Chapter 8 - Exercise 3: Wine

Cho dữ liệu wine nằm trong tập tin wine.data.txt

(Xem chi tiết tại: http://archive.ics.uci.edu/ml/datasets/Wine))

Yêu cầu: đọc dữ liệu về, chuẩn hóa dữ liệu (nếu cần) và áp dụng thuật toán SVM để thực hiện việc dự đoán loại rượu dựa trên thông tin được cung cấp

- 1. Tạo X_train, X_test, y_train, y_test từ dữ liệu đọc được với tỷ lệ dữ liệu test là 0.3
- 2. Áp dụng thuật toán SVM
- 3. Tìm kết quả
- 4. Kiểm tra độ chính xác
- 5. Với X new = [[13.71, 5.65, 2.45, 20.5, 95, 1.68, .61, .52, 1.06, 7.7, .64, 1.74, 740]

```
[12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], \\ [13.2, 1.78, 2.14, 11.2, 100, 2.65, 2.76, .26, 1.28, 4.38, 1.05, 3.4, 1050]], thi y_new có kết quả?
```

- 6. So sánh hiệu suất của 4 thuật toán: RandomForestClassifier, SVC, GaussianNB, LogisticRegression
- 7. Trực quan hóa kết quả

```
In [1]: # from google.colab import drive
# drive.mount("/content/gdrive", force_remount=True)
```

In [2]: # %cd '/content/gdrive/My Drive/LDS6_MachineLearning/practice/Chapter8_SVM/'

```
In [3]: import matplotlib.pyplot as plt
    from sklearn import datasets
    from sklearn import svm
    from sklearn.model_selection import train_test_split
    import numpy as np
    import pandas as pd
```

In [4]: import warnings
warnings.filterwarnings("ignore", category=FutureWarning)

```
In [5]:
        data = pd.read csv('wine.data.txt', sep=',', header= None)
         data.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 178 entries, 0 to 177
        Data columns (total 14 columns):
               178 non-null int64
        1
               178 non-null float64
               178 non-null float64
        2
        3
               178 non-null float64
        4
               178 non-null float64
        5
               178 non-null int64
               178 non-null float64
        6
        7
               178 non-null float64
               178 non-null float64
        8
               178 non-null float64
        9
        10
               178 non-null float64
        11
               178 non-null float64
        12
               178 non-null float64
        13
               178 non-null int64
        dtypes: float64(11), int64(3)
        memory usage: 19.6 KB
        # data.head()
In [6]:
In [7]: | X = data.iloc[:, 1:14]
         y = data.iloc[:, 0]
In [8]: X.head()
Out[8]:
               1
                    2
                                   5
                                             7
                                                       9
                                                           10
                         3
                                        6
                                                  8
                                                                11
                                                                     12
                                                                          13
           14.23 1.71 2.43
                           15.6
                                127 2.80
                                          3.06 0.28
                                                    2.29
                                                         5.64
                                                              1.04
                                                                   3.92
                                                                        1065
                                                              1.05
            13.20 1.78 2.14
                            11.2
                                100 2.65 2.76 0.26
                                                   1.28
                                                         4.38
                                                                   3.40
                                                                        1050
            13.16 2.36 2.67 18.6
                                 101
                                     2.80
                                          3.24 0.30 2.81
                                                         5.68
                                                              1.03
                                                                  3.17
                                                                        1185
            14.37 1.95 2.50
                           16.8
                                 113
                                     3.85 3.49 0.24
                                                   2.18
                                                         7.80
                                                              0.86
                                                                   3.45
                                                                        1480
            13.24 2.59 2.87 21.0
                                 118 2.80 2.69 0.39 1.82 4.32
                                                              1.04 2.93
                                                                         735
In [9]: y.head()
Out[9]: 0
              1
              1
        1
        2
              1
        3
              1
        Name: 0, dtype: int64
        from sklearn.model selection import train test split
         X_train, X_test, y_train, y_test = train_test_split(X, y,
                                                                test_size=0.3)
```

```
In [11]: | clf = svm.SVC(kernel='linear')
         clf.fit(X_train, y_train)
Out[11]: SVC(C=1.0, cache size=200, class weight=None, coef0=0.0,
             decision_function_shape='ovr', degree=3, gamma='auto_deprecated',
             kernel='linear', max_iter=-1, probability=False, random_state=None,
             shrinking=True, tol=0.001, verbose=False)
In [12]: | y_pred = clf.predict(X_test)
In [13]: y pred
Out[13]: array([2, 2, 2, 1, 1, 3, 2, 3, 1, 3, 2, 3, 2, 2, 1, 3, 3, 1, 3, 1, 2, 1,
                1, 1, 3, 3, 2, 1, 2, 3, 1, 3, 1, 1, 2, 3, 1, 1, 2, 2, 2, 2, 2, 3,
                3, 1, 2, 2, 2, 1, 1, 1, 2], dtype=int64)
In [14]: from sklearn.metrics import accuracy score
         print("Accuracy is ", accuracy_score(y_test,y_pred)*100,"%")
         Accuracy is 94.444444444444 %
In [15]:
         from sklearn.metrics import classification report, confusion matrix
         print(confusion matrix(y test,y pred))
         print(classification report(y test,y pred))
         [[19 2 0]
          [ 0 19 1]
          [ 0 0 13]]
                       precision
                                    recall f1-score
                                                        support
                    1
                            1.00
                                      0.90
                                                0.95
                                                             21
                    2
                            0.90
                                      0.95
                                                0.93
                                                             20
                    3
                            0.93
                                      1.00
                                                0.96
                                                             13
                                                0.94
                                                             54
             accuracy
            macro avg
                            0.94
                                      0.95
                                                0.95
                                                             54
         weighted avg
                            0.95
                                      0.94
                                                0.94
                                                             54
```

```
In [16]: # Score of Training and Testing data
print("Training R^2 Score", clf.score(X_train, y_train))
print("Testing R^2 Score", clf.score(X_test, y_test))
```

Summary about the model:

- High accuracy: ~0.94
- High precision: ~0.94, High recall: ~0.95
- High training R^2 score and High testing score, nearly the same
- => The good model

```
Chapter8 Ex3 Wine - Jupyter Notebook
In [17]: X new = [[13.71,5.65,2.45,20.5,95,1.68,.61,.52,1.06,7.7,.64,1.74,740],
                   [12.29,1.61,2.21,20.4,103,1.1,1.02,.37,1.46,3.05,.906,1.82,870],
                  [13.2,1.78,2.14,11.2,100,2.65,2.76,.26,1.28,4.38,1.05,3.4,1050]]
         y new = clf.predict(X new)
         y_new
Out[17]: array([3, 2, 1], dtype=int64)
In [18]: # Tính đô chính xác theo: Logistic, Naive Bayes, SVM, KNN
          from sklearn.ensemble import RandomForestClassifier
         from sklearn.svm import SVC
         from sklearn.naive_bayes import GaussianNB
          from sklearn.tree import DecisionTreeClassifier
          from sklearn.model selection import cross val score
         models = [
              RandomForestClassifier(n_estimators=200),
              SVC(kernel='linear'),
             GaussianNB(),
             DecisionTreeClassifier()
          1
         CV = 5
         cv df = pd.DataFrame(index=range(CV * len(models)))
         entries = []
         i=0
         for model in models:
              scores = []
              for j in range(CV):
                  model_name = model.__class__.__name__
                  model.fit(X,y)
                  score = model.score(X,y)
                  scores.append(score)
              print(model.__class__.__name__, scores)
              entries.append([model name, np.array(scores).mean()])
         cv_df = pd.DataFrame(entries, columns=['model_name', 'score_mean'])
         RandomForestClassifier [1.0, 1.0, 1.0, 1.0, 1.0]
         SVC [0.9943820224719101, 0.9943820224719101, 0.9943820224719101, 0.994382022471
         9101, 0.9943820224719101]
         GaussianNB [0.9887640449438202, 0.9887640449438202, 0.9887640449438202, 0.98876
         40449438202, 0.98876404494382021
         DecisionTreeClassifier [1.0, 1.0, 1.0, 1.0, 1.0]
```

In [19]: cv_df

Out[19]:

	model_name	score_mean
0	RandomForestClassifier	1.000000
1	SVC	0.994382
2	GaussianNB	0.988764
3	DecisionTreeClassifier	1.000000

```
In [20]: plt.bar(cv_df['model_name'],cv_df['score_mean'])
    plt.xlabel('model_name')
    plt.ylabel('accuracy')
    plt.xticks(rotation='vertical')
    plt.title("Accuracies of Algorithms")
    plt.show()
```

