```
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Roll No = 001811001078

Class =IT 4th year 1st semester

Subject = Machine Learning

Question no 1

Import required header files

import pandas as pd

```
from sklearn.datasets import load_wine # import datasets
from matplotlib import pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, classification_report,
confusion_matrix
import seaborn as sns
```

from sklearn.svm import SVC #import SVM classifier

from sklearn.tree import DecisionTreeClassifier # import decision tree
classifier

from sklearn.ensemble import RandomForestClassifier # import random
forest classifier

from sklearn.naive_bayes import GaussianNB # import naive bayes
classifier

Load Wine Dataset

```
# load wine dataset
wine = load_wine()
dir(wine)

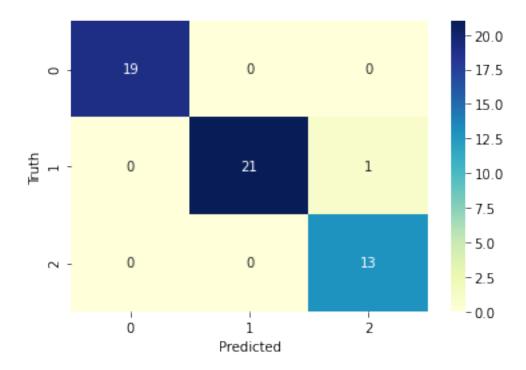
['DESCR', 'data', 'feature_names', 'frame', 'target', 'target_names']
wine.feature_names

['alcohol',
    'malic acid',
```

```
'ash',
 'alcalinity of ash',
 'magnesium',
 'total phenols',
 'flavanoids',
 'nonflavanoid phenols',
 'proanthocyanins',
 'color intensity',
 'hue',
 'od280/od315 of diluted wines',
 'proline']
df = pd.DataFrame(wine.data, columns=wine.feature names)
df.head()
   alcohol malic acid
                         ash alcalinity of ash magnesium
total phenols \
     14.23
                  1.71
                        2.43
                                            15.6
                                                       127.0
2.80
                  1.78 2.14
                                            11.2
                                                       100.0
     13.20
1
2.65
2
     13.16
                  2.36 2.67
                                            18.6
                                                       101.0
2.80
3
     14.37
                  1.95 2.50
                                            16.8
                                                       113.0
3.85
                  2.59 2.87
                                            21.0
     13.24
                                                       118.0
4
2.80
   flavanoids nonflavanoid phenols proanthocyanins color intensity
hue \
0
         3.06
                                0.28
                                                 2.29
                                                                   5.64
1.04
         2.76
                                0.26
                                                 1.28
                                                                   4.38
1.05
                                                                   5.68
         3.24
                                0.30
                                                 2.81
2
1.03
         3.49
                                0.24
                                                 2.18
                                                                   7.80
3
0.86
         2.69
                                0.39
                                                 1.82
                                                                   4.32
4
1.04
   od280/od315 of diluted wines
                                  proline
                            3.92
0
                                   1065.0
                            3.40
1
                                   1050.0
2
                            3.17
                                   1185.0
3
                            3.45
                                   1480.0
                            2.93
                                    735.0
df['target'] = wine.target
df.head()
```

```
alcohol malic acid
                        ash alcalinity of ash magnesium
total_phenols \
                  1.71 2.43
                                            15.6
                                                       127.0
     14.23
2.80
     13.20
                  1.78
                       2.14
                                            11.2
                                                       100.0
1
2.65
                  2.36 2.67
                                            18.6
                                                       101.0
2
     13.16
2.80
3
     14.37
                  1.95 2.50
                                            16.8
                                                       113.0
3.85
4
     13.24
                  2.59 2.87
                                            21.0
                                                       118.0
2.80
   flavanoids nonflavanoid phenols proanthocyanins color intensity
hue \
                                0.28
                                                 2.29
         3.06
                                                                   5.64
1.04
         2.76
                                0.26
                                                 1.28
                                                                   4.38
1
1.05
2
         3.24
                                0.30
                                                 2.81
                                                                   5.68
1.03
         3.49
                                0.24
                                                 2.18
                                                                   7.80
3
0.86
         2.69
                                0.39
                                                 1.82
                                                                   4.32
1.04
   od280/od315 of diluted wines
                                  proline
                                           target
0
                                   1065.0
                            3.92
1
                            3.40
                                   1050.0
                                                 0
2
                            3.17
                                   1185.0
                                                0
3
                            3.45
                                   1480.0
                                                 0
4
                            2.93
                                    735.0
                                                 0
wine.target names
array(['class 0', 'class 1', 'class 2'], dtype='<U7')</pre>
df['target'].value counts()
1
     71
0
     59
     48
Name: target, dtype: int64
X = df.drop(['target'], axis='columns')
len(X)
178
y = df.target
len(y)
```

```
training and test data split
X_train, X_test, y_train, y_test = train_test_split(X, y,
test size=0.3, random state=0)
Work for SVM classifier
model = SVC(kernel='linear')
model.fit(X train, y train)
SVC(kernel='linear')
model.score(X_test,y_test)
0.9814814814814815
y pred = model.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: 98.14814814814815%
Confusion Matrix:
[[19 0 0]
[ 0 21 1]
[ 0 0 13]]
Classification Report:
              precision
                           recall f1-score
                                               support
           0
                   1.00
                              1.00
                                        1.00
                                                    19
           1
                   1.00
                              0.95
                                        0.98
                                                    22
           2
                   0.93
                              1.00
                                        0.96
                                                    13
                                        0.98
                                                    54
    accuracy
                                        0.98
                                                    54
                   0.98
                              0.98
   macro avq
weighted avg
                   0.98
                              0.98
                                        0.98
                                                    54
%matplotlib inline
sns.heatmap(cf matrix,annot=True,cmap="YlGnBu")
plt.xlabel('Predicted')
plt.ylabel('Truth')
Text(33.0, 0.5, 'Truth')
```

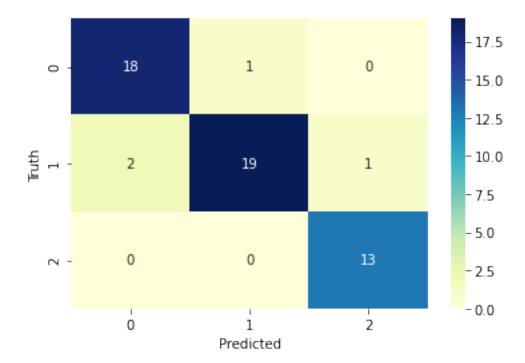


```
Work for Decision Tree classifier
model = DecisionTreeClassifier(criterion='entropy')
model.fit(X_train, y_train)
DecisionTreeClassifier(criterion='entropy')
model.score(X_test, y_test)
0.9259259259259259
y pred = model.predict(X_test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
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: 92.5925925925926%
Confusion Matrix:
[[18 1 0]
 [ 2 19 1]
 [ 0 0 13]]
Classification Report:
              precision recall f1-score
                                              support
```

```
0.90
                               0.95
                                          0.92
            0
                                                       19
            1
                    0.95
                               0.86
                                          0.90
                                                       22
            2
                    0.93
                               1.00
                                          0.96
                                                       13
                                          0.93
                                                       54
    accuracy
                    0.93
                               0.94
                                          0.93
                                                       54
   macro avg
                               0.93
                                                       54
weighted avg
                    0.93
                                          0.93
```

```
%matplotlib inline
sns.heatmap(cf_matrix,annot=True,cmap="YlGnBu")
plt.xlabel('Predicted')
plt.ylabel('Truth')
```

Text(33.0, 0.5, 'Truth')



Work for Random forest classifier

```
model = RandomForestClassifier()
model.fit(X_train,y_train)
RandomForestClassifier()
model.score(X_test,y_test)
0.9814814814815

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

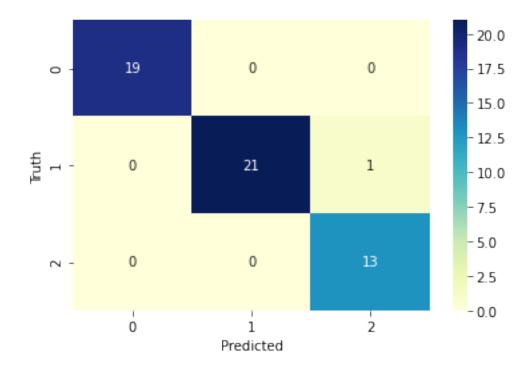
```
print("Confusion Matrix:")
print(cf_matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
Accuracy: 98.14814814814815%

Confusion Matrix:
[[19 0 0]
  [ 0 21 1]
  [ 0 0 13]]
```

Classification Report:

	precision recall f1-score		support	
0	1.00	1.00	1.00	19
1	1.00 0.93	0.95 1.00	0.98 0.96	22 13
Z	0.93	1.00	0.90	13
accuracy			0.98	54
macro avg	0.98	0.98	0.98	54
weighted avg	0.98	0.98	0.98	54

```
%matplotlib inline
sns.heatmap(cf_matrix,annot=True,cmap="YlGnBu")
plt.xlabel('Predicted')
plt.ylabel('Truth')
Text(33.0, 0.5, 'Truth')
```

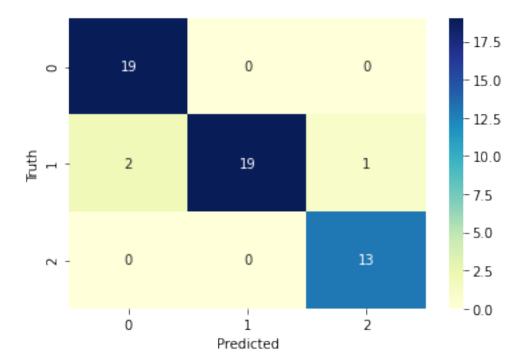


```
Work for Naive Bayes Classifier
model = GaussianNB()
model.fit(X_train,y_train)
GaussianNB()
model.score(X_test, y_test)
0.94444444444444
y pred = model.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf matrix)
print("\nClassification Report:\n")
print(classification report(y test,y pred))
Accuracy: 94.444444444444
Confusion Matrix:
[[19 0 0]
 [ 2 19 1]
 [ 0 0 13]]
Classification Report:
              precision
                           recall f1-score
                                              support
```

```
0
                    0.90
                               1.00
                                          0.95
                                                       19
            1
                    1.00
                               0.86
                                          0.93
                                                       22
            2
                    0.93
                               1.00
                                          0.96
                                                       13
                                                       54
                                          0.94
    accuracy
   macro avg
                    0.94
                               0.95
                                          0.95
                                                       54
weighted avg
                    0.95
                               0.94
                                          0.94
                                                       54
```

```
%matplotlib inline
sns.heatmap(cf_matrix,annot=True,cmap="YlGnBu")
plt.xlabel('Predicted')
plt.ylabel('Truth')
```

Text(33.0, 0.5, 'Truth')



Ionosphere Dataset

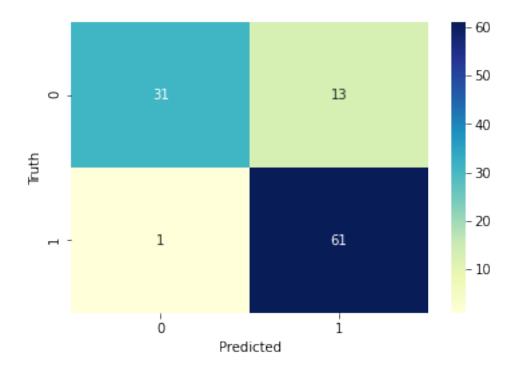
```
# load ionosphere dataset
df = pd.read_csv('ionosphere_data.csv')
df.head()
```

column_a	column_b	column_c	column_d	column_e	column_f	
column_g \	Га1 аа	0 00530	0 05000	0 05242	0 02200	
0 True 0.83398	ratse	0.99539	-0.05889	0.85243	0.02300	
1 True	False	1.00000	-0.18829	0.93035	-0.36156	-
0.10868						
2 True	False	1.00000	-0.03365	1.00000	0.00485	
1.00000						

```
False
       True
                         1.00000
                                   -0.45161
                                               1.00000
                                                          1.00000
0.71216
                 False
                                                          0.06531
       True
                         1.00000
                                   -0.02401
                                               0.94140
0.92106
             column i
                        column j
                                        column z
   column h
                                                   column aa
                                   . . .
column ab
  -0.37708
               1.00000
                         0.03760
                                        -0.51171
                                                     0.41078
                                                                -0.46168
                                   . . .
                                        -0.26569
1
  -0.93597
               1.00000
                        -0.04549
                                                    -0.20468
                                                                -0.18401
                                   . . .
2
  -0.12062
               0.88965
                         0.01198
                                        -0.40220
                                                     0.58984
                                                                -0.22145
                                   . . .
3
  -1.00000
               0.00000
                         0.00000
                                         0.90695
                                                     0.51613
                                                                 1.00000
                                   . . .
   -0.23255
               0.77152
                        -0.16399
                                        -0.65158
                                                     0.13290
                                                                -0.53206
4
                                   . . .
   column ac
               column ad
                          column_ae
                                      column_af
                                                  column_ag
                                                              column ah
column ai
     0.21266
                -0.34090
                             0.42267
                                       -0.54487
                                                    0.18641
                                                               -0.45300
0
g
1
    -0.19040
                -0.11593
                            -0.16626
                                       -0.06288
                                                   -0.13738
                                                               -0.02447
b
2
                             0.60436
                                       -0.24180
                                                    0.56045
     0.43100
                -0.17365
                                                               -0.38238
g
3
     1.00000
                -0.20099
                             0.25682
                                        1.00000
                                                   -0.32382
                                                                1.00000
b
4
     0.02431
                -0.62197
                            -0.05707
                                       -0.59573
                                                   -0.04608
                                                               -0.65697
[5 rows x 35 columns]
X = df.drop(['column ai'],axis='columns')
len(X)
351
y = df['column ai']
len(y)
351
training and test split
X_train, X_test, y_train, y_test = train_test_split(X, y,
test size=0.3, random state=0)
Work for SVM classifier
model = SVC(kernel='linear')
model.fit(X train,y train)
```

```
SVC(kernel='linear')
model.score(X test,y test)
0.8679245283018868
y pred = model.predict(X test)
print(f"Accuracy: {100 * accuracy score(y test,y pred)}%\n")
cf matrix = confusion matrix(y test,y pre\overline{d})
print("Confusion Matrix:")
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
                              0.70
                                        0.82
                                                    44
                              0.98
                                        0.90
                                                    62
                   0.82
           q
                                        0.87
                                                   106
    accuracy
                   0.90
                             0.84
                                        0.86
                                                   106
   macro avg
weighted avg
                   0.88
                              0.87
                                        0.86
                                                   106
%matplotlib inline
sns.heatmap(cf matrix,annot=True,cmap="YlGnBu")
plt.xlabel('Predicted')
plt.ylabel('Truth')
```

Text(33.0, 0.5, 'Truth')

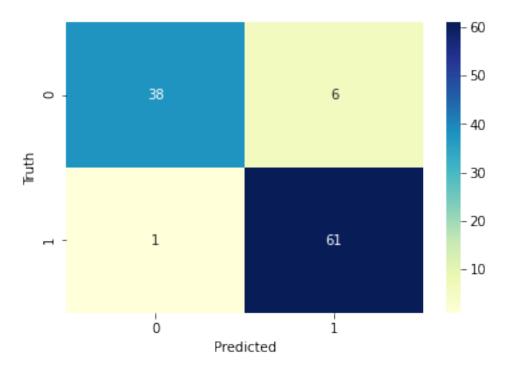


```
Work for Decision Tree classifier
model = DecisionTreeClassifier(criterion='entropy')
model.fit(X_train,y_train)
DecisionTreeClassifier(criterion='entropy')
model.score(X test,y test)
0.9339622641509434
y_pred = model.predict(X_test)
print(f"Accuracy: {100 * accuracy_score(y_test,y_pred)}%\n")
cf_matrix = confusion_matrix(y_test,y_pred)
print("Confusion Matrix:")
print(cf matrix)
print("\nClassification Report:\n")
print(classification_report(y_test,y_pred))
Accuracy: 93.39622641509435%
Confusion Matrix:
[[38 6]
 [ 1 61]]
Classification Report:
              precision recall f1-score
                                               support
           b
                   0.97
                             0.86
                                        0.92
                                                    44
```

```
0.91
                               0.98
                                          0.95
                                                       62
           g
                                          0.93
                                                      106
    accuracy
                                          0.93
                    0.94
                               0.92
                                                      106
   macro avg
weighted avg
                    0.94
                               0.93
                                          0.93
                                                      106
```

```
%matplotlib inline
sns.heatmap(cf_matrix,annot=True,cmap="YlGnBu")
plt.xlabel('Predicted')
plt.ylabel('Truth')
```

Text(33.0, 0.5, 'Truth')



Work for Random forest classifier

```
model = RandomForestClassifier()
model.fit(X_train,y_train)
RandomForestClassifier()
model.score(X_test,y_test)
0.9339622641509434
y_pred = model.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: 93.39622641509435%

Confusion Matrix:

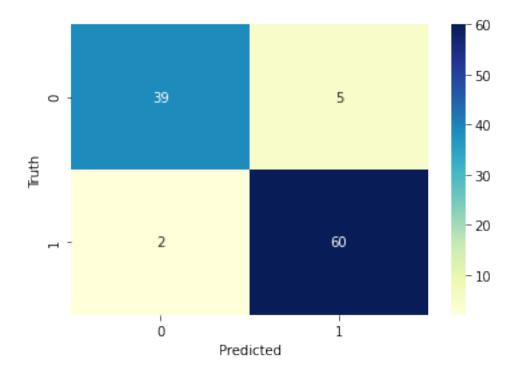
[[39 5] [2 60]]

Classification Report:

	precision	recall	f1-score	support
b q	0.95 0.92	0.89 0.97	0.92 0.94	44 62
3	0132	0.37		
accuracy macro avg	0.94	0.93	0.93 0.93	106 106
weighted avg	0.93	0.93	0.93	106

```
%matplotlib inline
sns.heatmap(cf_matrix,annot=True,cmap="YlGnBu")
plt.xlabel('Predicted')
plt.ylabel('Truth')
```

Text(33.0, 0.5, 'Truth')



```
Work for Nayes Bayes classifier
model = GaussianNB()
model.fit(X_train,y_train)
GaussianNB()
model.score(X test,y test)
0.9339622641509434
y pred = model.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: 93.39622641509435%
Confusion Matrix:
[[38 6]
 [ 1 61]]
Classification Report:
              precision recall f1-score
                                               support
                   0.97
                             0.86
                                        0.92
                                                    44
           b
                   0.91
                             0.98
                                        0.95
                                                    62
           g
                                        0.93
                                                   106
    accuracy
                   0.94
                             0.92
                                        0.93
                                                   106
   macro avg
weighted avg
                   0.94
                             0.93
                                        0.93
                                                   106
%matplotlib inline
sns.heatmap(cf_matrix,annot=True,cmap="YlGnBu")
plt.xlabel('Predicted')
plt.ylabel('Truth')
Text(33.0, 0.5, 'Truth')
```

