In [3]:	<pre>from pyspark.sql import SparkSession spark=SparkSession.builder.appName("empApp").getOrCreate()</pre>
<pre>In [4]: Out[4]:</pre>	spark SparkSession - in-memory
	Spark UI Spark UI
	Version v3.2.1 Master local[*] AppName empApp
In [5]:	df=spark.read.option("header", "true").csv('D:\employees.csv',inferSchema=True) df.show() *
In [6]:	df.count()
Out[6]: In [7]:	<pre>df1=df.na.drop(how='any') df1</pre>
Out[7]: In [8]:	DataFrame[FirstName: string, Gender: string, StartDate: string, LastLoginTime: string, Salary: int, Bonus: double, SeniorManagement: boolean, Team: string] df1.count()
Out[8]:	<pre>df1.printSchema() root</pre>
In [10]:	<pre>df1.select('SeniorManagement').distinct().show() ++ SeniorManagement </pre>
	true
In [11]: In [12]:	<pre>from pyspark.ml.feature import VectorAssembler from pyspark.ml.feature import StringIndexer</pre> f=df1_withColumn(SoniorManagement df1[SoniorManagement] cast(string))
[12] :	<pre>f=df1.withColumn('SeniorManagement', df1['SeniorManagement'].cast('string')) f.printSchema() root FirstName: string (nullable = true) Gender: string (nullable = true)</pre>
	StartDate: string (nullable = true) LastLoginTime: string (nullable = true) LastLoginTime: string (nullable = true) Salary: integer (nullable = true) Bonus: double (nullable = true) SeniorManagement: string (nullable = true) Team: string (nullable = true)
In [49]:	<pre>va=VectorAssembler(inputCols=['Salary', 'Bonus'], outputCol='Input Features') indexer=StringIndexer(inputCol='SeniorManagement', outputCol='SMData') df2=indexer.fit(f).transform(f)</pre>
	<pre>df3=va.transform(df2) df3.show() ++++++</pre>
	Douglas Male 8/6/1993 12:42 PM 97308 6.945 true Marketing 1.0 [97308.0, 6.945] Dennis Male 4/18/1987 1:35 AM 115163 10:125 false Legal 0.0 [115163.0, 10:125] Julie Female 10/26/1997 3:19 PM 102508 12:637 true Legal 1.0 [102508.0, 12:637] True Legal 1.0 [102508.0, 12:637] True Legal 1.0 [102508.0, 12:637] True Human Resources 1.0 [112807.0, 17.492] Gary Male 1/27/2008 11:40 PM 109831 5.831 false Sales 0.0 [109831.0, 5.831] Kimberly Female 1/14/1999 7:13 AM 41426 14.543 true Finance 1.0 [41426.0, 14.543] Lillian Female 6/5/2016 6:09 AM 59414 1.256 false Product 0.0 [59414.0, 1.256] Jeremy Male 9/21/2010 5:56 AM 90370 7.369 false Human Resources 0.0 [90370.0, 7.369] Shawn Male 12/77/1986 7.45 PM 111737 6.414 false Product 0.0 [111737.0, 6.414] Diana Female 10/23/1981 10:27 AM 132940 19.082 false Client Services 0.0 [132940.0, 19.082] Donna Female 7/22/2010 3:48 AM 81014 1.894 false Product 0.0 [13014.0, 1.894] Matthew Male 9/5/1995 2:12 AM 100612 13.645 false Marketing 0.0 [100612.0, 13.645] John Male 7/17/1992 10:08 PM 79550 13.873 false Client Services 0.0 [97550.0, 13.873] Craig Male 2/27/2000 7:45 AM 37588 7.757 true Marketing 1.0 [37598.0, 7.757] Terry Male 1/26/2005 10:06 PM 79529 7.008 true Legal 1.0 [79529.0, 7.008] Christina Female 8/6/2002 1:19 PM 118780 9.096 true Engineering 1.0 [118780.0, 9.096] Jean Female 10/10/2004 12:56 PM 95734 19.096 false Client Services 0.0 [95734.0, 19.096] Theresa Female 10/10/2004 12:56 PM 95734 19.096 false Client Services 0.0 [95734.0, 19.096] Theresa Female 10/10/2004 12:56 PM 95734 19.096 false Client Services 0.0 [95734.0, 19.096] Theresa Female 10/10/2004 12:56 PM 95734 19.096
In [50]:	only showing top 20 rows finaldata=df3.select('Input Features','SMData')
In [51]:	finaldata.show() Input Features SMData 197308.0,6.945] 1.0 [1315163.0,10.125] 0.0 [1012508.0,12.637] 1.0 [1012807.0,17.492] 1.0 [109831.0,5.831] 0.0 [109831.0,5.831] 0.0 [41426.0,14.5431] 1.0 [59414.0,1.256] 0.0 [80370.0,7.369] 0.0 [80370.0,7.369] 0.0 [80370.0,7.369] 0.0 [80370.0,7.369] 0.0 [81014.0,1.894] 0.0 [101612.0,13.645] 0.0 [101612.0,13.645] 0.0 [79529.0,7.080] 1.0 [179529.0,7.080] 1.0 [181620.0,13.09.996] 1.0 [181673.0,13.464] 0.0 [1816780.0,9.996] 1.0 [1816780.0,9.996] 1.0 [1816780.0,9.996] 0.0 [1816780.0,6.18] 0.0 [1816780.0,9.996] 1.0 [1816780.0,9.996] 0.0 [1816780.0,6.18] 0.0 [1816780.0,9.996] 0.0 [1816780.0,6.18] 0.0 [1816780.0,9.996] 0.0 [1816780.0,6.18] 0.0 [1816780.0,6.18] 0.0 [1816780.0,6.18] 0.0 [1816780.0,6.18] 0.0 [1816780.0,9.996] 0.0 [1816780.0,6.18] 0.0 [1816780.0,6.18] 0.0 [1816780.0,9.996] 0.0 [1816780.0,6.18] 0.0 [1816780.0,9.996] 0.0 [1816780.0,6.18] 0.0 [1816780.0
In [52]:	<pre>from pyspark.ml.classification import DecisionTreeClassifier</pre>
In [53]: In [54]:	<pre>dtcmodel=DecisionTreeClassifier(labelCol='SMData', featuresCol='Input Features') model=dtcmodel.fit(train) model</pre>
Out[54]: In [55]:	DecisionTreeClassificationModel: uid=DecisionTreeClassifier_a5aa92e9b100, depth=5, numNodes=33, numClasses=2, numFeatures=2 prediction_res=model.transform(test) prediction_res.show()
In [57]:	Input Features SMData rawPrediction probability prediction
In [60]:	SMData rawPrediction
In [64]:	evaluator=MulticlassClassificationEvaluator(labelCol='SMData', predictionCol='prediction') accuracy=evaluator.evaluate(prediction_res) print("Accuracy of model ", accuracy) print("Error of model ", (1-accuracy)) Accuracy of model 0 5213648824123502
In []:	Accuracy of model 0.5213648824123502 Error of model 0.47863511758764976

In [2]:

import findspark
findspark.init()