Q1. Apply the random forest classifier as a bagging approach to the given dataset. UC-Irvine Machine Learning Repository http://archive.ics.uci.edu/ml/) choose any dataset.

```
In [16]:
from sklearn import datasets
import pandas as pd
dataset = pd.read_csv('winequality-red.csv', sep=';')
dataset.head()
```

Out[16]:

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcohol	quality
0	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4	5
1	7.8	0.88	0.00	2.6	0.098	25.0	67.0	0.9968	3.20	0.68	9.8	5
2	7.8	0.76	0.04	2.3	0.092	15.0	54.0	0.9970	3.26	0.65	9.8	5
3	11.2	0.28	0.56	1.9	0.075	17.0	60.0	0.9980	3.16	0.58	9.8	6
4	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4	5

```
In [17]:
```

```
X = dataset.drop(columns = ['quality'])
Y = dataset['quality']
X, Y
```

Out[17]:

```
fixed acidity volatile acidity citric acid ... pH sulphates alcohol
(
                         0.700 0.00 ... 3.51 0.56 9.4
            7.4
             7.8
1
                          0.880
                                      0.00 ... 3.20
                                                        0.68
                                                                 9.8
2
             7.8
                          0.760
                                      0.04 ... 3.26
                                                                 9.8
                                                        0.65
                                      0.56 ... 3.16
                                                        0.58
3
                          0.280
                                                                 9.8
            11.2
                                     0.00 ... 3.51
                                                                 9.4
                          0.700
4
             7.4
                                                        0.56
                                           ... 3.45
             . . .
                            . . .
                                       . . .
                                                         . . .
                          0.600
                                     0.08
                                                        0.58
                                                               10.5
1594
             6.2
                                           ... 3.52
1595
             5.9
                          0.550
                                      0.10
                                                         0.76
                                                                11.2
1596
             6.3
                          0.510
                                      0.13 ... 3.42
                                                         0.75
                                                                11.0
                          0.645
                                                                10.2
1597
             5.9
                                      0.12 ... 3.57
                                                         0.71
1598
             6.0
                          0.310
                                      0.47 ... 3.39
                                                         0.66
                                                                11.0
```

```
[1599 rows x 11 columns], 0 5

1 5

2 5

3 6

4 5

...

1594 5

1595 6

1596 6

1597 5

1598 6

Name: quality, Length: 1599, dtype: int64)
```

In [18]:

```
from sklearn.model_selection import train_test_split

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2)
X_train, X_test, Y_train, Y_test
```

```
Out[18]:
      fixed acidity volatile acidity citric acid ... pH sulphates alcohol
                                             0.23 ... 3.21
1199
                7.9
                                0.580
                                                                  0.58
                                                                           9.5
1314
                7.0
                                0.360
                                             0.21
                                                   ... 3.40
                                                                   0.53
                                                                            10.1
1403
                7.2
                                             0.33
                                                   ... 3.23
                                                                          10.0
                               0.330
                                                                  1.10
                7.7
1200
                               0.570
                                             0.21 ... 3.16
                                                                  0.54
                                                                           9.8
692
                8.6
                               0.490
                                             0.51 ... 3.03
                                                                  1.17
                                                                            9.0
. . .
                . . .
                                . . .
                                             . . . . . . . .
                                                        . . .
                                                                   . . .
                                                                            . . .
                              0.390
                                             0.16 ... 3.34
                                                                 0.56
                                                                            9.3
2.5
               6.3
629
                7.6
                              0.685
                                            0.23 ... 3.21
                                                                  0.61
                                                                            9.3
                7.8
                               0.700
                                                                          10.9
1254
                                             0.06 ... 3.40
                                                                  0.69
779
                7.1
                               0.520
                                            0.03 ... 3.50
                                                                  0.60
                                                                           9.8
1379
                7.5
                               0.570
                                             0.02 ... 3.36
                                                                   0.62
                                                                           10.8
[1279 rows x 11 columns],
      fixed acidity volatile acidity citric acid ...
                                                         pH sulphates alcohol
                                                   ... 3.38
1182
               10.2
                               0.40
                                             0.40
                                                              0.86
                                                   ... 3.20
1020
               11.3
                                 0.36
                                             0.66
                                                                   0.53
                                                                           11.9
                                                   ... 3.17
54
               7.6
                                             0.15
                                 0.51
                                                                   0.63
                                                                           10.2
                                                   . . .
566
                8.7
                                 0.70
                                             0.24
                                                        3.32
                                                                   0.60
                                                                            9.0
                                                   ... 3.52
184
               6.7
                                0.62
                                             0.21
                                                                  0.58
                                                                            9.3
                                 . . .
                                              . . .
                                                   . . .
 . . .
                . . .
                                                         . . .
                                                                   . . .
                                                                            . . .
                                                   ... 3.36
1188
                6.7
                                 0.64
                                             0.23
                                                                   0.70
                                                                          10.9
                                                  ... 3.29
799
                9.4
                                0.50
                                             0.34
                                                                  0.52
                                                                           10.7
1339
                7.5
                                0.51
                                             0.02 ... 3.36
                                                                  0.54
                                                                           10.5
158
                7.1
                                0.68
                                             0.00 ... 3.48
                                                                   0.50
                                                                            9.3
                                0.69
                                                                           10.1
1306
                9.7
                                             0.32 ... 3.29
                                                                   0.62
 [320 rows x 11 columns],
1199 6
1314
       6
1403
1200
692
        5
2.5
629
        5
1254
        5
779
1379
Name: quality, Length: 1279, dtype: int64,
1182
1020
54
        6
566
        6
184
        6
1188
        5
799
        6
1339
158
1306
Name: quality, Length: 320, dtype: int64)
In [22]:
```

In [31]:

clf.fit(X train, Y train)

Y pred=clf.predict(X test)

```
from sklearn import metrics
import matplotlib.pyplot as plt
from sklearn.metrics import plot_confusion_matrix, mean_squared_error, mean_absolute_error
, make_scorer, classification_report, confusion_matrix, accuracy_score, roc_auc_score, roc_cur
```

from sklearn.ensemble import RandomForestClassifier

clf=RandomForestClassifier(n estimators=100)

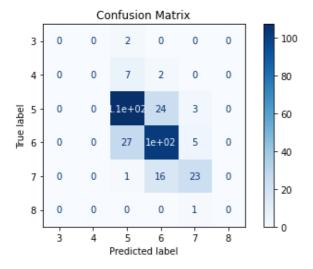
```
ve
print("Testing accuracy Score : ", metrics.accuracy_score(Y_test, Y_pred))
print(classification_report(Y_pred, Y_test))

# Plot confusion matrix
disp = plot_confusion_matrix(clf, X_test, Y_test, cmap=plt.cm.Blues)
disp.ax_.set_title("Confusion Matrix")
disp.figure_.tight_layout()

print("Confusion Matrix")
print(disp.confusion_matrix)
```

```
Testing accuracy Score: 0.725
              precision
                          recall f1-score
                                               support
           3
                   0.00
                             0.00
                                       0.00
                                                     0
                   0.00
                             0.00
                                       0.00
                                                     0
           5
                             0.74
                                       0.77
                   0.80
                                                   144
           6
                   0.76
                             0.71
                                       0.73
                                                   144
           7
                   0.57
                             0.72
                                       0.64
                                                    32
                             0.00
                                       0.00
                   0.00
                                       0.73
                                                   320
   accuracy
                   0.36
                             0.36
                                       0.36
  macro avg
                                                   320
                   0.76
                             0.72
                                       0.74
                                                   320
weighted avg
Confusion Matrix
0 11
       0 2
               0
                    0
                        0]
   Ω
       0
           7
               2
                    0
                        01
        0 107 24
                    3
                        01
   0
        0 27 102
                   5
   0
                        0.1
        0 1 16 23
   0
                        01
              0
                        0]]
```

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



Q2. Perform ADAboost algorithm for classification on any above dataset.

```
In [35]:
```

```
from sklearn.ensemble import AdaBoostClassifier
abc = AdaBoostClassifier(n_estimators=50,learning_rate=1)
model = abc.fit(X_train, Y_train)

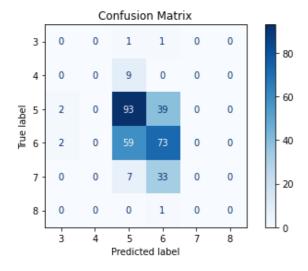
Y_pred = model.predict(X_test)
```

In [36]:

```
print("Testing accuracy Score : ", metrics.accuracy score(Y test, Y pred))
print(classification report(Y pred, Y test))
# Plot confusion matrix
disp = plot confusion matrix(abc, X test, Y test,cmap=plt.cm.Blues)
disp.ax .set title("Confusion Matrix")
disp.figure .tight layout()
print("Confusion Matrix")
print(disp.confusion_matrix)
```

```
Testing accuracy Score: 0.51875
              precision
                           recall
                                    f1-score
                                                support
           3
                    0.00
                              0.00
                                         0.00
                                                       4
           4
                    0.00
                              0.00
                                         0.00
                                                       0
           5
                    0.69
                              0.55
                                         0.61
                                                    169
           6
                    0.54
                              0.50
                                         0.52
                                                    147
           7
                    0.00
                              0.00
                                         0.00
                                                       0
           8
                                                       0
                    0.00
                              0.00
                                         0.00
                                         0.52
                                                    320
    accuracy
                    0.21
                              0.17
                                         0.19
                                                     320
   macro avg
weighted avg
                    0.62
                              0.52
                                         0.56
                                                    320
Confusion Matrix
[ [ 0 \ 0 \ 1 \ 1 \ 0 ]
                   01
     0 9 0 0
  0
                   01
 [
     0 93 39 0
   2
                   01
     0 59 73 0
   2
                   01
     0 7 33 0
```

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



0]

1 0 0]]

0

0

[0

[0