



# The Rise of Open Source Data Platforms: An Insider's View

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DataCon.TW 2017

Jonathan Hsieh, Cloudera

# Who Am I?



- **Tech Lead, Infrastructure @ Cloudera**
- **With Cloudera since 2009**
  - **Engineering Manager/Tech Lead** of Cloudera's HBase team
  - **Apache HBase** committer / PMC
  - **Apache Flume** founder / PMC
- **U of Washington:**
  - Research in Distributed Systems

# Evolution

- Big Data
- Data Science
- Machine Learning
- What's Next

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# The Rise of Open Source Big Data

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# Let's go back to 2006...

Mar '04: Google  
Google MapReduce  
OSDI '04

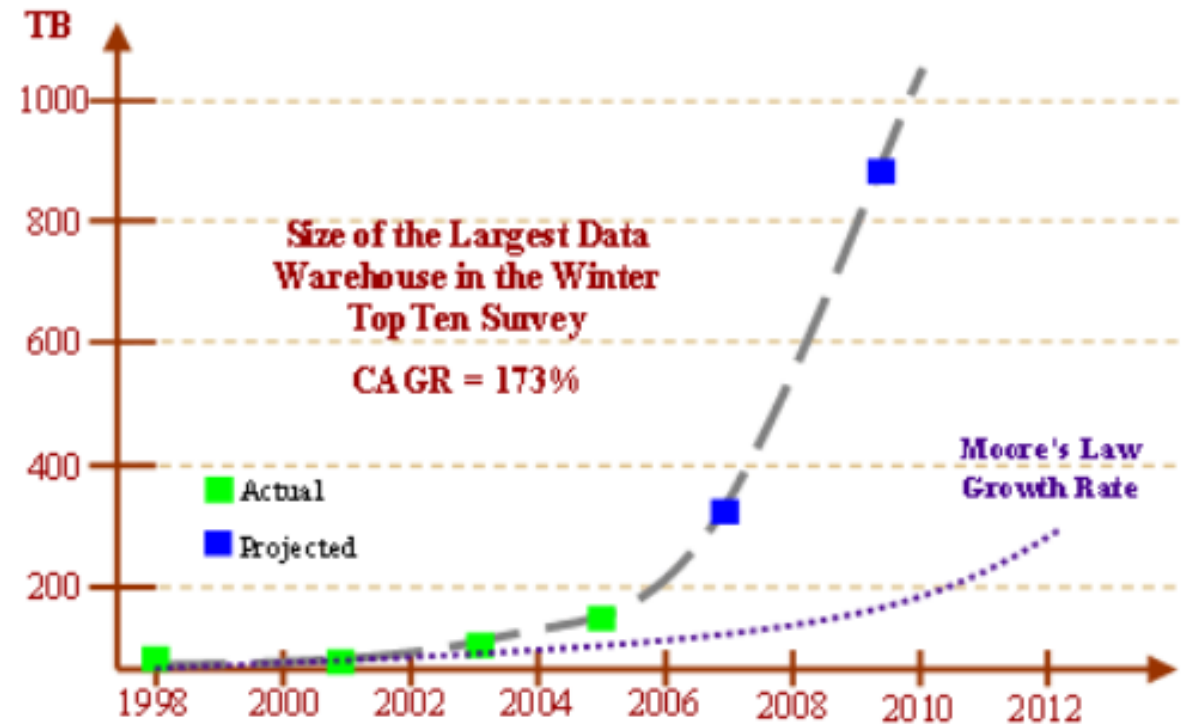
Oct '03 Google  
Google GFS  
SOSP '03



# Data Volumes were growing faster and faster

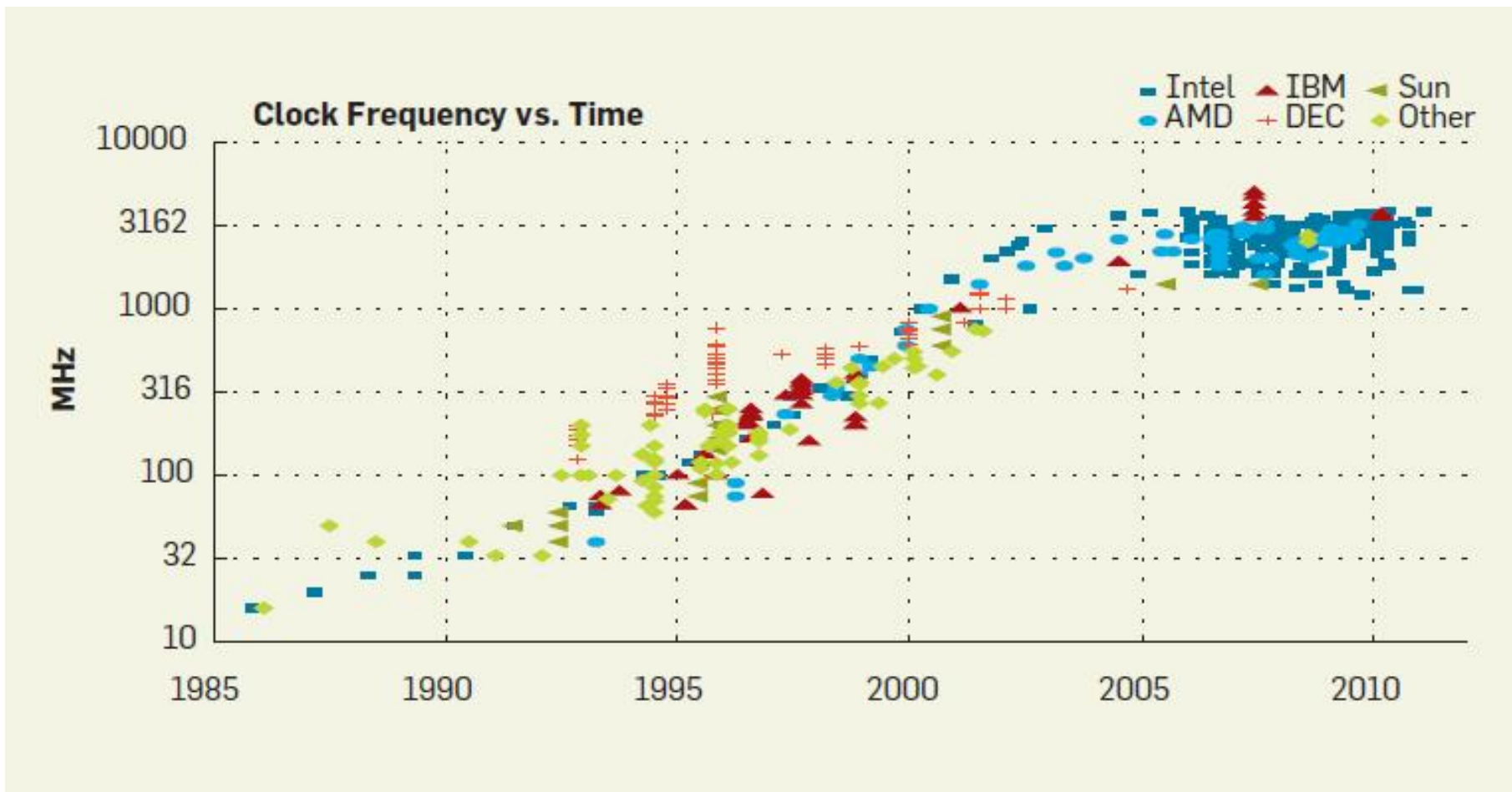
“Every two days we create as much information as we did from the dawn of civilization up until 2003.”

-- Eric Schmidt  
Google CEO, 2010



Source: Richard Winter, “Why are data warehouses growing so fast?” April 2008

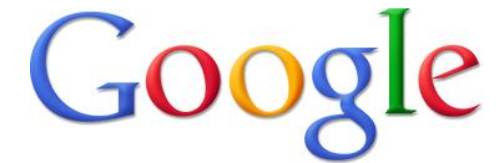
# CPUs weren't getting any faster



<http://cacm.acm.org/magazines/2012/4/147359-cpu-db-recording-microprocessor-history/abstract>

# Google's "Big Data" problem

- How to build an index of all the web's data?
- An algorithm:
  - Download the web (all of it)
  - Analyze it to build the index
    - Word count, page rank
    - Keep track of the links
  - Serve the index





# What is Big Data?

1. **Collect** the data
2. **Count** the data
3. **Report** the results

# Approach: Use Distributed Systems

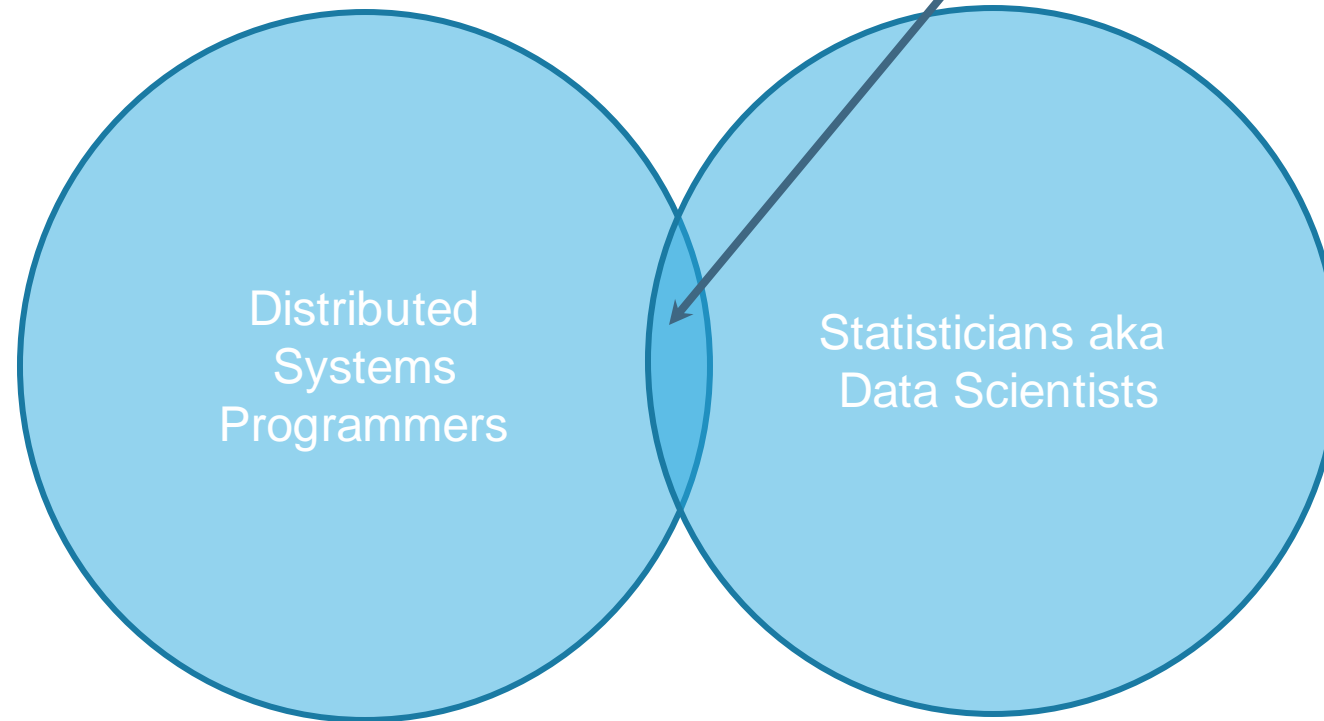
Instead of one expensive machine, we spread the work across many commodity machines

- **Horizontally Scalable**
  - More hardware means more capacity without diminishing returns
  - Deal with the **volume** of data
- **New Challenge: Reliability**
  - **Fault tolerant:** Tolerate failures, fail gracefully, have recovery mechanism
  - **Highly-Available:** Tolerate failures, recover immediately

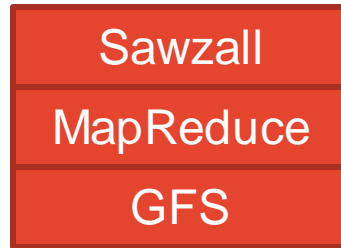
# Distributed systems are hard!



Unicorns



# Google built distributed storage and processing systems



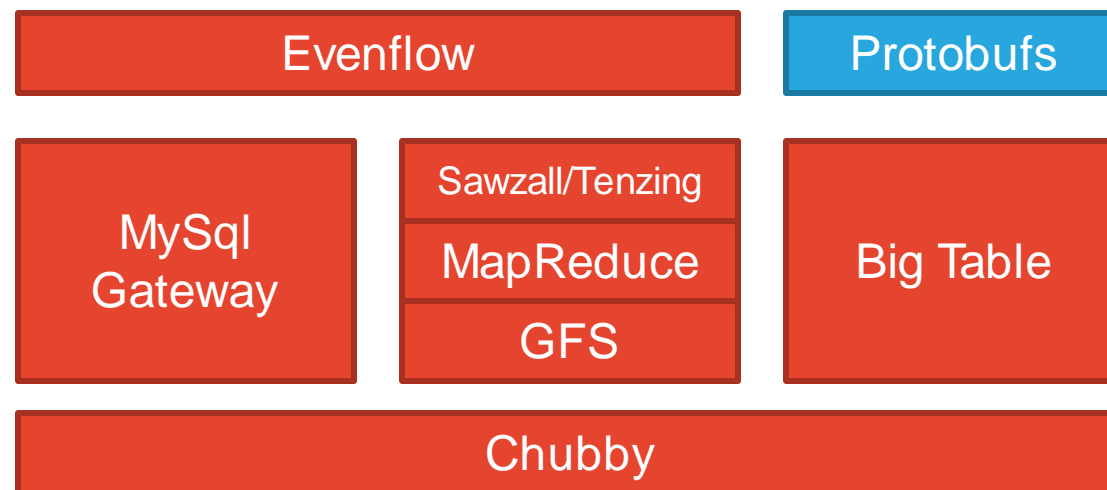
## Storage:

- Google File System (GFS)
- Store PB's of data

## Processing:

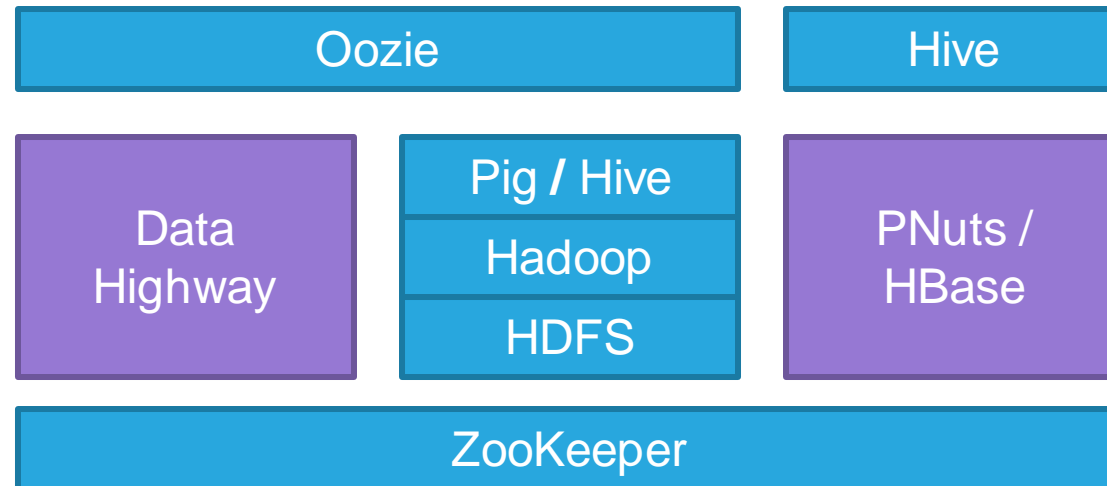
- MapReduce: Users specify a *map()* and a *reduce()* function
  - *map*:  $K_1, V_1 \rightarrow \text{list } K_2, V_2$
  - *reduce*:  $K_2, \text{iter}(V_2) \rightarrow \text{list}(K_3, V_3)$
- Sawzall: A DSL on top of MR

# Google built a Big Data Platform



- Feeding data to GFS/MapReduce
- Serving data generated by it
- Coordinating the different systems
- Managing chains of jobs.
- Managing metadata

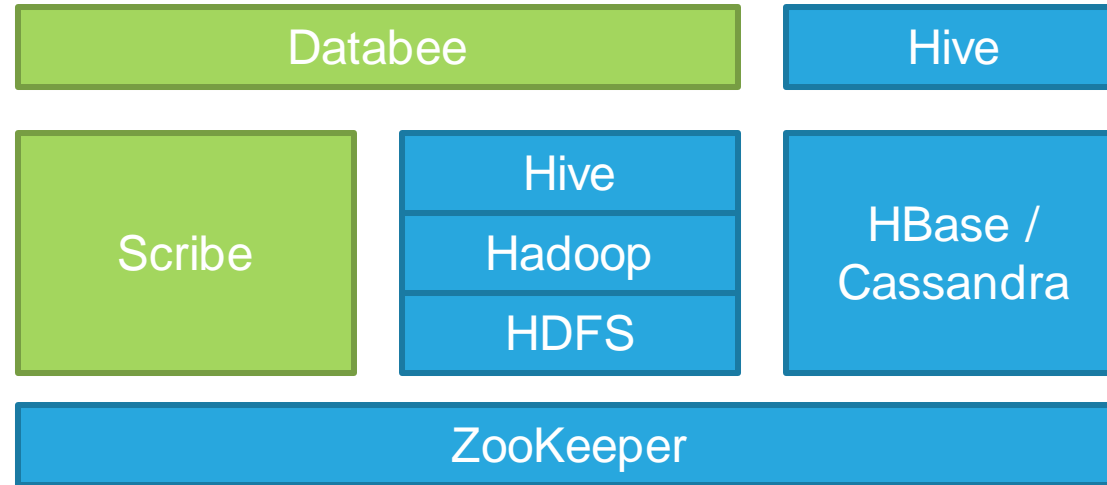
# Yahoo built a Big Data Platform



- Invested heavily in Hadoop (MapReduce clone) and HDFS (GFS clone)
- Created and open-sourced Oozie, Pig, and Zookeeper

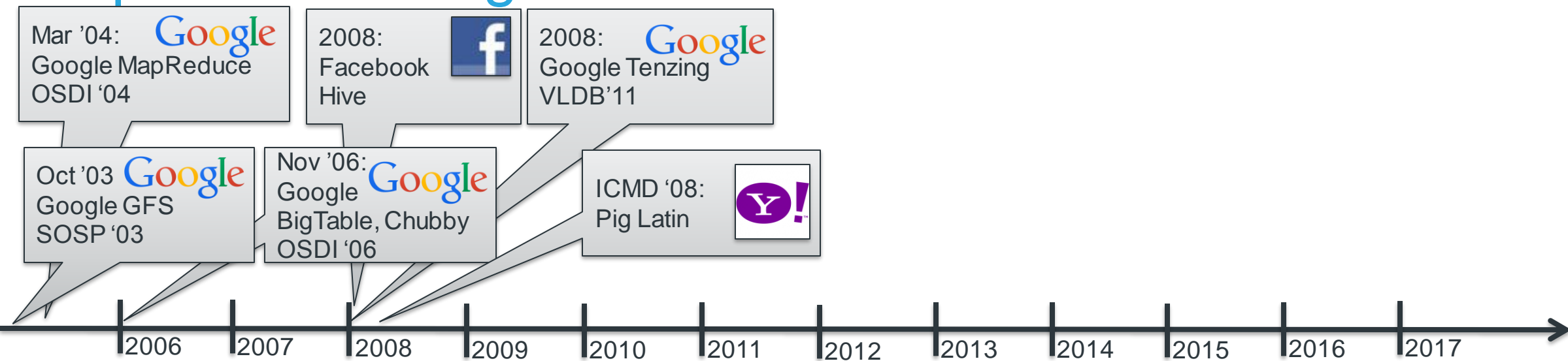
# Facebook built a Big Data Platform

**facebook**



- Created and open-sourced Scribe, Hive, and Cassandra
- Invested heavily on HBase

# Open Source Big Data Platform Timeline



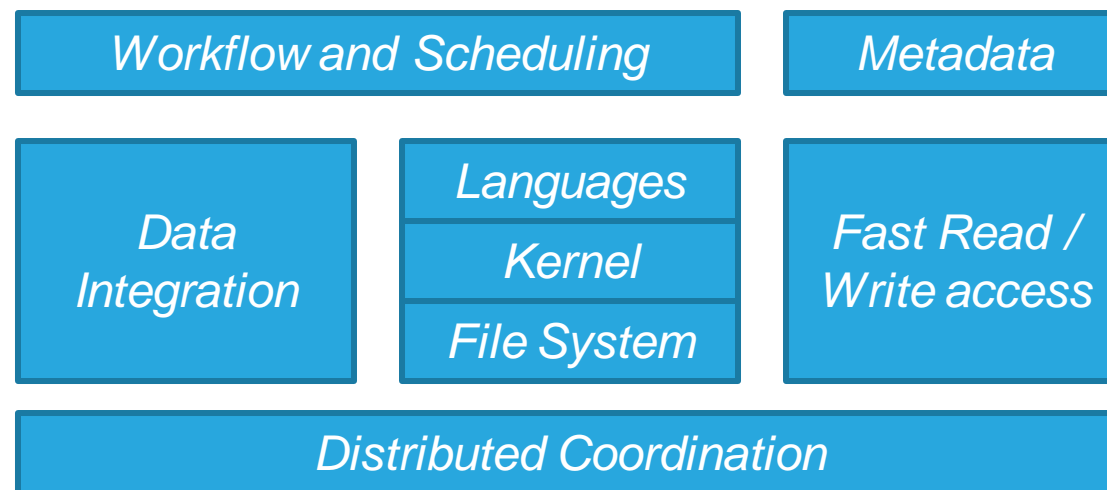


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**The future is already here — it's just not  
very evenly distributed.  
-- William Gibson**

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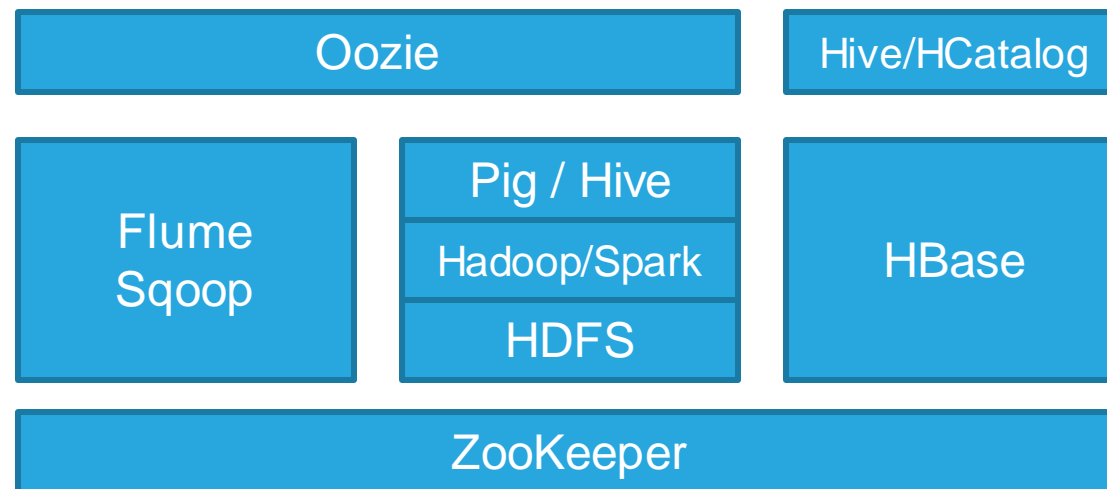
# The Core Big Data Platform



- Hadoop is an open MapReduce kernel
  - It is like the kernel of a Cluster Operating System.
- Each “friend” is a distributed service that has a similar workstation tool or abstraction.

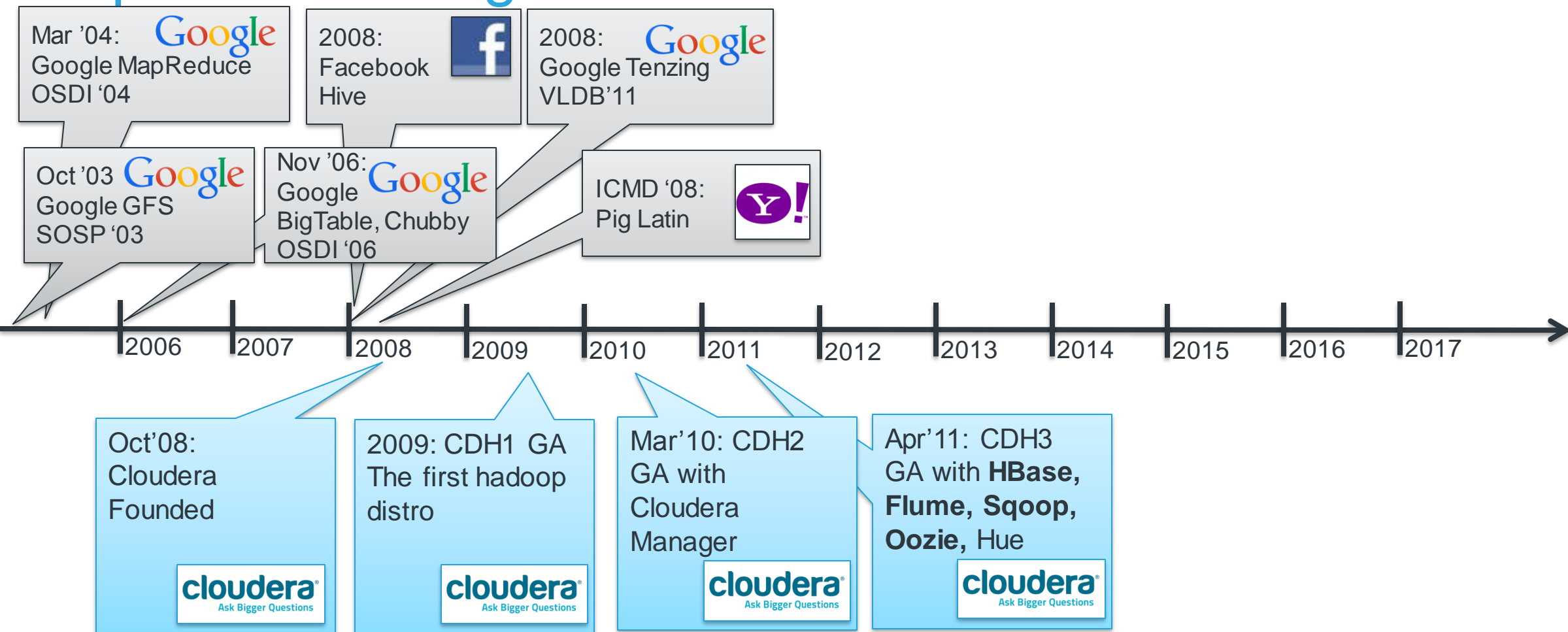
# Cloudera built a Big Data Platform

**cloudera**



- Created and open-sourced Flume and Sqoop
- Core stack is free and available for all to use

# Open Source Big Data Platform Timeline



# Functions move to Open Source

Function	Google	Yahoo!	Facebook	The Rest of Us
File Storage	GFS => Colossus	HDFS	HDFS	HDFS
Record storage (NoSQL)	BigTable => Megastore => Spanner	PNUTS => HBase	HBase	HBase Kudu
Batch processing	Google MapReduce	Hadoop MapReduce	Hadoop MapReduce	Hadoop MapReduce
Batch query	Sawzall, Tenzing, FlumeJava	Pig	Hive	Pig, Hive, Crunch
Resource Management	Borg => Omega	MR => Hadoop YARN	MR => Corona	Hadoop YARN
Ingest	EvenFlow Custom MySQL Proxy	Data Highway	Scribe / Calligraphus Custom MySQL proxy	Sqoop Flume
Coordination	Chubby	Zookeeper	Zookeeper	Zookeeper

# Limitations and next challenges

- High throughput, high latency batch processing with Map Reduce
- Deploy on prem
- Coupling storage and compute for **locality** is critical

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# The Rise of Data Science

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# Let's go back to 2012...



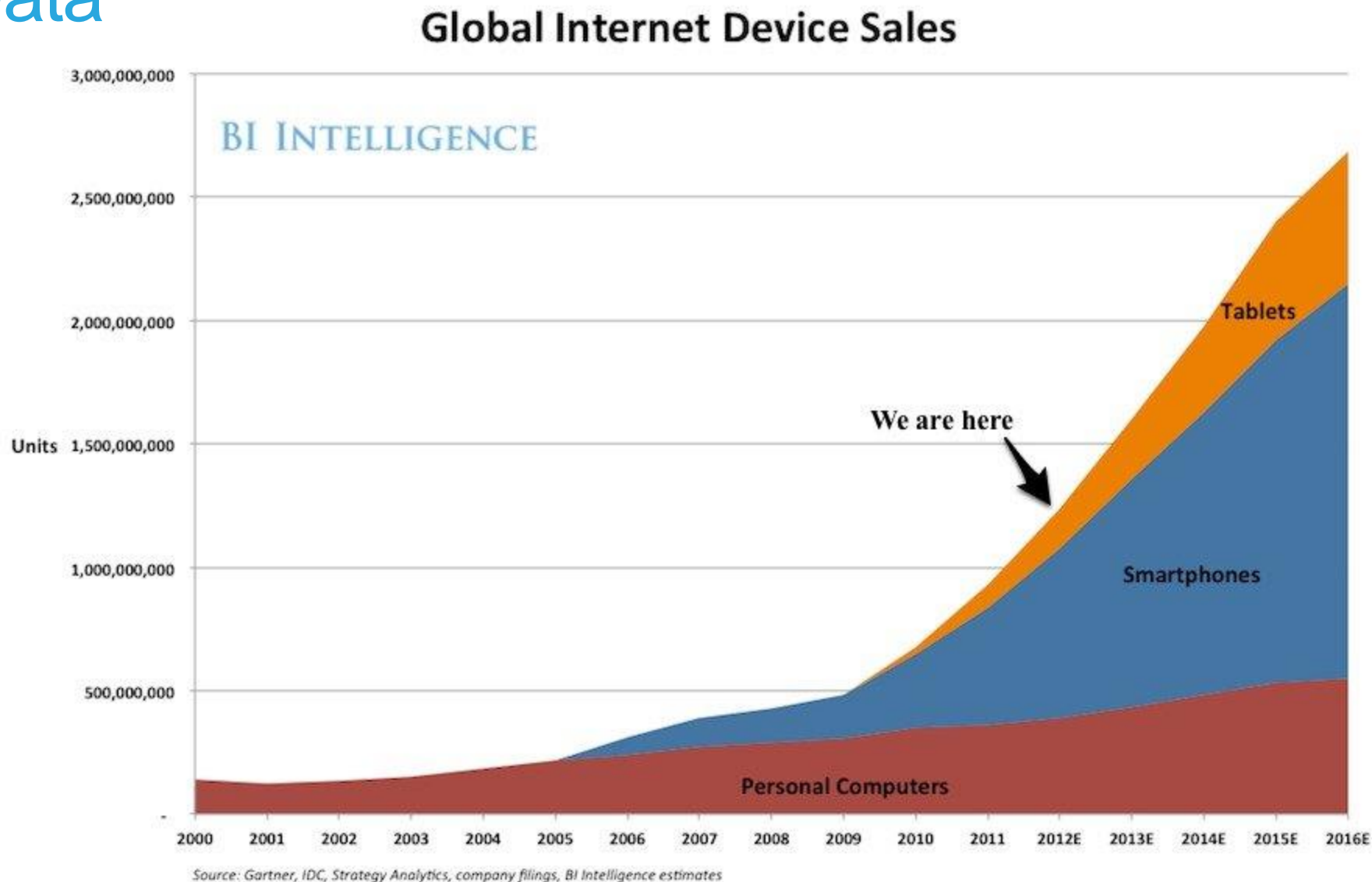
Apr'11: CDH3  
GA with HBase,  
Flume, Sqoop,  
Oozie, **HUE**





# Rise of Mobile Data

- Apple iPhone released in 2007
- Google Android released in 2008
- By 2012, smart phone sales volume outgrew PCs
- The Internet and Social Networking moved to mobile



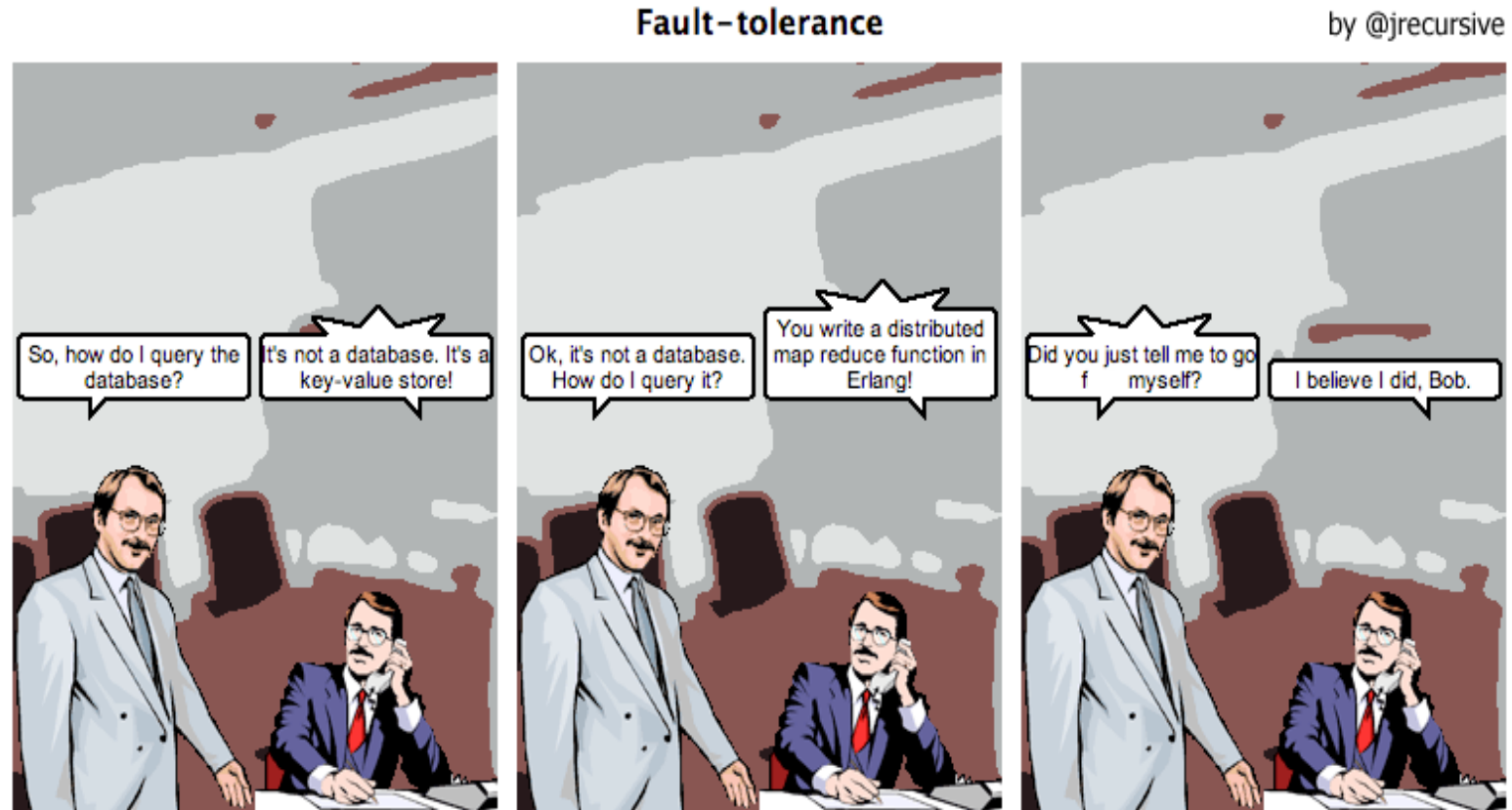
# What is Big Data?

1. **Collect** the data
2. **Count** the data
3. **Report** the results

# Big Data: Simplify and remove features to enable scaling

- Scalable and Fault tolerant first
- Exclude features to simplify
  - No transactions
  - No Schema
  - No Joins

**This is still hard to do!**



# Limitations and next challenges

- High throughput, high latency batch processing with MapReduce
- Low-latency SQL access with Impala
- Full text search with Solr
- Incremental processing with Spark
- Near-real time stream processing with Spark streaming
- Cluster is secured and governed for multitenancy
- Deploy on prem. Coupling storage and compute for **locality** is critical.
- Cloud is available for good for experimentation

# What is Big Data?

1. **Collect** the data
2. **Count** the data
3. **Report** the results

# What is Data Science?

1. **Wrangle** the data
2. **Explore** the data
3. **Interact** with the results

# Data Science and Analytics tools for your data



Interactive distributed SQL query engine



In-memory processing and stream processing



User friendly UX for queries



Fast storage for analytics

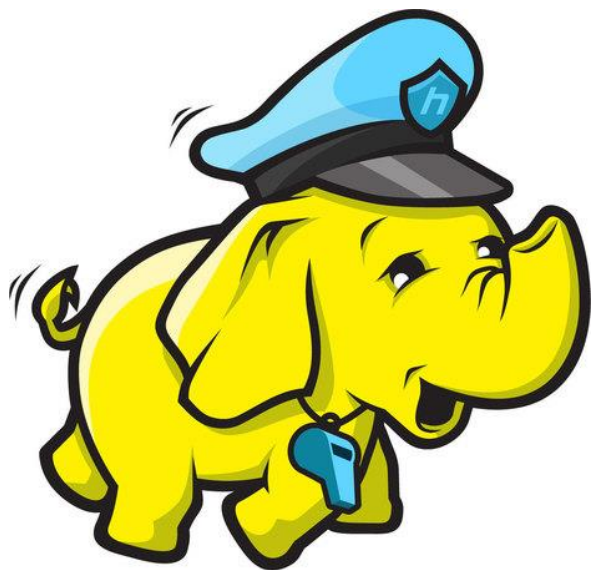


Incremental distributed full text search



R/Scala/Python Data Science Notebook for big data platforms (now CDSW)

# Sharing data requires security and governance



Kerberos Strong authentication  
Throughout the data platform



Optimization based on  
historical queries  
(now Navigator Optimizer)



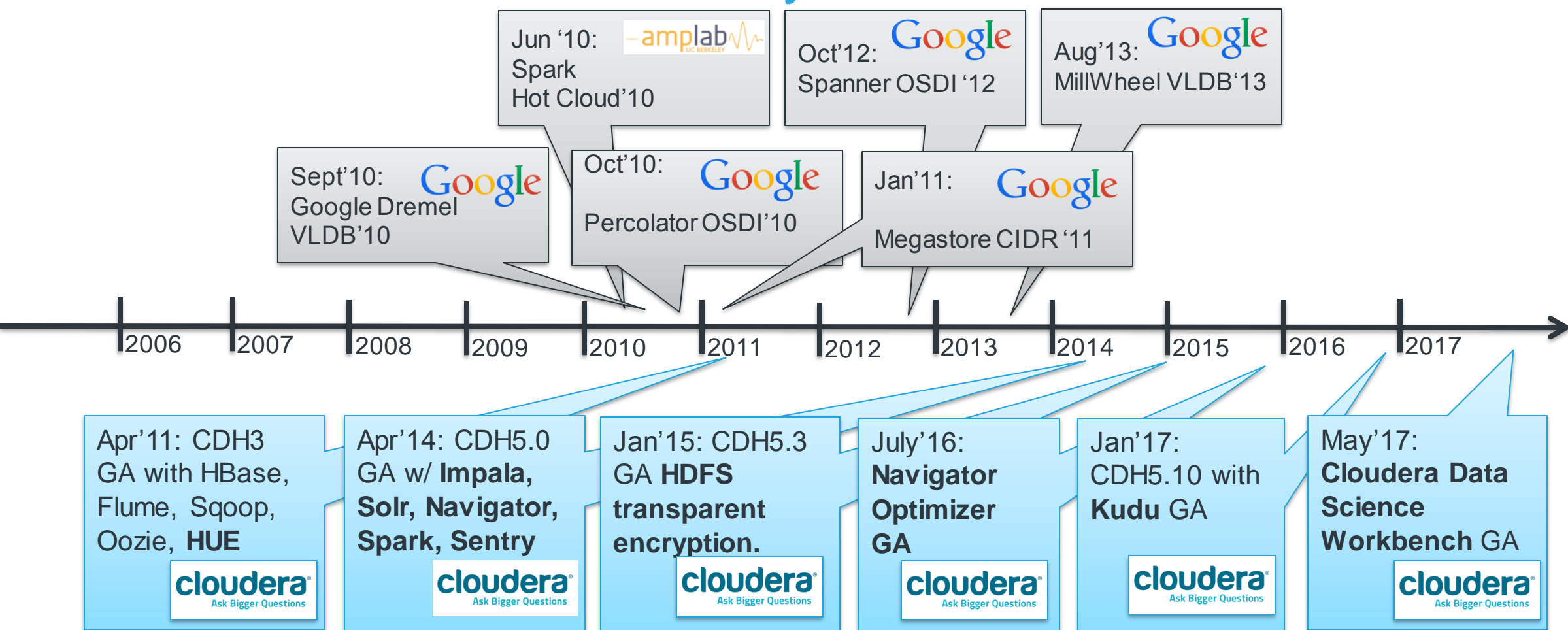
Lineage and auditing of data  
operations.



Strong Encryption for data at rest  
(now Navigator Encrypt + KeyTrustee)



# The Data Science and Analytics Stack



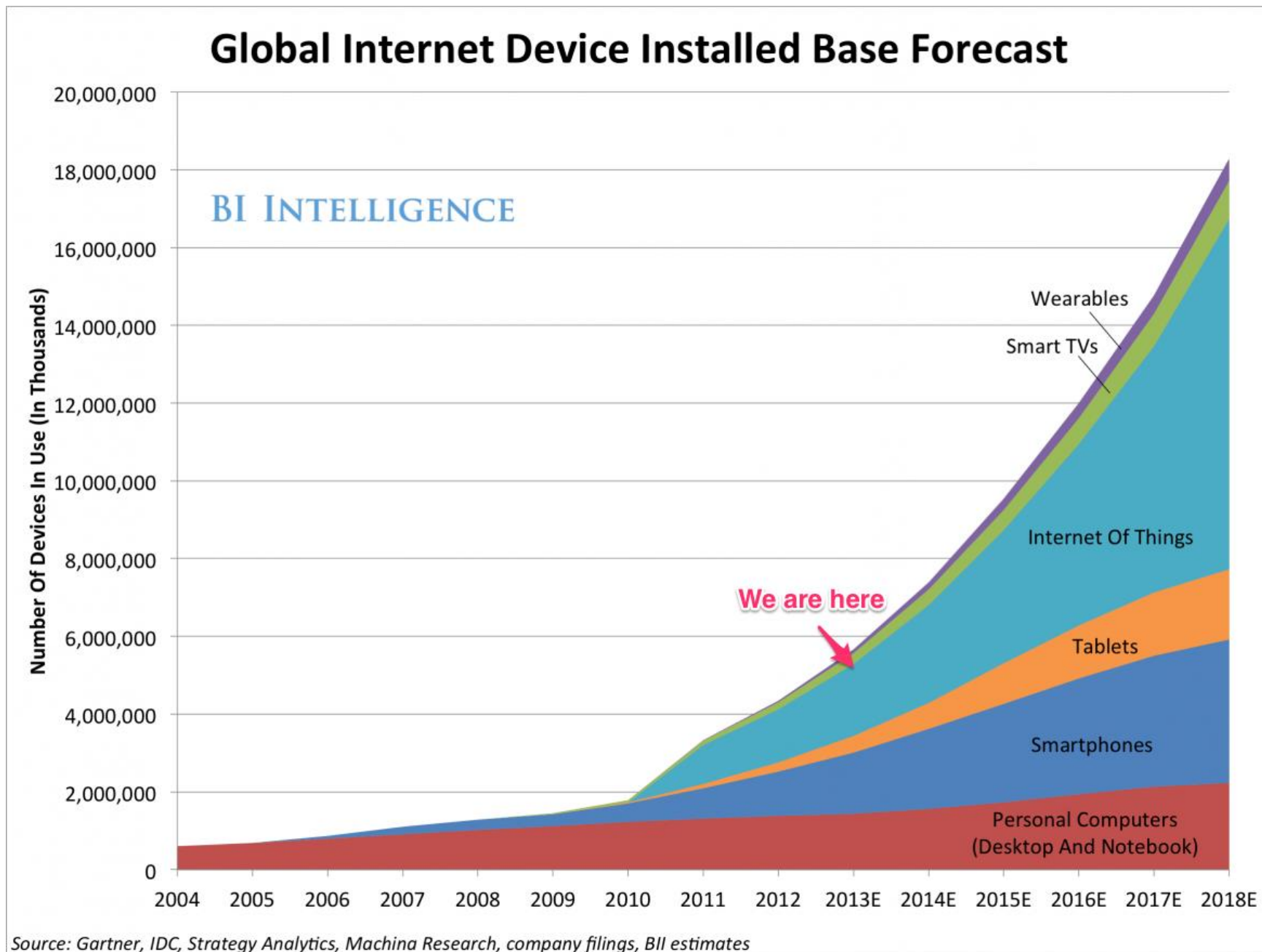
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# The Present: Enabling Machine Learning

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# Even More Data

- Sensors and IoT
  - Uber / Lyft / Didi
  - Fitbit / iWatch
- Mobile payments: Paypal, Square, AliPay, WePay



# What is Data Science?

1. **Wrangle** the data
2. **Explore** the data
3. **Interact** with the results

# What is Machine Learning?

1. **Wrangle** the data
2. **Model** the data
3. **Predict** the results

# More data usually beats better algorithms

- Use simple algorithms to learn about data from data
  - Supervised ML with training data
  - Unsupervised ML algorithms for clustering
  - Deep Neural nets that learn data using data.



## EXPERT OPINION

Contact Editor: **Brian Brannon**, [bbrannon@computer.org](mailto:bbrannon@computer.org)

## The Unreasonable Effectiveness of Data

Alon Halevy, Peter Norvig, and Fernando Pereira, *Google*

**E**ugene Wigner's article "The Unreasonable Effectiveness of Mathematics in the Natural Sciences"<sup>1</sup> examines why so much of physics can be neatly explained with simple mathematical formulas

behavior. So, this corpus could serve as the basis of a complete model for certain tasks—if only we knew how to extract the model from the data.

### Learning from Text at Web Scale

The biggest successes in natural-language-related

# THE DATA SCIENCE HIERARCHY OF NEEDS

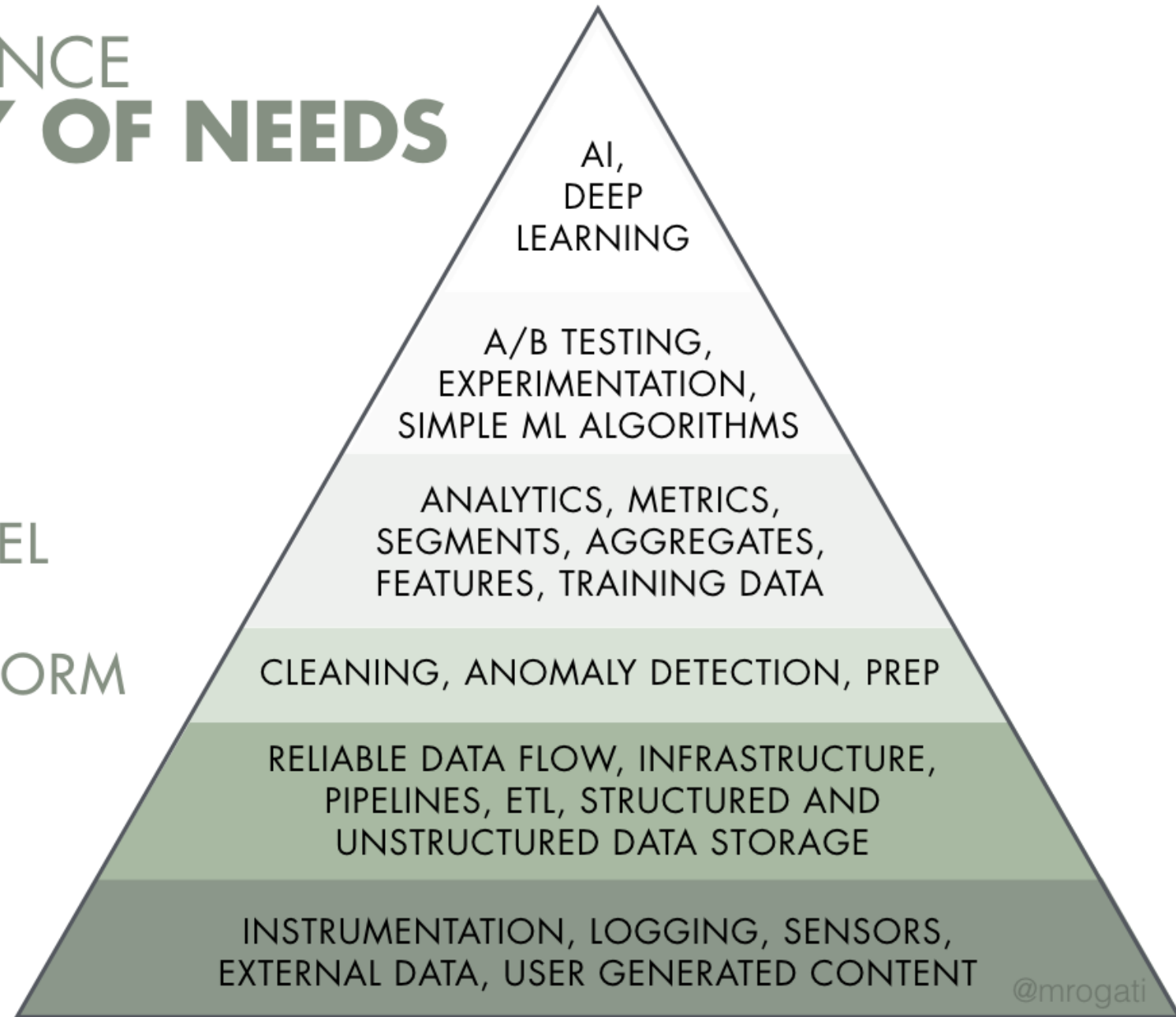
LEARN/OPTIMIZE

AGGREGATE/LABEL

EXPLORE/TRANSFORM

MOVE/STORE

COLLECT



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# What's Next

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## Apache Hadoop

Interest by region ?

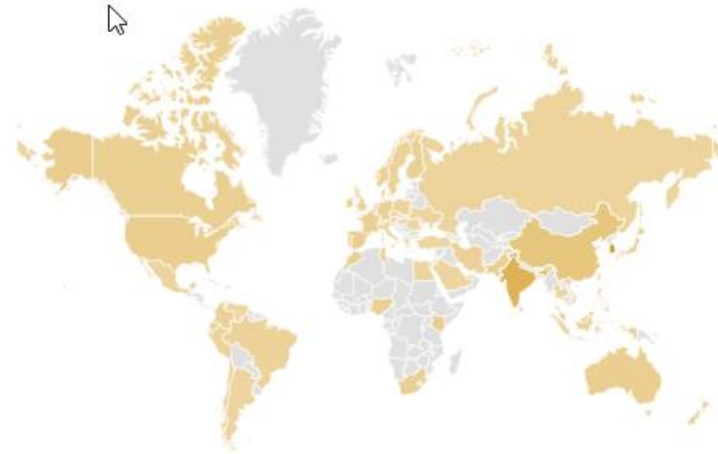
Region ▼ ☰ ⬇ <> 🔗



## Big data

Interest by region ?

Region ▼ ☰ ⬇ <> 🔗



## Machine learning

Interest by region ?

Region ▼ ☰ ⬇ <> 🔗



## Data Science

Interest by region ?

Region ▼ ☰ ⬇ <> 🔗



# Hardware comparison

## 2008

- CPU:
  - 8 Core 2GHz
- RAM:
  - 4GB
- NIC:
  - Dual 1GBe
- Disk:
  - 0.5 TB HDD

## 2017

- CPU: (4x)
  - 2x 14 Cores 2+Ghz
- RAM: (128x)
  - 512GB
- NIC: (5x)
  - 10GBe
- Disk: (250x)
  - 12-24x 6TB HDD (~144TB)
  - SSD System Disk

# From Cloudera's original pitch deck (Sept 2008)

## The Cloud Wars: \$100+ billion at stake

### ■ The Cloud - A multi-year shift in the computing paradigm

We are in the midst of a pronounced shift from client-server to Cloud computing, which is more analogous to centralized mainframe computing. Quantum improvements in Internet bandwidth, computing power and memory, coupled with enabling technologies like virtualization, parallel processing and multi-core chips, make it feasible to run large computing tasks on a centralized 'Cloud' infrastructure. The economics are truly compelling, with cost advantages of 3-5x for business apps, and 5-10x or better for personal productivity apps.

### Shift creates a \$100+ billion opportunity

Cloud equivalents exist today for most business and personal productivity apps. Starting in the enterprise as OnDemand apps, roughly a \$2 billion software segment. Cloud apps are moving into personal productivity (e.g., email, word processing). Cloud software is not as mature as client-server, but the trajectory is changing. The total \$160bn addressable market opportunity includes \$95 billion in business and productivity apps, and another \$65 billion in online advertising.

Source:  
Merrill Lynch  
Industry  
Overview,  
May 7, 2008

04/21/17

Cloudera Confidential

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## Cloudera Differentiators

- **Enabling Hadoop as an elastic platform with statistical multiplexing over many customers**
- **Multi-Tenant Support:** Concurrency, Priority, Namespace Isolation, Performance Isolation.
- **Monitoring, Reliability, and Availability**
- **Resilience and Fast Recovery:** A non-sexy problem that is **critical to enterprises**, no time to restart ETL job from scratch, otherwise misses SLA.
- **IDE to easily debug, deploy, and tune.**
- Integration with **data mining and analysis** functionality (R, Weka, SAS, SPSS)
- **Connector certification:** another non-sexy problem that is ignored by community, make sure system is compatible with other enterprise systems.

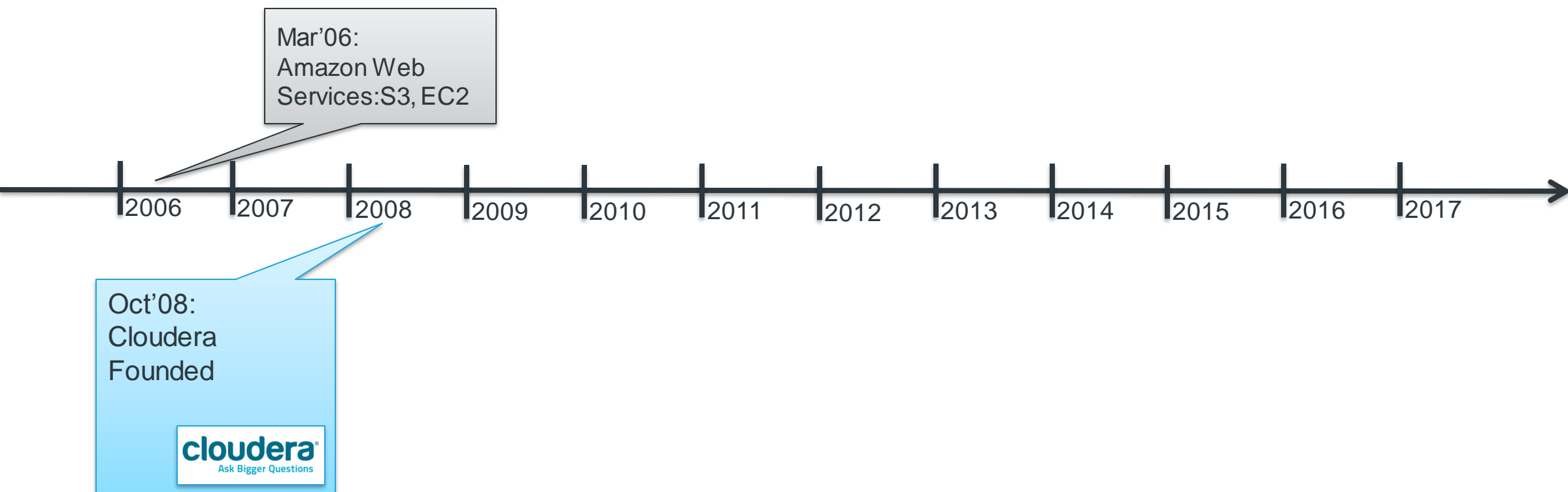
04/21/17

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<https://www.slideshare.net/AccelPartners/clouderas-original-pitch-deck-from-2008>

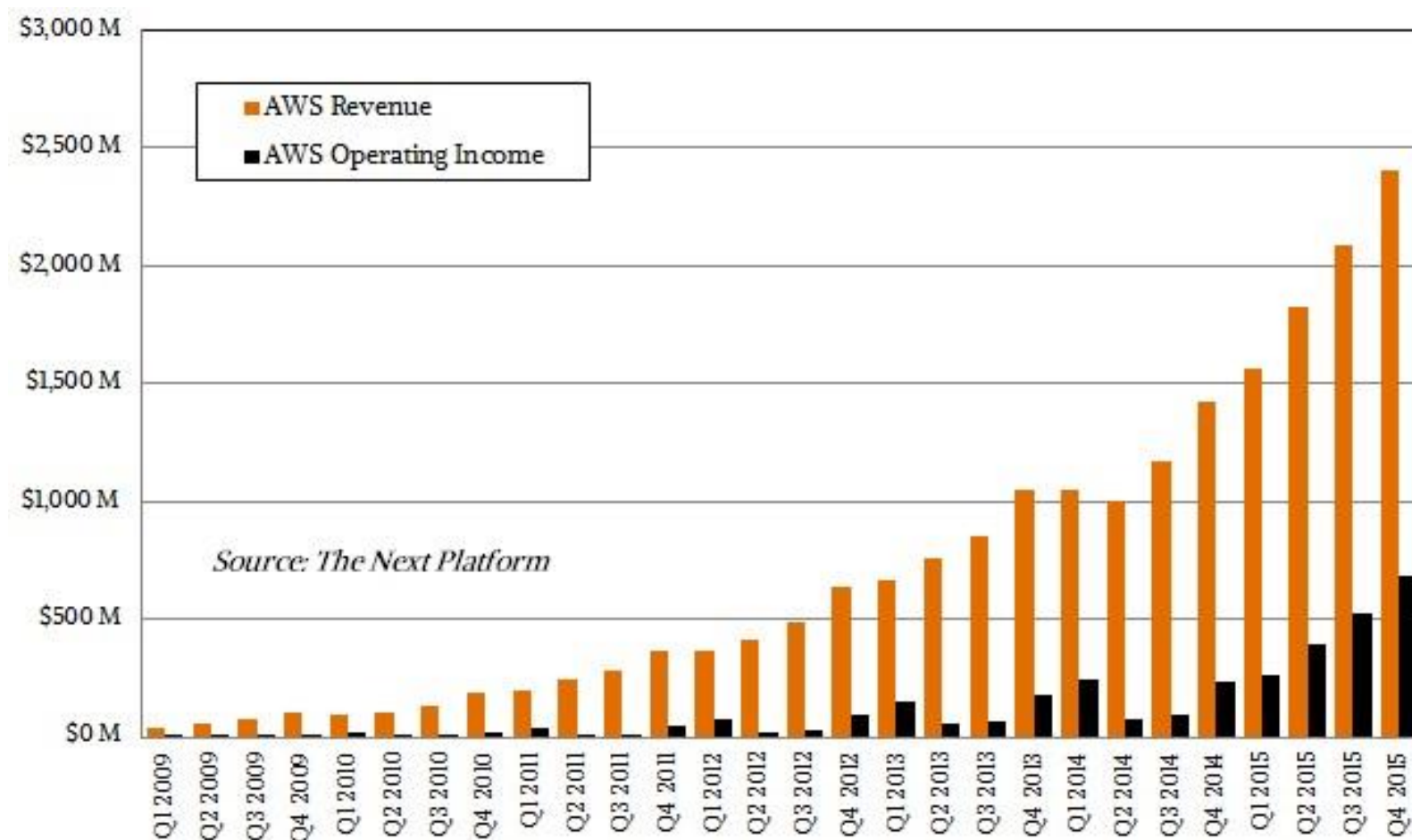
# Data Platforms in the Cloud





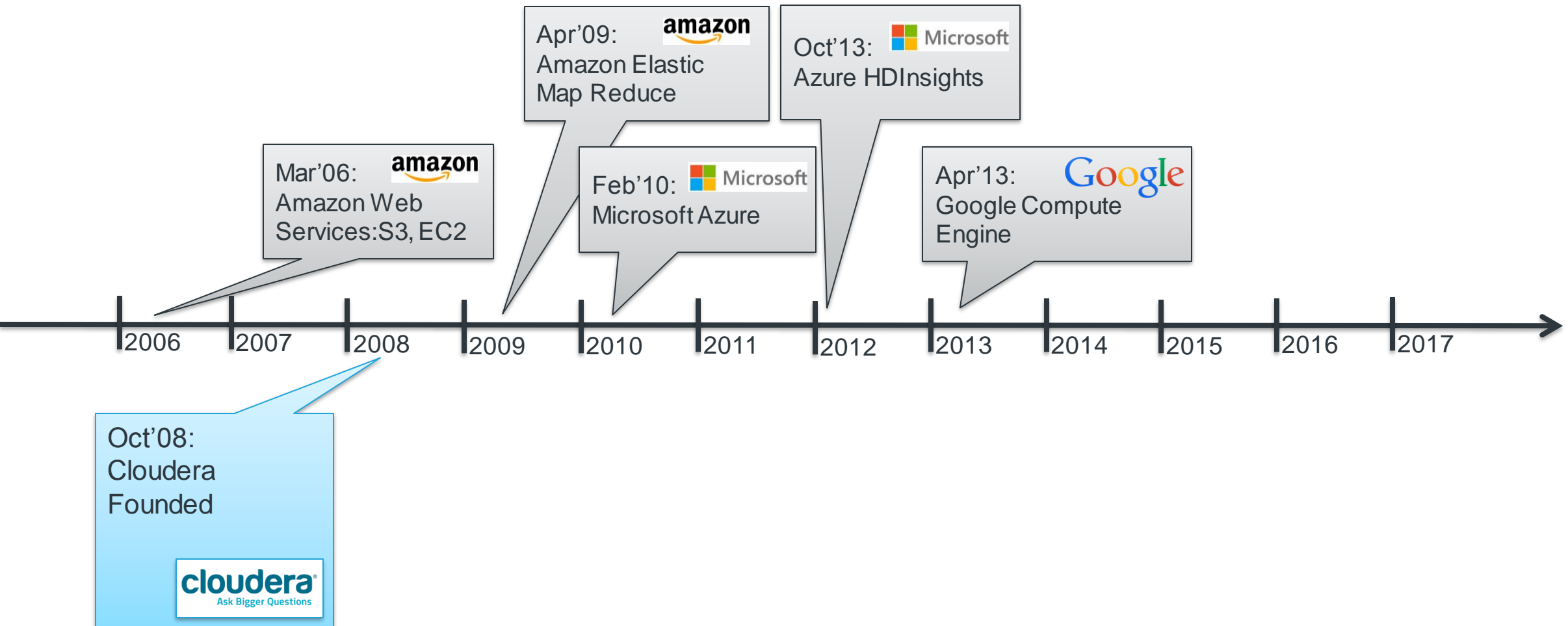
# The growth of cloud is now undeniable

- AWS revenues have a slow start
- But growth and competitor are here now.



<https://www.nextplatform.com/2016/02/01/how-long-can-aws-keep-climbing-its-steep-growth-curve/>

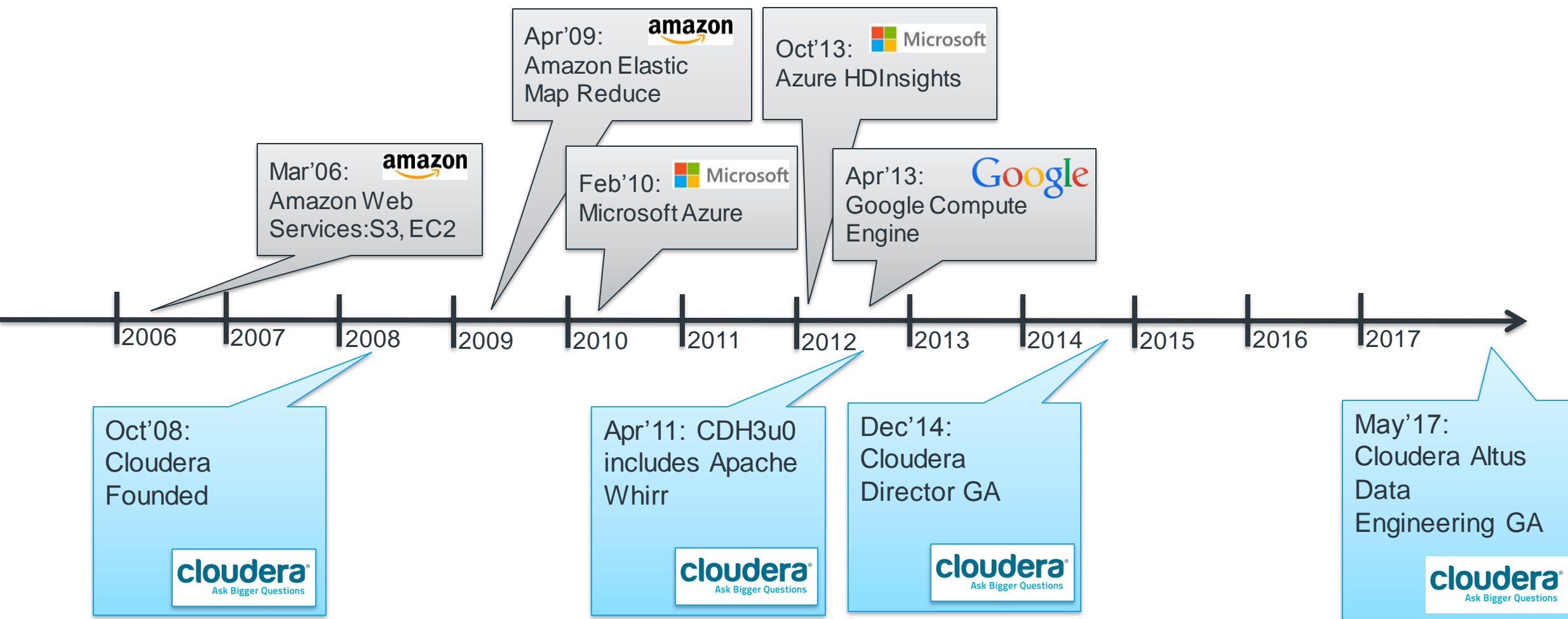
# Data Platforms in the Cloud



# New Assumptions with Data Platforms on Cloud

- On-prem Assumption
  - Machines are always on and added every few months. High Cap-Ex
  - Coupling Storage and compute for locality is critical for efficient performance
  - Deal with Services
  - Cost for maintaining services software and hardware health on Ops team
- Cloud Assumption
  - Machines are transient and elastic and added and removed every few hours. Low Cap-Ex
  - Decoupling storage from Compute is critical for efficiently lowering Op-Ex
  - Deal with Jobs or Requests
  - Cost for maintaining services software and hardware heal outsourced.

# Data Platforms in the Cloud





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**The only way you can predict the future is  
to build it.**  
**--Alan Kay**

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# Thank you!

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