Chapter 5 - Ex1: Social Network Ads

Cho dữ liệu Social_Network_Ads.csv

Sử dụng thuật toán KNN để dự đoán khách hàng mua (1) hay không mua sản phẩm (0) dựa trên các thông tin được cung cấp:

- 1. Đọc dữ liệu và gán cho biến data. Tiền xử lý dữ liệu (nếu cần)
- 2. Tạo inputs data với các cột trừ cột Purchased, và outputs data với 1 cột là Purchased
- 3. Từ inputs data và outputs data => Tạo X_train, X_test, y_train, y_test với tỷ lệ 70-30
- 4. Thực hiện KNN với X train, y train
- 5. Dự đoán y từ X test => so sánh với y test
- 6. Đánh giá mô hình => Nhận xét
- 7. Ghi mô hình (nếu mô hình tốt sau khi đánh giá)

```
In [39]:
         # link tham khảo: https://towardsdatascience.com/knn-in-python-835643e2fb53
 In [1]:
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         from sklearn.model selection import train test split
 In [2]:
         # import some data to play with
         data = pd.read csv("Social Network Ads.csv")
         data.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 400 entries, 0 to 399
         Data columns (total 5 columns):
         User ID
                            400 non-null int64
         Gender
                            400 non-null object
         Age
                            400 non-null int64
                            400 non-null int64
         EstimatedSalary
         Purchased
                            400 non-null int64
         dtypes: int64(4), object(1)
         memory usage: 15.8+ KB
 In [3]: data.shape
 Out[3]: (400, 5)
```

```
In [4]: data.head()
```

Out[4]:

	User ID	Gender	Age	EstimatedSalary	Purchased
0	15624510	Male	19	19000	0
1	15810944	Male	35	20000	0
2	15668575	Female	26	43000	0
3	15603246	Female	27	57000	0
4	15804002	Male	19	76000	0

```
In [5]: # data.tail()
In [6]: # thống kê số lượng các lớp
         data.groupby('Purchased').count()["Gender"]
Out[6]: Purchased
              257
         1
              143
         Name: Gender, dtype: int64
         from sklearn.preprocessing import LabelEncoder
In [7]:
In [8]:
         le = LabelEncoder()
         data['Gender_E'] = le.fit_transform(data.Gender)
         data.head()
In [9]:
Out[9]:
              User ID Gender Age EstimatedSalary Purchased Gender_E
         0 15624510
                                          19000
                                                        0
                                                                  1
                        Male
                              19
          1 15810944
                        Male
                              35
                                          20000
                                                                  1
            15668575 Female
                                          43000
                                                        0
                                                                  0
                              26
            15603246
                                          57000
                                                                  0
                     Female
                              27
            15804002
                        Male
                              19
                                          76000
                                                        0
                                                                  1
```

```
In [10]: # The columns that we will be making predictions with.
inputs = data.iloc[:,[2,3,5]]
inputs.shape
```

Out[10]: (400, 3)

```
In [11]: inputs.head()
```

Out[11]:

	Age	EstimatedSalary	Gender_E
0	19	19000	1
1	35	20000	1
2	26	43000	0
3	27	57000	0
4	19	76000	1

```
In [12]: # The column that we want to predict.
    outputs = data['Purchased']
    #outputs = np.array(outputs)
    outputs.shape
```

Out[12]: (400,)

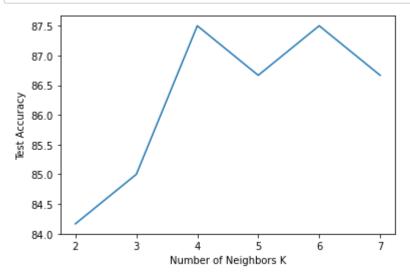
```
In [14]: # Can phải Scale dữ liệu vì Age và EstimatedSalary có thang đo khác nhau nhiều
    from sklearn.preprocessing import StandardScaler
    sc = StandardScaler()
    X_train = sc.fit_transform(X_train)
    X_test = sc.transform(X_test)
```

```
In [15]: from sklearn.neighbors import KNeighborsClassifier
    from sklearn.metrics import accuracy_score
    list_k = []
    list_acc = []
    for K_value in range(2,int((y_train.shape[0]**0.5)/2)):
    #for K_value in range(2,10):
        list_k.append(K_value)
        neigh = KNeighborsClassifier(n_neighbors = K_value)
        neigh.fit(X_train, y_train)
        y_pred = neigh.predict(X_test)
        acc = accuracy_score(y_test,y_pred)*100
        list_acc.append(acc)
        print("k = ", K_value,": Accuracy is ", accuracy_score(y_test,y_pred))
```

```
In [16]: vi_tri = list_acc.index(max(list_acc))
k = list_k[vi_tri]
print("The optimal number of neighbors is", k,"with", list_acc[vi_tri])
```

The optimal number of neighbors is 4 with 87.5

```
In [17]: plt.plot(list_k, list_acc)
    plt.xlabel('Number of Neighbors K')
    plt.ylabel('Test Accuracy')
    plt.show()
```



```
In [18]: from sklearn.neighbors import KNeighborsClassifier
```

```
k= 4 : The Train prediction accuracy is: 93.57142857142857 % ---- The Test prediction accuracy is: 87.5 % k= 6 : The Train prediction accuracy is: 93.21428571428572 % ---- The Test prediction accuracy is: 87.5 %
```

```
In [20]: knn = KNeighborsClassifier(n_neighbors=6)
knn.fit(X_train, y_train)
```

```
In [21]: # Kiểm tra độ chính xác
         print("The Train prediction accuracy is: ",
               knn.score(X_train,y_train)*100,"%")
         print("The Test prediction accuracy is: ",
               knn.score(X_test,y_test)*100,"%")
         The Train prediction accuracy is: 93.21428571428572 %
         The Test prediction accuracy is: 87.5 %
In [22]: y pred = knn.predict(X test)
         # y_pred
In [23]: # Đánh giá model
         from sklearn.metrics import confusion_matrix, classification_report
In [24]: | confusion_matrix(y_test, y_pred)
Out[24]: array([[61, 11],
                [ 4, 44]], dtype=int64)
In [25]:
         print(classification_report(y_test, y_pred))
                       precision
                                     recall f1-score
                                                        support
                    0
                             0.94
                                       0.85
                                                 0.89
                                                             72
                    1
                                       0.92
                             0.80
                                                 0.85
                                                             48
                                                 0.88
                                                            120
             accuracy
                             0.87
                                                 0.87
            macro avg
                                       0.88
                                                            120
         weighted avg
                             0.88
                                       0.88
                                                 0.88
                                                            120
```

Quan sát kết quả và đánh giá:

- Có sự chênh lệch giữa train r_score và test r_score ~7% => có thể tạm chấp nhận
- Mô hình có độ chính xác khá cao: 87.5%

Feature Selection

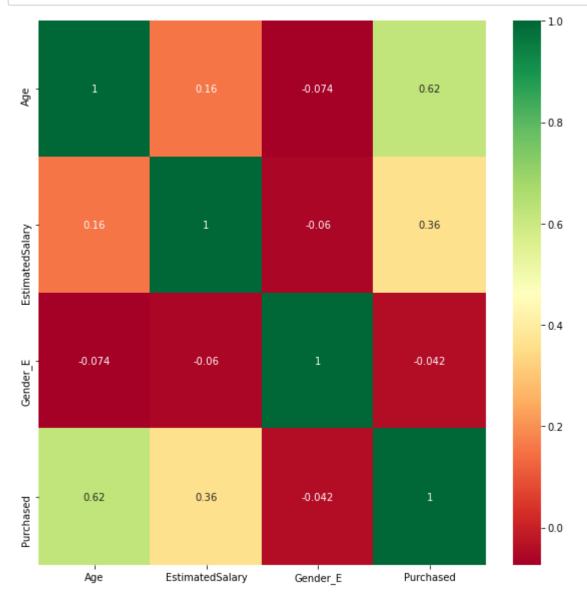
```
In [26]: #get correlations of each features in dataset
    data_sub = data.iloc[:,[2,3,5,4]]
    corrmat = data_sub.corr()
    top_corr_features = corrmat.index
```

In [27]: data_sub.corr()

Out[27]:

	Age	EstimatedSalary	Gender_E	Purchased
Age	1.000000	0.155238	-0.073741	0.622454
EstimatedSalary	0.155238	1.000000	-0.060435	0.362083
Gender_E	-0.073741	-0.060435	1.000000	-0.042469
Purchased	0.622454	0.362083	-0.042469	1.000000

```
In [28]: import seaborn as sns
    plt.figure(figsize=(10,10))
    #plot heat map
    g=sns.heatmap(data[top_corr_features].corr(), cmap="RdYlGn", annot=True)
# annot=True: néu muốn in cả giá trị
```



In [29]: | # => Select Age, EstimatedSalary

```
In [30]: from sklearn.feature selection import SelectKBest
         from sklearn.feature selection import chi2
In [31]: #apply SelectKBest class to extract all best features
         bestfeatures = SelectKBest(score func=chi2, k='all')
         fit = bestfeatures.fit(inputs,outputs)
         dfscores = pd.DataFrame(fit.scores )
         dfcolumns = pd.DataFrame(inputs.columns)
In [32]: #concat two dataframes for better visualization
         featureScores = pd.concat([dfcolumns,dfscores],axis=1)
         featureScores.columns = ['Specs','Score'] #naming the dataframe columns
         print(featureScores.nlargest(4,'Score')) #print 4 best features
                       Specs
                                      Score
            EstimatedSalary 872013.169231
                         Age
                                 451.155226
         2
                   Gender E
                                   0.367946
In [33]: # => select features EstimatedSalary & Age => KNN
In [34]:
         inputs_new = data.iloc[:,[2,3]]
         inputs new.head()
Out[34]:
             Age EstimatedSalary
          0
              19
                         19000
              35
                         20000
          1
          2
              26
                         43000
          3
              27
                         57000
              19
                         76000
In [35]: X_train_n, X_test_n, y_train_n, y_test_n = train_test_split(inputs_new,
                                                                       outputs,
                                                               test_size=0.30,
                                                               random_state=1)
In [36]: | sc = StandardScaler()
```

X train n = sc.fit transform(X train n)

X test n = sc.transform(X test n)

```
In [37]: list k = []
        list acc = []
        for K_value in range(2,int((y_train.shape[0]**0.5)/2)):
        #for K value in range(2,10):
           list_k.append(K_value)
           neigh = KNeighborsClassifier(n_neighbors = K_value)
           neigh.fit(X_train_n, y_train_n)
           y_pred = neigh.predict(X_test_n)
           acc = accuracy_score(y_test_n,y_pred)*100
           list_acc.append(acc)
           print("k = ", K_value,": Accuracy is ", accuracy_score(y_test_n,y_pred))
        k = 2 : Accuracy is 0.8416666666666667
        k = 3: Accuracy is 0.875
        k = 5: Accuracy is 0.875
        k = 6: Accuracy is 0.875
        In [38]: # Với k=4 có độ chính xác cao hơn khoảng 1%
```