# Project Development Phase SPRINT 1

| Date         | 29.10.2022                             |
|--------------|--|
| Team ID      | PNT2022TMID17773                       |
| Project Name | Project -Statistical Machine Learning  |
|              | Approaches To Liver Disease Prediction |

### **Executable Program**

import numpy as np
# for dataframes
import pandas as pd
# for easier visualization
import seaborn as sns
# for visualization and to display plots
from matplotlib import pyplot as plt
# import color maps
from matplotlib.colors import ListedColormap

df=pd.read\_csv('indian\_liver\_patient.csv')
df

| 13]: |        | Age  | Gender   | Total_Bilirubin | Direct_Bilirubin | Alkaline_Phosphotase | Alamine_Aminotransferase | Aspartate_Aminotransferase | Total_Protiens | Albumin | Albu |
|------|--------|------|----------|-----------------|------------------|----------------------|--------------------------|----------------------------|----------------|---------|------|
|      | 0      | 65   | Female   | 0.7             | 0.1              | 187                  | 16                       | 18                         | 6.8            | 3.3     |      |
|      | 1      | 62   | Male     | 10.9            | 5.5              | 699                  | 64                       | 100                        | 7.5            | 3.2     |      |
|      | 2      | 62   | Male     | 7.3             | 4.1              | 490                  | 60                       | 68                         | 7.0            | 3.3     |      |
|      | 3      | 58   | Male     | 1.0             | 0.4              | 182                  | 14                       | 20                         | 6.8            | 3.4     |      |
|      | 4      | 72   | Male     | 3.9             | 2.0              | 195                  | 27                       | 59                         | 7.3            | 2.4     |      |
|      |        |      |          |                 |                  |                      |                          |                            |                |         |      |
|      | 578    | 60   | Male     | 0.5             | 0.1              | 500                  | 20                       | 34                         | 5.9            | 1.6     |      |
|      | 579    | 40   | Male     | 0.6             | 0.1              | 98                   | 35                       | 31                         | 6.0            | 3.2     |      |
|      | 580    | 52   | Male     | 0.8             | 0.2              | 245                  | 48                       | 49                         | 6.4            | 3.2     |      |
|      | 581    | 31   | Male     | 1.3             | 0.5              | 184                  | 29                       | 32                         | 6.8            | 3.4     |      |
|      | 582    | 38   | Male     | 1.0             | 0.3              | 216                  | 21                       | 24                         | 7.3            | 4.4     |      |
| Ę    | 583 rc | ws × | 11 colur | nns             |                  |                      |                          |                            |                |         |      |
| <    |        |      |          |                 |                  |                      |                          |                            |                |         | )    |

#### df.shape

(583, 11)

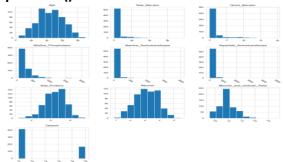
#### df.columns

## df.head()

| In [1: | 16]: | df.head() |        |                 |                  |                      |                          |                            |                |         |             |  |  |  |
|--------|------|-----------|--------|-----------------|------------------|----------------------|--------------------------|----------------------------|----------------|---------|-------------|--|--|--|
| Out[1  | 16]: | ge        | Gender | Total_Bilirubin | Direct_Bilirubin | Alkaline_Phosphotase | Alamine_Aminotransferase | Aspartate_Aminotransferase | Total_Protiens | Albumin | Albumin_and |  |  |  |
|        |      | 65        | Female | 0.7             | 0.1              | 187                  | 16                       | 18                         | 6.8            | 3.3     |             |  |  |  |
|        |      | 62        | Male   | 10.9            | 5.5              | 699                  | 64                       | 100                        | 7.5            | 3.2     |             |  |  |  |
|        |      | 62        | Male   | 7.3             | 4.1              | 490                  | 60                       | 68                         | 7.0            | 3.3     |             |  |  |  |
|        |      | 58        | Male   | 1.0             | 0.4              | 182                  | 14                       | 20                         | 6.8            | 3.4     |             |  |  |  |
|        |      | 72        | Male   | 3.9             | 2.0              | 195                  | 27                       | 59                         | 7.3            | 2.4     |             |  |  |  |
|        |      | <         |        |                 |                  |                      |                          |                            |                |         | >           |  |  |  |

Exploratory analysis filtering categorical data df.dtypes[df.dtypes=='object']

# Distribution of Numerical Features df.hist(figsize=(15,15), xrot=-45, bins=10) plt.show()



#### df.describe()

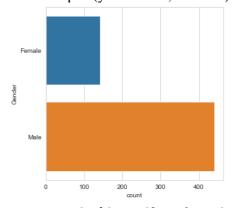
|       | cribe()    |                 |                  |                      |                          |                            |                |            |     |
|-------|------------|-----------------|------------------|----------------------|--------------------------|----------------------------|----------------|------------|-----|
|       | Age        | Total_Bilirubin | Direct_Bilirubin | Alkaline_Phosphotase | Alamine_Aminotransferase | Aspartate_Aminotransferase | Total_Protiens | Albumin    | ΑII |
| count | 583.000000 | 583.000000      | 583.000000       | 583.000000           | 583.000000               | 583.000000                 | 583.000000     | 583.000000 |     |
| mean  | 44.746141  | 3.298799        | 1.486106         | 290.576329           | 80.713551                | 109.910806                 | 6.483190       | 3.141852   |     |
| std   | 16.189833  | 6.209522        | 2.808498         | 242.937989           | 182.620356               | 288.918529                 | 1.085451       | 0.795519   |     |
| min   | 4.000000   | 0.400000        | 0.100000         | 63.000000            | 10.000000                | 10.000000                  | 2.700000       | 0.900000   |     |
| 25%   | 33.000000  | 0.800000        | 0.200000         | 175.500000           | 23.000000                | 25.000000                  | 5.800000       | 2.600000   |     |
| 50%   | 45.000000  | 1.000000        | 0.300000         | 208.000000           | 35.000000                | 42.000000                  | 6.600000       | 3.100000   |     |
| 75%   | 58.000000  | 2.600000        | 1.300000         | 298.000000           | 60.500000                | 87.000000                  | 7.200000       | 3.800000   |     |
| max   | 90.000000  | 75.000000       | 19.700000        | 2110.000000          | 2000.000000              | 4929.000000                | 9.600000       | 5.500000   |     |

Dataset i.e output value has '1' for liver disease and '2' for no liver disease so let's make it 0 for no disease to make it convineint

def partition(x):

if x == 2: return 0 return 1

df['Dataset'] = df['Dataset'].map(partition)
plt.figure(figsize=(5,5))
sns.countplot(y='Gender', data=df)

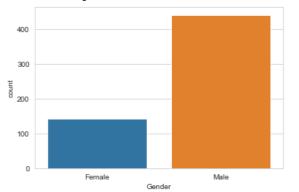


sns.countplot(data=df, x = 'Gender', label='Count')

M, F = df['Gender'].value\_counts()
print('Number of patients that are male: ',M)
print('Number of patients that are female: ',F)

Number of patients that are male: 441

Number of patients that are female: 142



Label Male as 0 and Female as 1

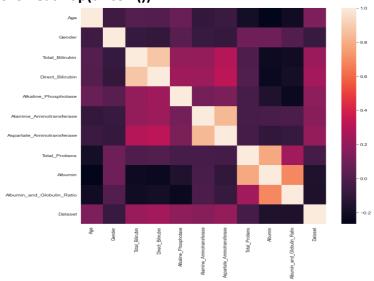
def partition(x):
 if x =='Male':
 return 0
 return 1

df['Gender'] = df['Gender'].map(partition)

df.corr()

plt.figure(figsize=(10,10))

sns.heatmap(df.corr())



Data cleaning

```
df = df.drop_duplicates()
print( df.shape )
  (564, 11)
Removing outliers
df = df[df.Aspartate_Aminotransferase <=3000 ]
df.shape
  (564, 11)
df = df[df.Aspartate_Aminotransferase <=2500 ]
df.shape
  (564, 11)
Dropping null values
df=df.dropna(how='any')</pre>
```

(564, 11)

df.shape