



**EDUCACIÓN**  
SECRETARÍA DE EDUCACIÓN PÚBLICA



TECNOLÓGICO  
NACIONAL DE MÉXICO®

TECNOLÓGICO NACIONAL DE MÉXICO

INSTITUTO TECNOLÓGICO DE TIJUANA

SUBDIRECCIÓN ACADÉMICA

DEPARTAMENTO DE SISTEMAS Y COMPUTACIÓN

NOMBRE DE LOS ALUMNOS:

GALAVIZ LONA OSCAR EDUARDO (N.CONTROL: 17212993)

MARQUEZ MILLAN SEASHELL VANESSA (N.CONTROL: )

Carrera: Ingeniería Informática

Semestre: 9no

MATERIA: Datos Masivos

PROFESOR: JOSE CHRISTIAN ROMERO HERNANDEZ

Practica 3

Unidad 2

## Development

we import the libraries that practice requires

```
import org.apache.spark.ml.Pipeline
import org.apache.spark.ml.classification.{RandomForestClassificationModel,
RandomForestClassifier}
import org.apache.spark.ml.evaluation.MulticlassClassificationEvaluator
import org.apache.spark.ml.feature.{IndexToString, StringIndexer, VectorIndexer}
```

here we load the data from a text file to be the dataframe

```
val data = spark.read.format("libsvm").load("sample.txt")

// Index labels, adding metadata to the label column.
// Fit on whole dataset to include all labels in index.
val labelIndexer = new
StringIndexer().setInputCol("label").setOutputCol("indexedLabel").fit(data)
```



```
usuario@ubuntu-20: ~
val labelIndexer = new StringIndexer().setInputCol("label").setOutputCol("indexedLabel").fit(data)

scala> val data = spark.read.format("libsvm").load("sample.txt")
21/11/06 04:23:37 WARN LibSVMFileFormat: 'numFeatures' option not specified
, determining the number of features by going through the input. If you know
the number in advance, please specify it via 'numFeatures' option to avoid
the extra scan.
[Stage 0:> (0 + 2)

data: org.apache.spark.sql.DataFrame = [label: double, features: vector]

scala> val labelIndexer = new StringIndexer().setInputCol("label").setOutputCol("indexedLabel").fit(data)
[Stage 1:> (0 + 1)

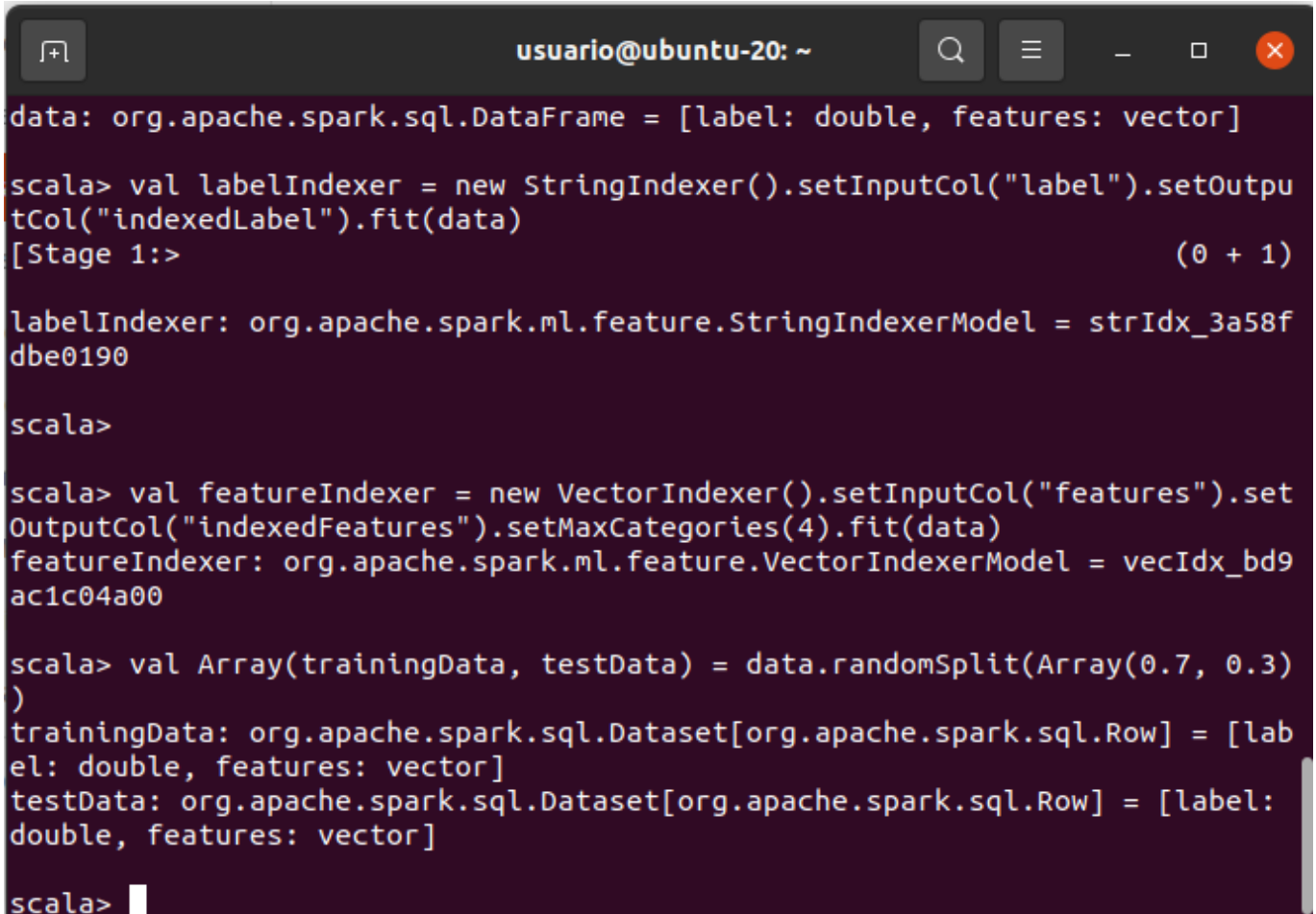
labelIndexer: org.apache.spark.ml.feature.StringIndexerModel = strIdx_3a58f
dbe0190

scala>
scala>
```

```
// Automatically identify categorical features, and index them.
// Set maxCategories so features with > 4 distinct values are treated as
```

```
continuous.
val featureIndexer = new
VectorIndexer().setInputCol("features").setOutputCol("indexedFeatures").setMaxCate
gories(4).fit(data)

// Split the data into training and test sets (30% held out for testing).
val Array(trainingData, testData) = data.randomSplit(Array(0.7, 0.3))
```



```
usuario@ubuntu-20: ~
data: org.apache.spark.sql.DataFrame = [label: double, features: vector]
scala> val labelIndexer = new StringIndexer().setInputCol("label").setOutput
Col("indexedLabel").fit(data)
[Stage 1:> (0 + 1)
labelIndexer: org.apache.spark.ml.feature.StringIndexerModel = strIdx_3a58f
dbe0190
scala>
scala> val featureIndexer = new VectorIndexer().setInputCol("features").set
OutputCol("indexedFeatures").setMaxCategories(4).fit(data)
featureIndexer: org.apache.spark.ml.feature.VectorIndexerModel = vecIdx_bd9
ac1c04a00
scala> val Array(trainingData, testData) = data.randomSplit(Array(0.7, 0.3)
)
trainingData: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [lab
el: double, features: vector]
testData: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [label:
double, features: vector]
scala>
```

Train a RandomForest model.

```
val rf = new
RandomForestClassifier().setLabelCol("indexedLabel").setFeaturesCol("indexedFeatur
es").setNumTrees(10)
```

```
scala> val rf = new RandomForestClassifier().setLabelCol("indexedLabel").se
tFeaturesCol("indexedFeatures").setNumTrees(10)
rf: org.apache.spark.ml.classification.RandomForestClassifier = rfc_910c4bf
0694d
```

Convert indexed labels back to original labels.

```
val labelConverter = new  
IndexToString().setInputCol("prediction").setOutputCol("predictedLabel").setLabels  
(labelIndexer.labels)
```

```
scala> val labelConverter = new IndexToString().setInputCol("prediction").s  
etOutputCol("predictedLabel").setLabels(labelIndexer.labels)  
labelConverter: org.apache.spark.ml.feature.IndexToString = idxToStr_688599  
962389
```

Chain indexers and forest in a Pipeline.

```
val pipeline = new Pipeline().setStages(Array(labelIndexer, featureIndexer, rf,  
labelConverter))
```

```
scala> val pipeline = new Pipeline().setStages(Array(labelIndexer, featureI  
ndexer, rf, labelConverter))  
pipeline: org.apache.spark.ml.Pipeline = pipeline_137c29a381c3
```

Train model. This also runs the indexers.

```
val model = pipeline.fit(trainingData)
```

```
scala> val model = pipeline.fit(trainingData)  
[Stage 5:> (0 + 1)  
[Stage 8:> (0 + 1)  
[Stage 9:> (0 + 1)  
model: org.apache.spark.ml.PipelineModel = pipeline_137c29a381c3
```

Make predictions.

```
val predictions = model.transform(testData)
```

```
scala> val predictions = model.transform(testData)  
predictions: org.apache.spark.sql.DataFrame = [label: double, features: vec  
tor ... 6 more fields]
```

Select example rows to display.

```
predictions.select("predictedLabel", "label", "features").show(5)
```



Select (prediction, true label) and compute test error.

```
val evaluator = new  
MulticlassClassificationEvaluator().setLabelCol("indexedLabel").setPredictionCol("  
prediction").setMetricName("accuracy")  
val accuracy = evaluator.evaluate(predictions)  
println(s"Test Error = ${(1.0 - accuracy)}")
```

```
scala> val evaluator = new MulticlassClassificationEvaluator().setLabelCol(  
"indexedLabel").setPredictionCol("prediction").setMetricName("accuracy")  
evaluator: org.apache.spark.ml.evaluation.MulticlassClassificationEvaluator  
= mcEval_13cf983d00f7  
  
scala> val accuracy = evaluator.evaluate(predictions)  
accuracy: Double = 1.0  
  
scala> println(s"Test Error = ${(1.0 - accuracy)}")  
Test Error = 0.0
```

```
val rfModel = model.stages(2).asInstanceOf[RandomForestClassificationModel]  
println(s"Learned classification forest model:\n ${rfModel.toDebugString}")
```



```
usuario@ubuntu-20: ~  
If (feature 512 <= 1.5)  
  If (feature 317 <= 164.5)  
    Predict: 0.0  
  Else (feature 317 > 164.5)  
    If (feature 296 <= 1.5)  
      Predict: 1.0  
    Else (feature 296 > 1.5)  
      Predict: 0.0  
  Else (feature 512 > 1.5)  
    Predict: 1.0  
Tree 8 (weight 1.0):  
  If (feature 462 <= 63.0)  
    If (feature 324 <= 251.5)  
      Predict: 1.0  
    Else (feature 324 > 251.5)  
      Predict: 0.0  
  Else (feature 462 > 63.0)  
    Predict: 0.0  
Tree 9 (weight 1.0):  
  If (feature 385 <= 4.0)  
    If (feature 545 <= 9.5)  
      If (feature 490 <= 15.5)  
        Predict: 1.0  
      Else (feature 490 > 15.5)  
        Predict: 0.0  
    Else (feature 545 > 9.5)  
      Predict: 0.0  
  Else (feature 385 > 4.0)  
    Predict: 1.0  
scala> 
```