```
from pyspark.sql import SparkSession
from pyspark.ml.feature import StringIndexer, VectorAssembler
from pyspark.ml.classification import LogisticRegression
from pyspark.ml.evaluation import BinaryClassificationEvaluator
spark = SparkSession.builder.getOrCreate()
suvDF = spark.read.csv("suv_data.csv",header = True, inferSchema = True)
suvDF.printSchema()
....
root
|-- User_ID: integer (nullable = true)
 |-- Gender: string (nullable = true)
|-- Age: integer (nullable = true)
 |-- EstimatedSalary: integer (nullable = true)
|-- Purchased: integer (nullable = true)
root
|-- User_ID: integer (nullable = true)
|-- Gender: string (nullable = true)
|-- Age: integer (nullable = true)
|-- EstimatedSalary: integer (nullable = true)
|-- Purchased: integer (nullable = true)
'\nroot\n |-- User_ID: integer (nullable = true)\n |-- Gender: string (nullable = true)\n |-- Age: integer (nullable = true)\n |-- E
suvDF.show(10,False)
|User_ID |Gender|Age|EstimatedSalary|Purchased|
|15624510|Male |19 |19000
                               10
|15810944|Male |35 |20000
                           |0
|0
|0
                              10
|15668575|Female|26 |43000
|15603246|Female|27 |57000
|15804002|Male |19 |76000
                               0
|15728773|Male |27 |58000
                               10
|15598044|Female|27 |84000
|15694829|Female|32 |150000
                               11
|15600575|Male |25 |33000
                               0
|15727311|Female|35 |65000
                               10
....
+-----
|User_ID |Gender|Age|EstimatedSalary|Purchased|
+-----
|15624510|Male |19 |19000
                               0 |
|15810944|Male |35 |20000
                               10
|15668575|Female|26 |43000
                              10
|15603246|Female|27 |57000
                              10
|15804002|Male |19 |76000
                               0 |
|15728773|Male |27 |58000
                               lΘ
|15598044|Female|27 |84000
|15694829|Female|32 |150000
                              11
|15600575|Male |25 |33000
                               0 |
|15727311|Female|35 |65000
                               0
only showing top 10 rows
```

```
suvDF.describe().show()
suvDF.createOrReplaceTempView("first_view")
|summary| User_ID|Gender| Age| EstimatedSalary| Purchased| | | | | | | |
| count | 400 | 400 | 400 | 400 | 400 | 400 | |
| mean | 1.56915397575E7 | null | 37.655 | 69742.5 | 0.3575 |
| stddev|71658.32158119006| null|10.482876597307927|34096.9602824248|0.4798639635968691|
| min| 15566689|Female| 18| 15000| 0|
| max| 15815236| Male| 60| 150000| 1|
|summary| User_ID|Gender| Age| EstimatedSalary| Purchased|
| count| 400| 400| 400| 400| 400| 400| 400| 0.3575| | 69742.5| 0.3575|
| stddev|71658.32158119006| null|10.482876597307927|34096.9602824248|0.4798639635968691|
| min| 15566689|Female| 18| 15000| 0| | max| 15815236| Male| 60| 150000| 1|
+----+
'\n+----+\n|summary|
                                                                              User_ID|Gender|
data_set_suv= spark.sql("select Gender, Age, EstimatedSalary, Purchased from first_view ")
data_set_suv.show(5,False)
|Gender|Age|EstimatedSalary|Purchased|
+----+
|Gender|Age|EstimatedSalary|Purchased|
|Male |19 |19000 |0
|Male |35 |20000
+----+
only showing top 5 rows
```

```
suv_indexer = StringIndexer(inputCol = "Gender", outputCol = "gen_label")
qen_label_df = suv_indexer.fit(data_set_suv).transform(data_set_suv)
print("after adding label")
gen_label_df.printSchema()
gen_label_df.show(5,False)
|-- Gender: string (nullable = true)
 |-- Age: integer (nullable = true)
 |-- EstimatedSalary: integer (nullable = true)
|-- Purchased: integer (nullable = true)
 |-- gen_label: double (nullable = false)
|Gender|Age|EstimatedSalary|Purchased|gen_label|
+----+
....
after adding label
root
|-- Gender: string (nullable = true)
|-- Age: integer (nullable = true)
|-- EstimatedSalary: integer (nullable = true)
|-- Purchased: integer (nullable = true)
|-- gen_label: double (nullable = false)
+----+
|Gender|Age|EstimatedSalary|Purchased|gen_label|
+----+
```

```
inputCols = ["Age","EstimatedSalary","gen_label"]
outputCol = "features"
suvDF_assembler = VectorAssembler(inputCols = inputCols,outputCol = outputCol)
featuresDf = suvDF_assembler.transform(gen_label_df)
print("featuresDF printSchema")
featuresDf.printSchema()
"""
root
|-- Gender: string (nullable = true)
|-- Age: integer (nullable = true)
|-- EstimatedSalary: integer (nullable = true)
|-- Purchased: integer (nullable = true)
|-- gen_label: double (nullable = false)
|-- featuresDF printSchema
featuresDF printSchema
```

+----+

only showing top 5 rows

root

|-- Gender: string (nullable = true)

```
|-- Age: integer (nullable = true)
 |-- EstimatedSalary: integer (nullable = true)
 |-- Purchased: integer (nullable = true)
 |-- gen_label: double (nullable = false)
 |-- features: vector (nullable = true)
'\nroot\n |-- Gender: string (nullable = true)\n |-- Age: integer (nullable = true)\n |-- EstimatedSalary: integer (nullable = true)
featuresDf.show(10,False)
print("featureDf show")
|Gender|Age|EstimatedSalary|Purchased|gen_label|features
|Gender|Age|EstimatedSalary|Purchased|gen_label|features
+----+
only showing top 10 rows
featureDf show
 \n+----+\n|Gender|Age|EstimatedSalary|Purchased|gen_label|features
suv_indexer = StringIndexer(inputCol = "Purchased", outputCol = "label")
label_df = suv_indexer.fit(featuresDf).transform(featuresDf)
print("after adding pur_label")
label_df.createOrReplaceTempView("main_df")
label_df.printSchema()
root
 |-- Gender: string (nullable = true)
 |-- Age: integer (nullable = true)
 |-- EstimatedSalary: integer (nullable = true)
 |-- Purchased: integer (nullable = true)
 |-- gen_label: double (nullable = false)
 |-- features: vector (nullable = true)
 |-- label: double (nullable = false)
....
```

```
Py4JJavaError: An error occurred while calling o945.fit.
: org.apache.spark.SparkException: Input column Purchased does not exist.
              at org. apache. spark. \verb|ml.feature.StringIndexerBase.$| anonfun$| validateAndTransformSchema$| 2(StringIndexer.scala:128)| anonfun$| 2(StringIndexer.scala:128)| anonfun$| 2(StringIndexer.scala:128)| anonfun$| 2(StringIndexer.scala:128
              at scala.collection.TraversableLike.$anonfun$flatMap$1(TraversableLike.scala:293)
               at scala.collection.IndexedSeqOptimized.foreach(IndexedSeqOptimized.scala:36)
              at scala.collection.IndexedSegOptimized.foreach$(IndexedSegOptimized.scala:33)
               at scala.collection.mutable.ArrayOps$ofRef.foreach(ArrayOps.scala:198)
              at scala.collection.TraversableLike.flatMap(TraversableLike.scala:293)
              at scala.collection.TraversableLike.flatMap$(TraversableLike.scala:290)
              at scala.collection.mutable.ArrayOps$ofRef.flatMap(ArrayOps.scala:198)
               at org.apache.spark.ml.feature.StringIndexerBase.validateAndTransformSchema(StringIndexer.scala:123)
               at org.apache.spark.ml.feature.StringIndexerBase.validateAndTransformSchema$(StringIndexer.scala:115)
               at org. a pache. spark. \verb|ml.feature.StringIndexer.validateAndTransformSchema(StringIndexer.scala:145)| \\
              at org.apache.spark.ml.feature.StringIndexer.transformSchema(StringIndexer.scala:252)
               at org.apache.spark.ml.PipelineStage.transformSchema(Pipeline.scala:71)
               at org.apache.spark.ml.feature.StringIndexer.fit(StringIndexer.scala:237)
               at org.apache.spark.ml.feature.StringIndexer.fit(StringIndexer.scala:145)
               at java.base/jdk.internal.reflect.NativeMethodAccessorImpl.invokeO(Native Method)
               at java.base/jdk.internal.reflect.NativeMethodAccessorImpl.invoke(NativeMethodAccessorImpl.java:62)
               at java.base/jdk.internal.reflect.DelegatingMethodAccessorImpl.invoke(DelegatingMethodAccessorImpl.java:43)
              at java.base/java.lang.reflect.Method.invoke(Method.java:566)
               at py4j.reflection.MethodInvoker.invoke(MethodInvoker.java:244)
               at py4j.reflection.ReflectionEngine.invoke(ReflectionEngine.java:357)
               at py4j.Gateway.invoke(Gateway.java:282)
               at py4j.commands.AbstractCommand.invokeMethod(AbstractCommand.java:132)
               at py4j.commands.CallCommand.execute(CallCommand.java:79)
               at py4j.ClientServerConnection.waitForCommands(ClientServerConnection.java:182)
               at pv4i.ClientServerConnection.run(ClientServerConnection.java:106)
               at java.base/java.lang.Thread.run(Thread.java:829)
```

label df.show(10,False) |Gender|Age|EstimatedSalary|Purchased|gen_label|features |label| |1.0 |1.0 |0.0 |0.0 |1.0 |1.0 [[19.0,19000.0,1.0] |0.0 |Male | 19 | 19000 0 [35.0,20000.0,1.0] [0.0 |Female|26 |43000 10 |[26.0,43000.0,0.0] |0.0 0 | [27.0,57000.0,0.0] [0.0 |Female|27 |57000 |Male |19 |76000 [[19.0,76000.0,1.0] |0.0 0 | 0 |Male | 127 | 158000 I[27.0.58000.0.1.0] | 10.0 |Female|27 |84000 0 | [27.0,84000.0,0.0] [0.0 0.0 |Female|32 |150000 |[32.0,150000.0,0.0]|1.0 |1 |Male | 125 | 133000 10 11.0 |[25.0,33000.0,1.0] |0.0 |Female|35 |65000 10 0.0 [35.0,65000.0,0.0] [0.0

Gender Age Es	stimatedSalary	Purchased	gen_label		label
Male 19 19				[19.0,19000.0,1.0]	
Male 35 20	0000	0	1.0	[35.0,20000.0,1.0]	0.0
Female 26 43	3000 I	0	0.0	[26.0,43000.0,0.0]	0.0
Female 27 57	7000 I	0	0.0	[27.0,57000.0,0.0]	0.0
Male 19 76	6000 I	0	1.0	[19.0,76000.0,1.0]	0.0
Male 27 58	3000 I	0	1.0	[27.0,58000.0,1.0]	0.0
Female 27 84	4000 I	0	0.0	[27.0,84000.0,0.0]	0.0
Female 32 15	50000 I	1	0.0	[32.0,150000.0,0.0]	1.0
Male 25 33	3000 I	0	1.0	[25.0,33000.0,1.0]	0.0
Female 35 65	5000 I	0	0.0	[35.0,65000.0,0.0]	0.0
+					++

only showing top 10 rows

'\n+----+\n|Gender|Age|EstimatedSalary|Purchased|gen_label|featu

```
trainingData,testdata = label_df.randomSplit([0.7,0.3],seed = 42)
print("display training data")
trainingData.show(10,False)
|Gender|Age|EstimatedSalary|Purchased|gen_label|features | label|
+----+---+---
display training data
+----+
|Gender|Age|EstimatedSalary|Purchased|gen_label|features | label|
+----+
only showing top 10 rows
'\n+-----+\n|Gender|Age|EstimatedSalary|Purchased|gen_label|featur
logisticRegression = LogisticRegression().setMaxIter(100).setRegParam(0.02).setElasticNetParam(0.8)
logisticRegressionModel = logisticRegression.fit(trainingData)
predictionDf = logisticRegressionModel.transform(testdata)
print("logisticregession model prediction")
predictionDf.show(10,False)
|Gender|Age|EstimatedSalary|Purchased|gen_label|features | label|rawPrediction
                                                             Iprobability
....
logisticregession model prediction
|Gender|Age|EstimatedSalary|Purchased|gen_label|features | label|rawPrediction
                                                             Iprobability
```

```
10
                                                                                             10.0
                                                                                                                      |Female|18 |86000
|Female|20 |36000
                                                                    10
                                                                                             0.0
                                                                                                                       [20.0,36000.0,0.0] | 0.0 | [4.504284227708987,-4.504284227708987] | [0.989059513373478
                                                                                             10.0
                                                                                                                      [20.0,82000.0,0.0] [0.0 [3.4398933877329734,-3.4398933877329734] [0.968928306305915]
||Female||20 ||82000
                                                                   10
|Female|21 |16000
                                                                                             0.0
                                                                                                                      10
                                                                                                                      [22.0,63000.0,0.0]|0.0 |[3.556295972998962,-3.556295972998962] |[0.972247810228541
|Female|22 |63000
                                                                   10
                                                                                             10.0
                                                                                                                       \hspace{0.5cm} 
|Female|23 |28000
                                                                    0 |
                                                                                             0.0
                                                                                                                      |[23.0,48000.0,0.0]|0.0 |[3.7417613878030007,-3.7417613878030007]|[0.976836949428207
|Female|23 |48000
                                                                                             10.0
                                                                   10
|Female|24 |32000
                                                                    10
                                                                                             0.0
                                                                                                                      [[24.0,32000.0,0.0]]0.0 [[3.950365733910866,-3.950365733910866] [[0.981115815416854
|Female|24 |89000
                                                                    0 |
                                                                                             0.0
                                                                                                                      [24.0,89000.0,0.0] | 0.0 | [2.631446649592763,-2.631446649592763] | [0.9328582152820763
|Female|26 |17000
                                                                    0 |
                                                                                             0.0
                                                                                                                      [26.0,17000.0,0.0] | 0.0 | [3.974212593961549, -3.974212593961549] | [0.981552607774936]
only showing top 10 rows
evaluator = BinaryClassificationEvaluator() .setLabelCol("label").setRawPredictionCol("prediction").setMetricName("areaUnderROC")
accuracy = evaluator.evaluate(predictionDf)
print("accuracy of the model")
print(accuracy * 100)
....
accuracy of the model
83.09386973180077
accuracy of the model
83.09386973180077
'\naccuracy of the model\n83.09386973180077\n\n'
df1 = spark.read.csv("car_suv.csv", header = True, inferSchema = True)
df1.createOrReplaceTempView("input_view")
df1.printSchema()
root
  |-- User_ID: integer (nullable = true)
  |-- Gender: string (nullable = true)
  |-- Age: integer (nullable = true)
 |-- EstimatedSalary: integer (nullable = true)
root
  |-- User_ID: integer (nullable = true)
  |-- Gender: string (nullable = true)
  |-- Age: integer (nullable = true)
  |-- EstimatedSalary: integer (nullable = true)
'\nroot\n |-- User_ID: integer (nullable = true)\n |-- Gender: string (nullable = true)\n |-- Age: integer (nullable = true)\n |-- E
```

```
df1= spark.sql("select Gender,Age,EstimatedSalary from input_view ")
df1.show(5,False)
|Gender|Age|EstimatedSalary|
|Male | 19 | 19000
|Male |35 |20000
|Female|26 |43000
|Female|27 |57000
|Male | 19 | 76000
|Gender|Age|EstimatedSalary|
+----+
|Male |19 |19000
|Male |35 |20000
|Female|26 |43000
|Female|27 |57000
|Male |19 |76000
+----+
only showing top 5 rows
'\n+----+\n|Gender|Age|EstimatedSalary|\n+----+\n|Male |19 |19000
                                                                                                  |\n|Male |3
input_indexer = StringIndexer(inputCol = "Gender", outputCol = "gen_label")
gen_label_input_df = input_indexer.fit(data_set_suv).transform(df1)
print("after adding label")
gen_label_input_df.printSchema()
gen_label_input_df.show(5,False)
after adding label
|-- Gender: string (nullable = true)
 |-- Age: integer (nullable = true)
 |-- EstimatedSalary: integer (nullable = true)
|-- gen_label: double (nullable = false)
|Gender|Age|EstimatedSalary|gen_label|
+----+
+----+
only showing top 5 rows
```

```
outputCol = "features"
input_assembler = VectorAssembler(inputCols = inputCols,outputCol = outputCol)
featuresDf = input_assembler.transform(gen_label_input_df)
print("featuresDF printSchema")
featuresDf.printSchema()
....
root
  |-- Gender: string (nullable = true)
  |-- Age: integer (nullable = true)
  |-- EstimatedSalary: integer (nullable = true)
  |-- gen_label: double (nullable = false)
  |-- features: vector (nullable = true)
featuresDf.show(5,False)
0.00
|Gender|Age|EstimatedSalary|gen_label|features |
 featuresDF printSchema
root
  |-- Gender: string (nullable = true)
  |-- Age: integer (nullable = true)
  |-- EstimatedSalary: integer (nullable = true)
  |-- gen_label: double (nullable = false)
 |-- features: vector (nullable = true)
|Gender|Age|EstimatedSalary|gen_label|features |
+----+
only showing top 5 rows
                                                                                                     --+----Pin-Gender|Age|EstimatedSalary|Purchased|gen_label|features
print("prediction_input")
input_pre = logisticRegressionModel.transform(featuresDf)
input_pre.show(5,False)
|Gender|Age|EstimatedSalary|gen_label|features |rawPrediction
                                                                                                                                                                                         probability
| Male | 19 | 19000 | 1.0 | [19.0,19000.0,1.0] | [5.059264614627391, -5.059264614627391] | [0.993689843271015, 0.006310156728985] | Male | 35 | 20000 | 1.0 | [35.0,20000.0,1.0] | [2.4502288072698626, -2.4502288072698626] | [0.920578181439154, 0.079421818560845] | Female | 26 | 43000 | 0.0 | [26.0,43000.0,0.0] | [3.372600380062064, -3.372600380062064] | [0.9668371685880653, 0.03316283141193] | Female | 27 | 57000 | 0.0 | [27.0,57000.0,0.0] | [2.8870367870551386, -2.8870367870551386] | [0.9472018861086681, 0.05279811389133] | Male | 19 | 76000 | 1.0 | [19.0,76000.0,1.0] | [3.7403455303092885, -3.7403455303092885] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510814264] | [0.9768048918573531, 0.02319510
 +----+---+---
....
```

inputCols = ["Age", "EstimatedSalary", "gen_label"]

prediction_input |Gender|Age|EstimatedSalary|gen_label|features |rawPrediction |1.0 ||[19.0,19000.0,1.0]|[5.059264614627391,-5.059264614627391] ||[0.993689843271015,0.0063101567289| |1.0 ||[35.0,20000.0,1.0]|[2.4502288072698626,-2.4502288072698626]|[0.920578181439154,0.0794218185608| |Male |19 |19000 |Male |35 |20000 11.0 |Female|26 |43000 0.0 [26.0,43000.0,0.0][3.372600380062064,-3.372600380062064] [0.9668371685880653,0.033162831411 0.0 [[27.0,57000.0,0.0]][2.8870367870551386,-2.8870367870551386][0.9472018861086681,0.052798113891] [[19.0,76000.0,1.0]][3.7403455303092885,-3.7403455303092885][0.9768048918573531,0.023195108142] |Female|27 |57000 |Male |19 |76000 1.0 only showing top 5 rows

```
final_out =spark.sql ("SELECT main_df.subject_1_gp,main_df.subject_2_gp,main_df.subject_3_gp," +
    "main_df.subject_4_gp,main_df.subject_5_gp,main_df.grade,main_df.label,input_marks_view.prediction FROM main_df " +
    "JOIN input_marks_view ON main_df.subject_1_gp = input_marks_view.subject_1_gp AND main_df.subject_2_gp = input_marks_view.subject_3_gp AND main_df.subject_4_gp = input_marks_view.subject_4_gp AND " +
    "main_df.subject_5_gp = input_marks_view.subject_5_gp GROUP BY main_df.subject_1_gp,main_df.subject_2_gp," +
    "main_df.subject_3_gp,main_df.subject_4_gp,main_df.subject_5_gp,main_df.grade,input_marks_view.prediction,main_df.label")
```