Final Project of Machine Learning with Spark

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In [3]: | # 2. Data Transformation
        #We need to remove the garbage values from the first row.
        from itertools import islice
        data = data.rdd.mapPartitionsWithIndex(lambda idx, it: islice(it, 1, None) if
        (idx == 0) else it).toDF()
        firstRow = data.head(1)
        #Second row contains the column names.
        columnNames = []
        for i in range(len(firstRow[0])):
            columnNames.append( str(firstRow[0][i]).replace(' ', '_') )
        data = data.rdd.mapPartitionsWithIndex(lambda idx, it: islice(it, 1, None) if
        (idx == 0) else it).toDF()
        #Next we tidy up the dataframe by setting the right column names and casting c
        olumn data to the right type
        for i in range(len(firstRow[0])):
            data = data.withColumnRenamed("_c" + str(i), columnNames[i])
            data = data.withColumn(columnNames[i],data[columnNames[i]].cast('double'))
            # Source: https://forums.databricks.com/questions/9147/how-to-infer-csv-sc
        hema-default-all-columns-like-s.html
        #Now we optimize our data for machine learning. That is extracting features an
        d then normalizing it.
        from pyspark.ml.feature import VectorAssembler
        from pyspark.ml.feature import StringIndexer
        from pyspark.ml.feature import VectorIndexer
        from pyspark.ml.feature import MinMaxScaler
        #Code taken from 1st assignment
        df = data
        list of columns = df.columns
        list of columns.pop() #Remove the target column from feature vector
        assembler = VectorAssembler().setInputCols(list_of_columns).setOutputCol("feat
        ures")
        transformed = assembler.transform(df)
        scaler = MinMaxScaler(inputCol="features",outputCol="scaledFeatures")
        scalerModel = scaler.fit(transformed.select("features"))
        scaledData = scalerModel.transform(transformed)
        #Remove unnecessary columns. We have all the data in vectorised column - scale
        dFeatures
        for col in scaledData.columns:
            if( (col == 'default payment next month') or (col == 'scaledFeatures')):
                pass
            else:
                scaledData = scaledData.drop(col)
        print("Dataframe structure used for training and evaluation:-")
        scaledData.printSchema()
```

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Dataframe structure used for training and evaluation:-
root
    |-- default_payment_next_month: double (nullable = true)
    |-- scaledFeatures: vector (nullable = true)
```

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In [5]: # 4. Model evaluation

# Make predictions.
predictions = model.transform(testData)

from pyspark.ml.evaluation import MulticlassClassificationEvaluator
# Select (prediction, true label) and compute test error
evaluator = MulticlassClassificationEvaluator(
    labelCol="default_payment_next_month", predictionCol="prediction", metricN ame="accuracy")

accuracy = evaluator.evaluate(predictions)
print("DecisionTreeClassifier accuracy = ", str(format((accuracy)*100, '.2f')) + "%")
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DecisionTreeClassifier accuracy = 81.69%

Accuracy of the model is around 82%

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In [6]: # Select example rows to display.
    print("Some values to verify our results:-")
    predictions.select("prediction", "default_payment_next_month").show(10)
```

Some values to verify our results:-

+	+
prediction default_payment_next_month	
+	+
0.0	0.0
0.0	0.0
0.0	0.0
0.0	0.0
0.0	0.0
0.0	0.0
0.0	0.0
0.0	0.0
0.0	0.0
0.0	0.0
+	+
only showing top 10 rows	