

MACHINE LEARNING LAB

ANSWER SCRIPT

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Subject: Machine Learning Lab (IT/PC/B/S/411)

Year: 4

Semester: 1

Department : Information Technology

GitHub Link: <https://github.com/sudiptajuit/ML-lab/tree/main/Final%20Evaluation%20of%20ML%20Lab>

Question1:

Apply different types of following machine learning models:
(CO1) (10)

- Support Vector Machine (SVM)
- Decision Tree
- Random Forest
- Naive Bayes

And compare as well as discuss the performances in terms of Accuracy, Precision, Recall and F-measure on the following datasets:

1. Wine Dataset:

<https://archive.ics.uci.edu/ml/datasets/wine>

2. Ionosphere Dataset:

<https://archive.ics.uci.edu/ml/datasets/Ionosphere>

Generate the respective confusion matrices with class labels (heat map).

IMPORT THE REQUIRED HEADERS

```
In [60]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import scikitplot as skplt

from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier from
sklearn.ensemble import RandomForestClassifier from
sklearn.naive_bayes import GaussianNB, BernoulliNB
```

Setting up the dataset

```
In [61]: def preprocess(X,y,te_size,label=False,scale=False,pca=False):
    if
label:
    from sklearn.preprocessing import LabelEncoder
    y = LabelEncoder().fit_transform(y)

    from sklearn.model_selection import train_test_split
    X_tr,X_te,y_tr,y_te = train_test_split(X,y,test_size=te_size)
    if
scale:
    from sklearn.preprocessing import StandardScaler
    sc = StandardScaler()
    X_tr = sc.fit_transform(X_tr)
    X_te = sc.transform(X_te)
    if
pca:
    from sklearn.decomposition import PCA
    pca = PCA(n_components='mle')
    X_tr = pca.fit_transform(X_tr)
    X_te = pca.transform(X_te)

    return X_tr,X_te,y_tr,y_te
```

Tester function to display

In [62]:

```
def tester(classi,X_t,y_t,y_p):

    from sklearn.metrics import
    classification_report,confusion_matrix,accuracy_score,    print("Confusion
    Matrix")    print(confusion_matrix(y_t,y_p))
    print('-----')
    print('-----\n\n')
    print('Preformance Evaluation:')
    print(classification_report(y_t,y_p))    print('-----
    -----')    print('-----
    -----\n\n')    print('Accuracy
    Score:')    print(accuracy_score(y_t,y_p))
    plot_confusion_matrix(classi,X_t,y_t)
    plt.title('Heat map for confusion matrix')
    plt.show()
    y_p_proba = classifier.predict_proba(X_t)
    skplt.metrics.plot_roc(y_t,y_p_proba)
    plt.show()
```

IMPORTING THE WINE DATASET

```
In [63]: df1 = pd.read_csv('wine.data', header=None)
        X = df1.iloc[:,1:] y = df1.iloc[:,0]
```

TRAIN-TEST SPLIT OF 50:50

In [64]:

```
X_train,X_test,y_train,y_test = preprocess(X,y,0.5,scale=True,pca=True)
```

SVM CLASSIFIERS with Linear, Polynomial(degree2), Polynomial(degree3), Gaussian, Sigmoid

In [65]:

```
#SVM Linear model
classifier = SVC(kernel='linear',
probability=True) classifier.fit(X_train,y_train)
y_pred = classifier.predict(X_test) print('SVC
Linear:')
tester(classifier,X_test,y_test,y_pred) print('-----
-----\n\n\n')

#SVM polynomial model degree 2
classifier = SVC(kernel='poly', degree=2,
probability=True) classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test) print('SVC Polynomial degree
2:') tester(classifier,X_test,y_test,y_pred) print('-----
-----\n\n\n')

#SVM polynomial model degree 3
classifier = SVC(kernel='poly', degree=3,
probability=True) classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test) print('SVC Polynomial degree
3:') tester(classifier,X_test,y_test,y_pred) print('-----
-----\n\n\n')

#SVM gaussian model
classifier = SVC(kernel='rbf',
probability=True)
classifier.fit(X_train,y_train) y_pred =
classifier.predict(X_test) print('SVC
Gaussian:')
tester(classifier,X_test,y_test,y_pred) print('-----
-----\n\n\n')

#SVM sigmoid model
classifier = SVC(kernel='sigmoid',
probability=True) classifier.fit(X_train,y_train)
y_pred = classifier.predict(X_test) print('SVC
Sigmoid:')
tester(classifier,X_test,y_test,y_pred) print('-----
-----\n\n\n')
```

SVC Linear:

Confusion Matrix

```
[[28  0  0]
 [ 4 33  0]
 [ 0  1 23]]
```


Preformance Evaluation:

precision

recall f1-score support

1	0.88	1.00	0.93	28
2	0.97	0.89	0.93	37
3	1.00	0.96	0.98	24

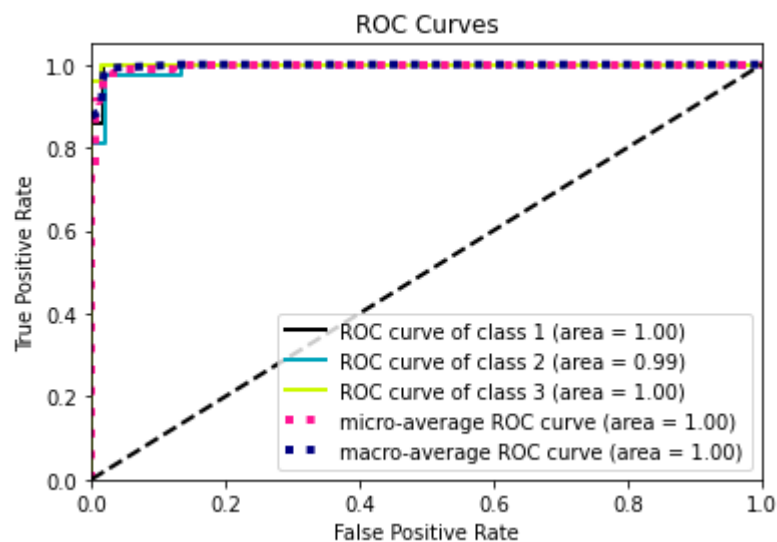
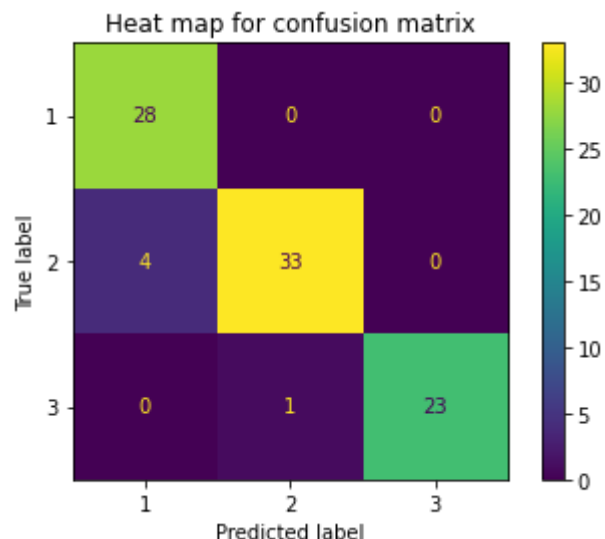
	accuracy		0.94
89	macro avg	0.95	0.95
89	weighted avg	0.95	0.94
89			

Accuracy Score:

0.9438202247191011

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator. warnings.warn(msg,
category=FutureWarning)



SVC Polynomial degree 2:

Confusion Matrix

```
[[24  3  1]
 [ 6 29  2]
 [ 1  0 23]]
```

```

Preformance Evaluation:
recall  f1-score  support      precision

1      0.77      0.86      0.81      28
2      0.91      0.78      0.84      37
3      0.88      0.96      0.92      24

accuracy 0.85
89  macro avg      0.86      0.87      0.86
89  weighted avg    0.86      0.85      0.85
89

```

```

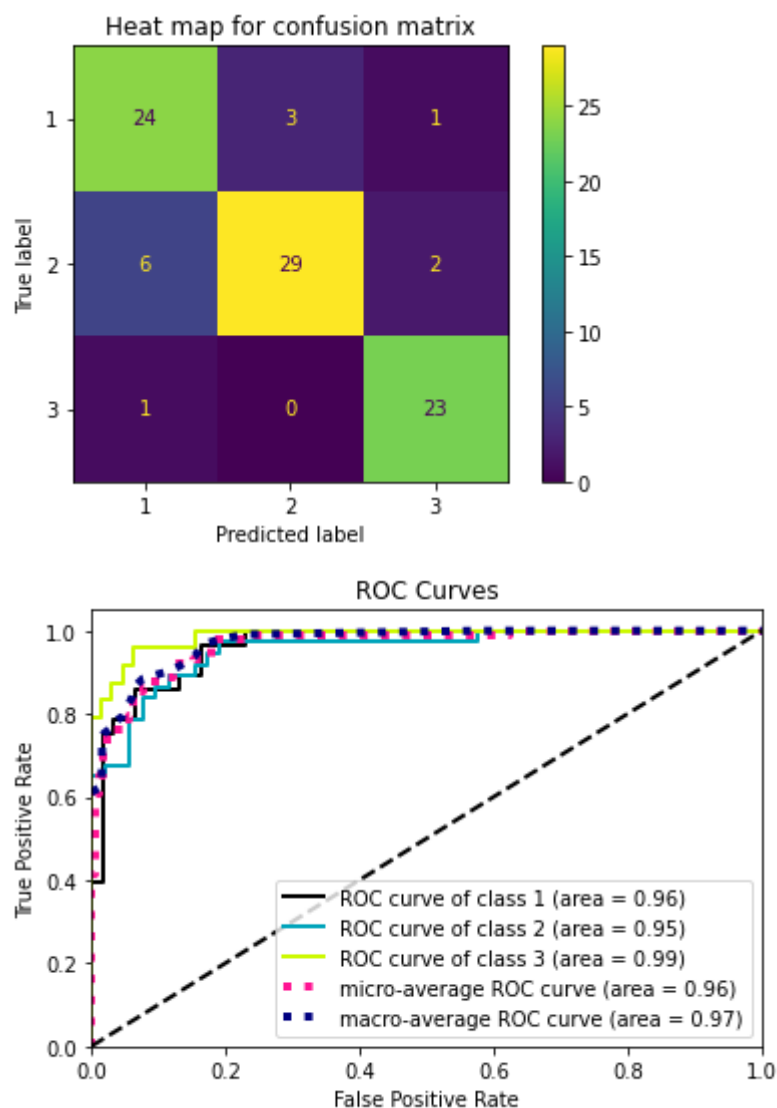
Accuracy Score:
0.8539325842696629

```

```

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:
FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator.  warnings.warn(msg,
category=FutureWarning)

```



SVC Polynomial degree 3:

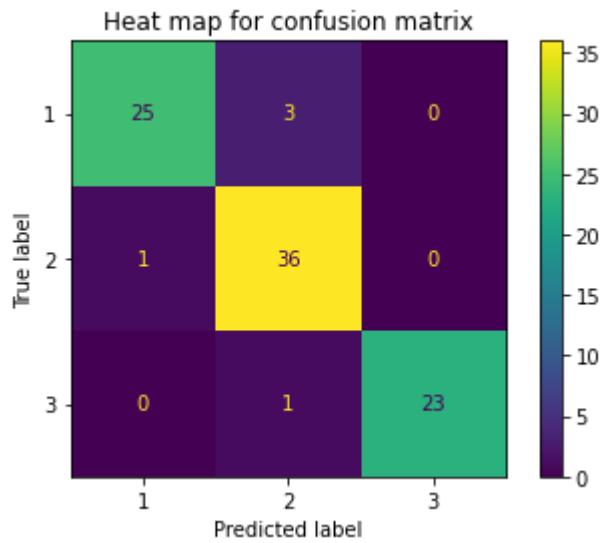
Confusion Matrix

```
[[25  3  0]
 [ 1 36  0]
 [ 0  1 23]]
```


Preformance Evaluation:				precision
recall	f1-score	support		
1	0.96	0.89	0.93	28
2	0.90	0.97	0.94	37
3	1.00	0.96	0.98	24
accuracy				0.94
89	macro avg	0.95	0.94	0.95
89	weighted avg	0.95	0.94	0.94
89				

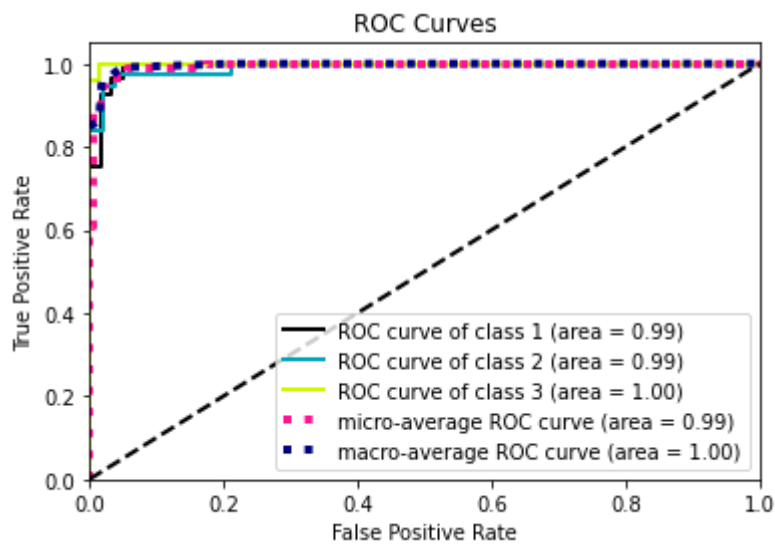
Accuracy Score:
0.9438202247191011

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:



FutureWarning: Function

plot_confusion_matrix is



deprecated; Function

`plot_confusion_matrix` is

deprecated in 1.0 and will be removed in 1.2. Use one of the class methods:

ConfusionMatrixDisplay.from_predictions or

```
ConfusionMatrixDisplay.from_estimator.    warnings.warn(msg,
```

```
category=FutureWarning) -----
```

SVC Gaussian:

Confusion Matrix

```
[[28  0  0]
 [ 0 37  0]
 [ 0  1 23]]
```

```
-----
-----

Preformance Evaluation:                precision
recall  f1-score  support

1         1.00      1.00      1.00         28
2         0.97      1.00      0.99         37
          3         1.00      0.96      0.98         24

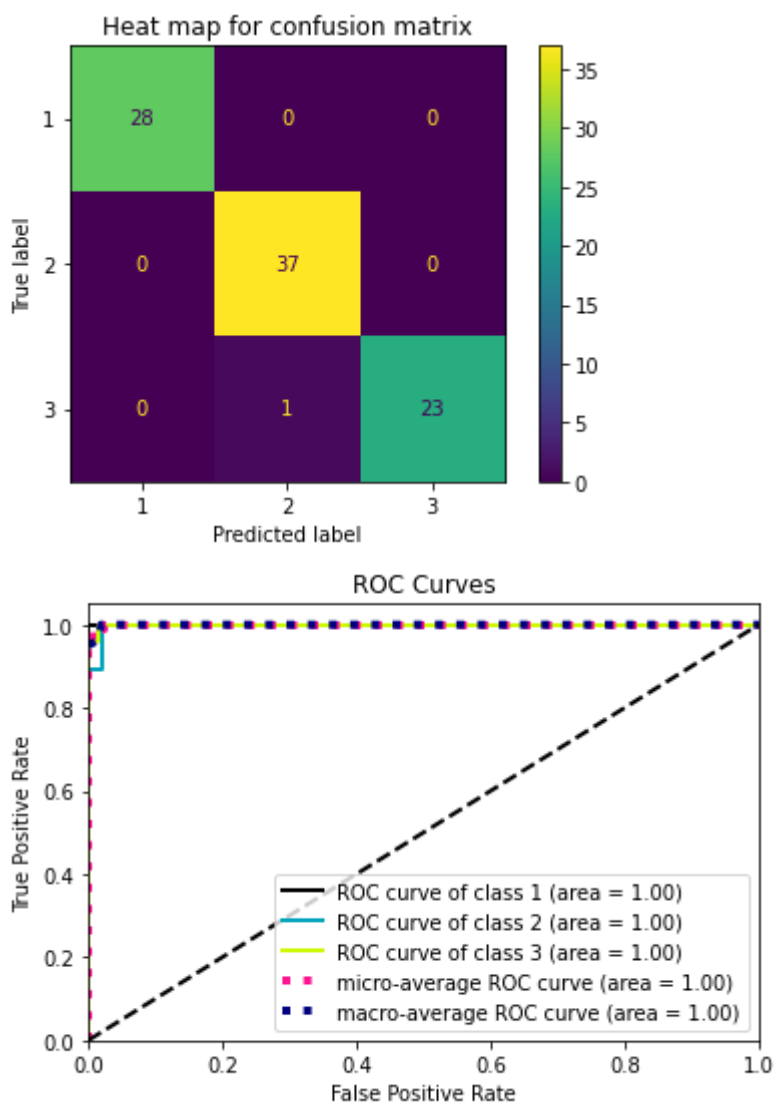
      accuracy                0.99
89  macro avg          0.99      0.99      0.99
89  weighted avg          0.99      0.99      0.99
89
```

Accuracy Score:

0.9887640449438202

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator. warnings.warn(msg,
category=FutureWarning)



SVC Sigmoid:
 Confusion Matrix

```
[[28  0  0]
 [ 1 35  1]
 [ 0  0 24]]
```

Preformance Evaluation:

	recall	f1-score	support	precision
1	0.97	1.00	0.98	28
2	1.00	0.95	0.97	37
3	0.96	1.00	0.98	24
accuracy	0.98			
89 macro avg	0.98	0.98	0.98	
89 weighted avg	0.98	0.98	0.98	
89				

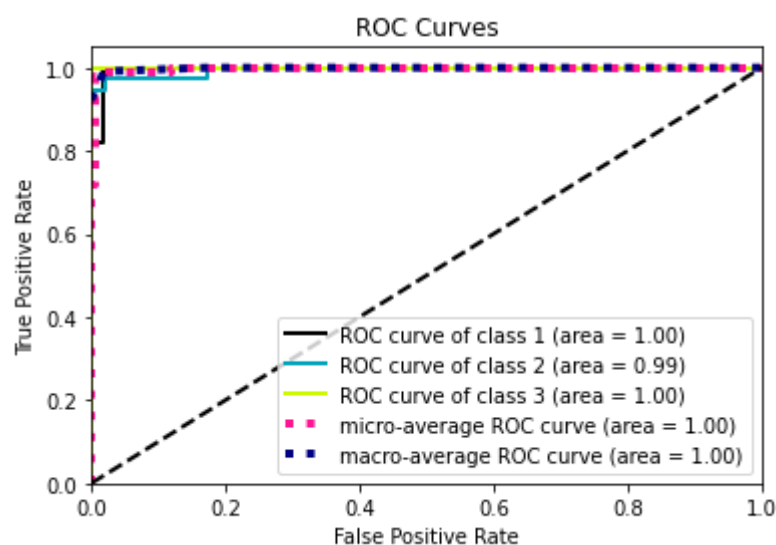
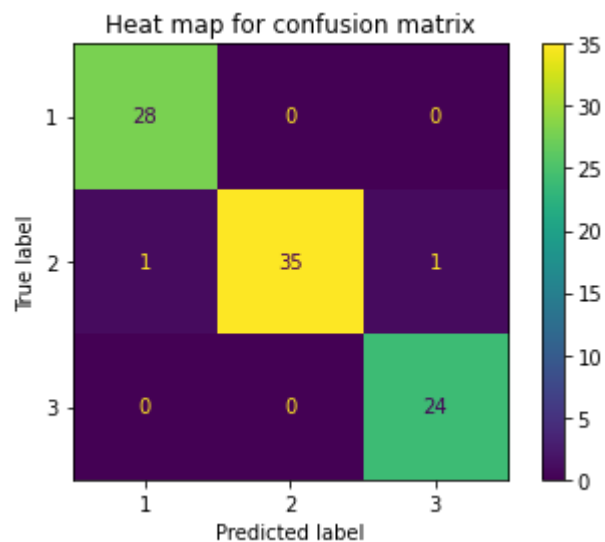
Accuracy Score:

0.9775280898876404

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator.

warnings.warn(msg, category=FutureWarning)



DECISION TREE CLASSIFIER

```
In [66]: #Decision tree classifier
classifier=DecisionTreeClassifier()
classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test)
print('Decision Tree Classifier:')
tester(classifier,X_test,y_test,y_pred)
print('-----\n\n\n')
```

Decision Tree Classifier:

Confusion Matrix

[[28 0 0]

```
[ 5 32  0]
[ 0  0 24]]
```

```
-----
-----

Preformance Evaluation:                precision
recall  f1-score  support

1         0.85      1.00      0.92      28
2         1.00      0.86      0.93      37
3         1.00      1.00      1.00      24
    accuracy                                0.94
89    macro avg              0.95      0.95      0.95
89 weighted avg              0.95      0.94      0.94
89
```

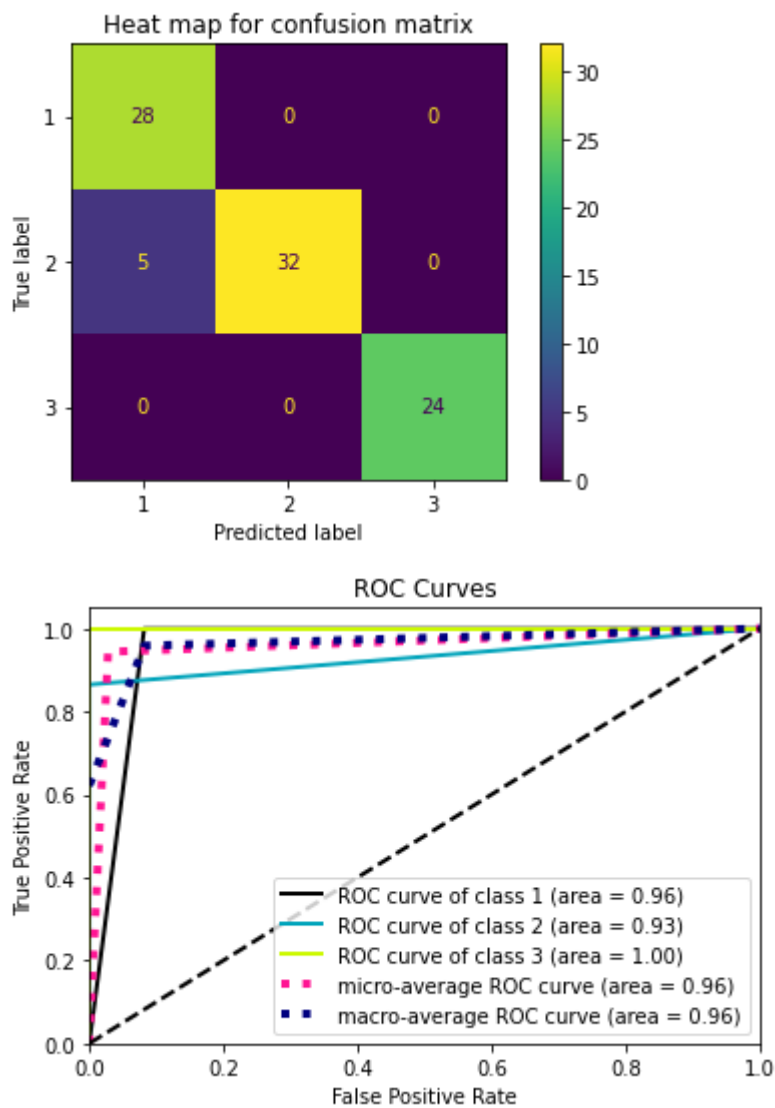
```
-----
-----

Accuracy Score:
```

```
0.9438202247191011
```

```
/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:
```

```
FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator.  warnings.warn(msg,
category=FutureWarning)
```



RANDOM FOREST MODEL

```
In [67]: #Random forest model
classifier=RandomForestClassifier()
classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test)
print('Random Forest Classifier:')
tester(classifier,X_test,y_test,y_pred)
print('-----\n\n\n')
```

Random Forest Classifier:
Confusion Matrix
[[28 0 0]
[3 34 0]
[0 0 24]]

```

Preformance Evaluation:
recall  f1-score  support      precision

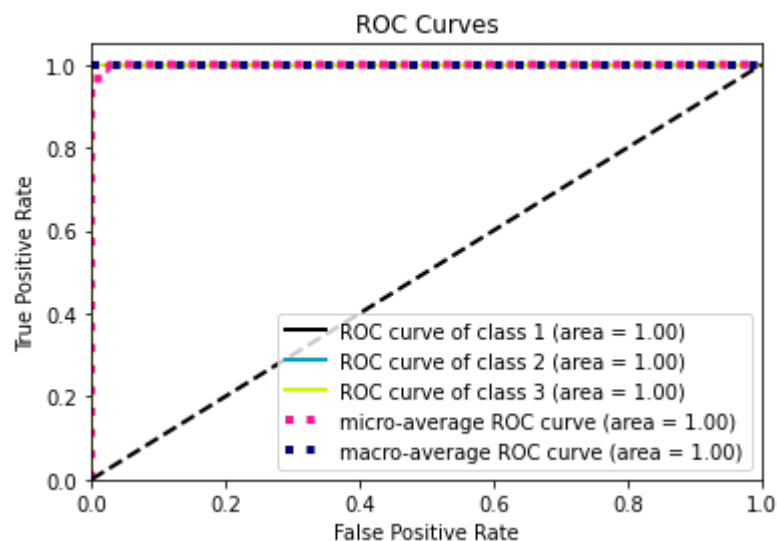
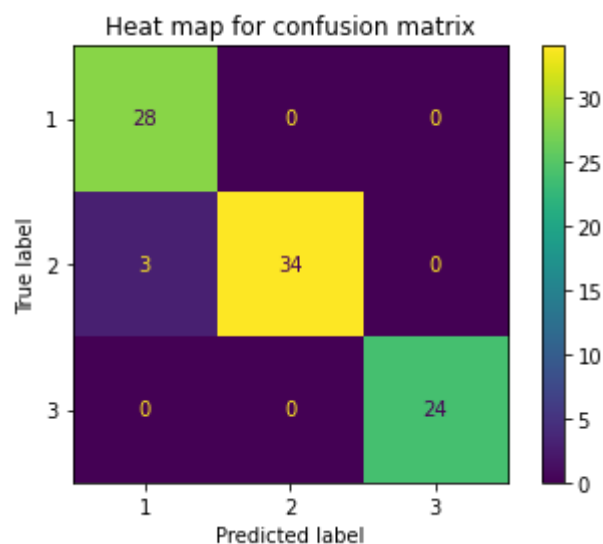
1      0.90      1.00      0.95      28
2      1.00      0.92      0.96      37
3      1.00      1.00      1.00      24

accuracy 0.97
89 macro avg      0.97      0.97      0.97
89 weighted avg    0.97      0.97      0.97
89

```

Accuracy Score:
0.9662921348314607

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:
FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator.
warnings.warn(msg, category=FutureWarning)



NAIVE BAYES CLASSIFIER

In [68]:

```
#Gaussian naive bayes
classifier=GaussianNB()
classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test)
print('Gaussian Naive Bayes:')
tester(classifier,X_test,y_test,y_pred)
print('-----\n\n\n')
#Bernoulli naive bayes
classifier=BernoulliNB()
classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test)
print('Bernoulli Naive Bayes:')
tester(classifier,X_test,y_test,y_pred)
print('-----\n\n\n')
```

Gaussian Naive Bayes:

Confusion Matrix

```
[[28  0  0]
 [ 0 37  0]
 [ 0  0 24]]
```


Preformance Evaluation:

precision

recall f1-score support

1	1.00	1.00	1.00	28
2	1.00	1.00	1.00	37
3	1.00	1.00	1.00	24

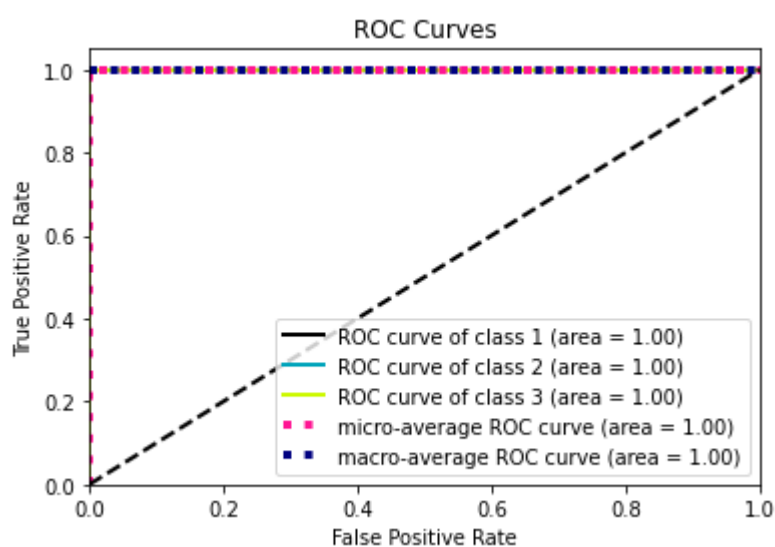
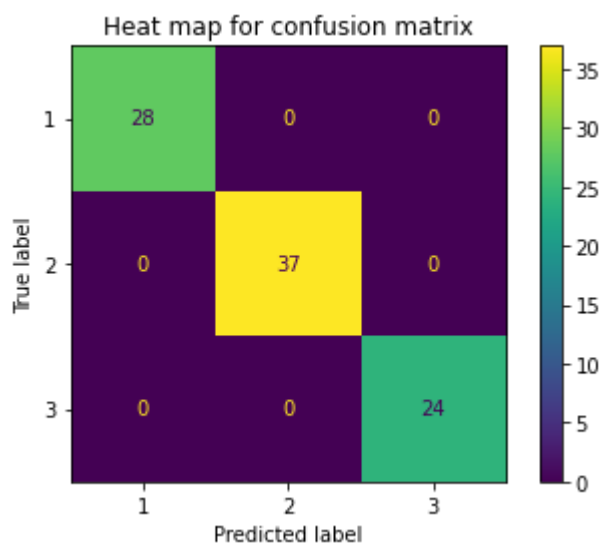
	accuracy			1.00
89	macro avg	1.00	1.00	1.00
89	weighted avg	1.00	1.00	1.00
89				

Accuracy Score:

1.0

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator. warnings.warn(msg,
category=FutureWarning)



Bernoulli Naive Bayes:

Confusion Matrix

```
[[26  2  0]
 [ 1 36  0]
 [ 0  1 23]]
```

Preformance Evaluation:

recall f1-score support

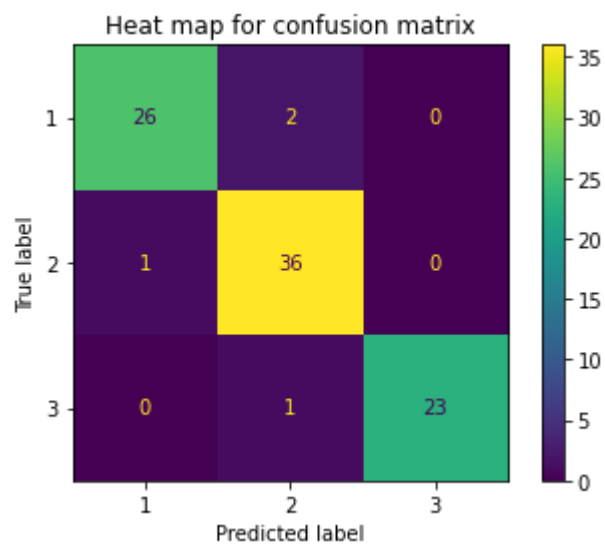
	recall	f1-score	support	precision
1	0.96	0.93	0.95	28
2	0.92	0.97	0.95	37
3	1.00	0.96	0.98	24

	accuracy		0.96
89 macro avg	0.96	0.95	0.96
89 weighted avg	0.96	0.96	0.96
89			

Accuracy Score:

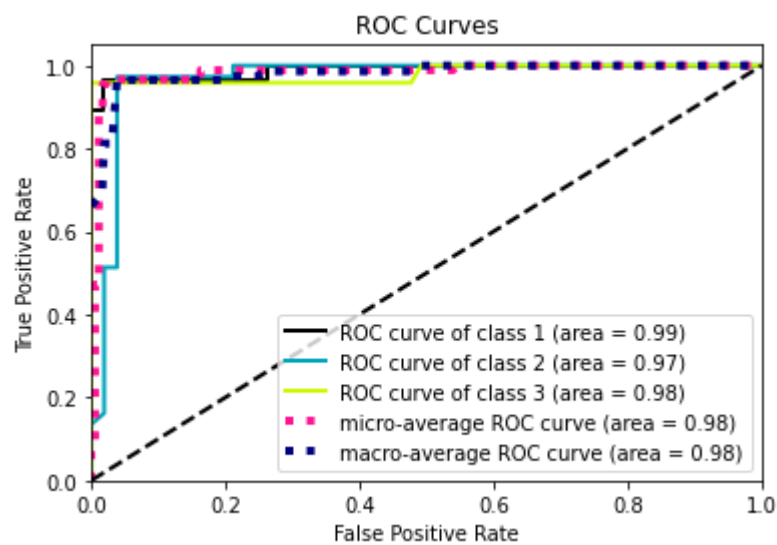
0.9550561797752809

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:



FutureWarning: Function

plot_confusion_matrix is



deprecated; Function

`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one

of the class methods: `ConfusionMatrixDisplay.from_predictions` or

ConfusionMatrixDisplay.from_estimator.

warnings.warn(msg,

category=FutureWarning) -----

WINE DATASET

	Accuracy	Precision	Recall	F-Measure
SVM CLASSIFIER				
Linear	0.943	0.95	0.95	0.95
Polynomial (2)	0.853	0.86	0.87	0.86
Polynomial (3)	0.943	0.95	0.94	0.95
Gaussian	0.988	0.99	0.99	0.99
Sigmoid	0.977	0.98	0.98	0.98
DECISION TREE CLASSIFIER				
Decision Tree	0.943	0.95	0.94	0.95
RANDOM FOREST CLASSIFIER				
Random Forest	0.966	0.97	0.97	0.97
NAÏVE BAYES CLASSIFIER				
Gaussian	1	1	1	1
Bernoulli	0.955	0.96	0.95	0.96

From the above tabulated data, we can conclude that at **50:50** train test split ratio in the **WINE** dataset, **Gaussian Naïve Bayes Classifier** produces the maximum accuracy followed by the **SVM Gaussian classifier**.

IMPORT THE REQUIRED HEADERS

```
In [41]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import scikitplot as skplt

from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier from
sklearn.ensemble import RandomForestClassifier from
sklearn.naive_bayes import GaussianNB, BernoulliNB
```

Setting up the dataset

```
In [26]: def preprocess(X,y,te_size,label=False,scale=False,pca=False):
    if
label:
    from sklearn.preprocessing import LabelEncoder
    y = LabelEncoder().fit_transform(y)

    from sklearn.model_selection import train_test_split
    X_tr,X_te,y_tr,y_te = train_test_split(X,y,test_size=te_size)
    if
scale:
    from sklearn.preprocessing import StandardScaler
    sc = StandardScaler()
    X_tr = sc.fit_transform(X_tr)
    X_te = sc.transform(X_te)
    if
pca:
    from sklearn.decomposition import PCA
    pca = PCA(n_components='mle')
    X_tr = pca.fit_transform(X_tr)
    X_te = pca.transform(X_te)

    return X_tr,X_te,y_tr,y_te
```

Tester function to display

In [42]:

```
def tester(classi,X_t,y_t,y_p):

    from sklearn.metrics import
    classification_report,confusion_matrix,accuracy_score,    print("Confusion
    Matrix")    print(confusion_matrix(y_t,y_p))
    print('-----')
    print('-----\n\n')
    print('Preformance Evaluation:')
    print(classification_report(y_t,y_p))    print('-----
    -----')    print('-----
    -----\n\n')    print('Accuracy
    Score:')    print(accuracy_score(y_t,y_p))
    plot_confusion_matrix(classi,X_t,y_t)
    plt.title('Heat map for confusion matrix')
    plt.show()
    y_p_proba = classifier.predict_proba(X_t)
    skplt.metrics.plot_roc(y_t,y_p_proba)
    plt.show()
```

IMPORTING THE IONOSPHERE DATASET

```
In [28]: df1 = pd.read_csv('ionosphere.data', header=None)
        X = df1.iloc[:, :-1] y = df1.iloc[:, -1]
```

TRAIN-TEST SPLIT OF 50:50

In [35]:

```
X_train,X_test,y_train,y_test = preprocess(X,y,0.5,scale=True,pca=True)
```

SVM CLASSIFIERS with Linear, Polynomial(degree2), Polynomial(degree3), Gaussian, Sigmoid

In [43]:

```
#SVM Linear model
classifier = SVC(kernel='linear',
probability=True) classifier.fit(X_train,y_train)
y_pred = classifier.predict(X_test) print('SVC
Linear:')
tester(classifier,X_test,y_test,y_pred) print('-----
-----\n\n\n')

#SVM polynomial model degree 2
classifier = SVC(kernel='poly', degree=2,
probability=True) classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test) print('SVC Polynomial degree
2:') tester(classifier,X_test,y_test,y_pred) print('-----
-----\n\n\n')

#SVM polynomial model degree 3
classifier = SVC(kernel='poly', degree=3,
probability=True) classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test) print('SVC Polynomial degree
3:') tester(classifier,X_test,y_test,y_pred) print('-----
-----\n\n\n')

#SVM gaussian model
classifier = SVC(kernel='rbf',
probability=True)
classifier.fit(X_train,y_train) y_pred =
classifier.predict(X_test) print('SVC
Gaussian:')
tester(classifier,X_test,y_test,y_pred) print('-----
-----\n\n\n')

#SVM sigmoid model
classifier = SVC(kernel='sigmoid',
probability=True) classifier.fit(X_train,y_train)
y_pred = classifier.predict(X_test) print('SVC
Sigmoid:')
tester(classifier,X_test,y_test,y_pred) print('-----
-----\n\n\n')
```

SVC Linear:

Confusion Matrix

```
[[ 52  19]
 [  2 103]]
```


Preformance Evaluation:

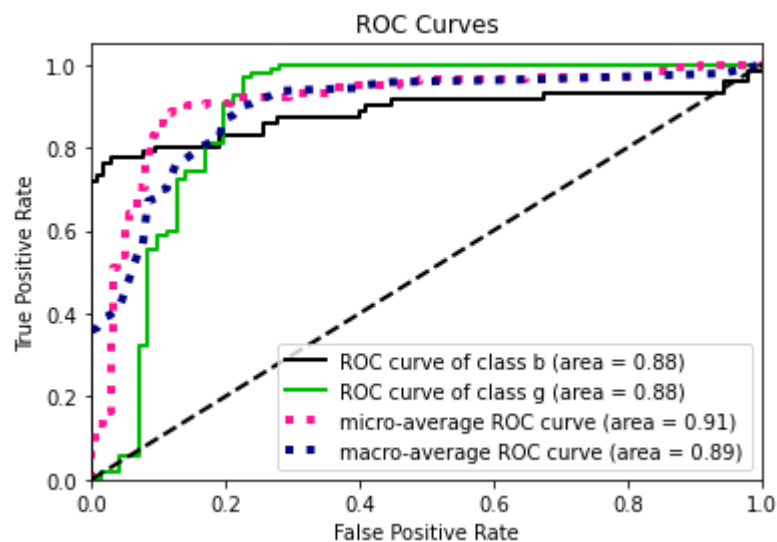
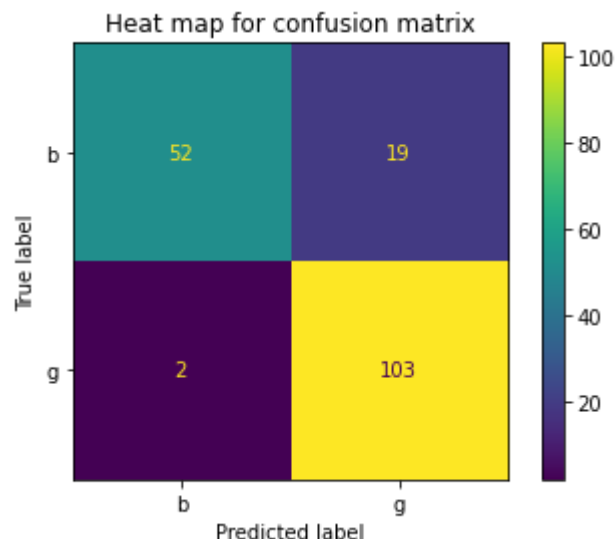
		precision	recall	f1-score
support				
	b	0.96	0.73	0.83
71	g	0.84	0.98	0.91
105				
	accuracy			0.88
176	macro avg	0.90	0.86	0.87
176	weighted avg	0.89	0.88	0.88
176				

Accuracy Score:

0.8806818181818182

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator. warnings.warn(msg,
category=FutureWarning)



SVC Polynomial degree 2:

Confusion Matrix

```
[[ 33  38]
 [  0 105]]
```


Performance Evaluation:

recall f1-score support

precision

	b	1.00	0.46	0.63
71	g	0.73	1.00	0.85
105				

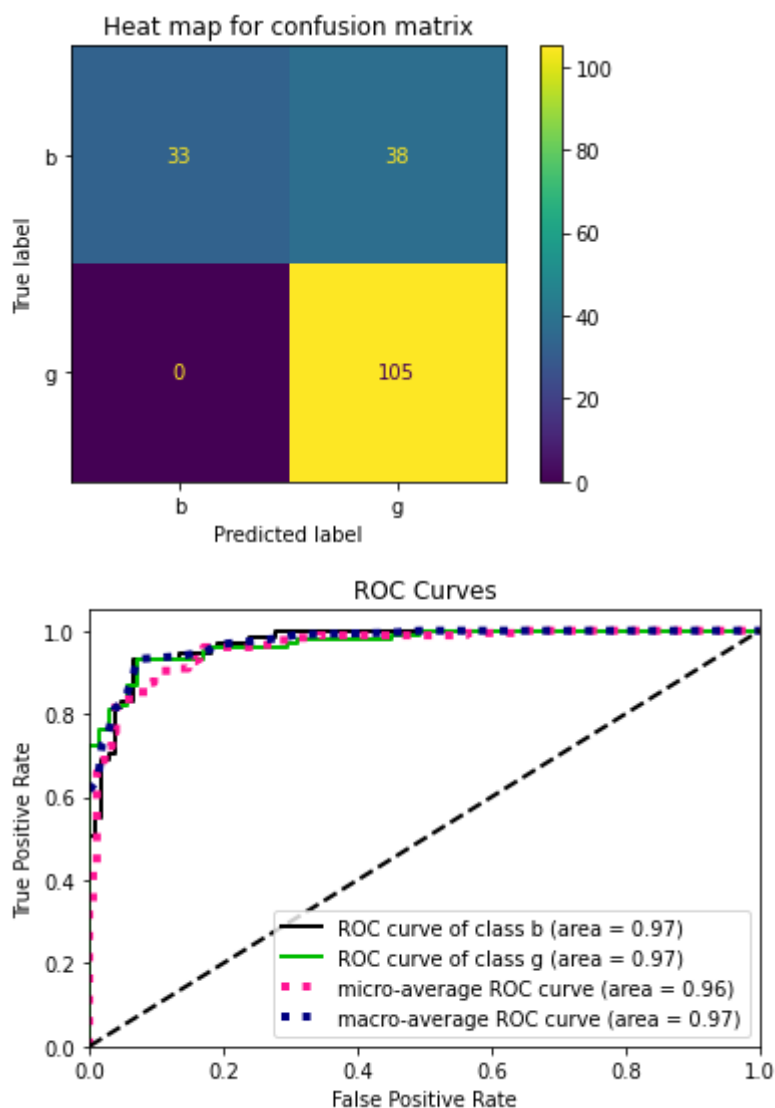
	accuracy		0.78
176	macro avg	0.87	0.73
176	weighted avg	0.84	0.78
176			

Accuracy Score:

0.7840909090909091

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function plot_confusion_matrix is deprecated; Function `plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: ConfusionMatrixDisplay.from_predictions or ConfusionMatrixDisplay.from_estimator. warnings.warn(msg, category=FutureWarning)



SVC Polynomial degree 3:

Confusion Matrix

```
[[ 13  58]
 [   0 105]]
```


Preformance Evaluation: precision

recall f1-score support

	b	1.00	0.18	0.31
71	g	0.64	1.00	0.78
105				

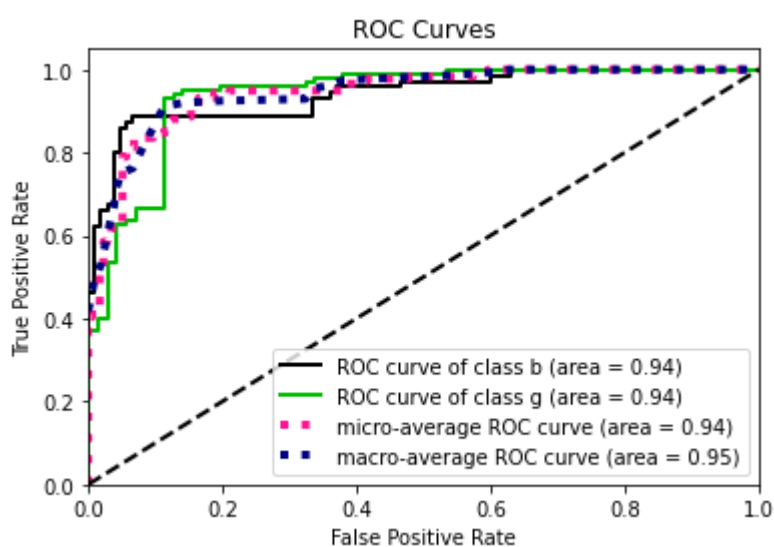
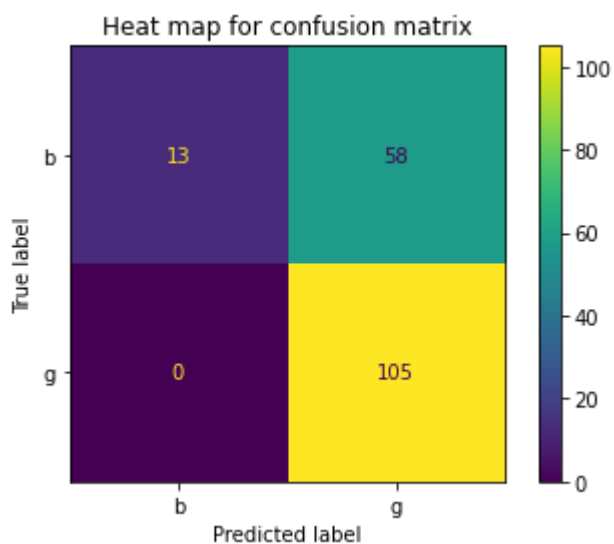
	accuracy		0.67	
176	macro avg	0.82	0.59	0.55
176	weighted avg	0.79	0.67	0.59
176				

Accuracy Score:

0.6704545454545454

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator. warnings.warn(msg,
category=FutureWarning)



SVC Gaussian:
Confusion Matrix
[[58 13]
[2 103]]

Preformance Evaluation:

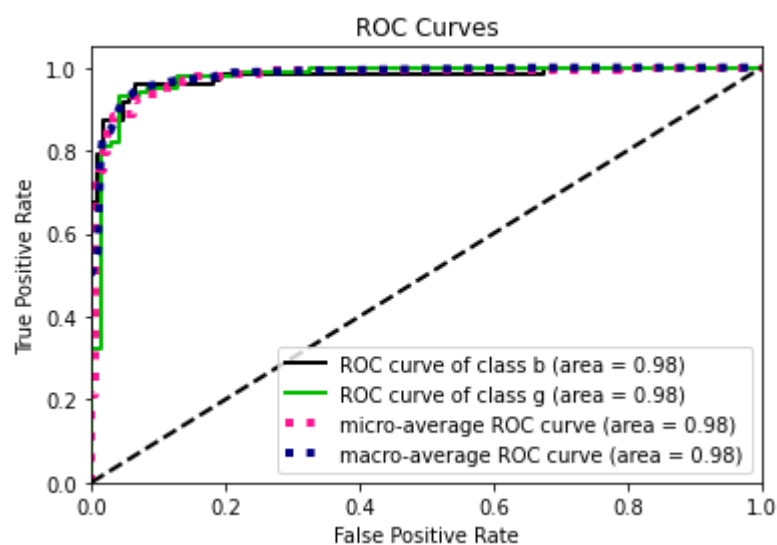
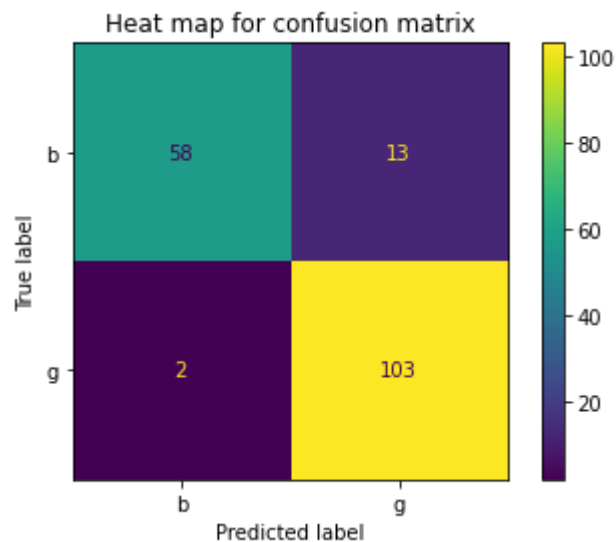
		recall	f1-score	support	precision
71	b	0.97	0.82	0.89	
	g	0.89	0.98	0.93	
105					
	accuracy				0.91
176	macro avg	0.93	0.90	0.91	
176	weighted avg	0.92	0.91	0.91	
176					

Accuracy Score:

0.9147727272727273

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator. warnings.warn(msg,
category=FutureWarning)



SVC Sigmoid:

Confusion Matrix

```
[[ 47  24]
 [   1 104]]
```

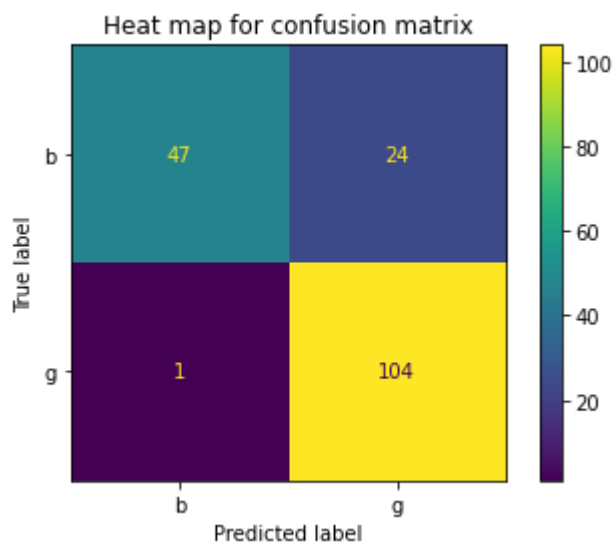
Preformance Evaluation: precision
recall f1-score support

	b	0.98	0.66	0.79
71	g	0.81	0.99	0.89
105				

	accuracy		0.86
176	macro avg	0.90	0.83
176	weighted avg	0.88	0.86
176			0.85

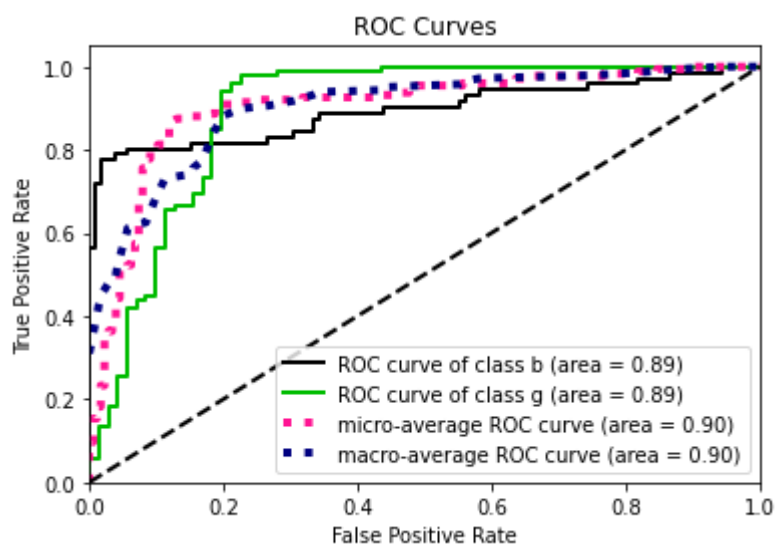
```
-----
-----
```

Accuracy Score:
0.8579545454545454
/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:



FutureWarning: Function

plot_confusion_matrix is



deprecated; Function

`plot_confusion_matrix` is

deprecated in 1.0 and will be removed in 1.2. Use one of the class methods:

ConfusionMatrixDisplay.from_predictions or

ConfusionMatrixDisplay.from_estimator. warnings.warn(msg,

category=FutureWarning) -----

DECISION TREE CLASSIFIER

```
In [46]: #Decision tree classifier
classifier=DecisionTreeClassifier()
classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test)
print('Decision Tree Classifier:')
tester(classifier,X_test,y_test,y_pred)
print('-----\n\n\n')
```

Decision Tree Classifier:

Confusion Matrix

```
[[ 59 12]
 [ 5 100]]
```

Performance Evaluation: precision

recall f1-score support

	b	0.92	0.83	0.87
71	g	0.89	0.95	0.92
105				

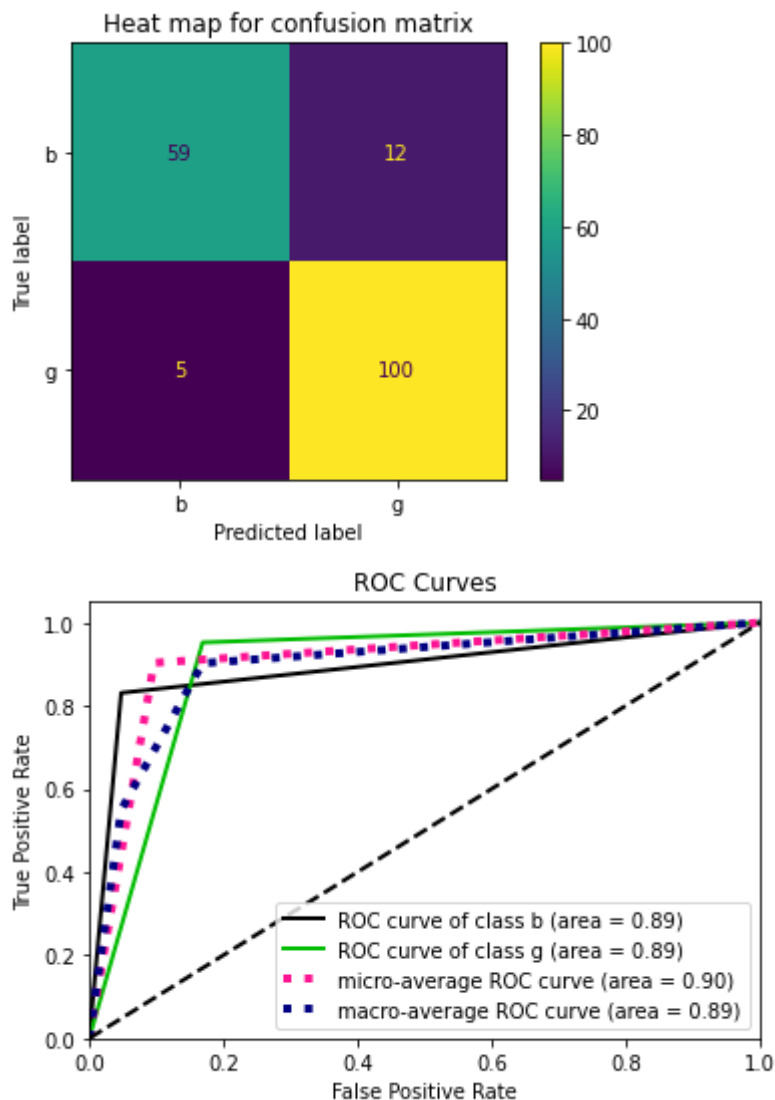
	accuracy		0.90	
176	macro avg	0.91	0.89	0.90
176	weighted avg	0.90	0.90	0.90
176				

Accuracy Score:

0.9034090909090909

```
/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:
FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator.
```

```
warnings.warn(msg, category=FutureWarning)
```



RANDOM FOREST MODEL

```
In [45]: #Random forest model
classifier=RandomForestClassifier()
classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test)
print('Random Forest Classifier:')
tester(classifier,X_test,y_test,y_pred)
print('-----\n\n\n')
```

```
Random Forest Classifier:
Confusion Matrix
[[ 63   8]
 [  4 101]]
```



```

-----
-----
Preformance Evaluation:                precision
recall  f1-score  support

```

```

          b          0.94          0.89          0.91
71          g          0.93          0.96          0.94
105

```

```

          accuracy                0.93
176  macro avg          0.93          0.92          0.93
176 weighted avg          0.93          0.93          0.93
176

```

```

-----
-----
Accuracy Score:

```

```

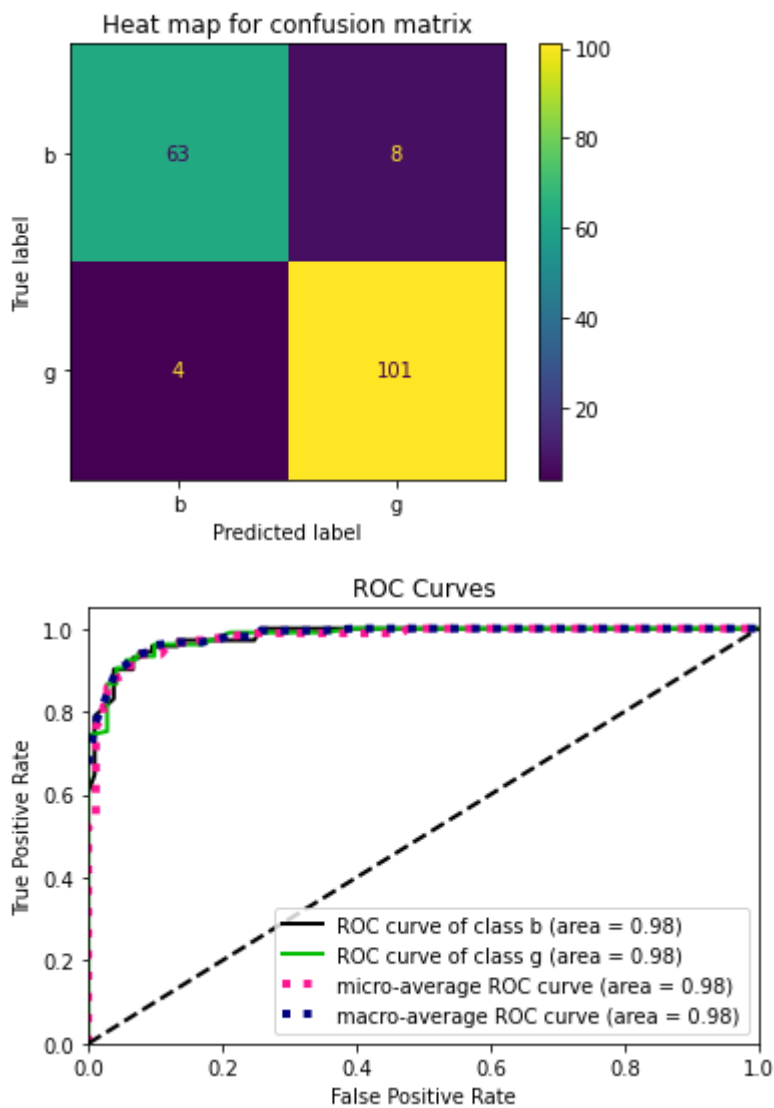
0.9318181818181818

```

```

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:
FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator.
  warnings.warn(msg, category=FutureWarning)

```



NAIVE BAYES CLASSIFIER

In [47]:

```
#Gaussian naive bayes
classifier=GaussianNB()
classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test)
print('Gaussian Naive Bayes:')
tester(classifier,X_test,y_test,y_pred)
print('-----\n\n\n')

#Bernoulli naive bayes
classifier=BernoulliNB()
classifier.fit(X_train,y_train) y_pred
= classifier.predict(X_test)
print('Bernoulli Naive Bayes:')
tester(classifier,X_test,y_test,y_pred)
print('-----\n\n\n')
```

Gaussian Naive Bayes:
Confusion Matrix
[[65 6]

[10 95]]

```
-----
-----
Preformance Evaluation:                precision
recall  f1-score  support
      b      0.87      0.92      0.89
71      g      0.94      0.90      0.92
105

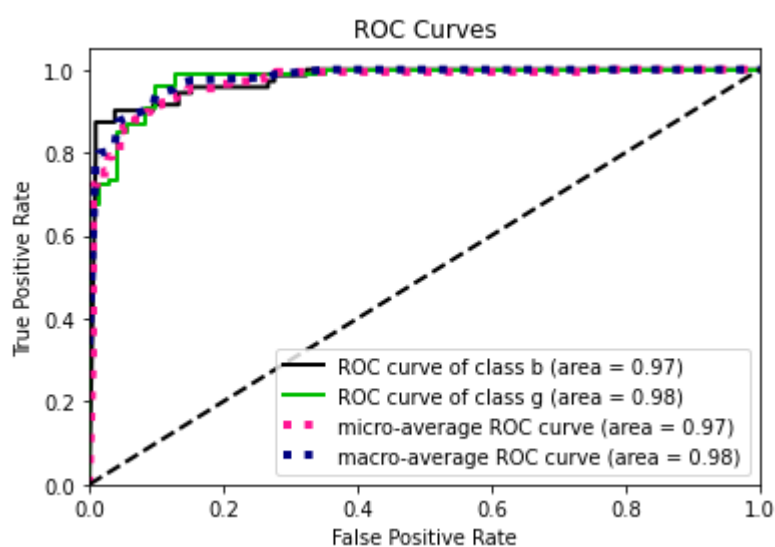
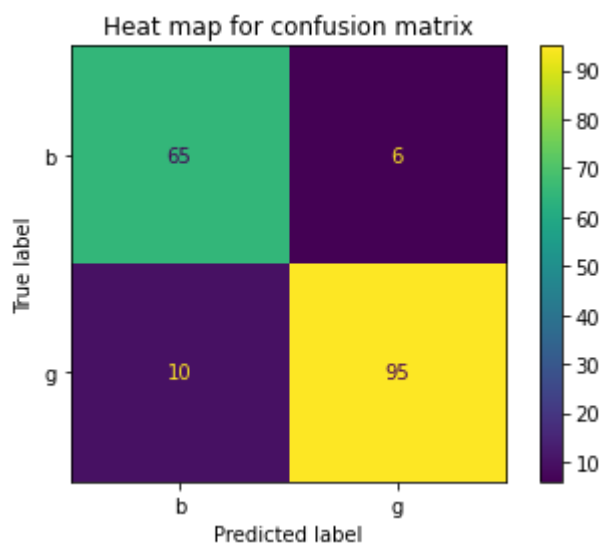
      accuracy                0.91
176  macro avg      0.90      0.91      0.91
176  weighted avg      0.91      0.91      0.91
176
```

Accuracy Score:

0.9090909090909091

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one
of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator. warnings.warn(msg,
category=FutureWarning)



Bernoulli Naive Bayes:

Confusion Matrix

```
[[53 18]
 [14 91]]
```

Preformance Evaluation:

recall f1-score support

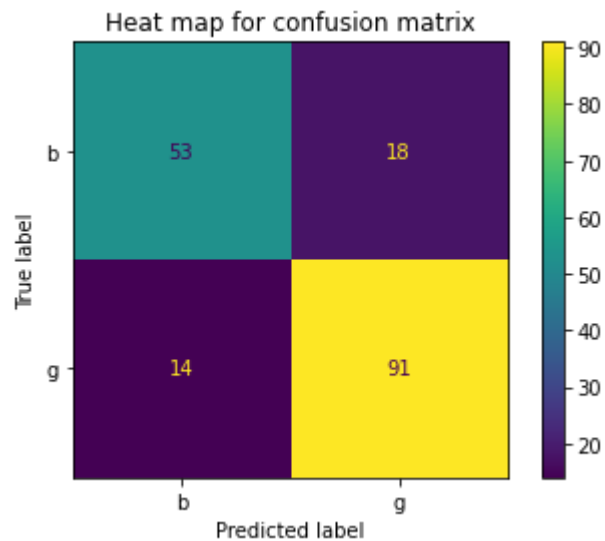
		b	g	support
71		0.79	0.75	0.77
105		0.83	0.87	0.85

		accuracy	0.82
176	macro avg	0.81	0.81
176	weighted avg	0.82	0.82
176			

Accuracy Score:

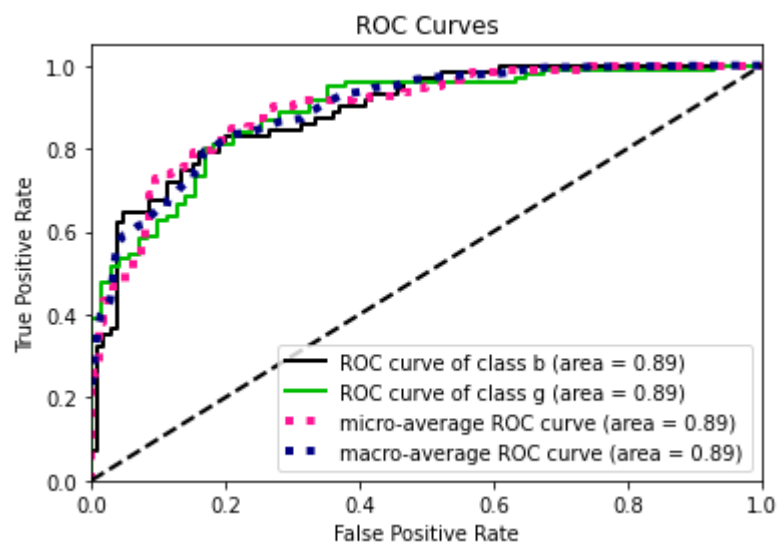
0.8181818181818182

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:



FutureWarning: Function

plot_confusion_matrix is



deprecated; Function

`plot_confusion_matrix` is

deprecated in 1.0 and will be removed in 1.2. Use one of the class methods:

ConfusionMatrixDisplay.from_predictions or

```
ConfusionMatrixDisplay.from_estimator.    warnings.warn(msg,
```

```
category=FutureWarning) -----
```

IONOSPHERE DATASET

	Accuracy	Precision	Recall	F-Measure
SVM CLASSIFIER				
Linear	0.880	0.90	0.86	0.87
Polynomial (2)	0.784	0.87	0.73	0.74
Polynomial (3)	0.670	0.82	0.59	0.55
Gaussian	0.91	0.93	0.90	0.91
Sigmoid	0.857	0.90	0.83	0.84
DECISION TREE CLASSIFIER				
Decision Tree	0.903	0.91	0.89	0.90
RANDOM FOREST CLASSIFIER				
Random Forest	0.931	0.93	0.92	0.93
NAÏVE BAYES CLASSIFIER				
Gaussian	0.909	0.90	0.91	0.91
Bernoulli	0.818	0.81	0.81	0.81

From the above tabulated data, we can conclude that at **50:50** train test split ratio in the

IONOSPHERE dataset, **Random Forest Classifier** produces the maximum accuracy followed by the **SVM Gaussian classifier**.

Question2:

Implement an ANN based model for classification task on the following datasets: **(CO2)** (10)

1. Iris plants dataset:

<https://archive.ics.uci.edu/ml/datasets/Iris/>

2. Diabetes dataset:

<https://www4.stat.ncsu.edu/~boos/var.select/diabetes.html>

3. Wisconsin Breast Cancer Dataset:

4. [https://archive.ics.uci.edu/ml/datasets/Breast+Cancer+Wisconsin+\(Diagnostic\)](https://archive.ics.uci.edu/ml/datasets/Breast+Cancer+Wisconsin+(Diagnostic))

Iris Dataset

In [4]:

```
import pandas as pd
import numpy as np

# Dataset Preparation
df = pd.read_csv("iris.data", header=None)

col_name = ['Sepal Length', 'Sepal Width', 'Petal Length', 'Petal Width', 'Class']

df.columns = col_name

X = df.drop(['Class'], axis=1)
y = df['Class']

from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y, train_size=0.7, test_size=0.3, random_state=42)

# Feature Scaling
from sklearn.preprocessing import StandardScaler

sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

In [5]:

```
# Classification using MLP
from sklearn.neural_network import MLPClassifier

classifier = MLPClassifier()
classifier.fit(X_train,y_train)

y_pred = classifier.predict(X_test)

from sklearn.metrics import classification_report, confusion_matrix, accuracy_score

print("Confusion Matrix:")
print(confusion_matrix(y_test, y_pred))

print("-----")
print("-----")

print("Performance Evaluation")
print(classification_report(y_test, y_pred))

print("-----")
print("-----")

print("Accuracy:")
print(accuracy_score(y_test, y_pred))

import matplotlib.pyplot as plt
from sklearn.metrics import plot_confusion_matrix
plot_confusion_matrix(classifier, X_test, y_test)
plt.show()
```

Confusion Matrix:

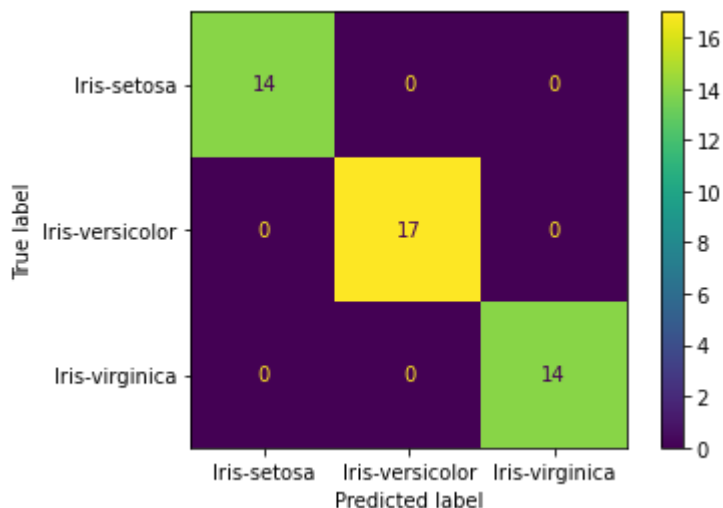
```
[[14  0  0]
 [ 0 17  0]
 [ 0  0 14]]
```


	recall	f1-score	support	precision	
Iris-setosa	1.00	1.00	1.00	14	
Iris-versicolor	1.00	1.00	1.00	17	Iris-virginica
	1.00	1.00	1.00	14	
accuracy				1.00	45
macro avg	1.00	1.00	1.00	45	
weighted avg	1.00	1.00	1.00	45	

Accuracy:

1.0

```
/usr/local/lib/python3.7/dist-packages/sklearn/neural_network/_multilayer_perceptron.py:696:
ConvergenceWarning: Stochastic Optimizer: Maximum iterations (200)
reached and the optimization hasn't converged yet.
ConvergenceWarning,
/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87: FutureWarning: Function plot_confusion_matrix is deprecated; Function `plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: ConfusionMatrixDisplay.from_predictions or ConfusionMatrixDisplay.from_estimator.
warnings.warn(msg, category=FutureWarning)
```



Diabetes Dataset

In [6]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.datasets import load_diabetes

# preparing the dataset

dataset = load_diabetes()

X = np.delete(dataset.data,1,1)
y = dataset.data[:,1]

# as in the dataset Male or Female is not mentioned properly so we assume the first unique

data_sex_type = np.unique(y);
y = list(map(lambda x : 'M' if x == data_sex_type[0] else 'F' , y));

target_name = ['M','F']
feature_name = list(filter(lambda x : x != 'sex',dataset.feature_names));

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

In [7]:

```
# Classification
from sklearn.neural_network import MLPClassifier
classifier = MLPClassifier(max_iter=100)

#####
# Showing all the parameters

from pprint import pprint
# Look at parameters used by our current forest
print('Parameters currently in use:\n')
pprint(classifier.get_params())

#####
# Creating a set of important sample features

parameter_space = {
    'hidden_layer_sizes': [(50,50,50), (50,100,50), (100,)],
    'activation': ['tanh', 'relu'],
    'solver': ['sgd', 'adam'],
    'alpha': [0.0001, 0.05],
    'learning_rate': ['constant', 'adaptive'],
}
pprint(parameter_space)

#####

from sklearn.model_selection import GridSearchCV

# Use the random grid to search for best hyperparameters
# First create the base model to tune
classifier = MLPClassifier(max_iter=100)
# Random search of parameters, using 3 fold cross validation,
# search across 100 different combinations, and use all available cores

rf_random = GridSearchCV(classifier, parameter_space, n_jobs=-1, cv=3)
rf_random.fit(X_train, y_train)

y_pred = rf_random.predict(X_test)

from sklearn.metrics import classification_report, confusion_matrix, accuracy_score

print("Confusion Matrix:")
print(confusion_matrix(y_test, y_pred))

print("-----")
print("-----")

print("Performance Evaluation")
print(classification_report(y_test, y_pred))

print("-----")
print("-----")

print("Accuracy:")
print(accuracy_score(y_test, y_pred))
```

```
import matplotlib.pyplot as plt
from sklearn.metrics import plot_confusion_matrix
plot_confusion_matrix(rf_random, X_test, y_test)
plt.show()
```

Parameters currently in use:

```
{'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': 100,
 '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}
{'activation': ['tanh', 'relu'],
 'alpha': [0.0001, 0.05],
 'hidden_layer_sizes': [(50, 50, 50), (50, 100, 50), (100,)],
 'learning_rate': ['constant', 'adaptive'],
 'solver': ['sgd', 'adam']}
```

Confusion Matrix:

```
[[42 20]
 [22 49]]
```

Performance Evaluation

recall f1-score support

precision

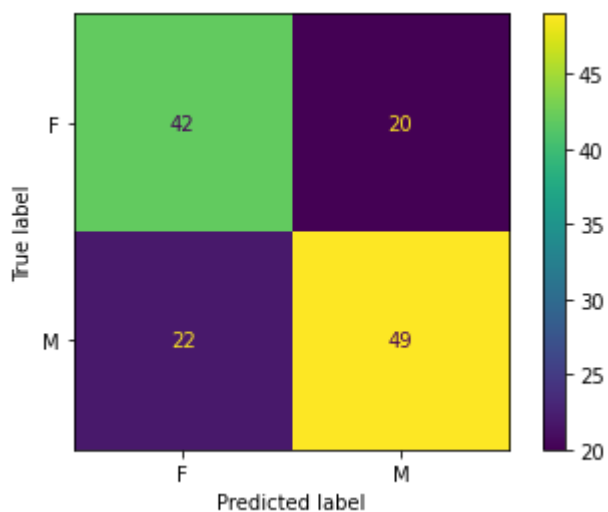
```
      F      0.66      0.68      0.67      62
M      0.71      0.69      0.70      71
```

```
      accuracy      0.68      133
macro avg      0.68      0.68      0.68      133 weighted
avg      0.69      0.68      0.68      133
```

Accuracy:

0.6842105263157895

```
/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:
FutureWarning: Function plot_confusion_matrix is deprecated; Function
`plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2.
Use one of the class methods: ConfusionMatrixDisplay.from_predictions or
ConfusionMatrixDisplay.from_estimator. warnings.warn(msg,
category=FutureWarning)
```



Wisconsin Breast Cancer Dataset

In [9]:

```
import pandas as pd
import numpy as np

# Dataset Preparation
df = pd.read_csv("wdbc.data", header=None)

col_name = ['1', 'Class', '3', '4', '5', '6', '7', '8', '9', '10', '11', '12', '13', '14', '15', '16', '17',
            '20', '21', '22', '23', '24', '25', '26', '27', '28', '29', '30', '31', '32']

df.columns = col_name

X = df.drop(['1', 'Class'], axis=1)
y = df['Class']

from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y, train_size=0.7, test_size=0.3, random_state=42)

# Feature Scaling
from sklearn.preprocessing import StandardScaler

sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)

# Classification using MLP
from sklearn.neural_network import MLPClassifier

classifier = MLPClassifier()
classifier.fit(X_train, y_train)

y_pred = classifier.predict(X_test)

from sklearn.metrics import classification_report, confusion_matrix, accuracy_score

print("Confusion Matrix:")
print(confusion_matrix(y_test, y_pred))

print("-----")
print("-----")

print("Performance Evaluation")
print(classification_report(y_test, y_pred))

print("-----")
print("-----")

print("Accuracy:")
print(accuracy_score(y_test, y_pred))

import matplotlib.pyplot as plt
from sklearn.metrics import plot_confusion_matrix
plot_confusion_matrix(classifier, X_test, y_test)
plt.show()
```

Confusion Matrix:

```
[[111  1]
 [  2 57]]
```

Performance Evaluation

recall f1-score support

precision

	B	0.98	0.99	0.99	112
M	0.98	0.97	0.97	59	
	accuracy			0.98	171
macro avg	0.98	0.98	0.98	0.98	171 weighted
avg	0.98	0.98	0.98	171	

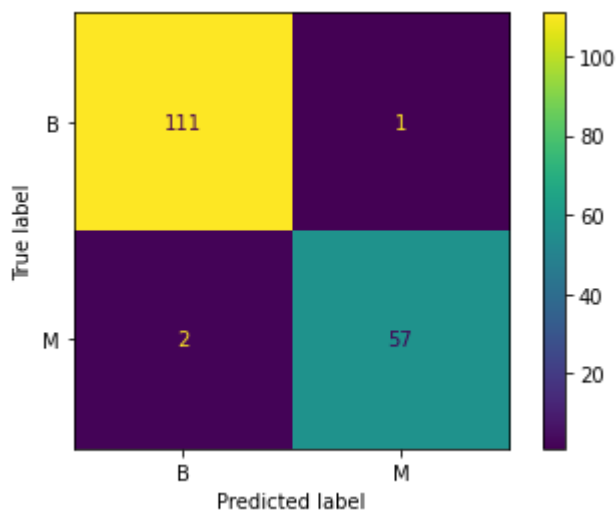
Accuracy:

0.9824561403508771

/usr/local/lib/python3.7/dist-packages/sklearn/neural_network/_multilayer_perceptron.py:696: ConvergenceWarning: Stochastic Optimizer: Maximum iterations (200) reached and the optimization hasn't converged yet.

ConvergenceWarning,

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87: FutureWarning: Function plot_confusion_matrix is deprecated; Function `plot_confusion_matrix` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: ConfusionMatrixDisplay.from_predictions or ConfusionMatrixDisplay.from_estimator. warnings.warn(msg, category=FutureWarning)



Question5:

Apply any two of the comparisons for the clustering task: **(CO4) (10)**

- K-means versus, K-medoids/PAM,
- Dendrogram versus AGNES versus BIRCH
- DBSCAN versus OPTICS

on **the Wine Dataset:**

<https://archive.ics.uci.edu/ml/datasets/wine>

and use the following performance measures

1. Silhouette Coefficient
2. Calinski-Harabasz Index
3. Davies-Bouldin Index

K-Means

In [1]:

```
#importing libraries
import numpy as np
import pandas as pd
import sklearn as sk
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.cluster import KMeans
from sklearn.datasets import load_wine
```

In [2]:

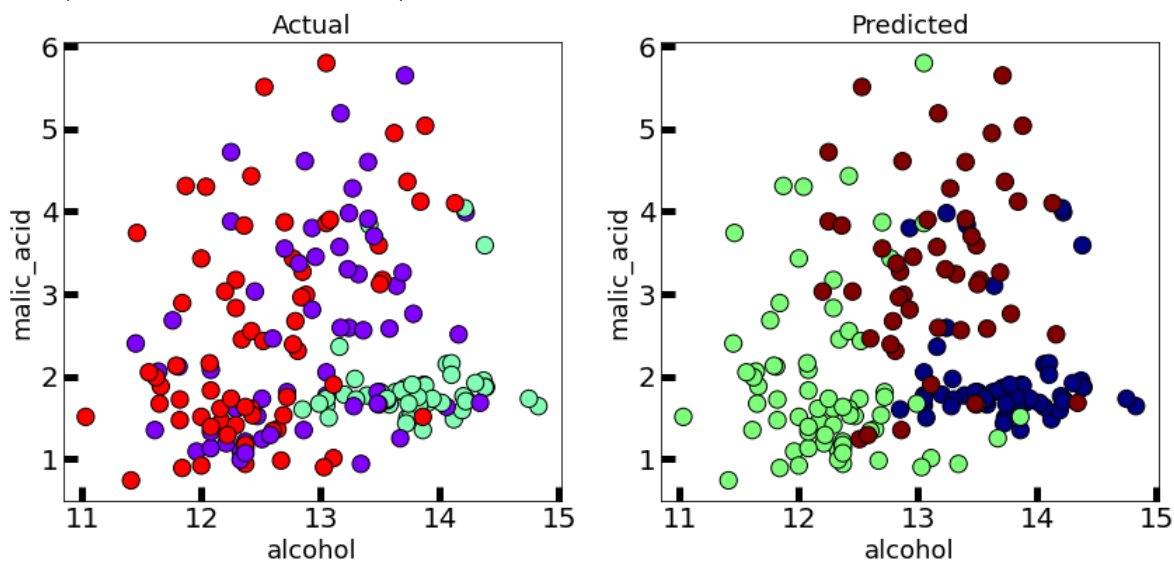
```
wine=load_wine()
x = wine.data
df=pd.DataFrame(data=x, columns=wine.feature_names)
kmeans = KMeans(init="random", n_clusters=3, n_init=10, max_iter=300, random_state=42)
y = kmeans.fit_predict(x)
```

In [3]:

```
fig, axes = plt.subplots(1, 2, figsize=(14,6))
axes[0].scatter(x=df['alcohol'], y=df['malic_acid'], c=y, cmap='rainbow',edgecolor='k', s=1)
axes[1].scatter(x=df['alcohol'], y=df['malic_acid'], c=wine.target,
cmap='jet',edgecolor='k axes[0].set_xlabel('alcohol', fontsize=18)
axes[0].set_ylabel('malic_acid', fontsize=18) axes[1].set_xlabel('alcohol', fontsize=18)
axes[1].set_ylabel('malic_acid', fontsize=18)
axes[0].tick_params(direction='in', length=10, width=5, colors='k', labels=20)
axes[1].tick_params(direction='in', length=10, width=5, colors='k',
labels=20) axes[0].set_title('Actual', fontsize=18)
axes[1].set_title('Predicted', fontsize=18)
```

Out[3]:

Text(0.5, 1.0, 'Predicted')



[4]:

In

```
from sklearn.metrics import silhouette_score
print("The silhouette score is :")
silhouette_score(x, kmeans.labels_)
```

The silhouette score is :

Out[4]:

0.5711381937868844

In [5]:

```
from sklearn.metrics import calinski_harabasz_score
print("The calinski harabasz score is :")
calinski_harabasz_score(x, kmeans.labels_)
```

The calinski harabasz score is :

Out[5]:

561.815657860671

K-medoids

In [7]:

```
#importing libraries
import numpy as np
import pandas as pd
import sklearn as sk
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn_extra.cluster import KMedoids
from sklearn.datasets import load_wine
```

In [8]:

```
wine=load_wine()
x = wine.data
df=pd.DataFrame(data=x, columns=wine.feature_names)
kmedoid = KMedoids(init="heuristic", n_clusters=3, max_iter=300, random_state=42)
y = kmedoid.fit_predict(x)
```

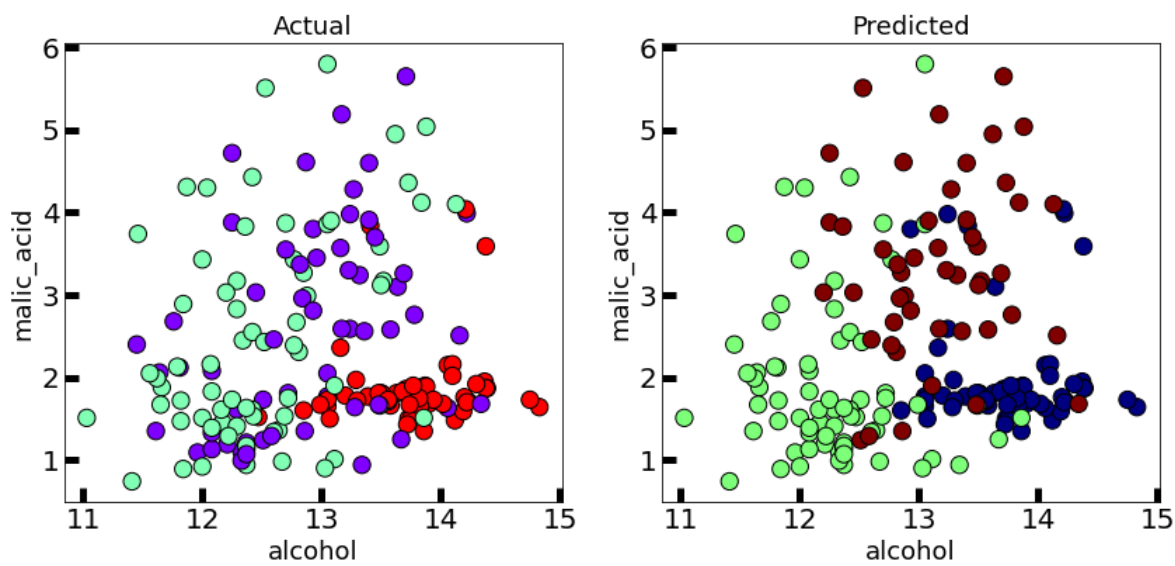
[9]:

```
fig, axes = plt.subplots(1, 2, figsize=(14,6))
axes[0].scatter(x=df['alcohol'], y=df['malic_acid'], c=y, cmap='rainbow', edgecolor='k', s=1)
axes[1].scatter(x=df['alcohol'], y=df['malic_acid'], c=wine.target,
cmap='jet', edgecolor='k')
axes[0].set_xlabel('alcohol', fontsize=18)
axes[0].set_ylabel('malic_acid', fontsize=18)
axes[1].set_xlabel('alcohol', fontsize=18)
axes[1].set_ylabel('malic_acid', fontsize=18)
axes[0].tick_params(direction='in', length=10, width=5, colors='k', labelsz=20)
axes[1].tick_params(direction='in', length=10, width=5, colors='k',
labelsz=20)
axes[0].set_title('Actual', fontsize=18)
axes[1].set_title('Predicted', fontsize=18)
```

In

```
Out[9]: Text(0.5, 1.0,
```

```
'Predicted')
```



In [10]:

```
from sklearn.metrics import silhouette_score
print("The silhouette score is :")
silhouette_score(x, kmedoid.labels_)
```

The silhouette score is :

Out[10]:

0.5666480408636575

[11]:

```
from sklearn.metrics import calinski_harabasz_score
print("The calinski harabasz score is :")
calinski_harabasz_score(x, kmedoid.labels_)
```

The calinski harabasz score is :

Out[11]:

539.3792353535451

In

Dendrogram

In [12]:

```
#importing libraries
import numpy as np
import pandas as pd
import sklearn as sk
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.datasets import load_wine
```

In [13]:

```
wine=load_wine()
x = wine.data
df=pd.DataFrame(data=x, columns=wine.feature_names)
```

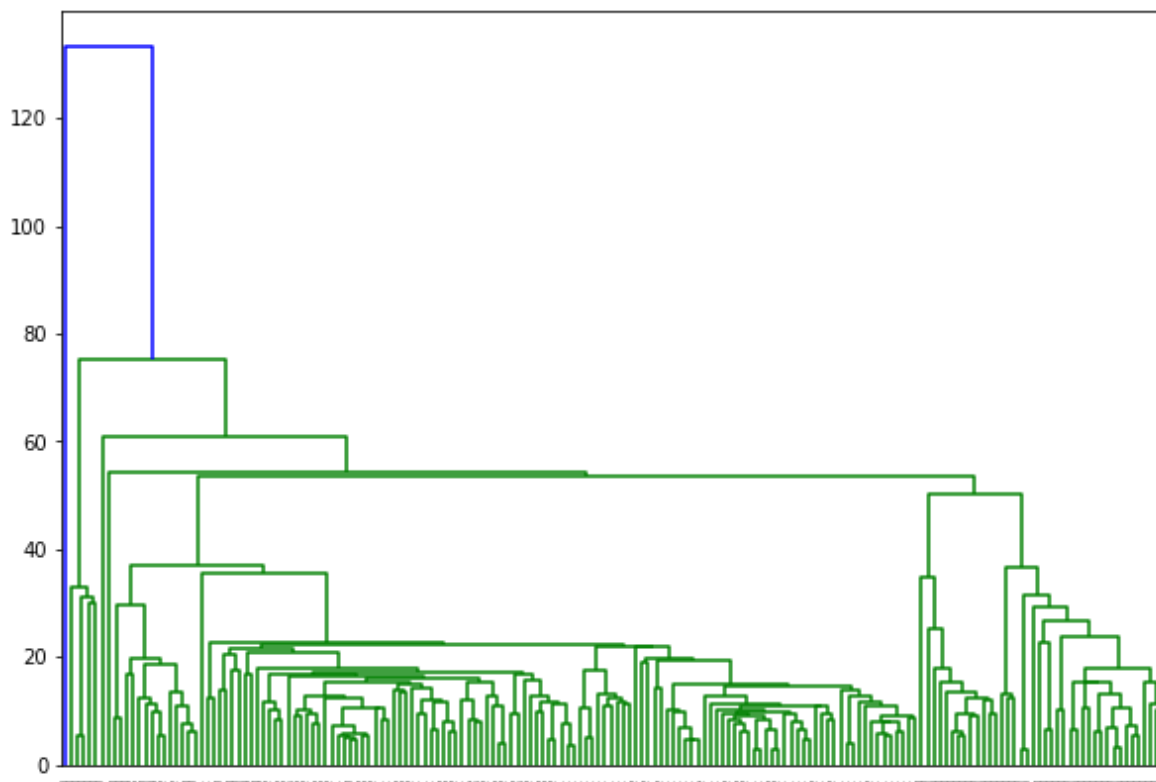
[14]:

```
from scipy.cluster.hierarchy import dendrogram, linkage

linked = linkage(x, 'single')
plt.figure(figsize=(10,7))

dendrogram(linked,
            orientation='top',
            labels=wine.target,
            distance_sort='descending',
            show_leaf_counts=True)

plt.show()
```



In

Since dendrogram illustrates how each cluster is composed by drawing a U-shaped link between a nonsingleton cluster and its children, evaluation metrics cannot be applied on this

Agnes

In [15]:

```
#importing libraries
import numpy as np
import pandas as pd
import sklearn as sk
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.datasets import load_wine
```

[16]:

```
wine=load_wine()
x = wine.data
df=pd.DataFrame(data=x, columns=wine.feature_names)
```

In [17]:

```
from sklearn.cluster import AgglomerativeClustering

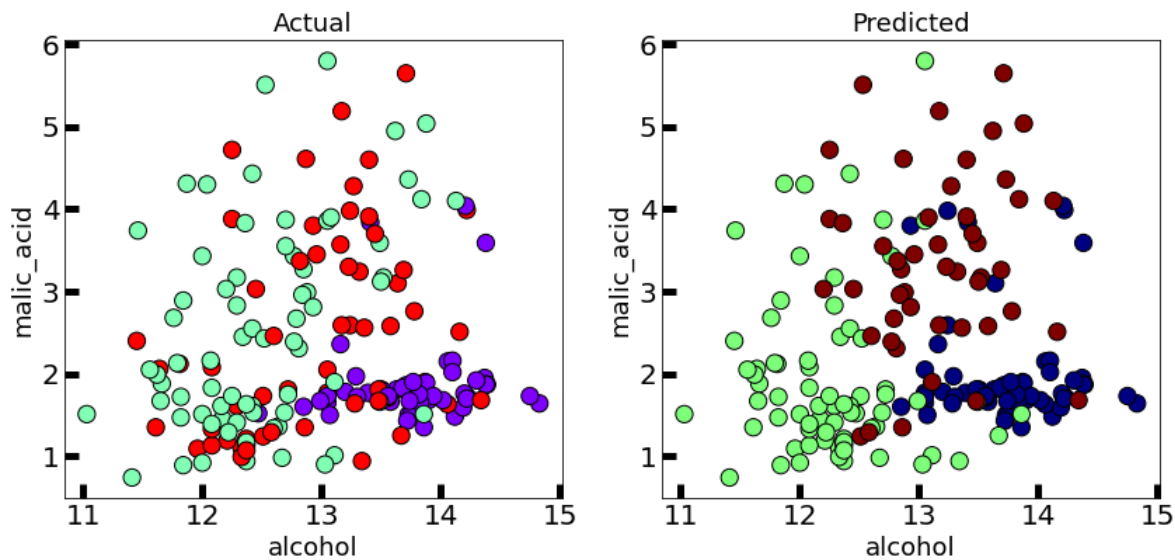
cluster = AgglomerativeClustering(n_clusters=3, affinity='euclidean', linkage='ward')
y = cluster.fit_predict(x)
```

In [18]:

```
fig, axes = plt.subplots(1, 2, figsize=(14,6))
axes[0].scatter(x=df['alcohol'], y=df['malic_acid'], c=y, cmap='rainbow',edgecolor='k', s=1)
axes[1].scatter(x=df['alcohol'], y=df['malic_acid'], c=wine.target,
cmap='jet',edgecolor='k axes[0].set_xlabel('alcohol', fontsize=18)
axes[0].set_ylabel('malic_acid', fontsize=18) axes[1].set_xlabel('alcohol', fontsize=18)
axes[1].set_ylabel('malic_acid', fontsize=18)
axes[0].tick_params(direction='in', length=10, width=5, colors='k', labelsz=20)
axes[1].tick_params(direction='in', length=10, width=5, colors='k',
labelsz=20) axes[0].set_title('Actual', fontsize=18)
axes[1].set_title('Predicted', fontsize=18)
```

Out[18]: Text(0.5, 1.0,
'Predicted')

In



In [19]:

```
from sklearn.metrics import silhouette_score
print("The silhouette score is :")
silhouette_score(x, cluster.labels_)
```

The silhouette score is :

Out[19]:

0.5644796401732074

[20]:

```
from sklearn.metrics import calinski_harabasz_score
print("The calinski harabasz score is :")
calinski_harabasz_score(x, cluster.labels_)
```

The calinski harabasz score is :

Out[20]:

552.851711505718

Birch

In [21]:

```
#importing libraries
import numpy as np
import pandas as pd
import sklearn as sk
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.datasets import load_wine
```

In [22]:

```
wine=load_wine()
x = wine.data
df=pd.DataFrame(data=x, columns=wine.feature_names)
```

In
In [23]:

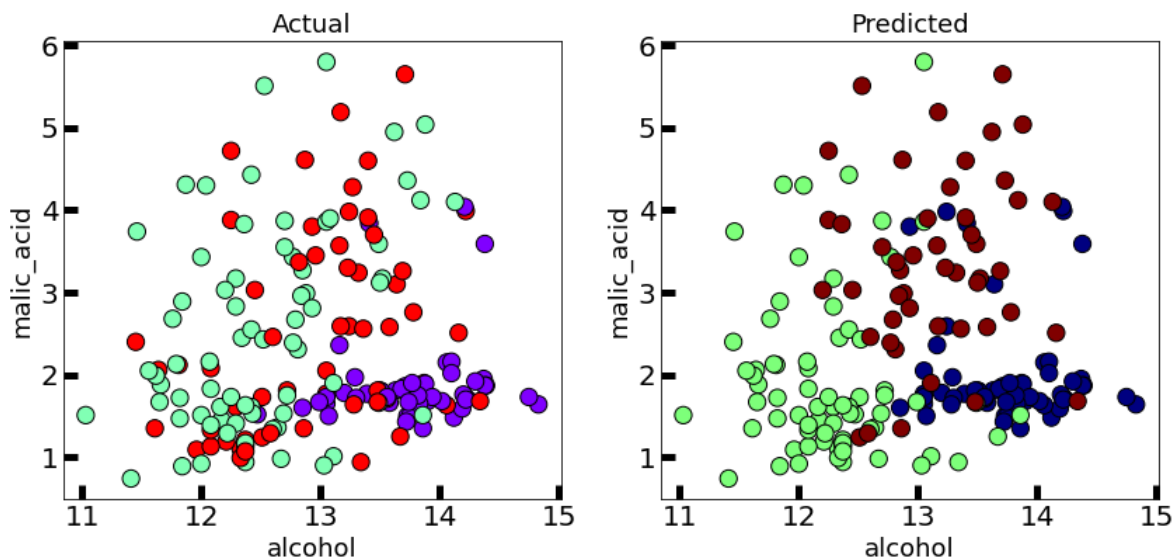
```
from sklearn.cluster import Birch

birch = Birch(n_clusters=3, compute_labels=True, branching_factor=50)
y = birch.fit_predict(x)
```

[24]:

```
fig, axes = plt.subplots(1, 2, figsize=(14,6))
axes[0].scatter(x=df['alcohol'], y=df['malic_acid'], c=y, cmap='rainbow', edgecolor='k', s=1)
axes[1].scatter(x=df['alcohol'], y=df['malic_acid'], c=wine.target,
cmap='jet', edgecolor='k')
axes[0].set_xlabel('alcohol', fontsize=18)
axes[0].set_ylabel('malic_acid', fontsize=18)
axes[1].set_xlabel('alcohol', fontsize=18)
axes[1].set_ylabel('malic_acid', fontsize=18)
axes[0].tick_params(direction='in', length=10, width=5, colors='k', labelsz=20)
axes[1].tick_params(direction='in', length=10, width=5, colors='k',
labelsz=20)
axes[0].set_title('Actual', fontsize=18)
axes[1].set_title('Predicted', fontsize=18)
```

Out[24]: Text(0.5, 1.0,
'Predicted')



In [25]:

```
from sklearn.metrics import silhouette_score
print("The silhouette score is :")
silhouette_score(x, birch.labels_)
```

The silhouette score is :

Out[25]:

0.5644796401732074

In [26]:

```
from sklearn.metrics import calinski_harabasz_score
print("The calinski harabasz score is :")
calinski_harabasz_score(x, birch.labels_)
```

The calinski harabasz score is :

In

Out[26]:

552.851711505718

DBSCAN

[27]:

```
#importing libraries
import numpy as np
import pandas as pd
import sklearn as sk
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.datasets import load_wine

wine=load_wine()
x = wine.data
df=pd.DataFrame(data=x, columns=wine.feature_names)
```

In [28]:

```
from sklearn.cluster import DBSCAN

dbscan = DBSCAN(eps=35, algorithm='auto', metric='euclidean')
y = dbscan.fit_predict(x)
```

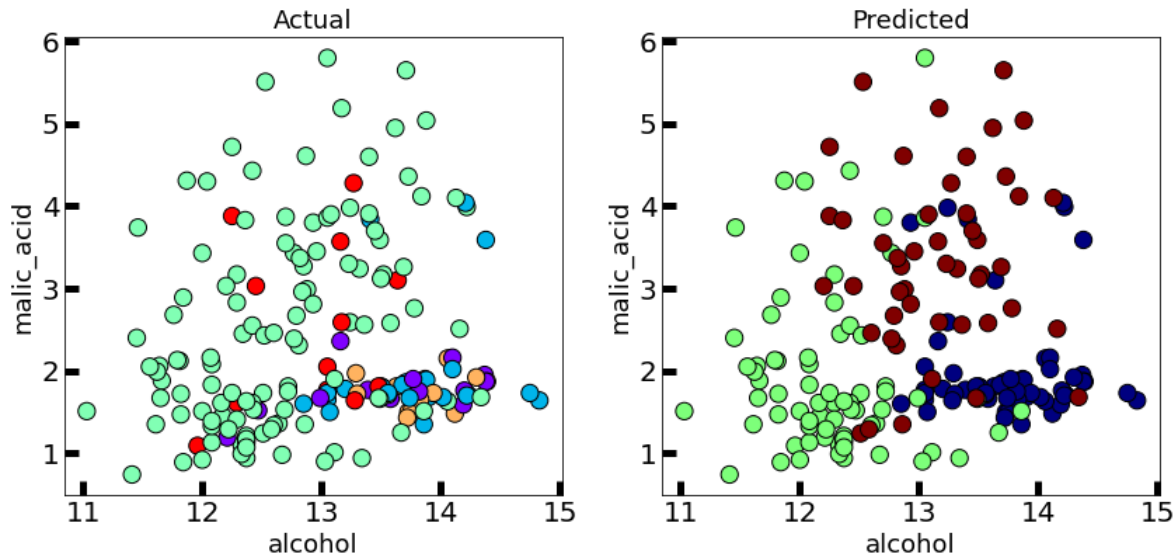
In [29]:

```
fig, axes = plt.subplots(1, 2, figsize=(14,6))
axes[0].scatter(x=df['alcohol'], y=df['malic_acid'], c=y, cmap='rainbow',edgecolor='k', s=1)
axes[1].scatter(x=df['alcohol'], y=df['malic_acid'], c=wine.target,
cmap='jet',edgecolor='k') axes[0].set_xlabel('alcohol', fontsize=18)
axes[0].set_ylabel('malic_acid', fontsize=18) axes[1].set_xlabel('alcohol', fontsize=18)
axes[1].set_ylabel('malic_acid', fontsize=18)
axes[0].tick_params(direction='in', length=10, width=5, colors='k', labelsz=20)
axes[1].tick_params(direction='in', length=10, width=5, colors='k',
labelsz=20) axes[0].set_title('Actual', fontsize=18)
axes[1].set_title('Predicted', fontsize=18)
```

Out[29]:

Text(0.5, 1.0, 'Predicted')

In



[30]:

```
from sklearn.metrics import silhouette_score
print("The silhouette score is :")
silhouette_score(x, dbscan.labels_)
```

The silhouette score is :

Out[30]:

0.4413295944891938

In [31]:

```
from sklearn.metrics import calinski_harabasz_score
print("The calinski harabasz score is :")
calinski_harabasz_score(x, dbscan.labels_)
```

The calinski harabasz score is :

Out[31]:

208.9449395725058

OPTICS

In [32]:

```
#importing libraries
import numpy as np
import pandas as pd
import sklearn as sk
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
from sklearn.datasets import load_wine

wine=load_wine()
x = wine.data
df=pd.DataFrame(data=x, columns=wine.feature_names)
```

In [33]:

In

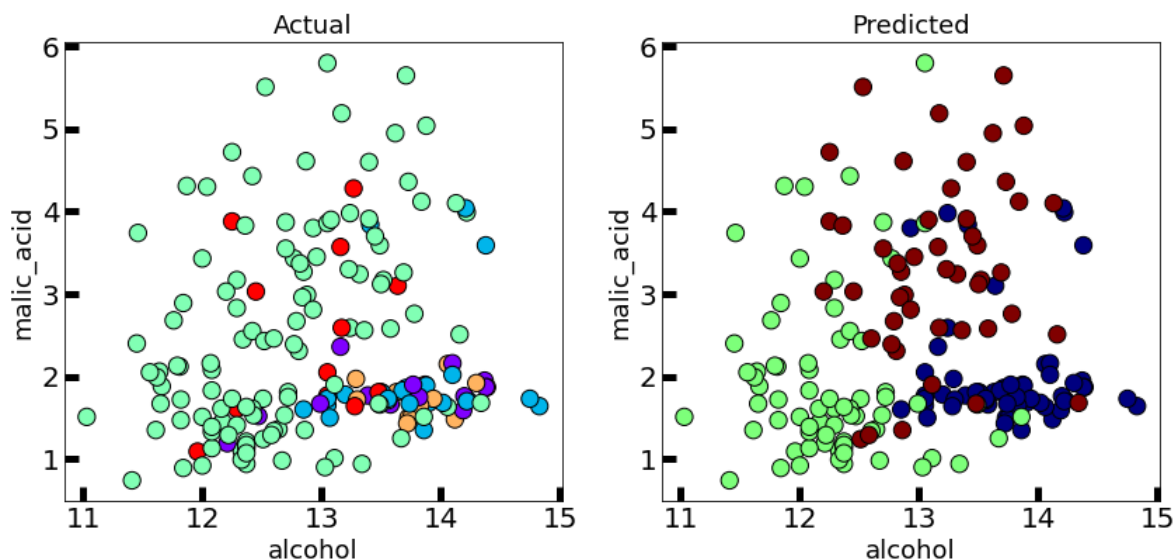
```
from sklearn.cluster import DBSCAN

dbscan = DBSCAN(eps=35, algorithm='auto', metric='euclidean')
y = dbscan.fit_predict(x)
```

[34]:

```
fig, axes = plt.subplots(1, 2, figsize=(14,6))
axes[0].scatter(x=df['alcohol'], y=df['malic_acid'], c=y, cmap='rainbow', edgecolor='k', s=1)
axes[1].scatter(x=df['alcohol'], y=df['malic_acid'], c=wine.target,
cmap='jet', edgecolor='k axes[0].set_xlabel('alcohol', fontsize=18)
axes[0].set_ylabel('malic_acid', fontsize=18) axes[1].set_xlabel('alcohol', fontsize=18)
axes[1].set_ylabel('malic_acid', fontsize=18)
axes[0].tick_params(direction='in', length=10, width=5, colors='k', labels=20)
axes[1].tick_params(direction='in', length=10, width=5, colors='k',
labels=20) axes[0].set_title('Actual', fontsize=18)
axes[1].set_title('Predicted', fontsize=18)
```

Out[34]: Text(0.5, 1.0,
'Predicted')



In [35]:

```
from sklearn.metrics import silhouette_score
print("The silhouette score is :")
silhouette_score(x, dbscan.labels_)
```

The silhouette score is :

In

Out[35]:

0.4413295944891938

[36]:

```
from sklearn.metrics import calinski_harabasz_score
print("The calinski harabasz score is :")
calinski_harabasz_score(x, dbscan.labels_)
```

The calinski harabasz score is :

Out[36]:

208.9449395725058