Big Data Engineering

Conclusions and Recap

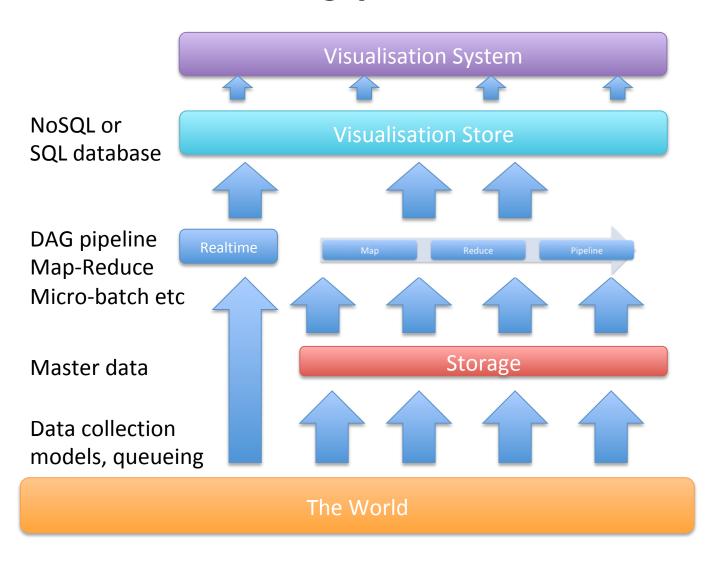


Contents

- Understanding the bigger picture
- What are the different components
- Message queueing and collection systems
- Map-Reduce and DAG systems
- Realtime Systems
- Fast databases for speed
- Visualisation and Dashboards



The big picture





The big picture

- You have immutable master data
- You create a set of processes to:
 - Collect that data
 - Store master data
 - Process data
 - Visualise and present
- Some of those processes act on batch and others on real-time data



How to choose the components?

- Two main approaches:
 - Best of breed
 - Choose the best available component in each space
 - Stack
 - Choose a curated stack that a team or organization is providing/selling/supporting

Approach

- Minimise the pain
 - Choose what you need when you need it
 - Don't over engineer

How do I ingest data?

- File transfer
- Live stream
 - Sockets
 - Syslog
 - Messaging system
- From existing databases

How do I store data?

- HDFS
- NoSQL database only
 - Mongo / HBase / Cassandra
- zFS / GlusterFS / NFS etc
- Apache Parquet, CSV, or speci

How do I process data?

- Simple Map Reduce
- Hive / Pig
- DAG
- Pipeline
- etc

How do I visualise data

- From a SQL database?
- From a NoSQL database?
- Generate charts in Python Spark?
- Etc?

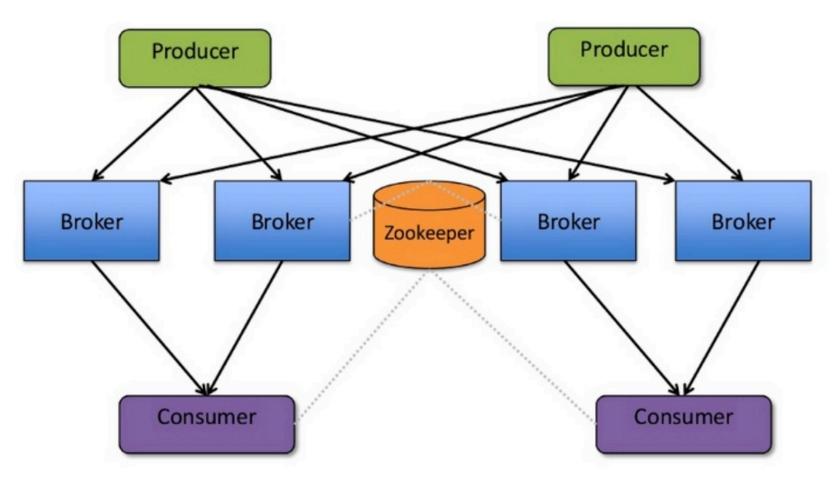


Collection / Queuing systems

- Two ways of making the choice
 - The protocol
 - The middleware
- Protocols
 - ZeroMQ, MQTT, AMQP, STOMP, Kafka Protocol, Rendevouz, etc
- Middleware
 - Kafka, Apollo, Mosquitto, QPid, WSO2, etc

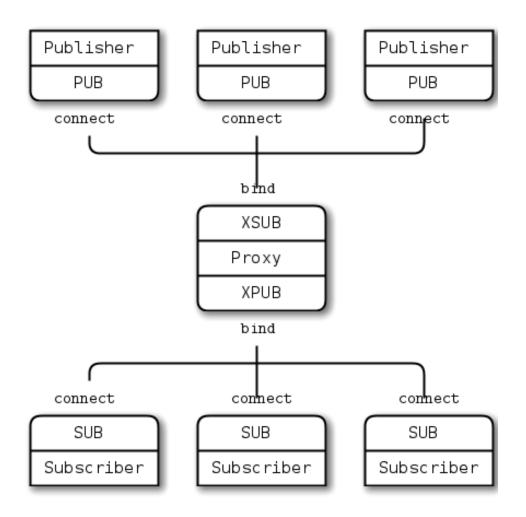


Apache Kafka





ZeroMQ





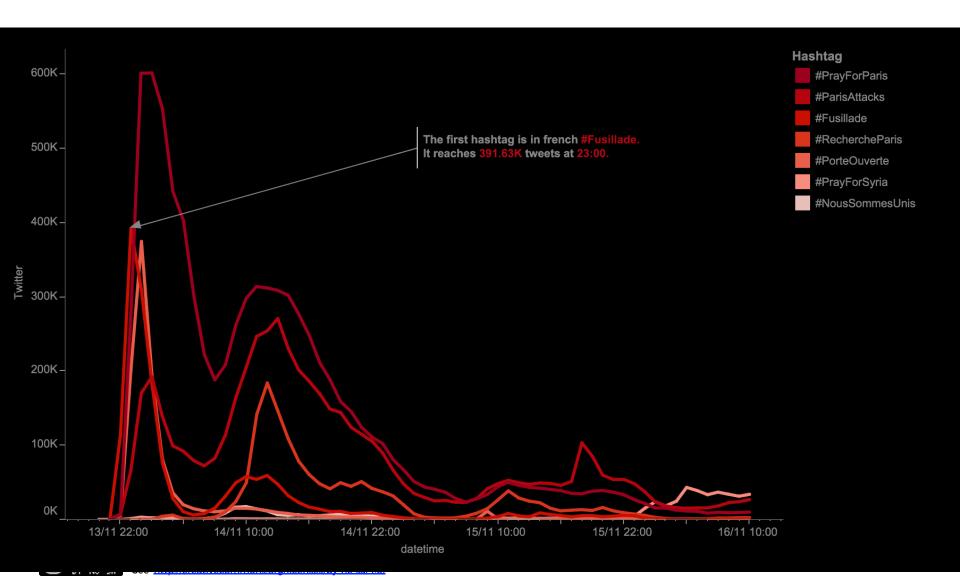
Processing approaches

- Covered in detail already
- Hadoop
- Spark
- Tez
- etc

Cluster Management

- Spark
- YARN
- Mesos
- Kubernetes
- etc

Visualisation



Visualisation approaches

- Full products
 - Tableau, Qlik, SAS, GoodData
- Web-based systems
 - Tableau Public, Datawrapper, Raw, Plotly
- Developer oriented
 - D3.js, dygraphs, Python charting, Leaflet,
 Fusion Charts, Google Charts, etc



Fortune top 10 big data companies

fortune.com/2014/06/13/these-big-data-companies-are-ones-to-watch/

- MapR Apache Hadoop
- MemSQL
- Databricks Apache Spark
- Platfora Apache Hadoop
- Splunk
- Teradata Apache Hadoop
- Palantir Hadoop, Cassandra, Lucene
- Premise
- Datameer Apache Hadoop
- Cloudera Apache Hadoop
- Hortonworks Apache Hadoop
- MongoDB MongoDB
- Trifacta Apache Hadoop



Hortonworks



Enhanced SQL Semantics in Apache Hive

Hive adds time intervals and UNION semantics, 2.5x performance improvements and improved query scheduling, along with a more streamlined user interface for Hive within Ambari.

Solr on YARN

The Solr search engine is being built to run on YARN and is now in technical preview. This critical advancement allows customers to reduce their total cost of ownership by deploying Solr within the same cluster as other workloads – eliminating the need for a "side cluster" dedicated to indexing data and delivering search results.

New capabilities for feature-rich Spark applications

Apache Spark on YARN is enhanced with the new DataFrame API, machine learning algorithms such as clustering, frequent patternmining algorithms and a technology preview of SparkSQL.

Databricks



Notebooks

Dashboards

Jobs

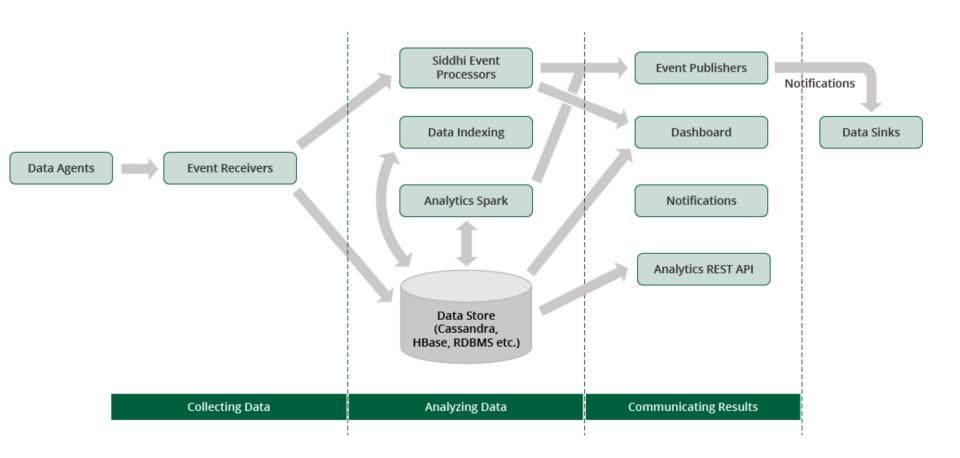
Third-Party Apps

Cluster Manager



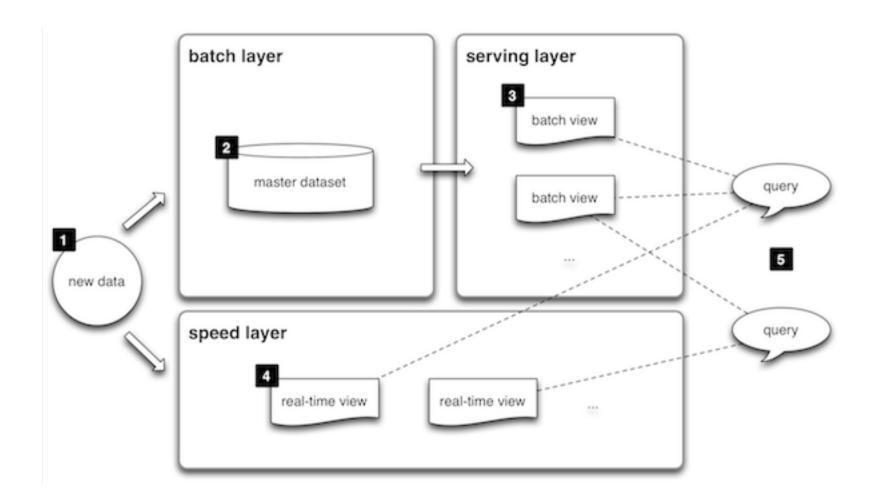


WSO2 DAS





Lambda





The real answer

You are on the bleeding edge -Expect to have some pain



Questions?

