Credit Risk

January 4, 2023

Prepare dataset

```
[]: import pandas as pd
     import numpy as np
     df = pd.read_csv("data/UCI_Credit_Card.csv")
     df["y"] = df["default.payment.next.month"]
     df.drop("default.payment.next.month", axis=1, inplace=True)
     df.drop("ID", axis=1, inplace=True)
     df.head()
                                                    PAY_0
[]:
        LIMIT_BAL
                   SEX
                        EDUCATION
                                    MARRIAGE
                                               AGE
                                                           PAY_2
                                                                   PAY_3 PAY_4
                                                                                 \
          20000.0
                                                24
                                                                2
                      2
                                 2
                                            1
                                                         2
                                                                       -1
                                                                              -1
         120000.0
                      2
                                 2
                                            2
                                                                2
     1
                                                26
                                                        -1
                                                                       0
                                                                               0
     2
          90000.0
                      2
                                 2
                                            2
                                                                0
                                                                               0
                                                34
                                                         0
                                                                       0
                                 2
     3
          50000.0
                      2
                                                37
                                                        0
                                                                0
                                                                       0
                                                                               0
     4
          50000.0
                                 2
                                                57
                                                        -1
                                                                       -1
                                                                               0
                  BILL_AMT4
                              BILL_AMT5
                                          BILL_AMT6
                                                     PAY_AMT1 PAY_AMT2
                                                                          PAY_AMT3
        PAY_5
     0
           -2
                         0.0
                                    0.0
                                                0.0
                                                           0.0
                                                                   689.0
                                                                                0.0
     1
            0
                      3272.0
                                 3455.0
                                             3261.0
                                                           0.0
                                                                  1000.0
                                                                             1000.0
     2
            0
                     14331.0
                                14948.0
                                            15549.0
                                                        1518.0
                                                                  1500.0
                                                                             1000.0
     3
            0
                     28314.0
                                28959.0
                                            29547.0
                                                        2000.0
                                                                  2019.0
                                                                             1200.0
                     20940.0
                                            19131.0
                                                        2000.0
                                                                 36681.0
            0
                                19146.0
                                                                            10000.0
        PAY_AMT4
                  PAY_AMT5 PAY_AMT6 y
     0
             0.0
                        0.0
                                  0.0 1
          1000.0
     1
                        0.0
                               2000.0 1
     2
          1000.0
                     1000.0
                               5000.0 0
     3
          1100.0
                     1069.0
                               1000.0 0
     4
          9000.0
                      689.0
                                679.0 0
```

[5 rows x 24 columns]

Calculate imbalance in dataset

```
[]: df["y"].value_counts()[0] / len(df["y"])
```

[]: 0.7788

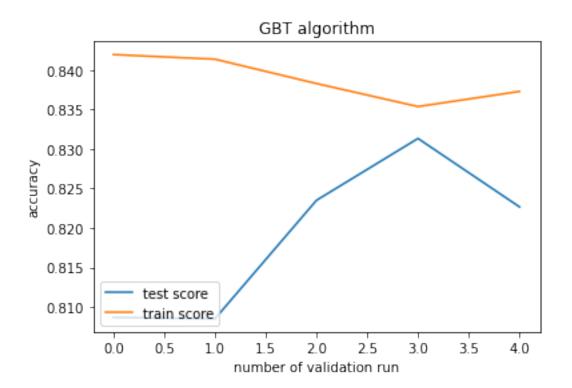
Split data into test and validation sets

GBT Algorithm

	precision	recall	f1-score	support
	_			
0	0.84	0.95	0.89	7060
1	0.68	0.35	0.46	1940
accuracy			0.82	9000
macro avg	0.76	0.65	0.68	9000
weighted avg	0.81	0.82	0.80	9000

```
[]: from sklearn.model_selection import cross_validate
import matplotlib.pylab as plt

cv = cross_validate(model_gbt, X, y, return_train_score=True, cv=5)
plt.plot(cv["test_score"], label="test score")
plt.plot(cv["train_score"], label="train score")
plt.legend(loc="lower left")
plt.xlabel("number of validation run")
plt.ylabel("accuracy")
plt.title("GBT algorithm")
plt.show()
```



AdaBoost

```
precision
                             recall f1-score
                                                 support
           0
                    0.84
                               0.96
                                          0.89
                                                     7060
           1
                    0.69
                               0.31
                                          0.43
                                                     1940
                                                     9000
                                          0.82
    accuracy
   macro avg
                    0.76
                               0.64
                                          0.66
                                                     9000
                               0.82
                                                     9000
weighted avg
                    0.80
                                          0.79
```

```
[]: cv = cross_validate(model_ada, X, y, return_train_score=True, cv=5)
   plt.plot(cv["test_score"], label="test score")
   plt.plot(cv["train_score"], label="train score")
```

```
plt.legend(loc="lower left")
plt.xlabel("number of validation run")
plt.ylabel("accuracy")
plt.title("AdaBoost")
plt.show()
```



XGBoost

```
[]: import xgboost as xgb

model_xgb = xgb.XGBClassifier(learning_rate=0.1, max_depth=6, n_estimators=100)
model_xgb.fit(X_train, y_train)
y_pred = model_xgb.predict(X_val)
print(classification_report(y_true=y_val, y_pred=y_pred))
```

	precision	recall	f1-score	support
0	0.84 0.67	0.95 0.36	0.89	7060 1940
1	0.07	0.50	0.47	1340
accuracy			0.82	9000
macro avg	0.76	0.66	0.68	9000
weighted avg	0.81	0.82	0.80	9000

```
[]: cv = cross_validate(model_xgb, X, y, return_train_score=True, cv=5)
    plt.plot(cv["test_score"], label="test score")
    plt.plot(cv["train_score"], label="train score")
    plt.legend(loc="lower left")
    plt.xlabel("number of validation run")
    plt.ylabel("accuracy")
    plt.title("XGBoost")
    plt.show()
```

