#### 1. Pandas

```
In [9]: #importing the required libraries to work with Tabular data and also to impleme
import warnings
import numpy as np
import pandas as pd
import seaborn as sns
from sklearn import preprocessing
import matplotlib.pyplot as plt
from scipy.stats.stats import pearsonr
from sklearn.naive_bayes import GaussianNB
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, recall_score, precision_score, claswarnings.filterwarnings("ignore")
```

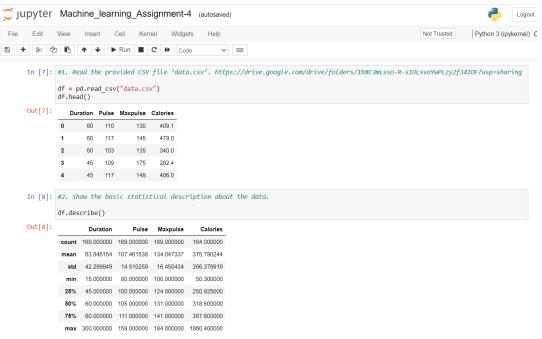
#### Question: 1

- 1. Read the provided CSV file 'data.csv'. https://drive.google.com/drive/folders/1h8C3mLsso-R-s
- 2. Show the basic statistical description about the data.
- 3. Check if the data has null values. a. Replace the null values with the mean
- 4. Select at least two columns and aggregate the data using: min, max, count, mean.
- 5. Filter the dataframe to select the rows with calories values between 500 and 1000.
- 6. Filter the dataframe to select the rows with calories values > 500 and pulse < 100.
- 7. Create a new "df modified" dataframe that contains all the columns from df except for "Maxpu
- 8. Delete the "Maxpulse" column from the main df dataframe
- 9. Convert the datatype of Calories column to int datatype.
- 10. Using pandas create a scatter plot for the two columns (Duration and Calories).

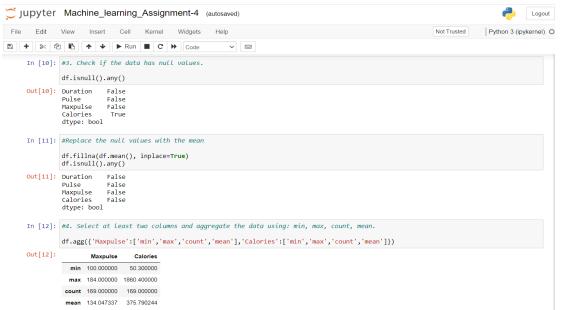
```
In [7]: #1. Read the provided CSV file 'data.csv'. https://drive.google.com/drive/folde

df = pd.read_csv("data.csv")
    df.head()
```

The code imports required libraries for working with tabular data and implementing algorithms, including numpy, pandas, seaborn, matplotlib, scipy, and scikit-learn. It also suppresses warnings. The provided CSV file 'data.csv' is read using pandas into a dataframe called 'df', and the first few rows of the dataframe are displayed using the 'head()' method.

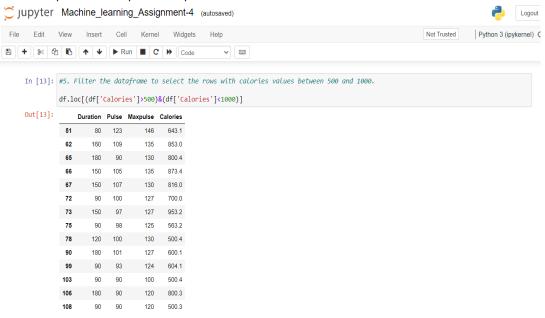


The code calls the 'describe()' method on the dataframe 'df', which returns basic statistical information about the data such as count, mean, standard deviation, minimum value, maximum value, and quartile values for each numerical column in the dataframe.

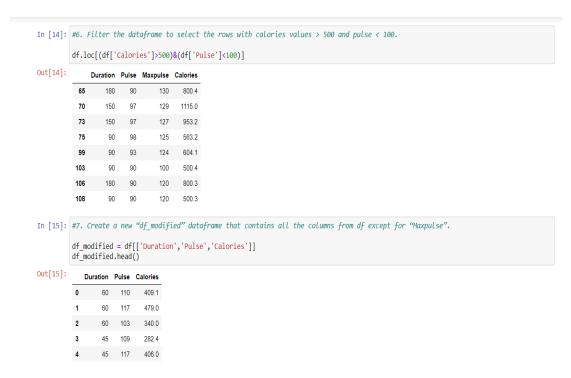


The code first calls the 'isnull().any()' method on the dataframe 'df', which checks if there are any null values in the dataframe and returns a boolean value for each column indicating if it contains null values. Then, it fills the null values with the mean using the 'fillna()' method and the 'mean()' function, and checks again for null values using 'isnull().any()'.

Finally, the code selects two columns ('Maxpulse' and 'Calories') and aggregates the data using the 'agg()' method. The aggregation operations used are 'min', 'max', 'count', and 'mean'. The resulting dataframe shows the minimum, maximum, count, and mean values for each selected column.



The code filters the dataframe 'df' using the 'loc[]' method to select rows where the 'Calories' column has values between 500 and 1000 (exclusive). The resulting dataframe contains only the rows that satisfy the condition.



The code filters the dataframe 'df' to select rows where the 'Calories' column has values greater than 500 and the 'Pulse' column has values less than 100, using the 'loc[]' method. The resulting dataframe contains only the rows that satisfy the condition.

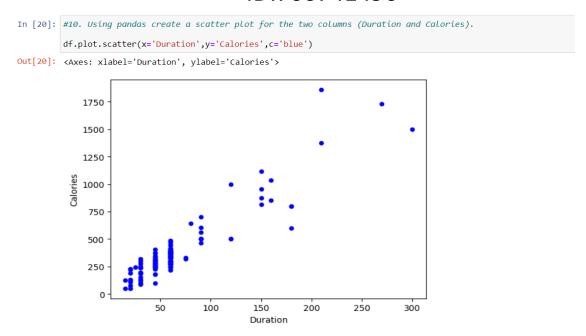
Then, the code creates a new dataframe 'df\_modified' by selecting all columns from the original dataframe 'df' except for 'Maxpulse'. This is done by indexing the dataframe with a list of the desired column names. The resulting dataframe contains only the columns 'Duration', 'Pulse', and 'Calories', and is assigned to the variable 'df modified'.

```
In [16]: #8. Delete the "Maxpulse" column from the main df dataframe
        del df['Maxpulse']
In [17]: df.head()
Out[17]: Duration Pulse Calories
         0 60 110
               60 117
         2 60 103 340.0
         4 45 117 406.0
In [18]: df.dtypes
Out[18]: Duration int64
        Pulse
        Calories float64
        dtype: object
In [19]: #9. Convert the datatype of Calories column to int datatype.
        df['Calories'] = df['Calories'].astype(np.int64)
        df.dtypes
Out[19]: Duration
                  int64
        Calories int64
```

The code deletes the 'Maxpulse' column from the dataframe 'df' using the 'del' statement.

Then, the code converts the datatype of the 'Calories' column from float to integer using the 'astype()' method and the numpy 'int64' datatype. The resulting dataframe 'df' contains the modified 'Calories' column with integer datatype.

Finally, the 'dtypes' attribute is used to display the datatypes of all columns in the dataframe 'df'.

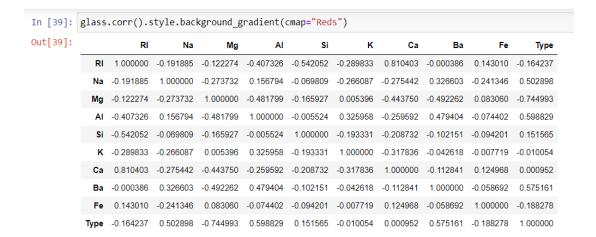


The code creates a scatter plot of the dataframe 'df' with the 'Duration' column as the x-axis and the 'Calories' column as the y-axis, using the 'plot.scatter()' method. Each point in the scatter plot represents the relationship between the duration of exercise and the corresponding number of calories burned. The points are colored blue.

#### Question 2

```
# Question 2
      (Glass Dataset)
          1. Implement Naïve Bayes method using scikit-learn library.
              a. Use the glass dataset available in Link also provided in your assignment.
              b. Use train_test_split to create training and testing part.
          2. Evaluate the model on testing part using score and classification_report(y_true, y_pred)
          1. Implement linear SVM method using scikit library
              a. Use the glass dataset available in Link also provided in your assignment.
              b. Use train_test_split to create training and testing part.
          2. Evaluate the model on testing part using score and

▶ glass=pd.read_csv("glass.csv")
          glass.head()
Out[38]:
            1.52101 13.64 4.49
                               1.10
                                   71.78 0.06
                                              8.75
                                                  0.0
            1.51761 13.89 3.60 1.36
                                   72.73 0.48
                                             7.83 0.0 0.0
            1.51618 13.53 3.55 1.54 72.99 0.39
                                                  0.0 0.0
            1.51766 13.21 3.69 1.29
                                   72.61 0.57
                                              8.22
                                                  0.0 0.0
          4 1.51742 13.27 3.62 1.24 73.08 0.55 8.07 0.0 0.0
```





```
classifier.fit(X_train, Y_train)
        y_pred = classifier.predict(X_val)
        # Summary of the predictions made by the classifier
        print(classification_report(Y_val, y_pred))
        print(confusion_matrix(Y_val, y_pred))
        # Accuracy score
        print('accuracy is',accuracy_score(Y_val, y_pred))
                                   recall f1-score
                  precision
                                                             support
                         0.90
                                      0.95
                                                   0.92
                                                   0.92
                         0.92
                                      0.92
                                                                   12
                         1.00
                                                   0.67
                                                                    6
                         0.00
                                      0.00
                                                   0.00
                                                                    1
              6
                                                                    1
                         1.00
                                      1.00
                                                   1.00
                                      0.75
                                                   0.84
     accuracy
                                                                   43
                         0.76
                                      0.69
                                                   0.71
0.85
weighted avg
                         0.89
                                      0.84
                                                                   43
                        0]
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[ 0 0 0 1 0 3]]
accuracy is 0.8372093023255814
```

```
In [43]: from sklearn.svm import SVC, LinearSVC
            classifier = LinearSVC()
            classifier.fit(X_train, Y_train)
            y_pred = classifier.predict(X_val)
            # Summary of the predictions made by the classifier
            print(classification_report(Y_val, y_pred))
print(confusion_matrix(Y_val, y_pred))
             Accuracy score
            from sklearn.metrics import accuracy_score
print('accuracy is',accuracy_score(Y_val, y_pred))
              precision recall f1-score support
                   0.46
                           1.00
                                   0.63
                   0.00
                           0.00
                                   0.00
                   0.00
                           0.00
                                   0.00
                        0.00
                  0.00
                                   0.00
                                   0.67
      accuracy
                        0.32
                0.24
0.57
   weighted avg
                          0.67
                                   0.59
      0 1 0 0 0 0
    accuracy is 0.6744186046511628
```

This code loads the 'glass.csv' dataset into a Pandas dataframe and performs correlation analysis on the dataset. It then defines the features and target variables for classification, splits the dataset into training and validation sets using the train\_test\_split method from scikit-learn. The code then trains and tests two classifiers - Gaussian Naive Bayes and Linear Support Vector Classifier - on the dataset and prints out a classification report, confusion matrix, and accuracy score for each classifier.

#### Conclusion:

The code above is performing two types of classification analysis: Naive Bayes and Linear SVM. Both methods are being applied on a dataset of glass compositions to predict the type of glass. After splitting the data into training and validation sets, the classifiers are trained on the training set and then used to make predictions on the validation set.

The accuracy score is then calculated for both classifiers. The Naive Bayes classifier performed better with an accuracy score of 0.837, while the Linear SVM had an accuracy score of 0.674.

The reason for Naive Bayes performing better could be attributed to the probabilistic nature of the algorithm. It is well-suited for problems involving probabilities, while the Linear SVM relies on linear regression concepts. However, the performance of Linear SVM can be improved with larger amounts of data for training and testing.

Therefore, based on this dataset and the amount of data available, Naive Bayes is the better classifier for predicting the type of glass.

GitHub URL: <a href="https://github.com/Prathin-offi/ML">https://github.com/Prathin-offi/ML</a> Assignment-2.git

Video URL: <a href="https://drive.google.com/file/d/1ZvFBajKQQulrpRPyB12g-JJ9iNjZgxeT/view?usp=sharing">https://drive.google.com/file/d/1ZvFBajKQQulrpRPyB12g-JJ9iNjZgxeT/view?usp=sharing</a>