## Chapter 8 - Exercise 3: Wine

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

1. Tạo X\_train, X\_test, y\_train, y\_test từ dữ liệu đọc được với tỷ lệ dữ liệu test là 0.3

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

2. Áp dụng thuật toán SVM

4. Kiểm tra độ chính xác

3. Tìm kết quả

In [ ]: X.head()

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

```
5. V\acute{\sigma}i 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], [12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], [12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], [12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], [12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], [12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], [12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], [12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], [12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], [12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], [12.29, 1.61, 2.21, 20.4, 103, 1.1, 1.02, .37, 1.46, 3.05, .906, 1.82, 870], [12.29, 1.61, 2.21, 20.4, 103, 2.21, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 20.4, 
                             [13.2,1.78,2.14,11.2,100,2.65,2.76,.26,1.28,4.38,1.05,3.4,1050]], thì y_new có kết quả?
                        6. So sánh hiệu suất của 5 thuật toán: Logistic, Naive Bayes, SVM, RandomForestClassifier, DecisionTreeClassifier
                        7. Trực quan hóa kết quả
In [ ]: from google.colab import drive
                   drive.mount("/content/gdrive", force_remount=True)
                   Mounted at /content/gdrive
                 %cd '/content/gdrive/My Drive/LDS6_MachineLearning/practice_2020/Chapter8_SVM/'
                   /content/gdrive/My Drive/LDS6_MachineLearning/practice_2020/Chapter8_SVM
                   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
                  import warnings
                   warnings.filterwarnings("ignore", category=FutureWarning)
                   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):
                               Column Non-Null Count Dtype
                                                  178 non-null
                                                                                       int64
                                                  178 non-null
                                                                                      float64
                                                  178 non-null
                                                                                      float64
                                                  178 non-null
                                                                                       float64
                                                  178 non-null
                                                                                       float64
                                                  178 non-null
                                                                                       int64
                                                  178 non-null
                                                                                       float64
                                                  178 non-null
                                                                                       float64
                                                  178 non-null
                                                                                       float64
                                                  178 non-null
                                                                                       float64
                     10
                               10
                                                  178 non-null
                                                                                       float64
                     11 11
                                                 178 non-null
                                                                                       float64
                     12 12
                                                 178 non-null
                                                                                       float64
                     13 13
                                                  178 non-null
                                                                                       int64
                   dtypes: float64(11), int64(3)
                   memory usage: 19.6 KB
In [ ]: # data.head()
In [ ]: X = data.iloc[:, 1:14]
                  y = data.iloc[:, 0]
```

```
0 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 13.20 1.78 2.14 11.2 100 2.65 2.76 0.26 1.28 4.38 1.05 3.40 1050
         2 13.16 2.36 2.67 18.6 101 2.80 3.24 0.30 2.81 5.68 1.03 3.17 1185
        3 14.37 1.95 2.50 16.8 113 3.85 3.49 0.24 2.18 7.80 0.86 3.45 1480
         4 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 [ ]: y.head()
Out[]: 0
        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 [ ]: clf = svm.SVC(kernel='linear')
        clf.fit(X_train, y_train)
Out[]: SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
             decision_function_shape='ovr', degree=3, gamma='scale', kernel='linear',
             max_iter=-1, probability=False, random_state=None, shrinking=True,
             tol=0.001, verbose=False)
In [ ]: y_pred = clf.predict(X_test)
In [ ]: y_pred
Out[]: array([2, 2, 2, 3, 1, 1, 3, 2, 2, 3, 1, 2, 2, 2, 2, 1, 3, 2, 3, 1, 1, 3,
               3, 2, 3, 2, 2, 1, 3, 2, 1, 1, 3, 2, 2, 2, 1, 1, 1, 3, 1, 1, 2, 1,
                2, 1, 2, 2, 2, 2, 1, 2, 3, 1])
In [ ]: from sklearn.metrics import accuracy_score
        print("Accuracy is ", accuracy_score(y_test,y_pred)*100,"%")
        Accuracy is 98.14814814814815 %
        from sklearn.metrics import classification_report, confusion_matrix
        print(confusion_matrix(y_test,y_pred))
         print(classification_report(y_test,y_pred))
         [[18 0 0]
           0 23 0]
          [ 0 1 12]]
                                    recall f1-score
                       precision
                                                        support
                            1.00
                                                             18
                                      1.00
                                                1.00
                                                             23
                            0.96
                                      1.00
                                                0.98
                                                             13
                                      0.92
                            1.00
                                                 0.96
                                                 0.98
                                                             54
             accuracy
                                      0.97
                                                             54
                            0.99
                                                0.98
           macro avg
        weighted avg
                                                             54
                            0.98
                                      0.98
                                                0.98
In [ ]: # 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))
        Training R^2 Score 1.0
        Testing R^2 Score 0.9814814814814815
        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

In []: X_{\text{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[]: array([3, 2, 1])
```

2 3 4 5 6 7 8 9 10 11 12 13

Out[]:

```
In []: # Tính độ chính xác mdoel theo:
        # Logistic, Naive Bayes, SVM, RandomForestClassifier, DecisionTreeClassifier
        # Khi dùng KNN thì cần chọn k phù hợp trước
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.svm import SVC
        from sklearn.naive_bayes import GaussianNB
        from sklearn.tree import DecisionTreeClassifier
        from sklearn.linear_model import LogisticRegression
        import datetime
        from sklearn.model_selection import cross_val_score
        models = [
            LogisticRegression(max_iter=200),
            GaussianNB(),
            SVC(kernel='linear'),
            RandomForestClassifier(n_estimators=200),
            DecisionTreeClassifier()
        CV = 10
        entries = []
        i=0
        for model in models:
            scores_train = []
            scores_test = []
            times = []
            abs_scores = []
            for j in range(CV):
                X_train, X_test, y_train, y_test = train_test_split(X, y,
                                                             test_size=0.3)
                t1 = datetime.datetime.now()
                model_name = model.__class__._name__
                model.fit(X_train,y_train)
                t2 = datetime.datetime.now()
                d = round((t2 - t1).microseconds/1000,1) # => miliseconds
                score_train = model.score(X_train,y_train)
                score_test = model.score(X_test,y_test)
                 abs_score = abs(score_train - score_test)
                 scores_train.append(score_train)
                 scores_test.append(score_test)
                 abs_scores.append(abs_score)
                times.append(d)
            print(model.__class__._name__, scores_test)
            entries.append([model_name, np.array(scores_train).mean(),
                            np.array(scores_test).mean(), np.array(abs_scores).mean(),
                            np.array(times).mean()])
            i += 1
        cv_df = pd.DataFrame(entries,
                              columns=['model_name', 'score_train_mean',
                                       'score_test_mean', 'abs|score|', 'time_mean'])
```

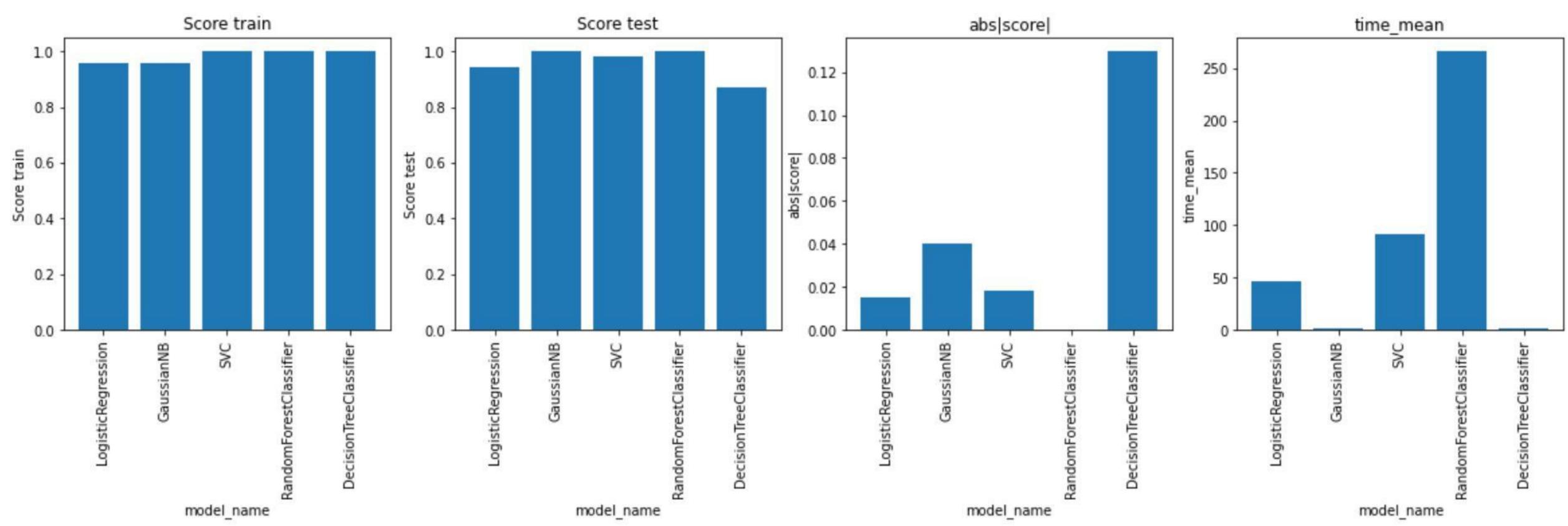
In [ ]: cv\_df

Out[ ]: model\_name score\_train\_mean score\_test\_mean abs|score| time\_mean 0 LogisticRegression 0.959677 0.944444 0.015233 46.96 GaussianNB 1.000000 0.040323 1.44 0.959677 2 SVC 1.000000 0.018519 91.18 0.981481 3 RandomForestClassifier 1.000000 265.96 1.000000 0.000000 1.24 DecisionTreeClassifier 1.000000 0.870370 0.129630

```
plt.figure(figsize=(20, 4))
plt.subplot(1, 4, 1)
plt.bar(cv_df['model_name'],cv_df['score_train_mean'])
plt.xlabel('model_name')
plt.ylabel('Score train')
plt.xticks(rotation='vertical')
plt.title("Score train")
plt.subplot(1, 4, 2)
plt.bar(cv_df['model_name'],cv_df['score_test_mean'])
plt.xlabel('model_name')
plt.ylabel('Score test')
plt.xticks(rotation='vertical')
plt.title("Score test")
plt.subplot(1, 4, 3)
plt.bar(cv_df['model_name'],cv_df['abs|score|'])
plt.xlabel('model_name')
plt.ylabel('abs | score | ')
plt.xticks(rotation='vertical')
plt.title("abs|score|")
```

```
plt.subplot(1, 4, 4)
plt.bar(cv_df['model_name'],cv_df['time_mean'])
plt.xlabel('model_name')
plt.ylabel('time_mean')
plt.xticks(rotation='vertical')
plt.title("time_mean")

plt.show()
```



In []: # nhan xet
 # chon model nao?
# build model do tu A\_Z