## SparkSession and SparkContext

```
In [1]: from pyspark.sql import SparkSession
        spark = SparkSession.builder.appName("classification").getOrCreate()
        sc = spark.sparkContext
```

#### **Schema**

```
In [2]: from pyspark.sql.types import StructType, StructField, IntegerType, Doub
        leType
In [3]: colNames = ["Elevation", "Aspect", "Slope",
        "Horizontal_Distance_To_Hydrology", "Vertical_Distance_To_Hydrology",
        "Horizontal_Distance_To_Roadways",
        "Hillshade_9am", "Hillshade_Noon", "Hillshade_3pm",
        "Horizontal Distance To Fire Points"]
        for i in range(4):
            colNames += ["Wilderness Area "+str(i),]
        for i in range(40):
            colNames += ["Soil_Type_"+str(i),]
        colNames += ["Cover Type",]
In [4]: schema = StructType()
        for name in colNames:
            if name == "Cover Type":
                schema.add(StructField(name, DoubleType(), True))
                schema.add(StructField(name, IntegerType(), True))
```

## Read CSV with the prepared schema

```
In [5]: data = spark.read.csv('../Dropbox/pj ss/covtype/covtype.data', header=Fa
        lse, schema=schema)
```

In [6]: data.printSchema()

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```
3 Classification
root
 -- Elevation: integer (nullable = true)
 -- Aspect: integer (nullable = true)
 -- Slope: integer (nullable = true)
  -- Horizontal Distance To Hydrology: integer (nullable = true)
 -- Vertical Distance To Hydrology: integer (nullable = true)
  -- Horizontal Distance To Roadways: integer (nullable = true)
  -- Hillshade 9am: integer (nullable = true)
 -- Hillshade_Noon: integer (nullable = true)
  -- Hillshade_3pm: integer (nullable = true)
 -- Horizontal Distance To Fire Points: integer (nullable = true)
  -- Wilderness_Area_0: integer (nullable = true)
  -- Wilderness Area 1: integer (nullable = true)
  -- Wilderness Area 2: integer (nullable = true)
  -- Wilderness Area 3: integer (nullable = true)
 -- Soil_Type_0: integer (nullable = true)
  -- Soil Type 1: integer (nullable = true)
  -- Soil Type 2: integer (nullable = true)
  -- Soil Type 3: integer (nullable = true)
  -- Soil Type 4: integer (nullable = true)
  -- Soil Type 5: integer (nullable = true)
  -- Soil_Type_6: integer (nullable = true)
  -- Soil Type 7: integer (nullable = true)
  -- Soil Type 8: integer (nullable = true)
  -- Soil Type 9: integer (nullable = true)
  -- Soil Type 10: integer (nullable = true)
  -- Soil Type 11: integer (nullable = true)
 -- Soil Type 12: integer (nullable = true)
  -- Soil Type 13: integer (nullable = true)
  -- Soil Type 14: integer (nullable = true)
  -- Soil Type 15: integer (nullable = true)
  -- Soil Type 16: integer (nullable = true)
  -- Soil Type 17: integer (nullable = true)
  -- Soil Type 18: integer (nullable = true)
  -- Soil Type 19: integer (nullable = true)
  -- Soil Type 20: integer (nullable = true)
  -- Soil Type 21: integer (nullable = true)
  -- Soil Type 22: integer (nullable = true)
  -- Soil Type 23: integer (nullable = true)
  -- Soil Type 24: integer (nullable = true)
  -- Soil_Type_25: integer (nullable = true)
  -- Soil Type 26: integer (nullable = true)
  -- Soil Type 27: integer (nullable = true)
  -- Soil Type 28: integer (nullable = true)
  -- Soil Type 29: integer (nullable = true)
  -- Soil Type 30: integer (nullable = true)
  -- Soil Type 31: integer (nullable = true)
  -- Soil Type 32: integer (nullable = true)
  -- Soil Type 33: integer (nullable = true)
  -- Soil Type 34: integer (nullable = true)
  -- Soil Type 35: integer (nullable = true)
 -- Soil Type 36: integer (nullable = true)
```

-- Soil Type 37: integer (nullable = true) -- Soil Type 38: integer (nullable = true) -- Soil Type 39: integer (nullable = true) -- Cover Type: double (nullable = true)

```
In [7]: data.take(1)
```

#### **Assembler**

```
In [8]: from pyspark.ml.linalg import Vectors
from pyspark.ml.feature import VectorAssembler
```

```
In [9]: (trainData, testData) = data.randomSplit([0.9, 0.1])
```

```
In [10]: trainData
```

Out[10]: DataFrame[Elevation: int, Aspect: int, Slope: int, Horizontal Distance To Hydrology: int, Vertical Distance To Hydrology: int, Horizontal Dist ance To Roadways: int, Hillshade 9am: int, Hillshade Noon: int, Hillsha de 3pm: int, Horizontal Distance To Fire Points: int, Wilderness Area 0: int, Wilderness Area 1: int, Wilderness Area 2: int, Wilderness Area \_3: int, Soil\_Type\_0: int, Soil\_Type\_1: int, Soil\_Type\_2: int, Soil\_Typ e 3: int, Soil Type 4: int, Soil Type 5: int, Soil Type 6: int, Soil Ty pe 7: int, Soil Type 8: int, Soil Type 9: int, Soil Type 10: int, Soil Type 11: int, Soil Type 12: int, Soil Type 13: int, Soil Type 14: int, Soil Type 15: int, Soil Type 16: int, Soil Type 17: int, Soil Type 18: int, Soil Type 19: int, Soil Type 20: int, Soil Type 21: int, Soil Type \_22: int, Soil\_Type\_23: int, Soil\_Type\_24: int, Soil\_Type\_25: int, Soil Type 26: int, Soil Type 27: int, Soil Type 28: int, Soil Type 29: int, Soil Type 30: int, Soil Type 31: int, Soil Type 32: int, Soil Type 33: int, Soil Type 34: int, Soil Type 35: int, Soil Type 36: int, Soil Type 37: int, Soil Type 38: int, Soil Type 39: int, Cover Type: double]

```
In [11]: inputCols = trainData.drop('Cover_Type').columns
```

```
In [12]:
          inputCols
Out[12]: ['Elevation',
           'Aspect',
           'Slope',
           'Horizontal_Distance_To_Hydrology',
           'Vertical Distance To Hydrology',
           'Horizontal_Distance_To_Roadways',
           'Hillshade_9am',
           'Hillshade Noon',
           'Hillshade 3pm',
           'Horizontal_Distance_To_Fire_Points',
           'Wilderness Area 0',
           'Wilderness_Area_1',
           'Wilderness_Area_2',
           'Wilderness_Area_3',
           'Soil Type 0',
           'Soil_Type_1',
           'Soil Type 2',
           'Soil_Type_3',
           'Soil_Type_4',
           'Soil_Type_5',
           'Soil_Type_6',
           'Soil_Type_7',
           'Soil_Type_8',
           'Soil_Type_9',
           'Soil_Type_10',
           'Soil Type 11',
           'Soil Type 12',
           'Soil Type 13',
           'Soil_Type_14',
           'Soil Type 15',
           'Soil_Type_16',
           'Soil Type 17',
           'Soil Type 18',
           'Soil Type 19',
           'Soil_Type_20',
           'Soil Type 21'
           'Soil Type 22',
           'Soil_Type_23',
           'Soil Type 24',
           'Soil Type 25',
           'Soil Type 26',
           'Soil Type 27',
           'Soil_Type_28',
           'Soil Type 29',
           'Soil Type 30',
           'Soil Type 31',
           'Soil_Type_32',
           'Soil Type 33',
           'Soil Type 34',
           'Soil Type 35',
           'Soil Type 36',
           'Soil Type 37',
           'Soil Type 38',
```

'Soil\_Type\_39']

```
In [16]: | assembler = VectorAssembler(
             inputCols=inputCols,
             outputCol="featureVector")
In [17]:
         assembler
Out[17]: VectorAssembler_4bb3bb34a9cf0b6fd419
In [18]:
         assembledTrainData = assembler.transform(trainData)
In [19]: assembledTrainData
Out[19]: DataFrame[Elevation: int, Aspect: int, Slope: int, Horizontal_Distance_
         To Hydrology: int, Vertical Distance To Hydrology: int, Horizontal Dist
         ance To Roadways: int, Hillshade 9am: int, Hillshade Noon: int, Hillsha
         de 3pm: int, Horizontal Distance To Fire Points: int, Wilderness Area
         0: int, Wilderness Area 1: int, Wilderness Area 2: int, Wilderness Area
         3: int, Soil Type 0: int, Soil Type 1: int, Soil Type 2: int, Soil Typ
         e 3: int, Soil Type 4: int, Soil Type 5: int, Soil Type 6: int, Soil Ty
         pe_7: int, Soil_Type_8: int, Soil_Type_9: int, Soil_Type_10: int, Soil_
         Type_11: int, Soil_Type_12: int, Soil_Type_13: int, Soil_Type_14: int,
         Soil Type 15: int, Soil Type 16: int, Soil Type 17: int, Soil Type 18:
         int, Soil Type 19: int, Soil Type 20: int, Soil Type 21: int, Soil Type
         22: int, Soil Type 23: int, Soil Type 24: int, Soil Type 25: int, Soil
          Type 26: int, Soil Type 27: int, Soil Type 28: int, Soil Type 29: int,
         Soil Type 30: int, Soil Type 31: int, Soil Type 32: int, Soil Type 33:
         int, Soil Type 34: int, Soil Type 35: int, Soil Type 36: int, Soil Type
         37: int, Soil Type 38: int, Soil Type 39: int, Cover Type: double, fea
         tureVector: vector]
In [20]: assembledTrainData.select('featureVector').show(3, truncate=False)
         |featureVector
         (54,[0,1,2,3,4,5,6,7,8,9,13,15],[1863.0,37.0,17.0,120.0,18.0,90.0,217.
         0,202.0,115.0,769.0,1.0,1.0])
         |(54,[0,1,2,3,4,5,6,7,8,9,13,18],[1879.0,28.0,19.0,30.0,12.0,95.0,209.
         0,196.0,117.0,778.0,1.0,1.0])
         (54,[0,1,2,3,4,5,6,7,8,9,13,15],[1888.0,33.0,22.0,150.0,46.0,108.0,20
         9.0,185.0,103.0,735.0,1.0,1.0])
         only showing top 3 rows
```

#### **Decision Tree**

In [21]: from pyspark.ml.classification import DecisionTreeClassifier

In [25]: print(model.toDebugString)

DecisionTreeClassificationModel (uid=DecisionTreeClassifier\_44ac953bdb2 a9ad6e065) of depth 5 with 63 nodes If (feature 0 <= 3034.5)</pre> If (feature 0 <= 2479.5)</pre> If (feature 3 <= 15.0)</pre> If (feature 13 <= 0.5) If (feature 17 <= 0.5)</pre> Predict: 6.0 Else (feature 17 > 0.5) Predict: 6.0 Else (feature 13 > 0.5) If (feature 23 <= 0.5) Predict: 4.0 Else (feature 23 > 0.5) Predict: 3.0 Else (feature 3 > 15.0) If (feature 16 <= 0.5)</pre> If (feature 9 <= 576.5) Predict: 3.0 Else (feature 9 > 576.5) Predict: 3.0 Else (feature 16 > 0.5) If (feature 9 <= 1295.0) Predict: 3.0 Else (feature 9 > 1295.0) Predict: 4.0 Else (feature 0 > 2479.5) If (feature 17 <= 0.5) If (feature 15 <= 0.5)</pre> If (feature  $0 \le 2942.5$ ) Predict: 2.0 Else (feature 0 > 2942.5) Predict: 2.0 Else (feature 15 > 0.5) If (feature 9 <= 1380.5)</pre> Predict: 3.0 Else (feature 9 > 1380.5) Predict: 3.0 Else (feature 17 > 0.5) If (feature 0 <= 2692.5)</pre> If (feature 0 <= 2635.5)</pre> Predict: 3.0 Else (feature 0 > 2635.5) Predict: 3.0 Else (feature 0 > 2692.5) If (feature 5 <= 1200.5) Predict: 5.0 Else (feature 5 > 1200.5) Predict: 2.0 Else (feature 0 > 3034.5) If (feature 0 <= 3309.5)</pre> If (feature 7 <= 238.5)</pre> If (feature 0 <= 3101.5)</pre> If (feature 3 <= 191.0)</pre> Predict: 1.0 Else (feature 3 > 191.0)

Predict: 2.0

```
Else (feature 0 > 3101.5)
   If (feature 5 <= 998.0)
    Predict: 1.0
   Else (feature 5 > 998.0)
    Predict: 1.0
 Else (feature 7 > 238.5)
  If (feature 3 <= 333.0)</pre>
   If (feature 0 <= 3186.5)
    Predict: 1.0
   Else (feature 0 > 3186.5)
    Predict: 1.0
  Else (feature 3 > 333.0)
   If (feature 0 <= 3206.5)</pre>
    Predict: 2.0
   Else (feature 0 > 3206.5)
    Predict: 1.0
Else (feature 0 > 3309.5)
 If (feature 12 <= 0.5)</pre>
  If (feature 3 <= 290.0)
   If (feature 6 <= 207.5)
    Predict: 1.0
   Else (feature 6 > 207.5)
    Predict: 7.0
  Else (feature 3 > 290.0)
   If (feature 10 <= 0.5)</pre>
    Predict: 1.0
   Else (feature 10 > 0.5)
    Predict: 1.0
 Else (feature 12 > 0.5)
  If (feature 45 <= 0.5)
   If (feature 0 <= 3369.5)</pre>
    Predict: 7.0
   Else (feature 0 > 3369.5)
    Predict: 7.0
  Else (feature 45 > 0.5)
   If (feature 5 <= 998.0)
    Predict: 7.0
   Else (feature 5 > 998.0)
    Predict: 1.0
```

#### In [26]: print(model.featureImportances)

(54,[0,3,5,6,7,9,10,12,13,15,16,17,23,45],[0.8133742673793523,0.0309786,0477331104,0.014747043247755682,0.0023999635668687604,0.025523595174575,0.00646850931723158,0.0031668991984998784,0.011192823508553537,0.0027142714212239146,0.03035888049891708,0.0033263230701791355,0.03807455883308248,0.0009625115515200479,0.01671174845892951])

In [27]: | predictions = model.transform(assembledTrainData)

## **Evaluation**

## **Pipeline**

```
In [34]: from pyspark.ml import Pipeline
In [35]: inputCols = trainData.columns[:-1]
```

```
In [36]:
          inputCols
Out[36]: ['Elevation',
           'Aspect',
           'Slope',
           'Horizontal_Distance_To_Hydrology',
           'Vertical Distance To Hydrology',
           'Horizontal_Distance_To_Roadways',
           'Hillshade_9am',
           'Hillshade Noon',
           'Hillshade 3pm',
           'Horizontal_Distance_To_Fire_Points',
           'Wilderness Area 0',
           'Wilderness_Area_1',
           'Wilderness_Area_2',
           'Wilderness_Area_3',
           'Soil Type 0',
           'Soil_Type_1',
           'Soil Type 2',
           'Soil_Type_3',
           'Soil_Type_4',
           'Soil_Type_5',
           'Soil_Type_6',
           'Soil_Type_7',
           'Soil_Type_8',
           'Soil_Type_9',
           'Soil_Type_10',
           'Soil Type 11',
           'Soil Type 12',
           'Soil Type 13',
           'Soil_Type_14',
           'Soil Type 15',
           'Soil_Type_16',
           'Soil Type 17',
           'Soil Type 18',
           'Soil Type 19',
           'Soil_Type_20',
           'Soil Type 21'
           'Soil Type 22',
           'Soil_Type_23',
           'Soil Type 24',
           'Soil Type 25',
           'Soil Type 26',
           'Soil Type 27',
           'Soil_Type_28',
           'Soil Type 29',
           'Soil Type 30',
           'Soil Type 31',
           'Soil_Type_32',
           'Soil Type 33',
           'Soil Type 34',
           'Soil Type 35',
           'Soil Type 36',
           'Soil Type 37',
           'Soil Type 38',
```

'Soil\_Type\_39']

```
In [37]:
         assembler = VectorAssembler(inputCols=inputCols, outputCol="featureVecto
         r")
In [38]: classifier = DecisionTreeClassifier(labelCol="Cover Type",
                                              featuresCol="featureVector",
                                              predictionCol="prediction")
In [39]: classifier
Out[39]: DecisionTreeClassifier_41c58c49ca4ad38a5bc6
In [40]: pipeline = Pipeline(stages=[assembler, classifier])
In [41]: pipeline
Out[41]: Pipeline 4b6aad06f5c5ac75d01b
In [42]: from pyspark.ml.tuning import ParamGridBuilder
In [46]: paramGrid = ParamGridBuilder()\
              .addGrid(classifier, ["gini", "entropy"])\
              .addGrid(classifier, [1, 20])\
              .addGrid(classifier, [40, 300])\
              .addGrid(classifier, [0.0, 0.05])\
              .build()
In [47]: multiclassEval = MulticlassClassificationEvaluator(
             labelCol="Cover Type",
             predictionCol="prediction",
             metricName="accuracy")
In [48]: | multiclassEval.evaluate(predictions)
Out[48]: 0.7022924849787542
In [49]: from pyspark.ml.tuning import TrainValidationSplit
In [50]: validator = TrainValidationSplit(
             estimator=pipeline,
             estimatorParamMaps=paramGrid,
             evaluator=multiclassEval,
             trainRatio=0.9)
In [51]: validatorModel = validator.fit(trainData)
In [52]: bestModel = validatorModel.bestModel
In [53]: | bestModel
Out[53]: PipelineModel 44b3ad4e4f4671d6a89b
```

In [54]: bestModel.stages[-1].extractParamMap()

Out[54]: {Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='cach eNodeIds', doc='If false, the algorithm will pass trees to executors to match instances with nodes. If true, the algorithm will cache node IDs for each instance. Caching can speed up training of deeper trees.'): Fa lse,

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='chec kpointInterval', doc='set checkpoint interval (>= 1) or disable checkpo int (-1). E.g. 10 means that the cache will get checkpointed every 10 i terations. Note: this setting will be ignored if the checkpoint directo ry is not set in the SparkContext'): 10,

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='feat
uresCol', doc='features column name'): 'featureVector',

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='impu rity', doc='Criterion used for information gain calculation (case-insen sitive). Supported options: entropy, gini'): 'gini',

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='labe 1Col', doc='label column name'): 'Cover\_Type',

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='maxB ins', doc='Max number of bins for discretizing continuous features. Mu st be >=2 and >= number of categories for any categorical feature.'): 3 2,

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='maxD epth', doc='Maximum depth of the tree. (>= 0) E.g., depth 0 means 1 lea f node; depth 1 means 1 internal node + 2 leaf nodes.'): 5,

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='maxM emoryInMB', doc='Maximum memory in MB allocated to histogram aggregatio n.'): 256,

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='minI nfoGain', doc='Minimum information gain for a split to be considered at a tree node.'): 0.0,

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='minI nstancesPerNode', doc='Minimum number of instances each child must have after split. If a split causes the left or right child to have fewer t han minInstancesPerNode, the split will be discarded as invalid. Should be >= 1.'): 1,

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='pred
ictionCol', doc='prediction column name'): 'prediction',

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='prob abilityCol', doc='Column name for predicted class conditional probabilities. Note: Not all models output well-calibrated probability estimate s! These probabilities should be treated as confidences, not precise probabilities'): 'probability',

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='rawP redictionCol', doc='raw prediction (a.k.a. confidence) column name'): 'rawPrediction',

Param(parent='DecisionTreeClassifier\_41c58c49ca4ad38a5bc6', name='see
d', doc='random seed'): 7750783964360596788}

In [55]: paramsAndMetrics = validatorModel.validationMetrics

In [56]: paramsAndMetrics

Out[56]: [0.7012882447665056, 0.7012882447665056]

```
In [57]: multiclassEval.evaluate(bestModel.transform(testData))
Out[57]: 0.7040250662810316
```

# **Undoing Onehot Coding**

```
In [58]: wildernessCols = []
         for i in range(4):
             wildernessCols += ["Wilderness_Area_"+str(i),]
In [63]: wildernessAssembler = VectorAssembler(
             inputCols=wildernessCols,
             outputCol="wilderness")
In [64]:
         from pyspark.sql.functions import udf
         from pyspark.sql.types import ArrayType, DoubleType, StructType
In [65]:
         unhotudf = udf(lambda x: float(x.toArray().nonzero()[0]), DoubleType())
         withWilderness = wildernessAssembler.transform(data)
In [66]:
In [67]: withWilderness.take(1)
Out[67]: [Row(Elevation=2596, Aspect=51, Slope=3, Horizontal Distance To Hydrolo
         qy=258, Vertical Distance To Hydrology=0, Horizontal Distance To Roadwa
         ys=510, Hillshade 9am=221, Hillshade Noon=232, Hillshade 3pm=148, Horiz
         ontal Distance To Fire Points=6279, Wilderness Area 0=1, Wilderness Are
         a 1=0, Wilderness Area 2=0, Wilderness Area 3=0, Soil Type 0=0, Soil Ty
         pe_1=0, Soil_Type_2=0, Soil_Type_3=0, Soil_Type_4=0, Soil_Type_5=0, Soi
         1 Type 6=0, Soil Type 7=0, Soil Type 8=0, Soil Type 9=0, Soil Type 10=
         0, Soil Type 11=0, Soil Type 12=0, Soil Type 13=0, Soil Type 14=0, Soil
         Type 15=0, Soil Type 16=0, Soil Type 17=0, Soil Type 18=0, Soil Type 1
         9=0, Soil Type 20=0, Soil Type 21=0, Soil Type 22=0, Soil Type 23=0, So
         il Type 24=0, Soil Type 25=0, Soil Type 26=0, Soil Type 27=0, Soil Type
         28=1, Soil Type 29=0, Soil Type 30=0, Soil Type 31=0, Soil Type 32=0,
         Soil Type 33=0, Soil Type 34=0, Soil_Type_35=0, Soil_Type_36=0, Soil_Ty
         pe 37=0, Soil Type 38=0, Soil Type 39=0, Cover Type=5.0, wilderness=Spa
         rseVector(4, {0: 1.0}))]
In [68]: withWilderness = withWilderness\
             .drop(*wildernessCols)\
             .withColumn("wilderness", unhotudf(withWilderness['wilderness']))
```

```
In [69]: withWilderness.take(1)
Out[69]: [Row(Elevation=2596, Aspect=51, Slope=3, Horizontal Distance To Hydrolo
         gy=258, Vertical_Distance_To_Hydrology=0, Horizontal Distance To Roadwa
         ys=510, Hillshade 9am=221, Hillshade Noon=232, Hillshade 3pm=148, Horiz
         ontal_Distance_To_Fire_Points=6279, Soil_Type_0=0, Soil_Type_1=0, Soil_
         Type 2=0, Soil Type 3=0, Soil Type 4=0, Soil Type 5=0, Soil Type 6=0, S
         oil Type 7=0, Soil Type 8=0, Soil Type 9=0, Soil Type 10=0, Soil Type 1
         1=0, Soil_Type_12=0, Soil_Type_13=0, Soil_Type_14=0, Soil_Type_15=0, So
         il Type 16=0, Soil Type 17=0, Soil Type 18=0, Soil Type 19=0, Soil Type
         20=0, Soil_Type_21=0, Soil_Type_22=0, Soil_Type_23=0, Soil_Type_24=0,
         Soil_Type_25=0, Soil_Type_26=0, Soil_Type_27=0, Soil_Type_28=1, Soil_Ty
         pe 29=0, Soil Type 30=0, Soil Type 31=0, Soil Type 32=0, Soil Type 33=
         0, Soil Type 34=0, Soil Type 35=0, Soil Type 36=0, Soil Type 37=0, Soil
         Type 38=0, Soil Type 39=0, Cover Type=5.0, wilderness=0.0)]
In [70]: soilCols = []
         for i in range(40):
             soilCols += ["Soil_Type_"+str(i),]
In [71]: soilAssembler = VectorAssembler(
                 inputCols=soilCols,
                 outputCol="soil")
In [72]: withWilderness = soilAssembler.transform(withWilderness)
In [73]: unencodedData = withWilderness\
             .drop(*soilCols)\
              .withColumn("soil", unhotudf(withWilderness['soil']))
In [74]: unencodedData.take(2)
Out[74]: [Row(Elevation=2596, Aspect=51, Slope=3, Horizontal Distance To Hydrolo
         gy=258, Vertical Distance To Hydrology=0, Horizontal Distance To Roadwa
         ys=510, Hillshade 9am=221, Hillshade Noon=232, Hillshade 3pm=148, Horiz
         ontal Distance To Fire Points=6279, Cover Type=5.0, wilderness=0.0, soi
         1=28.0),
          Row(Elevation=2590, Aspect=56, Slope=2, Horizontal_Distance_To_Hydrolo
         gy=212, Vertical Distance To Hydrology=-6, Horizontal Distance To Roadw
         ays=390, Hillshade 9am=220, Hillshade Noon=235, Hillshade 3pm=151, Hori
         zontal Distance To Fire Points=6225, Cover Type=5.0, wilderness=0.0, so
         il=28.0)1
```

### **Decision Tree with Unencoded Data**

```
In [75]: (unencTrainData, unencTestData) = unencodedData.randomSplit([0.9, 0.1])
In [76]: from pyspark.ml.feature import VectorIndexer
In [77]: inputCols = unencTrainData.drop('Cover_Type').columns
```

```
In [78]:
         inputCols
Out[78]: ['Elevation',
           'Aspect',
           'Slope',
           'Horizontal_Distance_To_Hydrology',
           'Vertical Distance To Hydrology',
           'Horizontal_Distance_To_Roadways',
           'Hillshade_9am',
           'Hillshade Noon',
           'Hillshade_3pm',
           'Horizontal_Distance_To_Fire_Points',
           'wilderness',
           'soil']
In [79]: assembler = VectorAssembler(
              inputCols=inputCols,
              outputCol="featureVector")
In [80]:
         indexer = VectorIndexer(
             maxCategories=40,
              inputCol="featureVector",
              outputCol="indexedVector")
In [81]: classifier = DecisionTreeClassifier(
              seed=42,
              labelCol="Cover Type",
              featuresCol="indexedVector",
              predictionCol="prediction")
In [82]: pipeline = Pipeline(stages=[assembler, indexer, classifier])
```

## **Random Forrest Classifier**

In [92]: forestModel = bestModel.stages[-1]
 print(forestModel.extractParamMap())

{Param(parent='RandomForestClassifier 4fa2ae5257b23789060c', name='cach eNodeIds', doc='If false, the algorithm will pass trees to executors to match instances with nodes. If true, the algorithm will cache node IDs for each instance. Caching can speed up training of deeper trees.'): Fa lse, Param(parent='RandomForestClassifier\_4fa2ae5257b23789060c', name ='checkpointInterval', doc='set checkpoint interval (>= 1) or disable c heckpoint (-1). E.g. 10 means that the cache will get checkpointed ever y 10 iterations. Note: this setting will be ignored if the checkpoint d irectory is not set in the SparkContext'): 10, Param(parent='RandomFore stClassifier 4fa2ae5257b23789060c', name='featureSubsetStrategy', doc ='The number of features to consider for splits at each tree node. Supp orted options: auto, all, onethird, sqrt, log2, (0.0-1.0], [1-n].'): 'a uto', Param(parent='RandomForestClassifier\_4fa2ae5257b23789060c', name ='featuresCol', doc='features column name'): 'indexedVector', Param(par ent='RandomForestClassifier\_4fa2ae5257b23789060c', name='impurity', doc ='Criterion used for information gain calculation (case-insensitive). S upported options: entropy, gini'): 'gini', Param(parent='RandomForestCl assifier\_4fa2ae5257b23789060c', name='labelCol', doc='label column nam e'): 'Cover\_Type', Param(parent='RandomForestClassifier 4fa2ae5257b2378 9060c', name='maxBins', doc='Max number of bins for discretizing contin uous features. Must be >=2 and >= number of categories for any categor ical feature.'): 40, Param(parent='RandomForestClassifier\_4fa2ae5257b23 789060c', name='maxDepth', doc='Maximum depth of the tree. (>= 0) E.g., depth 0 means 1 leaf node; depth 1 means 1 internal node + 2 leaf node s.'): 5, Param(parent='RandomForestClassifier 4fa2ae5257b23789060c', na me='maxMemoryInMB', doc='Maximum memory in MB allocated to histogram ag gregation.'): 256, Param(parent='RandomForestClassifier 4fa2ae5257b2378 9060c', name='minInfoGain', doc='Minimum information gain for a split t o be considered at a tree node.'): 0.0, Param(parent='RandomForestClass ifier\_4fa2ae5257b23789060c', name='minInstancesPerNode', doc='Minimum n umber of instances each child must have after split. If a split causes the left or right child to have fewer than minInstancesPerNode, the spl it will be discarded as invalid. Should be >= 1.'): 1, Param(parent='Ra ndomForestClassifier\_4fa2ae5257b23789060c', name='numTrees', doc='Numbe r of trees to train (>= 1)'): 1, Param(parent='RandomForestClassifier 4 fa2ae5257b23789060c', name='predictionCol', doc='prediction column nam e'): 'prediction', Param(parent='RandomForestClassifier 4fa2ae5257b2378 9060c', name='probabilityCol', doc='Column name for predicted class con ditional probabilities. Note: Not all models output well-calibrated pro bability estimates! These probabilities should be treated as confidence s, not precise probabilities'): 'probability', Param(parent='RandomFore stClassifier\_4fa2ae5257b23789060c', name='rawPredictionCol', doc='raw p rediction (a.k.a. confidence) column name'): 'rawPrediction', Param(par ent='RandomForestClassifier 4fa2ae5257b23789060c', name='seed', doc='ra ndom seed'): 42, Param(parent='RandomForestClassifier 4fa2ae5257b237890 60c', name='subsamplingRate', doc='Fraction of the training data used f or learning each decision tree, in range (0, 1].'): 1.0}

In [93]: forestModel.getNumTrees

Out[93]: 1

```
In [94]: sorted(list(zip(inputCols, forestModel.featureImportances)), key=lambda
         x: x[1], reverse=True)
Out[94]: [('Elevation', 0.7822498415213323),
          ('soil', 0.11297264686002277),
          ('wilderness', 0.037105461899423466),
          ('Horizontal_Distance_To_Roadways', 0.03535837398448177),
          ('Horizontal_Distance_To_Fire_Points', 0.012732741797356662),
          ('Hillshade_Noon', 0.011420066167183551),
          ('Horizontal_Distance_To_Hydrology', 0.0050684199555167024),
          ('Vertical_Distance_To_Hydrology', 0.003092447814683044),
          ('Aspect', 0.0),
          ('Slope', 0.0),
          ('Hillshade_9am', 0.0),
          ('Hillshade_3pm', 0.0)]
In [95]: testAccuracy = multiclassEval.evaluate(bestModel.transform(unencTestData
In [96]: testAccuracy
Out[96]: 0.6989563173901023
```