```
import numpy as np
In [1]:
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
In [2]:
          df=pd.read_csv("/home/aparna/Downloads/winequality-red.csv")
          df.head(10)
                                                           free
                                                                    total
Out[2]:
              fixed
                     volatile
                             citric
                                    residual
                                             chlorides
                                                          sulfur
                                                                   sulfur
                                                                          density
                                                                                        sulphates alcohol quality
             acidity
                     acidity
                              acid
                                      sugar
                                                        dioxide
                                                                 dioxide
          0
                        0.70
                              0.00
                                                           11.0
                                                                                                                5
                7.4
                                         1.9
                                                 0.076
                                                                    34.0
                                                                          0.9978
                                                                                  3.51
                                                                                             0.56
                                                                                                       9.4
          1
                7.8
                        0.88
                              0.00
                                         2.6
                                                 0.098
                                                           25.0
                                                                    67.0
                                                                           0.9968
                                                                                  3.20
                                                                                             0.68
                                                                                                       9.8
                                                                                                                5
          2
                              0.04
                7.8
                        0.76
                                         2.3
                                                 0.092
                                                           15.0
                                                                    54.0
                                                                           0.9970
                                                                                  3.26
                                                                                             0.65
                                                                                                       9.8
                                                                                                                5
          3
               11.2
                        0.28
                              0.56
                                         1.9
                                                 0.075
                                                           17.0
                                                                    60.0
                                                                           0.9980
                                                                                  3.16
                                                                                             0.58
                                                                                                       9.8
                                                                                                                6
          4
                        0.70
                              0.00
                                         1.9
                                                 0.076
                                                           11.0
                                                                    34.0
                                                                           0.9978
                                                                                  3.51
                                                                                             0.56
                                                                                                       9.4
                                                                                                                5
                7.4
          5
                              0.00
                                                           13.0
                                                                                  3.51
                                                                                                                5
                7.4
                        0.66
                                         1.8
                                                 0.075
                                                                    40.0
                                                                           0.9978
                                                                                             0.56
                                                                                                       9.4
          6
                7.9
                                                           15.0
                                                                                                                5
                        0.60
                              0.06
                                         1.6
                                                 0.069
                                                                    59.0
                                                                           0.9964
                                                                                  3.30
                                                                                             0.46
                                                                                                      9.4
                7.3
                        0.65
                              0.00
                                         1.2
                                                 0.065
                                                           15.0
                                                                    21.0
                                                                           0.9946
                                                                                  3.39
                                                                                             0.47
                                                                                                      10.0
                                                                                                                7
          8
                        0.58
                              0.02
                                         2.0
                                                 0.073
                                                            9.0
                                                                    18.0
                                                                                  3.36
                                                                                             0.57
                                                                                                      9.5
                                                                                                                7
                7.8
                                                                           0.9968
          9
                7.5
                        0.50
                              0.36
                                         6.1
                                                 0.071
                                                           17.0
                                                                   102.0
                                                                           0.9978
                                                                                  3.35
                                                                                             0.80
                                                                                                      10.5
                                                                                                                5
          df.info()
In [3]:
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 1599 entries, 0 to 1598
          Data columns (total 12 columns):
           #
                Column
                                          Non-Null Count
                                                              Dtype
                _ _ _ _ _ _
          - - -
                                           -----
                                                              ----
                fixed acidity
           0
                                          1599 non-null
                                                              float64
                                                              float64
           1
                volatile acidity
                                           1599 non-null
           2
                citric acid
                                          1599 non-null
                                                              float64
                                                              float64
           3
                residual sugar
                                          1599 non-null
                chlorides
                                          1599 non-null
                                                              float64
           4
           5
                free sulfur dioxide
                                          1599 non-null
                                                              float64
                total sulfur dioxide
                                                              float64
           6
                                          1599 non-null
           7
                density
                                          1599 non-null
                                                              float64
           8
                                          1599 non-null
                                                              float64
                рН
           9
                sulphates
                                          1599 non-null
                                                              float64
                                                              float64
           10
                alcohol
                                          1599 non-null
                quality
                                          1599 non-null
                                                              int64
           11
```

dtypes: float64(11), int64(1)

#find the statistical measures

citric acid

0.270976

1599.000000

residual

2.538806

1599.000000

sugar

free sulfur

1599.000000

15.874922

dioxide

chlorides

0.087467

1599.000000

total sulfur

1599.000000

46.467792

dioxide

densit

1599.00000

0.99674

volatile

acidity

1599.000000

0.527821

memory usage: 150.0 KB

fixed acidity

1599.000000

8.319637

df.shape

count

(1599, 12)

df.describe()

In [4]:

Out[4]:

In [5]:

Out[5]:

st	d 1.741096	0.179060	0.194801	1.409928	0.047065	10.460157	32.895324	0.00188
mi	n 4.600000	0.120000	0.000000	0.900000	0.012000	1.000000	6.000000	0.99007
259	7.100000	0.390000	0.090000	1.900000	0.070000	7.000000	22.000000	0.99560
50 9	7.900000	0.520000	0.260000	2.200000	0.079000	14.000000	38.000000	0.99675
759	% 9.200000	0.640000	0.420000	2.600000	0.090000	21.000000	62.000000	0.99783
ma	x 15.900000	1.580000	1.000000	15.500000	0.611000	72.000000	289.000000	1.00369

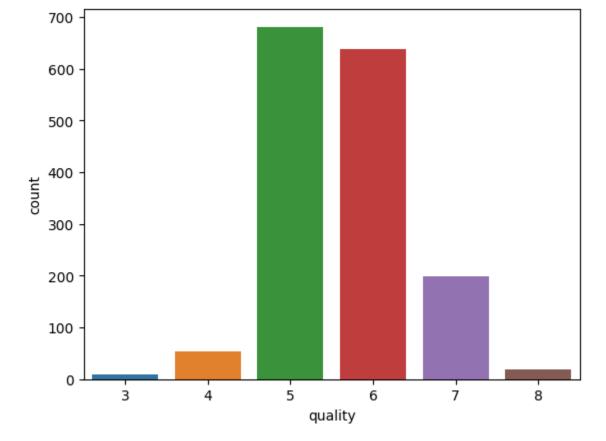
```
In [6]: df.isna().sum()
                                 0
        fixed acidity
Out[6]:
        volatile acidity
                                 0
        citric acid
                                 0
        residual sugar
                                 0
        chlorides
                                 0
        free sulfur dioxide
                                 0
        total sulfur dioxide
                                 0
        density
                                 0
                                 0
        рН
        sulphates
                                 0
        alcohol
                                 0
        quality
                                 0
        dtype: int64
In [7]: df['quality'].value_counts()
             681
Out[7]:
             638
             199
        4
              53
        8
              18
        3
              10
        Name: quality, dtype: int64
In [8]: #number of values for each quality
        sns.countplot('quality', data=df)
        /home/aparna/.local/lib/python3.8/site-packages/seaborn/_decorators.py:36: FutureWarnin
        g: Pass the following variable as a keyword arg: x. From version 0.12, the only valid po
        sitional argument will be `data`, and passing other arguments without an explicit keywor
```

d will result in an error or misinterpretation.

<AxesSubplot:xlabel='quality', ylabel='count'>

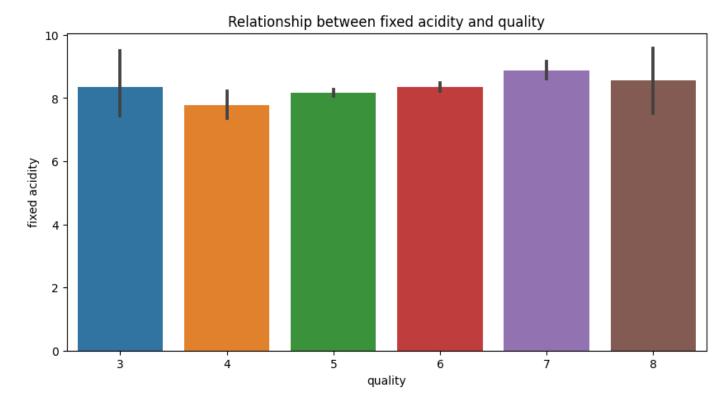
warnings.warn(

Out[8]:



```
In [9]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality',y='fixed acidity',data=df)
    plt.title('Relationship between fixed acidity and quality')
```

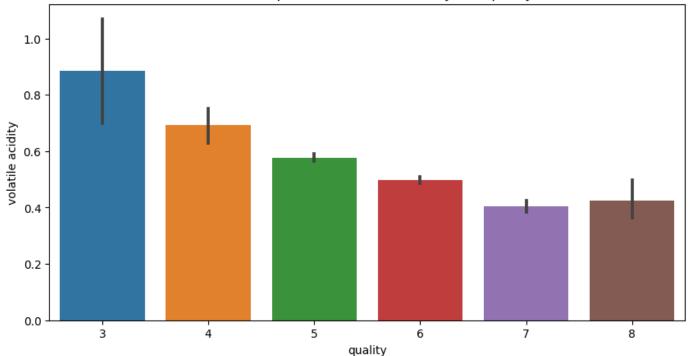
Out[9]: Text(0.5, 1.0, 'Relationship between fixed acidity and quality')



```
In [10]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality',y='volatile acidity',data=df)
    plt.title('Relationship between volatile acidity and quality')
```

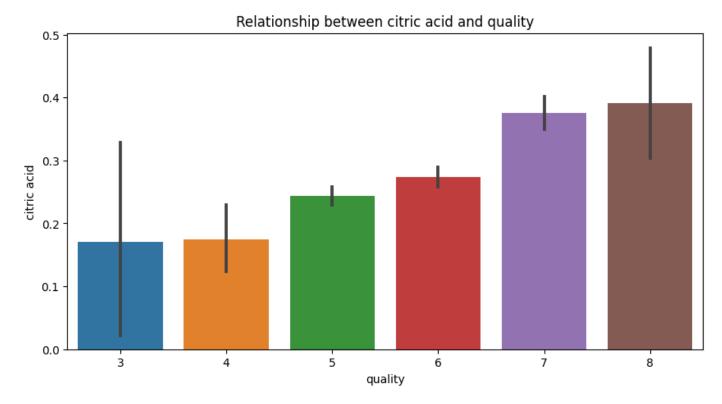
Out[10]: Text(0.5, 1.0, 'Relationship between volatile acidity and quality')

Relationship between volatile acidity and quality



```
In [11]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality',y='citric acid',data=df)
    plt.title('Relationship between citric acid and quality')
```

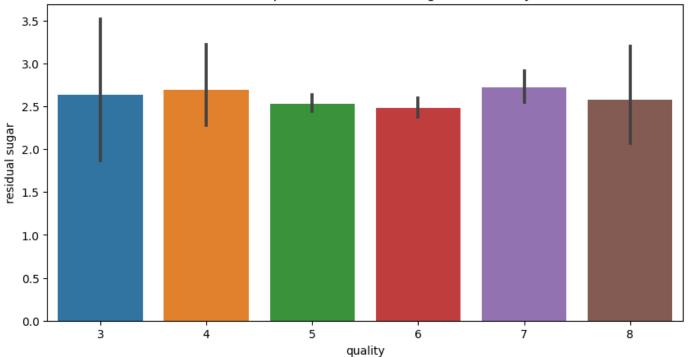
 ${\tt Out[11]:}$ Text(0.5, 1.0, 'Relationship between citric acid and quality')



```
In [18]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality',y='residual sugar',data=df)
    plt.title('Relationship between Residual sugar and Quality')
```

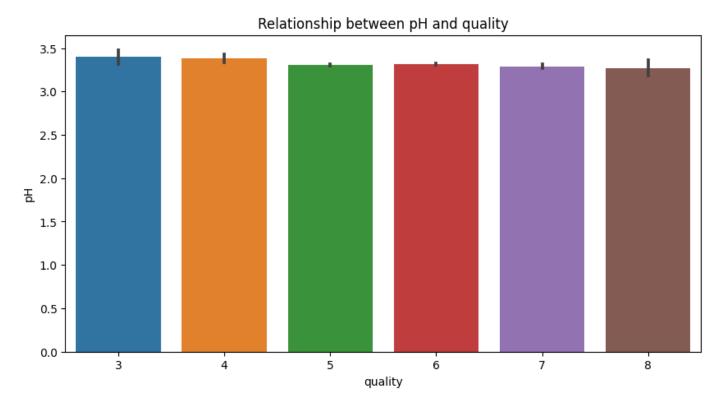
 ${\tt Out[18]:}$ Text(0.5, 1.0, 'Relationship between Residual sugar and Quality')

Relationship between Residual sugar and Quality



```
In [12]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality',y='pH',data=df)
    plt.title('Relationship between pH and quality')
```

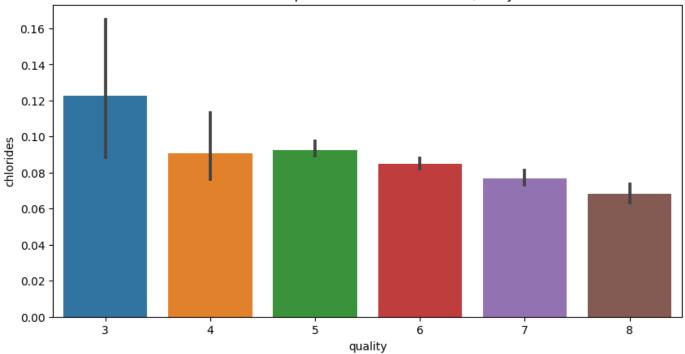
Out[12]: Text(0.5, 1.0, 'Relationship between pH and quality')



```
In [19]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality',y='chlorides',data=df)
    plt.title('Relationship between Chlorides and Quality')
```

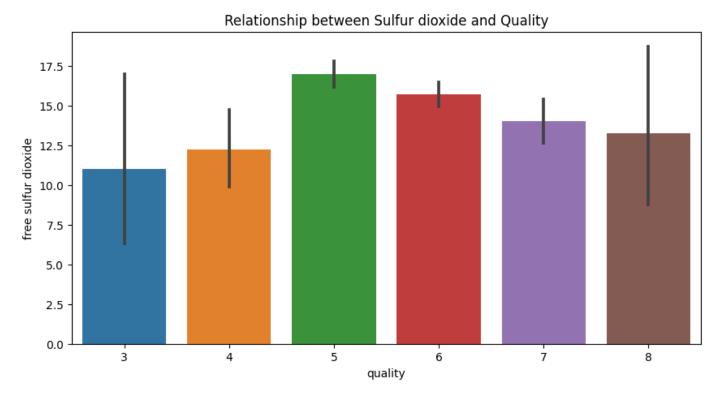
 ${\tt Out[19]:}$ Text(0.5, 1.0, 'Relationship between Chlorides and Quality')

Relationship between Chlorides and Quality



```
In [20]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality',y='free sulfur dioxide',data=df)
    plt.title('Relationship between Sulfur dioxide and Quality')
```

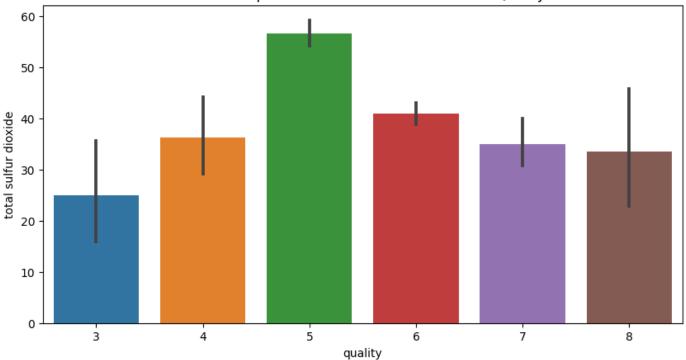
Out[20]: Text(0.5, 1.0, 'Relationship between Sulfur dioxide and Quality')



```
In [21]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality', y='total sulfur dioxide', data=df)
    plt.title('Relationship between Total sulfur dioxide and Quality')
```

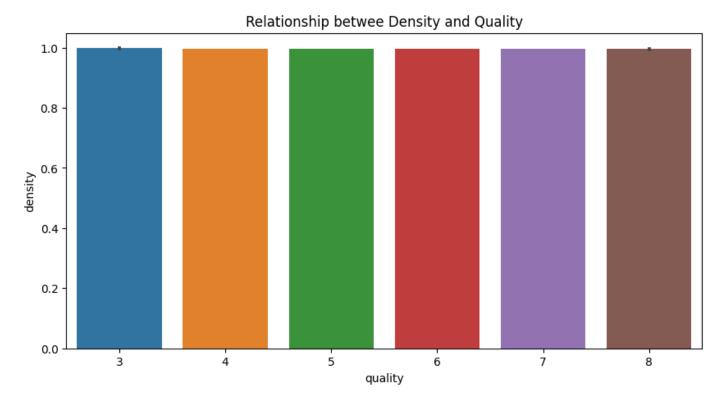
Out[21]: Text(0.5, 1.0, 'Relationship between Total sulfur dioxide and Quality')

Relationship between Total sulfur dioxide and Quality



```
In [22]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality', y='density', data=df)
    plt.title('Relationship betwee Density and Quality')
```

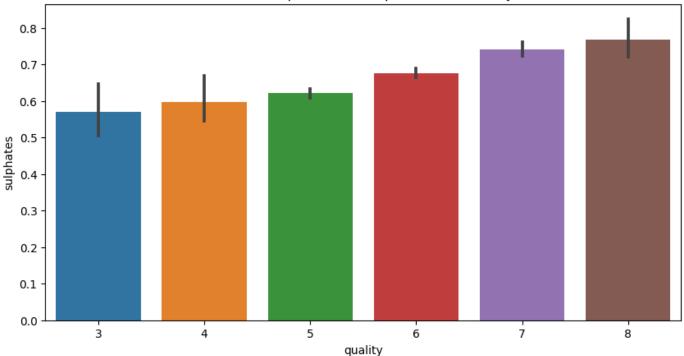
Out[22]: Text(0.5, 1.0, 'Relationship betwee Density and Quality')



```
In [23]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality', y='sulphates', data=df)
    plt.title('Relationship between Sulphates and Quality')
```

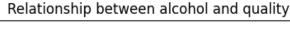
Out[23]: Text(0.5, 1.0, 'Relationship between Sulphates and Quality')

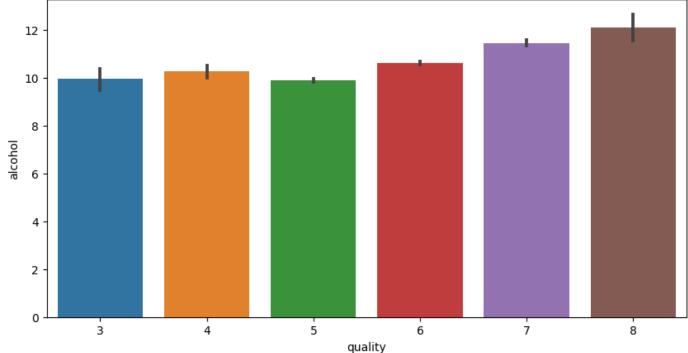
Relationship between Sulphates and Quality



```
In [13]: plt.figure(figsize=(10,5))
    sns.barplot(x='quality',y='alcohol',data=df)
    plt.title('Relationship between alcohol and quality')
```

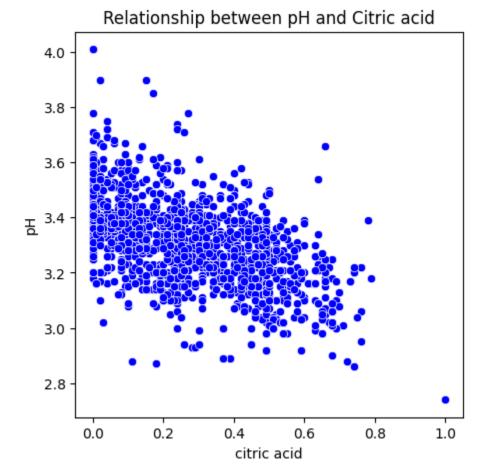
Out[13]: Text(0.5, 1.0, 'Relationship between alcohol and quality')





```
In [14]: plt.figure(figsize=(5,5))
    sns.scatterplot(x='citric acid',y='pH',data=df,color='b')
    plt.title("Relationship between pH and Citric acid")
```

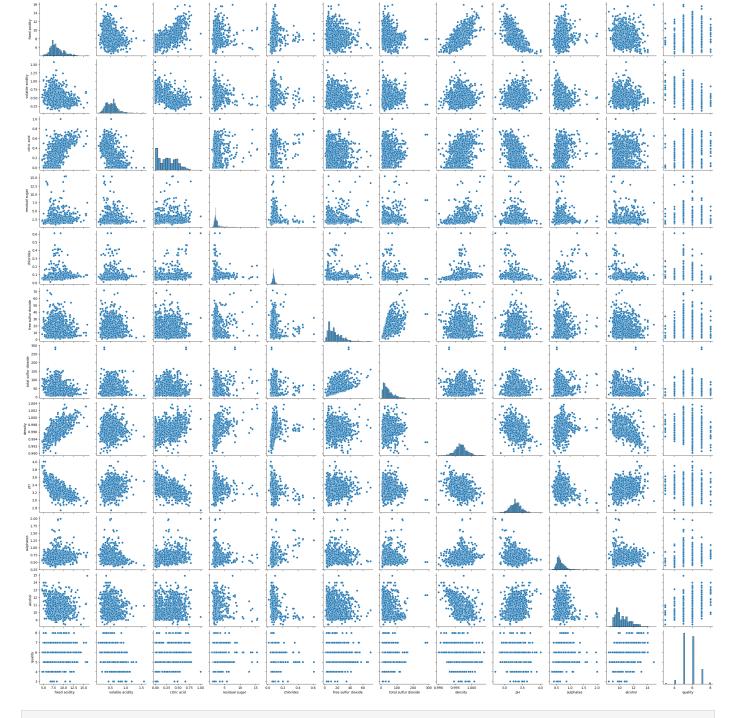
 $\operatorname{Out}[14]$: Text(0.5, 1.0, 'Relationship between pH and Citric acid')



```
In [15]: plt.figure(figsize=(10,10))
sns.pairplot(df)
```

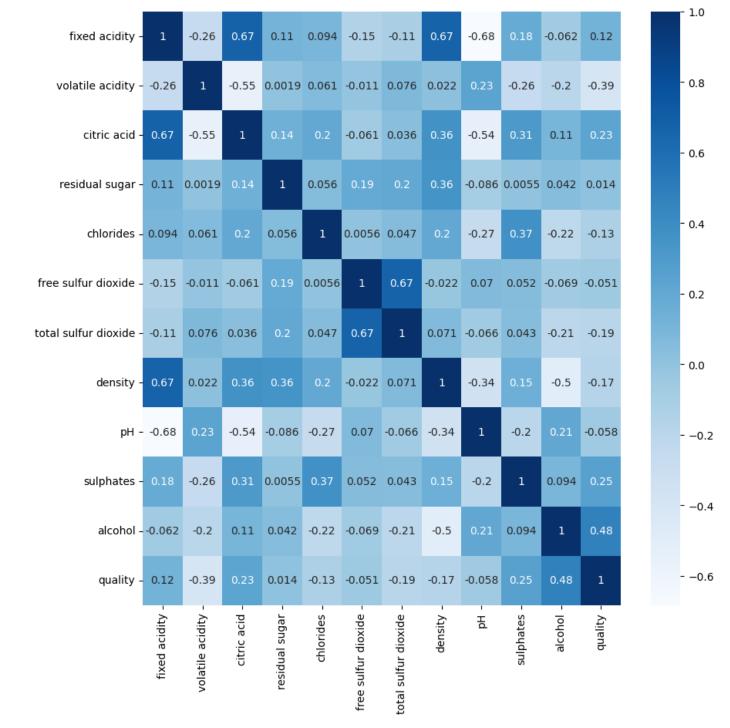
Out[15]: <seaborn.axisgrid.PairGrid at 0x7f03aa19b190>

<Figure size 1000x1000 with 0 Axes>



```
In []:
In [16]: #correlation between all the columns
   plt.figure(figsize=(10,10))
   sns.heatmap(df.corr(),annot=True,cmap='Blues')
```

Out[16]: <AxesSubplot:>



In [24]: x=df.drop(['quality'],axis=1)
x

Out[24]:

fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcohol
7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	9.4
7.8	0.880	0.00	2.6	0.098	25.0	67.0	0.99680	3.20	0.68	9.8
7.8	0.760	0.04	2.3	0.092	15.0	54.0	0.99700	3.26	0.65	9.8
11.2	0.280	0.56	1.9	0.075	17.0	60.0	0.99800	3.16	0.58	9.8
7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	9.4
6.2	0.600	0.08	2.0	0.090	32.0	44.0	0.99490	3.45	0.58	10.5
5.9	0.550	0.10	2.2	0.062	39.0	51.0	0.99512	3.52	0.76	11.2
	7.4 7.8 7.8 11.2 7.4 6.2	acidity acidity 7.4 0.700 7.8 0.880 7.8 0.760 11.2 0.280 7.4 0.700 6.2 0.600	acidity acidity acid 7.4 0.700 0.00 7.8 0.880 0.00 7.8 0.760 0.04 11.2 0.280 0.56 7.4 0.700 0.00 6.2 0.600 0.08	acidity acidity acid sugar 7.4 0.700 0.00 1.9 7.8 0.880 0.00 2.6 7.8 0.760 0.04 2.3 11.2 0.280 0.56 1.9 7.4 0.700 0.00 1.9 6.2 0.600 0.08 2.0	acidity acidity acid sugar chlorides 7.4 0.700 0.00 1.9 0.076 7.8 0.880 0.00 2.6 0.098 7.8 0.760 0.04 2.3 0.092 11.2 0.280 0.56 1.9 0.075 7.4 0.700 0.00 1.9 0.076 6.2 0.600 0.08 2.0 0.090	fixed acidity volatile acidity citric acid residual sugar chlorides sulfur dioxide 7.4 0.700 0.00 1.9 0.076 11.0 7.8 0.880 0.00 2.6 0.098 25.0 7.8 0.760 0.04 2.3 0.092 15.0 11.2 0.280 0.56 1.9 0.075 17.0 7.4 0.700 0.00 1.9 0.076 11.0 6.2 0.600 0.08 2.0 0.090 32.0	fixed acidity volatile acidity citric acid residual sugar chlorides sulfur dioxide sulfur dioxide 7.4 0.700 0.00 1.9 0.076 11.0 34.0 7.8 0.880 0.00 2.6 0.098 25.0 67.0 7.8 0.760 0.04 2.3 0.092 15.0 54.0 11.2 0.280 0.56 1.9 0.075 17.0 60.0 7.4 0.700 0.00 1.9 0.076 11.0 34.0 6.2 0.600 0.08 2.0 0.090 32.0 44.0	fixed acidity volatile acidity cltric acid residual sugar chlorides sulfur dioxide sulfur dioxide density 7.4 0.700 0.00 1.9 0.076 11.0 34.0 0.99780 7.8 0.880 0.00 2.6 0.098 25.0 67.0 0.99680 7.8 0.760 0.04 2.3 0.092 15.0 54.0 0.99700 11.2 0.280 0.56 1.9 0.075 17.0 60.0 0.99800 7.4 0.700 0.00 1.9 0.076 11.0 34.0 0.99780	fixed acidity Volatile acidity cltric acid residual sugar sugar chlorides sulfur dioxide sulfur dioxide density pH 7.4 0.700 0.00 1.9 0.076 11.0 34.0 0.99780 3.51 7.8 0.880 0.00 2.6 0.098 25.0 67.0 0.99680 3.20 7.8 0.760 0.04 2.3 0.092 15.0 54.0 0.99700 3.26 11.2 0.280 0.56 1.9 0.075 17.0 60.0 0.99800 3.16 7.4 0.700 0.00 1.9 0.076 11.0 34.0 0.99780 3.51 6.2 0.600 0.08 2.0 0.090 32.0 44.0 0.99490 3.45	fixed acidity Volatile acidity cltric acid residual sugar sugar chlorides dioxide sulfur dioxide sulfur dioxide density pH sulphates 7.4 0.700 0.00 1.9 0.076 11.0 34.0 0.99780 3.51 0.56 7.8 0.880 0.00 2.6 0.098 25.0 67.0 0.99680 3.20 0.68 7.8 0.760 0.04 2.3 0.092 15.0 54.0 0.99700 3.26 0.65 11.2 0.280 0.56 1.9 0.075 17.0 60.0 0.99800 3.16 0.58 7.4 0.700 0.00 1.9 0.076 11.0 34.0 0.99780 3.51 0.56

```
1596
                                                0.076
                                                             29.0
                                                                        40.0 0.99574 3.42
                                                                                                    0.75
          6.3
                  0.510
                           0.13
                                       2.3
                                                                                                             11.0
1597
          5.9
                  0.645
                           0.12
                                       2.0
                                                0.075
                                                             32.0
                                                                        44.0 0.99547 3.57
                                                                                                    0.71
                                                                                                             10.2
1598
          6.0
                  0.310
                                                                                                    0.66
                           0.47
                                       3.6
                                                0.067
                                                             18.0
                                                                        42.0 0.99549 3.39
                                                                                                             11.0
```

1599 rows × 11 columns

```
#label binarization
In [25]:
          y=df['quality'].apply(lambda x:1 if x>=7 else 0)
                  0
Out[25]:
                  0
         2
                  0
         3
                  0
         4
                  0
                 . .
         1594
                  0
         1595
                  0
         1596
                  0
         1597
                  0
         1598
         Name: quality, Length: 1599, dtype: int64
         from sklearn.model_selection import train_test_split
In [26]:
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=1)
```

Out[26]:

x_test

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcohol
75	8.8	0.41	0.64	2.2	0.093	9.0	42.0	0.99860	3.54	0.66	10.5
1283	8.7	0.63	0.28	2.7	0.096	17.0	69.0	0.99734	3.26	0.63	10.2
408	10.4	0.34	0.58	3.7	0.174	6.0	16.0	0.99700	3.19	0.70	11.3
1281	7.1	0.46	0.20	1.9	0.077	28.0	54.0	0.99560	3.37	0.64	10.4
1118	7.1	0.39	0.12	2.1	0.065	14.0	24.0	0.99252	3.30	0.53	13.3
1596	6.3	0.51	0.13	2.3	0.076	29.0	40.0	0.99574	3.42	0.75	11.0
985	7.4	0.58	0.00	2.0	0.064	7.0	11.0	0.99562	3.45	0.58	11.3
671	8.2	0.73	0.21	1.7	0.074	5.0	13.0	0.99680	3.20	0.52	9.5
1379	7.5	0.57	0.02	2.6	0.077	11.0	35.0	0.99557	3.36	0.62	10.8
1169	7.6	0.50	0.29	2.3	0.086	5.0	14.0	0.99502	3.32	0.62	11.5

480 rows × 11 columns

```
Θ,
              0, 0, 1, 1, 0, 0, 1,
                               0, 0,
                                       0, 1, 0,
                                               0, 0, 0, 0,
                                                         0, 0,
              0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0])
        model.predict([[11.2,0.280,0.56 ,1.9 ,0.075 ,17.0,60.0,0.99800,3.16 ,0.58 ,9.8]])
In [28]:
        /home/aparna/.local/lib/python3.8/site-packages/sklearn/base.py:450: UserWarning: X does
        not have valid feature names, but KNeighborsClassifier was fitted with feature names
         warnings.warn(
        array([0])
Out[28]:
        from sklearn.metrics import accuracy_score,classification_report,ConfusionMatrixDisplay
In [291:
        print(accuracy_score(y_test,y_pred))
        0.885416666666666
        print(classification_report(y_test,y_pred))
In [30]:
                    precision
                               recall f1-score
                                               support
                 0
                        0.92
                                 0.95
                                         0.94
                                                   425
                        0.50
                                 0.40
                                         0.44
                                                    55
                                         0.89
                                                   480
           accuracy
                                                   480
          macro avg
                        0.71
                                 0.67
                                         0.69
                        0.88
                                 0.89
                                         0.88
                                                   480
        weighted avg
        from sklearn.metrics import mean_absolute_error
In [31]:
        print('MAE', mean_absolute_error(y_test, y_pred))
        MAE 0.114583333333333333
        ConfusionMatrixDisplay.from_predictions(y_test,y_pred)
In [32]:
        <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x7f039bce2910>
Out[321:
```

Θ,

Θ,

0, 0,

0, 0, 1,

0, 0,

0, 0,

0, 0, 0,

Θ,

Θ,

0, 0, 0,

0, 0, 0,

0, 0, 0, 1, 0,

0, 0, 0, 0, 0,

0, 0, 0,

0, 0, 0, 0, 0,

0, 0, 0, 1, 0,

0, 0, 0, 0, 0,

0, 0,

0, 1,

0, 0,

Θ,

0, 0,

0, 0, 1,

0, 0,

0, 0,

Θ,

0, 0,

Θ,

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0, 0,

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1, 0, 0,

0, 0,

1, 0,

0, 0,

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Θ,

1,

Θ,

Θ,

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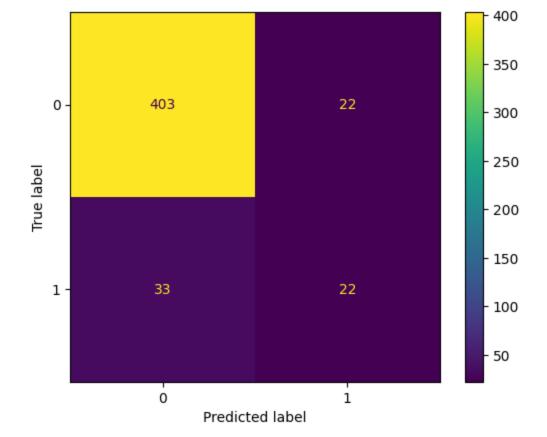
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In []: