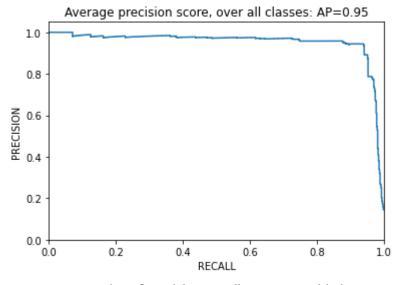
```
#!/usr/bin/env python3
# -*- coding: utf-8 -*-
Created on Mon May 18 11:29:26 2020
@author: yash
import pandas as pd
from sklearn.preprocessing import label_binarize
from sklearn.preprocessing import StandardScaler
from sklearn.linear model import LogisticRegression
from sklearn.multiclass import OneVsRestClassifier
from sklearn.metrics import accuracy_score, confusion_matrix
from itertools import cycle
from sklearn.metrics import precision recall curve
from sklearn.metrics import average_precision_score, classification_report
import matplotlib.pyplot as plt
dataset = pd.read_csv('labelledfeatures.csv')
X = dataset.iloc[:,:-2].values
y = dataset.iloc[:,328].values
y = label_binarize(y, classes=[0, 1, 2, 3, 4, 5, 6])
X_{train} = X[0:5600]
X_{\text{test}} = X[5600:6301]
y train = y[0:5600]
y_{\text{test}} = y[5600:6301]
# Feature Scaling
sc = StandardScaler()
X train = sc.fit transform(X train)
X test = sc.transform(X test)
model = LogisticRegression(solver = 'newton-cg', C = 1, random_state = 42, max_iter = 500)
ovr = OneVsRestClassifier(model)
ovr.fit(X_train, y_train)
y_pred = ovr.predict(X_test)
acc = accuracy_score(y_test, y_pred)
# conf_matrix = confusion_matrix(y_test, y_pred)
print("Accuracy of the model is:")
print(acc)
# print("The confusion matrix is:")
# print(conf_matrix)
print(classification_report(y_test, y_pred, target_names=['BHO', 'CeeInject','FakeRean', 'OnL
y score = ovr.decision function(X test)
```

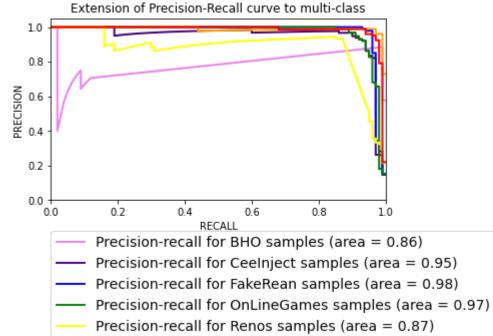
```
n classes = 7
precision = dict()
recall = dict()
average_precision = dict()
for i in range(n classes):
   precision[i], recall[i], _ = precision_recall_curve(y_test[:, i], y_score[:, i])
   average_precision[i] = average_precision_score(y_test[:, i], y_score[:, i])
# A "micro-average": quantifying score on all classes jointly
precision["micro"], recall["micro"], = precision recall curve(y test.ravel(), y score.ravel
average_precision["micro"] = average_precision_score(y_test, y_score, average="micro")
#print('Average precision score over all classes: {0:0.2f}'.format(average precision["micro"]
plt.figure()
plt.step(recall['micro'], precision['micro'], where='post')
plt.xlabel('RECALL')
plt.ylabel('PRECISION')
plt.ylim([0.0, 1.05])
plt.xlim([0.0, 1.0])
plt.title('Average precision score, over all classes: AP={0:0.2f}'.format(average_precision["
plt.show()
# setup plot details
colors = cycle(['violet', 'indigo', 'blue', 'green', 'yellow', 'orange', 'red'])
lines = []
labels = []
for i, color in zip(range(n_classes), colors):
   tgt = ['BHO', 'CeeInject', 'FakeRean', 'OnLineGames', 'Renos', 'Vobfus', 'Winwebsec']
   1, = plt.plot(recall[i], precision[i], color=color, lw=2)
   lines.append(1)
   labels.append('Precision-recall for {0} samples (area = {1:0.2f})'''.format(tgt[i], avera
fig = plt.gcf()
fig.subplots_adjust(bottom=0.25)
plt.xlim([0.0, 1.0])
plt.ylim([0.0, 1.05])
plt.xlabel('RECALL')
plt.ylabel('PRECISION')
plt.title('Extension of Precision-Recall curve to multi-class')
plt.legend(lines, labels, loc=(0, -1.0), prop=dict(size=14))
plt.show()
```

Accuracy of the model is: 0.9257142857142857

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| вно | 0.88 | 0.99 | 0.93 | 100 |
| CeeInject | 0.90 | 0.94 | 0.92 | 100 |
| FakeRean | 0.98 | 0.96 | 0.97 | 100 |
| OnLineGames | 0.92 | 0.93 | 0.93 | 100 |
| Renos | 0.94 | 0.85 | 0.89 | 100 |
| Vobfus | 0.98 | 0.97 | 0.97 | 100 |
| Winwebsec | 0.98 | 0.94 | 0.96 | 100 |
| | | | | |
| micro avg | 0.94 | 0.94 | 0.94 | 700 |
| macro avg | 0.94 | 0.94 | 0.94 | 700 |
| weighted avg | 0.94 | 0.94 | 0.94 | 700 |
| samples avg | 0.93 | 0.94 | 0.94 | 700 |

/usr/local/lib/python3.6/dist-packages/sklearn/metrics/_classification.py:1272: Undefine _warn_prf(average, modifier, msg_start, len(result))





Precision-recall for Vobfus samples (area = 0.99)

riecision-recan for willwebsec samples (area – 0.50)