Wine Quality data test

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
%matplotlib inline
from sklearn.ensemble import RandomForestClassifier,
GradientBoostingClassifier
from sklearn.svm import SVC, LinearSVC
from sklearn.linear model import SGDClassifier
from sklearn.metrics import confusion matrix, classification report
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model selection import train test split, GridSearchCV,
cross val score, StratifiedKFold
from sklearn.linear model import LogisticRegression
from sklearn.neighbors import KNeighborsClassifier
from sklearn.naive bayes import GaussianNB
from sklearn.metrics import accuracy score
df = pd.read csv('Wine Quality Data.csv')
df.head(10)
   fixed acidity volatile acidity citric acid residual sugar
chlorides
                               0.70
             7.4
                                            0.00
                                                              1.9
0
0.076
             7.8
                               0.88
                                            0.00
                                                              2.6
1
0.098
             7.8
                               0.76
                                            0.04
                                                              2.3
2
0.092
3
            11.2
                               0.28
                                            0.56
                                                              1.9
0.075
             7.4
                               0.70
                                            0.00
                                                              1.9
0.076
             7.4
                               0.66
                                            0.00
                                                              1.8
5
0.075
             7.9
                               0.60
                                            0.06
                                                              1.6
6
0.069
7
             7.3
                               0.65
                                            0.00
                                                              1.2
0.065
             7.8
                                                              2.0
                               0.58
                                            0.02
8
0.073
             7.5
                               0.50
                                            0.36
                                                              6.1
0.071
   free sulfur dioxide total sulfur dioxide density
                                                               sulphates
                                                          Hq
0
                  11.0
                                         34.0
                                                0.9978 3.51
                                                                    0.56
1
                  25.0
                                         67.0
                                                0.9968 3.20
                                                                    0.68
```

```
0.9970 3.26
2
                   15.0
                                           54.0
                                                                       0.65
3
                   17.0
                                           60.0
                                                   0.9980
                                                           3.16
                                                                       0.58
4
                   11.0
                                           34.0
                                                   0.9978 3.51
                                                                       0.56
5
                   13.0
                                           40.0
                                                   0.9978 3.51
                                                                       0.56
6
                   15.0
                                           59.0
                                                  0.9964
                                                          3.30
                                                                       0.46
7
                   15.0
                                           21.0
                                                   0.9946 3.39
                                                                       0.47
8
                    9.0
                                           18.0
                                                   0.9968 3.36
                                                                       0.57
9
                   17.0
                                          102.0
                                                  0.9978 3.35
                                                                       0.80
             quality color
   alcohol
0
       9.4
                   5
                        red
                   5
1
       9.8
                        red
2
                   5
       9.8
                        red
3
                   6
       9.8
                        red
                   5
4
       9.4
                        red
5
                   5
       9.4
                        red
6
                   5
       9.4
                        red
7
                   7
      10.0
                        red
8
                   7
       9.5
                        red
                   5
9
      10.5
                        red
```

df['color'].value_counts()

white 4898 red 1599

Name: color, dtype: int64

X=df[['fixed_acidity',
 'volatile_acidity',
 'citric_acid',
 'residual_sugar',
 'chlorides',
 'free_sulfur_dioxide',
 'total_sulfur_dioxide',
 'density',
 'pH',
 'sulphates',
 'alcohol',
 'quality']].values
X[0:5]

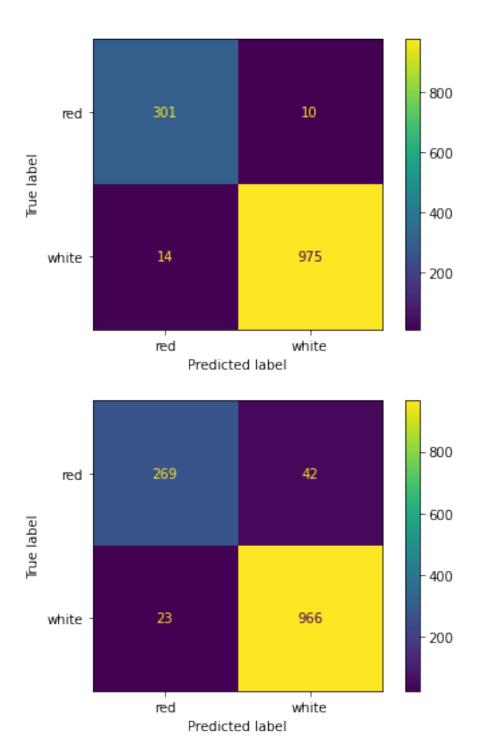
```
y=df[['color']].values
y[0:5]
array([['red'],
       ['red'],
       ['red'],
       ['red'],
       ['red']], dtype=object)
from sklearn import preprocessing
le color = preprocessing.LabelEncoder()
le_color.fit(['white','red'])
y[:,-1] = le color.transform(y[:,-1])
df['color'].value counts().plot.bar(color=['green', 'red'])
<AxesSubplot:>
  5000
  4000
  3000
  2000
  1000
     0
Y = df.color
X = df.drop('color', axis=1)
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size =
0.2, random state = 0)
def models(X train,Y train):
  #Using Logistic Regression Algorithm to the Training Set
  from sklearn.linear model import LogisticRegression
  log = LogisticRegression(random_state = 0)
  log.fit(X_train, Y_train)
```

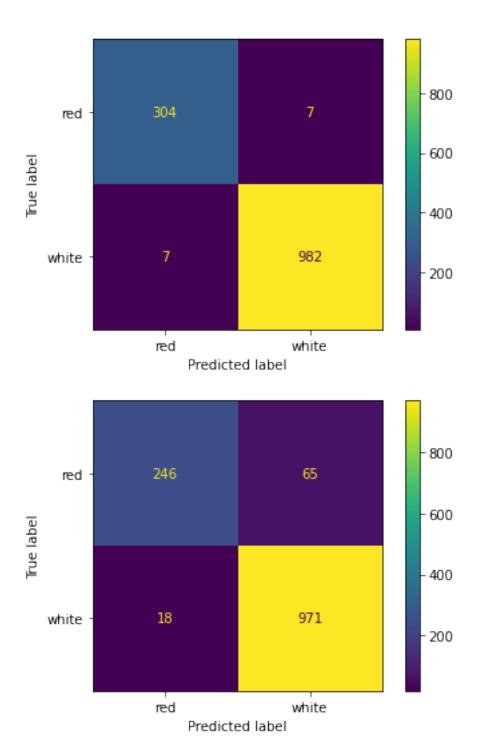
```
#Using KNeighborsClassifier Method of neighbors class to use Nearest
Neighbor algorithm
  from sklearn.neighbors import KNeighborsClassifier
  knn = KNeighborsClassifier(n neighbors = 5, metric = 'minkowski', p
  knn.fit(X train, Y train)
  #Using SVC method of svm class to use Support Vector Machine
Algorithm
  from sklearn.svm import SVC
  svc lin = SVC(kernel = 'linear', random state = 0)
  svc_lin.fit(X_train, Y train)
  #Using SVC method of svm class to use Kernel SVM Algorithm
  from sklearn.svm import SVC
  svc rbf = SVC(kernel = 'rbf', random state = 0)
  svc rbf.fit(X train, Y train)
  #Using GaussianNB method of naïve bayes class to use Naïve Bayes
Algorithm
  from sklearn.naive bayes import GaussianNB
  gauss = GaussianNB()
  gauss.fit(X train, Y train)
  #Using DecisionTreeClassifier of tree class to use Decision Tree
Alaorithm
  from sklearn.tree import DecisionTreeClassifier
  tree = DecisionTreeClassifier(criterion = 'entropy', random state =
0)
  tree.fit(X train, Y train)
  #Using RandomForestClassifier method of ensemble class to use Random
Forest Classification algorithm
  from sklearn.ensemble import RandomForestClassifier
  forest = RandomForestClassifier(n estimators = 10, criterion =
'entropy', random state = 0)
  forest.fit(X train, Y train)
  #print model accuracy on the training data.
  print('[0]Logistic Regression Training Accuracy:',
log.score(X train, Y train))
  print('[1]K Nearest Neighbor Training Accuracy:', knn.score(X train,
Y train))
  print('[2]Support Vector Machine (Linear Classifier) Training
Accuracy:', svc_lin.score(X_train, Y train))
  print('[3]Support Vector Machine (RBF Classifier) Training
Accuracy:', svc_rbf.score(X_train, Y_train))
  print('[4]Gaussian Naive Bayes Training Accuracy:',
gauss.score(X train, Y train))
```

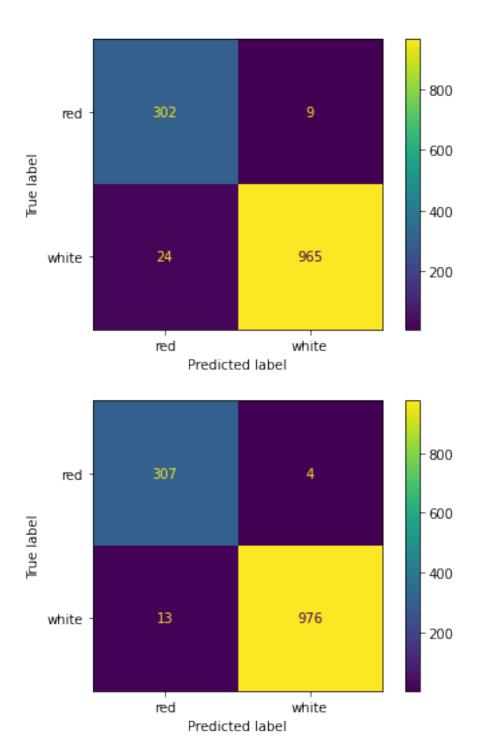
```
print('[5]Decision Tree Classifier Training Accuracy:',
tree.score(X train, Y train))
  print('[6]Random Forest Classifier Training Accuracy:',
forest.score(X train, Y train))
  return log, knn, svc lin, svc rbf, gauss, tree, forest
model = models(X train,y train)
c:\Python310\lib\site-packages\sklearn\linear model\ logistic.py:814:
ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
Increase the number of iterations (max iter) or scale the data as
shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear model.html#logistic-
regression
  n iter i = check optimize result(
[0]Logistic Regression Training Accuracy: 0.9790263613623245
[1]K Nearest Neighbor Training Accuracy: 0.9570906292091591
[2] Support Vector Machine (Linear Classifier) Training Accuracy:
0.9878776217048297
[3]Support Vector Machine (RBF Classifier) Training Accuracy:
0.9363094092745815
[4]Gaussian Naive Bayes Training Accuracy: 0.9697902636136232
[5]Decision Tree Classifier Training Accuracy: 1.0
[6]Random Forest Classifier Training Accuracy: 0.9998075812969021
from sklearn.metrics import confusion matrix
from sklearn.metrics import plot confusion matrix
l1=list()
accuracy scores = list()
precision scores = list()
for i in range(len(model)):
   cm = confusion_matrix(y test, model[i].predict(X test))
   #extracting TN, FP, FN, TP
   TN, FP, FN, TP = confusion matrix(y test,
model[i].predict(X test)).ravel()
   print(cm)
   print('Model[{}] Testing Accuracy = "{} !"'.format(i, (TP + TN) /
(TP + TN + FN + FP))
   acc=(TP + TN) / (TP + TN + FN + FP)
   print('Model[{}] Testing Precision = "{} !"'.format(i, (TP)) / (TP)
+ FP)))
   prec= (TP) / (TP + FP)
   print()# Print a new line
   plot confusion matrix(model[i], X test, y test)
```

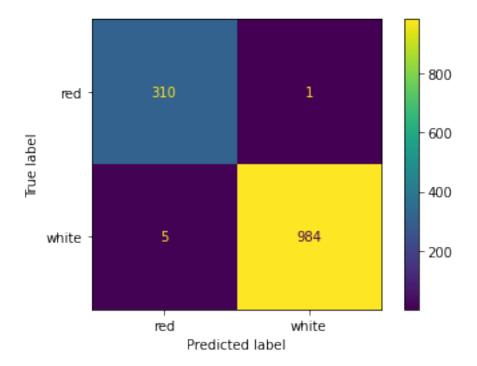
```
accuracy scores.append(acc)
   precision scores.append(prec)
   l1.append(pd.Series({'model': model[i],
'accuracy':acc,'precision':prec }))
[[301 10]
 [ 14 975]]
Model[0] Testing Accuracy = "0.9815384615384616 !"
Model[0] Testing Precision = "0.9898477157360406 !"
c:\Python310\lib\site-packages\sklearn\utils\deprecation.py:87:
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)
[[269 42]
 [ 23 96611
Model[1] Testing Accuracy = "0.95 !"
[[304 7]
7 98211
Model[2] Testing Accuracy = "0.9892307692307692 !"
Model[2] Testing Precision = "0.9929221435793731 !"
c:\Python310\lib\site-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:
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ConfusionMatrixDisplay.from estimator.
  warnings.warn(msg, category=FutureWarning)
c:\Python310\lib\site-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)
[[246 65]
 [ 18 971]]
Model[3] Testing Accuracy = "0.9361538461538461 !"
Model[3] Testing Precision = "0.9372586872586872 !"
[[302
       91
```

```
[ 24 96511
Model[4] Testing Accuracy = "0.9746153846153847 !"
Model[4] Testing Precision = "0.9907597535934292 !"
c:\Python310\lib\site-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:
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ConfusionMatrixDisplay.from estimator.
  warnings.warn(msg, category=FutureWarning)
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FutureWarning: Function plot confusion matrix is deprecated; Function
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1.2. Use one of the class methods:
ConfusionMatrixDisplay.from predictions or
ConfusionMatrixDisplay.from estimator.
  warnings.warn(msg, category=FutureWarning)
[[307
 [ 13 976]]
Model[5] Testing Accuracy = "0.9869230769230769 !"
Model[5] Testing Precision = "0.9959183673469387 !"
[[310
        11
 [ 5 984]]
Model[6] Testing Accuracy = "0.9953846153846154 !"
Model[6] Testing Precision = "0.9989847715736041 !"
c:\Python310\lib\site-packages\sklearn\utils\deprecation.py:87:
FutureWarning: Function plot confusion matrix is deprecated; Function
plot_confusion_matrix` is \overline{deprecated in 1.0} and will be removed in
1.2. Use one of the class methods:
ConfusionMatrixDisplay.from predictions or
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c:\Python310\lib\site-packages\sklearn\utils\deprecation.py:87:
FutureWarning: Function plot confusion matrix is deprecated; Function
plot_confusion_matrix` is \overline{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)
```









```
models=['LogisticRegression','Knn','SVC_Linear','SVC_RBF','GaussianNB'
,'DecisionTree','RandomForest']
performance df =
```

pd.DataFrame({'Algorithm':models,'Accuracy':accuracy_scores,'Precision
':precision_scores}).sort_values('Precision',ascending=False)

performance_df

```
Algorithm Accuracy
                                 Precision
         RandomForest 0.995385
6
                                  0.998985
5
         DecisionTree 0.986923
                                  0.995918
2
           SVC_Linear 0.989231
                                  0.992922
4
           GaussianNB 0.974615
                                  0.990760
   LogisticRegression 0.981538
0
                                  0.989848
1
                  Knn 0.950000
                                  0.958333
3
              SVC RBF
                       0.936154
                                  0.937259
```

performance_df1 = pd.melt(performance_df, id_vars = "Algorithm")
performance_df1

	Algorithm	variable	value
0	RandomForest	Accuracy	0.995385
1	DecisionTree	Accuracy	0.986923
2	SVC_Linear	Accuracy	0.989231
3	GaussianNB	Accuracy	0.974615
4	LogisticRegression	Accuracy	0.981538
5	Knn	Accuracy	0.950000
6	SVC_RBF	Accuracy	0.936154
7	RandomForest	Precision	0.998985

```
DecisionTree
                           Precision
                                       0.995918
8
9
             SVC_Linear
                           Precision
                                       0.992922
10
             GaussianNB
                           Precision
                                       0.990760
11
    LogisticRegression
                           Precision
                                       0.989848
12
                     Knn
                           Precision
                                       0.958333
13
                SVC_RBF
                           Precision
                                       0.937259
kind='bar',height=5)
plt.ylim(0.5,1.0)
plt.xticks(rotation='vertical')
plt.show()
     1.0
     0.9
     0.8
  value
     0.7
                                                              variable
                                                               Accuracy
                                                               Precision
     0.6
     0.5
           RandomForest
                                       LogisticRegression
                                              Ē
                  DecisionTree
                                GaussianNB
                         SVC_Linear
                                                     SVC RBF
                             Algorithm
```