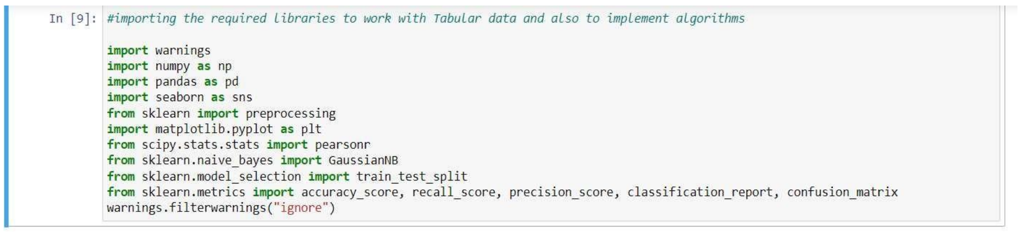
ASSIGNMENT - 4

Github link: https://github.com/tanikonda7441/Assignment4

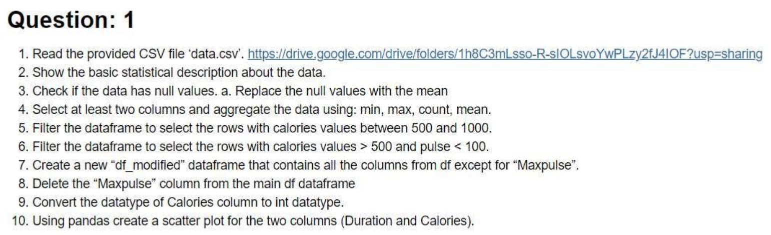
1. Pandas

Importing all the required libraries to work with tabular data and also implement algorithms.



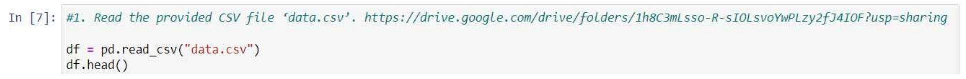
```
In [9]: #importing the required libraries to work with Tabular data and also to implement algorithms

        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, classification_report, confusion_matrix
        warnings.filterwarnings("ignore")
```



Question: 1

1. Read the provided CSV file 'data.csv'. https://drive.google.com/drive/folders/1h8C3mLsso-R-sIOLsvoYwPLzy2fJ4IOF?usp=sharing
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 "Maxpulse".
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).



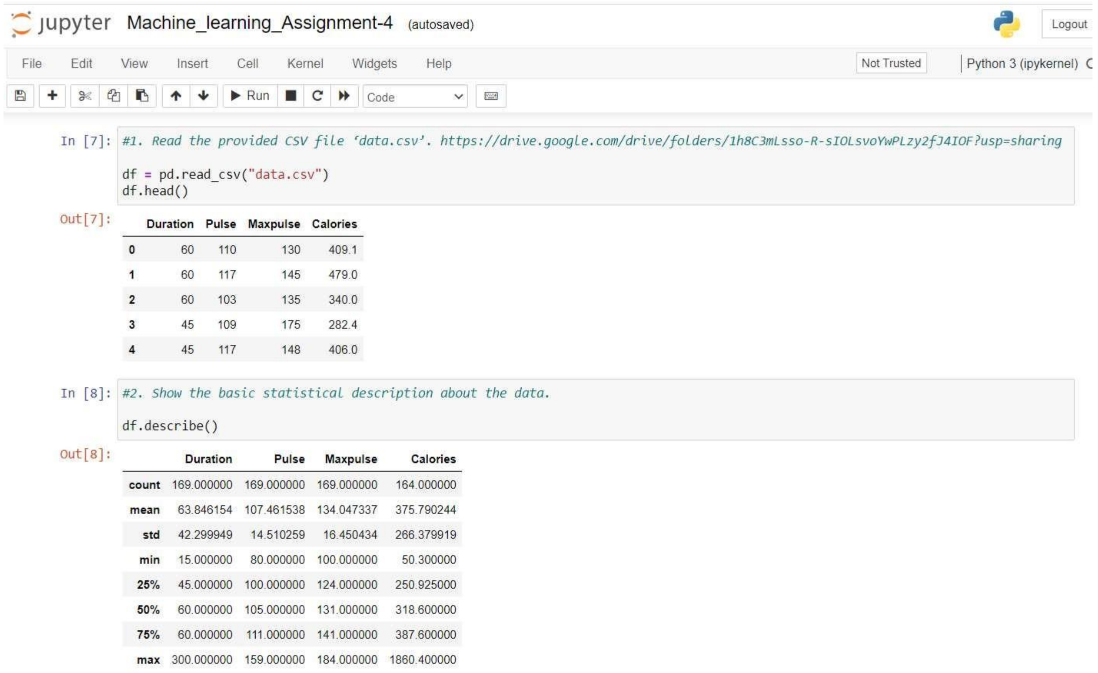
```
In [7]: #1. Read the provided CSV file 'data.csv'. https://drive.google.com/drive/folders/1h8C3mLsso-R-sIOLsvoYwPLzy2fJ4IOF?usp=sharing

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

using the Pandas library to read a CSV file named "data.csv" and store its contents in a DataFrame object called df.

The pd.read_csv() function is a method provided by the Pandas library that reads the CSV file and creates a DataFrame object from it.

The df.head() function is then called to display the first five rows of the DataFrame.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Python 3 (ipykernel)
Run Code

In [7]:
```
#1. Read the provided CSV file 'data.csv'. https://drive.google.com/drive/folders/1h8C3mLsso-R-sIOLsvoYWPLzy2fJ4IOF?usp=sharing

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

Out[7]:

	Duration	Pulse	Maxpulse	Calories
0	60	110	130	409.1
1	60	117	145	479.0
2	60	103	135	340.0
3	45	109	175	282.4
4	45	117	148	406.0

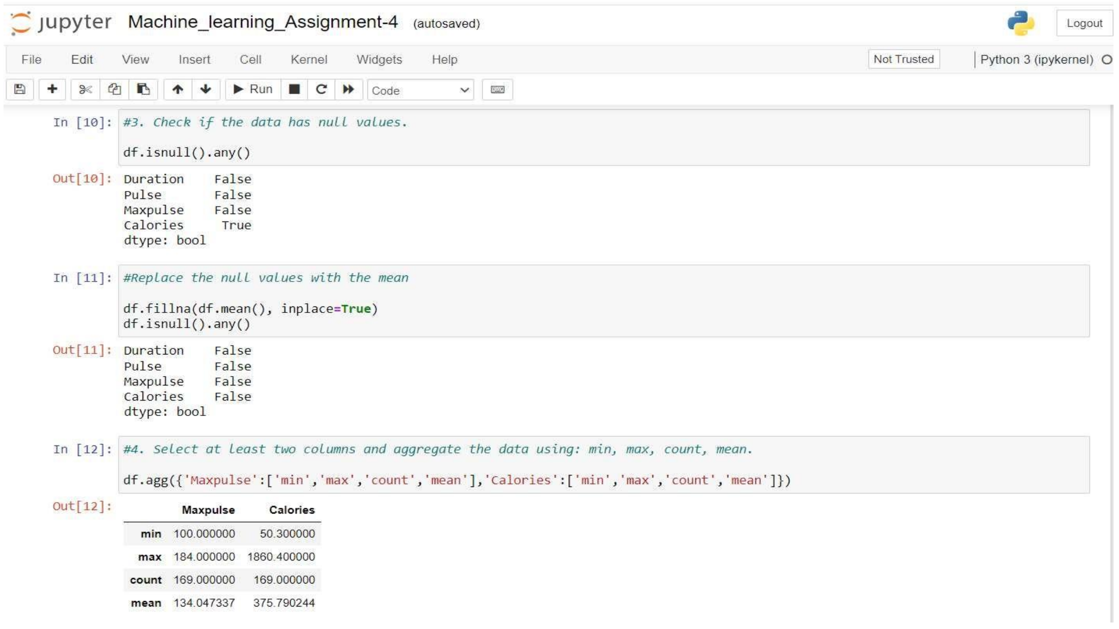
In [8]:
```
#2. Show the basic statistical description about the data.

df.describe()
```

Out[8]:

	Duration	Pulse	Maxpulse	Calories
count	169.000000	169.000000	169.000000	164.000000
mean	63.846154	107.461538	134.047337	375.790244
std	42.299949	14.510259	16.450434	266.379919
min	15.000000	80.000000	100.000000	50.300000
25%	45.000000	100.000000	124.000000	250.925000
50%	60.000000	105.000000	131.000000	318.600000
75%	60.000000	111.000000	141.000000	387.600000
max	300.000000	159.000000	184.000000	1860.400000

The df.describe() method is a built-in function in Pandas that generates descriptive statistics of the DataFrame df. It includes the count, mean, standard deviation, minimum, 25th percentile, 50th percentile (median), 75th percentile, and maximum values for each numeric column in the dataframe.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Python 3 (ipykernel)
Run Code

In [10]:
```
#3. Check if the data has null values.

df.isnull().any()
```

Out[10]:
```
Duration     False
Pulse        False
Maxpulse     False
Calories      True
dtype: bool
```

In [11]:
```
#Replace the null values with the mean

df.fillna(df.mean(), inplace=True)
df.isnull().any()
```

Out[11]:
```
Duration     False
Pulse        False
Maxpulse     False
Calories     False
dtype: bool
```

In [12]:
```
#4. Select at least two columns and aggregate the data using: min, max, count, mean.

df.agg({'Maxpulse':['min','max','count','mean'],'Calories':['min','max','count','mean']})
```

Out[12]:

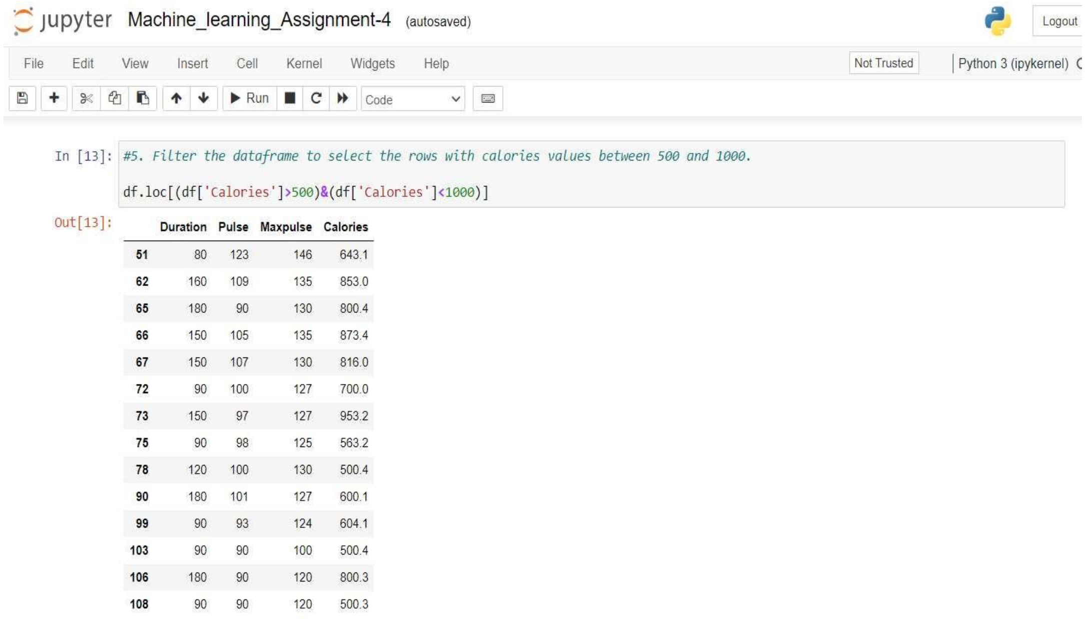
	Maxpulse	Calories
min	100.000000	50.300000
max	184.000000	1860.400000
count	169.000000	169.000000
mean	134.047337	375.790244

The df.isnull().any() code checks whether there are any missing values (also known as NaN or null values) in each column of the DataFrame df.

The first line of this code fills in any missing values in the DataFrame df with the mean value of each column using the fillna() method.

The second line of this code uses df.isnull().any() to check if there are any missing values remaining in the DataFrame after filling in the missing values with the mean.

This code uses the agg() method to perform aggregate calculations on the Maxpulse and Calories columns of the DataFrame df.


jupyter Machine_learning_Assignment-4 (autosaved)
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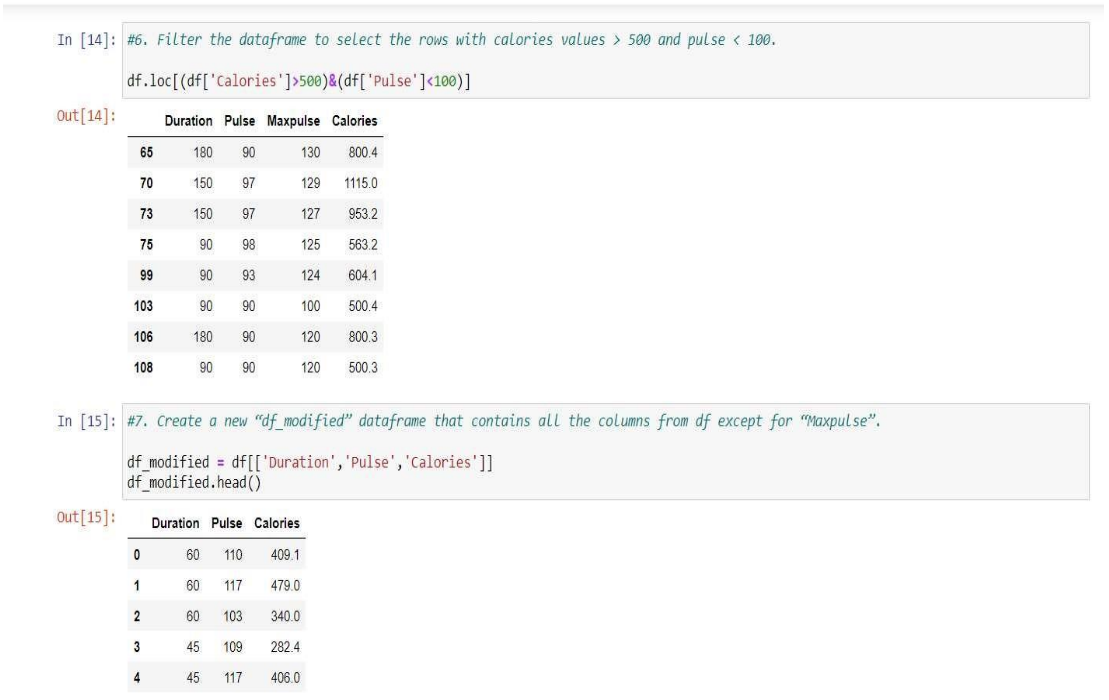
Python 3 (ipykernel)
Run
Code

```
In [13]: #5. Filter the dataframe to select the rows with calories values between 500 and 1000.

         df.loc[(df['Calories']>500)&(df['Calories']<1000)]
```

Out[13]:

	Duration	Pulse	Maxpulse	Calories
51	80	123	146	643.1
62	160	109	135	853.0
65	180	90	130	800.4
66	150	105	135	873.4
67	150	107	130	816.0
72	90	100	127	700.0
73	150	97	127	953.2
75	90	98	125	563.2
78	120	100	130	500.4
90	180	101	127	600.1
99	90	93	124	604.1
103	90	90	100	500.4
106	180	90	120	800.3
108	90	90	120	500.3

This code uses the loc[] method of the DataFrame df to select rows where the value in the Calories column is greater than 500 and less than 1000.



```
In [14]: #6. Filter the dataframe to select the rows with calories values > 500 and pulse < 100.

         df.loc[(df['Calories']>500)&(df['Pulse']<100)]
```

Out[14]:

	Duration	Pulse	Maxpulse	Calories
65	180	90	130	800.4
70	150	97	129	1115.0
73	150	97	127	953.2
75	90	98	125	563.2
99	90	93	124	604.1
103	90	90	100	500.4
106	180	90	120	800.3
108	90	90	120	500.3

```
In [15]: #7. Create a new “df_modified” dataframe that contains all the columns from df except for “Maxpulse”.

         df_modified = df[['Duration','Pulse','Calories']]
         df_modified.head()
```

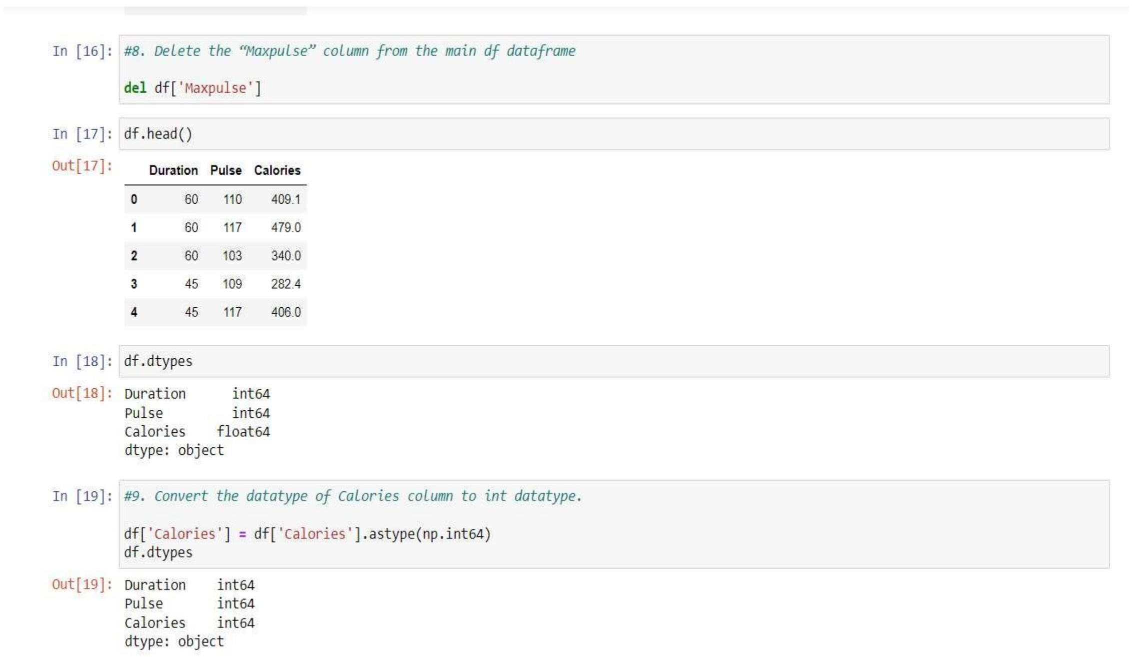
Out[15]:

	Duration	Pulse	Calories
0	60	110	409.1
1	60	117	479.0
2	60	103	340.0
3	45	109	282.4
4	45	117	406.0

This code uses the loc[] method of the DataFrame df to select rows where the value in the Calories column is greater than 500 and the value in the Pulse column is less than 100.

This code creates a new DataFrame df_modified that includes only the Duration, Pulse, and Calories columns of the original DataFrame df.

The head() method is then called on the df_modified DataFrame to display the first five rows of the new DataFrame.



```
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	409.1
1	60	117	479.0
2	60	103	340.0
3	45	109	282.4
4	45	117	406.0

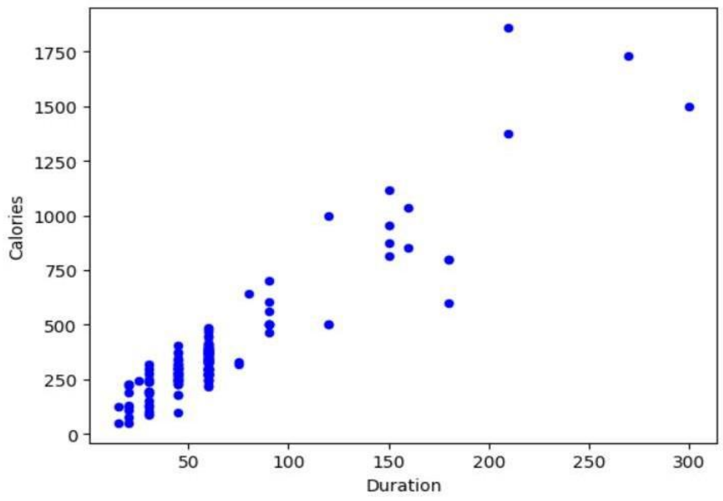
```
In [18]: df.dtypes
Out[18]: Duration      int64
         Pulse         int64
         Calories    float64
         dtype: object

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
         Pulse       int64
         Calories    int64
         dtype: object
```

After running this code, the Maxpulse column will no longer be present in the DataFrame df. If the column was present in the original DataFrame df, then it has now been removed permanently. The resulting DataFrame will have one less column than the original DataFrame.

The astype() method is a Pandas method that is used to cast a column of a DataFrame to a specified data type. In this case, the Calories column is being cast to the 64-bit integer data type using the np.int64 NumPy data type.

The dtypes attribute of the DataFrame is then used to display the data types of each column in the DataFrame. After running this code, the Calories column will have a data type of int64.

```
In [20]: #10. Using pandas create a scatter plot for the two columns (Duration and Calories).

         df.plot.scatter(x='Duration',y='Calories',c='blue')

Out[20]: <Axes: xlabel='Duration', ylabel='Calories'>
```



a scatter plot of the Duration and Calories columns in the DataFrame df will be displayed, where each data point is represented by a blue marker. The x-axis will correspond to the Duration column values and the y-axis to the Calories column values.

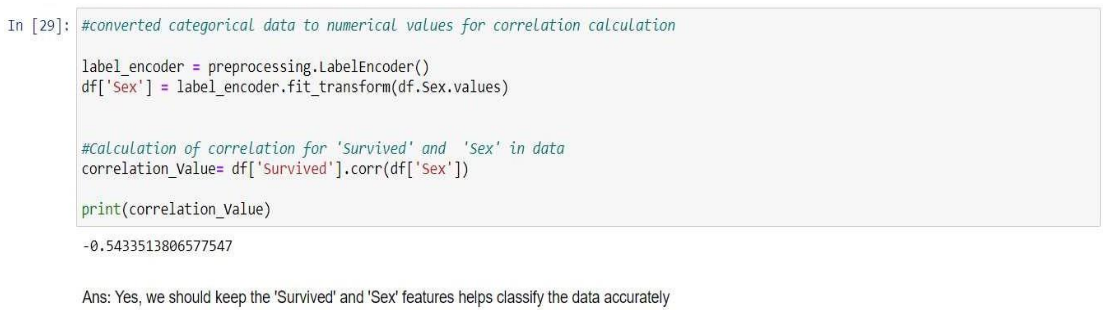
Question: 2

Titanic Dataset

1. Find the correlation between 'survived' (target column) and 'sex' column for the Titanic use case in class. a. Do you think we should keep this feature?
2. Do at least two visualizations to describe or show correlations.
3. Implement Naïve Bayes method using scikit-learn library and report the accuracy

```
In [28]: #Loading the data file into te program
         df=pd.read_csv("train.csv")

         df.head()

Out[28]:
```

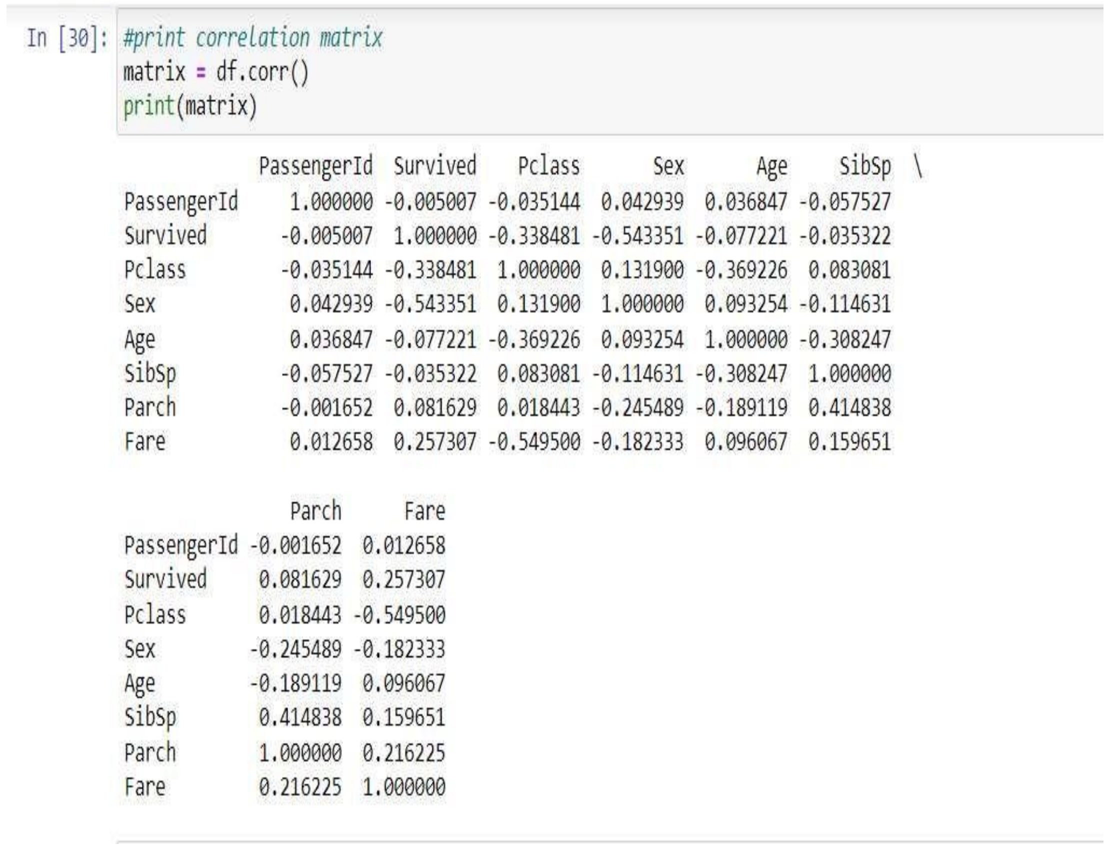
	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833	C85	C
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	S
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	S
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	S



```
In [29]: #converted categorical data to numerical values for correlation calculation

         label_encoder = preprocessing.LabelEncoder()
         df['Sex'] = label_encoder.fit_transform(df.Sex.values)


         #Calculation of correlation for 'Survived' and  'Sex' in data
         correlation_Value= df['Survived'].corr(df['Sex'])

         print(correlation_Value)
```

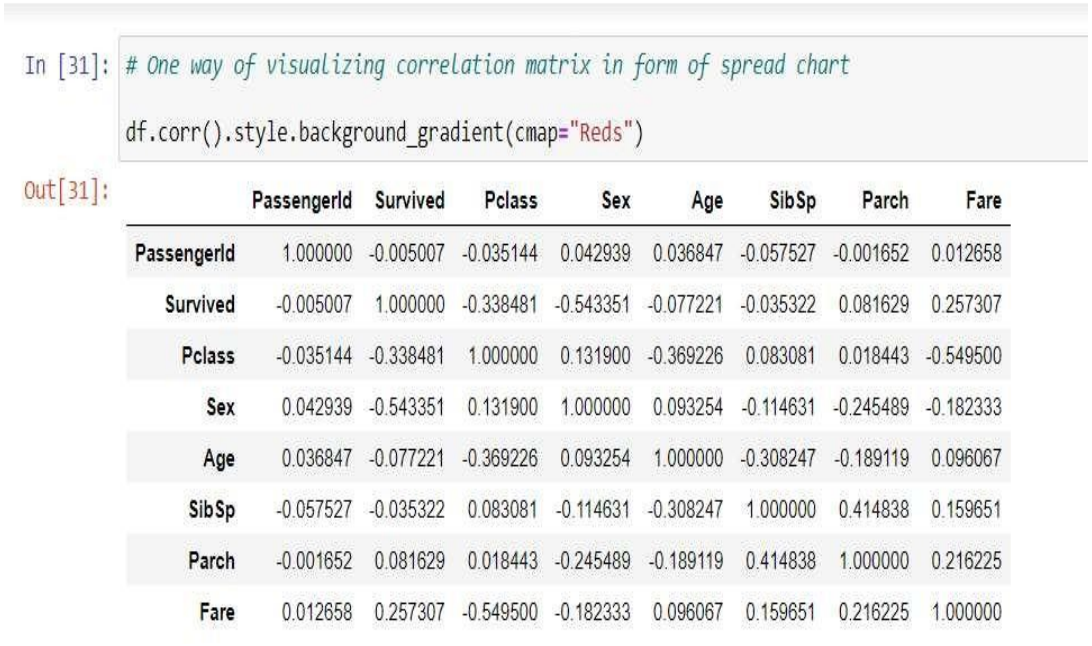
```
-0.5433513806577547
```

Ans: Yes, we should keep the 'Survived' and 'Sex' features helps classify the data accurately



```
In [30]: #print correlation matrix
         matrix = df.corr()
         print(matrix)
```

```
             PassengerId  Survived    Pclass       Sex       Age     SibSp  \
PassengerId     1.000000 -0.005007 -0.035144  0.042939  0.036847 -0.057527
Survived       -0.005007  1.000000 -0.338481 -0.543351 -0.077221 -0.035322
Pclass         -0.035144 -0.338481  1.000000  0.131900 -0.369226  0.083081
Sex             0.042939 -0.543351  0.131900  1.000000  0.093254 -0.114631
Age             0.036847 -0.077221 -0.369226  0.093254  1.000000 -0.308247
SibSp          -0.057527 -0.035322  0.083081 -0.114631 -0.308247  1.000000
Parch          -0.001652  0.081629  0.018443 -0.245489 -0.189119  0.414838
Fare            0.012658  0.257307 -0.549500 -0.182333  0.096067  0.159651

                Parch      Fare
PassengerId -0.001652  0.012658
Survived     0.081629  0.257307
Pclass       0.018443 -0.549500
Sex         -0.245489 -0.182333
Age         -0.189119  0.096067
SibSp        0.414838  0.159651
Parch        1.000000  0.216225
Fare         0.216225  1.000000
```

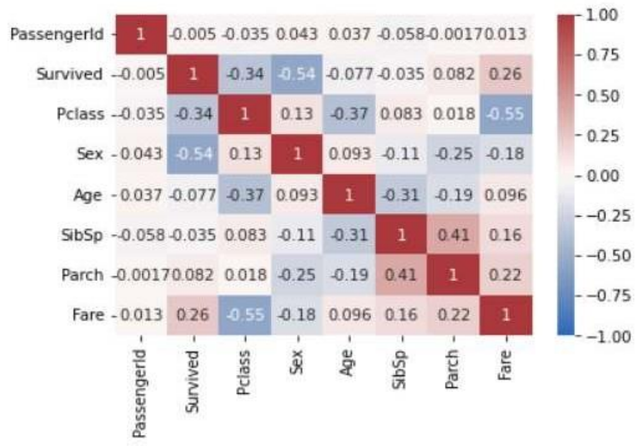


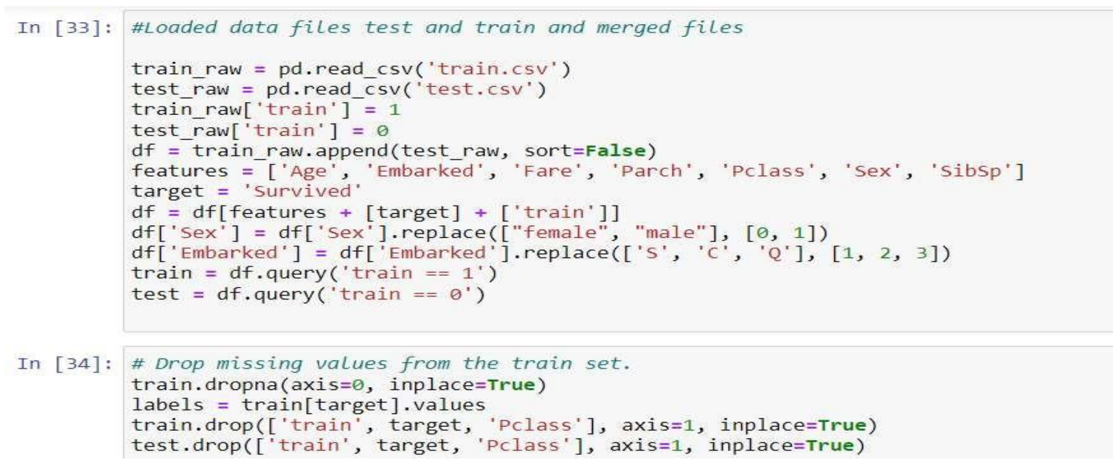
```
In [31]: # One way of visualizing correlation matrix in form of spread chart

         df.corr().style.background_gradient(cmap="Reds")
```

Out[31]:

	PassengerId	Survived	Pclass	Sex	Age	SibSp	Parch	Fare
PassengerId	1.000000	-0.005007	-0.035144	0.042939	0.036847	-0.057527	-0.001652	0.012658
Survived	-0.005007	1.000000	-0.338481	-0.543351	-0.077221	-0.035322	0.081629	0.257307
Pclass	-0.035144	-0.338481	1.000000	0.131900	-0.369226	0.083081	0.018443	-0.549500
Sex	0.042939	-0.543351	0.131900	1.000000	0.093254	-0.114631	-0.245489	-0.182333
Age	0.036847	-0.077221	-0.369226	0.093254	1.000000	-0.308247	-0.189119	0.096067
SibSp	-0.057527	-0.035322	0.083081	-0.114631	-0.308247	1.000000	0.414838	0.159651
Parch	-0.001652	0.081629	0.018443	-0.245489	-0.189119	0.414838	1.000000	0.216225
Fare	0.012658	0.257307	-0.549500	-0.182333	0.096067	0.159651	0.216225	1.000000

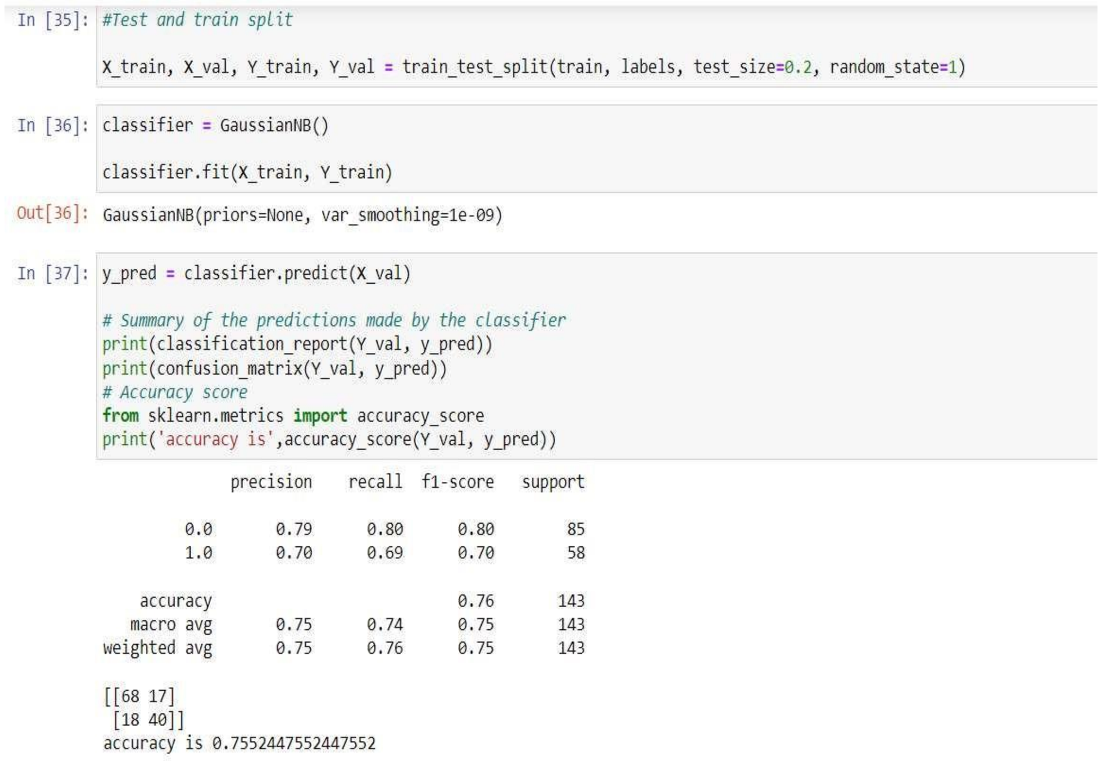
```
In [32]: #Second form of visuaizing correlation matriX using heatmap() from seaborn

         sns.heatmap(matrix, annot=True, vmax=1, vmin=-1, center=0, cmap='vlag')
         plt.show()
```





```
In [33]: #Loaded data files test and train and merged files

         train_raw = pd.read_csv('train.csv')
         test_raw = pd.read_csv('test.csv')
         train_raw['train'] = 1
         test_raw['train'] = 0
         df = train_raw.append(test_raw, sort=False)
         features = ['Age', 'Embarked', 'Fare', 'Parch', 'Pclass', 'Sex', 'SibSp']
         target = 'Survived'
         df = df[features + [target] + ['train']]
         df['Sex'] = df['Sex'].replace(["female", "male"], [0, 1])
         df['Embarked'] = df['Embarked'].replace(['S', 'C', 'Q'], [1, 2, 3])
         train = df.query('train == 1')
         test = df.query('train == 0')
```

```
In [34]: # Drop missing values from the train set.
         train.dropna(axis=0, inplace=True)
         labels = train[target].values
         train.drop(['train', target, 'Pclass'], axis=1, inplace=True)
         test.drop(['train', target, 'Pclass'], axis=1, inplace=True)
```

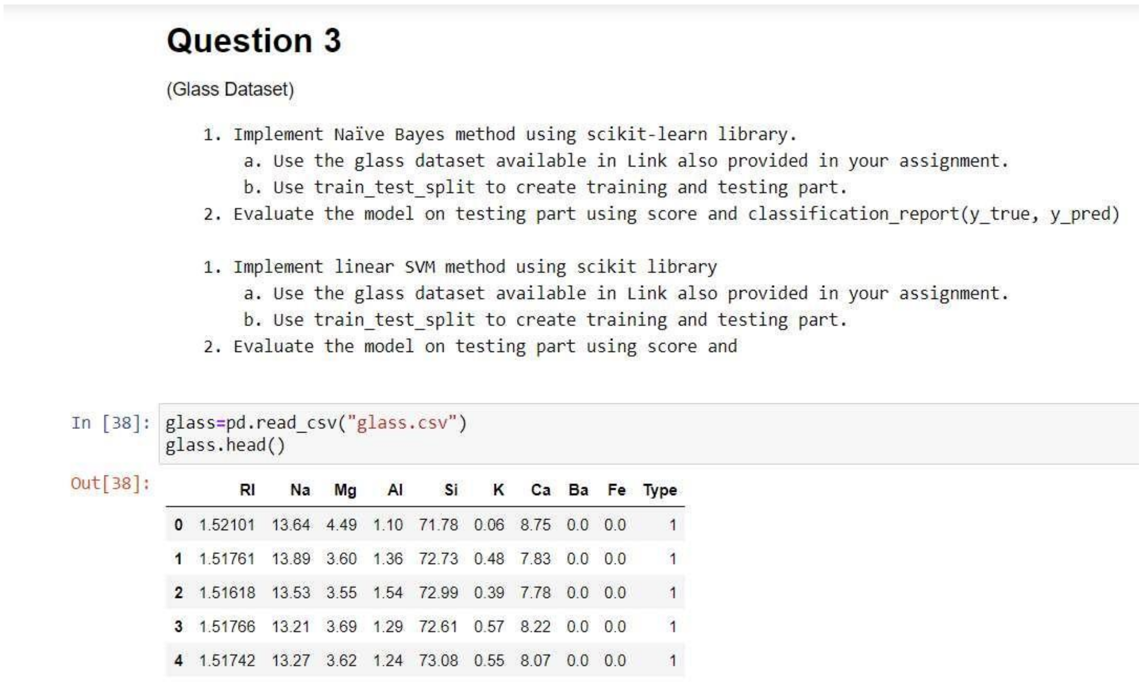


```
In [35]: #Test and train split

         X_train, X_val, Y_train, Y_val = train_test_split(train, labels, test_size=0.2, random_state=1)
```

```
In [36]: classifier = GaussianNB()

         classifier.fit(X_train, Y_train)

Out[36]: GaussianNB(priors=None, var_smoothing=1e-09)
```

```
In [37]: 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

         0.0       0.79      0.80      0.80        85
         1.0       0.70      0.69      0.70        58

    accuracy                           0.76       143
   macro avg       0.75      0.74      0.75       143
weighted avg       0.75      0.76      0.75       143

[[68 17]
 [18 40]]
accuracy is 0.7552447552447552
```



Question 3

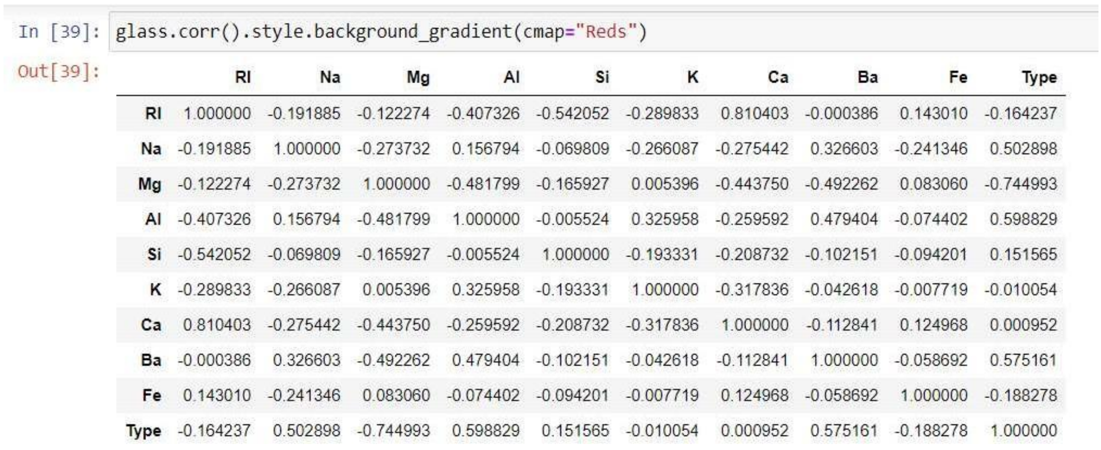
(Glass Dataset)

```
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
```

In [38]:
```
glass=pd.read_csv("glass.csv")
glass.head()
```

Out[38]:

	RI	Na	Mg	Al	Si	K	Ca	Ba	Fe	Type
0	1.52101	13.64	4.49	1.10	71.78	0.06	8.75	0.0	0.0	1
1	1.51761	13.89	3.60	1.36	72.73	0.48	7.83	0.0	0.0	1
2	1.51618	13.53	3.55	1.54	72.99	0.39	7.78	0.0	0.0	1
3	1.51766	13.21	3.69	1.29	72.61	0.57	8.22	0.0	0.0	1
4	1.51742	13.27	3.62	1.24	73.08	0.55	8.07	0.0	0.0	1

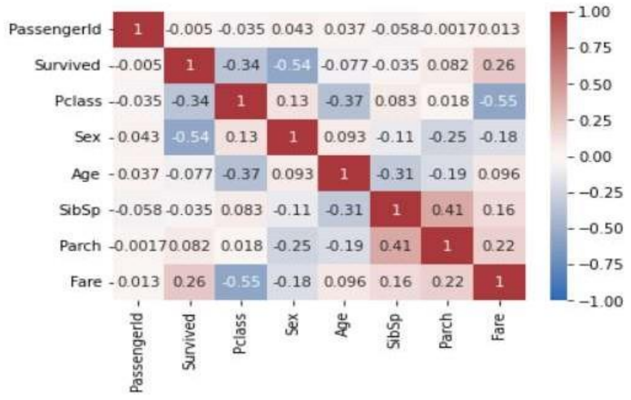


In [39]:
```
glass.corr().style.background_gradient(cmap="Reds")
```

Out[39]:

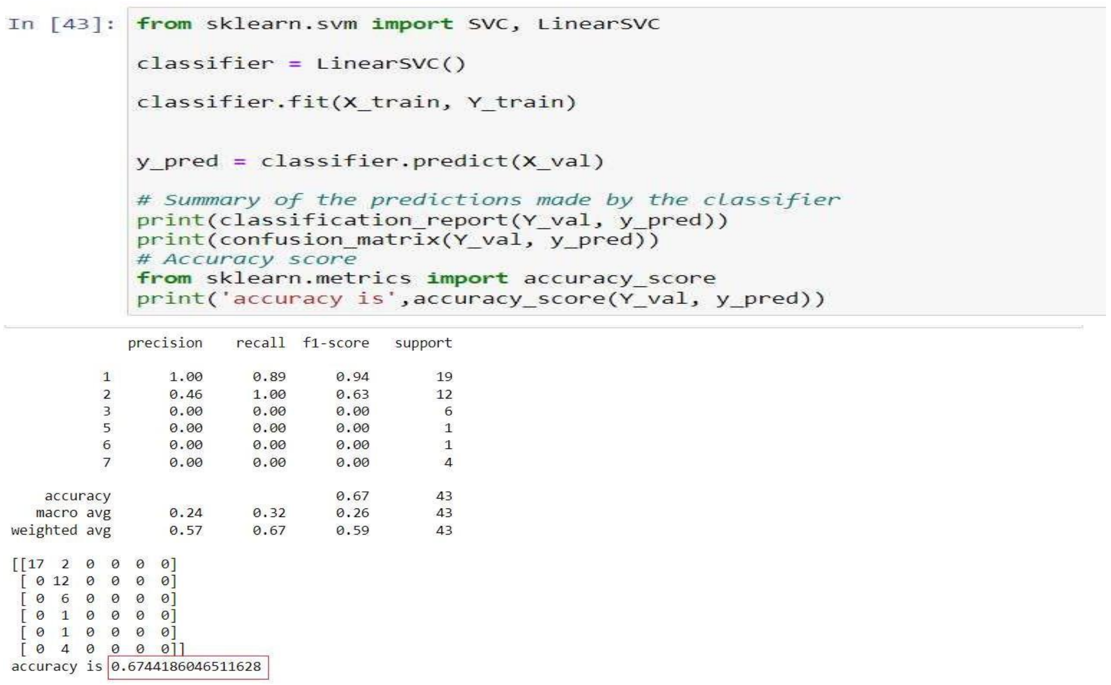
	RI	Na	Mg	Al	Si	K	Ca	Ba	Fe	Type
RI	1.000000	-0.191885	-0.122274	-0.407326	-0.542052	-0.289833	0.810403	-0.000386	0.143010	-0.164237
Na	-0.191885	1.000000	-0.273732	0.156794	-0.069809	-0.266087	-0.275442	0.326603	-0.241346	0.502898
Mg	-0.122274	-0.273732	1.000000	-0.481799	-0.165927	0.005396	-0.443750	-0.492262	0.083060	-0.744993
Al	-0.407326	0.156794	-0.481799	1.000000	-0.005524	0.325958	-0.259592	0.479404	-0.074402	0.598829
Si	-0.542052	-0.069809	-0.165927	-0.005524	1.000000	-0.193331	-0.208732	-0.102151	-0.094201	0.151565
K	-0.289833	-0.266087	0.005396	0.325958	-0.193331	1.000000	-0.317836	-0.042618	-0.007719	-0.010054
Ca	0.810403	-0.275442	-0.443750	-0.259592	-0.208732	-0.317836	1.000000	-0.112841	0.124968	0.000952
Ba	-0.000386	0.326603	-0.492262	0.479404	-0.102151	-0.042618	-0.112841	1.000000	-0.058692	0.575161
Fe	0.143010	-0.241346	0.083060	-0.074402	-0.094201	-0.007719	0.124968	-0.058692	1.000000	-0.188278
Type	-0.164237	0.502898	-0.744993	0.598829	0.151565	-0.010054	0.000952	0.575161	-0.188278	1.000000

```
In [40]: sns.heatmap(matrix, annot=True, vmax=1, vmin=-1, center=0, cmap='vlag')
         plt.show()
```



```
In [41]: features = ['Rl', 'Na', 'Mg', 'Al', 'Si', 'K', 'Ca', 'Ba', 'Fe']
         target = 'Type'


         X_train, X_val, Y_train, Y_val = train_test_split(glass[::-1], glass['Type'],test_size=0.2, random_state=1)

         classifier = GaussianNB()

         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))
```

```
              precision    recall  f1-score   support

           1       0.90      0.95      0.92        19
           2       0.92      0.92      0.92        12
           3       1.00      0.50      0.67         6
           5       0.00      0.00      0.00         1
           6       1.00      1.00      1.00         1
           7       0.75      0.75      0.75         4

    accuracy                           0.84        43
   macro avg       0.76      0.69      0.71        43
weighted avg       0.89      0.84      0.85        43

[[18  1  0  0  0  0]
 [ 1 11  0  0  0  0]
 [ 1  0  3  2  0  0]
 [ 0  0  0  0  0  1]
 [ 0  0  0  0  1  0]
 [ 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

           1       1.00      0.89      0.94        19
           2       0.46      1.00      0.63        12
           3       0.00      0.00      0.00         6
           5       0.00      0.00      0.00         1
           6       0.00      0.00      0.00         1
           7       0.00      0.00      0.00         4

    accuracy                           0.67        43
   macro avg       0.24      0.32      0.26        43
weighted avg       0.57      0.67      0.59        43

[[17  2  0  0  0  0]
 [ 0 12  0  0  0  0]
 [ 0  6  0  0  0  0]
 [ 0  1  0  0  0  0]
 [ 0  1  0  0  0  0]
 [ 0  4  0  0  0  0]]
accuracy is 0.6744186046511628
```

Justification:

We got better accuracy for Naïve Bayes method which is 0.8372093023255814. Naive Bayes analysis works well with probabilistic concepts where as Linear SVM works better with linear regression logics. But to perform more accurately SVM requires large amounts of data to train and test the data. So, due to the amount of data Naive Bayes algorith gives better accuracy compared to Linear SVM.