Ex 1: Click Prediction

Cho dữ liệu Click_prediction_small (link: https://www.openml.org/d/1220 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 click 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.

```
In [ ]: import pandas as pd
        from sklearn.datasets import fetch_openml
In [ ]: 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 [ ]: 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
                           39948 non-null float64
        position
                           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 [ ]: #X = X.drop('user_id', axis=1)
In [ ]: X.head()
                          ad_id advertiser_id depth position keyword_id title_id description_id
Out[]:
           impression
                                                                                              user_id
                      8343295.0
                                     11700.0
                                               3.0
                                                       3.0
                                                               21264.0 27892.0
                                                                                                  0.0
        0
                 1.0
                                                                                    1559.0
                  1.0 20017077.0
                                     23798.0
                                               1.0
                                                       1.0
                                                               35498.0
                                                                          4.0
                                                                                    36476.0
                                                                                             562934.0
         2
                                                       1.0
                  1.0 21348354.0
                                               1.0
                                                               19975.0 36105.0
                                                                                    33292.0 11621116.0
                                     36654.0
                                                                                            8778348.0
        3
                  1.0 20366086.0
                                               3.0
                                                       3.0
                                                                       4057.0
                                     33280.0
                                                                5942.0
                                                                                    4390.0
                      6803526.0
                                               2.0
                                                       1.0
                                                               60593.0 25242.0
         4
                  1.0
                                     10790.0
                                                                                     1679.0 12118311.0
In [ ]: y.head()
Out[]: 0
        dtype: int32
        #target count: 1: click, 0: not click
        y.value_counts()
Out[]: 0
              33220
               6728
        dtype: int64
In [ ]: 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 [ ]: from sklearn.naive_bayes import GaussianNB
        import numpy as np
        from sklearn.utils.validation import column_or_1d
In [ ]: #Create a Gaussian Classifier
        model = GaussianNB()
        # Train the model using the training sets
        model.fit(X_train, y_train)
Out[]: GaussianNB(priors=None, var_smoothing=1e-09)
In [ ]: model.score(X_train, y_train)
Out[]: 0.8086512466206068
       model.score(X_test, y_test)
Out[]: 0.8078502052668469
       y_pred = model.predict(X_test)
In [ ]: from sklearn.metrics import accuracy_score
        print("Accuracy is ", accuracy_score(y_test,y_pred)*100,"%")
        Accuracy is 80.78502052668469 %
       from sklearn.metrics import confusion_matrix
        confusion_matrix(y_test, y_pred)
Out[]: array([[7992, 313],
                       76]], dtype=int64)
               [1606,
In []: # Đánh giá model
        from sklearn. metrics import classification_report, roc_auc_score, roc_curve
        print(classification_report(y_test, y_pred))
                                   recall f1-score
                      precision
                                                     support
                                              0.89
                                                        8305
                                     0.96
                           0.83
                                                        1682
                           0.20
                                     0.05
                                               0.07
                                               0.81
                                                        9987
            accuracy
                                     0.50
                           0.51
                                               0.48
                                                        9987
           macro avg
                                                        9987
        weighted avg
                                               0.75
                           0.73
                                     0.81
In [ ]: y_prob = model.predict_proba(X_test)
        y_prob
Out[]: 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 [ ]: roc_auc_score(y_test, y_prob[:, 1])
Out[]: 0.549416529875775
In [ ]: import matplotlib.pyplot as plt
In [ ]: # 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()
```

```
1.0 - 0.8 - 0.6 - 0.4 - 0.6 - 0.8 - 1.0
```

```
In []: # 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

Out[]: 0.5521173035705006

```
In [ ]: from imblearn.under_sampling import RandomUnderSampler
        from imblearn.over_sampling import SMOTE
        Using TensorFlow backend.
In [ ]: rus = RandomUnderSampler()
        X_resampled, y_resampled = rus.fit_resample(X, y)
In [ ]: 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:
         [6728 6728]]
In [ ]: 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 [ ]: #Create a Gaussian Classifier
        model_r = GaussianNB()
        # Train the model using the training sets
        model_r.fit(X_train_r, y_train_r)
Out[]: GaussianNB(priors=None, var_smoothing=1e-09)
In [ ]: y_pred_r = model.predict(X_test_r)
In [ ]: confusion_matrix(y_test_r, y_pred_r)
Out[]: array([[1633,
                        68],
               [1596, 67]], dtype=int64)
        print(classification_report(y_test_r, y_pred_r))
                                   recall f1-score
                      precision
                                                      support
                                     0.96
                                               0.66
                           0.51
                                                         1701
                           0.50
                                     0.04
                                               0.07
                                                         1663
                                               0.51
                                                         3364
            accuracy
                                     0.50
                           0.50
                                               0.37
                                                         3364
           macro avg
        weighted avg
                           0.50
                                     0.51
                                               0.37
                                                         3364
In [ ]: y_prob_r = model.predict_proba(X_test_r)
        y_prob_r
Out[]: 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]])
        roc_auc_score(y_test_r, y_prob_r[:, 1])
```

```
In []: # 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 [ ]: # 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 []: # 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)
```