

INSTALLING AND RUNNING SPARK

First of all



Virtual Machine with all that you would need



VMWare Player:

https://my.vmware.com/en/web/vmware/free#desktop_en_d_user_computing/vmware_workstation_player/14_0

User: spark | Pass: spark | RootPass: spark

https://mega.nz/#!DdRnhAYJ!7PeO-U28OAadLb8KFYBsa0CDSSX74t9c6QdqWmEdA2A





- Downloading and Installing Spark
- □ Datasets: SUSY
- □Datasets: ECBDL_14
- MLlib: Machine Learning on Spark
 - Statistical
 - Spark-packages
 - □kNN-IS

Scala

- wget http://scala-lang.org/files/archive/scala-2.11.6.tgz
- □ tar xzf scala-2.11.6.tgz -C /usr/local
- cd /usr/local/
- □ In -s scala-2.11.6 scala
- vi /etc/profile.d/scala.sh



export SCALA_HOME=/usr/local/scala
export PATH=\${SCALA_HOME}/bin:\${PATH}

Maven

- wget http://mirror.cc.columbia.edu/pub/software/
 apache/maven/maven-3/3.0.5/binaries/apache maven-3.0.5-bin.tar.gz
- \square tar xzf apache-maven-3.0.5-bin.tar.gz -C /opt
- mv /opt/apache-maven-3.0.5/ /opt/maven
- vi /etc/profile.d/maven.sh
 export M2_HOME=/opt/maven
 - export PATH=\${M2_HOME}/bin:\${PATH}

Spark

- wget http://ftp.cixug.es/apache/spark/spark-
 2.2.0/spark-2.2.0-bin-hadoop2.7.tgz
- tar xzf spark-2.2.0-bin-hadoop2.7.tgz -C /opt/
- vi /etc/profile.d/spark.sh
 export SPARK_HOME=/opt/spark-2.2.0-bin-hadoop2.7
 export PATH=\${SPARK_HOME}/bin:\${PATH}

Scala IDE

- □ wget http://downloads.typesafe.com/scalaide-pack/4.7.0- vfinal-oxygen-212-20170929/scala-SDK-4.7.0-vfinal-2.12-linux.gtk.x86 64.tar.gz
- □ tar -xvzf scala-SDK-4.7.0-vfinal-2.12-linux.gtk.x86_64 -C opt

 chmod -R +r /opt/eclipse/

 vi /usr/bin/eclipse

 Scala | □ E
- □ vi /usr/bin/eclipse

#!/bin/sh

export ECLIPSE_HOME="/opt/eclipse"

\$ECLIPSE_HOME/eclipse \$*

chmod 755 /usr/bin/eclipse

Scala IDE

vi /usr/share/applications/eclipse.desktop

[Desktop Entry]

Encoding=UTF-8

Name=Eclipse



Exec=eclipse

lcon=/opt/eclipse/icon.xpm

Terminal=false

Type=Application

Categories=GNOME; Application; Development;

StartupNotify=true

Update Path

- source /etc/profile.d/*.sh
- □Log Out & Log In
- □ Restart

Outline



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Dataset: SUSY



SUSY

- □ Subset of 10,000 instances in the VM. (5M original)
- □ Number of instances: 10.000 + 10.000 (18 features)

training test

- Classification problem
- Real features



Dataset: SUSY



SUSY: Path

Local (Virtual Machine):

- Training: /home/spark/datasets/susy-10k-tra.data
- Test: /home/ spark /datasets/susy-10k-tst.data
- Header: /home/ spark /datasets/susy.header

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Dataset: ECBDL_14



ECBDL_14

Number of instances: 499,449 + 500,353 (631 features)
 training test

 \square Imbalance Ratio (IR): 70 % + 30 %

label: "0.0" "1.0"

Dataset: ECBDL_14



ECBDL_14: Path

□ Cluster (HDFS):

- Training: /user/spark/datasets/ECBDL14_mbd/ecbdl14tra.data
- Test: /user/spark/datasets/ECBDL14_mbd/ecbdl14tst.data
- Header: /user/spark/datasets/ECBDL14_mbd/ecbdl4.header

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Mllib: Running

Virtual Machine examples

Interactive Shell Scala (Virtual Machine):

- | /opt/spark-2.2.0-bin-hadoop2.7/bin/spark-shell
- Spark Submit (Virtual Machine):
- /opt/spark-2.2.0-bin-hadoop2.7/bin/spark-submit

Interactive Shell R (Virtual Machine):

/opt/spark-2.2.0-bin-hadoop2.7/bin/sparkR

Mllib: Running

Cluster examples

Interactive Shell Scala (Virtual Machine):

/opt/spark-2.2.0/bin/spark-shell

Spark Submit (Virtual Machine):

/opt/spark-2.2.0/bin/spark-submit

Interactive Shell R (Virtual Machine):

/opt/spark-2.2.0/bin/sparkR

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Mllib with Scala: Statistical

Basic Statistics

- Summary statistics
- Correlations
- Stratified sampling
- Hypothesis testing
 - Streaming Significance Testing
- Random data generation



https://github.com/JMailloH/BasicStatisticsSpark

https://spark.apache.org/docs/2.2.0/mllib-statistics.html#summary-statistics

Statistical: import

```
import org.apache.spark.SparkConfemport org.apache.spark.SparkConfimport org.apache.spark.mllib.linalg._
import org.apache.spark.mllib.regression.LabeledPoint
import org.apache.spark.mllib.stat.{ MultivariateStatisticalSummary, Statistics }
import scala.collection.mutable.ListBuffer
```

Statistical: Read

```
val train = sc.textFile(pathTrain).map { line =>
           val array = line.split(",")
           var arrayDouble = array.map(f => f.toDouble)
           val featureVector = Vectors.dense(arrayDouble.init)
           val label = arrayDouble.last
           LabeledPoint(label, featureVector)
}.persist
val test = sc.textFile(pathTest).map { line =>
           val array = line.split(",")
           var arrayDouble = array.map(f => f.toDouble)
           val featureVector = Vectors.dense(arrayDouble.init)
           val label = arrayDouble.last
           LabeledPoint(label, featureVector)
}.persist
```

Statistical: StatisticalSummary Spa

val observationsTrain = train.map(_.features)
val observationsTest = test.map(_.features)

val summaryTrain: MultivariateStatisticalSummary = Statistics.colStats(observationsTrain)
val summaryTest: MultivariateStatisticalSummary = Statistics.colStats(observationsTest)

summaryTrain.max is a dense vector containing the max value for each column

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Statistical: Result and write

```
var outputString = new ListBuffer[String]
outputString += "*****TRAIN*****\n\n"
outputString += "@Max (0) --> " + summaryTrain.max(0) + "\n"
outputString += "@Min (0) --> " + summaryTrain.min(0) + "\n"
outputString += "@Mean (0) --> " + summaryTrain.mean(0) + "\n"
outputString += "@Variance (0) --> " + summaryTrain.variance(0) + "\n"
outputString += "@NumNonZeros (0) --> " + summaryTrain.numNonzeros(0) + "\n"
outputString += "\n\n*****TEST*****\n\n"
outputString += "@Max (0) --> " + summaryTest.max(0) + "\n"
outputString += "@Min (0) --> " + summaryTest.min(0) + "\n"
outputString += "@Mean (0) --> " + summaryTest.mean(0) + "\n"
outputString += "@Variance (0) --> " + summaryTest.variance(0) + "\n"
outputString += "@NumNonZeros (0) --> " + summaryTest.numNonzeros(0) + "\n"
```

Statistical: StatisticalSummary Span

val predictionsTxt = sc.parallelize(outputString, 1)

predictionsTxt.saveAsTextFile(pathOutput)

Number of partition = 1 to obtain only 1 part SaveAsTextFile for readable stile

Statistical: compile.sh & clean.sh

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Build the .jar file

mvn package -Dmaven.test.skip=true

Remove old compilation

mvn clean

Run the example with

./clean.sh; ./compile.sh; ./run.sh



Statistical: run.sh

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compile.sh



pom.xml





/opt/spark-2.2.0-bin-hadoop2.7/bin/spark-submit

--master local[*]

--class org.apache.spark.run.runBasicStatistics

./target/BasicStatistics-1.0.jar

home/spark/datasets/susy-10k-tra.data

/home/spark/datasets/susy-10k-tst.data

./outputBasicStatistics



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Next >

Spark-packages

https://spark-packages.org/

SparkPackages Feedback Register a package Login Find a package

A community index of third-party packages for Apache Spark.

Showing packages 1 - 50 out of 394

All Core Sources (14)

Data Machine Streaming Graph PySpark Applications Deployment Examples Tools Learning (79) (54) (18) (17) (14) (12) (24) (30)

- Large number of "utils" implementations
- Machine learning algorithms
- Many other third parties algorithm for spark

Spark-packages



https://spark-packages.org/package/h2oai/sparkling-water

sparkling-water (homepage)

Sparkling Water provides H2O algorithms inside Spark cluster

Sparkling Water connects worlds of H2O and Spark. It enables launching H2O on top of Spark and using H2O capabilities including various ML algorithms, graphical user interface, or R integration.

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Michal Malohlava • 3 years ago

More about Sparkling Water is available at h2o blog: http://h2o.ai/blog

Example code is part of GitHub repository: https://github.com/h2oai/sp...

Spark-packages

https://spark-packages.org/package/lensacom/sparkit-learn

sparkit-learn (homepage)

PySpark + Scikit-learn = Sparkit-learn

@lensacom / *** (\$\(\begin{align*}2\))

Sparkit-learn aims to provide scikit-learn functionality and API on PySpark. The main goal of the library is to create an API that stays close to sklearn's.

The driving principle was to "Think locally, execute distributively." To accommodate this concept, the basic data block is always an array or a (sparse) matrix and the operations are executed on block level.

Spark-packages



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PySpark + Scikit-learn = Sparkit-learn

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How to

This package doesn't have any releases published in the Spark Packages repo, or with maven coordinates supplied. You may have to build this package from source, or it may simply be a script. To use this Spark Package, please follow the instructions in the README.

Spark-packages



https://spark-packages.org/package/Azure/mmlspark

mmlspark (homepage)

Microsoft Machine Learning for Apache Spark

@Azure / *********** (23)

MMLSpark provides a number of deep learning and data science tools for Apache Spark, including seamless integration of Spark Machine Learning pipelines with Microsoft Cognitive Toolkit (CNTK) and OpenCV, enabling you to quickly create powerful, highly-scalable predictive and analytical models for large image and text datasets.

Tags

3 ml 3 machine learning 3 Microsoft 3 Azure 3 pyspark 3 cntk 3 Microsoft Machine Learning 2 tools

How to [+]

Include this package in your Spark Applications using:

spark-shell, pyspark, or spark-submit

> \$SPARK_HOME/bin/spark-shell --packages Azure:mmlspark:0.11

\$\$PARK_HOME/bin/spark-shell -packages Azure:mmlspark:0.11

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Mllib: Spark-packages

https://spark-packages.org/package/JMailloH/kNN_IS

kNN_IS (homepage)



kNN-IS: An Iterative Spark-based design of the k-Nearest Neighbors classifier for big data.

@JMailloH / ****** (**1**4)

This is an open-source Spark package about an exact k-nearest neighbors classification based on Apache Spark. We take advantage of its in-memory operations to simultaneously classify big amounts of unseen cases against a big training dataset. The map phase computes the k-nearest neighbors in different splits of the training data. Afterwards, multiple reducers process the definitive neighbors from the list obtained in the map phase. The key point of this proposal lies on the management of the test set, maintaining it in memory when it is possible. Otherwise, this is split into a minimum number of pieces, applying a MapReduce per chunk, using the caching skills of Spark to reuse the previously partitioned training set.

>\$: /opt/spark-2.2.0-bin-hadoop2.7/bin/spark-shell --packages JMailloH:kNN_IS:3.0

Mllib: Script with kNN-IS

https://spark-packages.org/package/JMailloH/kNN IS







- clean.sh
- compile.sh
- kNN_IS.scala
- pom.xml
- run.sh









kNN-IS: import

```
import org.apache.spark.SparkContext
import org.apache.spark.SparkConf
import org.apache.spark.rdd._
import org.apache.spark.mllib.linalg.Vectors
import org.apache.spark.mllib.regression.LabeledPoint
import org.apache.spark.mllib.evaluation.MulticlassMetrics
import org.apache.spark.mllib.evaluation.BinaryClassificationMetrics
import org.apache.spark.mllib.classification.kNN_IS.kNN_IS
import utils.keel.KeelParser
import scala.collection.mutable.ListBuffer
```

kNN-IS: Read



val converter = new KeelParser(sc, "/home/hadoop/datasets/susy.header")

val train = sc.textFile("/home/hadoop/datasets/susy-10k-tra.data", 10).
map(line => converter.parserToLabeledPoint(line)).persist

val test = sc.textFile("/home/hadoop/datasets/susy-10k-tst.data", 10).
map(line => converter.parserToLabeledPoint(line)).persist

kNN-IS: Param.

```
val k = 5
val dist = 2 //euclidean
val numClass = converter.getNumClassFromHeader()
val numFeatures = converter.getNumFeaturesFromHeader()
val numPartitionMap = 10
val numReduces = 2
val numiterations = 1
val maxWeight = 5
```

val knn = kNN_IS.setup(train, test, k, dist, numClass, numFeatures, numPartitionMap, numReduces, numIterations, maxWeight)

kNN-IS: Classify

```
val predictions = knn.predict(sc)
val metrics = new MulticlassMetrics(predictions)
val precision = metrics.precision
val cm = metrics.confusionMatrix
val binaryMetrics = new BinaryClassificationMetrics(predictions)
val AUC = binaryMetrics.areaUnderROC
```



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