

Data Pipelines

cs5356

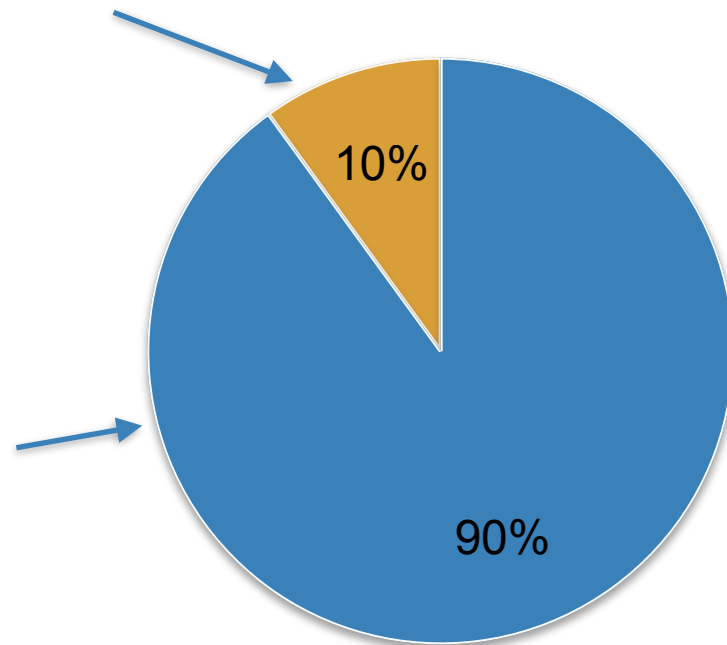
Daniel Doubrovkine

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Job: Data Scientist

Machine Learning Big Data in the Cloud ...

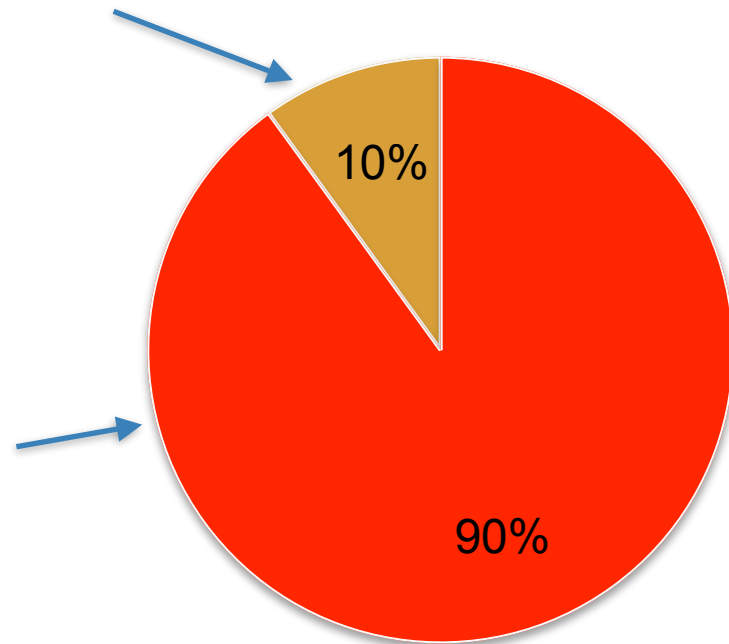
Moving Data Around



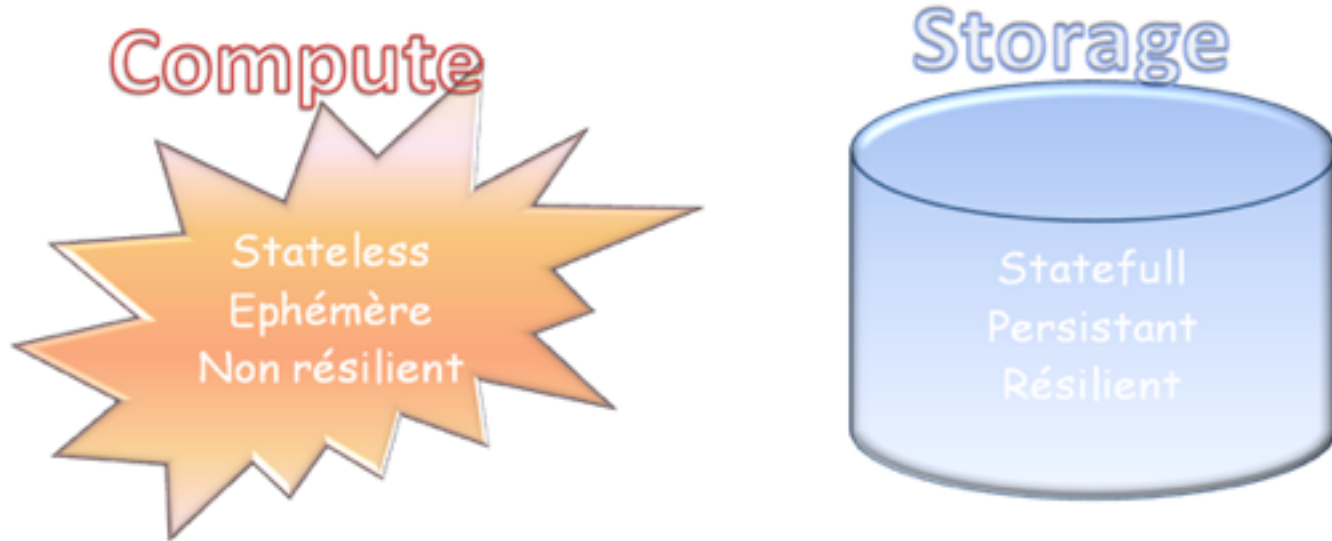
Data Pipelines

Machine Learning Big Data in the Cloud ...

Moving Data Around



Moving Data Around



Typical Warehousing: Batch

Event Data

billions of data points
terabytes of data



**Extract
Transform
Load**



Insight

Too slow ... continuous processing



Clickstream

- What users do, usually in log files.

date, time, IP, URL

- Page views per URL over time?
 - Top N page views of all time?
 - What products do visitors buy together?
-

Aggregate

- Distributed Messaging System
 - **In:** Logs
 - **Out:** Domain Data



Hadoop

- Scalable
- Reliable
- Fault Tolerant
- Simple

One Big Computer



<https://hadoop.apache.org>

Hadoop

- Storage: HDFS

Linux ext3

Replicated blocks

Write Once / Read Many

Retry transfers

- Execution: Map/Reduce

In parallel on many servers

Retry on failure

Map + Reduce

Servers can join or leave

Map/Reduce Canonical Example

- Word Count
 - **Map**: transform text to { word: 1 }
 - **Reduce**: sum by key
-

Storage



YMMV

Amazon RDS



APACHE
HBASE

Amazon Redshift



mongoDB

Replacing Parts of Hadoop

- Spark
 - Microsoft Dryad
 - Apache Tez
 - Impala
 - Google Big Query
 - Google Cloud Dataflow
-

Spark

- Richer API

`filter()`, `join()`, `distinct()`, `groupByKey()`

- Maintain State

`updateStateByKey()`

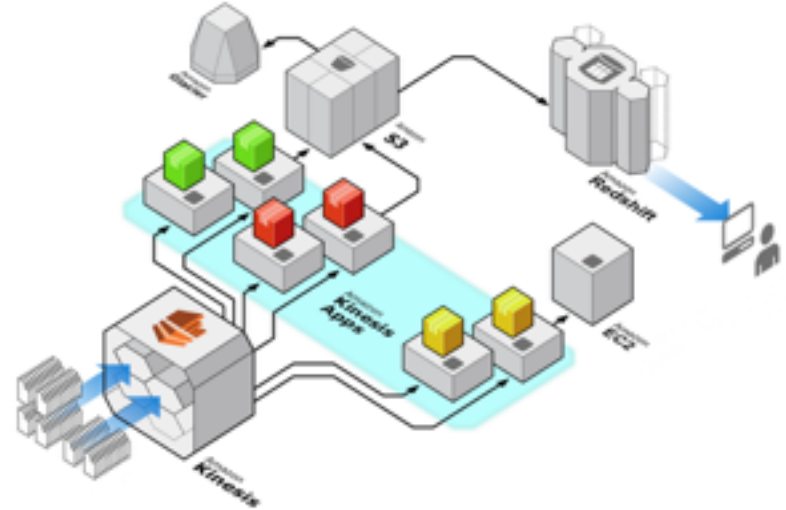
- Window Operations

window size and slide interval

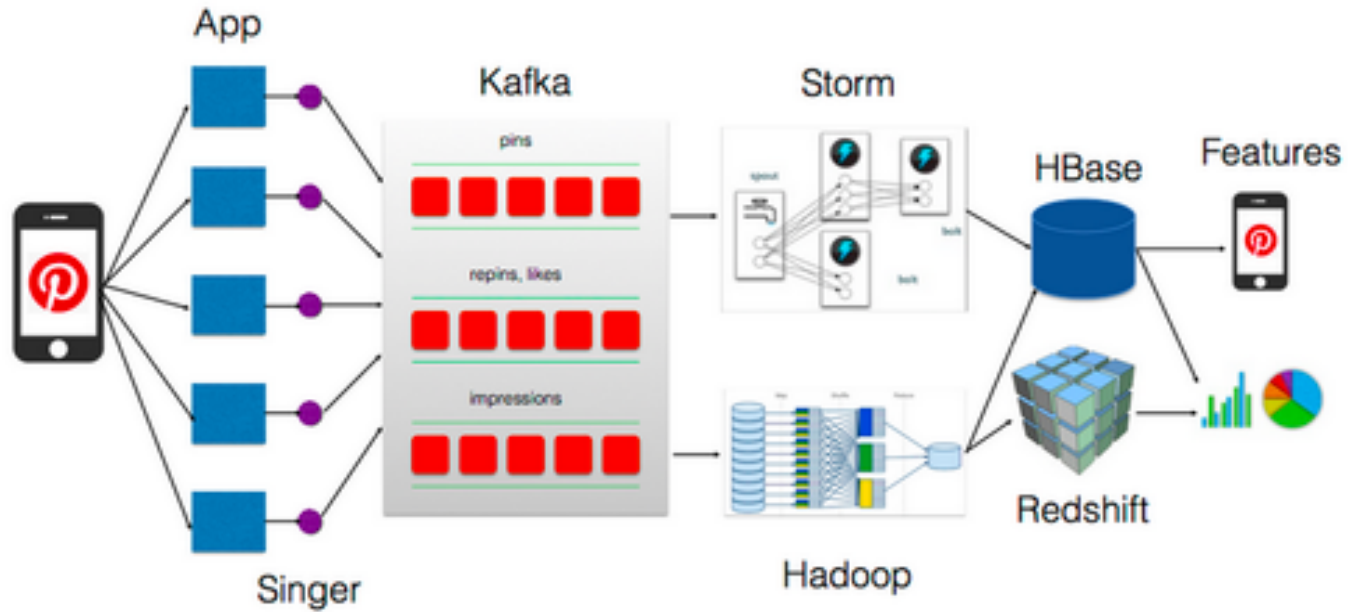


Amazon Kinesis

- Easy
- Real-Time
- High Throughput
- S3/DynamoDB/Redshift



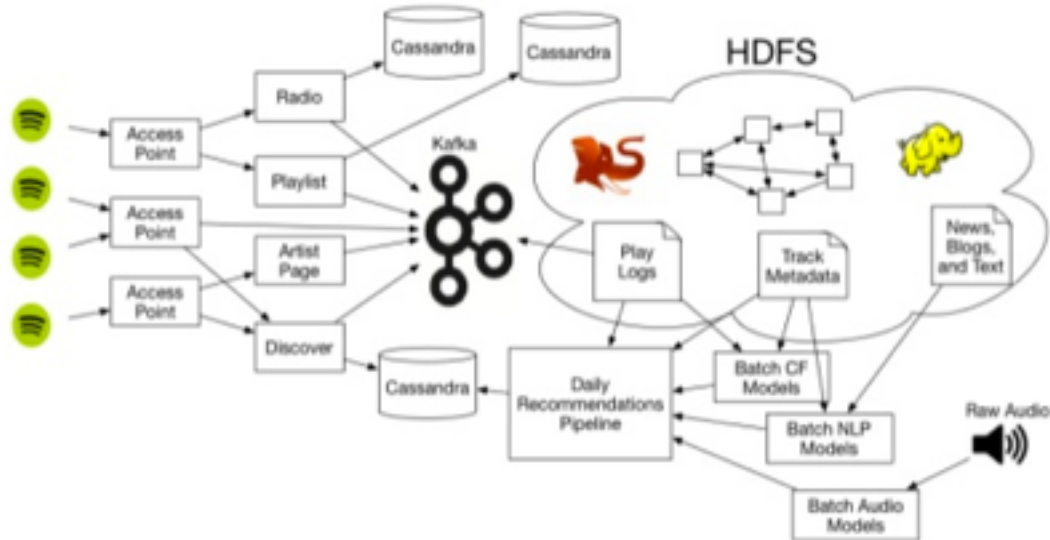
Pinterest



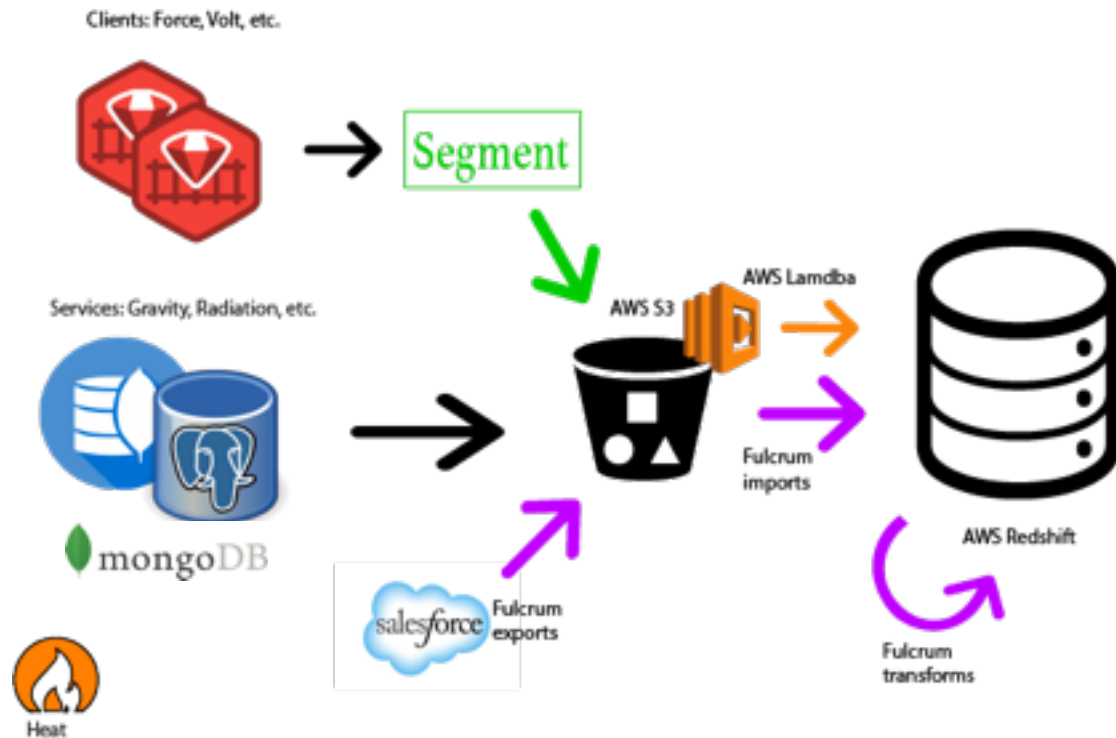
Data Architecture overview

Spotify

Recommendation systems



Artsy



MongoDB (after the break)



Shannon Bradshaw

