  - Choose the system and features that reduces I/O in general
- Pick the systems best for your workload













IOPs/s/disk


















# The physics of Hadoop Storage Systems

Workload	HBase	HDFS
Low latency	 ms, cached	 mins, MR  seconds, Impala
Random Read	 primary index	 index?, small files problem

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Random Write	 log structured	 not supported
Sequential Write	 HBase overhead  bulk load	 minimal overhead
Updates	 log structured	 not supported