Building 1000 node **Spork** cluster on EMR

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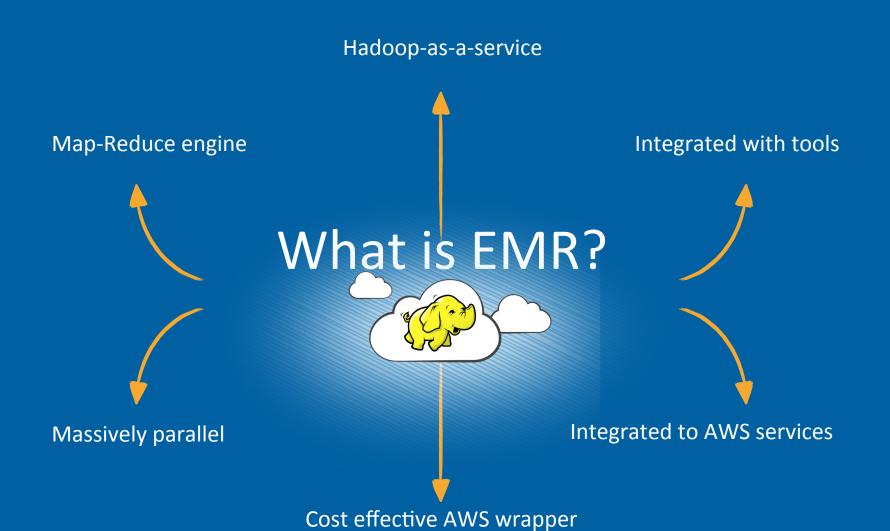


What is EMR?

Amazon Elastic MapReduce



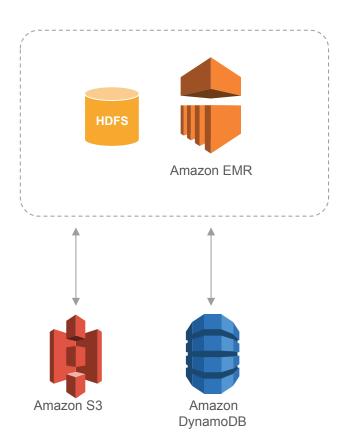






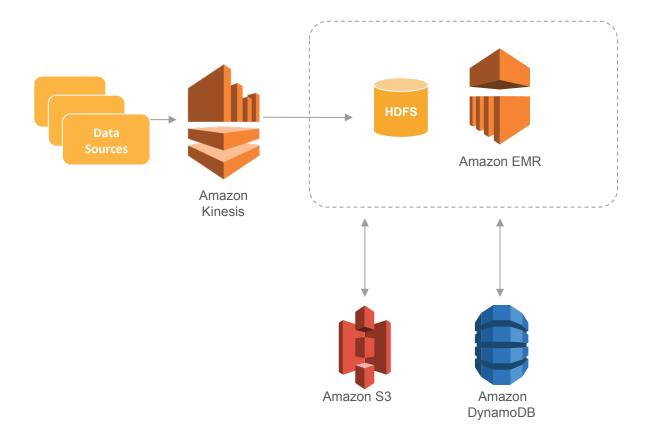




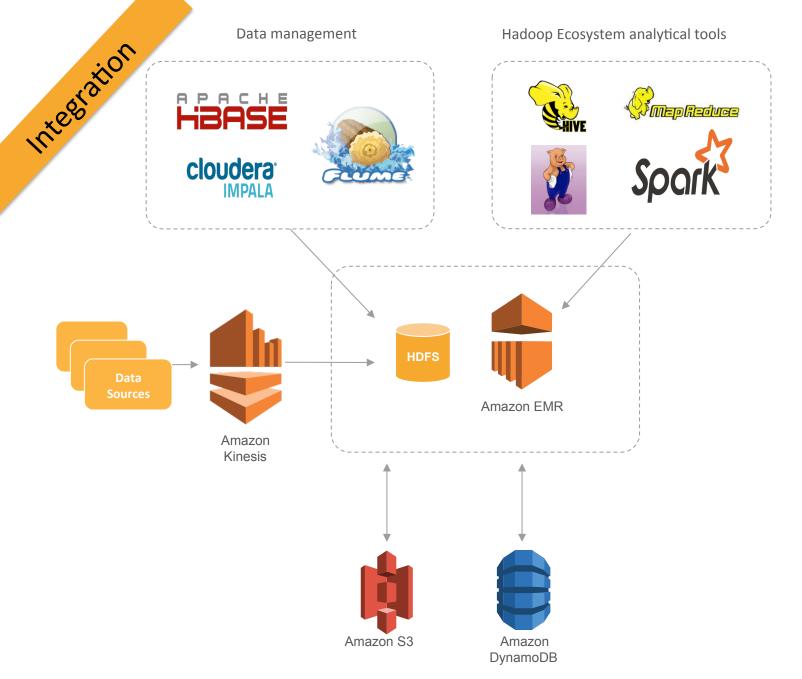




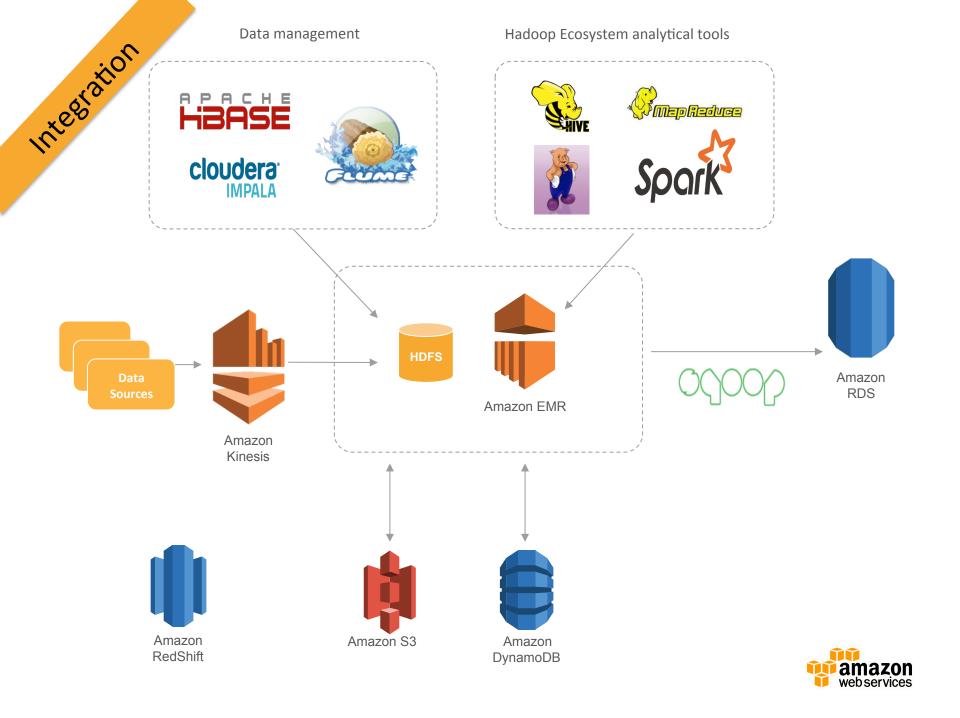
Integration

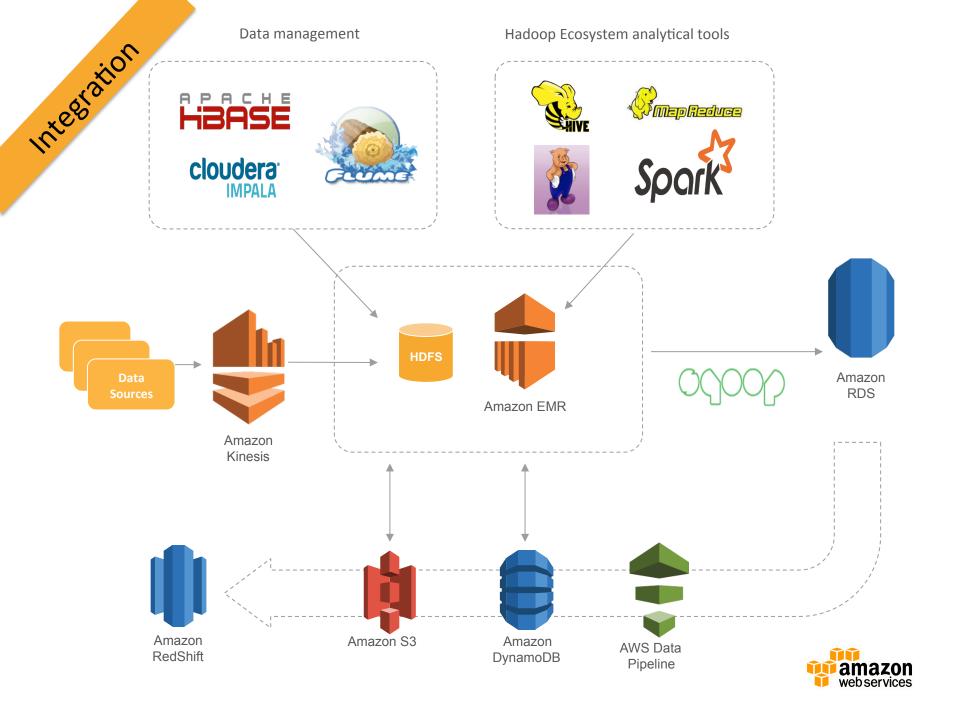












Amazon EMR Concepts

Master Node

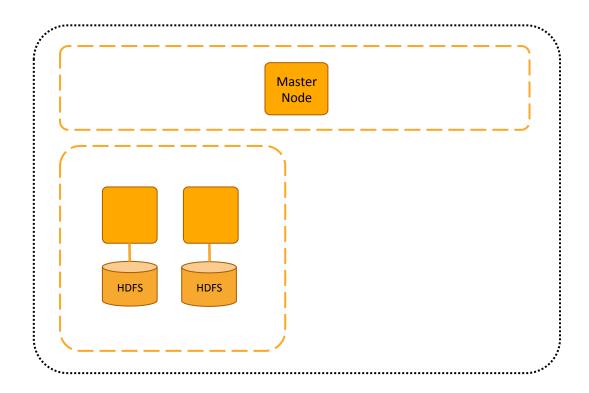
Core Nodes

Task Nodes



Core Nodes

DataNode (HDFS)





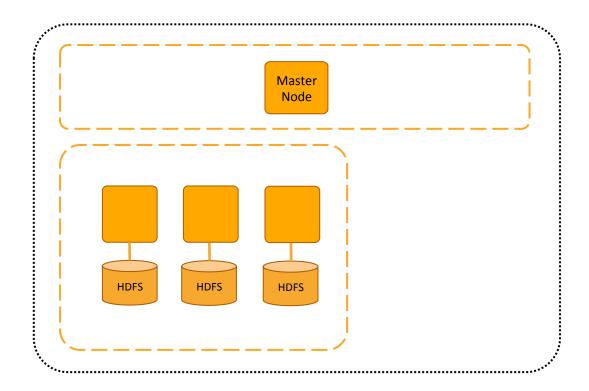
Core Nodes

Can Add Core Nodes:

More CPU

More Memory

More HDFS Space

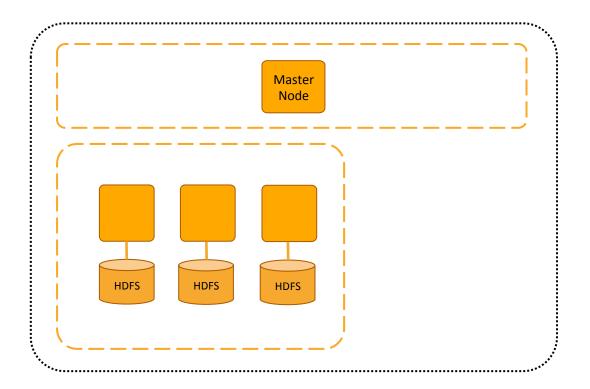




Core Nodes

Can't remove core nodes:

HDFS corruption





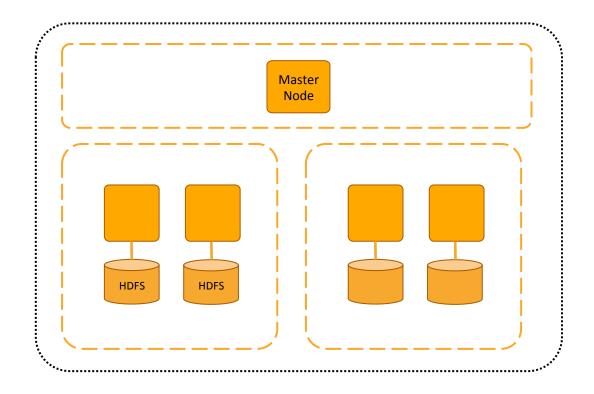
Task Nodes

No HDFS

Provides compute resources:

CPU

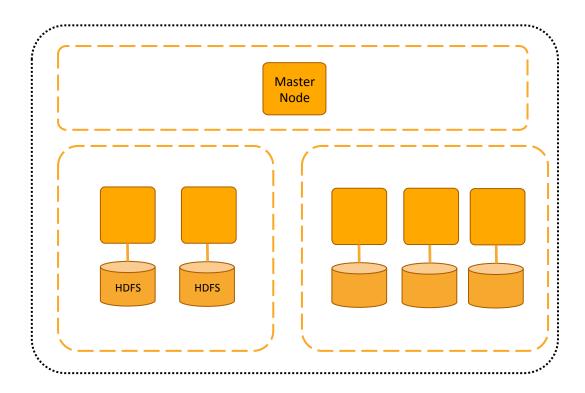
Memory





Task Nodes

Can add and remove task nodes





Spark On Amazon EMR





Bootstrap Actions

- Ability to run or install additional packages/ software on EMR nodes
- Simple bash script stored on S3
- Script gets executed during node/instance boot time
- Script gets executed on every node that gets added to the cluster



Spark on Amazon EMR

Bootstrap action installs Spark on EMR nodes

Currently on Spark 0.8.1 & upgrading to 1.0 very soon



Why Spark on Amazon EMR?

- Deploy small and large Spark clusters in minutes
- EMR manages your cluster, and handles node recover in case of failures
- Integration with EC2 Spot Market, Amazon Redshift, Amazon Data pipeline, Amazon Cloudwatch and etc
- Tight integration with Amazon S3



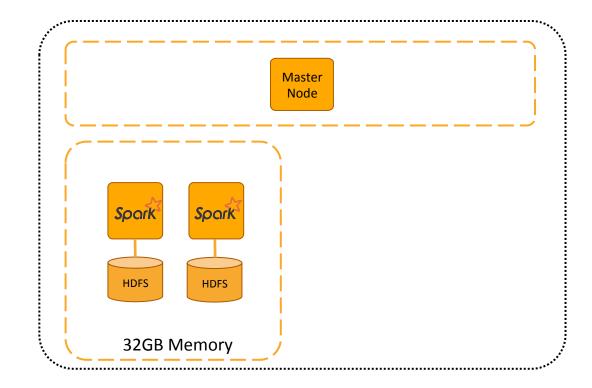
Why Spark on Amazon EMR?

Shipping Spark logs to S3 for debugging

Define S3 bucket at cluster deploy time

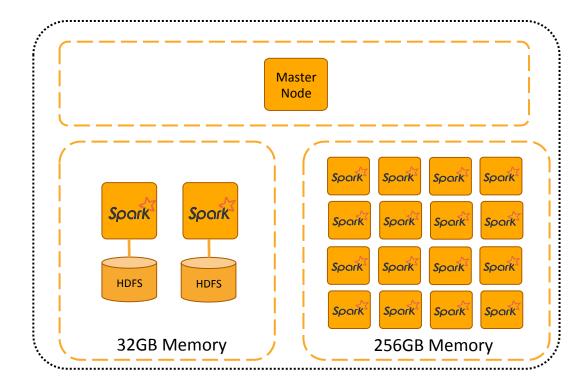


- Launch initial
 Spark cluster with
 core nodes
- HDFS to store and checkpoint RDDs





 Add Task nodes in spot market to increase memory capacity

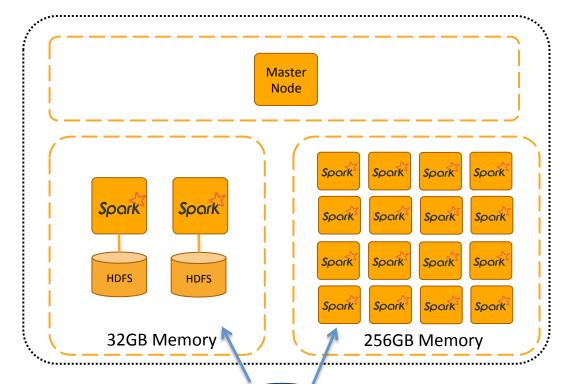




 Create RDDs from HDFS or Amazon S3 with:

sc.textFile
OR
sc.sequenceFile

 Run Computation on RDDs

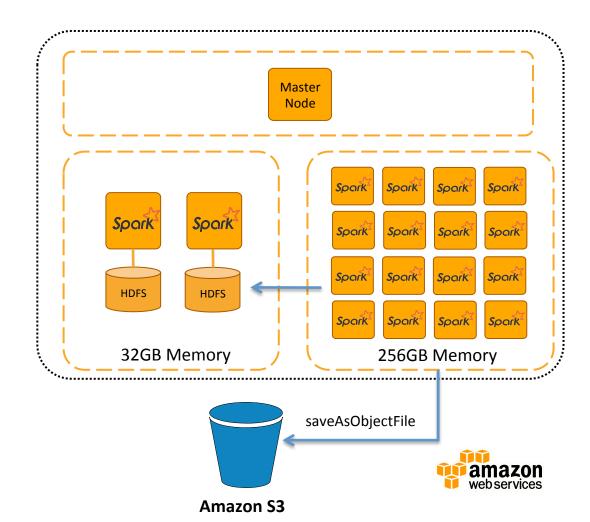




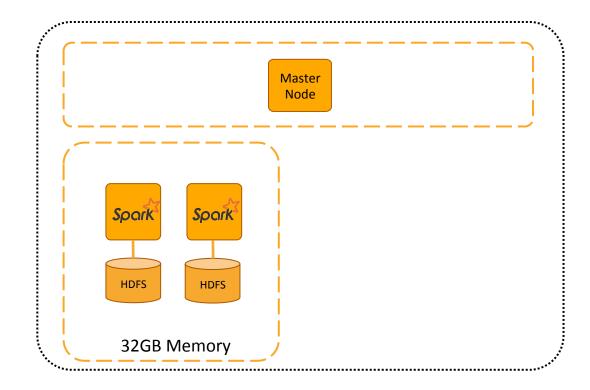


 Save the resulting RDDs to HDFS or S3 with:

> rdd.saveAsSequenceFile OR rdd.saveAsObjectFile

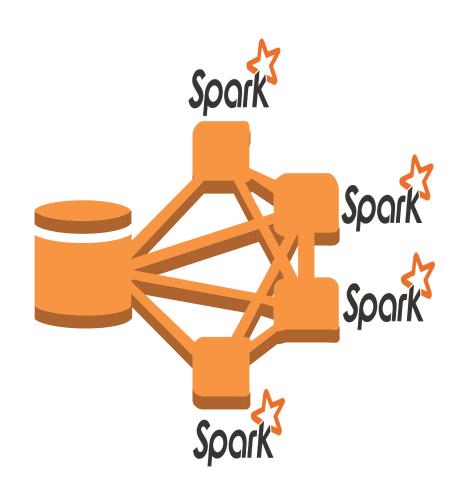


Shutdown
 TaskNodes when
 your job is done



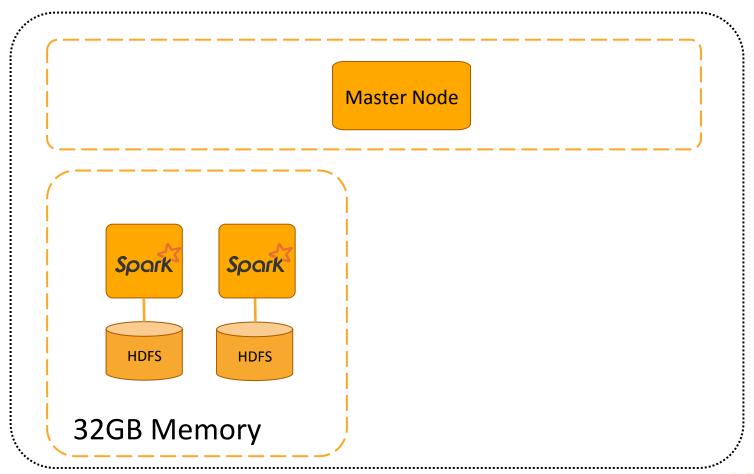


Elastic Spark With Amazon EMR



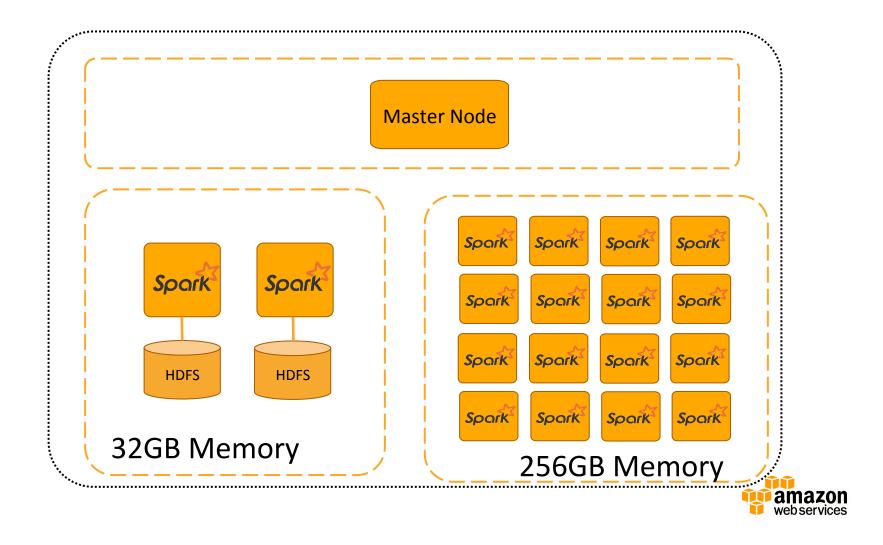


Autoscaling Spark





Autoscaling Spark



Elastic Spark

- When to Scale?
 - Depends on your job
- CPU bounded or Memory intensive?
 - Probably both for Spark jobs
- Use CPU/Memory util. metrics to decide when to scale



Spark Autoscaling Based Memory

Spark needs memory

– Lots of it!!

How to scale based on the memory usage?







- What do I need to launch a cluster?
 - AWS Account
 - Amazon EMR CLI



Easy to launch – 1 command

```
./elastic-mapreduce --create -alive
--name "Spark/Shark Cluster" \
--bootstrap-action s3://elasticmapreduce/samples/spark/0.8.1/install-spark-shark.sh
--bootstrap-name "Spark/Shark"
--instance-type m1.xlarge
--instance-count 1000
```

Comes up in 15-20mins



Adding Task nodes

- --add-instance-group *TASK*
- --instance-count *INSTANCE_COUNT*
- --instance-type INSTANCE_TYPE



Is cluster ready?

Cluster will be in Waiting state





Elastic MapReduce V Cluster List > Cluster Details

Add step

Resize

Clone

Terminate

Cluster: Spark/Shark Cluster Waiting Waiting for steps to run

Master public DNS: ec2-107-20-0-141.compute-1.amazonaws.com

Tags: -- View All / Edit

Summary

ID: j-MGQ0H4L0JJKG

Creation date: 2014-06-18 18:02 (UTC-7)

Elapsed time: 20 minutes

Auto-terminate: No

Termination Off Change

protection:

Configuration Details

AMI version: 2.4.2

Hadoop Amazon 1.0.3

distribution:

Applications: --

Log URI: s3://chayel-spark-logs/

-

Security/Network

Availability us-east-1a

zone:

Subnet ID: --

Key name: keypair

EC2 role: --

Visible to all None Change

users:

Hardware

Master: Running 1 m1.xlarge

Core: Running 999 m1.xlarge

Task: --



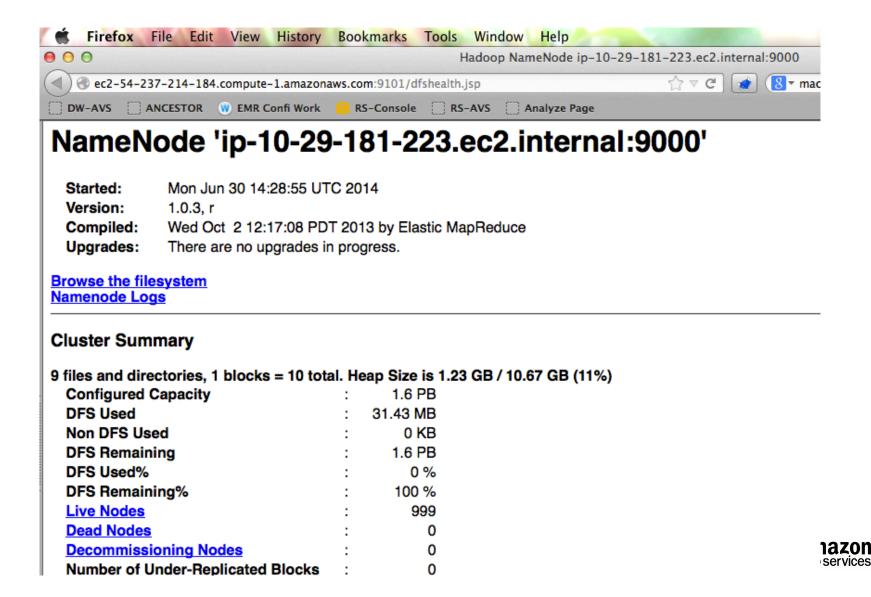
Lynx Interface

lynx http://localhost:9101

```
Namenoae Logs
Cluster Summary
9 files and directories, 1 blocks = 10 total. Heap Size is 314 MB / 2 GB (15%)
 Configured Capacity
                      : 1.6 PB
 DFS Used
                                  : 31.43 MB
 Non DFS Used
                                  : 0 KB
 DFS Remaining
                                  : 1.6 PB
 DFS Used%
                                  : 0 %
 DFS Remaining%
                                  : 100 %
Live Nodes
                                  : 999
 Dead Nodes
                                  : 0
 Decommissioning Nodes
 Number of Under-Replicated Blocks: 0
NameNode Storage:
```



Web Interface



Spark UI



Workers

ld	Address
worker-20140630142856-ip-10-16-139-142.ec2.internal-35690	ip-10-16-139-142.ec2.internal:7077
worker-20140630142856-ip-10-183-44-115.ec2.internal-49815	ip-10-183-44-115.ec2.internal:7077
worker-20140630142858-ip-10-146-177-207.ec2.internal-37928	ip-10-146-177-207.ec2.internal:7077
worker-20140630142858-ip-10-231-18-32.ec2.internal-53245	ip-10-231-18-32.ec2.internal:7077
worker-20140630142900-ip-10-146-248-171.ec2.internal-58652	ip-10-146-248-171.ec2.internal:7077
worker-20140630142900-ip-10-65-132-221.ec2.internal-43700	ip-10-65-132-221.ec2.internal:7077
worker-20140630142901-in-10-146-249-189 ec2 internal-41691	in-10-146-249-189 ec2 internal:7077

Dataset

- Wikipedia article traffic statistics
 - -4.5 TB
 - 104 Billion records

- Stored in Amazon S3
 - s3://bigdata-spark-demo/wikistats/



Dataset

- File structure (pagecount-DATE-HOUR.gz)
- Period: Dec-2007 to Feb-2014

- Format of File (tsv)
- Feilds

projectcode, pagename, pageviews, and bytes



Sample dataset

```
Projectcode pagename pageviews bytes
```

```
en Barack_Obama 997 123091092
en Barack_Obama%27s_first_100_days 8 850127
en Barack_Obama,_Jr 1 144103
en Barack_Obama,_Sr. 37 938821
en Barack_Obama_%22HOPE%22_poster 4 81005
en Barack Obama %22Hope%22 poster 5 102081
```



Loading data





Analyze options







Table structure

```
create external table wikistats
projectcode string,
pagename string,
pageviews int,
pagesize int
ROW FORMAT
DELIMITED FIELDS
TERMINATED BY ' '
LOCATION 's3n://bigdata-spark-demo/wikistats/';
ALTER TABLE wikistats add partition(dt='2007-12') location 's3n://bigdata-spark-
demo//wikistats/2007/2007-12';
Adding partitions for every month till 2014-04
```



Analyze using Shark

Top 10 Page Views in Jan 2014

Exec time: 26 secs

Scanning 250GB of data



Analyze using Shark

Query 2:

Top 10 Page Views Overall

- Exec time: 45 sec
- Scanning 4.5TB of data



Analyze using Shark

Query 3: No of pages in each projectcodes

Exec time: 48 secs

Scanning 4.5TB of data / 104 Billion records



Spark Streaming and Amazon Kinesis





Amazon Kinesis

CreateStream

Creates a new Data Stream within the Kinesis Service

PutRecord

Adds new records to a Kinesis Stream

DescribeStream

Provides metadata about the Stream, including name, status, Shards, etc.

GetNextRecord

Fetches next record for processing by user business logic

MergeShard / SplitShard

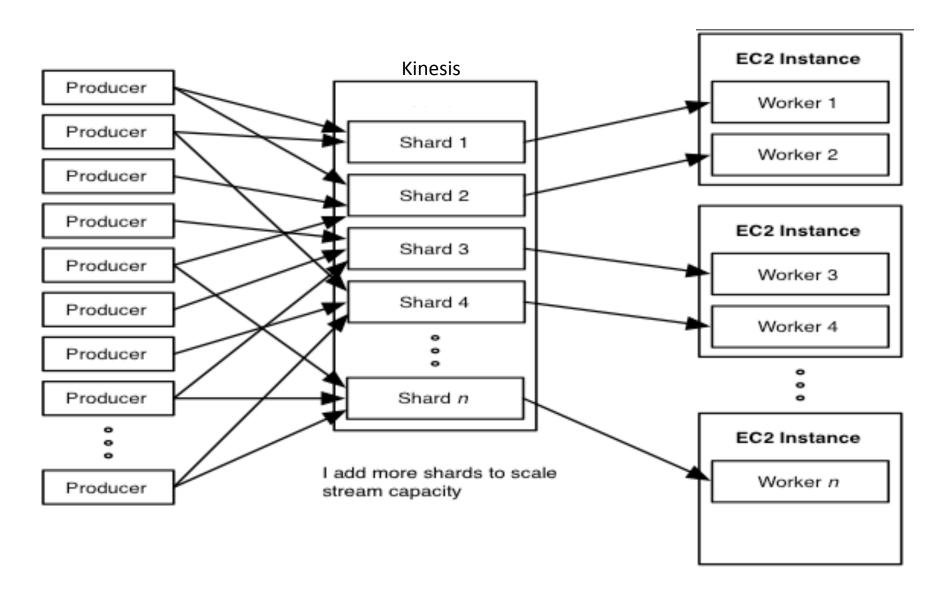
Scales Stream up/ down

DeleteStream

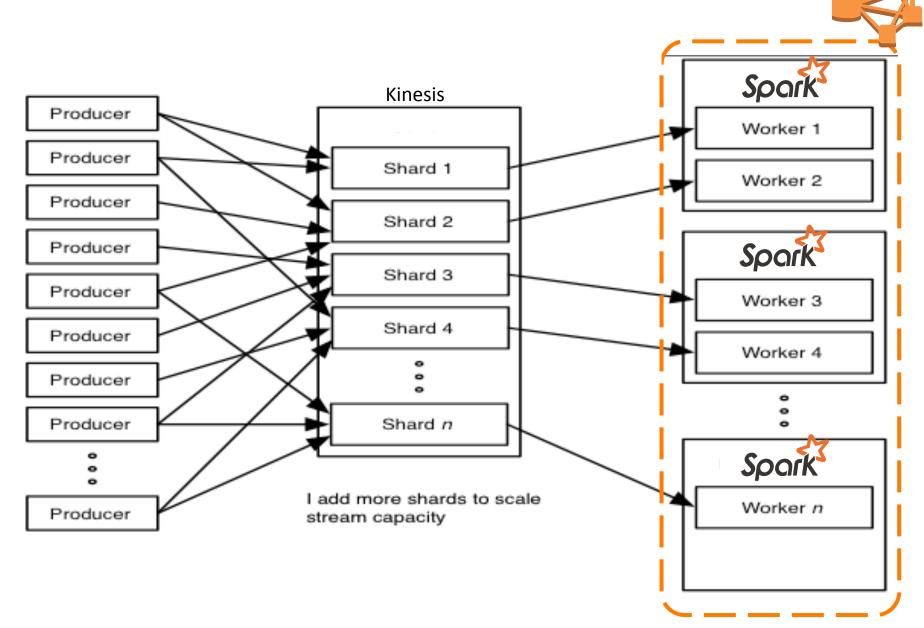
Deletes the Stream



Amazon Kinesis



Amazon Kinesis



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http://bit.ly/sparkemr

