→ Assignment 4

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21BIT0023

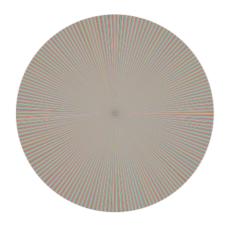
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

df=pd.read_csv('/content/winequality-red.csv')
df

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН
0	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51
1	7.8	0.880	0.00	2.6	0.098	25.0	67.0	0.99680	3.20
2	7.8	0.760	0.04	2.3	0.092	15.0	54.0	0.99700	3.26
3	11.2	0.280	0.56	1.9	0.075	17.0	60.0	0.99800	3.16
4	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51
1594	6.2	0.600	0.08	2.0	0.090	32.0	44.0	0.99490	3.45
1595	5.9	0.550	0.10	2.2	0.062	39.0	51.0	0.99512	3.52
1596	6.3	0.510	0.13	2.3	0.076	29.0	40.0	0.99574	3.42
1597	5.9	0.645	0.12	2.0	0.075	32.0	44.0	0.99547	3.57
1598	6.0	0.310	0.47	3.6	0.067	18.0	42.0	0.99549	3.39
4									-

plt.pie(df.quality)
plt.show

<function matplotlib.pyplot.show(close=None, block=None)>



sns.distplot(df.quality)

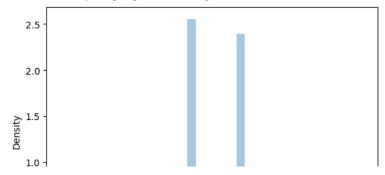
<ipython-input-4-e8684199aa87>:1: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

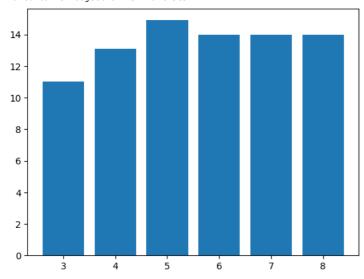
For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

sns.distplot(df.quality)
<Axes: xlabel='quality', ylabel='Density'>



plt.bar(df.quality,df.alcohol)

<BarContainer object of 1599 artists>

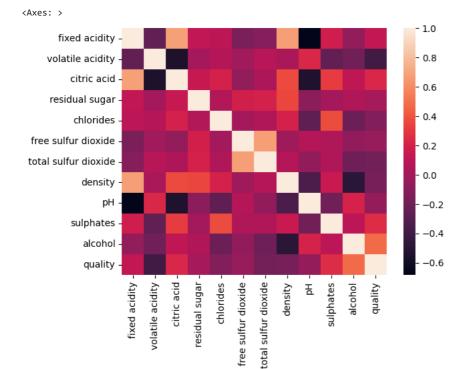


sns.jointplot(x='quality',y='pH',data=df)

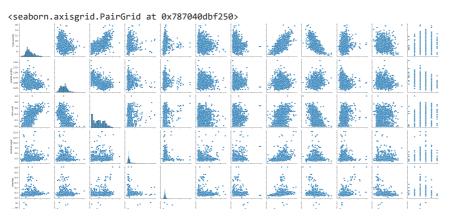
<seaborn.axisgrid.JointGrid at 0x78703f2eb430>



sns.heatmap(df.corr())



sns.pairplot(df)



df.describe()

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	
count	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	1599.000000	15
mean	8.319637	0.527821	0.270976	2.538806	0.087467	15.874922	
std	1.741096	0.179060	0.194801	1.409928	0.047065	10.460157	
min	4.600000	0.120000	0.000000	0.900000	0.012000	1.000000	
25%	7.100000	0.390000	0.090000	1.900000	0.070000	7.000000	
50%	7.900000	0.520000	0.260000	2.200000	0.079000	14.000000	
75%	9.200000	0.640000	0.420000	2.600000	0.090000	21.000000	
max	15.900000	1.580000	1.000000	15.500000	0.611000	72.000000	2

df.isnull().any()

fixed acidity	False
volatile acidity	False
citric acid	False
residual sugar	False
chlorides	False
free sulfur dioxide	False
total sulfur dioxide	False
density	False
рН	False
sulphates	False
alcohol	False
quality	False
dtype: bool	

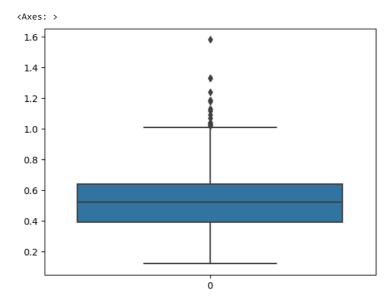
df.median()

fixed acidity	7.90000
volatile acidity	0.52000
citric acid	0.26000
residual sugar	2.20000
chlorides	0.07900
free sulfur dioxide	14.00000
total sulfur dioxide	38.00000
density	0.99675
рН	3.31000
sulphates	0.62000
alcohol	10.20000
quality	6.00000
dtype: float64	

sns.boxplot(df['fixed acidity'])

```
<Axes: >
       16
       14
       12
       10
q1=df['fixed acidity'].quantile(0.25)
q3=df['fixed acidity'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+(1.5*IQR)
lower_limit=q1-(1.5*IQR)
        ۲٥
df['fixed acidity']=np.where(df['fixed acidity']>upper_limit,7.9,df['fixed acidity'])
sns.boxplot(df['fixed acidity'])
     <Axes: >
       12
       11
       10
        9
        8
        7
        6
        5
                                             0
q1=df['fixed acidity'].quantile(0.25)
q3=df['fixed acidity'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+(1.5*IQR)
lower_limit=q1-(1.5*IQR)
df['fixed acidity']=np.where(df['fixed acidity']>upper_limit,7.9,df['fixed acidity'])
sns.boxplot(df['fixed acidity'])
```

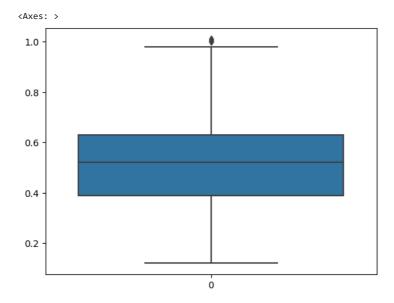
sns.boxplot(df['volatile acidity'])



```
q1=df['volatile acidity'].quantile(0.25)
q3=df['volatile acidity'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+(1.5*IQR)
lower_limit=q1-(1.5*IQR)
```

df['volatile acidity']=np.where(df['volatile acidity']>upper_limit,0.52,df['volatile acidity'])

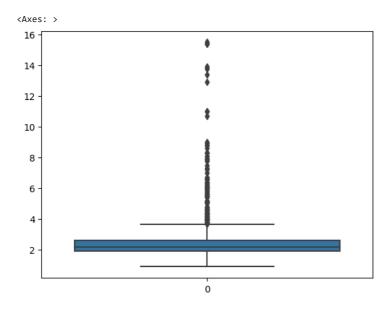
sns.boxplot(df['volatile acidity'])



sns.boxplot(df['citric acid'])

```
<Axes: >
      1.0
      Λ 0
q1=df['citric acid'].quantile(0.25)
q3=df['citric acid'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR
df['citric acid'] = np.where(df['citric acid']>upper_limit,0.26,df['citric acid'])
sns.boxplot(df['citric acid'])
     <Axes: >
      0.8
      0.7
      0.6
      0.5
      0.4
      0.3
      0.2
      0.1
      0.0
```

sns.boxplot(df['residual sugar'])



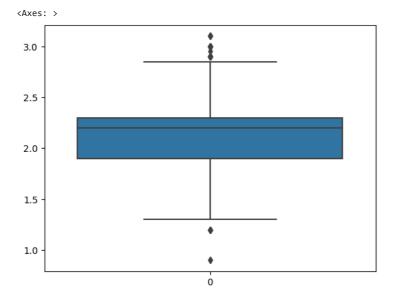
```
q1=df['residual sugar'].quantile(0.25)
q3=df['residual sugar'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

df['residual sugar'] = np.where(df['residual sugar']>upper_limit,2.2,df['residual sugar'])
sns.boxplot(df['residual sugar'])
```

q1=df['residual sugar'].quantile(0.25)
q3=df['residual sugar'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

df['residual sugar'] = np.where(df['residual sugar']>upper_limit,2.2,df['residual sugar'])

sns.boxplot(df['residual sugar'])

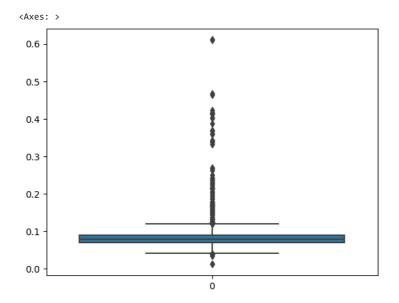


```
q1=df['residual sugar'].quantile(0.25)
q3=df['residual sugar'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

df['residual sugar'] = np.where(df['residual sugar']<lower_limit,2.2,df['residual sugar'])
sns.boxplot(df['residual sugar'])</pre>
```

```
<Axes: >
      3.00
      2.75
      2.50
q1=df['residual sugar'].quantile(0.25)
q3=df['residual sugar'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR
      1./27
df['residual sugar'] = np.where(df['residual sugar']>upper_limit,2.2,df['residual sugar'])
sns.boxplot(df['residual sugar'])
     <Axes: >
      2.8
      2.6
      2.4
      2.2
      2.0
      1.8
      1.6
      1.4
                                           0
```

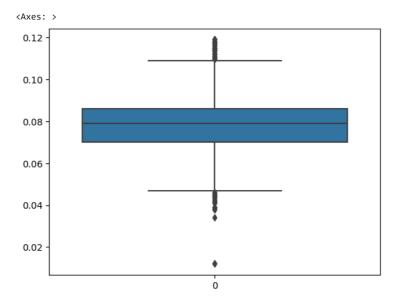
sns.boxplot(df['chlorides'])



```
q1=df['chlorides'].quantile(0.25)
q3=df['chlorides'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

df['chlorides'] = np.where(df['chlorides']>upper_limit,0.08,df['chlorides'])
```

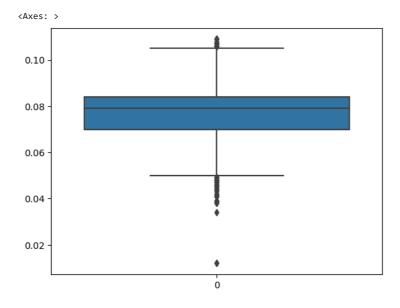
sns.boxplot(df['chlorides'])



```
q1=df['chlorides'].quantile(0.25)
q3=df['chlorides'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR
```

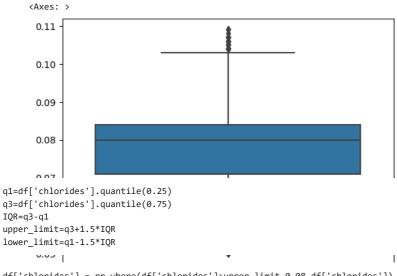
df['chlorides'] = np.where(df['chlorides']>upper_limit,0.08,df['chlorides'])

sns.boxplot(df['chlorides'])



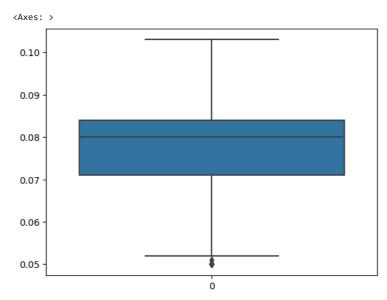
```
q1=df['chlorides'].quantile(0.25)
q3=df['chlorides'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

df['chlorides'] = np.where(df['chlorides']<lower_limit,0.08,df['chlorides'])
sns.boxplot(df['chlorides'])</pre>
```

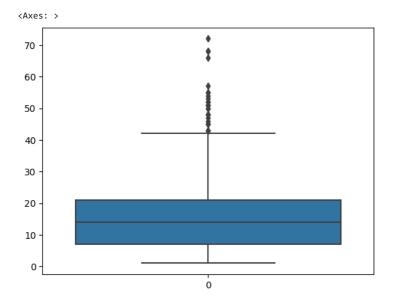


 $\label{eq:df['chlorides']} {\tt df['chlorides']} {\tt upper_limit,0.08,df['chlorides']} \\$

sns.boxplot(df['chlorides'])



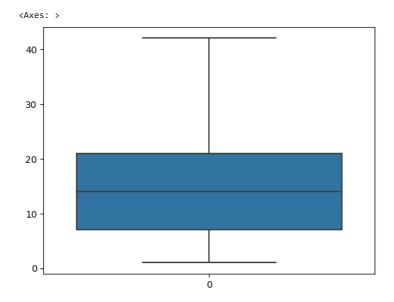
sns.boxplot(df['free sulfur dioxide'])



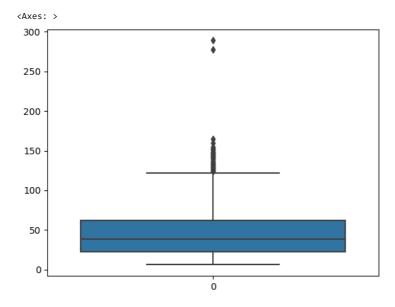
```
q1=df['free sulfur dioxide'].quantile(0.25)
q3=df['free sulfur dioxide'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR
```

df['free sulfur dioxide'] = np.where(df['free sulfur dioxide']>upper_limit,14,df['free sulfur dioxide'])

sns.boxplot(df['free sulfur dioxide'])

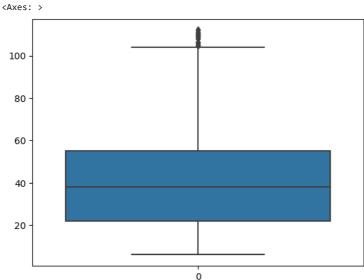


sns.boxplot(df['total sulfur dioxide'])

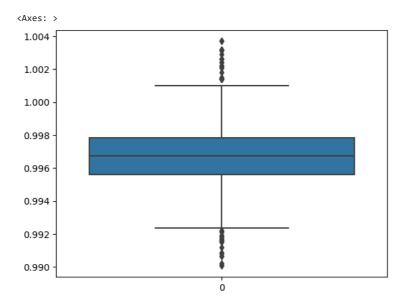


```
q1=df['total sulfur dioxide'].quantile(0.25)
q3=df['total sulfur dioxide'].quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

df['total sulfur dioxide'] = np.where(df['total sulfur dioxide']>upper_limit,38,df['total sulfur dioxide'])
sns.boxplot(df['total sulfur dioxide'])
```

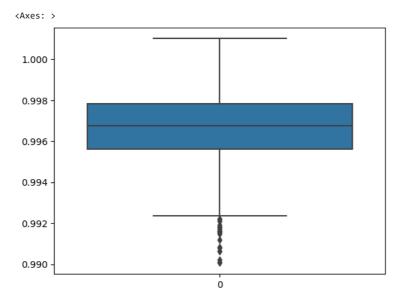


sns.boxplot(df.density)



```
q1=df.density.quantile(0.25)
q3=df.density.quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

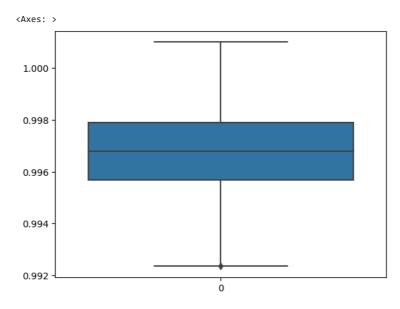
df.density = np.where(df.density>upper_limit,1,df.density)
sns.boxplot(df.density)
```



q1=df.density.quantile(0.25)
q3=df.density.quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

df.density = np.where(df.density<lower_limit,1,df.density)</pre>

sns.boxplot(df.density)

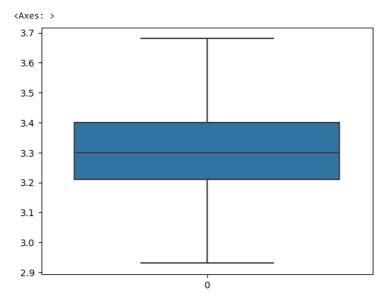


sns.boxplot(df.pH)

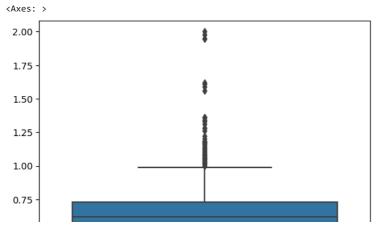
q1=df.pH.quantile(0.25)
q3=df.pH.quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

df.pH = np.where(df.pH>upper_limit,3.3,df.pH)

sns.boxplot(df.pH)



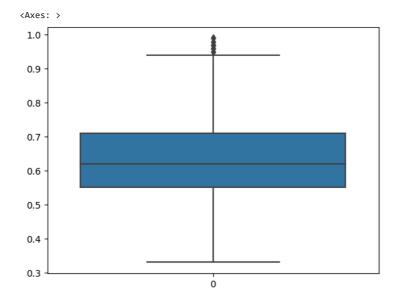
sns.boxplot(df.sulphates)



q1=df.sulphates.quantile(0.25)
q3=df.sulphates.quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR

df.sulphates = np.where(df.sulphates>upper_limit,0.62,df.sulphates)

sns.boxplot(df.sulphates)



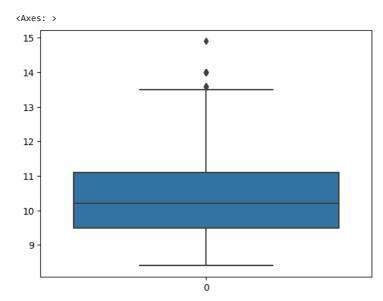
```
q1=df.sulphates.quantile(0.25)
q3=df.sulphates.quantile(0.75)
IQR=q3-q1
upper_limit=q3+1.5*IQR
lower_limit=q1-1.5*IQR
```

df.sulphates = np.where(df.sulphates>upper_limit,0.62,df.sulphates)

sns.boxplot(df.sulphates)



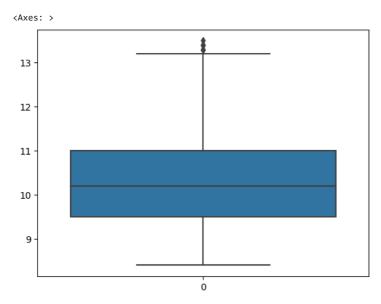
sns.boxplot(df.alcohol)



q1=df.alcohol.quantile(0.25) q3=df.alcohol.quantile(0.75) IQR=q3-q1 upper_limit=q3+1.5*IQR lower_limit=q1-1.5*IQR

df.alcohol = np.where(df.alcohol>upper_limit,10.2,df.alcohol)

sns.boxplot(df.alcohol)



sns.boxplot(df.quality)

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1599 entries, 0 to 1598
Data columns (total 12 columns):

Data	COTUMNIS (COCAT 12 COT	uiii 13) •	
#	Column	Non-Null Count	Dtype
0	fixed acidity	1599 non-null	float64
1	volatile acidity	1599 non-null	float64
2	citric acid	1599 non-null	float64
3	residual sugar	1599 non-null	float64
4	chlorides	1599 non-null	float64
5	free sulfur dioxide	1599 non-null	float64
6	total sulfur dioxide	1599 non-null	float64
7	density	1599 non-null	float64
8	рН	1599 non-null	float64
9	sulphates	1599 non-null	float64
10	alcohol	1599 non-null	float64
11	quality	1599 non-null	int64

dtypes: float64(11), int64(1)
memory usage: 150.0 KB

x=df.drop(columns=['quality'],axis=1)

x.head()

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcohol	
0	7.4	0.70	0.00	1.9	0.076	11.0	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	
2	7.8	0.76	0.04	2.3	0.092	15.0	54.0	0.9970	3.26	0.65	9.8	
3	11.2	0.28	0.56	1.9	0.075	17.0	60.0	0.9980	3.16	0.58	9.8	
4	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4	

```
y=df.quality
y.head()
```

0 5 1 5

2 5

2 :

Name: quality, dtype: int64

from sklearn.preprocessing import MinMaxScaler

scale=MinMaxScaler()

 $x_s = pd.DataFrame(scale.fit_transform(x),columns = x.columns)$

from sklearn.model_selection import train_test_split

 $x_train, x_test, y_train, y_test=train_test_split(x_s, y, test_size=0.2, random_state=0)$

 $x_train.shape, y_train.shape$

((1279, 11), (1279,))

x_test.shape,y_test.shape

((320, 11), (320,))

 \blacksquare

```
# Random Forest
from sklearn.ensemble import RandomForestClassifier
model=RandomForestClassifier(n_estimators=400)
model.fit(x_train,y_train)
               RandomForestClassifier
     RandomForestClassifier(n_estimators=400)
y_pred=model.predict(x_test)
y_pred_train=model.predict(x_train)
from sklearn.metrics import accuracy_score
print(accuracy_score(y_train,y_pred_train))
     1.0
print(accuracy_score(y_test,y_pred))
     0.70625
from \ sklearn.metrics \ import \ classification\_report, confusion\_matrix
print(classification_report(y_test,y_pred))
                   precision
                                recall f1-score
                                                   support
                3
                        0.00
                                  0.00
                                            0.00
                4
                        0.00
                                  0.00
                                             0.00
                                                        11
                5
                        0.76
                                  0.81
                                             0.78
                                                        135
                        0.71
                                  0.73
                                            0.72
                6
                                                        142
                7
                        0.47
                                  0.52
                                            0.49
                                                        27
                8
                        0.00
                                  0.00
                                            0.00
                                                         3
         accuracy
                                             0.71
                                                        320
        macro avg
                        0.32
                                  0.34
                                             0.33
                                                        320
     weighted avg
                        0.68
                                  0.71
                                            0.69
                                                        320
     /usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-score are
       _warn_prf(average, modifier, msg_start, len(result))
     /usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-score are
       _warn_prf(average, modifier, msg_start, len(result))
     /usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-score are
       _warn_prf(average, modifier, msg_start, len(result))
    4
confusion_matrix(y_test, y_pred)
                                        0],
     array([[ 0,
                    0.
                         0.
                              2.
                                   0.
               0,
                    0,
                         6,
                              5,
                                   0,
                                        0],
                    0, 109,
                                        0],
               0,
                            24,
                                   2,
                                        0],
               0,
                    0,
                       27, 103,
                                  12,
               0,
                    0,
                         1, 10,
                                        2],
               0.
                         0.
                                        0]])
model.predict([[7.5, 0.8, 0.0, 2, 0.75, 15, 38, 0.96, 3.5, 0.5, 9.8]])
     /usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but RandomForestClas
       warnings.warn(
     array([6])
model.predict([[7.4, 0.7, 0.1, 2, 0.75, 14, 35, 0.96, 3.5, 0.5, 9.9]])
     /usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but RandomForestClas
       warnings.warn(
     array([6])
    4
model.predict([[7.9, 0.52, 0.26, 2.2, 0.08, 14, 38, 0.99, 3.3, 0.62, 10.2]])
     /usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but RandomForestClas
      warnings.warn(
     array([6])
```