## **Chapter4 - Exercise 1: Click Prediction**

Cho dữ liệu Click\_prediction\_small (link: <a href="https://www.openml.org/d/1220">https://www.openml.org/d/1220</a> (<a href="https://www.openml.org/d/1220">https://www.openml.org/d/1220</a>) hoặc sklearn.datasets import fetch\_openml với data\_id = 1220).

Yêu cầu: Hãy đọc dữ liệu từ tập tin này, áp dụng Naive Bayes để thực hiện việc xác định người dùng có clịck vào một quảng cáo hay không dựa trên các thông tin được cung cấp.

## Yêu cầu:

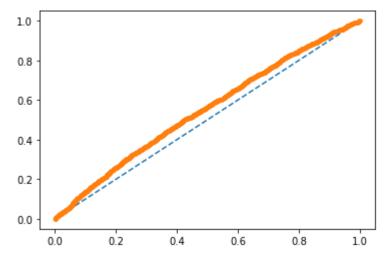
- 1. Hãy chuẩn hóa dữ liệu cho phù hợp
- 2. Áp dụng Naive Bayes. Nhận xét kết quả.
- 3. Model trên có phù hợp không? Nếu không thì đưa ra phương án khác.

```
import pandas as pd
In [1]:
        from sklearn.datasets import fetch openml
In [2]: data = fetch openml(data id=1220)
        X = pd.DataFrame(
            data['data'],
            columns=data['feature_names']
        ).astype(float)
        y = pd.Series(data['target']).astype(int)
In [3]: X.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 39948 entries, 0 to 39947
        Data columns (total 9 columns):
        impression
                          39948 non-null float64
        ad id
                          39948 non-null float64
        advertiser_id 39948 non-null float64
                          39948 non-null float64
        depth
        position
                          39948 non-null float64
                    39948 non-null float64
        keyword_id
        title id
                          39948 non-null float64
        description_id 39948 non-null float64
                          39948 non-null float64
        user id
        dtypes: float64(9)
        memory usage: 2.7 MB
In [4]: \#X = X.drop('user id', axis=1)
```

```
In [5]:
          X.head()
 Out[5]:
                             ad_id advertiser_id depth position
              impression
                                                              keyword_id
                                                                          title_id description_id
           0
                    1.0
                         8343295.0
                                        11700.0
                                                  3.0
                                                          3.0
                                                                  21264.0 27892.0
                                                                                        1559.0
           1
                    1.0 20017077.0
                                        23798.0
                                                  1.0
                                                          1.0
                                                                  35498.0
                                                                             4.0
                                                                                       36476.0
                                                                                                 56
           2
                    1.0 21348354.0
                                        36654.0
                                                  1.0
                                                          1.0
                                                                  19975.0 36105.0
                                                                                       33292.0 116
           3
                    1.0 20366086.0
                                        33280.0
                                                  3.0
                                                          3.0
                                                                   5942.0
                                                                          4057.0
                                                                                        4390.0
                                                                                                877
                         6803526.0
                                        10790.0
                                                  2.0
                                                                                        1679.0 121
                    1.0
                                                          1.0
                                                                  60593.0 25242.0
 In [6]: y.head()
 Out[6]: 0
               0
          1
               1
          2
               0
          3
               0
               0
          4
          dtype: int32
 In [7]: #target count: 1: click, 0: not click
          y.value_counts()
 Out[7]: 0
               33220
                6728
          dtype: int64
 In [8]:
          from sklearn.model selection import train test split
          X_train, X_test, y_train, y_test = train_test_split(X, y,
                                                                   test_size=0.25,
                                                                   random state=42)
 In [9]:
          from sklearn.naive_bayes import GaussianNB
          import numpy as np
          from sklearn.utils.validation import column or 1d
In [10]:
          #Create a Gaussian Classifier
          model = GaussianNB()
          # Train the model using the training sets
          model.fit(X_train, y_train)
Out[10]: GaussianNB(priors=None, var smoothing=1e-09)
In [11]: model.score(X_train, y_train)
Out[11]: 0.8086512466206068
In [12]: model.score(X_test, y_test)
Out[12]: 0.8078502052668469
```

```
In [13]: y pred = model.predict(X test)
         from sklearn.metrics import accuracy score
In [14]:
         print("Accuracy is ", accuracy_score(y_test,y_pred)*100,"%")
         Accuracy is 80.78502052668469 %
In [15]: from sklearn.metrics import confusion matrix
In [16]: confusion_matrix(y_test, y_pred)
Out[16]: array([[7992,
                        313],
                [1606,
                         76]], dtype=int64)
         # Đánh giá model
In [17]:
         from sklearn. metrics import classification_report, roc_auc_score, roc_curve
In [18]: | print(classification report(y test, y pred))
                        precision
                                     recall f1-score
                                                        support
                    0
                             0.83
                                       0.96
                                                 0.89
                                                           8305
                    1
                             0.20
                                       0.05
                                                 0.07
                                                           1682
                                                 0.81
                                                           9987
             accuracy
                             0.51
                                       0.50
                                                 0.48
                                                           9987
            macro avg
         weighted avg
                             0.73
                                       0.81
                                                 0.75
                                                           9987
In [19]: y_prob = model.predict_proba(X_test)
         y_prob
Out[19]: array([[0.90958662, 0.09041338],
                [0.8898664 , 0.1101336 ],
                [0.85271593, 0.14728407],
                [0.88644901, 0.11355099],
                [0.9088923 , 0.0911077 ],
                [0.76539178, 0.23460822]])
In [20]: roc_auc_score(y_test, y_prob[:, 1])
Out[20]: 0.549416529875775
         import matplotlib.pyplot as plt
In [21]:
```

```
In [22]: # calculate roc curve
fpr, tpr, thresholds = roc_curve(y_test, y_prob[:, 1])
# plot no skill
plt.plot([0, 1], [0, 1], linestyle='--')
plt.plot(fpr, tpr, marker='.')
plt.show()
```



```
In [1]: # Dựa trên kết quả trên ta thấy precision và recall của class 1 đều rất thấp # ROC thấp # Dựa trên tất cả các đánh giá # => model chưa phù hợp có thể do mất cân bằng dữ liệu ???
```

## Resampling data

```
In [24]: from imblearn.under_sampling import RandomUnderSampler
from imblearn.over_sampling import SMOTE
```

Using TensorFlow backend.

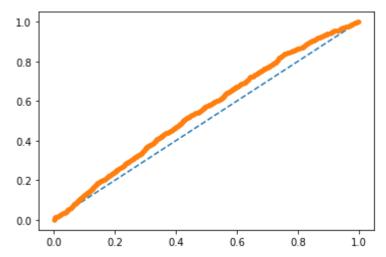
```
In [25]: rus = RandomUnderSampler()
X_resampled, y_resampled = rus.fit_resample(X, y)
```

```
In [26]: unique_elements, counts_elements = np.unique(y_resampled, return_counts=True)
    print("Frequency of each category:")
    print(np.asarray((unique_elements, counts_elements)))
```

```
Frequency of each category: [[ 0 1] [6728 6728]]
```

```
In [27]: | X_train_r, X_test_r, y_train_r, y_test_r = train_test_split(X_resampled,
                                                                       y_resampled,
                                                               test size=0.25,
                                                               random state=42)
In [28]: #Create a Gaussian Classifier
         model r = GaussianNB()
         # Train the model using the training sets
         model_r.fit(X_train_r, y_train_r)
Out[28]: GaussianNB(priors=None, var smoothing=1e-09)
In [29]: y pred r = model.predict(X test r)
In [30]: confusion_matrix(y_test_r, y_pred_r)
Out[30]: array([[1633,
                          68],
                 [1596,
                          67]], dtype=int64)
In [31]: | print(classification_report(y_test_r, y_pred_r))
                        precision
                                     recall f1-score
                                                        support
                    0
                             0.51
                                       0.96
                                                 0.66
                                                            1701
                    1
                             0.50
                                       0.04
                                                 0.07
                                                            1663
                                                 0.51
                                                            3364
             accuracy
            macro avg
                             0.50
                                       0.50
                                                 0.37
                                                            3364
         weighted avg
                             0.50
                                       0.51
                                                 0.37
                                                            3364
         y prob r = model.predict proba(X test r)
In [32]:
         y_prob_r
Out[32]: array([[9.14505765e-01, 8.54942351e-02],
                 [9.10578135e-01, 8.94218651e-02],
                 [9.50145867e-09, 9.99999990e-01],
                 [9.16230159e-01, 8.37698413e-02],
                 [1.64622602e-02, 9.83537740e-01],
                 [8.93668515e-01, 1.06331485e-01]])
In [33]: roc auc score(y test r, y prob r[:, 1])
Out[33]: 0.5521173035705006
```

```
In [34]: # calculate roc curve
fpr, tpr, thresholds = roc_curve(y_test_r, y_prob_r[:, 1])
# plot no skill
plt.plot([0, 1], [0, 1], linestyle='--')
plt.plot(fpr, tpr, marker='.')
plt.show()
```



```
In [35]: # Kết quả với resampling cũng không tốt hơn
# => thay đổi thuật toán khác???
# Thống kê về các giải pháp và kết quả: https://www.openml.org/t/7295

In [36]: # import pickle
# # Save to file in the current working directory
# pkl_filename = "pickle_model.pkl"
# with open(pkl_filename, 'wb') as file:
# pickle.dump(model, file)
In [37]: # with open(pkl_filename, 'rb') as file:
# pickle_model = pickle.load(file)
```