```
In [1]: import pandas as pd
          import numpy as np
          import seaborn as sns
          import matplotlib.pyplot as plt
          from sklearn.model_selection import train_test_split
          from sklearn.preprocessing import StandardScaler
          from sklearn.metrics import accuracy_score, confusion_matrix,classification_report
          from sklearn.naive_bayes import BernoulliNB
          from sklearn.naive_bayes import GaussianNB
          from sklearn.naive_bayes import MultinomialNB
 In [2]: data = pd.read_csv('D:\\24 - Machine_Learning\\download files\\glass.data', sep=',')
          data
                1 1.52101 13.64 4.49 1.10 71.78 0.06 8.75 0.00 0.00.1 1.1
 Out[2]:
            0 2 1.51761 13.89 3.60 1.36 72.73 0.48 7.83 0.00 0.00 1
            1 3 1.51618 13.53 3.55 1.54 72.99 0.39 7.78 0.00
            2 4 1.51766 13.21 3.69 1.29 72.61 0.57 8.22 0.00
                                                                0.00
               5 1.51742 13.27 3.62 1.24 73.08 0.55 8.07 0.00
                                                               0.00 1
                6 1.51596 12.79 3.61 1.62 72.97 0.64 8.07 0.00
                                                               0.26
                                                                    1
          208 210 1.51623 14.14 0.00 2.88 72.61 0.08 9.18 1.06
                                                                0.00 7
          209 211 1.51685 14.92 0.00 1.99 73.06 0.00 8.40 1.59
          210 212 1.52065 14.36 0.00 2.02 73.42 0.00 8.44 1.64
          211 213 1.51651 14.38 0.00 1.94 73.61 0.00 8.48 1.57
          212 214 1.51711 14.23 0.00 2.08 73.36 0.00 8.62 1.67 0.00 7
         213 rows × 11 columns
 In [3]: data.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 213 entries, 0 to 212
          Data columns (total 11 columns):
          # Column Non-Null Count Dtype
          --- ----- ------ -----
                         213 non-null
          0 1
                                         int64
          1
               1.52101 213 non-null
                                         float64
               13.64
                        213 non-null
                                         float64
               4.49
                        213 non-null
                                         float64
               1.10
                        213 non-null
                                         float64
              71.78
                        213 non-null
          5
                                         float64
                        213 non-null
          6 0.06
                                         float64
                        213 non-null
          7
               8.75
                                         float64
                        213 non-null
          8 0.00
                                         float64
           9 0.00.1 213 non-null
                                         float64
                        213 non-null
          10 1.1
                                         int64
          dtypes: float64(9), int64(2)
          memory usage: 18.4 KB
 In [4]: col_names = ['Id', 'RI', 'NA', 'Mg', 'Al', 'Si', 'K', 'Ca', 'Ba', 'Fe', 'Type']
 In [5]: data.columns = col_names
 In [6]: data.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 213 entries, 0 to 212
          Data columns (total 11 columns):
               Column Non-Null Count Dtype
              -----
                       213 non-null
           0
               Ιd
                                        int64
               RΙ
                       213 non-null
                                        float64
          1
                       213 non-null
                                        float64
               NA
           3
               Mg
                       213 non-null
                                         float64
           4
               Αl
                       213 non-null
                                        float64
                                         float64
               Si
                       213 non-null
               K
                       213 non-null
                                        float64
               Ca
                       213 non-null
                                        float64
               Ва
                       213 non-null
                                        float64
               Fe
                       213 non-null
                                        float64
           10 Type
                       213 non-null
                                        int64
          dtypes: float64(9), int64(2)
          memory usage: 18.4 KB
 In [7]: data
 Out[7]:
              2 1.51761 13.89 3.60 1.36 72.73 0.48 7.83 0.00
           1 3 1.51618 13.53 3.55 1.54 72.99 0.39 7.78 0.00 0.00
            2 4 1.51766 13.21 3.69 1.29 72.61 0.57 8.22 0.00 0.00
              5 1.51742 13.27 3.62 1.24 73.08 0.55 8.07 0.00 0.00
            4 6 1.51596 12.79 3.61 1.62 72.97 0.64 8.07 0.00 0.26
          208 210 1.51623 14.14 0.00 2.88 72.61 0.08 9.18 1.06 0.00
          209 211 1.51685 14.92 0.00 1.99 73.06 0.00 8.40 1.59 0.00
          210 212 1.52065 14.36 0.00 2.02 73.42 0.00 8.44 1.64 0.00
          211 213 1.51651 14.38 0.00 1.94 73.61 0.00 8.48 1.57 0.00
          212 214 1.51711 14.23 0.00 2.08 73.36 0.00 8.62 1.67 0.00
         213 rows × 11 columns
 In [8]: sns.heatmap(data.isna());
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          209
                                                       -0.100
               Id RI NA Mg Al Si K Ca Ba Fe Type
 In [9]: cor = data.corr()
          cor
                                                          ΑI
 Out[9]:
                     Id
                              RI
                                       NA
                                                Mg
                                                                   Si
                                                                                    Ca
                                                                                              Ba
                                                                                                       Fe
                                                                                                              Type
                                                                                        0.451437 -0.078151
            ld 1.000000 -0.065721 0.380767 -0.647082 0.488475
                                                             0.052634
                                                                      -0.002306
                                                                               0.090257
                                                                               0.812495 0.001062 0.145791 -0.161322
            RI -0.065721 1.000000 -0.193436 -0.128118 -0.405671 -0.540010 -0.287900
               0.380767 -0.193436
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                                                                               -0.275314 0.327233
                                                                                                 -0.240802
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               -0.647082 -0.128118 -0.276486
                                           1.000000 -0.480035
                                                             -0.160359
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            Al 0.488475 -0.405671 0.157928
                                          -0.480035 1.000000
                                                             -0.009226
                                                                      0.324484
                                                                               -0.160359 -0.009226 1.000000 -0.197684
            Si 0.052634 -0.540010 -0.068519
                                                                               -0.210141 -0.104361 -0.097674 0.147767
             K -0.002306 -0.287900 -0.265520
                                           0.009397
                                                    0.324484
                                                             -0.197684
                                                                      1.000000
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            Ca 0.090257
                         0.812495 -0.275314
                                          -0.444559
                                                    -0.260372
                                                             -0.210141 -0.318649
                                                                               1.000000 -0.113121
                                                                                                 0.124674
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            Fe -0.078151 0.145791 -0.240802 0.086906 -0.076456 -0.097674 -0.009586
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          Type 0.878113 -0.161322 0.504983 -0.744004 0.597754 0.147767 -0.012765 0.000372 0.574896 -0.191090 1.000000
In [10]: sns.set(rc = {'figure.figsize':(15,8)})
          sns.heatmap(cor,annot=True)
          <AxesSubplot:>
Out[10]:
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                         RI
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                ld
                                                                                  Ba
                                                                                                  Type
 In [ ]
In [11]: nb = BernoulliNB()
          gnb = GaussianNB()
          mnb = MultinomialNB()
In [12]: x = data.drop(columns=['Type'])
          y = data['Type']
In [13]:
         xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size=0.2, random_state=4)
In [14]: nb.fit(xtrain, ytrain)
          gnb.fit(xtrain, ytrain)
          mnb.fit(xtrain, ytrain)
          MultinomialNB()
Out[14]:
In [15]:
          ypred = nb.predict(xtest)
In [16]:
         accuracy_score(ytest,ypred)
          0.4883720930232558
Out[16]:
In [17]:
          print(classification_report(ytest,ypred))
                                       recall f1-score
                                                          support
                         precision
                              0.39
                                         0.92
                                                   0.55
                                                                13
                              0.67
                                         0.11
                                                   0.19
                                                                18
                              0.00
                                         0.00
                                                   0.00
                                                                 4
                              0.50
                                        1.00
                                                   0.67
                              0.86
                                         0.86
                                                   0.86
                                                                43
              accuracy
                                        0.58
             macro avg
                              0.48
                                                   0.45
                                                                43
          weighted avg
                              0.55
                                        0.49
                                                   0.40
          D:\24-Annaconda\lib\site-packages\sklearn\metrics\_classification.py:1318: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted sa
          mples. Use `zero_division` parameter to control this behavior.
            _warn_prf(average, modifier, msg_start, len(result))
          D:\24-Annaconda\lib\site-packages\sklearn\metrics\_classification.py:1318: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted sa
          mples. Use `zero_division` parameter to control this behavior.
            _warn_prf(average, modifier, msg_start, len(result))
          D:\24-Annaconda\lib\site-packages\sklearn\metrics\_classification.py:1318: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted sa
          mples. Use `zero_division` parameter to control this behavior.
           _warn_prf(average, modifier, msg_start, len(result))
In [18]: sns.heatmap (cf,annot=True)
          plt.axis('equal')
          plt.show()
                                                      Traceback (most recent call last)
          Input In [18], in <cell line: 1>()
          ----> 1 sns.heatmap (cf, annot=True)
                2 plt.axis('equal')
                3 plt.show()
          NameError: name 'cf' is not defined
         cf = confusion_matrix(ytest,ypred)
          cf
 In [ ]: ypred = gnb.predict(xtest)
         accuracy_score(ytest,ypred)
         print(classification_report(ytest,ypred))
 In [ ]: sns.heatmap (cf,annot=True)
          plt.axis('equal')
          plt.show()
 In [ ]: cf = confusion_matrix(ytest,ypred)
          cf
 In [ ]: ypred = mnb.predict(xtest)
         accuracy_score(ytest,ypred)
         print(classification_report(ytest,ypred))
 In [ ]: cf = confusion_matrix(ytest,ypred)
          cf
 In [ ]: sn.heatmap (cf,annot=True)
          plt.axis('equal')
          plt.show()
          The accuracy BernoulliNB is 0.4883720930232558 The accuracy GaussianNB is 0.7906976744186046 The accuracy MultinomialNB is 0.7906976744186046
```