# Hands on



Exploring Machine learning with Spark & Real-world Scalable Architecture

### PREREQUISITES

- Python Programming
- Higher level understanding of Distributed computing
- Map-Reduce Paradigm



#### WHAT WAS COVERED EARLIER?

Installation

- Resilient Distributed Datasets (RDD)
  - Create RDDs
  - O RDD transformations
- Spark SQL
- Introduction Spark Streaming



#### WHAT WE WILL COVER TODAY?

- Review of Apache Spark basic concepts
- Text Classification using MLlib
- Exploring Real-world Architectures
  - Scalable Data Processing Pipeline



#### REVIEW OF APACHE SPARK CONCEPTS

Obligatory program of *Map-Reduce*: Word Count



# BUSINESS PROBLEM AT CPISOURCE

- Clinical Text Medical Records of patients
- Distributions of Medical Charts
  - $\circ$  Client  $\rightarrow$  Operations Team  $\rightarrow$  Team Leads  $\rightarrow$  Medical Coders
- ICD-10 Coding by Medical Coders
  - International Classification of Diseases codes for diseases, signs and symptoms, abnormal findings, complaints, social circumstances, and external causes of injury or diseases
  - Category Hierarchical Condition Category (HCC) vs No HCC

Example: <u>Viral Fever</u> - A92.8 - HCC - billable <u>Whooping cough</u> - A37 - No HCC - Non Billable

- Revenue based on no. of HCC Charts being coded
  - Each Medical Coder codes 10-12 charts/day depending on expertise

#### CHALLENGES:

- Processing of thousands of files
- Dynamic distribution of files for processing
- Fault tolerant architecture
- Data security is the priority
- Cost minimization
- Fully Automated data pipeline



#### GENERAL WAY TO APPROACH

Maintain a separate data store

Keep dedicated servers running all day to process files.

Server maintenance is an overhead

Less flexibility to scale the resources.

#### SERVERLESS PATH

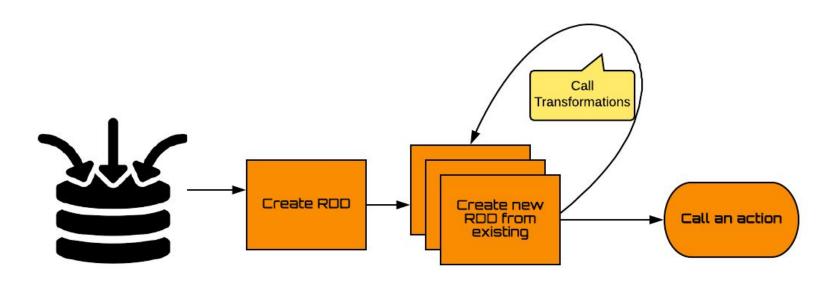
- Event-based asynchronous approach to application design
- Launch clusters only when required.
- Three simple steps:
  - Start
  - Process
  - Terminate
- Reduced operational costs
- Deploy in Minutes
- Enhanced scalability, better performance, and lower infrastructure and maintenance costs

#### WHAT MAKES IT SERVERLESS?

- EMR Cluster is launched only when there are files to process.
- Lambda function is triggered by a cloudwatch event
- Lambda function contains the code to launch an EMR Cluster.
- Cluster is terminated automatically once the processing is over.

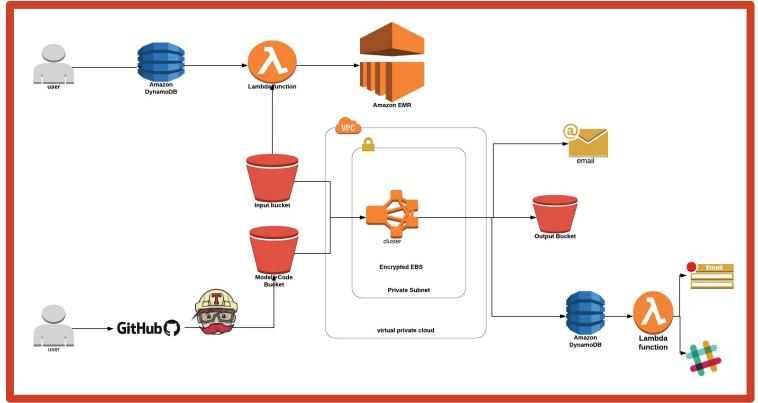


#### IDEAL SPARK APPLICATIONS



**Input Data** 

# Serverless Data processing pipeline



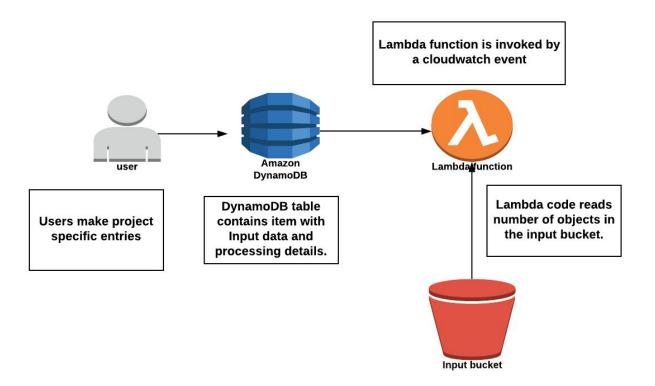
#### WHERE IS DATA STORED?

- Scales just keep putting files, and it will never fill up.
- Upload and download your data with SSL encrypted end points
- Provides multiple options for encrypting data at rest.
- **Low Cost -** \$0.023 per GB
- Reading data in a Spark application is as simple as calling -

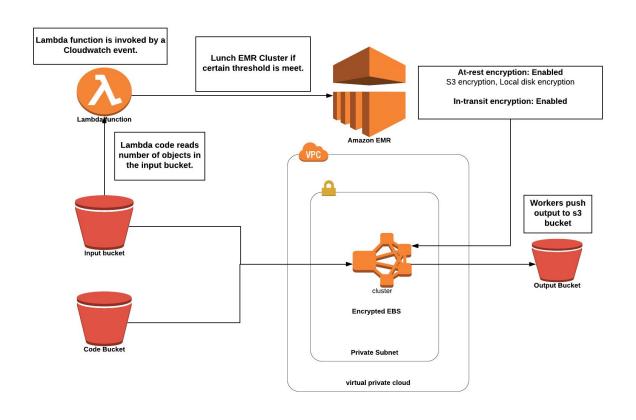
sc.textFile("s3n://<bucketname>")



#### WHAT TRIGGERS THE PIPELINE?

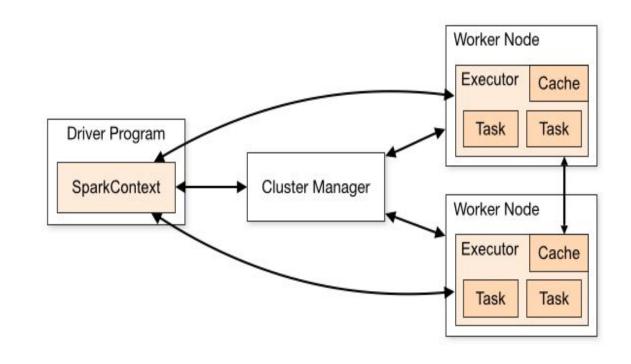


#### BATCH PROCESSING WITH EMR



#### APACHE SPARK ARCHITECTURE

- Single Master & Multiple Workers.
- Driver Program resides on Master node.
- Multiple executors may reside on a worker node.



#### WHAT ROLE DOES A DRIVER PLAY IN A SPARK ARCHITECTURE?

- Runs the main () function of the application and is the place where the Spark Context is created
- Translates the RDD's into the execution graph and splits the graph into multiple stages
- Converts User Application into tasks
- Exposes the information about the running spark application through a Web UI at port 4040.



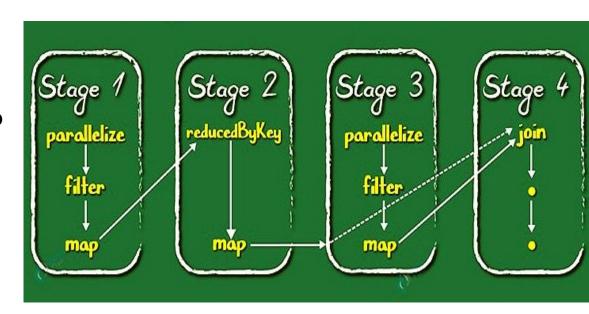
So, Smart Driver is more important!

#### ROLE OF CLUSTER MANAGER

- Responsible for acquiring resources on the spark cluster and allocating them to a spark job.
- 3 different types of cluster managers:
  - Hadoop YARN
  - Apache Mesos
  - Simple standalone spark cluster manager
- Allocation and deallocation of various physical resources such as memory for client spark jobs, CPU memory, etc.

#### DAG SCHEDULER

- Stages depends on transformations on RDDs
- Narrow transformation:
   Doesn't require the data to
   be shuffled across the
   partitions. for example,
   Map, filter etc.
- Wide transformation requires the data to be shuffled for example, reduceByKey etc.



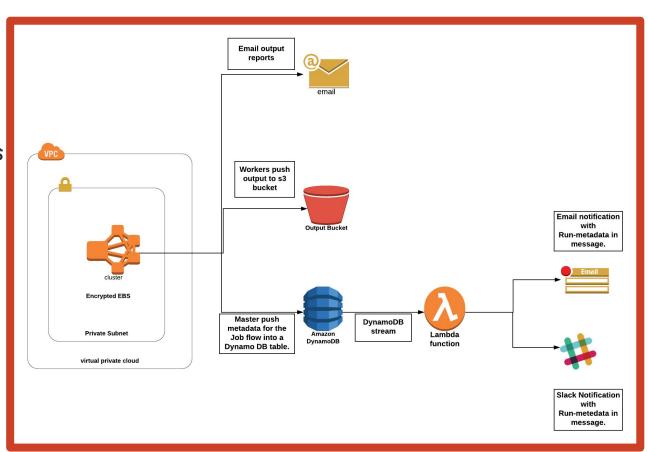
Wide transformations results in stages

#### RUN TIME ARCHITECTURE OF A SPARK APPLICATION

- 1. Creates a logical DAG from the RDD transformation and actions given in the application
- 2. Physical execution plan with set of stages
- 3. Creates tasks and send them to the Cluster
- 4. Cluster manager then launches executors on the worker nodes depending on resource allocation
- 5. Executors registers themselves with the driver program.
- 6. Executors start executing the various tasks assigned by the driver program
- 7. If main()  $\Rightarrow$  exits the call SparkContext.stop()

#### NOTIFICATION SYSTEM

After every run, a notification is sent to the user which contains run-metadata in the message

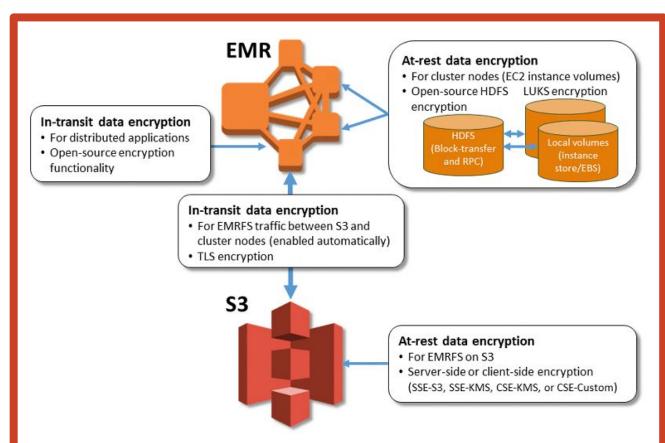


#### DATA SECURITY IN OUR ARCHITECTURE

Security configurations:

Templates for security configurations which can be attached to EMR Clusters.

- SC provides:
  - At-rest data encryption
  - In-transit data encryption



#### CODE MAINTENANCE

- Our code base resides on Git.
- EMR fetches code from s3
   bucket so it's necessary to
   create a sync between Git and
   s3
- Travis CI continuous integration service
- Travis CI sync the S3 bucket with git repo on every commit to the master branch



## QUESTIONS?

