



Overview

- Spark at MediaCrossing
- Choosing, maintaining and possibly compiling the right combination of packages to work with Spark
- Data serialization/deserialization formats
- Performance issues with small data
- Configuration/Deployment automation
- Monitoring
- Exposing Spark for non-developer usage



Spark at MediaCrossing

- We make it easier and more efficient for advertisers and publishers to trade digital media in milliseconds
- Since we started trading in 2013, we were able to avoid ever having to use Map/Reduce for our big data thanks to Spark
- Inspired by Nathan Marz's "Lambda Architecture" principles, our team leverages a unified view of realtime and historic user behavior to constantly adjust our buying and selling models



Target Audience

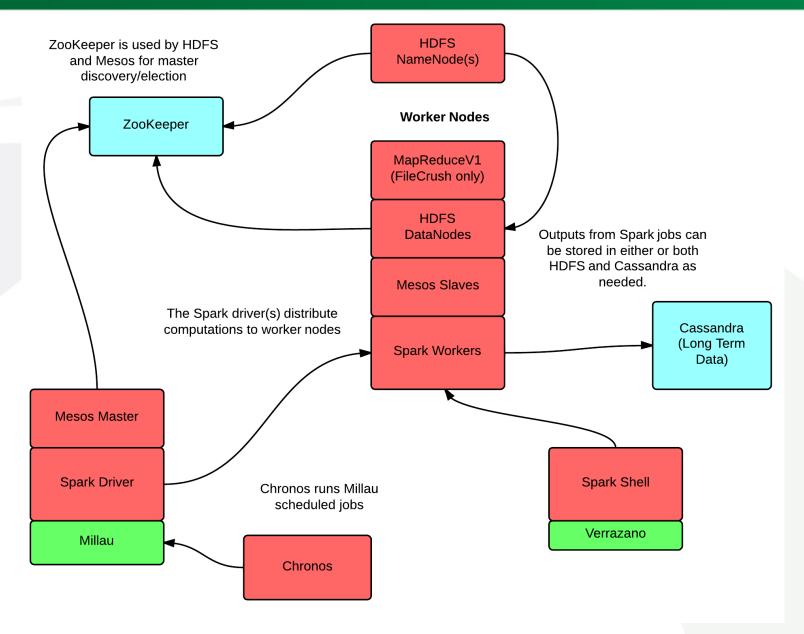
- You are building a new data/analytics system and have determined that you have big data (> 5TB)
- 'New' in this case means no prior Hadoop/HDFS installation
- This is your first rodeo with an analytics system based on Spark



Building the Stack

- Where will data be stored?
 - -HDFS/Cassandra are two major options
 - -If not already published, may need to compile Spark against your HDFS distribution version
- How to choose a cluster manager?
 - -Standalone/Yarn/Mesos
 - -We went with Mesos Berkeley stack preference
- How will data get persisted into your system?
 - -Continuous streaming data Apache Flume, Spark Streaming
 - -Large batches Spark jobs!
- Reliable Job Scheduling?
 - -Chronos (runs on Mesos) fault-tolerant job scheduler







Storing your data

- Text Files
 - -Human Readable
 - Not splittable when compressed out of box (negatively affects parallelization)
- Sequence Files (container for binary key-value pairs)
 - -Not human readable
 - -Great for storing key-value pairs without parsing
 - -Splittable (helps parallelize jobs)
 - -Storage efficient
 - -Protobuf friendly



Small File Woes

- HDFS Default Block Size 64MB
- Data partitioning in Spark is done based on the number of blocks each source file takes up.
- Spark/HDFS performs orders of magnitude better with 100's of files on the order of 64/128 MB in size vs 1000's/10's of 1000's of much smaller files.
- Options:
 - -Batch your data into large files at write time
 - -Write your own job to aggregate small files into large ones
 - -Use an existing library like https://github.com/edwardcapriolo/filecrush



Configuration/Deployment

- Automate from day 1 Ansible, Puppet, Chef, etc
- Like you would with source code, version control your configuration and deployment scripts
- Co-locate Spark workers (Mesos slaves) with HDFS data nodes – noticeably better performance
- Ideal World: Separate research cluster from 'production'



Monitoring

- Spark covers this extensively in their docs: https://spark.apache.org/docs/latest/monitoring.html
- In practice, we rely on Nagios to tell us when a server is struggling and Munin for diagnosing less urgent, long running problems
- Running on Mesos, we rely on their user interface for visually seeing what work is taking place across the cluster, and when we need to increase our cluster size.



Non-developer Tooling

- Built-in PySpark for interactive querying
- SparkR http://amplab-extras.github.io/SparkR-pkg/
- Shark (Hive on top of Spark) SQL query language



Thank you for your time!

- We continue to increase our usage of Spark for research and other critical business functions
- If what we are doing sounds interesting...
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