





# Tecnológico nacional de México Instituto tecnológico de Tijuana

Subdirección Académica Departamento de sistemas y computación

SEMESTRE: Febrero - agosto 2022

CARRERA: Ingeniería en Sistemas Computacionales

> MATERIA: Datos masivos

NOMBRE TRABAJO:
Practice 3 Random Forest Classifier

UNIDAD: 2

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# // Import libraries

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}

# // Load and parse the data file, converting it to a DataFrame.

val data =

spark.read.format("libsvm").load("C:\Spark\spark-2.4.8-bin-hadoop2.7\data\mll ib\sample\_libsvm\_data.txt)

scala> val data = spark.read.format("libsvm").load("C:/Spark/spark-2.4.8-bin-hadoop2.7/data/mllib/sample\_libsvm\_data.txt")
22/85/04 21:52:44 WARN LibsVWF3leFormat: 'numFeatures' option not specified, determining the number of features by going though the input. If you know the number in ad
ance, please specify it via 'numFeatures' option to avoid the extra scan.
data: org.apache.spark.sql.DataFrame = [label: double, features: vector]

// 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)
- // 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")
- .setMaxCategories(4)

scala> val featureIndexer = new VectorIndexer() .setInputCol("features") .setOutputCol("indexedFeatures") .setMaxCategories(4) .fit(data) featureIndexer : new anache .snak: mal feature | vectorIndexerModal = vector dexed

# // Split the data into training and test sets (30% held out for testing).

val Array(trainingData, testData) = data.randomSplit(Array(0.7, 0.3))

```
scala> val Array(trainingData, testData) = data.randomSplit(Array(0.7, 0.3))
trainingData: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [label: double, features: vector]
testData: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [label: double, features: vector]
```

#### // Train a RandomForest model.

val rf = new RandomForestClassifier()

- .setLabelCol("indexedLabel")
- .setFeaturesCol("indexedFeatures")
- .setNumTrees(10)

```
scala> val rf = new RandomForestRegressor() .setLabelCol("label") .setFeaturesCol("indexedFeatures") rf: org.apache.spark.ml.regression.RandomForestRegressor = rfr_7dd95ee99dc0
```

# // Convert indexed labels back to original labels.

val labelConverter = new IndexToString()

- .setInputCol("prediction")
- .setOutputCol("predictedLabel")
- .setLabels(labelIndexer.labelsArray(0))

#### // Chain indexers and forest in a Pipeline.

val pipeline = new Pipeline()

.setStages(Array(labelIndexer, featureIndexer, rf, labelConverter))

```
scala> val pipeline = new Pipeline() .setStages(Array(featureIndexer, rf))
pipeline: org.apache.spark.ml.Pipeline = pipeline_6bad1b60306c
```

#### // Train model. This also runs the indexers.

val model = pipeline.fit(trainingData)

```
scala> val model = pipeline.fit(trainingData)
model: org.apache.spark.ml.PipelineModel = pipeline_6bad1b60306c
```

# // Make predictions.

val predictions = model.transform(testData)

```
scala> val predictions = model.transform(testData)
predictions: org.apache.spark.sql.DataFrame = [label: double, features: vector ... 2 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)}")

#### val rfModel =

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

```
scala> val evaluator = new RegressionEvaluator() .setLabelCol("label") .setPredictionCol("prediction") .setMetricName("rmse")
evaluator: org.apache.spark.ml.evaluation.RegressionEvaluator = regEval_98db33ea9ecd
scala> println(s"Root Mean Squared Error (RMSE) on test data = $rmse")
```

```
scala> println(s"Root Mean Squared Error (RMSE) on test data = $rmse")
Root Mean Squared Error (RMSE) on test data = 0.16422453217986943
scala> val rfModel = model.stages(1).asInstanceOf[RandomForestRegressionModel]
rfModel: org.apache.spark.ml.regression.RandomForestRegressionModel = RandomForestRegressionModel (uid=rfr_7dd95ee99dc0) with 20 trees
```

```
scalar val rfModel = model.stages(1).asInstanceOf[RandomForestRegressionModel]

rfModel: org.apache.spark.ml.regression.RandomForestRegressionModel = RandomForestRegressionModel (uid=rfr_7dd95ee99dc0) with 20 trees

scalar println(s*Learned regression forest model:\n $(rfModel.toDebugString)*)

learned regression forest model:

Ince (weight 1.0):

If (feature 433 < 52.5)

Predict: 0.0

Else (feature 433 > 52.5)

Predict: 1.0

Tree 1 (weight 1.0):

If (feature 490 < 44.5)

Predict: 0.0

Else (feature 490 > 44.5)

Predict: 1.0

Tree 2 (weight 1.0):

If (feature 490 < 40.5)

Predict: 0.0

Else (feature 490 > 40.5)

Predict: 0.0

Else (feature 490 > 40.5)

Predict: 0.0

Else (feature 490 > 40.5)

Predict: 0.0

If (feature 490 > 40.5)

Predict: 0.0

Else (feature 490 > 40.5)

Predict: 0.0

If (feature 490 > 40.5)

Predict: 0.0

Else (feature 490 > 40.5)

Else (feature 490 > 40.5)
```