CS 696 Intro to Big Data: Tools and Methods Fall Semester, 2020 Doc 19 Spark on AWS Mar 24, 2020

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Virus infection Simulation

Assumptions
Infect 10% of people you meet
Contact 20 people per day

New Cases Each Day

0%	50%	80%	
Immune	Immune Immur		
I	1	I	
2	1	0	
6	2	1	
18	4	I	
54	8	1	
162	16	2	
486	32	2	
I,458	64	3	
4,374	128	4	
13,122	256	6	
39,366	512	8	
118,098	1,024	12	
354,294	2,048	16	

Assumptions
Infect 10% of people you meet
0% Immune

New Cases Each Day

Contact 5 People	Contact 10 People	Contact 20 People	Contact 25 People
1	I	I	I
I	1	2	3
1	2	6	9
1	4	18	31
2	8	54	107
3	16	162	375
4	32	486	1,313
6	64	1,458	4,596
9	128	4,374	16,085
13	256	13,122	56,297
19	512	39,366	197,039
29	1,024	118,098	689,637
43	2,048	354,294	2,413,729

https://medium.com/@tomaspueyo/coronavirus-act-today-or-people-will-die-f4d3d9cd99ca https://tinyurl.com/w49ms3s

Chart 18: Slide from a Webinar of the American Hospital Association, communicating best guesses on the impact of the Coronavirus in the US healthcare system in 2020

Best Guess Epidemiology

• Ro = 2.5; Doubling time 7-10 days Community epi wave 2 months

• Community attack rate = 30-40% US: 96 million cases

• Cases requiring hospitalization = 5% US: 4.8 million admissions

• Cases requiring ICU care = 1-2% US: 1.9 million ICU

• Cases requiring ventilatory support = 1% US: 1 PPV

• CFR = 0.5% US: 480,000 deaths

PREPARE FOR DISEASE BURDEN ROUGHLY 10X SEVERE FLU SEASON

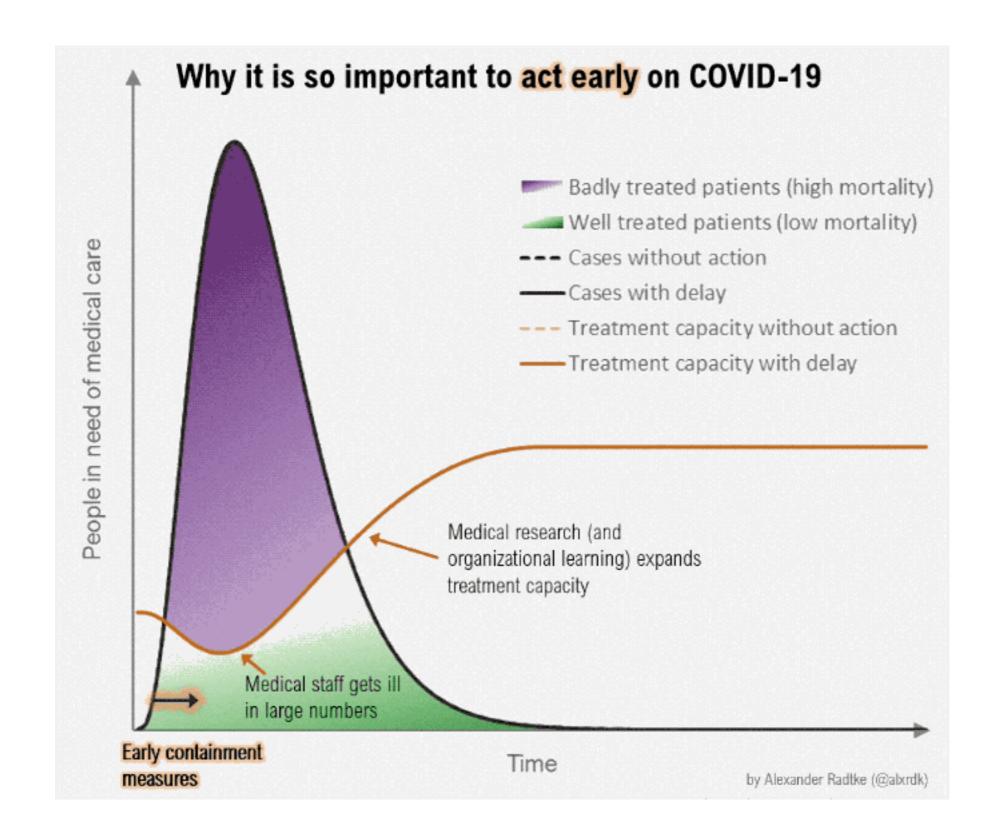




AHA webinar

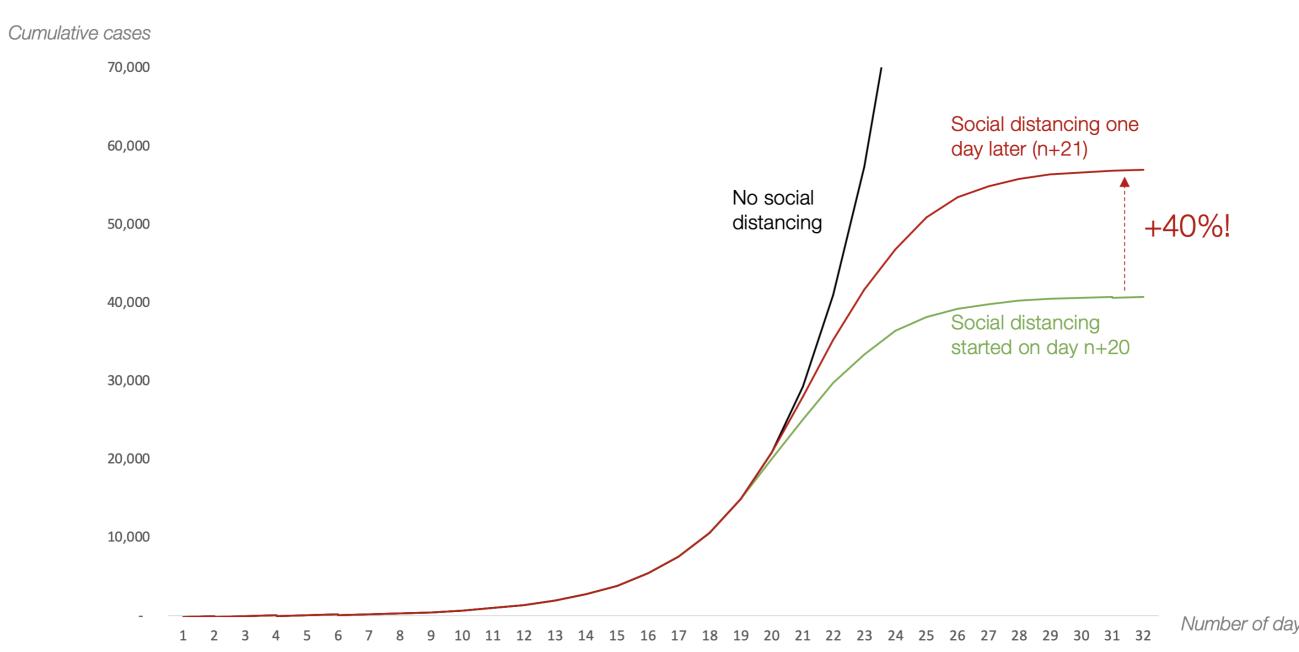
Source: Dr. James Lawler, professor at the University of Nebraska Medical Center, for the American Hospital Association, via Business Insider, https://www.businessinsider.com/presentation-us-hospitals-preparing-for-millions-of-hospitalizations-2020-3

https://tinyurl.com/w49ms3s



https://tinyurl.com/w49ms3s

Chart 23: Model of Cumulative Cases of Coronavirus with Social Distancing Measures Taken One Day Apart



Why We Sleep

By Mathew Walker

Professor of neuroscience and psychology

Director of the Center for Human Sleep Science at the University of California, Berkeley

Effect of lack of sleep Immune system Learning

Amazon Elastic Map-Reduce (EMR)

Hadoop, Hive, Spark, etc on Cluster

Predefined set of languages/tools available

Can create cluster of machines

https://aws.amazon.com

Create new account

Get 12 months free access

AWS Free Tier

12 months free

EC2 - compute instances
740 hours per month
Billed in hour increments
Billed per instance

S3 - storage 5 GB 20,000 Get requests

RDS - MySQL, PostgresSQL, SQL Sever 20 GB 750 hours

EC2 Container - Docker images 500 MB

I and students were charged last year

AWS Educate

https://aws.amazon.com/education/awseducate/

SDSU is an institutional member

Students get \$100 credit

EC2 Pricing

	Price Per Hour		
	On Demand	Spot	
a I .medium	\$0.0255	\$0.0050	
t3.nano	\$0.0058	\$0.0016	
m5.large	\$0.0960	\$0.0202	
c5.large	\$0.0850	\$0.0200	
p3.2xlarge (GPU)	\$3.0600	\$0.9413	

Basic Outline

Develop & test Spark locally

Upload program file & data to S3

Configure & launch cluster
AWS Management Console
AWS CLI
SDKs

Monitor cluster

Make sure you terminate cluster when done

Simple Storage System - S3

Files are stored in buckets

Bucket names are global

Supports

s3 - files divided in to block s3n

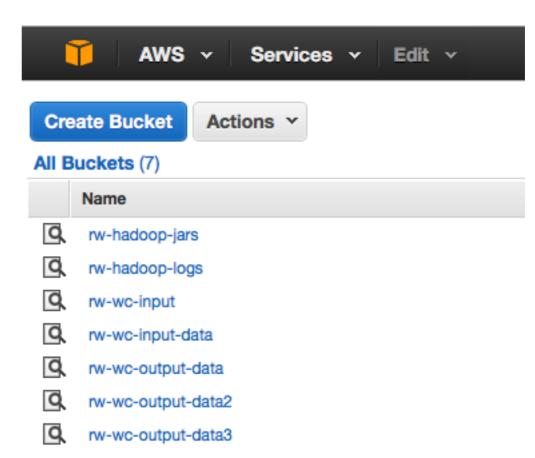
Accessing files

S3 console

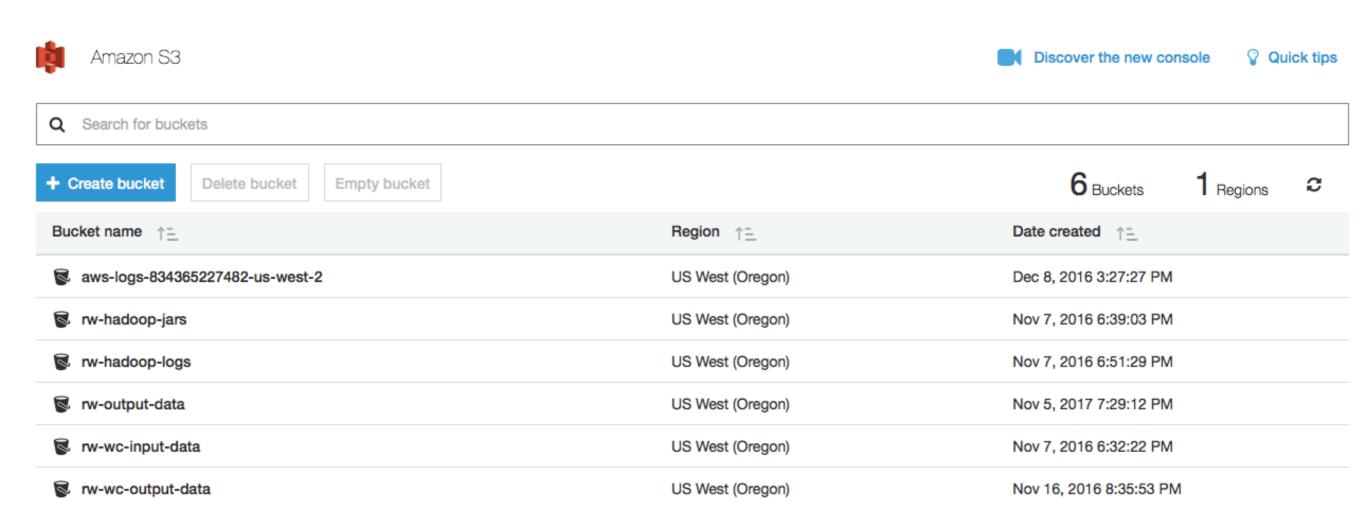
Third party

REST

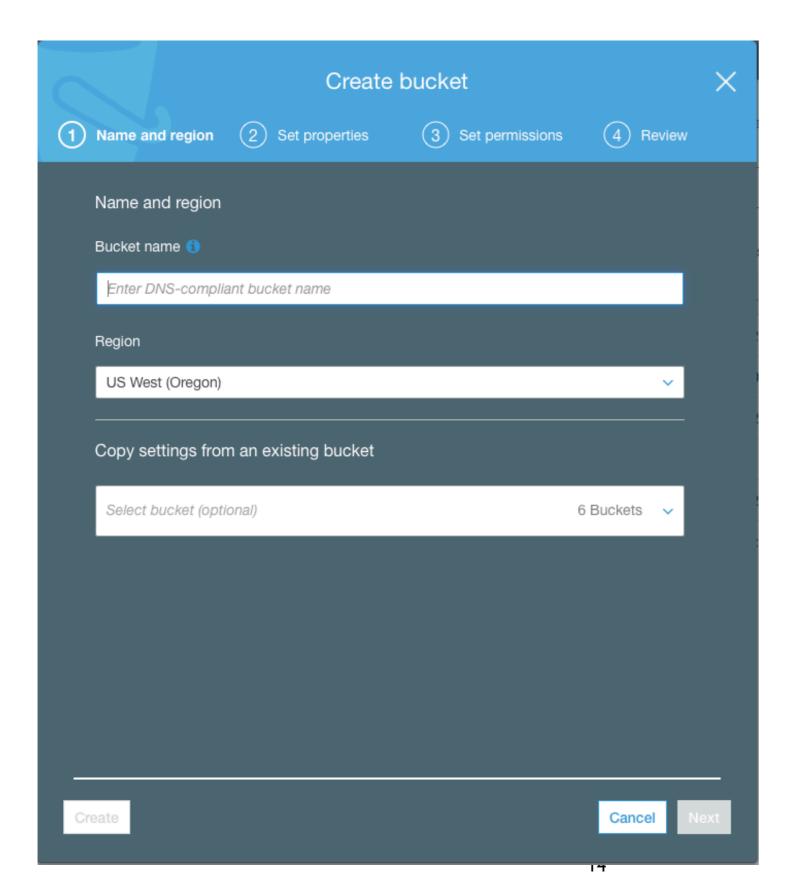
Java, C#, etc



Amazon S3



S3 Creating a Bucket



S3 Costs

AWS Free Usage Tier

New AWS customers receive each month for one year 5 GB of Amazon S3 storage in the Standard Storage class, 20,000 Get Requests, 2,000 Put Requests, and 15 GB of data transfer out

	Standard Storage	Standard - Infrequent Access Storage	Glacier Storage
First 50 TB / month	\$0.023 per GB	\$0.0125 per GB	\$0.004 per GB
Next 450 TB / month	\$0.022 per GB	\$0.0125 per GB	\$0.004 per GB
Over 500 TB / month	\$0.021 per GB	\$0.0125 per GB	\$0.004 per GB

S3 Objects

Objects contain
Object data
Metadata

Size

1 byte to 5 gigabytes per object

Object data

Just bytes

No meaning associated with bytes

Metadata

Name-value pairs to describe the object Some http headers used Content-Type

S3 Buckets

Namespace for objects

No limitation on number of object per bucket

Only 100 buckets per account

Each bucket has a name
Up to 255 bytes long
Cannot be same as existing bucket name by any S3 user

Bucket Names

Bucket names must

Contain lowercase letters, numbers, periods (.), underscores (_), and dashes (-)

Start with a number or letter

Be between 3 and 255 characters long

Not be in an IP address style (e.g., "192.168.5.4")

To conform with DNS requirements, Amazon recommends

Bucket names should not contain underscores (_)

Bucket names should be between 3 and 63 characters long

Bucket names should not end with a dash

Bucket names cannot contain dashes next to periods (e.g.,

"my-.bucket.com" and "my.-bucket" are invalid

Key

Unique identifier for an object within a bucket

Object Url

http://buckerName.s3.amazonaws.com/Key

http://doc.s3.amazonaws.com/2006-03-01/AmazonS3.wsdl

Bucket = doc

Key = 2006-03-01/AmazonS3.wsdI

Access Control Lists (ACL)

Each Bucket has an ACL
Determines who has read/write access

Each Object can have an ACL
Determines who has read/write access

ACL consists of a list of grants

Grant contains

One grantee

One permission

S3 Data Consistency Model

Updates to a single object at a key in a bucket are atomic

But a read after a write may return the old value Changes may take time to progate

No object locking

If two writes to same object occur at the same time

The one with later timestamp wins

CAP Theorem

CAP theorem says in a distributed system you can not have all three

Consistency

Availability

Tolerance to network Partitions

Consistency

Machine 1

Machine 2

$$A = 2$$

$$A = 2$$

Not Consistent

$$A = 2$$

$$A = 3$$

Partition

$$A = 2$$

$$A = 2$$

Partitioned

$$A = 2$$

$$A = 2$$

Machine 1 cannot talk to machine 2

But how does machine 1 tell the difference between no connection and a very slow connection or busy machine 2?

Latency

Latency

Time between making a request and getting a response

Distributed systems always have latency

In practice detect a partition by latency

When no response in a given time frame assume we are partitioned

Available

Machine 1

Machine 2

Client

A = 2

A = 2

Client

A = 2

A = 2

Client can not access value of A

What does not available mean?

No connection

Slow connection

What is the difference?

Some say high available - meaning low latency

In practice available and latency are related

Consistency over Latency

Machine 1

Machine 2

Set A to 3

A = 2

A = 2

Set A to 3

A = 2

Lock A

A = 2

Set A to 3

A = 2

Set A to 3

A = 2

Increased latency

System still available

queued until unlocked

Write requests

Set A to 3

A = 3

Unlock A

A = 3

A = 3

A = 3

Latency over Consistency

Machine 1

Machine 2

Set A to 3

$$A = 2$$

$$A = 2$$

$$A = 3$$

$$A = 2$$

$$A = 3$$

$$A = 3$$

$$A = 3$$

Set A to 3

Latency over Consistency - Write Conflicts

Machine 1

Machine 2

Set A to 3

$$A = 2$$

$$A = 2$$

Subtract 1 from A

$$A = 3$$

$$A = 2$$

$$A = 3$$

$$A = I$$

Subtract 1 from A

$$A = ?$$

$$A = ?$$

Need policy to make system consistent

Partition

					4
\mathbb{N}	la	ch	nin	9	1
IV	ıu	VI.			

Machine 2

$$A = 2$$

$$A = 2$$

$$A = 2$$

$$A = 2$$

Set A to 3

$$A = 3$$

$$A = I$$

Subtract 1 from A

$$A = ?$$

$$A = ?$$

Need policy to make system consistent

CAP Theorem

Not a theorem

Too simplistic
What is availability
What is a partition of the network

Misleading

Intent of CAP was to focus designers attention on the tradeoffs in distributed systems

How to handle partitions in the network Consistency
Latency

Availability

CAP & S3

S3 favors latency over consistency

Running Program on AWS EMR

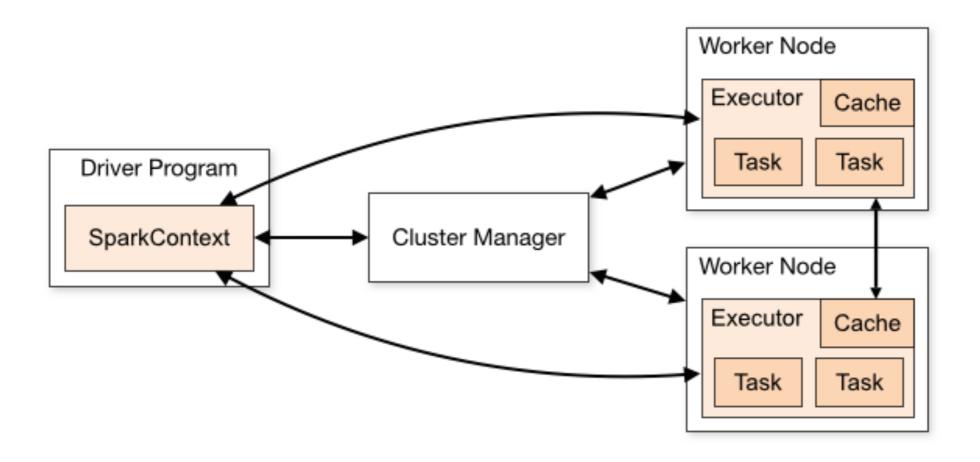
Make sure program runs locally

Create program file containing code

```
Create s3 bucket(s) for program file file logs input output
```

Upload program & data files to s3

Spark Components



Terms

Application

User program built on Spark

Driver program + executors

Driver program

The process running the main() function of the application and creating the SparkContext

Cluster manager

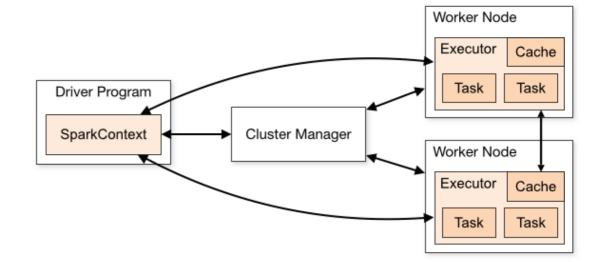
External service for acquiring resources on the cluster

Deploy mode

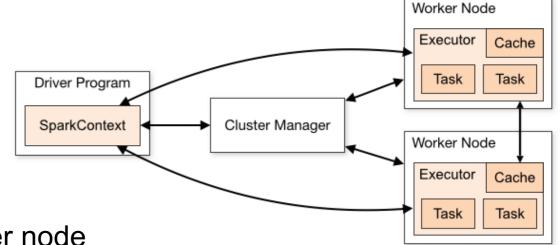
Where the driver process runs

"cluster" - the driver inside of the cluster

"client" - the driver outside of the cluster



Terms



Executor

A process launched for an application on a worker node

Runs tasks and keeps data in memory or disk storage across them.

Each application has its own executors

Task

A unit of work that will be sent to one executor

Job

A parallel computation consisting of multiple tasks

Gets spawned in response to a Spark action (e.g. save, collect)

Stage

Job divided into smaller tasks called stages

Depend on each other

Test Program 1 - Pi

from random import random from operator import add

from pyspark import SparkContext

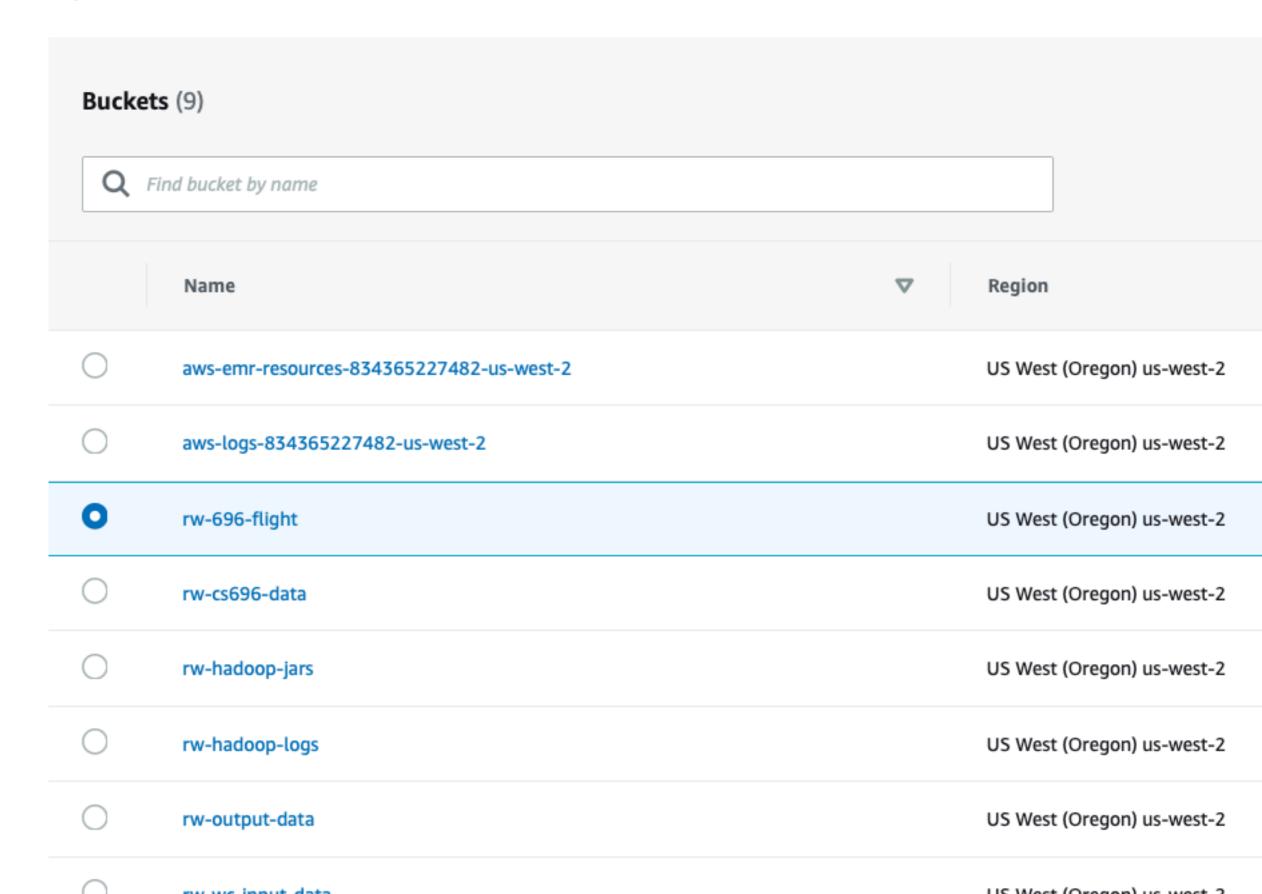
```
if _name__ == "__main__":
  sc = SparkContext(appName="PythonPi")
  partitions = 3
  n = 100000 * partitions
  def f():
     x = random() * 2 - 1
     y = random() * 2 - 1
     return 1 if x ** 2 + y ** 2 < 1 else 0
  count = sc.parallelize(range(1, n + 1), partitions).map(f).reduce(add)
  print("Pi is roughly %f" % (4.0 * count / n))
  sc.stop()
```

Designed to have no

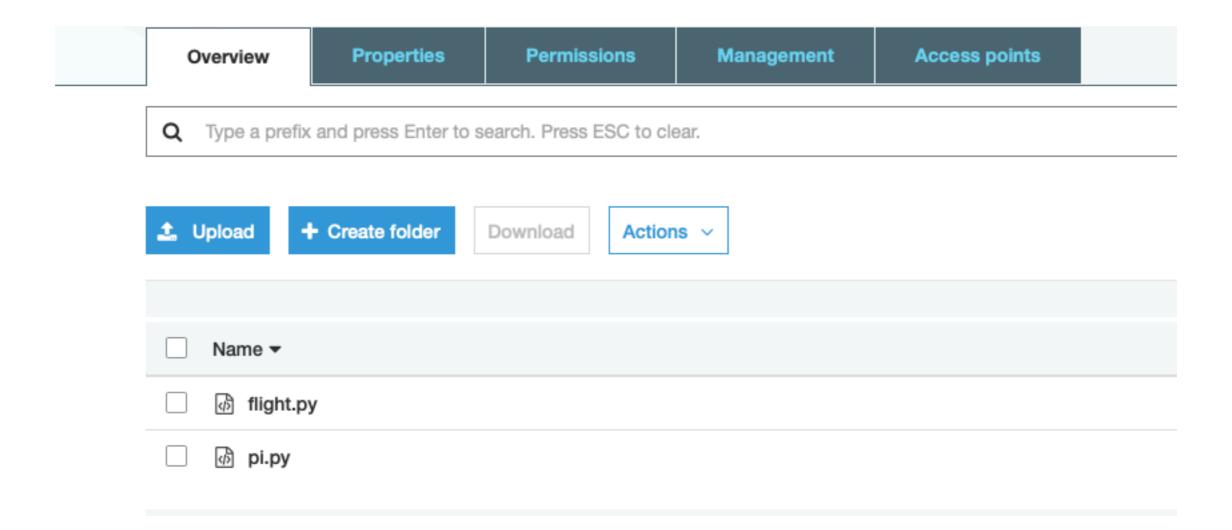
Command line dependancies

No input or output files

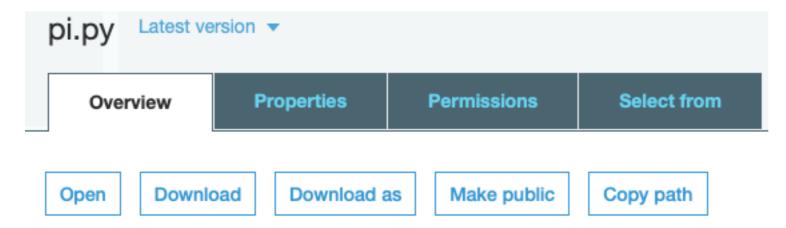
My S3 Buckets



My S3 Buckets



My S3 Buckets



Owner

rwhitney

Last modified

Mar 27, 2019 8:58:15 PM GMT-0700

Etag

2a37d19cc32e1a51c0472473e26f72d1

Storage class

Standard

Server-side encryption

None

Size

498.0 B

Key

pi.py

Object URL

https://rw-696-flight.s3-us-west-2.amazonaws.com/pi.py

Spark on AWS - EMR Console

Amazon EMR Clusters

Security configurations

Block public access

VPC subnets

Events

Notebooks

Git repositories

Help

What's new

Welcome to Amazon Elastic MapReduce

Create

Amazon Elastic MapReduce (Amazon EMR) is a web service that enables businesses, researchers, data analysts, and developers to easily and cost-effectively process vast amounts of data.

You do not appear to have any clusters. Create one now:

Create cluster

How Elastic MapReduce Works

Upload

Upload your data and processing application to S3.

Learn more



Configure and create your cluster by specifying data inputs, outputs, cluster size, security settings, etc.

Learn more

Monitor



Monitor the health and progress of your cluster. Retrieve the output in S3.

Learn more

Using Quick Options

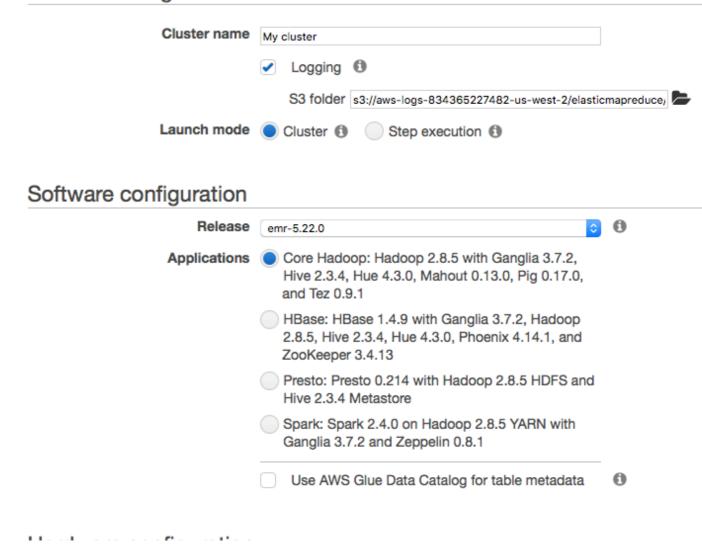
General Configuration

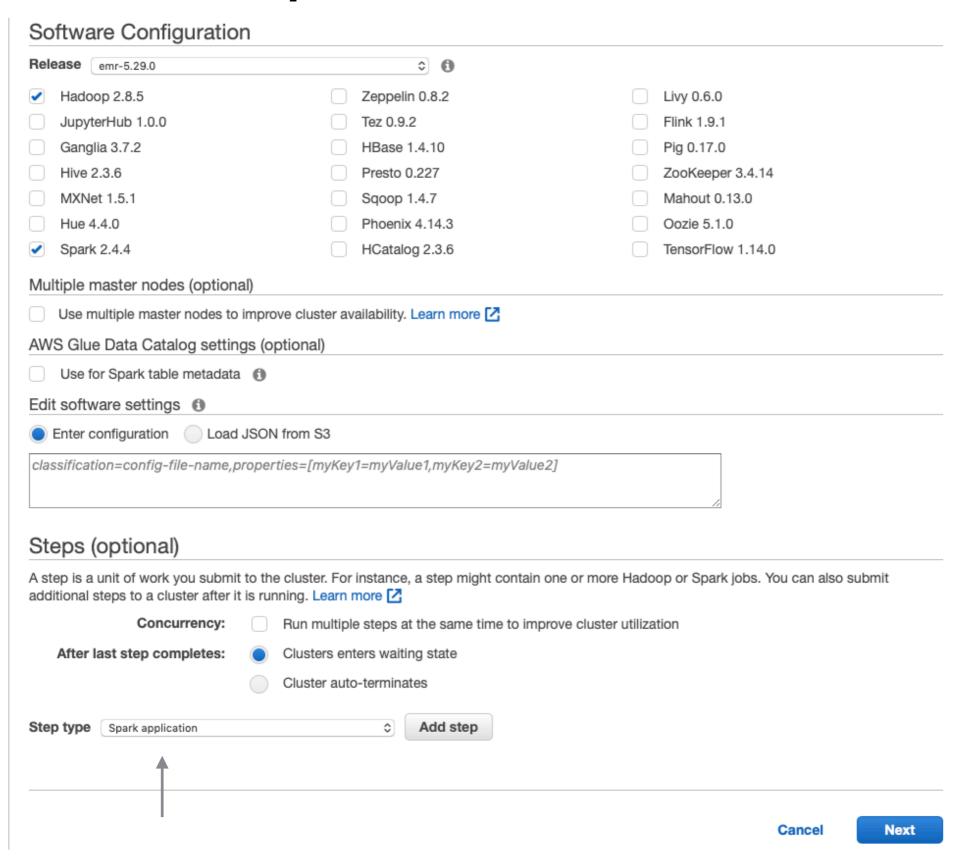


Use Advanced Options

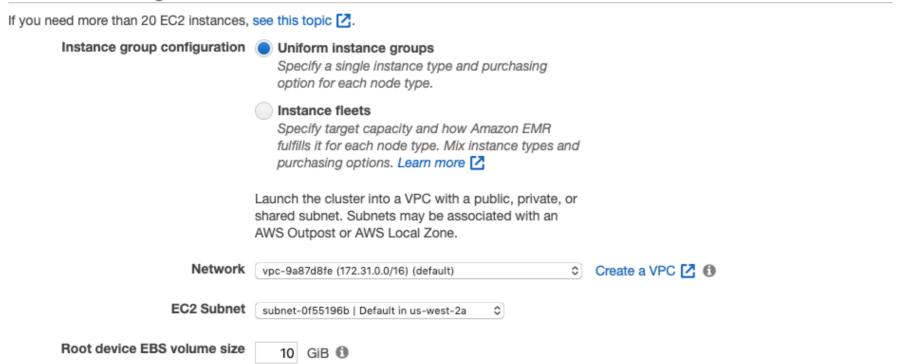
Create Cluster - Quick Options Go to advanced options

General Configuration

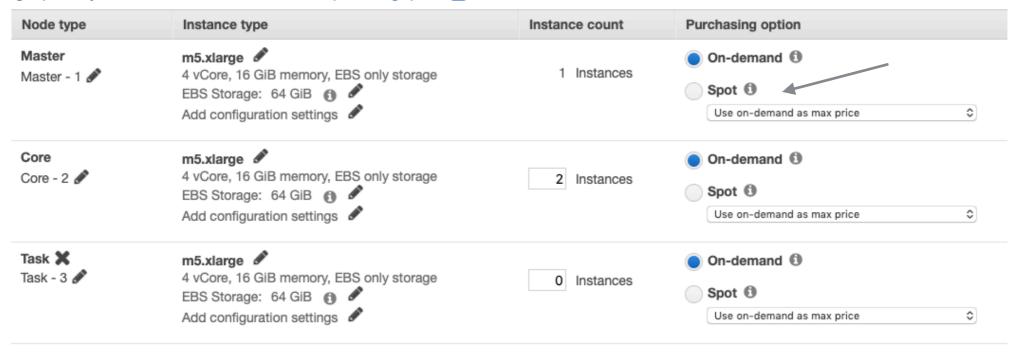


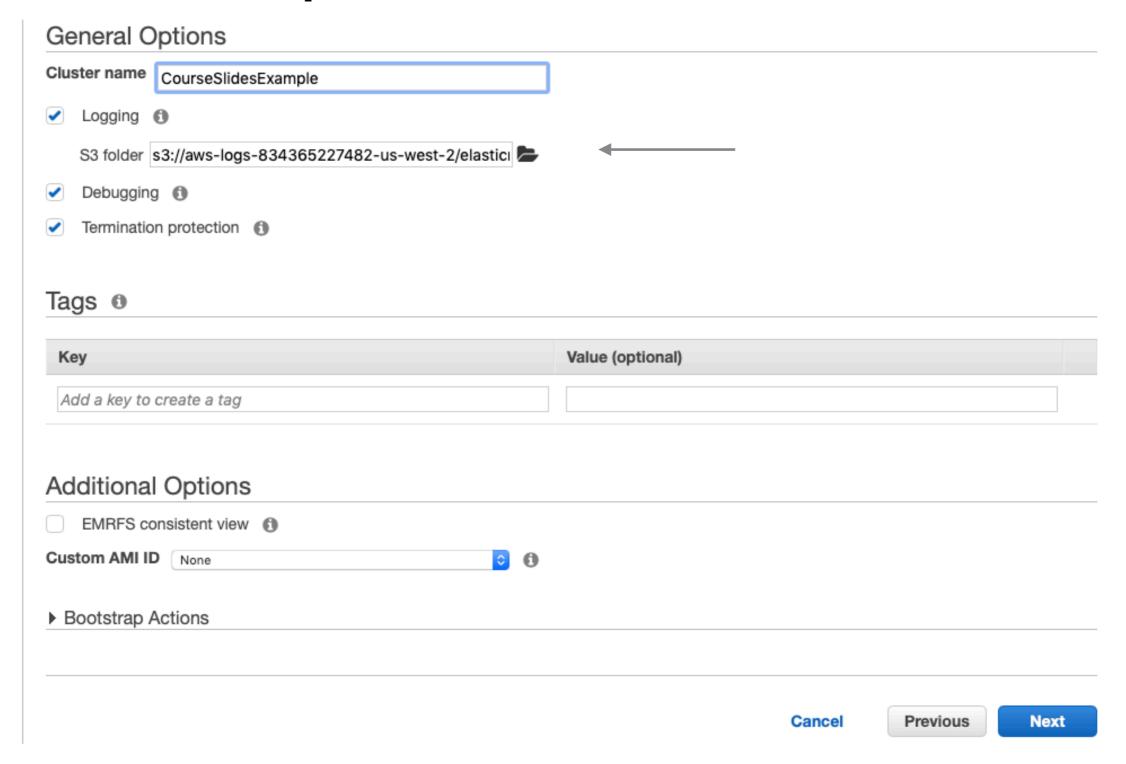


Hardware Configuration 6



Choose the instance type, number of instances, and a purchasing option. You can choose to use On-Demand Instances, Spot Instances, or both. The instance type and purchasing option apply to all EC2 instances in each instance group, and you can only specify these options for an instance group when you create it. Learn more about instance purchasing options





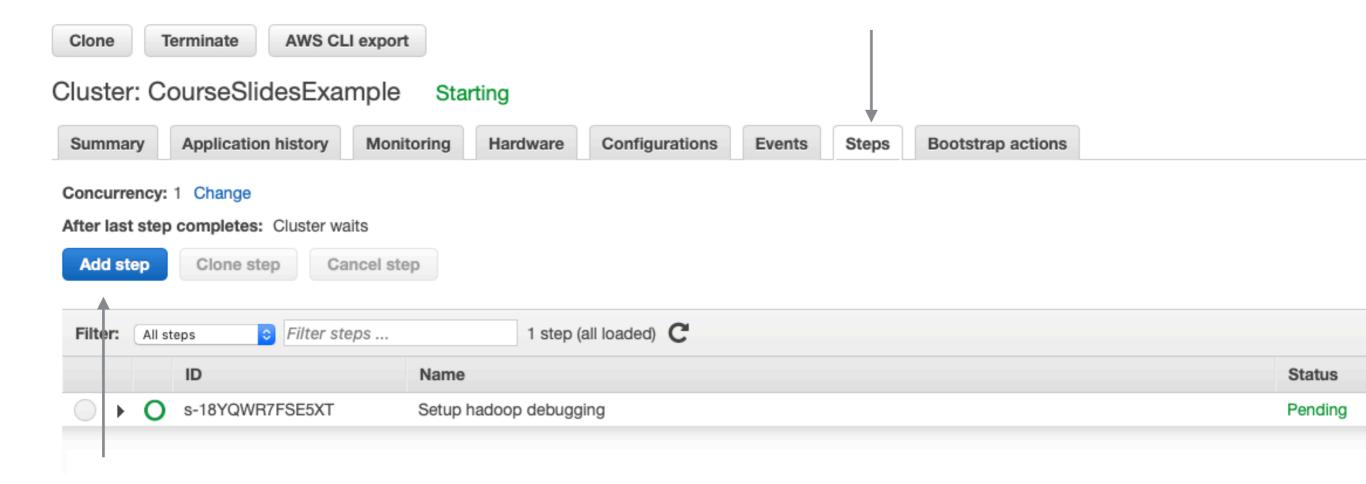
Security Options EC2 key pair Proceed without an EC2 key pair Cluster visible to all IAM users in account Permissions (1) Custom Default Use default IAM roles. If roles are not present, they will be automatically created for you with managed policies for automatic policy updates. EMR role EMR_DefaultRole [2] 1 EC2 instance profile EMR_EC2_DefaultRole [2] (1) Auto Scaling role EMR_AutoScaling_DefaultRole [2] Security Configuration ▶ EC2 security groups

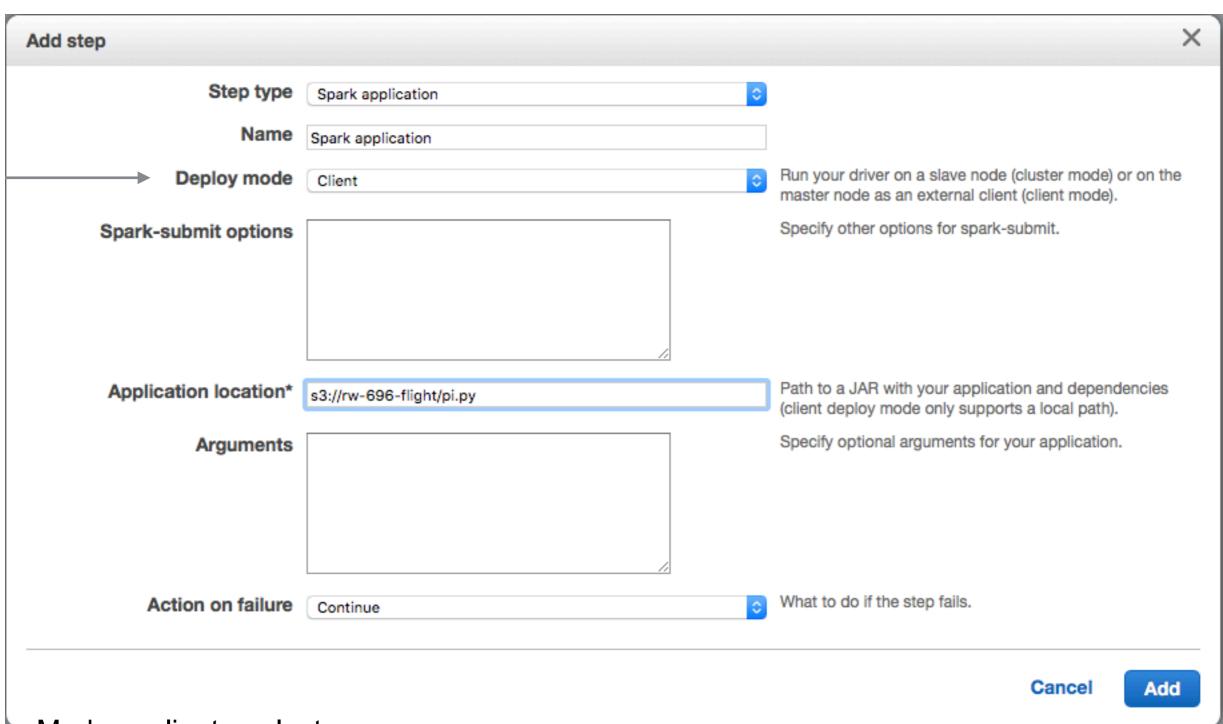


Cluster Created - Either Quick or Advanced



Adding a Step

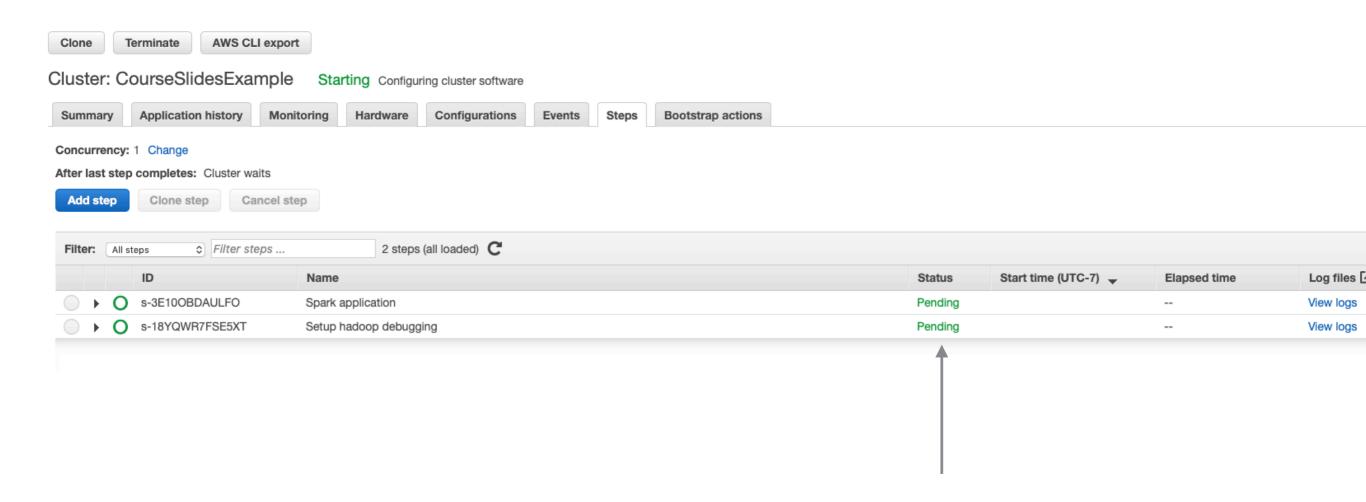


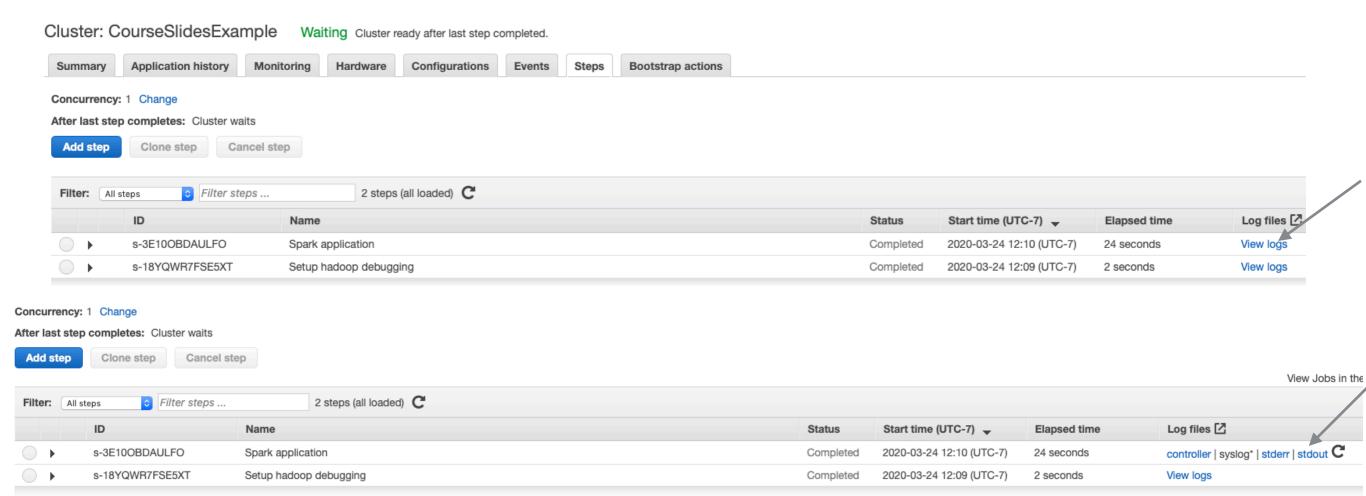


Modes - client or cluster

Either works

client mode gives access to standard out





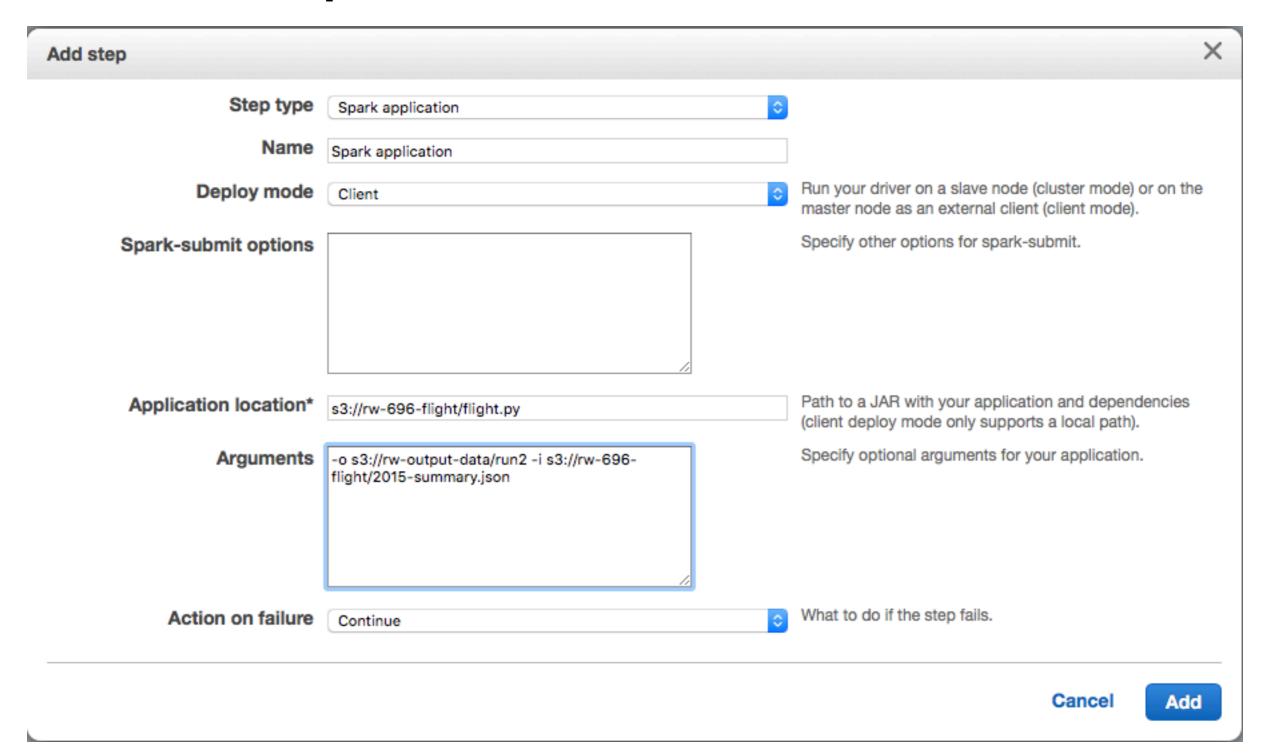
Example 2

```
def flight(input, output):
  import pyspark.sql.functions as F
  from pyspark.sql import SparkSession
  spark = SparkSession.builder \
     .appName("Fight") \
     .getOrCreate()
  flight df = spark.read.json(input)
  grouped_df = flight_df.groupBy('DEST_COUNTRY_NAME').agg(F.sum('count'))
  grouped df.write.format('csv').save(output)
def files_from_args():
    import argparse
    parser = argparse.ArgumentParser()
    parser.add_argument('-i', '--input', default='input')
    parser.add_argument('-o', '--output',default='output')
    args = parser.parse_args()
    return (args.input, args.output)
if name == " main ":
    inputfile, outputfile = files_from_args()
    flight(inputfile, outputfile)
                                            53
```

S3 Buckets

▶ 词 aws-emr-resources-834365227482-us-west-2	•	
▶ 3 aws-logs-834365227482-us-west-2		
▼ 🧓 rw-696-flight		
2015-summary.json	21 KB	3/26/19
flight.py	2 KB	3/26/19
ø pi.py	49tes	3/27/19
▶ 🗊 rw-cs696-data		
▶		

Added Step



S3 output

▼ 👼 rw-output-data			
▼ im run2			
_SUCCESS	Zero KB	3/28/19	
part-00000-d6b0488c-30c4-472a-9a53-0a	Zero KB	3/28/19	
part-00002-d6b0488c-30c4-472a-9a53-0a	35tes	3/28/19	
part-00008-d6b0488c-30c4-472a-9a53-0a	11tes	3/28/19	
part-00009-d6b0488c-30c4-472a-9a53-0a	11tes	3/28/19	
part-00010-d6b0488c-30c4-472a-9a53-0a	12tes	3/28/19	
part-00012-d6b0488c-30c4-472a-9a53-0a	10tes	3/28/19	
part-00013-d6b0488c-30c4-472a-9a53-0abd52629b65-c000.csv			
part-00014-d6b0488c-30c4-472a-9a53-0a	11tes	3/28/19	
part-00015-d6b0488c-30c4-472a-9a53-0a	23tes	3/28/19	
part-00016-d6b0488c-30c4-472a-9a53-0a	8 bytes	3/28/19	
part-00017-d6b0488c-30c4-472a-9a53-0a	11tes	3/28/19	
part-00021-d6b0488c-30c4-472a-9a53-0a	7 bytes	3/28/19	
part-00022-d6b0488c-30c4-472a-9a53-0a	13tes	3/28/19	
part-00026-d6b0488c-30c4-472a-9a53-0a	10tes	3/28/19	
part-00029-d6b0488c-30c4-472a-9a53-0a	9 bytes	3/28/19	
part-00030-d6b0488c-30c4-472a-9a53-0a	40tes	3/28/19	
part-00031-d6b0488c-30c4-472a-9a53-0a	10tes	3/28/19	
part-00032-d6b0488c-30c4-472a-9a53-0a	9 bytes	3/28/19	
part-00038-d6b0488c-30c4-472a-9a53-0a	50tes	3/28/19	
part-00040-d6b0488c-30c4-472a-9a53-0a	10tes	3/28/19	
part-00044-d6b0488c-30c4-472a-9a53-0a	14tes	3/28/19	
part-00045-d6b0488c-30c4-472a-9a53-0a	22tes	3/28/19	
I			

Warning on AWS

It can take 5-10 minutes to start cluster

Logs do not show your logging statements

When you configure Steps incorrectly they fail Error messages are not very helpful

SSH to your Master Node

Create Amazon EC2 Key pair

Instructions

http://docs.aws.amazon.com/AWSEC2/latest/UserGuide/ec2-key-pairs.html#having-ec2-create-your-key-pair

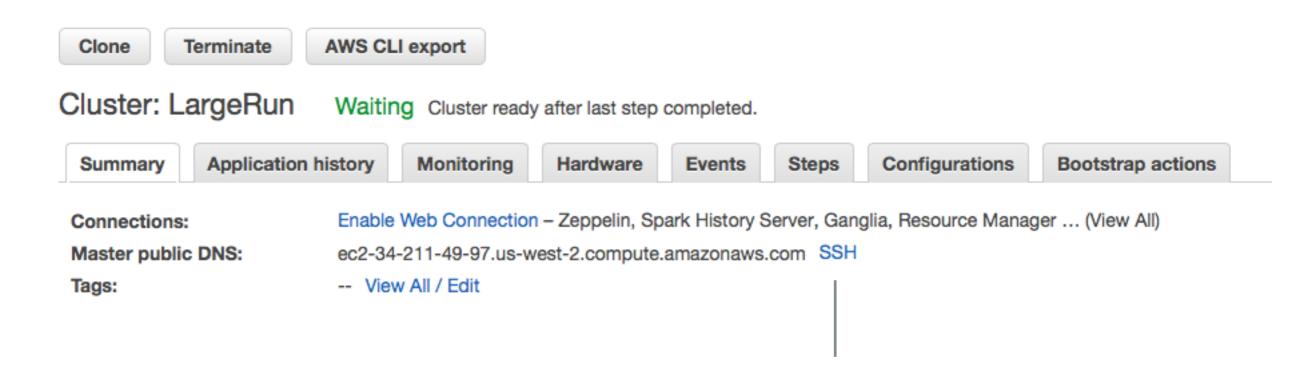
Open EC2 Dashboard - Select Key Pairs

SSH to your Master Node

EC2 instance profile EMR_EC2_DefaultRole

In Create Cluster - Quick Options HIVE 2.3.0 Metastore Spark: Spark 2.2.0 on Hadoop 2.7.3 YARN with Ganglia 3.7.2 and Zeppelin 0.7.2 Use AWS Glue Data Catalog for table metadata Hardware configuration Instance type 0 m3.xlarge Number of instances 3 (1 master and 2 core nodes) Security and access EC2 key pair Choose an option Learn how to create an EC2 key pair. Permissions Default Custom Use default IAM roles. If roles are not present, they will be automatically created for you with managed policies for automatic policy updates. EMR role EMR_DefaultRole (1)

SSH to your Master Node



Click for Instructions

Command-line Tools

Flintrock

Open-source command-line tool for launching Apache Spark clusters

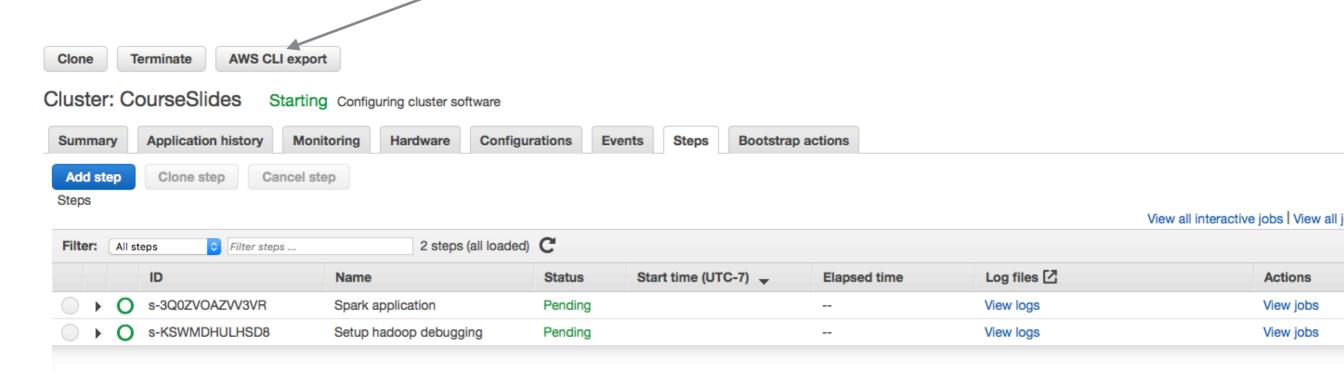
https://github.com/nchammas/flintrock

aws cli

Amazon's command line tool

https://aws.amazon.com/cli/

Generating the Command Line



AWS CLI export



aws emr create-cluster --termination-protected --applications Name=Hadoop Name=Spark --ec2-attributes
'{"InstanceProfile":"EMR_EC2_DefaultRole", "SubnetId": "subnet-0f55196b", "EmrManagedSlaveSecurityGroup": "sg-65bffa1c", "EmrManagedMasterSecurityGroup": "sg-62bffa1b"}' --release-label emr-5.22.0 --log-uri 's3n://aws-logs-834365227482-us-west-2/elasticmapreduce/' --steps '[{"Args":["spark-submit","--deploy-mode", "client", "s3://rw-696-flight/pi.py"], "Type": "CUSTOM_JAR", "ActionOnFailure": "CONTINUE", "Jar": "command-runner.jar", "Properties": "", "Name": "Spark application"}]' --instance-groups
'[{"InstanceCount":1, "InstanceGroupType": "MASTER", "InstanceType": "m3.xlarge", "Name": "Master - 1"},
{"InstanceCount":2, "InstanceGroupType": "CORE", "InstanceType": "m3.xlarge", "Name": "Core - 2"}]' --auto-scaling-role EMR_AutoScaling_DefaultRole --ebs-root-volume-size 10 --service-role EMR_DefaultRole --enable-debugging --name 'CourseSlides' --scale-down-behavior TERMINATE_AT_TASK_COMPLETION --region us-west-2

Hadoop Ecosystem

```
Hadoop
 HDFS
 MapReduce
 YARN
Tez
Pig
Hive
Hbase
Sqoop
Oozie
Falcon
Spark
ZooKeeper
Mahout
Phoenix
BigTop
+ others
```

Apache Pig

Programming Map-Reduce can be low level

Apache Pig - high-level platform for creating programs for Hadoop

Pig Latin

ordered_word_count = ORDER word_count BY count DESC; STORE ordered_word_count INTO '/tmp/number-of-words-on-internet';

Apache Hive

SQL is common way to interact with data

Hive provides SQL like query language for HDFS, Amazon S3 data

HiveQL - converted into MapReduce

DROP TABLE IF EXISTS docs;

CREATE TABLE docs (line STRING);

LOAD DATA INPATH 'input_file' OVERWRITE INTO TABLE docs;

CREATE TABLE word_counts AS

SELECT word, count(1) AS count FROM

(SELECT explode(split(line, '\s')) AS word FROM docs) temp

GROUP BY word

ORDER BY word;

Apache HBase

BigTable for Hadoop

Non-relational distributed database

Fault-tolerant way of storing large quantites of sparse data

Apache Sqoop

People have data in non-hadoop databases

Sqoop

Transferring data between relational databases & Hadoop

Apache Phoenix

But SQL is common

Phoenix

Massively parallel relational database for Hadoop Uses HBase to store data

Apache Spark

Hadoop has latency issues - reads data from disk MapReduce is not conducive to solving all problems

Spark

Uses distributed shared memory: Resilient distributed dataset (RDD) Iterative algorithms
Implemented in Scala

Spark Core
Spark SQL
Dataframes & SQL
Spark Streaming
Spark MLlib
Machine learning

Apache Mahout

Hadoop does not have machine learning libraries

Mahout

Environment for quickly creating scalable machine learning applications Samsara - R-line syntax & environment

Apache Flink, Apache Storm

Hadoop does batch jobs Spark streaming has delays

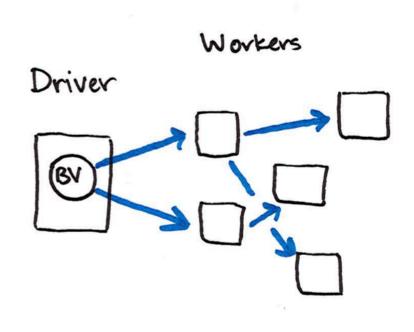
Fling & Storm
Each calin to have high throughput and low latency streaming

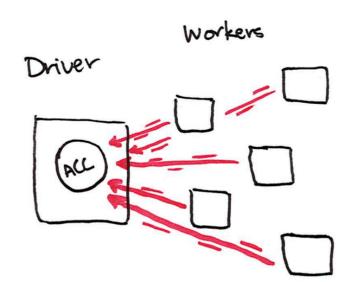
Distributed Variables

Broadcast

Read-only data shared among workers

Accumulator
Write only by workers
Read only on master





Broadcast Example

from pyspark.sql import SparkSession

```
spark = SparkSession\
     .builder\
     .appName("variables")\
     .getOrCreate()
courseSize = 45
courseSizeBroadcast = spark.sparkContext.broadcast(courseSize)
courseSizeBroadcast.value
data = spark.sparkContext.parallelize((1,2,3,4,5,6,7,8), 2)
data.map(lambda x: x + courseSizeBroadcast.value).collect()
```

Using ComplexType

```
sampleMap = {'a': 10,'bat':1 }
sampleBroadCast = spark.sparkContext.broadcast(sampleMap)
sampleBroadCast.value
```

```
import org.apache.spark.sql.SparkSession
val blockSize = "4096"
val spark = SparkSession.builder().
   appName("Broadcast Test").
   config("spark.broadcast.blockSize", blockSize).
   getOrCreate()
val sc = spark.sparkContext
val slices = 2
val num = 10000000
val arr1 = (0 until num).toArray
for (i <- 0 until 3) {
  println("Iteration " + i)
  println("=======")
  val startTime = System.nanoTime
  val barr1 = sc.broadcast(arr1)
  val observedSizes = sc.parallelize(1 to 10, slices).map(_ => barr1.value.length)
  observedSizes.collect().foreach(i => println(i))
  println("Iteration %d took %.0f milliseconds".format(i, (System.nanoTime - startTime) / 1E6))
```

Accumulator Example

```
from pyspark.sql import SparkSession
spark = SparkSession\
     .builder\
     .appName("variables")\
     .getOrCreate()
counter = spark.sparkContext.accumulator(0)
def count(item):
  global counter
  print("item: ", item.id)
  counter.add(1)
df = spark.range(16)
smaller = df.coalesce(4)
smaller.foreach(count)
counter.value
```

Output

16

Accumulator add() value

Numbers only

Can create custom accumulators

Machine Learning in Spark

MLlib

RDD-based org.apache.spark.mllib Maintenance mode

DataFrame based (Spark ML)
org.apache.spark.ml
Pipelines
Inspired by Python scikit-learn

Classification

Regression

Clustering

Collaborative Filtering

Dimension reduction

Linear Algebra

Statistics

http://spark.apache.org/docs/latest/ml-guide.html



Overview

Programming Guides *

API Docs ▼

Deploying *

More ▼

MLlib: Main Guide

- Basic statistics
- Data sources
- Pipelines
- Extracting, transforming and selecting features
- Classification and Regression
- Clustering
- Collaborative filtering
- Frequent Pattern Mining
- Model selection and tuning
- Advanced topics

Machine Learning Library (MLlib) Guide

MLlib is Spark's machine learning (ML) library. Its goal is to make practical machine learning scalable provides tools such as:

- ML Algorithms: common learning algorithms such as classification, regression, clustering, and co
- Featurization: feature extraction, transformation, dimensionality reduction, and selection
- Pipelines: tools for constructing, evaluating, and tuning ML Pipelines
- Persistence: saving and load algorithms, models, and Pipelines
- Utilities: linear algebra, statistics, data handling, etc.

Announcement: DataFrame-based API is prima

The MLlib RDD-based API is now in maintenance mode.

As of Spark 2.0, the RDD-based APIs in the spark.mllib package have entered maintenance mode.

Python Examples

\$SPARK_INSTALL_DIR/examples/src/main/python/ml

```
aft_survival_regression.py
                                                 index_to_string_example.py
als_example.py
                                                 isotonic_regression_example.py
binarizer_example.py
                                                 kmeans_example.py
bisecting_k_means_example.py
                                                 lda_example.py
bucketed_random_projection_lsh_example.py
                                                 linear_regression_with_elastic_net.py
bucketizer_example.py
                                                 linearsvc.py
chi_square_test_example.py
                                                 logistic_regression_summary_example.py
chisq_selector_example.py
                                                 logistic_regression_with_elastic_net.py
correlation_example.py
                                                 max_abs_scaler_example.py
count_vectorizer_example.py
                                                 min_hash_lsh_example.py
cross_validator.py
                                                 min_max_scaler_example.py
dataframe_example.py
                                                 multiclass_logistic_regression_with_elastic_i
dct_example.py
                                                 multilayer_perceptron_classification.py
decision_tree_classification_example.py
                                                 n_gram_example.py
decision_tree_regression_example.py
                                                 naive_bayes_example.py
elementwise_product_example.py
                                                 normalizer_example.py
estimator_transformer_param_example.py
                                                 one_vs_rest_example.py
feature_hasher_example.py
                                                 onehot_encoder_estimator_example.py
fpgrowth_example.py
                                                 pca_example.py
gaussian_mixture_example.py
                                                 pipeline_example.py
generalized_linear_regression_example.py
                                                 polynomial_expansion_example.py
gradient_boosted_tree_classifier_example.py
                                                 prefixspan_example.py
gradient_boosted_tree_regressor_example.py
                                                 quantile_discretizer_example.py
imputer_example.py
                                                 random_forest_classifier_example.py
                                         81
```