

★ ML with PySpark

· Classify/Predict

Datasource

• https://archive.ics.uci.edu/ml/datasets/HCV+data

```
# Load our Pkgs
from pyspark import SparkContext

sc = SparkContext(master='local[2]')

# Spark UI
sc
```

SparkContext

Spark UI

```
Version
v3.5.3
Master
local[2]
AppName
pyspark-shell
```

```
# Load Pkgs
from pyspark.sql import SparkSession
```

```
# Spark
spark = SparkSession.builder.appName("MLwithSpark").getOrCreate()
```

WorkFlow

- Data Prep
- Feature Engineering
- · Build Model
- Evaluate

★ Task

- Predict if a patient is Hep or not based parameter
- The data set contains laboratory values of blood donors and Hepatitis C patients and demographic values like age.

```
# Load our dataset
df = spark.read.csv("hcvdata.csv",header=True,inferSchema=True)
# Preview Dataset
df.show()
```

```
| c0|
        Category|Age|Sex| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT|
        -----
  1|0=Blood Donor| 32| m|38.5|52.5| 7.7|22.1| 7.5| 6.93|3.23|106.0|12.1| 69|
  2|0=Blood Donor| 32| m|38.5|70.3| 18|24.7| 3.9|11.17| 4.8| 74.0|15.6|76.5|
  3|0=Blood Donor| 32| m|46.9|74.7|36.2|52.6| 6.1| 8.84| 5.2| 86.0|33.2|79.3|
  4|0=Blood Donor| 32| m|43.2| 52|30.6|22.6|18.9| 7.33|4.74| 80.0|33.8|75.7|
  5|0=Blood Donor| 32|
                      m|39.2|74.1|32.6|24.8| 9.6| 9.15|4.32| 76.0|29.9|68.7|
                      m|41.6|43.3|18.5|19.7|12.3| 9.92|6.05|111.0|91.0| 74|
  6|0=Blood Donor| 32|
  7|0=Blood Donor| 32|
                      m|46.3|41.3|17.5|17.8| 8.5| 7.01|4.79| 70.0|16.9|74.5|
                      m|42.2|41.9|35.8|31.1|16.1| 5.82| 4.6|109.0|21.5|67.1|
  8|0=Blood Donor| 32|
  9|0=Blood Donor| 32|
                      m|50.9|65.5|23.2|21.2| 6.9| 8.69| 4.1| 83.0|13.7|71.3|
 10|0=Blood Donor| 32|
                      m|42.4|86.3|20.3|20.0|35.2| 5.46|4.45| 81.0|15.9|69.9|
 11|0=Blood Donor| 32|
                     m|44.3|52.3|21.7|22.4|17.2| 4.15|3.57| 78.0|24.1|75.4|
 12|0=Blood Donor| 33|
                      m|46.4|68.2|10.3|20.0| 5.7| 7.36| 4.3| 79.0|18.7|68.6|
                     m|36.3|78.6|23.6|22.0| 7.0| 8.56|5.38| 78.0|19.4|68.7|
 13|0=Blood Donor| 33|
 14|0=Blood Donor| 33|
                      m 39 51.7 15.9 24.0 6.8 6.46 3.38 65.0 7.0 70.4
 15|0=Blood Donor| 33|
                      m|38.7|39.8|22.5|23.0| 4.1| 4.63|4.97| 63.0|15.2|71.9|
| 16|0=Blood Donor| 33|
                     m|41.8| 65|33.1|38.0| 6.6| 8.83|4.43| 71.0|24.0|72.7|
                      m|40.9| 73|17.2|22.9|10.0| 6.98|5.22| 90.0|14.7|72.4|
 17 0=Blood Donor 33
 18|0=Blood Donor| 33|
                      m|45.2|88.3|32.4|31.2|10.1| 9.78|5.51|102.0|48.5|76.5|
 19|0=Blood Donor| 33| m|36.6|57.1|38.9|40.3|24.9| 9.62| 5.5|112.0|27.6|69.3|
20|0=Blood Donor| 33| m| 42|63.1|32.6|34.9|11.2| 7.01|4.05|105.0|19.1|68.1|
+---+----+---+---+---+---+----+
only showing top 20 rows
```

```
# check for columns
print(df.columns)

['_c0', 'Category', 'Age', 'Sex', 'ALB', 'ALP', 'ALT', 'AST', 'BIL', 'CHE', 'CHOL', '

# Rearrange
df = df.select('Age', 'Sex', 'ALB', 'ALP', 'ALT', 'AST', 'BIL', 'CHE', 'CHOL', 'CREA', 'G

df.show(5)
```

```
|Age|Sex| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT|
                                                          Category
    32 m 38.5 52.5 7.7 22.1 7.5 6.93 3.23 106.0 12.1 69 0 = Blood Donor
    | 32| m|38.5|70.3| 18|24.7| 3.9|11.17| 4.8| 74.0|15.6|76.5|0=Blood Donor|
    32 m|46.9|74.7|36.2|52.6| 6.1| 8.84| 5.2| 86.0|33.2|79.3|0=Blood Donor|
    32 m 43.2 52 30.6 22.6 18.9 7.33 4.74 80.0 33.8 75.7 0=Blood Donor
    32 m 39.2 74.1 32.6 24.8 9.6 9.15 4.32 76.0 29.9 68.7 0 = Blood Donor
    only showing top 5 rows
# Check for datatypes
# Before InferSchema=True
df.dtypes
    [('Age', 'int'),
     ('Sex', 'string'),
     ('ALB', 'string'),
     ('ALP', 'string'),
     ('ALT', 'string'),
     ('AST', 'double'),
     ('BIL', 'double'),
     ('CHE', 'double'),
    ('CHOL', 'string'), ('CREA', 'double'),
     ('GGT', 'double'),
     ('PROT', 'string'),
     ('Category', 'string')]
# After InferSchema
df.dtypes
    [('Age', 'int'),
     ('Sex', 'string'),
     ('ALB', 'string'),
     ('ALP', 'string'),
     ('ALT', 'string'),
     ('AST', 'double'),
     ('BIL', 'double'),
     ('CHE', 'double'),
     ('CHOL', 'string'),
     ('CREA', 'double'),
     ('GGT', 'double'),
     ('PROT', 'string'),
     ('Category', 'string')]
# Check for the Schema
df.printSchema()
    root
     |-- Age: integer (nullable = true)
```

```
|-- Sex: string (nullable = true)
     |-- ALB: string (nullable = true)
     |-- ALP: string (nullable = true)
     |-- ALT: string (nullable = true)
     |-- AST: double (nullable = true)
     |-- BIL: double (nullable = true)
     |-- CHE: double (nullable = true)
     |-- CHOL: string (nullable = true)
     |-- CREA: double (nullable = true)
     |-- GGT: double (nullable = true)
     |-- PROT: string (nullable = true)
     |-- Category: string (nullable = true)
# Descriptive summary
print(df.describe().show())
    |summary|
                 Age| Sex|
                                          ALB
                                                           ALP
                        615 | 615 |
                                           615 l
      mean | 47.40813008130081 | NULL | 41.62019543973941 | 68.28391959798999 | 28.45081433224
    | stddev|10.055105445519239|NULL|5.780629404103076|26.028315300123676|25.469688813870
                         19|
                                         14.9
                            fl
                                    NA|
                         77
                                                            NA
        max
                             m
    None
# Value Count
df.groupBy('Category').count().show()
    +----+
             Category | count |
          0=Blood Donor | 533|
            3=Cirrhosis | 30|
            2=Fibrosis| 21|
    |Os=suspect Blood ...|
        1=Hepatitis 24
#### Feature Engineering
# + Numberical Values
# + Vectorization
# + Scaling
df.show(5)
```

```
|Age|Sex| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT|
                                                             Category
    32 m 38.5 52.5 7.7 22.1 7.5 6.93 3.23 106.0 12.1 69 0 = Blood Donor
    32 m 38.5 70.3 18 24.7 3.9 11.17 4.8 74.0 15.6 76.5 0 = Blood Donor
    32 m 46.9 74.7 36.2 52.6 6.1 8.84 5.2 86.0 33.2 79.3 0 = Blood Donor
    32 m 43.2 52 30.6 22.6 18.9 7.33 4.74 80.0 33.8 75.7 0 = Blood Donor
    32 m 39.2 74.1 32.6 24.8 9.6 9.15 4.32 76.0 29.9 68.7 0 = Blood Donor
    only showing top 5 rows
import pyspark.ml
dir(pyspark.ml)
    ['Estimator',
     'Model',
     'Pipeline',
     'PipelineModel',
     'PredictionModel',
     'Predictor',
     'TorchDistributor',
     'Transformer',
     'UnaryTransformer',
     ' all ',
      __builtins___',
      _cached_',
      _doc__',
_file__',
      _loader___',
     '_name_',
      __package__',
      path_',
      spec__',
     'base',
     'classification',
     'clustering',
     'common',
     'evaluation',
     'feature',
     'fpm',
     'image',
     'linalg',
     'param',
     'pipeline',
     'recommendation',
     'regression',
     'stat',
     'torch',
     'tree',
     'tuning',
     'util',
     'wrapper']
```

```
# Load ML Pkgs
from pyspark.ml.feature import VectorAssembler,StringIndexer
df.show(4)
   |Age|Sex| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT|
   32 m 38.5 52.5 7.7 22.1 7.5 6.93 3.23 106.0 12.1 69 0 = Blood Donor
   32 m 38.5 70.3 18 24.7 3.9 11.17 4.8 74.0 15.6 76.5 0 = Blood Donor
   32 m|46.9|74.7|36.2|52.6| 6.1| 8.84| 5.2| 86.0|33.2|79.3|0=Blood Donor|
   32 m 43.2 52 30.6 22.6 18.9 7.33 4.74 80.0 33.8 75.7 0 = Blood Donor
   only showing top 4 rows
# Unique Values for Sex
df.select('Sex').distinct().show()
   +---+
   Sex
   +---+
    | m|
     f|
   +---+
# Convert the string into numerical code
# label encoding
genderEncoder = StringIndexer(inputCol='Sex',outputCol='Gender').fit(df)
df = genderEncoder.transform(df)
df.show(5)
   |Age|Sex| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT| Category|Gender|
    32 m|38.5|52.5| 7.7|22.1| 7.5| 6.93|3.23|106.0|12.1| 69|0=Blood Donor
                                                             0.0
   32 m 38.5 70.3 18 24.7 3.9 11.17 4.8 74.0 15.6 76.5 0 = Blood Donor
                                                             0.0
    32 m|46.9|74.7|36.2|52.6| 6.1| 8.84| 5.2| 86.0|33.2|79.3|0=Blood Donor|
                                                            0.0
   32 m|43.2 52|30.6|22.6|18.9| 7.33|4.74| 80.0|33.8|75.7|0=Blood Donor
                                                             0.0
   32 m 39.2 74.1 32.6 24.8 9.6 9.15 4.32 76.0 29.9 68.7 0 = Blood Donor
   only showing top 5 rows
# Encoding for Category
```

```
# Label Encoding
catEncoder = StringIndexer(inputCol='Category',outputCol='Target').fit(df)
df = catEncoder.transform(df)
df.show(5)
    |Age|Sex| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT|
                                                            Category | Gender | Tar
    32 m 38.5 52.5 7.7 22.1 7.5 6.93 3.23 106.0 12.1 69 0 Blood Donor
                                                                       0.0
          m|38.5|70.3| 18|24.7| 3.9|11.17| 4.8| 74.0|15.6|76.5|0=Blood Donor|
    | 32|
                                                                       0.0
          m|46.9|74.7|36.2|52.6| 6.1| 8.84| 5.2| 86.0|33.2|79.3|0=Blood Donor|
    32|
                                                                       0.01
          m|43.2| 52|30.6|22.6|18.9| 7.33|4.74| 80.0|33.8|75.7|0=Blood Donor|
    | 32|
                                                                       0.0
    32 m 39.2 74.1 32.6 24.8 9.6 9.15 4.32 76.0 29.9 68.7 0 = Blood Donor
                                                                       0.0
    +---+---+----+
    only showing top 5 rows
# Get the labels
catEncoder.labels
    ['0=Blood Donor',
     '3=Cirrhosis',
     '1=Hepatitis',
     '2=Fibrosis',
     '0s=suspect Blood Donor']
# IndexToString
from pyspark.ml.feature import IndexToString
converter = IndexToString(inputCol='Target',outputCol='orig_cat')
converted df = converter.transform(df)
converted_df.show()
    |Age|Sex| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT|
                                                            Category | Gender | Tar
          m|38.5|52.5| 7.7|22.1| 7.5| 6.93|3.23|106.0|12.1| 69|0=Blood Donor|
                                                                       0.01
      32|
          m|38.5|70.3| 18|24.7| 3.9|11.17| 4.8| 74.0|15.6|76.5|0=Blood Donor|
      32
                                                                       0.0
      32
          m|46.9|74.7|36.2|52.6| 6.1| 8.84| 5.2| 86.0|33.2|79.3|0=Blood Donor|
                                                                       0.01
          m|43.2| 52|30.6|22.6|18.9| 7.33|4.74| 80.0|33.8|75.7|0=Blood Donor|
      32
                                                                       0.01
          m|39.2|74.1|32.6|24.8| 9.6| 9.15|4.32| 76.0|29.9|68.7|0=Blood Donor|
      32
                                                                       0.0
          m|41.6|43.3|18.5|19.7|12.3| 9.92|6.05|111.0|91.0| 74|0=Blood Donor|
      32
                                                                       0.01
          m|46.3|41.3|17.5|17.8| 8.5| 7.01|4.79| 70.0|16.9|74.5|0=Blood Donor|
      32|
                                                                       0.0
          m|42.2|41.9|35.8|31.1|16.1| 5.82| 4.6|109.0|21.5|67.1|0=Blood Donor|
      32
                                                                       0.0
          m|50.9|65.5|23.2|21.2| 6.9| 8.69| 4.1| 83.0|13.7|71.3|0=Blood Donor|
      32|
                                                                       0.01
     32
          m|42.4|86.3|20.3|20.0|35.2| 5.46|4.45| 81.0|15.9|69.9|0=Blood Donor|
                                                                       0.0|
```

```
m|44.3|52.3|21.7|22.4|17.2| 4.15|3.57| 78.0|24.1|75.4|0=Blood Donor|
                                                                       0.0
 32
 33|
      m|46.4|68.2|10.3|20.0| 5.7| 7.36| 4.3| 79.0|18.7|68.6|0=Blood Donor|
                                                                       0.01
 33
      m|36.3|78.6|23.6|22.0| 7.0| 8.56|5.38| 78.0|19.4|68.7|0=Blood Donor|
                                                                       0.0
 33
          39|51.7|15.9|24.0| 6.8| 6.46|3.38| 65.0| 7.0|70.4|0=Blood Donor|
                                                                       0.0
      m|38.7|39.8|22.5|23.0| 4.1| 4.63|4.97| 63.0|15.2|71.9|0=Blood Donor|
 33|
                                                                       0.0
 33
              65|33.1|38.0| 6.6| 8.83|4.43| 71.0|24.0|72.7|0=Blood Donor|
                                                                       0.01
      m|40.9| 73|17.2|22.9|10.0| 6.98|5.22| 90.0|14.7|72.4|0=Blood Donor|
 33
                                                                       0.01
      m|45.2|88.3|32.4|31.2|10.1| 9.78|5.51|102.0|48.5|76.5|0=Blood Donor|
33
                                                                       0.0
      m|36.6|57.1|38.9|40.3|24.9| 9.62| 5.5|112.0|27.6|69.3|0=Blood Donor|
 33 l
                                                                       0.0
      m 42 63.1 32.6 34.9 11.2 7.01 4.05 105.0 19.1 68.1 0=Blood Donor
33
                                                                       0.0
only showing top 20 rows
```

```
### Feature
df.show()
```

```
|Age|Sex| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT|
                                                          Category | Gender | Tar
m|38.5|52.5| 7.7|22.1| 7.5| 6.93|3.23|106.0|12.1| 69|0=Blood Donor|
 32 l
                                                                     0.01
 32
      m|38.5|70.3| 18|24.7| 3.9|11.17| 4.8| 74.0|15.6|76.5|0=Blood Donor|
                                                                     0.0
      m|46.9|74.7|36.2|52.6| 6.1| 8.84| 5.2| 86.0|33.2|79.3|0=Blood Donor|
 32
                                                                     0.01
 32
      m|43.2| 52|30.6|22.6|18.9| 7.33|4.74| 80.0|33.8|75.7|0=Blood Donor|
                                                                     0.0
      m | 39.2 | 74.1 | 32.6 | 24.8 | 9.6 | 9.15 | 4.32 | 76.0 | 29.9 | 68.7 | 0 = Blood Donor |
 32
                                                                     0.0
 32
      m|41.6|43.3|18.5|19.7|12.3| 9.92|6.05|111.0|91.0| 74|0=Blood Donor|
                                                                     0.01
 32
      m|46.3|41.3|17.5|17.8| 8.5| 7.01|4.79| 70.0|16.9|74.5|0=Blood Donor|
                                                                     0.0
      m|42.2|41.9|35.8|31.1|16.1| 5.82| 4.6|109.0|21.5|67.1|0=Blood Donor|
 32
                                                                     0.0
      m|50.9|65.5|23.2|21.2| 6.9| 8.69| 4.1| 83.0|13.7|71.3|0=Blood Donor|
 32|
                                                                     0.0
 32
      m|42.4|86.3|20.3|20.0|35.2| 5.46|4.45| 81.0|15.9|69.9|0=Blood Donor|
                                                                     0.0
 32
      m|44.3|52.3|21.7|22.4|17.2| 4.15|3.57| 78.0|24.1|75.4|0=Blood Donor|
                                                                     0.01
      m|46.4|68.2|10.3|20.0| 5.7| 7.36| 4.3| 79.0|18.7|68.6|0=Blood Donor|
 33|
                                                                     0.0
      m|36.3|78.6|23.6|22.0| 7.0| 8.56|5.38| 78.0|19.4|68.7|0=Blood Donor|
 33
                                                                     0.0
         39|51.7|15.9|24.0| 6.8| 6.46|3.38| 65.0| 7.0|70.4|0=Blood Donor|
 33|
                                                                     0.0
 33|
      m|38.7|39.8|22.5|23.0| 4.1| 4.63|4.97| 63.0|15.2|71.9|0=Blood Donor|
                                                                     0.0
      m|41.8| 65|33.1|38.0| 6.6| 8.83|4.43| 71.0|24.0|72.7|0=Blood Donor|
 33
                                                                     0.0
      m|40.9| 73|17.2|22.9|10.0| 6.98|5.22| 90.0|14.7|72.4|0=Blood Donor|
 33 l
                                                                     0.01
 33
      m|45.2|88.3|32.4|31.2|10.1| 9.78|5.51|102.0|48.5|76.5|0=Blood Donor|
                                                                     0.01
 33|
      m|36.6|57.1|38.9|40.3|24.9| 9.62| 5.5|112.0|27.6|69.3|0=Blood Donor|
                                                                     0.0
      m 42 63.1 32.6 34.9 11.2 7.01 4.05 105.0 19.1 68.1 0=Blood Donor
33|
                                                                     0.01
```

only showing top 20 rows

```
('ALB', 'string'),
      ('ALP', 'string'),
('ALT', 'string'),
      ('AST', 'double'),
      ('BIL', 'double'),
      ('CHE', 'double'),
      ('CHOL', 'string'),
      ('CREA', 'double'),
      ('GGT', 'double'),
      ('PROT', 'string'),
      ('Category', 'string'),
      ('Gender', 'double'),
('Target', 'double')]
df2 = df.select('Age','Gender', 'ALB', 'ALP', 'ALT', 'AST', 'BIL', 'CHE', 'CHOL', 'CREA',
df2.printSchema()
     root
      |-- Age: integer (nullable = true)
      |-- Gender: double (nullable = false)
      |-- ALB: string (nullable = true)
      |-- ALP: string (nullable = true)
      |-- ALT: string (nullable = true)
      |-- AST: double (nullable = true)
      |-- BIL: double (nullable = true)
      |-- CHE: double (nullable = true)
      |-- CHOL: string (nullable = true)
      |-- CREA: double (nullable = true)
      |-- GGT: double (nullable = true)
      |-- PROT: string (nullable = true)
      |-- Target: double (nullable = false)
# df2.fillna(0,subset=['col1'])
df2 = df2.toPandas().replace('NA',0).astype(float)
type(df2)
```

```
type(df)
# Convert To PySpark Dataframe
new df = spark.createDataFrame(df2)
new_df.show()
     | Age|Gender| ALB| ALP| ALT| AST| BIL|
                                         CHE|CHOL| CREA| GGT|PROT|Target|
    32.0
             0.0|38.5|52.5| 7.7|22.1| 7.5| 6.93|3.23|106.0|12.1|69.0|
                                                                     0.01
     |32.0|
             0.0|38.5|70.3|18.0|24.7| 3.9|11.17| 4.8| 74.0|15.6|76.5|
                                                                     0.0
     32.0
             0.0 46.9 74.7 36.2 52.6 6.1 8.84 5.2 86.0 33.2 79.3
                                                                     0.0
             0.0|43.2|52.0|30.6|22.6|18.9| 7.33|4.74| 80.0|33.8|75.7|
     |32.0|
                                                                     0.01
             0.0|39.2|74.1|32.6|24.8| 9.6| 9.15|4.32| 76.0|29.9|68.7|
     |32.0|
                                                                     0.01
     |32.0|
             0.0|41.6|43.3|18.5|19.7|12.3| 9.92|6.05|111.0|91.0|74.0|
                                                                     0.0
     32.0
             0.0 | 46.3 | 41.3 | 17.5 | 17.8 | 8.5 | 7.01 | 4.79 | 70.0 | 16.9 | 74.5 |
                                                                     0.0
     |32.0|
             0.0|42.2|41.9|35.8|31.1|16.1| 5.82| 4.6|109.0|21.5|67.1|
                                                                     0.0
     |32.0|
             0.0|50.9|65.5|23.2|21.2| 6.9| 8.69| 4.1| 83.0|13.7|71.3|
                                                                     0.0
     32.0
             0.0|42.4|86.3|20.3|20.0|35.2| 5.46|4.45| 81.0|15.9|69.9|
                                                                     0.0
     |32.0|
             0.0 44.3 52.3 21.7 22.4 17.2 4.15 3.57 78.0 24.1 75.4
                                                                     0.0
     |33.0|
             0.0 46.4 68.2 10.3 20.0 5.7 7.36 4.3 79.0 18.7 68.6
                                                                     0.0
     33.0
             0.0|36.3|78.6|23.6|22.0| 7.0| 8.56|5.38| 78.0|19.4|68.7|
                                                                     0.0
     |33.0|
             0.0|39.0|51.7|15.9|24.0| 6.8| 6.46|3.38| 65.0| 7.0|70.4|
                                                                     0.0
     |33.0|
             0.0|38.7|39.8|22.5|23.0| 4.1| 4.63|4.97| 63.0|15.2|71.9|
                                                                     0.0
     33.0
             0.0|41.8|65.0|33.1|38.0| 6.6| 8.83|4.43| 71.0|24.0|72.7|
                                                                     0.0
             0.0|40.9|73.0|17.2|22.9|10.0| 6.98|5.22| 90.0|14.7|72.4|
     |33.0|
                                                                     0.0
     |33.0|
             0.0|45.2|88.3|32.4|31.2|10.1| 9.78|5.51|102.0|48.5|76.5|
                                                                     0.01
     33.0
             0.0|36.6|57.1|38.9|40.3|24.9| 9.62| 5.5|112.0|27.6|69.3|
                                                                     0.0
     33.0
             0.0|42.0|63.1|32.6|34.9|11.2| 7.01|4.05|105.0|19.1|68.1|
                                                                     0.0
    only showing top 20 rows
# Check For DTYpes and Schema
new_df.printSchema()
     |-- Age: double (nullable = true)
```

```
|-- Gender: double (nullable = true)
     |-- ALB: double (nullable = true)
     |-- ALP: double (nullable = true)
     |-- ALT: double (nullable = true)
     |-- AST: double (nullable = true)
     |-- BIL: double (nullable = true)
     |-- CHE: double (nullable = true)
     |-- CHOL: double (nullable = true)
     |-- CREA: double (nullable = true)
     |-- GGT: double (nullable = true)
     |-- PROT: double (nullable = true)
     |-- Target: double (nullable = true)
required_features = ['Age','Gender', 'ALB', 'ALP', 'ALT', 'AST', 'BIL', 'CHE', 'CHOL', 'C
# VectorAsm
vec_assembler = VectorAssembler(inputCols=required_features,outputCol='features')
vec df = vec assembler.transform(new df)
vec_df.show(5)
    | Age|Gender| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT|Target|
    0.0|38.5|52.5| 7.7|22.1| 7.5| 6.93|3.23|106.0|12.1|69.0| 0.0|[32.0,0.0,38.
    32.0
    |32.0| 0.0|38.5|70.3|18.0|24.7| 3.9|11.17| 4.8| 74.0|15.6|76.5| 0.0|[32.0,0.0,38.
           0.0|46.9|74.7|36.2|52.6|6.1|8.84|5.2|86.0|33.2|79.3|0.0|[32.0,0.0,46.
    |32.0|
           0.0|43.2|52.0|30.6|22.6|18.9| 7.33|4.74| 80.0|33.8|75.7| 0.0|[32.0,0.0,43.
    [32.0]
           0.0|39.2|74.1|32.6|24.8| 9.6| 9.15|4.32| 76.0|29.9|68.7|
                                                           0.0|[32.0,0.0,39.
    only showing top 5 rows
 Train, Test Split
train df,test df = vec df.randomSplit([0.7,0.3])
train df.count()
    432
train df.show(4)
    | Age|Gender| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT|Target|
                                                                        f
```

```
0.0|38.5|70.3|18.0|24.7| 3.9|11.17| 4.8| 74.0|15.6|76.5|
                                                           0.0|[32.0,0.0,38.
           0.0|41.6|43.3|18.5|19.7|12.3| 9.92|6.05|111.0|91.0|74.0|
    |32.0|
                                                          0.0|[32.0,0.0,41.
           0.0|42.2|41.9|35.8|31.1|16.1| 5.82| 4.6|109.0|21.5|67.1|
    32.0
                                                           0.0|[32.0,0.0,42.
    |32.0|
           0.0|42.4|86.3|20.3|20.0|35.2| 5.46|4.45| 81.0|15.9|69.9|
                                                           0.0|[32.0,0.0,42.
    only showing top 4 rows
# #### Model Building
# # + Pyspark.ml: DataFrame
# + Pyspark.mllib: RDD /Legacy
from pyspark.ml.classification import LogisticRegression, DecisionTreeClassifier
# Logist Model
lr = LogisticRegression(featuresCol='features',labelCol='Target')
lr_model = lr.fit(train_df)
y pred = lr model.transform(test df)
y_pred.show()
```

```
| Age|Gender| ALB| ALP| ALT| AST| BIL| CHE|CHOL| CREA| GGT|PROT|Target|
32.0
        0.0|38.5| 52.5| 7.7|22.1| 7.5| 6.93|3.23|106.0|12.1|69.0|
                                                                   0.0|[32.0,0.0,38
|32.0|
        0.0|39.2| 74.1|32.6|24.8| 9.6| 9.15|4.32| 76.0|29.9|68.7|
                                                                   0.0 | [32.0,0.0,39
|32.0|
        0.0|44.3| 52.3|21.7|22.4|17.2| 4.15|3.57| 78.0|24.1|75.4|
                                                                   0.0 | [32.0,0.0,44
32.0
        0.0 46.9 74.7 36.2 52.6 6.1 8.84 5.2 86.0 33.2 79.3
                                                                   0.0|[32.0,0.0,46]
|33.0|
        0.0|36.6| 57.1|38.9|40.3|24.9| 9.62| 5.5|112.0|27.6|69.3|
                                                                   0.0 | [33.0,0.0,36
33.0
        0.0|44.3| 49.8|32.1|21.6|13.1| 7.44|5.59|103.0|30.2|74.0|
                                                                   0.0 [33.0,0.0,44
34.0
        0.0 | 29.0 | 41.6 | 29.1 | 16.1 | 4.8 | 6.82 | 4.03 | 62.0 | 14.5 | 53.2 |
                                                                   0.0 [34.0,0.0,29
34.0
        0.0|43.6| 58.9|47.1|31.1|18.5| 9.14|4.99| 95.0|22.2|69.3|
                                                                   0.0|[34.0,0.0,43]
        0.0|44.8| 77.7|36.9|31.0|19.5|10.51|5.59| 80.0|23.7|78.9|
34.0
                                                                   0.0|[34.0,0.0,44
        0.0|44.5| 70.3|26.2|25.1| 5.1|10.12|4.69| 82.0|20.7|67.2|
35.0
                                                                   0.0 | [35.0,0.0,44
        0.0|42.6| 65.3|35.8|27.1|15.7|10.66|4.38| 96.0|34.7|71.0|
|36.0|
                                                                   0.0|[36.0,0.0,42]
|36.0|
        0.0|46.1| 58.5|26.8|25.3| 6.0| 6.61|5.07| 71.0|10.5|79.6|
                                                                   0.0|[36.0,0.0,46
        0.0|40.8|118.9|17.2|19.2| 3.2| 9.17|4.26| 88.0|13.5|72.0|
37.0
                                                                   0.0 | [37.0,0.0,40
37.0
        0.0|43.6| 72.8|51.4|43.7|13.8| 8.16|4.88| 70.0|94.5|75.2|
                                                                   0.0|[37.0,0.0,43
37.0
        0.0|46.4| 53.3|20.2|24.9| 8.7| 8.63| 5.9| 86.0|23.3|78.9|
                                                                   0.0 | [37.0,0.0,46
        0.0|48.7| 62.3|21.0|21.1|41.9| 9.71|4.02| 84.0|16.0|75.1|
37.0
                                                                   0.0 | [37.0,0.0,48
        0.0|40.5| 61.7|18.6|24.7| 6.7| 8.47|6.05| 89.0|19.6|75.6|
|38.0|
                                                                   0.0 | [38.0,0.0,40
38.0
        0.0|42.0| 42.7|34.8|42.2| 3.3| 6.1|4.74| 96.0|14.6|66.7|
                                                                   0.0 | [38.0,0.0,42
38.0
        0.0|44.7| 69.4|47.4|35.1|16.7| 6.9|4.14| 67.0|17.3|70.1|
                                                                   0.0|[38.0,0.0,44]
        0.0|38.8| 52.5|54.3|31.3|10.1|10.68|6.26| 81.0|31.5|77.2|
139.01
                                                                   0.0|[39.0,0.0,38]
only showing top 20 rows
```

```
print(y_pred.columns)
     ['Age', 'Gender', 'ALB', 'ALP', 'ALT', 'AST', 'BIL', 'CHE', 'CHOL', 'CREA', 'GGT', 'P
y_pred.select('target','rawPrediction', 'probability', 'prediction').show()
```

```
probability|prediction|
|target|
              rawPrediction
+----+
   0.0|[101.642773965286...|[1.0,1.2996356446...|
                                                      0.0
   0.0|[100.685400347765...|[1.0,2.9875755900...|
                                                      0.0
   0.0|[95.5397494065700...|[1.0,5.0698301252...|
                                                      0.0
   0.0 | [108.023431331312... | [1.0, 2.6499082283... |
                                                      0.0
   0.0 | [90.7458371264038... | [1.0,2.9392236820... |
                                                      0.0
   0.0|[107.125675388156...|[1.0,1.1367421106...|
                                                      0.0
   0.0 | [89.8138911596561... | [0.99999999999992... |
                                                      0.0
   0.0|[111.212065814535...|[1.0,4.4675788873...|
                                                      0.0
   0.0|[99.9904476721623...|[1.0,1.4964236132...|
                                                      0.0
   0.0 | [118.089253726244... | [1.0, 3.1657793756... |
                                                      0.0
   0.0 | [103.996663549684... | [1.0,1.0199994422... |
                                                      0.0
   0.0 | [106.293920030025... | [1.0,4.5660193153... |
                                                      0.0
   0.0 | [124.395758477991... | [1.0,1.9894080921... |
                                                      0.0
   0.0|[84.8910388865561...|[1.0,1.1625273040...|
                                                      0.0
   0.0|[102.744211158330...|[1.0,8.8429670519...|
                                                      0.0
   0.0 | [93.9041136697406... | [0.99999985749169... |
                                                      0.0
   0.0|[99.1396325342617...|[1.0,9.1969669784...|
                                                      0.0
   0.0|[114.710840838580...|[1.0,3.3847603408...|
                                                      0.0
   0.0 | [111.063146444288... | [1.0, 2.5865412225... |
                                                      0.0
   0.0 | [90.8409206745444... | [1.0,8.7340976197... |
                                                      0.0
+----+
```

only showing top 20 rows

Model Evaluation

```
from pyspark.ml.evaluation import MulticlassClassificationEvaluator
# How to Check For Accuracy
multi_evaluator = MulticlassClassificationEvaluator(labelCol='Target',metricName='accurac
multi_evaluator.evaluate(y_pred)
     0.9180327868852459
```

Precision,F1 Score,Recall : Classification Report

```
from pyspark.mllib.evaluation import MulticlassMetrics
lr_metric = MulticlassMetrics(y_pred['target', 'prediction'].rdd)
     /usr/local/lib/python3.10/dist-packages/pyspark/sql/context.py:158: FutureWarning: De
       warnings.warn(
dir(lr_metric)
     ['__class__',
       '__del__',
        _delattr__',
      ' dict_',
        _dir__',
         _doc__
        _eq__',
       '__format__',
       '_ge_',
       '__getattribute__',
       '_gt_',
       __hash___',
       __
'__init__',
       '__init_subclass__',
       _le_',
__lt__',
        __module__',
       '_ne_',
        __new___',
        __reduce__',
       __reduce_ex__',
       _repr_',
       __setattr__',
__sizeof__',
       '_str_',
        __subclasshook___',
       '_weakref_',
       _java_model',
      '_sc',
      'accuracy',
      'call',
      'confusionMatrix',
      'fMeasure',
      'falsePositiveRate',
      'logLoss',
      'precision',
      'recall',
      'truePositiveRate',
      'weightedFMeasure',
      'weightedFalsePositiveRate',
      'weightedPrecision',
      'weightedRecall',
      'weightedTruePositiveRate']
```

```
print("Accuracy", lr_metric.accuracy)
     Accuracy 0.9180327868852459
print("Precision", lr_metric.precision(1.0))
print("Recall", lr_metric.recall(1.0))
print("F1Score", lr_metric.fMeasure(1.0))
     Precision 0.5
     Recall 0.7142857142857143
     F1Score 0.588235294117647
dir(lr_model)
     ['__abstractmethods__',
      '_annotations_',
        __class___',
      ___class_getitem__',
      '_del_',
       __delattr__',
       '_dict_',
        dir__',
       __doc__
        _eq__',
       __format__',
       _ge_',
        __getattribute___',
       _gt_',
        __hash___',
         __init__',
       init_subclass__',
        le_',
        __lt___',
       __module__',
       _ne_',
       __new__',
       __orig_bases__',
       __parameters__',
       _reduce_',
       __reduce_ex__',
        repr_',
        _setattr__',
        _sizeof__',
       slots_',
        _str__',
        __subclasshook___',
       _weakref_',
       _abc_impl',
       _call_java',
       _checkThresholdConsistency',
       _copyValues',
       '_copy_params',
```

```
'_create_from_java_class',
       _create_params_from_java',
      '_defaultParamMap',
'_dummy',
       _empty_java_param_map',
        _from_java',
       _is_protocol',
       '_java_obj',
       '_make_java_param_pair',
       _new_java_array',
       _new_java_obj',
       _paramMap',
       _params',
       _randomUID',
       _resetUid',
       '_resolveParam',
       _set',
       '_setDefault',
       '_shouldOwn',
      '_testOwnParam',
# Saving Model
lr_model.save("lr_model_30")
lr_model.write().save("mylr_model")
```