XGBoost.

November 8, 2021

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[]: import pandas as pd
     import numpy as np
     import seaborn as sns
     import matplotlib.pyplot as plt
     from imblearn.combine import SMOTETomek
     from sklearn.model_selection import train_test_split
     from sklearn import preprocessing
     from sklearn.metrics import confusion matrix, accuracy_score, precision_score,
     →recall_score, roc_curve, roc_auc_score
     import xgboost as xgb
     from xgboost import XGBClassifier
     from xgboost import cv
[]: #Import Drive API and authenticate
     from google.colab import drive
     #Mount Drive to the Colab VM
     drive.mount('/content/drive')
    Drive already mounted at /content/drive; to attempt to forcibly remount, call
    drive.mount("/content/drive", force_remount=True).
[]: #Load the dataset into pandas DataFrame
     df = pd.read_csv("/content/drive/MyDrive/Capstone_project/v2_credit_default.
     ⇔csv")
[]: #Seperate the independent and dependent variables.
     X = df.drop(['Default'], axis=1)
     y = df['Default']
[]: # train - test split
     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.3,__
     \rightarrowrandom_state = 1)
```

```
[]: # # define data_dmatrix
    data_dmatrix = xgb.DMatrix(data=X,label=y)

[]: # Scale predictors
    min_max_scaler = preprocessing.MinMaxScaler()
    X_train_scaled = min_max_scaler.fit_transform(X_train)
```

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[]: # balancing the training data:

X_smt, y_smt = SMOTETomek(random_state=1).fit_sample(X_train_scaled, y_train.

→ squeeze())

#Now to check the training data distribution is balanced:

pd.Series(y_smt.reshape(-1)).value_counts().plot(kind='bar', title='Training_u

→data - SMOTETomek');
```

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function safe_indexing is deprecated; safe_indexing is deprecated in version 0.22 and will be removed in version 0.24.

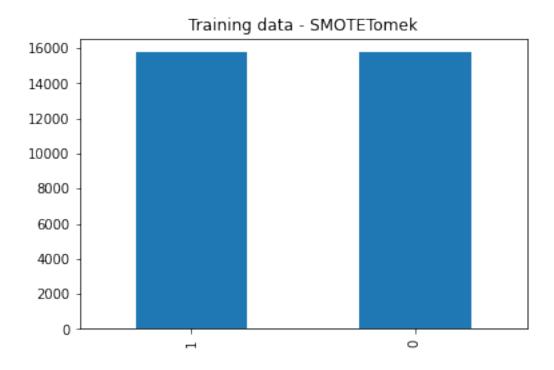
warnings.warn(msg, category=FutureWarning)

X_test_scaled = min_max_scaler.transform(X_test)

/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87:

FutureWarning: Function safe_indexing is deprecated; safe_indexing is deprecated in version 0.22 and will be removed in version 0.24.

warnings.warn(msg, category=FutureWarning)



```
[]: # declare XGBooster parameters
     params = {
                 'objective': 'binary:logistic',
                 'max_depth': 4,
                 'alpha': 10,
                 'learning_rate': 1.0,
                 'n estimators':100
             }
     # instantiate the classifier
     xgb_clf = XGBClassifier(**params)
     # fit the classifier to the training data
     xgb_clf.fit(X_smt, y_smt)
[]: XGBClassifier(alpha=10, base_score=0.5, booster='gbtree', colsample_bylevel=1,
                   colsample_bynode=1, colsample_bytree=1, gamma=0,
                   learning_rate=1.0, max_delta_step=0, max_depth=4,
                   min_child_weight=1, missing=None, n_estimators=100, n_jobs=1,
                   nthread=None, objective='binary:logistic', random_state=0,
                   reg_alpha=0, reg_lambda=1, scale_pos_weight=1, seed=None,
                   silent=None, subsample=1, verbosity=1)
[]: # make predictions on test data
     y_pred = xgb_clf.predict(X_test_scaled)
[]: # evaluation metrics
     cm = confusion_matrix(y_test, y_pred)
     print(cm)
     print('accuracy', accuracy_score(y_test, y_pred))
     print('precision', precision_score(y_test, y_pred))
     print('recall', recall_score(y_test, y_pred))
    [[6294 687]
     [1203 806]]
    accuracy 0.7897664071190211
    precision 0.5398526456798393
    recall 0.4011946241911399
[]: # cross validation with the metric: ROC AUC
     params = {"objective":"binary:logistic",'colsample_bytree': 0.3,'learning_rate':
     → 0.1,
                     'max_depth': 5, 'alpha': 10}
     xgb_cv = cv(dtrain=data_dmatrix, params=params, nfold=5,
```

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num_boost_round=50, early_stopping_rounds=10, u

→metrics="auc", as_pandas=True, seed=123)

xgb_cv.head()

[]: train-auc-mean train-auc-std test-auc-mean test-auc-std
```

```
0.745756
                            0.002224
                                           0.742265
                                                        0.008418
                            0.003481
             0.765999
                                           0.760508
                                                        0.004745
    1
    2
             0.770625
                            0.002431
                                           0.764191
                                                        0.006001
                                                        0.005666
             0.772996
                            0.001932
                                           0.765980
             0.776206
                            0.002038
                                           0.769430
                                                        0.004860
[]: #In the preprocessing step (MinMaxScaler) columns names are lost and replaced
     \rightarrowby f1, f1, f2 etc... Reintroduce names to the fitted model before plotting
     \rightarrow feature importance.
    xgb_clf.get_booster().feature_names = ['LIMIT_BAL', 'SEX', 'EDUCATION', __
     'Repay_Aug', 'Repay_July', 'Repay_June', 'Repay_May', 'Repay_Apr',
           'Bill_Sept', 'Bill_Aug', 'Bill_July', 'Bill_June', 'Bill_May',
            'Bill_Apr', 'Pay_Sept', 'Pay_Aug', 'Pay_July', 'Pay_June', 'Pay_May',
```

```
[]: # Feature importnace by weight - top 15

xgb.plot_importance(xgb_clf, title='Feature Importance by Weight',

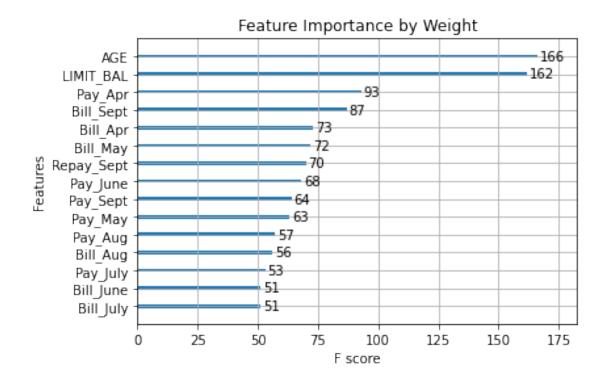
→max_num_features=15) # by default, importance_type='weight' or the number

→of times the feature appears in a tree

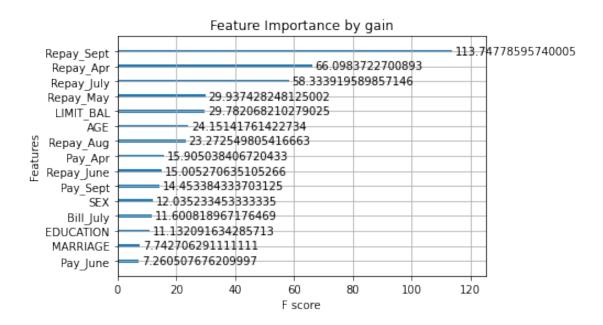
plt.figure(figsize = (26, 20))

plt.show()
```

'Pay_Apr']



<Figure size 1872x1440 with 0 Axes>



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