15Z332 Ex7 - NaiveBayes and SVM

October 17, 2018

1 Exercise 7

1.1 NaiveBayes and SVM Classifiers

Perform classification using NaiveBayes and SVM on Iris dataset. Display the confusion matrix for the various classifiers. Analyze the performance of SVM for different kernel functions. Represent the results of the classification in terms of precision and recall using graphs.

1.1.1 Step 1: Import Iris dataset

1.1.2 Step 2: Convert categorical data into numerical data.

Since the target attribute 'Species' is categorical, we convert it into numerical data using LabelEncoder

```
In [82]: print(df.Species.dtype)

object

In [83]: from sklearn import preprocessing

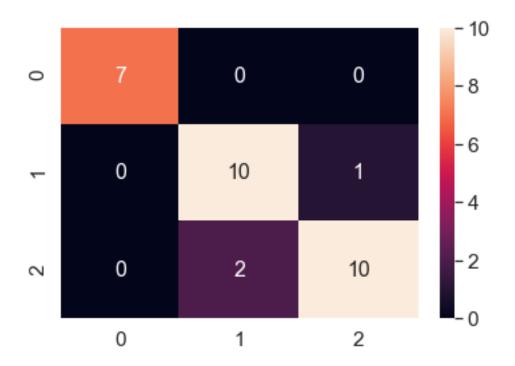
    if df.Species.dtype == 'object':
        lbl = preprocessing.LabelEncoder()
        lbl.fit(list(df.Species.values))
        df.Species = lbl.transform(list(df.Species.values))

        print(df.Species.dtype)

int64
```

1.1.3 Step 3: Split the dataset into train and test data

```
In [84]: from sklearn.model_selection import train_test_split
         Y = df.Species.values
         X = df.drop(['Species'],axis=1).values
         x_train, x_test, y_train, y_test = train_test_split(X,Y, test_size= 0.20, random_state
1.1.4 Step 4: Fit the data into NaiveBayes classifier
In [85]: from sklearn.naive_bayes import GaussianNB
         from sklearn.metrics import accuracy_score
         gnb = GaussianNB()
         NaiveBayesModel = gnb.fit(x_train, y_train)
         y_pred = NaiveBayesModel.predict(x_test)
         accuracy_score(y_test, y_pred)
Out[85]: 0.9
  Draw the confusion matrix
In [86]: from sklearn.metrics import confusion_matrix
         con_mat = confusion_matrix(y_test,y_pred)
         print(con_mat)
[[7 0 0]
 [ 0 10 1]
[ 0 2 10]]
In [87]: import seaborn as sn
         sn.set(font_scale=1.4)
         sn.heatmap(pd.DataFrame(con_mat), annot=True,annot_kws={"size": 16})
Out[87]: <matplotlib.axes._subplots.AxesSubplot at 0x11a8107f0>
```



1.1.5 Step 5: Fit the data into SVM classifier

Since the precision-recall score is 100%, we will try adding noise data and build the SVM model.

```
In [89]: import numpy as np
    # Add noisy features
    random_state = np.random.RandomState(0)
    n_samples, n_features = X.shape
```

```
X = np.c_[X, random_state.randn(n_samples, 200 * n_features)]
x_train, x_test, y_train, y_test = train_test_split(X[Y<2],Y[Y<2], test_size= 0.20)</pre>
```

1) Using Linear kernel

```
In [90]: from sklearn.metrics import precision_recall_curve
    import matplotlib.pyplot as plt

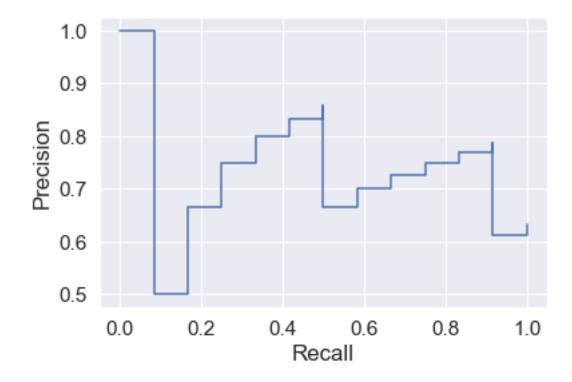
classifier = svm.SVC(kernel='linear', random_state=27)
    classifier.fit(x_train, y_train)
    y_score = classifier.decision_function(x_test)

print('Average precision-recall score:',average_precision_score(y_test, y_score))
    precision, recall, _ = precision_recall_curve(y_test, y_score)

plt.step(recall, precision, color='b')
    plt.xlabel('Recall')
    plt.ylabel('Precision')
```

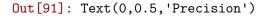
Average precision-recall score: 0.7725782988940885

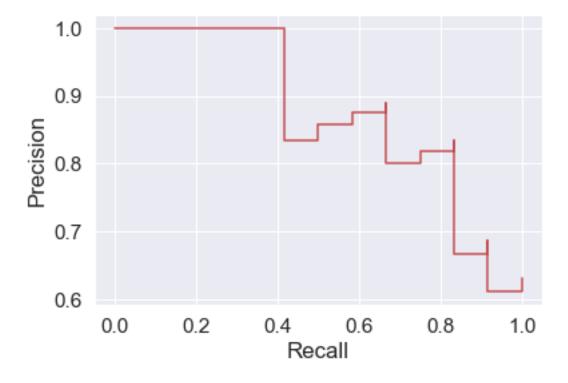
Out[90]: Text(0,0.5,'Precision')



2) Using Polynomial kernel

Average precision-recall score: 0.8826354870762766





3) Using RBF (Radial Basis Function) Kernel

```
print('Average precision-recall score:',average_precision_score(y_test, y_score))
precision, recall, _ = precision_recall_curve(y_test, y_score)

plt.step(recall, precision, color='g')
plt.xlabel('Recall')
plt.ylabel('Precision')
```

Average precision-recall score: 0.7730399230399231

Out[92]: Text(0,0.5,'Precision')

