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Class: B.E.I.T. 4th year 1st semester ML

Assignment-2

1. Iris Dataset

Import required modules

```
import numpy as np
import pandas as pd
from sklearn import datasets
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
import seaborn as sns
from sklearn.neural_network import MLPClassifier
from sklearn.ensemble import RandomForestClassifier
```

Load Dataset

```
iris = datasets.load iris() # it's source is same as : https://archive.ics.uci.edu/ml/dat
asets/Iris/
In [ ]:
dir(iris)
Out[]:
['DESCR', 'data', 'feature names', 'filename', 'target', 'target names']
In [ ]:
iris.data
Out[]:
array([[5.1, 3.5, 1.4, 0.2],
       [4.9, 3., 1.4, 0.2],
       [4.7, 3.2, 1.3, 0.2],
       [4.6, 3.1, 1.5, 0.2],
       [5., 3.6, 1.4, 0.2],
       [5.4, 3.9, 1.7, 0.4],
       [4.6, 3.4, 1.4, 0.3],
       [5., 3.4, 1.5, 0.2],
       [4.4, 2.9, 1.4, 0.2],
       [4.9, 3.1, 1.5, 0.1],
       [5.4, 3.7, 1.5, 0.2],
       [4.8, 3.4, 1.6, 0.2],
       [4.8, 3., 1.4, 0.1],
       [4.3, 3., 1.1, 0.1],
```

```
[5.8, 4., 1.2, 0.2],
[5.7, 4.4, 1.5, 0.4],
[5.4, 3.9, 1.3, 0.4],
[5.1, 3.5, 1.4, 0.3],
[5.7, 3.8, 1.7, 0.3],
[5.1, 3.8, 1.5, 0.3],
[5.4, 3.4, 1.7, 0.2],
[5.1, 3.7, 1.5, 0.4],
[4.6, 3.6, 1., 0.2],
[5.1, 3.3, 1.7, 0.5],
[4.8, 3.4, 1.9, 0.2],
[5., 3., 1.6, 0.2],
[5., 3.4, 1.6, 0.4],
[5.2, 3.5, 1.5, 0.2],
[5.2, 3.4, 1.4, 0.2],
[4.7, 3.2, 1.6, 0.2],
[4.8, 3.1, 1.6, 0.2],
[5.4, 3.4, 1.5, 0.4],
[5.2, 4.1, 1.5, 0.1],
[5.5, 4.2, 1.4, 0.2],
[4.9, 3.1, 1.5, 0.2],
[5., 3.2, 1.2, 0.2],
[5.5, 3.5, 1.3, 0.2],
[4.9, 3.6, 1.4, 0.1],
[4.4, 3., 1.3, 0.2],
[5.1, 3.4, 1.5, 0.2],
[5., 3.5, 1.3, 0.3],
[4.5, 2.3, 1.3, 0.3],
[4.4, 3.2, 1.3, 0.2],
[5., 3.5, 1.6, 0.6],
[5.1, 3.8, 1.9, 0.4],
[4.8, 3., 1.4, 0.3], [5.1, 3.8, 1.6, 0.2],
[4.6, 3.2, 1.4, 0.2],
[5.3, 3.7, 1.5, 0.2],
[5., 3.3, 1.4, 0.2],
[7., 3.2, 4.7, 1.4],
[6.4, 3.2, 4.5, 1.5],
[6.9, 3.1, 4.9, 1.5],
[5.5, 2.3, 4., 1.3],
[6.5, 2.8, 4.6, 1.5],
[5.7, 2.8, 4.5, 1.3],
[6.3, 3.3, 4.7, 1.6],
[4.9, 2.4, 3.3, 1.],
[6.6, 2.9, 4.6, 1.3],
[5.2, 2.7, 3.9, 1.4],
[5., 2., 3.5, 1.],
[5.9, 3., 4.2, 1.5],
[6., 2.2, 4., 1.],
[6.1, 2.9, 4.7, 1.4],
[5.6, 2.9, 3.6, 1.3],
[6.7, 3.1, 4.4, 1.4],
[5.6, 3., 4.5, 1.5],
[5.8, 2.7, 4.1, 1.],
[6.2, 2.2, 4.5, 1.5],
[5.6, 2.5, 3.9, 1.1],
[5.9, 3.2, 4.8, 1.8],
[6.1, 2.8, 4., 1.3],
[6.3, 2.5, 4.9, 1.5],
[6.1, 2.8, 4.7, 1.2],
```

```
[6.7, 3.1, 4.7, 1.5],
[6.3, 2.3, 4.4, 1.3],
[5.6, 3., 4.1, 1.3],
[5.5, 2.5, 4., 1.3],
[5.5, 2.6, 4.4, 1.2],
[6.1, 3., 4.6, 1.4],
[5.8, 2.6, 4., 1.2],
[5., 2.3, 3.3, 1.],
[5.6, 2.7, 4.2, 1.3],
[5.7, 3., 4.2, 1.2],
[5.7, 2.9, 4.2, 1.3],
[6.2, 2.9, 4.3, 1.3],
[5.1, 2.5, 3., 1.1],
[5.7, 2.8, 4.1, 1.3],
[6.3, 3.3, 6. , 2.5],
[5.8, 2.7, 5.1, 1.9],
[7.1, 3., 5.9, 2.1],
[6.3, 2.9, 5.6, 1.8],
[6.5, 3., 5.8, 2.2],
[7.6, 3., 6.6, 2.1],
[4.9, 2.5, 4.5, 1.7],
[7.3, 2.9, 6.3, 1.8],
[6.7, 2.5, 5.8, 1.8],
[7.2, 3.6, 6.1, 2.5],
[6.5, 3.2, 5.1, 2.],
[6.4, 2.7, 5.3, 1.9],
[6.8, 3., 5.5, 2.1],
[5.7, 2.5, 5., 2.],
[5.8, 2.8, 5.1, 2.4],
[6.4, 3.2, 5.3, 2.3],
[6.5, 3., 5.5, 1.8],
[7.7, 3.8, 6.7, 2.2],
[7.7, 2.6, 6.9, 2.3],
[6., 2.2, 5., 1.5],
[6.9, 3.2, 5.7, 2.3],
[5.6, 2.8, 4.9, 2.],
[7.7, 2.8, 6.7, 2.],
[6.3, 2.7, 4.9, 1.8],
[6.7, 3.3, 5.7, 2.1],
[7.2, 3.2, 6., 1.8],
[6.2, 2.8, 4.8, 1.8],
[6.1, 3., 4.9, 1.8],
[6.4, 2.8, 5.6, 2.1],
[7.2, 3., 5.8, 1.6],
[7.4, 2.8, 6.1, 1.9],
[7.9, 3.8, 6.4, 2.],
[6.4, 2.8, 5.6, 2.2],
[6.3, 2.8, 5.1, 1.5],
[6.1, 2.6, 5.6, 1.4],
[7.7, 3., 6.1, 2.3],
[6.3, 3.4, 5.6, 2.4],
[6.4, 3.1, 5.5, 1.8],
[6., 3., 4.8, 1.8], [6.9, 3.1, 5.4, 2.1],
[6.7, 3.1, 5.6, 2.4],
[6.9, 3.1, 5.1, 2.3],
[5.8, 2.7, 5.1, 1.9],
[6.8, 3.2, 5.9, 2.3],
[6.7, 3.3, 5.7, 2.5],
[6.7, 3., 5.2, 2.3],
[6.3, 2.5, 5., 1.9],
[6.5, 3., 5.2, 2.],
[6.2, 3.4, 5.4, 2.3],
[5.9, 3., 5.1, 1.8]])
```

```
df = pd.DataFrame(data=iris.data, columns=iris.feature_names)
df.head()
```

```
sepal length (cm) sepal width (cm) petal length (cm) petal width (cm)
                                                                  0.2
0
                                 3.5
                4.9
                                 3.0
                                                                  0.2
1
                                                  1.4
2
                4.7
                                 3.2
                                                  1.3
                                                                  0.2
3
                                                                  0.2
                4.6
                                 3.1
                                                  1.5
                5.0
                                                                  0.2
                                 3.6
                                                  1.4
In [ ]:
df["target"] = iris.target
df.head()
Out[]:
```

```
sepal length (cm) sepal width (cm) petal length (cm) petal width (cm) target
0
                  5.1
                                    3.5
                                                       1.4
                                                                         0.2
                                                                                   0
1
                  4.9
                                    3.0
                                                                         0.2
                                                       1.4
                                                                                   0
                                                                                   0
2
                                                                         0.2
                  4.7
                                    3.2
                                                       1.3
3
                  4.6
                                    3.1
                                                       1.5
                                                                         0.2
                                                                                   0
                  5.0
                                                                         0.2
                                                                                   0
                                    3.6
                                                       1.4
```

```
In []:
iris.target_names
Out[]:
array(['setosa', 'versicolor', 'virginica'], dtype='<U10')</pre>
```

DataFrame ready to perform

y = df.target
print(X.head())

```
In []:

df["flower_names"] = df.target.apply(lambda x: iris.target_names[x])
    df.head()

Out[]:
```

sepal length (cm) sepal width (cm) petal length (cm) petal width (cm) target flower_names 0 5.1 3.5 1.4 0.2 setosa 1 4.9 3.0 0.2 0 1.4 setosa 2 4.7 3.2 1.3 0.2 0 setosa 3 3.1 1.5 0.2 0 4.6 setosa 5.0 1.4 0.2 0 setosa

```
In []:
len(df)
Out[]:
150
In []:
X = df.drop(["target", "flower names"], axis="columns")
```

```
print(y.head())
   sepal length (cm)
                      sepal width (cm) petal length (cm) petal width (cm)
0
                 5.1
                                   3.5
                                                                         0.2
                                                       1.4
1
                 4.9
                                   3.0
                                                       1.4
                                                                         0.2
2
                 4.7
                                   3.2
                                                       1.3
                                                                         0.2
3
                 4.6
                                   3.1
                                                       1.5
                                                                         0.2
                 5.0
                                   3.6
                                                       1.4
                                                                         0.2
0
1
    0
2
     0
3
     0
4
     0
Name: target, dtype: int64
SVC Classfier
Linear SVC Classifier
In [ ]:
linear SVC classifier = SVC(kernel='linear')
linear SVC classifier
Out[]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='linear',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
train size : test size = 70% : 30%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=0)
# 70% training data, 30% testing data
In [ ]:
print(len(X train))
print(len(y_test))
105
45
In [ ]:
linear_SVC_classifier.fit(X_train, y_train)
Out[]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='linear',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = linear SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
```

Accuracy: 97.777777777777%

Confusion Matrix:

```
[[16 0 0]
[ 0 17 1]
[ 0 0 11]]
```

Classification Report:

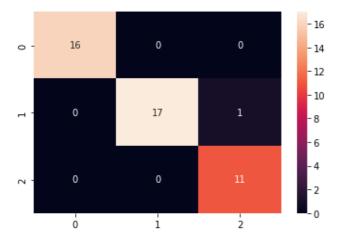
	precision	recall	f1-score	support
0	1.00	1.00	1.00	16 18
2	0.92	1.00	0.96	11
accuracy macro avg weighted avg	0.97 0.98	0.98 0.98	0.98 0.98 0.98	45 45 45

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f21bc5290>



train size : test size = 60% : 40%

In []:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
60% training data, 40% testing data

In []:

```
print(len(X_train))
print(len(y_test))
```

90

60

In []:

```
linear SVC classifier.fit(X train, y train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
```

Accuracy: 96.6666666666667%

Confusion Matrix:

[[16 0 0] [0 22 1] [0 1 20]]

Classification Report:

	precision	recall	f1-score	support
0 1 2	1.00 0.96 0.95	1.00 0.96 0.95	1.00 0.96 0.95	16 23 21
accuracy macro avg weighted avg	0.97 0.97	0.97 0.97	0.97 0.97 0.97	60 60 60

In []:

```
sns.heatmap(cf matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f19629690>



train size : test size = 50% : 50%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
# 50% training data, 50% testing data
```

In []:

```
print(len(X_train))
print(len(y_test))
```

75

```
linear SVC classifier.fit(X train, y train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In []:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 97.333333333333334%

Confusion Matrix:

[[21 0 0] [0 29 1] [0 1 23]]

Classification Report:

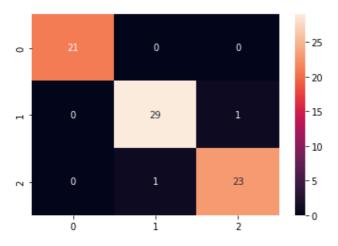
	precision	recall	f1-score	support
0	1.00	1.00	1.00	21
1	0.97	0.97	0.97	30
2	0.96	0.96	0.96	24
accuracy			0.97	75
macro avg	0.98	0.98	0.98	75
weighted avg	0.97	0.97	0.97	75

In []:

```
sns.heatmap(cf matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f8f195676d0>



train size : test size = 40% : 60%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

```
print(len(X_train))
print(len(y_test))
```

```
60
90
```

In []:

```
linear_SVC_classifier.fit(X_train, y_train)
```

Out[]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In []:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print(cf_matrix)
print("\nClassification_Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 96.6666666666667%

Confusion Matrix:

```
[[26 0 0]
[ 0 32 1]
[ 0 2 29]]
```

Classification Report:

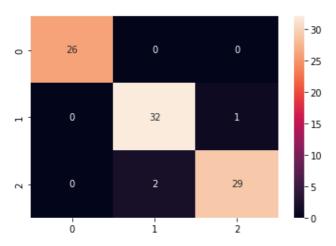
	precision	recall	f1-score	support
0 1 2	1.00 0.94 0.97	1.00 0.97 0.94	1.00 0.96 0.95	26 33 31
accuracy macro avg weighted avg	0.97 0.97	0.97 0.97	0.97 0.97 0.97	90 90 90

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f19515250>



train size: test size = 30%: 70%

```
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.7, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
105
In [ ]:
linear SVC classifier.fit(X_train, y_train)
Out[]:
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = linear SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 95.23809523809523%
Confusion Matrix:
[[33 0 0]
 [ 0 33 1]
 [ 0 4 34]]
Classification Report:
              precision recall f1-score support
           0
                   1.00
                             1.00
                                        1.00
                                                    33
                   0.89
                             0.97
                                        0.93
                                                    34
           1
                   0.97
                                        0.93
                                                    38
                             0.89
                                        0.95
                                                   105
   accuracy
                   0.95
                             0.96
                                        0.95
                                                   105
   macro avq
                   0.95
                             0.95
                                        0.95
                                                   105
weighted avg
In [ ]:
sns.heatmap(cf_matrix, annot=True)
Out[]:
<matplotlib.axes._subplots.AxesSubplot at 0x7f8f193e25d0>
                                      - 30
                  0
                            ٥
       33
                                      - 25
```

- 20

- 15

10

0

33

```
~ - 0 4 34 -5
0 1 2
```

Polynomial SVC Classifier

accuracy

macro avg

```
In [ ]:
poly SVC classifier = SVC(kernel='poly')
poly_SVC_classifier
Out[]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='poly',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
train size: test size = 70%: 30%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
105
45
In [ ]:
poly SVC classifier.fit(X train, y train)
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='poly',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = poly SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
Accuracy: 97.77777777777
Confusion Matrix:
 [[16 0 0]
 [ 0 17 1]
 [ 0 0 11]]
Classification Report:
              precision
                          recall f1-score support
           0
                   1.00
                            1.00
                                       1.00
                                                   16
                            0.94
           1
                   1.00
                                       0.97
                                                   18
                   0.92
                             1.00
                                       0.96
                                                   11
```

0.98

0.98

0.98

0.97

45

45

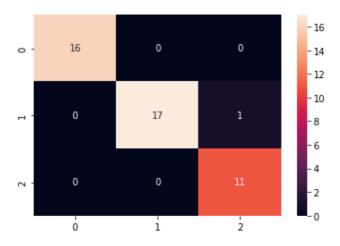
weighted avg 0.98 0.98 0.98 45

```
In [ ]:
```

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f1939d910>



train size : test size = 60% : 40%

```
In [ ]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

90

60

In []:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In []:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 90.0%

```
Confusion Matrix:
```

[[16 0 0] [0 22 1] [0 5 16]]

Classification Report:

precision recall f1-score support

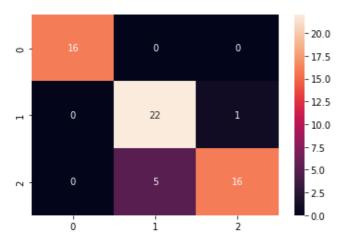
```
1.00
           0
                              1.00
                                         1.00
                                                      16
                              0.96
                                         0.88
           1
                    0.81
                                                      23
                    0.94
                              0.76
                                         0.84
                                                      21
                                         0.90
                                                      60
    accuracy
  macro avg
                    0.92
                              0.91
                                         0.91
                                                      60
weighted avg
                   0.91
                              0.90
                                         0.90
                                                      60
```

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f8f19265ad0>



train size: test size = 50%: 50%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

75 75

In []:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In []:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 92.0%

Confusion Matrix:
 [[21 0 0]

```
[ 0 29 1]
[ 0 5 19]]
```

Classification Report:

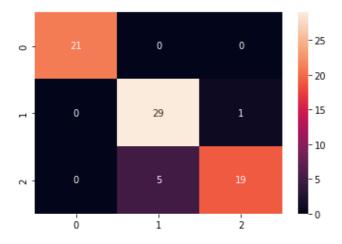
	precision	recall	f1-score	support
0	1.00	1.00	1.00	21
2	0.95	0.79	0.86	24
accuracy macro avo	0.93	0.92 0.92	0.92 0.92 0.92	75 75 75

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f8f1919ee10>



train size : test size = 40% : 60%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

60 90

In []:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
```

Classification Report:

[0 5 26]]

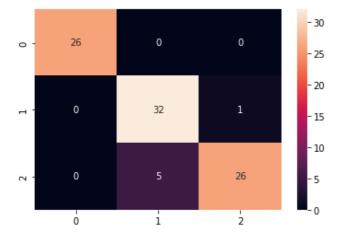
	precision	recall	f1-score	support
0	1.00	1.00	1.00	26
1	0.86	0.97	0.91	33
2	0.96	0.84	0.90	31
accuracy			0.93	90
macro avg	0.94	0.94	0.94	90
weighted avg	0.94	0.93	0.93	90

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f8f19158210>



train size : test size = 30% : 70%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

45 105

In []:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

```
In [ ]:
```

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 94.28571428571428%

Confusion Matrix: [[33 0 0] [0 34 0]

[0 6 32]]

Classification Report:

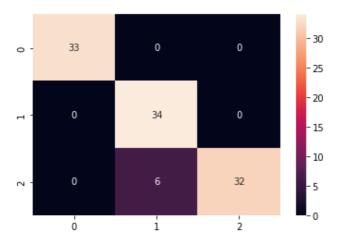
	precision	recall	f1-score	support
0	1.00	1.00	1.00	33
1	0.85	1.00	0.92	34
2	1.00	0.84	0.91	38
accuracy			0.94	105
macro avg	0.95	0.95	0.94	105
weighted avg	0.95	0.94	0.94	105

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f8f190d7310>



Gaussain SVC Classifier

```
In [ ]:
```

```
gaussain_SVC_classifier = SVC(kernel='rbf')
gaussain_SVC_classifier
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

train size : test size = 70% : 30%

```
In [ ]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
In [ ]:
print(len(X train))
print(len(y_test))
105
45
In [ ]:
gaussain SVC classifier.fit(X train, y train)
Out[]:
SVC(C=1.0, break_ties=False, cache_size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion matrix(y test, y pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 97.777777777778
Confusion Matrix:
 [[16 0 0]
 [ 0 17 1]
 [ 0 0 11]]
Classification Report:
                          recall f1-score
              precision
                                             support
           0
                             1.00
                                                    16
                   1.00
                                       1.00
                             0.94
                                        0.97
           1
                   1.00
                                                    18
           2
                   0.92
                             1.00
                                       0.96
                                                    11
   accuracy
                                       0.98
                                                    45
   macro avg
                   0.97
                             0.98
                                       0.98
                                                    45
weighted avg
                   0.98
                             0.98
                                       0.98
                                                    45
```

In []:

sns.heatmap(cf matrix, annot=True)

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f18fa53d0>



```
train size : test size = 60% : 40%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.4, random state=0)
In [ ]:
print(len(X_train))
print(len(y test))
60
In [ ]:
gaussain SVC classifier.fit(X train, y train)
Out[]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy score(y test, y pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
Accuracy: 93.333333333333333
Confusion Matrix:
 [[16 0 0]
 [ 0 22 1]
 [ 0 3 18]]
Classification Report:
              precision recall f1-score
                                             support
                   1.00
           0
                             1.00
                                       1.00
                                                    16
                   0.88
                             0.96
                                       0.92
                                                    23
           1
           2
                   0.95
                             0.86
                                       0.90
                                                    21
   accuracy
                                       0.93
                                                    60
                  0.94
                            0.94
                                      0.94
                                                    60
  macro avg
weighted avg
                  0.94
                             0.93
                                       0.93
                                                    60
In [ ]:
sns.heatmap(cf matrix, annot=True)
Out[]:
<matplotlib.axes. subplots.AxesSubplot at 0x7f8f18efba90>
                                      - 20.0
                  0
0
                                      - 17.5
                                      - 15.0
```

- 12.5

```
- 10.0
- 7.5
- 5.0
- 2.5
- 0.0
```

train size : test size = 50% : 50%

```
In [ ]:
```

```
X train, X test, y train, y test = train test split(X, y, test size=0.5, random state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

75 75

In []:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In []:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 94.66666666666667%

Confusion Matrix: [[21 0 0]

[0 29 1] [0 3 21]]

Classification Report:

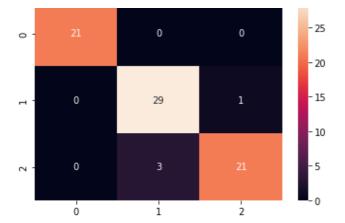
support	f1-score	recall	precision	
21	1.00	1.00	1.00	0
30	0.94	0.97	0.91	1
24	0.91	0.88	0.95	2
75	0.95			accuracy
75	0.95	0.95	0.95	macro avg
75	0.95	0.95	0.95	weighted avg

In []:

```
sns.heatmap(cf matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f18e43e50>



train size: test size = 40%: 60%

```
In [ ]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

60 90

In []:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In []:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 93.3333333333333333

Confusion Matrix:

[[26 0 0] [0 32 1] [0 5 26]]

Classification Report:

support	f1-score	recall	precision	
26	1.00	1.00	1.00	0
33	0.91	0.97	0.86	1
31	0.90	0.84	0.96	2
90	0.93			accuracy
90	0.94	0.94	0.94	macro avg
90	0.93	0.93	0.94	weighted avg

```
sns.heatmap(cf matrix, annot=True)
Out[]:
<matplotlib.axes._subplots.AxesSubplot at 0x7f8f1929f510>
                                         - 30
                    0
                               0
        26
 0
                                         - 25
                                         - 20
        0
                    32
                                         - 15
                                          - 10
        0
                               26
                                          - 5
                    i
        ò
                                ż
train size : test size = 30% : 70%
In [ ]:
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
In [ ]:
```

```
print(len(X train))
print(len(y_test))
```

45 105

In []:

```
gaussain SVC classifier.fit(X train, y train)
```

Out[]:

```
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
   decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
   max_iter=-1, probability=False, random_state=None, shrinking=True,
   tol=0.001, verbose=False)
```

In []:

```
y pred = gaussain SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
```

Accuracy: 88.57142857142857%

Confusion Matrix: [[33 0 0]

[0 34 0] [0 12 26]]

accu

Classification Report:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	33
1	0.74	1.00	0.85	34
2	1.00	0.68	0.81	38
ıracy			0.89	105

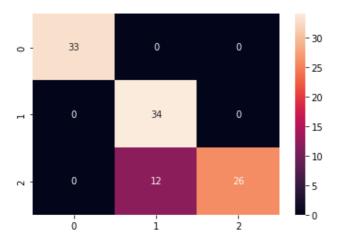
```
macro avg 0.91 0.89 0.89 105 weighted avg 0.92 0.89 0.88 105
```

```
In [ ]:
```

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f18d13e50>



Sigmoid SVC Classifier

```
In [ ]:
```

```
sigmoid_SVC_classifier = SVC(kernel='sigmoid')
sigmoid_SVC_classifier
```

Out[]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

train size : test size = 70% : 30%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

105 45

In []:

```
sigmoid_SVC_classifier.fit(X_train, y_train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
```

```
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 24.4444444444443%

Confusion Matrix:

[[0 0 16] [0 0 18] [0 0 11]]

Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	16
1	0.00	0.00	0.00	18
2	0.24	1.00	0.39	11
accuracy			0.24	45
macro avg	0.08	0.33	0.13	45
weighted avg	0.06	0.24	0.10	45

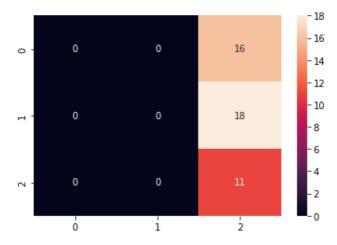
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))

In []:

sns.heatmap(cf matrix, annot=True)

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f8f18bfce50>



train size : test size = 60% : 40%

In []:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)

In []:

```
print(len(X_train))
print(len(y_test))
```

90 60

```
sigmoid_SVC_classifier.fit(X_train, y_train)
Out[]:
```

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In []:

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 26.6666666666668%

Confusion Matrix:

[[16 0 0] [23 0 0] [21 0 0]]

Classification Report:

	precision	recall	f1-score	support
0 1 2	0.27 0.00 0.00	1.00 0.00 0.00	0.42 0.00 0.00	16 23 21
accuracy macro avg weighted avg	0.09	0.33	0.27 0.14 0.11	60 60

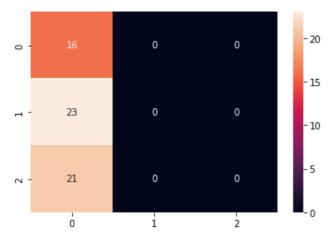
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f240a5590>



train size : test size = 50% : 50%

In []:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)

```
In []:
print(len(X_train))
print(len(y_test))
75
75
```

In []:

```
sigmoid_SVC_classifier.fit(X_train, y_train)
```

Out[]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In []:

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 28.000000000000004%

Confusion Matrix:

[[21 0 0] [30 0 0] [24 0 0]]

Classification Report:

	precision	recall	f1-score	support
0	0.28	1.00	0.44	21
2	0.00	0.00	0.00	24
accuracy			0.28	75
macro avg	0.09	0.33	0.15	75
weighted avg	0.08	0.28	0.12	75

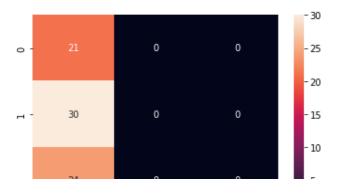
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior. warn prf(average, modifier, msg start, len(result))

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f8f18a70c90>



```
0 1 2
```

train size : test size = 40% : 60%

```
In [ ]:
X_train, X_test, y_train, y_test = train_test_split(X, y, test size=0.6, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
60
90
In [ ]:
sigmoid SVC classifier.fit(X train, y train)
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
   decision function shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
   max iter=-1, probability=False, random state=None, shrinking=True,
   tol=0.001, verbose=False)
In [ ]:
y pred = sigmoid SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Confusion Matrix:
 [[26 0 0]
 [33 0 0]
 [31 0 0]]
Classification Report:
             precision
                         recall f1-score
                                            support
                  0.29
                            1.00
                                      0.45
                                                  26
                            0.00
                                      0.00
                  0.00
                                                  33
                  0.00
                            0.00
                                      0.00
                                                  31
                                      0.29
                                                  90
   accuracy
                  0.10
                            0.33
                                      0.15
                                                  90
  macro avg
                            0.29
                                                  90
                  0.08
                                      0.13
weighted avg
```

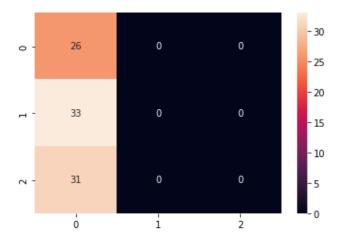
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior. _warn_prf(average, modifier, msg_start, len(result))

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f189b4190>



train size : test size = 30% : 70%

```
In [ ]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

45 105

In []:

```
sigmoid_SVC_classifier.fit(X_train, y_train)
```

Out[]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In []:

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 31.428571428571427%

Confusion Matrix:

[[33 0 0] [34 0 0] [38 0 0]]

Classification Report:

support	f1-score	recall	precision	
33 34 38	0.48 0.00 0.00	1.00 0.00 0.00	0.31 0.00 0.00	0 1 2
105	0.00	0.00	0.00	accuracy
105 105 105	0.16	0.33	0.10 0.10	macro avg weighted avg

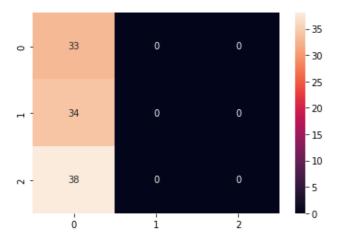
```
MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))
```

```
In [ ]:
```

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f8f188f7390>



MLP Classifier

```
In [ ]:
```

```
mlp_classifier = MLPClassifier(learning_rate='constant', max_iter=600)
mlp_classifier
```

Out[]:

train size: test size = 70%: 30%

```
In [ ]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
# 70% training data, 30% testing data
```

In []:

```
print(len(X_train))
print(len(y_test))
```

105 45

In []:

```
mlp_classifier.fit(X_train, y_train)
```

Out[]:

```
MLPClassifier(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant',
```

learning_rate_init=0.001, max_fun=15000, max_iter=600,
momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True,
power_t=0.5, random_state=None, shuffle=True, solver='adam',
tol=0.0001, validation_fraction=0.1, verbose=False,
warm start=False)

In []:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test, y_pred)}%\n")
cf_matrix = confusion_matrix(y_test, y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test, y_pred))
```

Accuracy: 97.77777777777%

Confusion Matrix:

[[16 0 0] [0 17 1] [0 0 11]]

Classification Report:

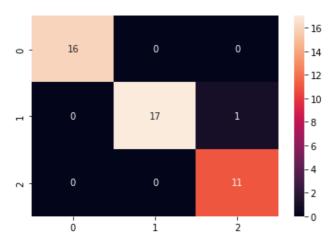
	precision	recall	f1-score	support
0	1.00	1.00	1.00 0.97	16 18
2	0.92	1.00	0.96	11
accuracy macro avg weighted avg	0.97 0.98	0.98 0.98	0.98 0.98 0.98	45 45 45

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f8f18839510>



train size : test size = 60% : 40%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
# 60% training data, 40% testing data
```

```
print(len(X train))
```

```
90
60
In [ ]:
mlp classifier.fit(X train, y train)
Out[]:
MLPClassifier(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9,
              beta 2=0.999, early stopping=False, epsilon=1e-08,
              hidden_layer_sizes=(100,), learning_rate='constant',
              learning rate init=0.001, max fun=15000, max iter=600,
              momentum=0.9, n iter no change=10, nesterovs momentum=True,
              power_t=0.5, random_state=None, shuffle=True, solver='adam',
              tol=0.0001, validation_fraction=0.1, verbose=False,
              warm start=False)
In [ ]:
y pred = mlp classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
Accuracy: 96.6666666666667%
Confusion Matrix:
[[16 0 0]
 [ 0 22 1]
 [ 0 1 20]]
Classification Report:
              precision
                          recall f1-score
                                               support
                             1.00
                                        1.00
                                                    16
                   1.00
                                        0.96
                                                    23
           1
                   0.96
                             0.96
                   0.95
                             0.95
                                        0.95
                                                    21
                                        0.97
                                                    60
    accuracy
                   0.97
                             0.97
   macro avg
                                       0.97
                                                    60
weighted avg
                   0.97
                             0.97
                                        0.97
                                                    60
In [ ]:
sns.heatmap(cf matrix, annot=True)
Out[]:
<matplotlib.axes. subplots.AxesSubplot at 0x7f8f1876f890>
                                      - 20 0
```

print(len(y test))

0

22

1

0

0

0

20

- 17.5 - 15.0 - 12.5

- 10.0 - 7.5 - 5.0

> - 2.5 - 0.0

```
train size: test size = 50%: 50%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.5, random state=0)
# 50% training data, 50% testing data
In [ ]:
print(len(X train))
print(len(y test))
7.5
In [ ]:
mlp classifier.fit(X train, y train)
Out[]:
MLPClassifier(activation='relu', alpha=0.0001, batch size='auto', beta 1=0.9,
              beta 2=0.999, early stopping=False, epsilon=1e-08,
              hidden layer sizes=(100,), learning rate='constant',
              learning rate init=0.001, max fun=15000, max iter=600,
              momentum=0.9, n iter no change=10, nesterovs momentum=True,
              power t=0.5, random state=None, shuffle=True, solver='adam',
              tol=0.0001, validation fraction=0.1, verbose=False,
              warm start=False)
In [ ]:
y pred = mlp classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
Accuracy: 97.333333333333334%
Confusion Matrix:
[[21 0 0]
[ 0 29 1]
 [ 0 1 23]]
Classification Report:
              precision
                          recall f1-score support
           0
                   1.00
                             1.00
                                       1.00
                                                    21
                   0.97
                             0.97
                                       0.97
                                                    30
           1
                   0.96
                             0.96
                                       0.96
                                                    24
                                       0.97
                                                   75
    accuracy
                   0.98
                            0.98
                                       0.98
                                                    75
   macro avg
                             0.97
                                       0.97
                                                    7.5
weighted avg
                   0.97
In [ ]:
sns.heatmap(cf matrix, annot=True)
Out[]:
<matplotlib.axes. subplots.AxesSubplot at 0x7f8f186af650>
```

```
-20
-15
-10
-5
0 1 23 -5
```

train size : test size = 40% : 60%

```
In [ ]:
```

```
X train, X test, y train, y test = train test split(X, y, test size=0.6, random state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

90

In []:

```
mlp_classifier.fit(X_train, y_train)
```

Out[]:

In []:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 97.7777777777778

Confusion Matrix:

[[26 0 0] [0 32 1] [0 1 30]]

Classification Report:

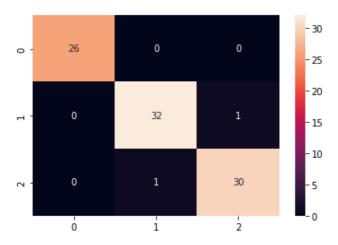
	precision	recall	f1-score	support
0 1 2	1.00 0.97 0.97	1.00 0.97 0.97	1.00 0.97 0.97	26 33 31
accuracy macro avg weighted avg	0.98 0.98	0.98 0.98	0.98 0.98 0.98	90 90 90

```
In [ ]:
```

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f240a50d0>



train size : test size = 30% : 70%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

105

In []:

```
mlp classifier.fit(X train, y train)
```

Out[]:

```
MLPClassifier (activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm start=False)
```

In []:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 97.14285714285714%

Confusion Matrix:

```
[[33 0 0]
[ 0 33 1]
```

[0 2 36]]

Classification Report:

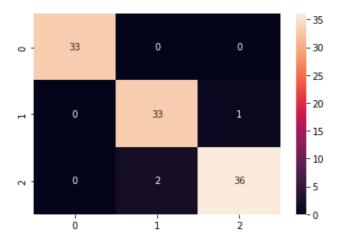
	precision	recall	f1-score	support
0 1 2	1.00 0.94 0.97	1.00 0.97 0.95	1.00 0.96 0.96	33 34 38
accuracy macro avg weighted avg	0.97 0.97	0.97 0.97	0.97 0.97 0.97	105 105 105

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f18b7a050>



Random Forest Classifier

In []:

```
rfc_classifier = RandomForestClassifier(n_estimators=20)
rfc_classifier
```

Out[]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm start=False)
```

train size : test size = 70% : 30%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
# 70% training data, 30% testing data
```

```
print(len(X_train))
print(len(y_test))
```

In []:

```
rfc_classifier.fit(X_train, y_train)
```

Out[]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm_start=False)
```

In []:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 97.77777777777%

Confusion Matrix:

[[16 0 0] [0 17 1] [0 0 11]]

Classification Report:

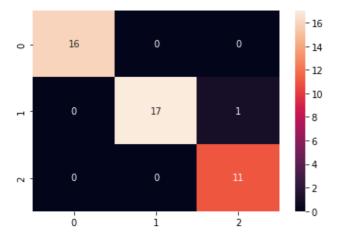
	precision	recall	f1-score	support
0 1 2	1.00 1.00 0.92	1.00 0.94 1.00	1.00 0.97 0.96	16 18 11
accuracy macro avg weighted avg	0.97 0.98	0.98	0.98 0.98 0.98	45 45 45

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f184b8750>



```
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.4, random state=0)
# 60% training data, 40% testing data
In [ ]:
print(len(X train))
print(len(y test))
60
In [ ]:
rfc classifier.fit(X train, y train)
Out[]:
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class weight=None,
                       criterion='gini', max depth=None, max features='auto',
                       max leaf nodes=None, max samples=None,
                       min impurity decrease=0.0, min impurity split=None,
                       min samples leaf=1, min samples split=2,
                       min weight fraction leaf=0.0, n estimators=20,
                       n_jobs=None, oob_score=False, random_state=None,
                       verbose=0, warm start=False)
In [ ]:
y pred = rfc classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion matrix(y test, y pred)
print("Confusion Matrix:")
print(cf matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
Accuracy: 93.3333333333333333
Confusion Matrix:
[[16 0 0]
 [ 0 23 0]
 [ 0 4 17]]
Classification Report:
              precision
                          recall f1-score
                                              support
           0
                   1.00
                             1.00
                                       1.00
                                                    16
                   0.85
                             1.00
                                       0.92
                                                    23
           1
           2
                   1.00
                             0.81
                                       0.89
                                                    21
                                       0.93
                                                    60
    accuracy
                   0.95
                             0.94
                                       0.94
   macro avg
                                                    60
weighted avg
                   0.94
                             0.93
                                       0.93
                                                    60
In [ ]:
sns.heatmap(cf matrix, annot=True)
Out[]:
<matplotlib.axes. subplots.AxesSubplot at 0x7f8f183f9850>
```

- 20

0

0

train size : test size = 60% : 40%

```
-15
-10
-10
-5
-10
-10
-5
-0
```

train size : test size = 50% : 50%

```
In [ ]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
# 50% training data, 50% testing data
```

In []:

```
print(len(X_train))
print(len(y_test))
```

75 75

In []:

```
rfc_classifier.fit(X_train, y_train)
```

Out[]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm start=False)
```

In []:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 93.3333333333333333

Confusion Matrix:

[[21 0 0] [0 29 1] [0 4 20]]

Classification Report:

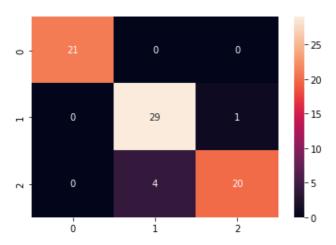
	precision	recall	f1-score	support
0 1 2	1.00 0.88 0.95	1.00 0.97 0.83	1.00 0.92 0.89	21 30 24
2	0.95	0.03	0.09	24
accuracy			0.93	75
macro avg weighted avg	0.94 0.94	0.93 0.93	0.94 0.93	75 75

In []:

```
sns.heatmap(cf matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f1832e890>



train size : test size = 40% : 60%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

60 90

In []:

```
rfc classifier.fit(X train, y train)
```

Out[]:

In []:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 96.6666666666667%

Confusion Matrix:

```
[[26 0 0]
[ 0 33 0]
[ 0 3 28]]
```

Classification Report:

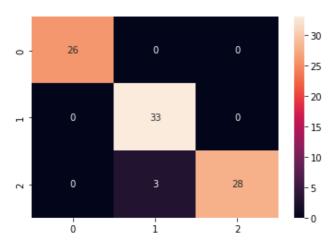
	precision	recall	f1-score	support
0 1 2	1.00 0.92 1.00	1.00 1.00 0.90	1.00 0.96 0.95	26 33 31
accuracy macro avg weighted avg	0.97 0.97	0.97	0.97 0.97 0.97	90 90 90

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f8f18263650>



train size : test size = 30% : 70%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

45 105

In []:

```
rfc_classifier.fit(X_train, y_train)
```

Out[]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm start=False)
```

In []:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
```

```
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 95.23809523809523%

Confusion Matrix:

[[33 0 0] [0 33 1] [0 4 34]]

Classification Report:

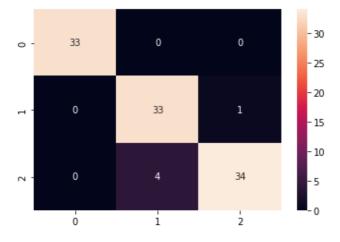
	precision	recall	f1-score	support
0	1.00	1.00	1.00	33
1	0.89	0.97	0.93	34
2	0.97	0.89	0.93	38
accuracy			0.95	105
macro avg	0.95	0.96	0.95	105
weighted avg	0.95	0.95	0.95	105

In []:

sns.heatmap(cf matrix, annot=True)

Out[]:

 ${\tt <matplotlib.axes._subplots.AxesSubplot}$ at ${\tt 0x7f8f181a1890>}$



2. Wine Dataset

Import required modules

['class 0' 'class 1' 'class 2']

```
In [1]:
import numpy as np
import pandas as pd
from sklearn import datasets
from sklearn.metrics import accuracy score, classification report, confusion matrix
from sklearn.model selection import train test split
from sklearn.svm import SVC
import seaborn as sns
from sklearn.neural network import MLPClassifier
from sklearn.ensemble import RandomForestClassifier
Load Dataset
In [2]:
wine = datasets.load wine() # it's source is same as : https://archive.ics.uci.edu/ml/dat
asets/wine
In [3]:
dir(wine)
Out[3]:
['DESCR', 'data', 'feature names', 'target', 'target names']
In [4]:
wine.data
Out[4]:
array([[1.423e+01, 1.710e+00, 2.430e+00, ..., 1.040e+00, 3.920e+00,
        1.065e+03],
       [1.320e+01, 1.780e+00, 2.140e+00, ..., 1.050e+00, 3.400e+00,
        1.050e+03],
       [1.316e+01, 2.360e+00, 2.670e+00, ..., 1.030e+00, 3.170e+00,
        1.185e+03],
       [1.327e+01, 4.280e+00, 2.260e+00, ..., 5.900e-01, 1.560e+00,
       8.350e+02],
       [1.317e+01, 2.590e+00, 2.370e+00, ..., 6.000e-01, 1.620e+00,
        8.400e+021,
       [1.413e+01, 4.100e+00, 2.740e+00, ..., 6.100e-01, 1.600e+00,
        5.600e+0211)
In [5]:
print(wine.feature names)
print(wine.target names)
print(wine.target)
['alcohol', 'malic acid', 'ash', 'alcalinity of ash', 'magnesium', 'total phenols', 'flav
anoids', 'nonflavanoid_phenols', 'proanthocyanins', 'color_intensity', 'hue', 'od280/od31
5_of_diluted_wines', 'proline']
```

```
In [6]:
df = pd.DataFrame(data=wine.data, columns=wine.feature names)
df.head()
Out[6]:
   alcohol malic_acid ash alcalinity_of_ash magnesium total_phenols flavanoids nonflavanoid_phenols proanthocyanins of
0
    14.23
               1.71 2.43
                                 15.6
                                          127.0
                                                      2.80
                                                               3.06
                                                                                 0.28
                                                                                               2.29
 1
    13.20
               1.78 2.14
                                 11.2
                                          100.0
                                                      2.65
                                                               2.76
                                                                                 0.26
                                                                                               1.28
                                          101.0
                                                                                 0.30
2
    13.16
               2.36 2.67
                                 18.6
                                                      2.80
                                                               3 24
                                                                                               2.81
3
    14.37
               1.95 2.50
                                 16.8
                                          113.0
                                                      3.85
                                                               3.49
                                                                                 0.24
                                                                                               2.18
                                                                                 0.39
    13.24
               2.59 2.87
                                 21.0
                                          118.0
                                                      2.80
                                                               2.69
                                                                                               1.82
4
df["target"] = wine.target
df.head()
Out[7]:
   alcohol malic_acid ash alcalinity_of_ash magnesium total_phenols flavanoids nonflavanoid_phenols proanthocyanins of
0
    14.23
               1.71 2.43
                                 15.6
                                          127.0
                                                      2.80
                                                               3.06
                                                                                 0.28
                                                                                               2.29
 1
    13.20
               1.78 2.14
                                 11.2
                                          100.0
                                                      2.65
                                                               2.76
                                                                                 0.26
                                                                                               1.28
2
    13.16
               2.36 2.67
                                 18.6
                                          101.0
                                                      2.80
                                                               3.24
                                                                                 0.30
                                                                                               2.81
3
    14.37
               1.95 2.50
                                 16.8
                                          113.0
                                                      3.85
                                                               3.49
                                                                                 0.24
                                                                                               2.18
 4
    13.24
               2.59 2.87
                                 21.0
                                          118.0
                                                      2.80
                                                               2.69
                                                                                 0.39
                                                                                               1.82
In [8]:
wine.target names
Out[8]:
array(['class_0', 'class_1', 'class_2'], dtype='<U7')</pre>
DataFrame ready to perform
In [9]:
len(df)
Out[9]:
178
In [10]:
X = df.drop(["target"], axis="columns")
y = df.target
print(X.head())
print(y.head())
   alcohol malic acid
                           ash
                                       hue
                                             od280/od315 of diluted wines
                                                                              proline
0
     14.23
                    1.71
                          2.43
                                      1.04
                                                                        3.92
                                                                               1065.0
                                 . . .
                                                                               1050.0
1
     13.20
                    1.78
                          2.14
                                      1.05
                                                                        3.40
                                 . . .
```

1.03

0.86

. . .

3.17

3.45

1185.0

1480.0

2

3

13.16

14.37

2.36

1.95

2.67

2.50

```
1
     Ω
2
     0
3
     0
4
     0
Name: target, dtype: int64
SVC Classfier
Linear SVC Classifier
In [11]:
linear SVC classifier = SVC(kernel='linear')
linear_SVC_classifier
Out[11]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max_iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
train size : test size = 70% : 30%
In [12]:
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
# 70% training data, 30% testing data
In [13]:
print(len(X train))
print(len(y test))
124
54
In [14]:
linear_SVC_classifier.fit(X_train, y_train)
Out[14]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='linear',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [15]:
y pred = linear SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test, y pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
Accuracy: 98.14814814814815%
Confusion Matrix:
[[19 0 0]
 [ 0 21 1]
```

2.93

735.0

13.24

Ω

[5 rows x 13 columns]

2.59 2.87 ... 1.04

```
[ 0 0 13]]
```

Classification Report:

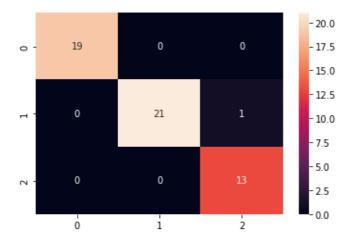
	precision	recall	f1-score	support
0 1 2	1.00 1.00 0.93	1.00 0.95 1.00	1.00 0.98 0.96	19 22 13
accuracy macro avg weighted avg	0.98 0.98	0.98	0.98 0.98 0.98	54 54 54

In [16]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[16]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa5a46ea190>



train size : test size = 60% : 40%

In [17]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
# 60% training data, 40% testing data
```

In [18]:

```
print(len(X_train))
print(len(y_test))
```

106 72

In [19]:

```
linear_SVC_classifier.fit(X_train, y_train)
```

Out[19]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In [20]:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
```

```
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 95.833333333333334%

Confusion Matrix:

[[22 0 0] [0 28 3] [0 0 19]]

Classification Report:

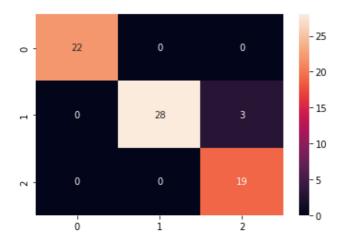
	precision	recall	f1-score	support
0	1.00	1.00	1.00	22
1	1.00	0.90	0.95	31
2	0.86	1.00	0.93	19
accuracy	0.00	1.00	0.93	72
macro avg	0.95	0.97	0.96	72
weighted avg	0.96	0.96		72

In [21]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[21]:

<matplotlib.axes._subplots.AxesSubplot at 0x7fa59c0e8790>



train size : test size = 50% : 50%

In [22]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
# 50% training data, 50% testing data
```

In [23]:

```
print(len(X_train))
print(len(y_test))
```

89

89

In [24]:

```
linear_SVC_classifier.fit(X_train, y_train)
```

Out[24]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
```

```
tol=0.001, verbose=False)
```

In [25]:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 92.13483146067416%

Confusion Matrix:

[[25 0 0] [3 34 3] [1 0 23]]

Classification Report:

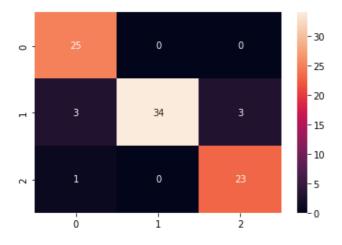
	precision	recall	f1-score	support
0	0.86	1.00	0.93	25
1	1.00	0.85	0.92	40
2	0.88	0.96	0.92	24
accuracy			0.92	89
macro avg	0.92	0.94	0.92	89
weighted avg	0.93	0.92	0.92	89

In [26]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[26]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59c0931d0>



train size : test size = 40% : 60%

In [27]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In [28]:

```
print(len(X_train))
print(len(y_test))
```

71 107

In [29]:

```
linear_SVC_classifier.fit(X_train, y_train)
```

Out[29]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [30]:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print(cf_matrix)
print("\nClassification_Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 88.78504672897196%

Confusion Matrix:

[[34 0 0] [4 38 4] [4 0 23]]

Classification Report:

	precision	recall	f1-score	support
0	0.81	1.00	0.89	34 46
2	0.85	0.85	0.85	27
accuracy			0.89	107
macro avg	0.89	0.89	0.88	107
weighted avg	0.90	0.89	0.89	107

In [31]:

sns.heatmap(cf_matrix, annot=True)

Out[31]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59bfaa3d0>



train size : test size = 30% : 70%

In [32]:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)

In [33]: print(len(X_train)) print(len(y test))

53 125

In [34]:

```
linear_SVC_classifier.fit(X_train, y_train)
```

Out[34]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [35]:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 88.0%

Confusion Matrix:

[[41 1 0] [5 44 3] [5 1 25]]

Classification Report:

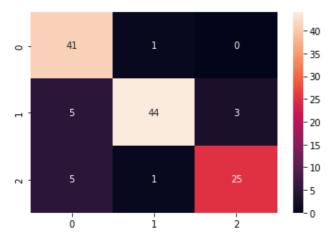
	precision	recall	il-score	support
0 1	0.80	0.98	0.88	42 52
2	0.89	0.81	0.85	31
accuracy			0.88	125
macro avg	0.88	0.88	0.88	125
weighted avg	0.89	0.88	0.88	125

In [36]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[36]:

 ${\tt <matplotlib.axes._subplots.AxesSubplot}$ at ${\tt 0x7fa59bf0f150>}$



Polynomial SVC Classifier

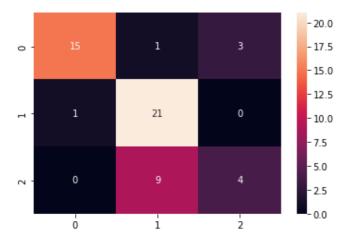
In [42]:

```
In [37]:
poly SVC classifier = SVC(kernel='poly')
poly SVC classifier
Out[37]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
train size: test size = 70%: 30%
In [38]:
X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=0)
In [39]:
print(len(X train))
print(len(y test))
124
54
In [40]:
poly SVC classifier.fit(X train, y train)
Out[40]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='poly',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [41]:
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion matrix(y test, y pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 74.07407407407408%
Confusion Matrix:
 [[15 1 3]
 [ 1 21 0]
 [ 0 9 4]]
Classification Report:
              precision recall f1-score support
           0
                   0.94
                            0.79
                                       0.86
                                                    19
                  0.68
                            0.95
                                      0.79
                                                    22
           1
                   0.57
           2
                             0.31
                                       0.40
                                                    13
                                       0.74
                                                   54
   accuracy
                   0.73
                                                    54
   macro avg
                             0.68
                                       0.68
weighted avg
                   0.74
                             0.74
                                       0.72
                                                    54
```

```
sns.heatmap(cf_matrix, annot=True)
```

Out[42]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59be50990>



train size : test size = 60% : 40%

In [43]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
```

In [44]:

```
print(len(X_train))
print(len(y_test))
```

106 72

In [45]:

```
poly SVC classifier.fit(X train, y train)
```

Out[45]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [46]:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 65.277777777779%

Confusion Matrix:

[[18 4 0] [229 0] [019 0]]

Classification Report:

	precision	recall	f1-score	support
0	0.90	0.82	0.86	22
1	0.56	0.94	0.70	31
2	0.00	0.00	0.00	19

```
      accuracy
      0.65
      72

      macro avg
      0.49
      0.58
      0.52
      72

      weighted avg
      0.52
      0.65
      0.56
      72
```

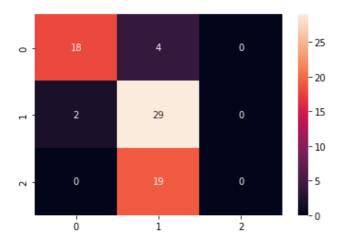
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))

In [47]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[47]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59bd52150>



train size: test size = 50%: 50%

In [48]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
```

In [49]:

```
print(len(X_train))
print(len(y_test))
```

89

In [50]:

```
poly SVC classifier.fit(X train, y train)
```

Out[50]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [51]:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification_Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 70.78651685393258%

Confusion Matrix: [[22 1 2]

```
[ 3 37 0]
[ 3 17 4]]
```

Classification Report:

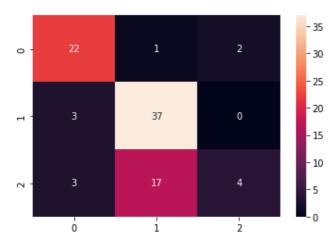
	precision	recall	f1-score	support
0	0.79	0.88	0.83	25
1	0.67	0.93	0.78	40
2	0.67	0.17	0.27	24
accuracy			0.71	89
macro avg	0.71	0.66	0.63	89
weighted avg	0.70	0.71	0.66	89

In [52]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[52]:

<matplotlib.axes._subplots.AxesSubplot at 0x7fa59bc80750>



train size: test size = 40%: 60%

In [53]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In [54]:

```
print(len(X_train))
print(len(y_test))
```

71 107

In [55]:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[55]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In [56]:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
```

print("\nClassification Report:\n") print(classification_report(y_test,y_pred))

Accuracy: 70.09345794392523%

Confusion Matrix:

[[30 1 3] [3 41 2] [4 19 4]]

Classification Report:

	precision	recall	f1-score	support
0	0.81	0.88	0.85	34
1	0.67	0.89	0.77	46
2	0.44	0.15	0.22	27
accuracy			0.70	107
macro avg	0.64	0.64	0.61	107
weighted avg	0.66		0.65	107

In [57]:

```
sns.heatmap(cf matrix, annot=True)
```

Out [57]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59bbb90d0>



train size : test size = 30% : 70%

In [58]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In [59]:

```
print(len(X_train))
print(len(y_test))
```

53 125

In [60]:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[60]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In [61]:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 70.399999999999998

Confusion Matrix:

[[38 1 3] [3 45 4]

[4 22 5]]

Classification Report:

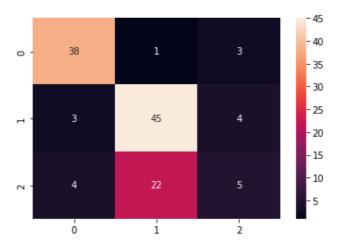
	precision	recall	f1-score	support
0	0.84	0.90	0.87	42
1	0.66	0.87	0.75	52
2	0.42	0.16	0.23	31
accuracy			0.70	125
macro avg	0.64	0.64	0.62	125
weighted avg	0.66	0.70	0.66	125

In [62]:

sns.heatmap(cf matrix, annot=True)

Out[62]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59bbaa250>



Gaussain SVC Classifier

In [63]:

```
gaussain_SVC_classifier = SVC(kernel='rbf')
gaussain_SVC_classifier
```

Out[63]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

train size : test size = 70% : 30%

```
In [64]:
```

In [65]:

```
print(len(X_train))
print(len(y_test))
```

X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=0)

124 54

In [66]:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[66]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [67]:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 77.77777777779%

Confusion Matrix:

[[17 0 2] [1 20 1] [1 7 5]]

Classification Report:

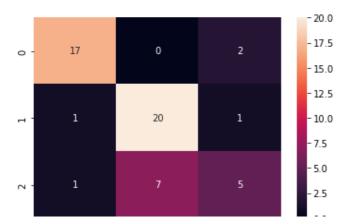
	precision	recall	f1-score	support
0	0.89	0.89	0.89	19
1	0.74	0.91	0.82	22
2	0.62	0.38	0.48	13
accuracy			0.78	54
macro avg	0.75	0.73	0.73	54
weighted avg	0.77	0.78	0.76	54

In [68]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[68]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59ba38ed0>



train size : test size = 60% : 40%

```
In [69]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
```

In [70]:

```
print(len(X_train))
print(len(y_test))
```

106 72

In [71]:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[71]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In [72]:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 68.0555555555556%

Confusion Matrix:

[[20 2 0]

[2 29 0]

[3 16 0]]

Classification Report:

	precision	recall	f1-score	support
0	0.80	0.91	0.85	22
1	0.62	0.94	0.74	31
2	0.00	0.00	0.00	19
accuracy			0.68	72
macro avg	0.47	0.61	0.53	72
weighted avg	0.51	0.68	0.58	72

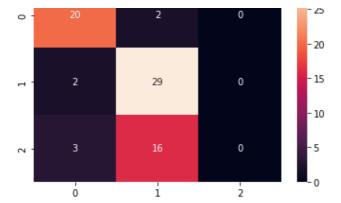
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior. warn prf(average, modifier, msg start, len(result))

In [73]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[73]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59b9826d0>



train size: test size = 50%: 50%

In [74]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
```

In [75]:

```
print(len(X_train))
print(len(y_test))
```

89 89

In [76]:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[76]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [77]:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 67.41573033707866%

Confusion Matrix:

[[22 2 1] [3 37 0] [3 20 1]]

Classification Report:

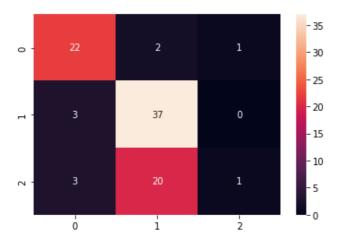
support	f1-score	recall	precision	
25	0.83	0.88	0.79	0
40	0.75	0.93	0.63	1
24	0.08	0.04	0.50	2
89	0.67			accuracy
89	0.55	0.62	0.64	macro avg
89	0.59	0.67	0.64	weighted avg

In [78]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[78]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59b8bc4d0>



train size : test size = 40% : 60%

In [79]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In [80]:

```
print(len(X_train))
print(len(y_test))
```

71 107

In [81]:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[81]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [82]:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 71.96261682242991%

Confusion Matrix:

[[30 0 4] [3 36 7] [4 12 11]]

Classification Report:

	precision	recall	fl-score	support
0	0.81	0.88	0.85	34
1	0.75	0.78	0.77	46
2	0.50	0.41	0.45	27
accuracy			0.72	107
macro avg	0.69	0.69	0.69	107

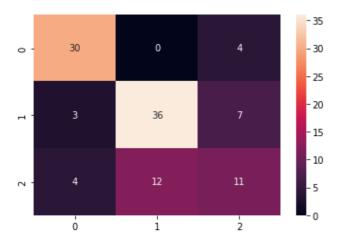
weighted avg 0.71 0.72 0.71 107

```
In [83]:
```

```
sns.heatmap(cf_matrix, annot=True)
```

Out[83]:

<matplotlib.axes._subplots.AxesSubplot at 0x7fa59b8beb90>



train size : test size = 30% : 70%

In [84]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In [85]:

```
print(len(X_train))
print(len(y_test))
```

53 125

In [86]:

```
gaussain SVC classifier.fit(X train, y train)
```

Out[86]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [87]:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 72.0%

```
Confusion Matrix:
[[38 0 4]
[ 3 36 13]
```

[4 11 16]]

Classification Report:

precision recall f1-score support

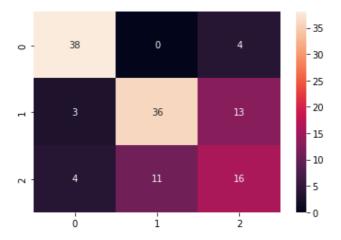
0	0.84	0.90	0.87	42
1	0.77	0.69	0.73	52
2	0.48	0.52	0.50	31
accuracy macro avg weighted avg	0.70 0.72	0.70 0.72	0.72 0.70 0.72	125 125 125

In [88]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[88]:

<matplotlib.axes._subplots.AxesSubplot at 0x7fa59b768750>



Sigmoid SVC Classifier

```
In [89]:
```

```
sigmoid_SVC_classifier = SVC(kernel='sigmoid')
sigmoid_SVC_classifier
```

Out[89]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

train size : test size = 70% : 30%

In [90]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
```

In [91]:

```
print(len(X_train))
print(len(y_test))
```

124 54

In [92]:

sigmoid SVC classifier.fit(X train, y train)

Out[92]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
 max_iter=-1, probability=False, random_state=None, shrinking=True,

```
tol=0.001, verbose=False)
```

In [93]:

```
y pred = sigmoid SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
```

Accuracy: 20.37037037037037%

Confusion Matrix:

[[0 19 0] [11 11 0] [8 5 0]]

Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	19
1	0.31	0.50	0.39	22
2	0.00	0.00	0.00	13
accuracy			0.20	54
macro avg	0.10	0.17	0.13	54
weighted avg	0.13	0.20	0.16	54

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/ classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

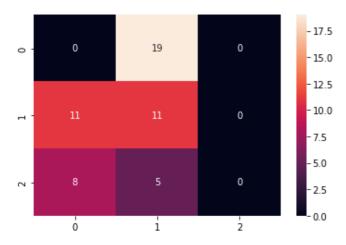
warn prf(average, modifier, msg start, len(result))

In [94]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[94]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59b6b2a90>



train size: test size = 60%: 40%

In [95]:

```
X train, X test, y train, y test = train test split(X, y, test size=0.4, random state=0)
```

In [96]:

```
print(len(X_train))
print(len(y test))
```

```
106
72
```

In [97]:

```
sigmoid_SVC_classifier.fit(X_train, y_train)
```

Out[97]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [98]:

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 18.055555555555554%

Confusion Matrix:

[[2 20 0] [20 11 0] [15 4 0]]

Classification Report:

	precision	recall	f1-score	support
0	0.05	0.09	0.07	22
1	0.31	0.35	0.33	31
2	0.00	0.00	0.00	19
accuracy			0.18	72
macro avg	0.12	0.15	0.13	72
weighted avg	0.15	0.18	0.16	72

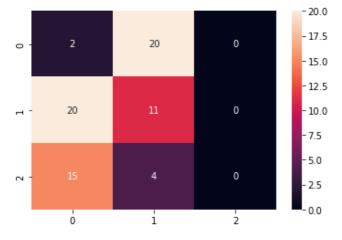
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))

In [99]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[99]:

<matplotlib.axes._subplots.AxesSubplot at 0x7fa59b5e2310>



```
train size: test size = 50%: 50%
In [100]:
X train, X test, y train, y test = train test split(X, y, test size=0.5, random state=0)
In [101]:
print(len(X train))
print(len(y test))
89
In [102]:
sigmoid SVC classifier.fit(X train, y train)
Out[102]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [103]:
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 17.97752808988764%
Confusion Matrix:
 [[7 18 0]
 [31 9 0]
 [24 0 0]]
Classification Report:
              precision recall f1-score support
           0
                   0.11
                            0.28
                                       0.16
                                                    25
           1
                   0.33
                            0.23
                                       0.27
                                                   40
           2
                   0.00
                             0.00
                                       0.00
                                                   24
                                       0.18
                                                   89
   accuracy
                   0.15
                             0.17
                                      0.14
                                                    89
  macro avo
                   0.18
                             0.18
                                       0.17
                                                    89
weighted avg
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/ classification.py:1272: Undefined
MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with
no predicted samples. Use `zero division` parameter to control this behavior.
 _warn_prf(average, modifier, msg_start, len(result))
In [104]:
sns.heatmap(cf matrix, annot=True)
Out[104]:
<matplotlib.axes. subplots.AxesSubplot at 0x7fa59b538b10>
                                     - 30
                            0
                                     - 25
```

```
- 31 9 0 -15
-10 -5
0 1 2
```

train size : test size = 40% : 60%

```
In [105]:
```

X train, X test, y train, y test = train test split(X, y, test size=0.6, random state=0)

In [106]:

```
print(len(X_train))
print(len(y_test))
```

71 107

In [107]:

```
sigmoid SVC classifier.fit(X train, y train)
```

Out[107]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [108]:

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 15.887850467289718%

Confusion Matrix:

[[7 27 0] [36 10 0] [27 0 0]]

Classification Report:

	precision	recall	f1-score	support
0 1 2	0.10 0.27 0.00	0.21 0.22 0.00	0.13 0.24 0.00	34 46 27
accuracy macro avg weighted avg	0.12 0.15	0.14 0.16	0.16 0.13 0.15	107 107 107

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))

In [109]:

```
sns.heatmap(cf matrix, annot=True)
Out[109]:
<matplotlib.axes. subplots.AxesSubplot at 0x7fa59b46cad0>
                                          - 35
                                 0
                                           - 30
                                           - 25
                                           - 20
        36
                    10
                                 0
                                           - 15
                                           - 10
                     0
                                 0
                                 ż
        Ó
                     1
train size : test size = 30% : 70%
In [110]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In [111]:

```
print(len(X train))
print(len(y test))
```

53 125

In [112]:

```
sigmoid SVC classifier.fit(X train, y train)
```

Out[112]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0, decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid', max iter=-1, probability=False, random state=None, shrinking=True, tol=0.001, verbose=False)

In [113]:

```
y pred = sigmoid SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
```

Accuracy: 41.6%

Confusion Matrix:

[[0 42 0] [0 52 0] [0 31 0]]

Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	42
1	0.42	1.00	0.59	52
2	0.00	0.00	0.00	31
accuracy			0.42	125

```
macro avg 0.14 0.33 0.20 125 weighted avg 0.17 0.42 0.24 125
```

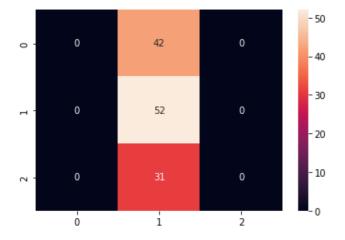
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))

In [114]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[114]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59b3b7850>



MLP Classifier

```
In [115]:
```

```
mlp_classifier = MLPClassifier(learning_rate='constant', max_iter=600)
mlp_classifier
```

Out[115]:

```
MLPClassifier (activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm_start=False)
```

train size : test size = 70% : 30%

In [116]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
# 70% training data, 30% testing data
```

In [117]:

```
print(len(X_train))
print(len(y_test))
```

124

54

In [118]:

```
mlp_classifier.fit(X_train, y_train)
```

Out[118]:

```
MLPClassifier(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm_start=False)
```

In [119]:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print(r\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 38.8888888888889%

Confusion Matrix:

[[19 0 0] [20 2 0] [12 1 0]]

Classification Report:

	precision	recall	f1-score	support
0 1 2	0.37 0.67 0.00	1.00 0.09 0.00	0.54 0.16 0.00	19 22 13
accuracy macro avg weighted avg	0.35 0.40	0.36 0.39	0.39 0.23 0.26	54 54 54

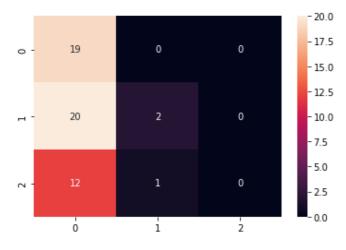
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))

In [120]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[120]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59b2eb810>



train size : test size = 60% : 40%

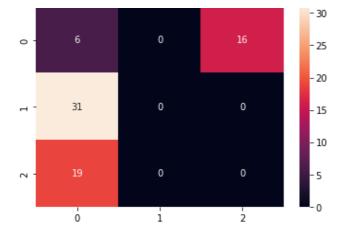
```
In [121]:
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
# 60% training data, 40% testing data
In [122]:
print(len(X train))
print(len(y test))
106
72
In [123]:
mlp_classifier.fit(X_train, y_train)
Out[123]:
MLPClassifier(activation='relu', alpha=0.0001, batch size='auto', beta 1=0.9,
              beta 2=0.999, early stopping=False, epsilon=1e-08,
              hidden_layer_sizes=(100,), learning_rate='constant',
              learning_rate_init=0.001, max_fun=15000, max_iter=600,
              momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True,
              power_t=0.5, random_state=None, shuffle=True, solver='adam',
              tol=0.0001, validation fraction=0.1, verbose=False,
              warm start=False)
In [124]:
y pred = mlp classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 8.33333333333333333
Confusion Matrix:
[[ 6 0 16]
 [31 0 0]
 [19 0 0]]
Classification Report:
                          recall f1-score support
              precision
           Ω
                            0.27
                   0.11
                                       0.15
                                                    22
                   0.00
                             0.00
                                       0.00
           1
                                                    31
                   0.00
                             0.00
                                       0.00
                                                   19
                                                   72
                                       0.08
   accuracy
                   0.04
                             0.09
                                       0.05
                                                   72
  macro avq
weighted avg
                   0.03
                             0.08
                                       0.05
                                                    72
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/ classification.py:1272: Undefined
MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with
no predicted samples. Use `zero division` parameter to control this behavior.
  warn prf(average, modifier, msg start, len(result))
```

In [125]:

Out[125]:

sns.heatmap(cf matrix, annot=True)

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59b22b450>



train size : test size = 50% : 50%

```
In [126]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
# 50% training data, 50% testing data
```

In [127]:

```
print(len(X_train))
print(len(y_test))
```

89 89

In [128]:

```
mlp_classifier.fit(X_train, y_train)
```

Out[128]:

```
MLPClassifier (activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm_start=False)
```

In [129]:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 37.07865168539326%

Confusion Matrix:

[[25 0 0] [10 0 30] [16 0 8]]

Classification Report:

	precision	recall	f1-score	support
0	0.49	1.00	0.66	25
1	0.00	0.00	0.00	40
2	0.21	0.33	0.26	24
accuracy			0.37	89

macro avg 0.23 0.44 0.31 89 weighted avg 0.19 0.37 0.25 89

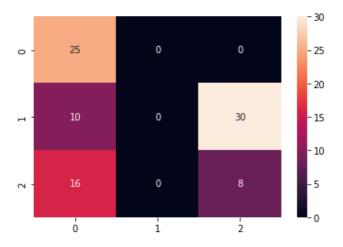
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))

In [130]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[130]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59b167c90>



train size: test size = 40%: 60%

In [131]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In [132]:

```
print(len(X_train))
print(len(y_test))
```

71 107

In [133]:

```
mlp classifier.fit(X train, y train)
```

/usr/local/lib/python3.7/dist-packages/sklearn/neural_network/_multilayer_perceptron.py:5 71: ConvergenceWarning: Stochastic Optimizer: Maximum iterations (600) reached and the optimization hasn't converged yet.

% self.max iter, ConvergenceWarning)

Out[133]:

```
MLPClassifier(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm_start=False)
```

In [134]:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
```

```
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 94.39252336448598%

Confusion Matrix:

```
[[33 1 0]
[ 3 41 2]
[ 0 0 27]]
```

Classification Report:

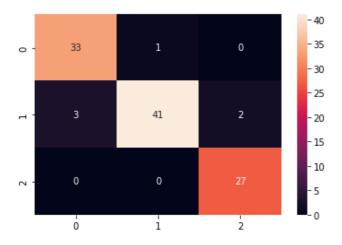
	precision	recall	f1-score	support
0	0.92	0.97	0.94	34
1	0.98	0.89	0.93	46
2	0.93	1.00	0.96	27
accuracy			0.94	107
macro avg	0.94	0.95	0.95	107
weighted avg	0.95	0.94	0.94	107

In [135]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[135]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59b09fbd0>



train size: test size = 30%: 70%

```
In [136]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In [137]:

```
print(len(X_train))
print(len(y_test))
```

53 125

In [138]:

```
mlp_classifier.fit(X_train, y_train)
```

Out[138]:

MLPClassifier(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9,

beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm start=False)

In [139]:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print(cf_matrix)
print("\nClassification_Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 18.4%

Confusion Matrix:

[[1 41 0] [30 22 0] [8 23 0]]

Classification Report:

	precision	recall	f1-score	support
0 1 2	0.03 0.26 0.00	0.02 0.42 0.00	0.02 0.32 0.00	42 52 31
accuracy macro avg weighted avg	0.09 0.12	0.15 0.18	0.18 0.11 0.14	125 125 125

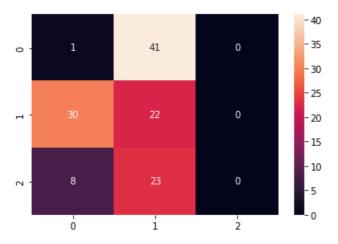
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior. warn prf(average, modifier, msg start, len(result))

In [140]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[140]:

<matplotlib.axes._subplots.AxesSubplot at 0x7fa59b4b7950>



Random Forest Classifier

In [141]:

```
rfc classifier = RandomForestClassifier(n estimators=20)
rfc classifier
Out[141]:
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None,
                       criterion='gini', max depth=None, max features='auto',
                       max leaf nodes=None, max samples=None,
                       min_impurity_decrease=0.0, min_impurity_split=None,
                       min samples leaf=1, min samples split=2,
                       min weight fraction leaf=0.0, n estimators=20,
                       n jobs=None, oob score=False, random state=None,
                       verbose=0, warm start=False)
train size: test size = 70%: 30%
In [142]:
X_train, X_test, y_train, y_test = train test split(X, y, test size=0.3, random state=0)
# 70% training data, 30% testing data
In [143]:
print(len(X train))
print(len(y test))
124
54
In [144]:
rfc classifier.fit(X train, y train)
Out[144]:
RandomForestClassifier(bootstrap=True, ccp alpha=0.0, class weight=None,
                       criterion='gini', max depth=None, max features='auto',
                       max leaf nodes=None, max samples=None,
                       min impurity decrease=0.0, min impurity split=None,
                       min samples leaf=1, min samples split=2,
                       min weight fraction leaf=0.0, n estimators=20,
                       n jobs=None, oob score=False, random state=None,
                       verbose=0, warm start=False)
In [145]:
y pred = rfc classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion matrix(y test, y pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 96.296296296298
Confusion Matrix:
[[19 0 0]
 [ 1 20 1]
 [ 0 0 13]]
Classification Report:
              precision
                          recall f1-score support
                   0.95
                             1.00
                                      0.97
           Ω
                                                   19
                            0.91
           1
                   1.00
                                      0.95
                                                   22
                   0.93
                            1.00
           2
                                      0.96
                                                   13
```

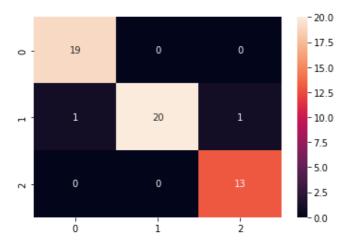
```
accuracy 0.96 54
macro avg 0.96 0.97 0.96 54
weighted avg 0.97 0.96 0.96 54
```

In [146]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[146]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59af17210>



train size : test size = 60% : 40%

In [147]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
# 60% training data, 40% testing data
```

In [148]:

```
print(len(X_train))
print(len(y_test))
```

72

In [149]:

```
rfc_classifier.fit(X_train, y_train)
```

Out[149]:

In [150]:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 94.4444444444444

Confusion Matrix:

```
[[22 0 0]
[2 27 2]
[0 0 19]]
```

Classification Report:

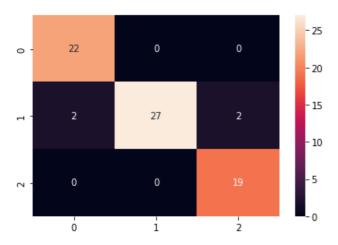
	precision	recall	f1-score	support
0 1	0.92	1.00 0.87	0.96 0.93	22 31
2	0.90	1.00	0.95	19
accuracy			0.94	72
macro avg	0.94	0.96	0.95	72
weighted avg	0.95	0.94	0.94	72

In [151]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[151]:

<matplotlib.axes._subplots.AxesSubplot at 0x7fa59aece9d0>



train size : test size = 50% : 50%

In [152]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
# 50% training data, 50% testing data
```

In [153]:

```
print(len(X_train))
print(len(y_test))
```

89 89

In [154]:

```
rfc_classifier.fit(X_train, y_train)
```

Out[154]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm_start=False)
```

In [155]:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 95.50561797752809%

Confusion Matrix:

[[24 1 0] [2 37 1] [0 0 24]]

Classification Report:

	precision	recall	f1-score	support
0 1 2	0.92 0.97 0.96	0.96 0.93 1.00	0.94 0.95 0.98	25 40 24
accuracy macro avg weighted avg	0.95 0.96	0.96 0.96	0.96 0.96 0.95	89 89 89

In [156]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[156]:

<matplotlib.axes._subplots.AxesSubplot at 0x7fa59ae51450>



train size : test size = 40% : 60%

In [157]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In [158]:

```
print(len(X_train))
print(len(y_test))
```

71 107

In [159]:

```
rfc classifier.fit(X train, y train)
```

Out[159]:

In [160]:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 95.32710280373831%

Confusion Matrix:

[[34 0 0] [3 41 2] [0 0 27]]

Classification Report:

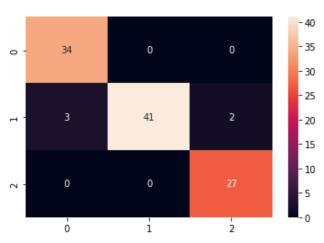
	precision	recall	f1-score	support
0 1 2	0.92 1.00 0.93	1.00 0.89 1.00	0.96 0.94 0.96	34 46 27
accuracy macro avg weighted avg	0.95 0.96	0.96 0.95	0.95 0.95 0.95	107 107 107

In [161]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[161]:

<matplotlib.axes._subplots.AxesSubplot at 0x7fa59ad3ea90>



train size : test size = 30% : 70%

In [162]:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)

```
In [163]:
print(len(X_train))
print(len(y_test))

53
125

In [164]:
rfc_classifier.fit(X_train, y_train)
Out[164]:
```

RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm start=False)

In [165]:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print(r\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 94.399999999999999

Confusion Matrix:

[[42 0 0] [4 45 3] [0 0 31]]

Classification Report:

	precision	recall	il-score	support
0 1 2	0.91 1.00 0.91	1.00 0.87 1.00	0.95 0.93 0.95	42 52 31
accuracy macro avg weighted avg	0.94 0.95	0.96 0.94	0.94 0.95 0.94	125 125 125

In [166]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[166]:

<matplotlib.axes. subplots.AxesSubplot at 0x7fa59ac71690>





3. lonosphere Dataset

Import required modules

```
In [1]:
```

```
import numpy as np
import pandas as pd
from sklearn import datasets
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
import seaborn as sns
from sklearn.neural_network import MLPClassifier
from sklearn.ensemble import RandomForestClassifier
```

Load Dataset

```
In [2]:
```

```
df = pd.read_csv('/content/ionosphere_data.csv')
df.head()
```

```
Out[2]:
```

	column_a	column_b	column_c	column_d	column_e	column_f	column_g	column_h	column_i	column_j	column_k	colu
0	True	False	0.99539	-0.05889	0.85243	0.02306	0.83398	-0.37708	1.00000	0.03760	0.85243	-0.1
1	True	False	1.00000	-0.18829	0.93035	-0.36156	-0.10868	-0.93597	1.00000	-0.04549	0.50874	-0.6
2	True	False	1.00000	-0.03365	1.00000	0.00485	1.00000	-0.12062	0.88965	0.01198	0.73082	0.0
3	True	False	1.00000	-0.45161	1.00000	1.00000	0.71216	-1.00000	0.00000	0.00000	0.00000	0.0
4	True	False	1.00000	-0.02401	0.94140	0.06531	0.92106	-0.23255	0.77152	-0.16399	0.52798	-0.2
4												F

```
In [3]:
```

```
df.column_ai.value_counts()
```

```
Out[3]:
```

g 225 b 126

print(X.head())

Name: column_ai, dtype: int64

DataFrame ready to perform

```
In [4]:
len(df)
Out[4]:
351
In [5]:

X = df.drop(["column_ai"], axis="columns")
y = df.column ai
```

```
print(y.head())
   column_a column_b column_c ... column_af column_ag column_ah
0
       True
                False
                       0.99539
                                       -0.54487
                                                   0.18641
                                                             -0.45300
1
                False
                        1.00000
                                       -0.06288
                                                   -0.13738
                                                              -0.02447
       True
                                 . . .
2
       True
                False
                        1.00000
                                       -0.24180
                                                   0.56045
                                                              -0.38238
                                 . . .
3
       True
                False
                        1.00000
                                 . . .
                                        1.00000
                                                   -0.32382
                                                               1.00000
                        1.00000 ...
                                      -0.59573
                                                  -0.04608
4
       True
                False
                                                             -0.65697
[5 rows x 34 columns]
0
    g
1
     h
2
     g
3
     b
4
Name: column ai, dtype: object
SVC Classfier
Linear SVC Classifier
In [6]:
linear SVC classifier = SVC(kernel='linear')
linear SVC classifier
Out[6]:
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='linear',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
train size : test size = 70% : 30%
In [7]:
X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=0)
# 70% training data, 30% testing data
In [8]:
print(len(X train))
print(len(y_test))
245
106
In [9]:
linear SVC classifier.fit(X train, y train)
Out[9]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='linear',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [10]:
y pred = linear SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
```

print(classification_report(y_test,y_pred))

Accuracy: 86.79245283018868%

Confusion Matrix:

[[31 13] [1 61]]

Classification Report:

	precision	recall	f1-score	support
b	0.97 0.82	0.70 0.98	0.82	44 62
accuracy macro avg weighted avg	0.90 0.88	0.84 0.87	0.87 0.86 0.86	106 106 106

In [11]:

sns.heatmap(cf_matrix, annot=True)

Out[11]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fd4a58650>



train size: test size = 60%: 40%

In [12]:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
60% training data, 40% testing data

In [13]:

```
print(len(X_train))
print(len(y_test))
```

210 141

In [14]:

```
linear_SVC_classifier.fit(X_train, y_train)
```

Out[14]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In [15]:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test, y_pred)}%\n")
cf_matrix = confusion_matrix(y_test, y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test, y_pred))
```

Accuracy: 84.39716312056737%

Confusion Matrix:

[[38 19] [3 81]]

Classification Report:

	precision	recall	f1-score	support
b	0.93	0.67	0.78	57
g	0.81	0.96	0.88	84
accuracy			0.84	141
macro avg	0.87	0.82	0.83	141
weighted avg	0.86	0.84	0.84	141

In [16]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[16]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fca484750>



train size : test size = 50% : 50%

In [17]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
# 50% training data, 50% testing data
```

In [18]:

```
print(len(X_train))
print(len(y_test))
```

175

176

In [19]:

```
linear_SVC_classifier.fit(X_train, y_train)
```

Out[19]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In [20]:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test, y_pred)}%\n")
cf_matrix = confusion_matrix(y_test, y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test, y_pred))
```

Accuracy: 81.25%

Confusion Matrix:

[[44 31] [2 99]]

Classification Report:

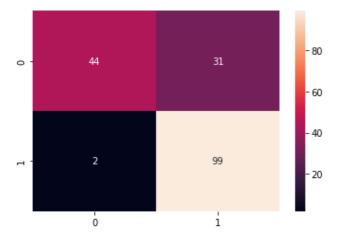
	precision	recall	f1-score	support
b	0.96 0.76	0.59	0.73 0.86	75 101
accuracy macro avg weighted avg	0.86 0.84	0.78 0.81	0.81 0.79 0.80	176 176 176

In [21]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[21]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc9f5da90>



train size : test size = 40% : 60%

In [22]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In [23]:

```
print(len(X_train))
print(len(y_test))
```

140

211

In [24]:

```
linear_SVC_classifier.fit(X_train, y_train)
```

Out[24]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [25]:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 79.14691943127961%

Confusion Matrix:

[[45 43] [1 122]]

Classification Report:

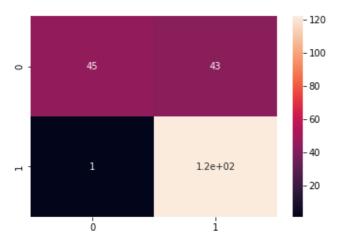
	precision	recall	f1-score	support
b g	0.98 0.74	0.51	0.67 0.85	88 123
accuracy			0.79	211
macro avg	0.86	0.75	0.76	211
weighted avg	0.84	0.79	0.77	211

In [26]:

sns.heatmap(cf matrix, annot=True)

Out[26]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc9e95450>



train size : test size = 30% : 70%

In [27]:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)

In [28]:

```
print(len(X train))
print(len(y_test))
105
246
In [29]:
linear SVC classifier.fit(X train, y train)
Out[29]:
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [30]:
y pred = linear SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
Accuracy: 83.73983739837398%
Confusion Matrix:
```

[[62 38] [2 144]]

Classification Report:

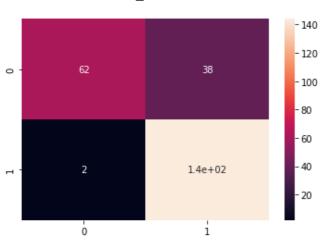
	precision	recall	f1-score	support
b g	0.97 0.79	0.62 0.99	0.76 0.88	100 146
accuracy macro avg weighted avg	0.88 0.86	0.80 0.84	0.84 0.82 0.83	246 246 246

In [31]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[31]:

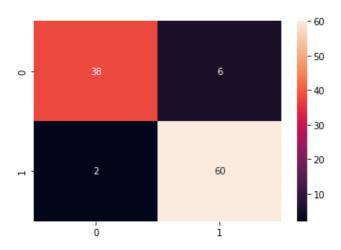
<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc9f65a50>



Polynomial SVC Classifier

```
In [32]:
poly SVC classifier = SVC(kernel='poly')
poly_SVC_classifier
Out[32]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='poly',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
train size : test size = 70% : 30%
In [33]:
X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=0)
In [34]:
print(len(X train))
print(len(y test))
245
106
In [35]:
poly SVC classifier.fit(X train, y train)
Out[35]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='poly',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [36]:
y pred = poly SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion matrix(y test, y pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 92.45283018867924%
Confusion Matrix:
 [[38 6]
 [ 2 60]]
Classification Report:
                          recall f1-score support
              precision
                   0.95
                             0.86
                                       0.90
                                                    44
           b
                   0.91
                             0.97
                                       0.94
                                                    62
                                       0.92
                                                   106
   accuracy
                   0.93
                            0.92
                                       0.92
                                                   106
   macro avq
                   0.93
                             0.92
                                       0.92
                                                   106
weighted avg
In [37]:
sns.heatmap(cf matrix, annot=True)
```

Out[37]:



train size : test size = 60% : 40%

In [38]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
```

In [39]:

```
print(len(X_train))
print(len(y_test))
```

210 141

In [40]:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[40]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [41]:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 75.88652482269504%

Confusion Matrix:

[[23 34] [0 84]]

Classification Report:

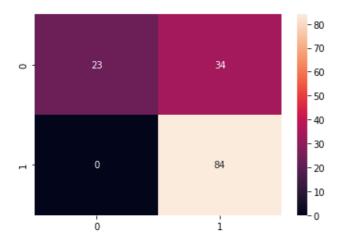
	precision	recall	f1-score	support
g	1.00 0.71	0.40	0.57 0.83	57 84
accuracy macro avg	0.86	0.70	0.76	141 141
weighted avg	0.83	0.76	0.73	141

In [42]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[42]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc9d0a3d0>



train size : test size = 50% : 50%

In [43]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
```

In [44]:

```
print(len(X_train))
print(len(y_test))
```

175176

In [45]:

```
poly SVC classifier.fit(X train, y train)
```

Out[45]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [46]:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 65.3409090909091%

Confusion Matrix:

[[14 61] [0 101]]

support	f1-score	recall	precision	
75 101	0.31 0.77	0.19	1.00 0.62	b g
176 176	0.65 0.54	0.59	0.81	accuracy macro avg

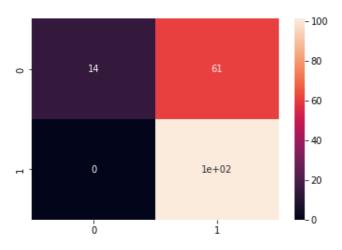
weighted avg 0.78 0.65 0.57 176

In [47]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[47]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fc9c31bd0>



train size : test size = 40% : 60%

In [48]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In [49]:

```
print(len(X_train))
print(len(y_test))
```

140

211

In [50]:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[50]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0, decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly', max_iter=-1, probability=False, random_state=None, shrinking=True, tol=0.001, verbose=False)
```

In [51]:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 63.507109004739334%

Confusion Matrix:

[[11 77] [0 123]]

Classification Report:

precision recall f1-score support

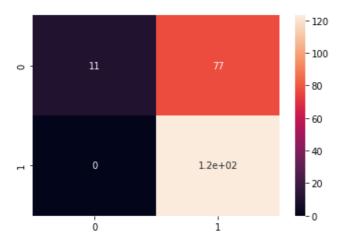
```
b
                   1.00
                           0.12
                                     0.22
                                                   88
                   0.61
                             1.00
                                      0.76
                                                  123
           g
                                       0.64
                                                  211
   accuracy
                            0.56
                                       0.49
                                                  211
                  0.81
  macro avg
                  0.78
                            0.64
                                       0.54
                                                  211
weighted avg
```

In [52]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out [52]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fc9c8f610>



train size: test size = 30%: 70%

In [53]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In [54]:

```
print(len(X_train))
print(len(y_test))
```

105

246

In [55]:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[55]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [56]:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 63.82113821138211%

Confusion Matrix:

[[11 89]

[0 146]]

Classification Report:

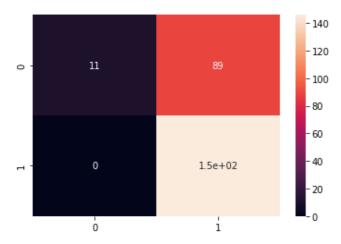
	precision	recall	f1-score	support
b g	1.00 0.62	0.11	0.20 0.77	100 146
accuracy macro avg weighted avg	0.81 0.78	0.56 0.64	0.64 0.48 0.54	246 246 246

In [57]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out [57]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fc9b11050>



Gaussain SVC Classifier

In [58]:

```
gaussain_SVC_classifier = SVC(kernel='rbf')
gaussain_SVC_classifier
```

Out[58]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

train size: test size = 70%: 30%

In [59]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
```

In [60]:

```
print(len(X_train))
print(len(y_test))
```

245

106

In [61]:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[61]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In [62]:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 95.28301886792453%

Confusion Matrix:

[[40 4] [161]]

Classification Report:

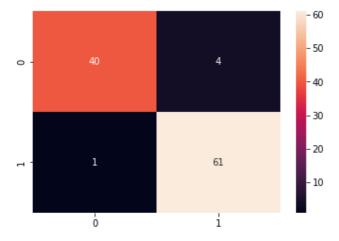
	precision	recall	f1-score	support
b g	0.98 0.94	0.91 0.98	0.94	44 62
accuracy macro avg weighted avg	0.96 0.95	0.95 0.95	0.95 0.95 0.95	106 106 106

In [63]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[63]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc9a3d310>



train size : test size = 60% : 40%

In [64]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
```

In [65]:

```
print(len(X_train))
print(len(y_test))
```

210

141

In [66]:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[66]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In [67]:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 94.32624113475178%

Confusion Matrix:

[[50 7] [1 83]]

Classification Report:

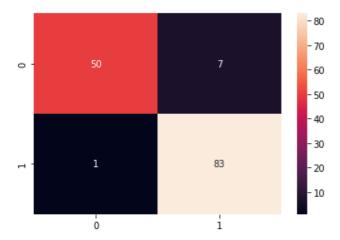
support	f1-score	recall	precision	
57 84	0.93 0.95	0.88	0.98 0.92	g
141	0.94			accuracy
141	0.94	0.93	0.95	macro avg
141	0.94	0.94	0.95	weighted avg

In [68]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[68]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc995f510>



train size : test size = 50% : 50%

In [69]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
```

In [70]:

```
print(len(X_train))
print(len(y_test))
```

```
175176
```

In [71]:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[71]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [72]:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 93.75%

Confusion Matrix:

[[65 10] [1 100]]

Classification Report:

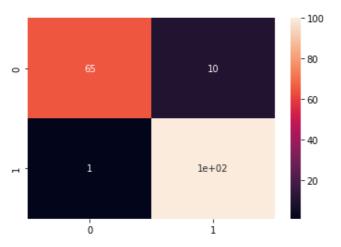
	precision	recall	f1-score	support
g	0.98 0.91	0.87	0.92 0.95	75 101
accuracy macro avg weighted avg	0.95 0.94	0.93 0.94	0.94 0.93 0.94	176 176 176

In [73]:

sns.heatmap(cf_matrix, annot=True)

Out[73]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fc9908110>



train size: test size = 40%: 60%

In [74]:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)

In [75]:

```
print(len(X_train))
print(len(y_test))
```

140 211

In [76]:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[76]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [77]:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 90.99526066350711%

Confusion Matrix:

[[69 19] [0 123]]

Classification Report:

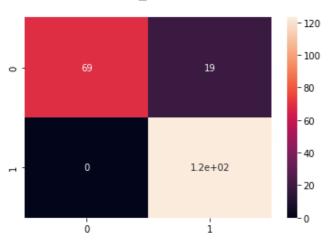
	precision	recall	f1-score	support
g p	1.00 0.87	0.78 1.00	0.88	88 123
accuracy macro avg weighted avg	0.93 0.92	0.89 0.91	0.91 0.90 0.91	211 211 211

In [78]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[78]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fc9cea990>



train size : test size = 30% : 70%

```
In [79]:
```

```
In [80]:
```

```
print(len(X_train))
print(len(y_test))
```

X train, X test, y train, y test = train test split(X, y, test size=0.7, random state=0)

105

246

In [81]:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[81]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [82]:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 90.2439024390244%

Confusion Matrix:

[[76 24] [0 146]]

Classification Report:

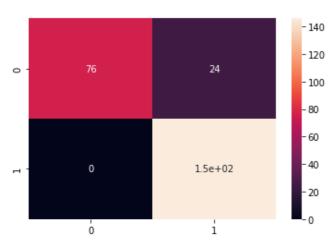
	precision	recall	f1-score	support
b b	1.00	0.76	0.86	100 146
accuracy macro avg weighted avg	0.93	0.88	0.90 0.89 0.90	246 246 246

In [83]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[83]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fc97ad810>



Sigmoid SVC Classifier

In [89]:

```
In [84]:
sigmoid SVC classifier = SVC(kernel='sigmoid')
sigmoid SVC classifier
Out[84]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max_iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
train size: test size = 70%: 30%
In [85]:
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
In [86]:
print(len(X train))
print(len(y_test))
245
106
In [87]:
sigmoid SVC classifier.fit(X train, y train)
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
In [88]:
y pred = sigmoid SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 84.90566037735849%
Confusion Matrix:
 [[29 15]
 [ 1 61]]
Classification Report:
              precision recall f1-score
                                             support
                   0.97
                            0.66
                                       0.78
                                                    44
                             0.98
                                       0.88
                   0.80
                                                    62
                                       0.85
                                                  106
   accuracy
                                      0.83
                  0.88
                           0.82
                                                  106
   macro avg
                  0.87
                             0.85
                                       0.84
                                                  106
weighted avg
```

```
sns.heatmap(cf_matrix, annot=True)
Out[89]:
```

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc96dcdd0>



train size : test size = 60% : 40%

```
In [90]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
```

In [91]:

```
print(len(X_train))
print(len(y_test))
```

210 141

In [92]:

```
sigmoid_SVC_classifier.fit(X_train, y_train)
```

Out[92]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [93]:

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 82.97872340425532%

Confusion Matrix:

[[34 23] [1 83]]

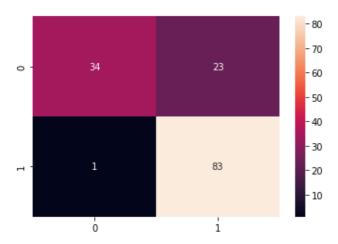
support	f1-score	recall	precision	
57 84	0.74 0.87	0.60 0.99	0.97 0.78	b g
141 141 141	0.83 0.81 0.82	0.79 0.83	0.88 0.86	accuracy macro avg weighted avg

```
In [94]:
```

```
sns.heatmap(cf matrix, annot=True)
```

Out[94]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc9673850>



train size : test size = 50% : 50%

In [95]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
```

In [96]:

```
print(len(X_train))
print(len(y_test))
```

175

176

In [97]:

```
sigmoid SVC classifier.fit(X train, y train)
```

Out[97]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In [98]:

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 83.522727272727278

Confusion Matrix:

[[48 27] [2 99]]

	precision	recall	f1-score	support
b	0.96	0.64	0.77	75
g	0.79	0.98	0.87	101

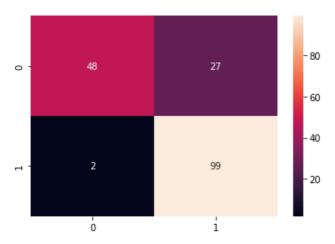
```
accuracy 0.84 176 macro avg 0.87 0.81 0.82 176 weighted avg 0.86 0.84 0.83 176
```

In [99]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out [99]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc95aa290>



train size: test size = 40%: 60%

```
In [100]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In [101]:

```
print(len(X_train))
print(len(y_test))
```

140

211

In [102]:

```
sigmoid_SVC_classifier.fit(X_train, y_train)
```

Out[102]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In [103]:

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 81.99052132701422%

```
Confusion Matrix:
  [[ 51  37]
  [ 1 122]]
```

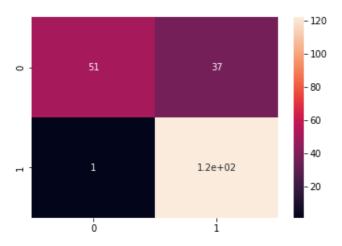
	precision	recall	f1-score	support
g p	0.98 0.77	0.58	0.73 0.87	88 123
accuracy			0.82	211
macro avg	0.87	0.79	0.80	211
weighted avg	0.86	0.82	0.81	211

In [104]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[104]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc94da4d0>



train size : test size = 30% : 70%

In [105]:

```
X train, X test, y train, y test = train test split(X, y, test size=0.7, random state=0)
```

In [106]:

```
print(len(X_train))
print(len(y_test))
```

105

246

In [107]:

```
sigmoid SVC classifier.fit(X_train, y_train)
```

Out[107]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In [108]:

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 82.11382113821138%

Confusion Matrix:

```
[[ 56 44]
[ 0 146]]
```

Classification Report:

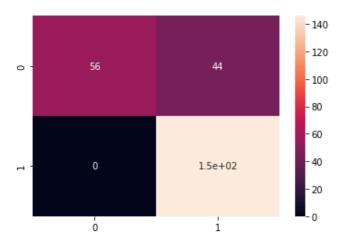
	precision	recall	f1-score	support
g	1.00 0.77	0.56 1.00	0.72 0.87	100 146
accuracy			0.82	246
macro avg	0.88	0.78	0.79	246
weighted avg	0.86	0.82	0.81	246

In [109]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[109]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc9476dd0>



MLP Classifier

In [110]:

```
mlp_classifier = MLPClassifier(learning_rate='constant', max_iter=600)
mlp_classifier
```

Out[110]:

MLPClassifier(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm start=False)

train size : test size = 70% : 30%

In [111]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
# 70% training data, 30% testing data
```

In [112]:

```
print(len(X_train))
print(len(y_test))
```

In [113]:

```
mlp_classifier.fit(X_train, y_train)
```

Out[113]:

MLPClassifier (activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm_start=False)

In [114]:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 93.39622641509435%

Confusion Matrix:

[[37 7] [0 62]]

Classification Report:

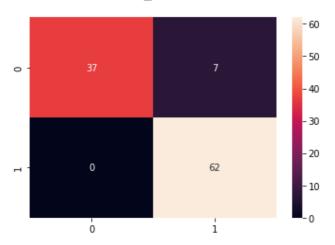
	precision	recall	f1-score	support
b g	1.00	0.84	0.91 0.95	44 62
accuracy macro avg weighted avg	0.95 0.94	0.92 0.93	0.93 0.93 0.93	106 106 106

In [115]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[115]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc93b3250>



train size : test size = 60% : 40%

```
In [116]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
# 60% training data, 40% testing data
```

In [117]:

```
print(len(X_train))
print(len(y_test))
```

210 141

In [118]:

```
mlp_classifier.fit(X_train, y_train)
```

Out[118]:

MLPClassifier (activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm start=False)

In [119]:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 90.0709219858156%

Confusion Matrix:

[[43 14] [0 84]]

Classification Report:

	precision	recall	f1-score	support
g	1.00	0.75 1.00	0.86 0.92	57 84
accuracy macro avg weighted avg	0.93 0.91	0.88	0.90 0.89 0.90	141 141 141

In [120]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[120]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc9340950>



```
-30
-20
-10
0 1
```

train size : test size = 50% : 50%

```
In [121]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
# 50% training data, 50% testing data
```

In [122]:

```
print(len(X_train))
print(len(y_test))
```

175

176

In [123]:

```
mlp_classifier.fit(X_train, y_train)
```

Out[123]:

```
MLPClassifier (activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm start=False)
```

In [124]:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 85.79545454545455%

Confusion Matrix:
[[50 25]

[0 101]]

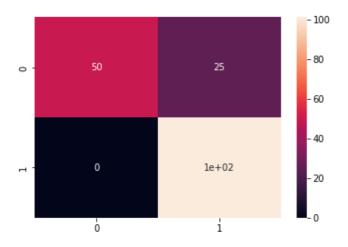
Classification Report:

	precision	recall	f1-score	support
þ	1.00	0.67 1.00	0.80	75 101
accuracy			0.86	176
macro avg	0.90	0.83	0.84	176
weighted avg	0.89	0.86	0.85	176

In [125]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[125]:



train size: test size = 40%: 60%

```
In [126]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In [127]:

```
print(len(X_train))
print(len(y_test))
```

140 211

In [128]:

```
mlp_classifier.fit(X_train, y_train)
```

Out[128]:

```
MLPClassifier (activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=le-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm start=False)
```

In [129]:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 84.36018957345972%

Confusion Matrix:

[[56 32] [1 122]]

	precision	recall	f1-score	support
b	0.98	0.64	0.77	88
g	0.79	0.99	0.88	123

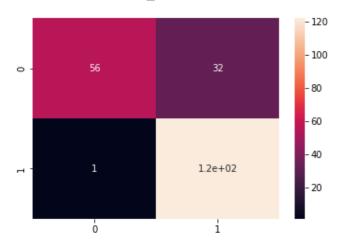
```
accuracy 0.84 211 macro avg 0.89 0.81 0.83 211 weighted avg 0.87 0.84 0.84 211
```

In [130]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[130]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fc91ac8d0>



train size: test size = 30%: 70%

In [131]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In [132]:

```
print(len(X_train))
print(len(y_test))
```

105

246

In [133]:

```
mlp_classifier.fit(X_train, y_train)
```

Out[133]:

MLPClassifier(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm start=False)

In [134]:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 84.5528455284553%

Confusion Matrix:

```
[[ 62 38]
[ 0 146]]
```

Classification Report:

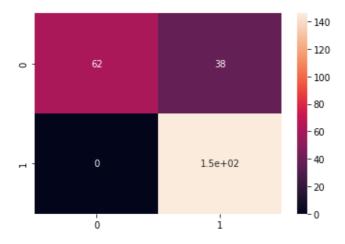
	precision	recall	f1-score	support
b g	1.00	0.62 1.00	0.77	100 146
accuracy macro avg weighted avg	0.90	0.81 0.85	0.85 0.83 0.84	246 246 246

In [135]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[135]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc90e4a50>



Random Forest Classifier

In [136]:

```
rfc_classifier = RandomForestClassifier(n_estimators=20)
rfc_classifier
```

Out[136]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm start=False)
```

train size: test size = 70%: 30%

In [137]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
# 70% training data, 30% testing data
```

In [138]:

```
print(len(X_train))
print(len(y_test))
```

```
245
106
```

In [139]:

```
rfc_classifier.fit(X_train, y_train)
```

Out[139]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm start=False)
```

In [140]:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 95.28301886792453%

Confusion Matrix:

[[41 3] [2 60]]

Classification Report:

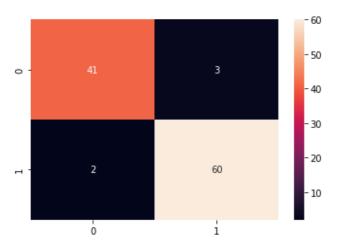
support	f1-score	recall	precision	
= =	0.94 0.96	0.93 0.97	0.95 0.95	b
106	0.95 0.95 0.95	0.95 0.95	0.95 0.95	accuracy macro avg weighted avg

In [141]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[141]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc95aa750>



train size: test size = 60%: 40%

```
In [142]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
# 60% training data, 40% testing data
```

In [143]:

```
print(len(X_train))
print(len(y_test))
```

210 141

In [144]:

```
rfc_classifier.fit(X_train, y_train)
```

Out[144]:

In [145]:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 90.0709219858156%

Confusion Matrix:

[[45 12] [2 82]]

Classification Report:

	precision	recall	f1-score	support
g p	0.96 0.87	0.79	0.87 0.92	57 84
accuracy macro avg weighted avg	0.91 0.91	0.88	0.90 0.89 0.90	141 141 141

In [146]:

```
sns.heatmap(cf matrix, annot=True)
```

Out[146]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f0fc8f97410>



```
2 82 -20 -10
```

train size : test size = 50% : 50%

```
In [147]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
# 50% training data, 50% testing data
```

In [148]:

```
print(len(X_train))
print(len(y_test))
```

175176

In [149]:

```
rfc_classifier.fit(X_train, y_train)
```

Out[149]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm_start=False)
```

In [150]:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 92.04545454545455%

Confusion Matrix:

[[62 13] [1 100]]

Classification Report:

support	f1-score	recall	precision	
75 101	0.90 0.93	0.83	0.98 0.88	þ
176	0.92			accuracy
176	0.92	0.91	0.93	macro avg
176	0.92	0.92	0.93	weighted avg

In [151]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[151]: <matplotlib.axes. subplots.AxesSubplot at 0x7f0fc8f42090> - 100 - 80 13 0 60 1e + 02ò train size : test size = 40% : 60% In [152]: X train, X test, y train, y test = train test split(X, y, test size=0.6, random state=0) In [153]: print(len(X train)) print(len(y test)) 140 211 In [154]: rfc classifier.fit(X train, y train) Out[154]: RandomForestClassifier(bootstrap=True, ccp alpha=0.0, class weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min impurity decrease=0.0, min impurity split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n jobs=None, oob score=False, random state=None, verbose=0, warm_start=False) In [155]: y pred = rfc classifier.predict(X test) print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n") cf_matrix = confusion_matrix(y_test,y_pred) print("Confusion Matrix:\n") print(cf_matrix) print("\nClassification Report:\n") print(classification report(y test, y pred)) Accuracy: 90.04739336492891% Confusion Matrix: [[75 13] [8 115]] Classification Report:

precision recall f1-score support

0.88

0.92

88

123

0.85

0.93

0.90

0.90

h

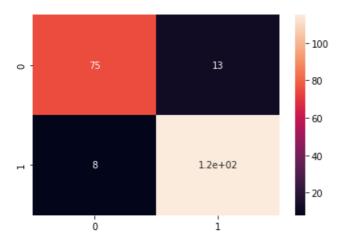
```
accuracy 0.90 211 macro avg 0.90 0.89 0.90 211 weighted avg 0.90 0.90 0.90 211
```

In [156]:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[156]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fc8e6b350>



train size: test size = 30%: 70%

In [157]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In [158]:

```
print(len(X_train))
print(len(y_test))
```

105

246

In [159]:

```
rfc_classifier.fit(X_train, y_train)
```

Out[159]:

In [160]:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 89.83739837398373%

Confusion Matrix:

[[84 16] [9 137]]

Classification Report:

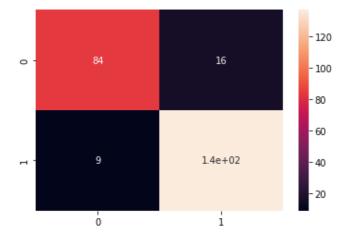
	precision	recall	f1-score	support
b	0.90 0.90	0.84	0.87 0.92	100 146
accuracy			0.90	246
macro avg	0.90	0.89	0.89	246
weighted avg	0.90	0.90	0.90	246

In [161]:

sns.heatmap(cf_matrix, annot=True)

Out[161]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f0fc8df4810>



4. Breast Cancer Dataset

Import required modules

```
In [ ]:
import numpy as np
import pandas as pd
from sklearn import datasets
from sklearn.metrics import accuracy score, classification report, confusion matrix
from sklearn.model selection import train test split
from sklearn.svm import SVC
import seaborn as sns
from sklearn.neural network import MLPClassifier
from sklearn.ensemble import RandomForestClassifier
Load Dataset
In [ ]:
b cancer = datasets.load breast cancer() # it's source is same as : https://archive.ics.u
ci.edu/ml/datasets/wine
In [ ]:
dir(b cancer)
Out[]:
['DESCR', 'data', 'feature names', 'filename', 'target', 'target names']
In [ ]:
b cancer.data
Out[]:
array([[1.799e+01, 1.038e+01, 1.228e+02, ..., 2.654e-01, 4.601e-01,
        1.189e-01],
       [2.057e+01, 1.777e+01, 1.329e+02, ..., 1.860e-01, 2.750e-01,
        8.902e-021,
       [1.969e+01, 2.125e+01, 1.300e+02, ..., 2.430e-01, 3.613e-01,
        8.758e-02],
       [1.660e+01, 2.808e+01, 1.083e+02, ..., 1.418e-01, 2.218e-01,
       7.820e-02],
       [2.060e+01, 2.933e+01, 1.401e+02, ..., 2.650e-01, 4.087e-01,
        1.240e-01],
       [7.760e+00, 2.454e+01, 4.792e+01, ..., 0.000e+00, 2.871e-01,
        7.039e-0211)
In [ ]:
print(b cancer.feature names)
print(b_cancer.target_names)
print(b cancer.target)
['mean radius' 'mean texture' 'mean perimeter' 'mean area'
 'mean smoothness' 'mean compactness' 'mean concavity'
 'mean concave points' 'mean symmetry' 'mean fractal dimension'
```

'radius error' 'texture error' 'perimeter error' 'area error' 'smoothness error' 'compactness error' 'concavity error'

'concave points error' 'symmetry error' 'fractal dimension error' 'worst radius' 'worst texture' 'worst perimeter' 'worst area'

```
'worst smoothness' 'worst compactness' 'worst concavity'
      'worst concave points' 'worst symmetry' 'worst fractal dimension']
['malignant' 'benign']
 1 \;\; 0 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 0 \;\; 0 \;\; 1 \;\; 1 \;\; 0 \;\; 1 \;\; 1 \;\; 1 \;\; 0 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 0 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\;
      1 \;\; 0 \;\; 1 \;\; 1 \;\; 0 \;\; 1 \;\; 0 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 0 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\;
      1 \;\; 1 \;\; 0 \;\; 1 \;\; 0 \;\; 1 \;\; 0 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 1 \;\; 0 \;\; 0 \;\; 1 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\; 0 \;\;
      1 1 1 1 1 1 1 0 0 0 0 0 0 1
```

df = pd.DataFrame(data=b_cancer.data, columns=b_cancer.feature_names)
df.head()

Out[]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	radius error	texture error
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0.2419	0.07871	1.0950	0.9053
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0.1812	0.05667	0.5435	0.7339
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0.2069	0.05999	0.7456	0.7869
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	0.2597	0.09744	0.4956	1.1560
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0.1809	0.05883	0.7572	0.7813
4)

In []:

```
df["target"] = b_cancer.target
df.head()
```

Out[]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	radius error	texture error
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0.2419	0.07871	1.0950	0.9053
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0.1812	0.05667	0.5435	0.7339
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0.2069	0.05999	0.7456	0.7869
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	0.2597	0.09744	0.4956	1.1560
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0.1809	0.05883	0.7572	0.7813
4												P

In []:

```
df.target.value_counts()
```

Out[]:

1 357 0 212

Name: target, dtype: int64

DataFrame ready to perform

train size : test size = 70% : 30%

70% training data, 30% testing data

linear SVC classifier.fit(X train, y train)

In []:

In []:

171

In []:

print(len(X train)) print(len(y test))

```
In [ ]:
len(df)
Out[]:
569
In [ ]:
X = df.drop(["target"], axis="columns")
y = df.target
print(X.head())
print(y.head())
   mean radius mean texture ... worst symmetry worst fractal dimension
                      10.38 ...
0
         17.99
                                          0.4601
                                                                   0.11890
                      17.77 ...
1
         20.57
                                          0.2750
                                                                   0.08902
                      21.25 ...
2
         19.69
                                          0.3613
                                                                   0.08758
3
         11.42
                      20.38 ...
                                          0.6638
                                                                   0.17300
4
        20.29
                      14.34 ...
                                          0.2364
                                                                   0.07678
[5 rows x 30 columns]
    0
1
     0
2
     0
3
     0
Name: target, dtype: int64
SVC Classfier
Linear SVC Classifier
In [ ]:
linear SVC classifier = SVC(kernel='linear')
linear SVC classifier
Out[]:
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='linear',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
```

X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=0)

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In []:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 95.90643274853801%

Confusion Matrix:

[[61 2] [5 103]]

Classification Report:

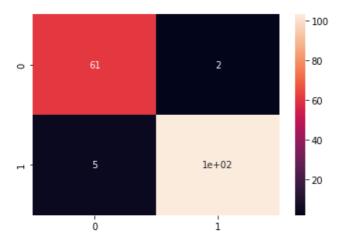
	precision	recall	f1-score	support
0	0.92	0.97	0.95	63
1	0.98	0.95	0.97	108
accuracy			0.96	171
macro avg	0.95	0.96	0.96	171
weighted avg	0.96	0.96	0.96	171

In []:

```
sns.heatmap(cf matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed652ca90>



train size : test size = 60% : 40%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
# 60% training data, 40% testing data
```

In []:

```
print(len(X_train))
print(len(y_test))
```

```
341
228
```

```
linear_SVC_classifier.fit(X_train, y_train)
```

Out[]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In []:

```
y_pred = linear_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 97.36842105263158%

Confusion Matrix:

[[80 3] [3 142]]

Classification Report:

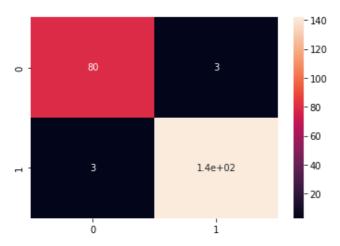
	precision	recall	f1-score	support
0 1	0.96 0.98	0.96	0.96 0.98	83 145
accuracy macro avg weighted avg	0.97 0.97	0.97 0.97	0.97 0.97 0.97	228 228 228

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed64fe090>



train size : test size = 50% : 50%

In []:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)

```
# 50% training data, 50% testing data
In [ ]:
print(len(X_train))
print(len(y test))
284
285
In [ ]:
linear SVC classifier.fit(X train, y train)
Out[]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='linear',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = linear SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 95.78947368421052%
Confusion Matrix:
[[ 95 6]
 [ 6 178]]
Classification Report:
              precision
                          recall f1-score
                                              support
           0
                   0.94
                             0.94
                                       0.94
                                                   101
                   0.97
                             0.97
                                       0.97
                                                   184
                                       0.96
                                                   285
   accuracy
```

0.95

0.96

285

285

In []:

macro avg

weighted avg

sns.heatmap(cf_matrix, annot=True)

0.95

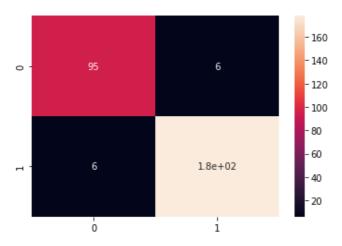
0.96

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed642f3d0>

0.95

0.96

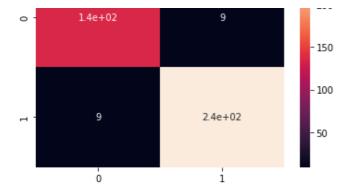


```
train size : test size = 40% : 60%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.6, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
342
In [ ]:
linear SVC classifier.fit(X train, y train)
Out[]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='linear',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = linear SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion matrix(y test, y pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
Accuracy: 95.90643274853801%
Confusion Matrix:
[[118 5]
9 210]]
Classification Report:
                          recall f1-score support
              precision
           0
                   0.93
                            0.96
                                       0.94
                                                   123
                   0.98
                            0.96
                                       0.97
                                                   219
           1
                                        0.96
                                                   342
   accuracy
   macro avg
                  0.95
                            0.96
                                       0.96
                                                   342
weighted avg
                   0.96
                             0.96
                                       0.96
                                                   342
In [ ]:
sns.heatmap(cf matrix, annot=True)
<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed63ca210>
                                      - 200
                                      - 175
        1.2e+02
0
                                      - 150
                                      - 125
                                      - 100
```

```
2.1e+02
train size : test size = 30% : 70%
In [ ]:
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
In [ ]:
print(len(X train))
print(len(y_test))
170
399
In [ ]:
linear SVC classifier.fit(X train, y train)
Out[]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='linear',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = linear SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion matrix(y test, y pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
Accuracy: 95.48872180451127%
Confusion Matrix:
[[137 9]
 [ 9 244]]
Classification Report:
                          recall f1-score support
              precision
           0
                   0.94
                             0.94
                                       0.94
                                                   146
                   0.96
                             0.96
                                       0.96
           1
                                                   253
                                       0.95
                                                   399
   accuracy
                   0.95
                             0.95
                                       0.95
                                                   399
  macro avg
                   0.95
                             0.95
                                       0.95
                                                   399
weighted avg
In [ ]:
sns.heatmap(cf_matrix, annot=True)
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed62fe750>

Out[]:



Polynomial SVC Classifier

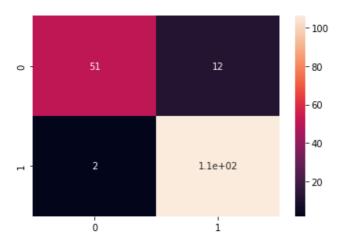
```
In [ ]:
poly SVC classifier = SVC(kernel='poly')
poly SVC classifier
Out[]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
train size: test size = 70%: 30%
In [ ]:
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
In [ ]:
print(len(X train))
print(len(y test))
398
171
In [ ]:
poly SVC classifier.fit(X train, y train)
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
Accuracy: 91.81286549707602%
Confusion Matrix:
 [[ 51 12]
 [ 2 106]]
Classification Report:
              precision recall f1-score support
```

```
0
                   0.96
                             0.81
                                        0.88
                                                     63
           1
                   0.90
                              0.98
                                        0.94
                                                   108
                                        0.92
                                                   171
   accuracy
                   0.93
                              0.90
                                        0.91
                                                   171
  macro avg
weighted avg
                  0.92
                             0.92
                                        0.92
                                                   171
```

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed6225310>



train size : test size = 60% : 40%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

341

228

In []:

```
poly SVC classifier.fit(X train, y train)
```

Out[]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In []:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 91.666666666666666

Confusion Matrix:
 [[65 18]

[1 144]]

Classification Report:

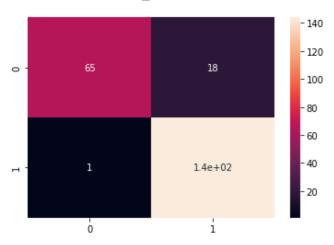
	precision	recall	f1-score	support
0 1	0.98	0.78 0.99	0.87 0.94	83 145
accuracy macro avg	0.94	0.89	0.92	228 228
weighted avg	0.92	0.92	0.91	228

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed621b950>



train size : test size = 50% : 50%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

284 285

In []:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0, decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly', max_iter=-1, probability=False, random_state=None, shrinking=True, tol=0.001, verbose=False)
```

In []:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

```
Accuracy: 90.87719298245615%
Confusion Matrix:
```

[[77 24] [2 182]]

Classification Report:

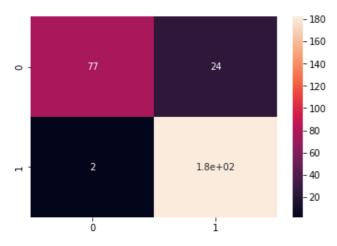
	precision	recall	f1-score	support
0 1	0.97 0.88	0.76 0.99	0.86 0.93	101 184
accuracy macro avg weighted avg	0.93 0.92	0.88 0.91	0.91 0.89 0.91	285 285 285

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed6159050>



train size: test size = 40%: 60%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

227 342

In []:

```
poly_SVC_classifier.fit(X_train, y_train)
```

Out[]

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In []:

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
```

```
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 90.05847953216374%

Confusion Matrix:

[[94 29] [5 214]]

Classification Report:

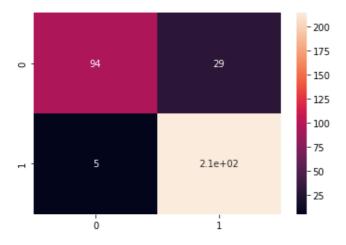
	precision	recall	f1-score	support
0	0.95 0.88	0.76 0.98	0.85 0.93	123 219
accuracy macro avg weighted avg	0.92 0.91	0.87	0.90 0.89 0.90	342 342 342

In []:

```
sns.heatmap(cf matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed608b150>



train size : test size = 30% : 70%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
```

In []:

```
print(len(X_train))
print(len(y_test))
```

170

399

In []:

```
poly SVC classifier.fit(X train, y train)
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='poly',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

```
y_pred = poly_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 89.72431077694235%

Confusion Matrix: [[108 38] [3 250]]

Classification Report:

	precision	recall	f1-score	support
0	0.97	0.74	0.84	146
1	0.87	0.99	0.92	253
accuracy			0.90	399
macro avg	0.92	0.86	0.88	399
weighted avg	0.91	0.90	0.89	399

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed5fb4ad0>



Gaussain SVC Classifier

```
In [ ]:
```

```
gaussain_SVC_classifier = SVC(kernel='rbf')
gaussain_SVC_classifier
```

Out[]:

```
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

train size: test size = 70%: 30%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
```

```
In [ ]:
```

```
print(len(X_train))
print(len(y_test))
```

398 171

In []:

```
gaussain_SVC_classifier.fit(X_train, y_train)
```

Out[]:

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
 max_iter=-1, probability=False, random_state=None, shrinking=True,
 tol=0.001, verbose=False)

In []:

```
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 92.39766081871345%

Confusion Matrix:

[[51 12] [1 107]]

Classification Report:

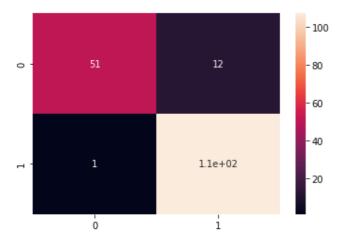
	precision	recall	fl-score	support
0	0.98	0.81	0.89	63
1	0.90	0.99	0.94	108
accuracy			0.92	171
macro avg	0.94	0.90	0.91	171
weighted avg	0.93	0.92	0.92	171

In []:

```
sns.heatmap(cf matrix, annot=True)
```

Out[]:

 ${\tt <matplotlib.axes._subplots.AxesSubplot}$ at ${\tt 0x7f4ed5f5d4d0>}$



```
train size : test size = 60% : 40%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.4, random state=0)
In [ ]:
print(len(X train))
print(len(y_test))
341
228
In [ ]:
gaussain SVC classifier.fit(X train, y train)
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = gaussain SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
Accuracy: 92.10526315789474%
Confusion Matrix:
 [[ 66 17]
 [ 1 144]]
Classification Report:
              precision recall f1-score support
                   0.99
           Λ
                             0.80
                                        0.88
                                                    83
                   0.89
                             0.99
                                        0.94
           1
                                                   145
                                        0.92
                                                   228
   accuracy
                  0.94
                            0.89
                                      0.91
                                                   228
  macro avq
                  0.93
                             0.92
                                       0.92
                                                   228
weighted avg
In [ ]:
sns.heatmap(cf matrix, annot=True)
Out[]:
<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed6fcf4d0>
                                     - 140
                                      - 120
          66
                          17
0
                                      - 100
                                      - 80
```

- 60

40

1.4e+02

```
train size: test size = 50%: 50%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.5, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
284
285
In [ ]:
gaussain SVC classifier.fit(X train, y train)
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 91.57894736842105%
Confusion Matrix:
 [[ 78 23]
 [ 1 183]]
Classification Report:
              precision
                          recall f1-score
                                               support
           0
                   0.99
                             0.77
                                        0.87
                                                   101
           1
                   0.89
                              0.99
                                        0.94
                                                   184
                                        0.92
                                                   285
   accuracy
                   0.94
                             0.88
                                        0.90
                                                   285
  macro avg
                   0.92
                             0.92
                                        0.91
                                                   285
weighted avg
In [ ]:
sns.heatmap(cf matrix, annot=True)
Out[]:
<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed5df8210>
                                     - 175
                                      - 150
                          23
0
                                      - 125
                                      - 100
```

```
1.8e+02
train size : test size = 40% : 60%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.6, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
227
342
In [ ]:
gaussain SVC classifier.fit(X train, y train)
Out[]:
SVC(C=1.0, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y_pred = gaussain_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
Accuracy: 90.64327485380117%
Confusion Matrix:
 [[ 94 29]
 [ 3 216]]
Classification Report:
                          recall f1-score support
              precision
           0
                   0.97
                             0.76
                                        0.85
                                                   123
           1
                   0.88
                             0.99
                                        0.93
                                                   219
                                        0.91
                                                   342
   accuracy
                   0.93
                             0.88
                                       0.89
                                                   342
   macro avg
weighted avg
                   0.91
                             0.91
                                       0.90
                                                   342
In [ ]:
sns.heatmap(cf matrix, annot=True)
Out[]:
<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed5d6d350>
```

- 200 - 175



train size: test size = 30%: 70%

```
In [ ]:
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
In [ ]:
print(len(X train))
print(len(y_test))
170
399
In [ ]:
gaussain SVC classifier.fit(X train, y train)
Out[]:
SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = gaussain SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 90.47619047619048%
Confusion Matrix:
 [[110 36]
 [ 2 251]]
Classification Report:
              precision recall f1-score support
                   0.98
                           0.75
           0
                                       0.85
                                                  146
           1
                   0.87
                             0.99
                                       0.93
                                                  253
                                       0.90
                                                  399
   accuracy
                  0.93
                            0.87
                                       0.89
```

In []:

macro avq weighted avg

```
sns.heatmap(cf_matrix, annot=True)
```

0.90

399

399

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed5cb6990>

0.90

0.91



[[10 53] [39 69]]

```
Sigmoid SVC Classifier
In [ ]:
sigmoid SVC classifier = SVC(kernel='sigmoid', C=0.9)
sigmoid SVC classifier
Out[]:
SVC(C=0.9, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
train size: test size = 70%: 30%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
398
171
In [ ]:
sigmoid SVC classifier.fit(X train, y train)
Out[]:
SVC(C=0.9, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = sigmoid SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion matrix(y test, y pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\\n")
print(classification report(y test, y pred))
Accuracy: 46.198830409356724%
Confusion Matrix:
```

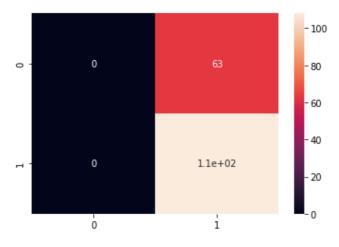
```
Classification Report:
          precision recall f1-score
                                 support
              0.20
                    0.16
                            0.18
                                      6.3
        1
              0.57
                     0.64
                             0.60
                                     108
                             0.46
   accuracy
                                     171
              0.38
                    0.40
                             0.39
                                     171
  macro avq
             0.43
                     0.46
                             0.44
                                     171
weighted avg
In [ ]:
from sklearn.model selection import GridSearchCV
In [ ]:
grid = GridSearchCV(SVC(),param grid,refit=True,verbose=2)
grid.fit(X train, y train)
Fitting 5 folds for each of 16 candidates, totalling 80 fits
[CV] C=0.1, gamma=1, kernel=sigmoid .....
[CV] ...... C=0.1, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=1, kernel=sigmoid ......
[CV] ...... C=0.1, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=1, kernel=sigmoid .....
[CV] ...... C=0.1, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=1, kernel=sigmoid .....
[CV] ...... C=0.1, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=1, kernel=sigmoid .....
[CV] ...... C=0.1, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.1, kernel=sigmoid .....
[CV] \dots C=0.1, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.1, kernel=sigmoid .....
[CV] \dots C=0.1, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.1, kernel=sigmoid .....
[CV] ...... C=0.1, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.1, kernel=sigmoid .....
[CV] ...... C=0.1, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.1, kernel=sigmoid .....
[CV] ...... C=0.1, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.01, kernel=sigmoid .....
[CV] ...... C=0.1, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=0.1, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.01, kernel=sigmoid ......
[CV] ..... C=0.1, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.01, kernel=sigmoid .....
[CV] ...... C=0.1, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=0.1, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.001, kernel=sigmoid .......
[CV] ..... C=0.1, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.001, kernel=sigmoid ......
[CV] ...... C=0.1, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.001, kernel=sigmoid ......
[CV] ...... C=0.1, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=0.1, gamma=0.001, kernel=sigmoid ......
[Parallel(n jobs=1)]: Using backend SequentialBackend with 1 concurrent workers.
[Parallel(n jobs=1)]: Done 1 out of 1 | elapsed:
                                       0.0s remaining:
[CV] ...... C=0.1, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=1, kernel=sigmoid ......
[CV] ...... C=1, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=1, kernel=sigmoid ......
[CV] ...... C=1, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=1, kernel=sigmoid ......
[CV] ...... C=1, gamma=1, kernel=sigmoid, total= 0.0s
```

```
[CV] C=1, gamma=1, kernel=sigmoid ......
[CV] ...... C=1, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=1, kernel=sigmoid ......
[CV] ...... C=1, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.1, kernel=sigmoid ......
[CV] C=1, gamma=0.1, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.1, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.1, kernel=sigmoid .....
[CV] ...... C=1, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.1, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.001, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.001, kernel=sigmoid .....
[CV] ...... C=1, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.001, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.001, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=1, gamma=0.001, kernel=sigmoid ......
[CV] ...... C=1, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=1, kernel=sigmoid ......
[CV] ...... C=10, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=1, kernel=sigmoid ......
[CV] ...... C=10, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=1, kernel=sigmoid ......
[CV] ...... C=10, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=1, kernel=sigmoid ......
[CV] ...... C=10, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=1, kernel=sigmoid ......
[CV] ..... C=10, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.1, kernel=sigmoid .....
[CV] ...... C=10, gamma=0.1, kernel=sigmoid, total= 0.0s [CV] C=10, gamma=0.1, kernel=sigmoid .....
[CV] ...... C=10, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.1, kernel=sigmoid .....
[CV] ...... C=10, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.1, kernel=sigmoid ......
[CV] ...... C=10, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.1, kernel=sigmoid ......
[CV] ...... C=10, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=10, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=10, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=10, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=10, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.01, kernel=sigmoid ......
[CV] ...... C=10, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.001, kernel=sigmoid ......
[CV] ...... C=10, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.001, kernel=sigmoid ......
[CV] ...... C=10, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.001, kernel=sigmoid ......
[CV] ...... C=10, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=10, gamma=0.001, kernel=sigmoid .....
[CV] ..... C=10, gamma=0.001, kernel=sigmoid, total= 0.0s
```

```
[CV] C=10, gamma=0.001, kernel=sigmoid ......
[CV] \dots C=10, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=1, kernel=sigmoid ......
[CV] ...... C=100, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=1, kernel=sigmoid .....
[CV] ...... C=100, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=1, kernel=sigmoid .....
[CV] ...... C=100, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=1, kernel=sigmoid ......
[CV] ...... C=100, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=1, kernel=sigmoid .....
[CV] ...... C=100, gamma=1, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.1, kernel=sigmoid .......................
[CV] ...... C=100, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.1, kernel=sigmoid ......
[CV] ...... C=100, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.1, kernel=sigmoid ......
[CV] ...... C=100, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.1, kernel=sigmoid ......
[CV] ...... C=100, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.1, kernel=sigmoid ......
[CV] ..... C=100, gamma=0.1, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.01, kernel=sigmoid ......
[CV] ..... C=100, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.01, kernel=sigmoid .....
[CV] ..... C=100, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.01, kernel=sigmoid ......
[CV] ..... C=100, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.01, kernel=sigmoid ......
[CV] ..... C=100, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.01, kernel=sigmoid .....
[CV] ..... C=100, gamma=0.01, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.001, kernel=sigmoid .....
[CV] ...... C=100, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.001, kernel=sigmoid .....
[CV] ...... C=100, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.001, kernel=sigmoid .....
[CV] ...... C=100, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.001, kernel=sigmoid .....
[CV] ..... C=100, gamma=0.001, kernel=sigmoid, total= 0.0s
[CV] C=100, gamma=0.001, kernel=sigmoid .....
[CV] ...... C=100, gamma=0.001, kernel=sigmoid, total= 0.0s
[Parallel(n jobs=1)]: Done 80 out of 80 | elapsed:
Out[]:
GridSearchCV(cv=None, error score=nan,
          estimator=SVC(C=1.0, break ties=False, cache size=200,
                     class weight=None, coef0=0.0,
                     decision function shape='ovr', degree=3,
                     gamma='scale', kernel='rbf', max_iter=-1,
                     probability=False, random state=None, shrinking=True,
                     tol=0.001, verbose=False),
          iid='deprecated', n_jobs=None,
          param grid={'C': [0.1, 1, 10, 100], 'gamma': [1, 0.1, 0.01, 0.001],
                    'kernel': ['sigmoid']},
          pre dispatch='2*n jobs', refit=True, return train score=False,
          scoring=None, verbose=2)
In [ ]:
param grid = {'C': [0.1,1, 10, 100], 'gamma': [1,0.1,0.01,0.001], 'kernel': ['sigmoid']}
In [ ]:
print(grid.best estimator )
SVC(C=0.1, break ties=False, cache size=200, class weight=None, coef0=0.0,
   decision function shape='ovr', degree=3, gamma=1, kernel='sigmoid',
   max iter=-1, probability=False, random state=None, shrinking=True,
   tol=0.001, verbose=False)
```

```
import matplotlib.pyplot as plt
grid_predictions = grid.predict(X_test)
print(confusion_matrix(y_test,grid_predictions))
plt.show(sns.heatmap(confusion_matrix(y_test,grid_predictions), annot=True))
print(classification_report(y_test,grid_predictions))
print("Accuracy Score of RBF kernel", accuracy_score(y_test,grid_predictions))
```

```
[[ 0 63]
[ 0 108]]
```



support	f1-score	recall	precision	
63 108	0.00 0.77	0.00	0.00 0.63	0 1
171 171 171	0.63 0.39 0.49	0.50 0.63	0.32 0.40	accuracy macro avg weighted avg

Accuracy Score of RBF kernel 0.631578947368421

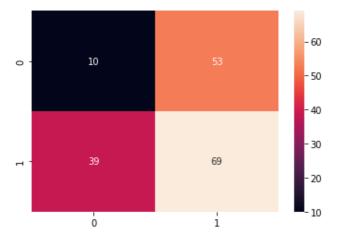
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1272: Undefined MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed6f6be90>



train size: test size = 60%: 40%

In []:

```
X train, X test, y train, y test = train test split(X, y, test size=0.4, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
341
228
In [ ]:
sigmoid SVC classifier.fit(X train, y train)
SVC(C=0.9, break_ties=False, cache_size=200, class weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion matrix(y test, y pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 47.80701754385965%
Confusion Matrix:
 [[15 68]
 [51 94]]
Classification Report:
              precision
                           recall f1-score
                                              support
           0
                   0.23
                              0.18
                                        0.20
                                                    83
           1
                   0.58
                              0.65
                                        0.61
                                                   145
                                        0.48
                                                   228
    accuracy
                   0.40
                             0.41
                                        0.41
                                                   228
   macro avg
                              0.48
                                        0.46
                                                   228
weighted avg
                   0.45
```

sns.heatmap(cf matrix, annot=True)

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed6f10c50>



```
train size: test size = 50%: 50%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.5, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
284
285
In [ ]:
sigmoid SVC classifier.fit(X train, y train)
Out[]:
SVC(C=0.9, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = sigmoid SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test, y pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
Accuracy: 46.6666666666664%
Confusion Matrix:
 [[ 24 77]
 [ 75 109]]
Classification Report:
              precision recall f1-score support
                             0.24
           0
                   0.24
                                        0.24
                                                   101
                             0.59
           1
                   0.59
                                        0.59
                                                   184
                                        0.47
                                                   285
    accuracy
                   0.41
                             0.42
                                        0.41
                                                   285
   macro avq
                   0.46
                             0.47
                                        0.47
                                                   285
weighted avg
In [ ]:
sns.heatmap(cf matrix, annot=True)
Out[]:
<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed6e50050>
                                      - 100
                                      - 90
          24
0
                                      - 80
```

- 70 - 60 - 50

1.1e+02

```
train size : test size = 40% : 60%
```

```
In [ ]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)
```

```
print(len(X_train))
print(len(y_test))
```

227

342

In []:

```
sigmoid_SVC_classifier.fit(X_train, y_train)
```

Out[]:

```
SVC(C=0.9, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

In []:

```
y_pred = sigmoid_SVC_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 50.29239766081871%

Confusion Matrix:

[[36 87]

[83 136]]

Classification Report:

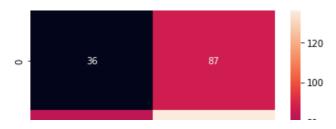
	precision	recall	f1-score	support
0 1	0.30 0.61	0.29	0.30 0.62	123 219
accuracy macro avg weighted avg	0.46 0.50	0.46	0.50 0.46 0.50	342 342 342

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed6d7d590>



```
-80
-60
-40
```

```
train size : test size = 30% : 70%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.7, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
170
399
In [ ]:
sigmoid SVC classifier.fit(X train, y train)
Out[]:
SVC(C=0.9, break ties=False, cache size=200, class weight=None, coef0=0.0,
    decision function shape='ovr', degree=3, gamma='scale', kernel='sigmoid',
    max_iter=-1, probability=False, random state=None, shrinking=True,
    tol=0.001, verbose=False)
In [ ]:
y pred = sigmoid SVC classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n", cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 50.37593984962406%
Confusion Matrix:
 [[ 43 103]
 [ 95 158]]
Classification Report:
              precision recall f1-score support
                   0.31
                            0.29
                                        0.30
                                                   146
                   0.61
           1
                             0.62
                                       0.61
                                                   253
                                        0.50
                                                   399
   accuracy
                   0.46
                             0.46
                                       0.46
                                                   399
   macro avg
                   0.50
                             0.50
                                       0.50
                                                   399
weighted avg
```

sns.heatmap(cf matrix, annot=True)

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed6cb9650>



MLP Classifier

print(cf matrix)

print("\nClassification Report:\n")

```
In [ ]:
mlp classifier = MLPClassifier(learning rate='constant', max iter=600)
mlp classifier
Out[]:
MLPClassifier(activation='relu', alpha=0.0001, batch size='auto', beta 1=0.9,
              beta 2=0.999, early stopping=False, epsilon=1e-08,
              hidden layer sizes=(100,), learning rate='constant',
              learning rate init=0.001, max fun=15000, max iter=600,
              momentum=0.9, n_iter_no change=10, nesterovs momentum=True,
              power t=0.5, random state=None, shuffle=True, solver='adam',
              tol=0.0001, validation fraction=0.1, verbose=False,
              warm start=False)
train size : test size = 70% : 30%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.3, random state=0)
# 70% training data, 30% testing data
In [ ]:
print(len(X train))
print(len(y_test))
171
In [ ]:
mlp_classifier.fit(X_train, y_train)
Out[]:
MLPClassifier(activation='relu', alpha=0.0001, batch size='auto', beta 1=0.9,
              beta 2=0.999, early stopping=False, epsilon=1e-08,
              hidden layer sizes=(100,), learning rate='constant',
              learning rate init=0.001, max fun=15000, max iter=600,
              momentum=0.9, n iter no change=10, nesterovs momentum=True,
              power t=0.5, random state=None, shuffle=True, solver='adam',
              tol=0.0001, validation fraction=0.1, verbose=False,
              warm start=False)
In [ ]:
y pred = mlp classifier.predict(X test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
```

print(classification_report(y_test,y_pred))

Accuracy: 95.32163742690058%

Confusion Matrix:

[[58 5] [3 105]]

Classification Report:

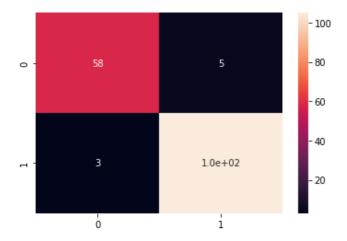
	precision	recall	f1-score	support
0 1	0.95 0.95	0.92 0.97	0.94	63 108
accuracy macro avg weighted avg	0.95 0.95	0.95 0.95	0.95 0.95 0.95	171 171 171

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed6d08790>



train size: test size = 60%: 40%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
# 60% training data, 40% testing data
```

In []:

```
print(len(X_train))
print(len(y_test))
```

341 228

In []:

```
mlp_classifier.fit(X_train, y_train)
```

Out[]:

```
MLPClassifier(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power t=0.5, random state=None, shuffle=True, solver='adam',
```

```
tol=0.0001, validation_fraction=0.1, verbose=False,
warm_start=False)
```

In []:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 96.49122807017544%

Confusion Matrix:

[[76 7] [1 144]]

Classification Report:

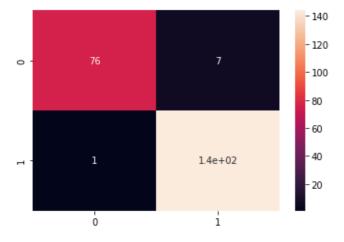
	precision	recall	f1-score	support
0 1	0.99	0.92	0.95 0.97	83 145
accuracy			0.96	228
macro avg	0.97	0.95	0.96	228
weighted avg	0.97	0.96	0.96	228

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed6b6fe90>



train size : test size = 50% : 50%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
# 50% training data, 50% testing data
```

In []:

```
print(len(X_train))
print(len(y_test))
```

284

285

In []:

```
mlp_classifier.fit(X_train, y_train)
Out[]:
```

MLPClassifier (activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9, beta_2=0.999, early_stopping=False, epsilon=1e-08, hidden_layer_sizes=(100,), learning_rate='constant', learning_rate_init=0.001, max_fun=15000, max_iter=600, momentum=0.9, n_iter_no_change=10, nesterovs_momentum=True, power_t=0.5, random_state=None, shuffle=True, solver='adam', tol=0.0001, validation_fraction=0.1, verbose=False, warm_start=False)

In []:

```
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 94.38596491228071%

Confusion Matrix:

[[92 9] [7 177]]

Classification Report:

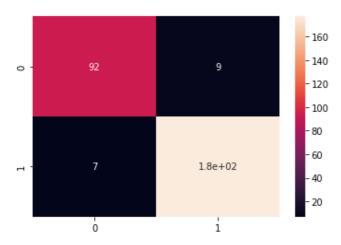
	precision	recall	f1-score	support
0 1	0.93 0.95	0.91 0.96	0.92 0.96	101 184
accuracy macro avg weighted avg	0.94 0.94	0.94	0.94 0.94 0.94	285 285 285

In []:

```
sns.heatmap(cf matrix, annot=True)
```

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed6aaf5d0>



train size : test size = 40% : 60%

In []:

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.6, random_state=0)

```
print(len(X train))
print(len(y test))
227
342
In [ ]:
mlp classifier.fit(X train, y train)
Out[]:
MLPClassifier(activation='relu', alpha=0.0001, batch size='auto', beta 1=0.9,
              beta 2=0.999, early stopping=False, epsilon=1e-08,
              hidden layer sizes=(100,), learning rate='constant',
              learning rate init=0.001, max fun=15000, max iter=600,
              momentum=0.9, n iter no change=10, nesterovs momentum=True,
              power t=0.5, random state=None, shuffle=True, solver='adam',
              tol=0.0001, validation fraction=0.1, verbose=False,
              warm start=False)
In [ ]:
y_pred = mlp_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 91.81286549707602%
Confusion Matrix:
[[ 99 24]
 [ 4 215]]
Classification Report:
                          recall f1-score
              precision
                                              support
           0
                   0.96
                             0.80
                                        0.88
                                                   123
                   0.90
                             0.98
                                        0.94
                                                   219
           1
                                        0.92
                                                   342
    accuracy
                   0.93
                                       0.91
                             0.89
                                                   342
   macro avg
                                        0.92
weighted avg
                   0.92
                             0.92
                                                   342
In [ ]:
sns.heatmap(cf matrix, annot=True)
```

Out[]:

In []:

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed6a37550>



```
train size : test size = 30% : 70%
In [ ]:
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.7, random_state=0)
In [ ]:
print(len(X train))
print(len(y test))
170
399
In [ ]:
mlp classifier.fit(X train, y train)
Out[]:
MLPClassifier(activation='relu', alpha=0.0001, batch size='auto', beta 1=0.9,
              beta 2=0.999, early stopping=False, epsilon=1e-08,
              hidden layer sizes=(100,), learning rate='constant',
              learning rate init=0.001, max fun=15000, max iter=600,
              momentum=0.9, n iter no change=10, nesterovs momentum=True,
              power t=0.5, random state=None, shuffle=True, solver='adam',
              tol=0.0001, validation fraction=0.1, verbose=False,
              warm start=False)
In [ ]:
y pred = mlp classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test, y pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 36.59147869674185%
Confusion Matrix:
[[146
        01
 [253
        0]]
Classification Report:
              precision recall f1-score support
                   0.37
           Λ
                             1.00
                                        0.54
                                                   146
                   0.00
                             0.00
                                        0.00
                                                   253
                                        0.37
                                                   399
   accuracy
                   0.18
                             0.50
                                        0.27
                                                   399
   macro avg
                   0.13
                             0.37
                                        0.20
                                                   399
weighted avg
/usr/local/lib/python3.7/dist-packages/sklearn/metrics/ classification.py:1272: Undefined
MetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with
```

no predicted samples. Use `zero_division` parameter to control this behavior.

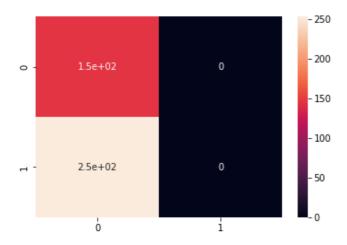
_warn_prf(average, modifier, msg_start, len(result))

sns.heatmap(cf matrix, annot=True)

In []:

Out[]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed693b090>



Random Forest Classifier

In []:

y pred = rfc classifier.predict(X test)

```
In [ ]:
rfc classifier = RandomForestClassifier(n estimators=20)
rfc classifier
Out[]:
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None,
                       criterion='gini', max depth=None, max features='auto',
                       max leaf nodes=None, max samples=None,
                       min impurity decrease=0.0, min impurity split=None,
                       min samples leaf=1, min samples split=2,
                       min weight fraction leaf=0.0, n estimators=20,
                       n jobs=None, oob score=False, random state=None,
                       verbose=0, warm start=False)
train size: test size = 70%: 30%
In [ ]:
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
# 70% training data, 30% testing data
In [ ]:
print(len(X train))
print(len(y test))
398
171
In [ ]:
rfc_classifier.fit(X_train, y_train)
Out[]:
RandomForestClassifier(bootstrap=True, ccp alpha=0.0, class weight=None,
                       criterion='gini', max depth=None, max features='auto',
                       max leaf nodes=None, max samples=None,
                       min impurity decrease=0.0, min impurity split=None,
                       min_samples_leaf=1, min_samples_split=2,
                       min weight fraction leaf=0.0, n estimators=20,
                       n jobs=None, oob score=False, random state=None,
                       verbose=0, warm_start=False)
```

```
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 95.90643274853801%

Confusion Matrix:

[[61 2] [5 103]]

Classification Report:

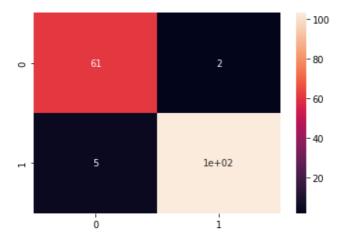
	precision	recall	f1-score	support
0 1	0.92 0.98	0.97 0.95	0.95 0.97	63 108
accuracy macro avg weighted avg	0.95 0.96	0.96	0.96 0.96	171 171 171

In []:

```
sns.heatmap(cf matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed68df550>



train size : test size = 60% : 40%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=0)
# 60% training data, 40% testing data
```

In []:

```
print(len(X_train))
print(len(y_test))
```

341 228

In []:

```
rfc_classifier.fit(X_train, y_train)
```

Out[]:

```
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, max_samples=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=None, oob_score=False, random_state=None, verbose=0, warm_start=False)
```

In []:

```
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf_matrix)
print(cf_matrix)
print("\nClassification_Report:\n")
print(classification_report(y_test,y_pred))
```

Accuracy: 96.05263157894737%

Confusion Matrix:

[[77 6] [3 142]]

Classification Report:

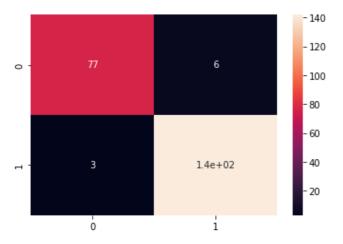
	precision	recall	f1-score	support
0 1	0.96 0.96	0.93	0.94 0.97	83 145
accuracy macro avg weighted avg	0.96 0.96	0.95 0.96	0.96 0.96 0.96	228 228 228

In []:

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed73f1cd0>



train size : test size = 50% : 50%

In []:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.5, random_state=0)
# 50% training data, 50% testing data
```

In []:

```
print(len(X train))
```

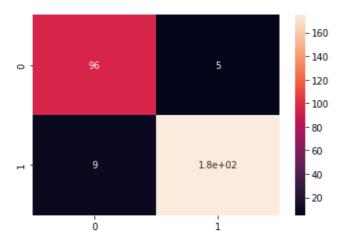
```
284
285
In [ ]:
rfc classifier.fit(X train, y train)
Out[]:
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight=None,
                       criterion='gini', max depth=None, max features='auto',
                       max leaf nodes=None, max samples=None,
                       min impurity decrease=0.0, min impurity split=None,
                       min samples leaf=1, min samples split=2,
                       min weight fraction leaf=0.0, n estimators=20,
                       n_jobs=None, oob_score=False, random_state=None,
                       verbose=0, warm_start=False)
In [ ]:
y pred = rfc classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion matrix(y test, y pred)
print("Confusion Matrix:")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
Accuracy: 95.08771929824562%
Confusion Matrix:
[[ 96 5]
 [ 9 175]]
Classification Report:
              precision recall f1-score support
                   0.91
                            0.95
                                       0.93
                                                  101
                   0.97
                             0.95
                                       0.96
                                                  184
                                       0.95
   accuracy
                                                  285
                                       0.95
                   0.94
                             0.95
                                                  285
   macro avg
                   0.95
                             0.95
                                       0.95
                                                  285
weighted avg
In [ ]:
```

```
sns.heatmap(cf_matrix, annot=True)
```

Out[]:

print(len(y test))

<matplotlib.axes. subplots.AxesSubplot at 0x7f4ed6781390>



```
train size : test size = 40% : 60%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.6, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
342
In [ ]:
rfc classifier.fit(X train, y train)
RandomForestClassifier(bootstrap=True, ccp alpha=0.0, class weight=None,
                       criterion='gini', max depth=None, max features='auto',
                       max leaf nodes=None, max samples=None,
                       min impurity decrease=0.0, min impurity split=None,
                       min samples leaf=1, min samples split=2,
                       min weight fraction leaf=0.0, n estimators=20,
                       n jobs=None, oob score=False, random state=None,
                       verbose=0, warm start=False)
In [ ]:
y_pred = rfc_classifier.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 94.15204678362574%
Confusion Matrix:
      91
[[114
 [ 11 208]]
Classification Report:
              precision
                          recall f1-score support
                   0.91
                             0.93
                                        0.92
                                                   123
           1
                   0.96
                             0.95
                                        0.95
                                                   219
                                        0.94
                                                   342
    accuracy
                             0.94
                  0.94
                                        0.94
                                                   342
  macro avg
weighted avg
                   0.94
                             0.94
                                        0.94
                                                   342
In [ ]:
sns.heatmap(cf matrix, annot=True)
Out[]:
<matplotlib.axes._subplots.AxesSubplot at 0x7f4ed66b9990>
                                      - 200
                                      - 175
        1.1e+02
0
                                      - 150
                                      - 125
```

```
- 100
- 75
- 50
- 25
```

```
train size : test size = 30% : 70%
In [ ]:
X train, X test, y train, y test = train test split(X, y, test size=0.7, random state=0)
In [ ]:
print(len(X train))
print(len(y test))
170
399
In [ ]:
rfc_classifier.fit(X_train, y_train)
Out[]:
RandomForestClassifier(bootstrap=True, ccp alpha=0.0, class weight=None,
                       criterion='gini', max depth=None, max features='auto',
                       max leaf nodes=None, max samples=None,
                       min impurity decrease=0.0, min impurity split=None,
                       min_samples_leaf=1, min_samples_split=2,
                       min weight fraction leaf=0.0, n estimators=20,
                       n jobs=None, oob score=False, random state=None,
                       verbose=0, warm start=False)
In [ ]:
y pred = rfc classifier.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test, y pred)}%\n")
cf matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:\n")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test, y pred))
Accuracy: 94.23558897243107%
Confusion Matrix:
[[130 16]
 7 246]]
Classification Report:
              precision recall f1-score support
                   0.95
                             0.89
                                        0.92
           0
                                                   146
                                        0.96
           1
                   0.94
                             0.97
                                                   253
                                        0.94
                                                   399
   accuracy
                   0.94
                             0.93
                                       0.94
                                                   399
   macro avq
```

In []:

weighted avg

sns.heatmap(cf_matrix, annot=True)

0.94

0.94

0.94

399

Out[]: <matplotlib.axes._subplots.AxesSubplot at 0x7f4ed665e0d0>

