```
import pandas as pd
import matplotlib.pyplot as plt
from wordcloud import WordCloud
from collections import Counter
from PIL import Image
import os
from os import path
from google.colab import drive
drive.mount('/content/drive')
     Mounted at /content/drive
result=pd.read_csv('/content/drive/My_Drive/GSCM/BankChurners_PCA.csv')
from imblearn.over sampling import SMOTE
# Split target & features
X = result.drop('Attrition_Flag', axis=1)
y = result['Attrition_Flag']
# Dealing with imbalanced dataset
from imblearn.over_sampling import SMOTE
# Upsampling with SMOTE algorithm
smote = SMOTE(random state=66)
X_smote, y_smote = smote.fit_resample(X, y)
print(f'''Shape of X before SMOTE: {X.shape}
Shape of X after SMOTE: {X_smote.shape}''')
print('\nBalance of positive and negative classes (%):')
y_smote.value_counts(normalize=True) * 100
     Shape of X before SMOTE: (10127, 11)
     Shape of X after SMOTE: (17000, 11)
     Balance of positive and negative classes (%):
     0 50 0
        50.0
     Name: Attrition_Flag, dtype: float64
from \quad sklearn.\,model\_selection \quad import \quad train\_test\_split
X_train, X_test, y_train, y_test = train_test_split(X_smote, y_smote, train_size=0.8, random_state=69)
```

Visualize Model Score

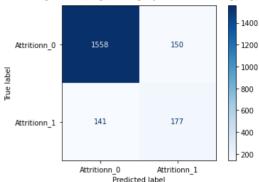
import numpy as np

```
from sklearn.ensemble import RandomForestClassifier ,AdaBoostClassifier,BaggingClassifier,ExtraTreesClassifier,GradientBoostingClassifier
from sklearn.model_selection import GridSearchCV, cross_val_score, StratifiedKFold, learning_curve ,KFold
from sklearn.metrics import roc_curve,accuracy_score,fl_score,auc,confusion_matrix,roc_auc_score,plot_confusion_matrix
from sklearn.metrics import XGBClassifier
from sklearn.metrics import classification_report,confusion_matrix,accuracy_score
from sklearn.tree import DecisionTreeClassifier
```

DecisionTreeClassifier

```
dt = DecisionTreeClassifier(random_state = 0)
dt.fit(X_train,y_train)
print("Accuracy: %.2f%%" % ((dt.score(X_test,y_test))*100.0))
```

/usr/local/lib/python3.8/dist-packages/sklearn/utils/deprecation.py:87: FutureWarning: Function plot_confusion_matrix is de warnings.warn(msg, category=FutureWarning)



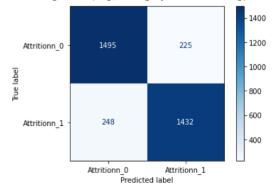
GradientBoostingClassifier

```
from sklearn.ensemble import GradientBoostingClassifier

gb = GradientBoostingClassifier(random_state = 0)
gb.fit(X_train, y_train)
print("Accuracy: %. 2f%%" % ((gb.score(X_test, y_test))*100.0))
```

Accuracy: 86.09%

/usr/local/lib/python3.8/dist-packages/sklearn/utils/deprecation.py:87: FutureWarning: Function plot_confusion_matrix is de warnings.warn(msg, category=FutureWarning)



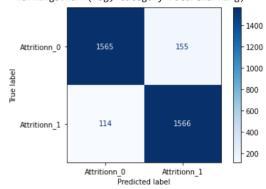
Random Forest

```
from sklearn.ensemble import RandomForestClassifier

RFC = RandomForestClassifier(random_state = 0)

RFC.fit(X_train, y_train)
print("Accuracy: %. 2f%%" % ((RFC.score(X_test, y_test))*100.0))
```

/usr/local/lib/python3.8/dist-packages/sklearn/utils/deprecation.py:87: FutureWarning: Function plot_confusion_matrix is de warnings.warn(msg, category=FutureWarning)



```
feature = pd. Series(RFC. feature_importances_, index = X_train. columns). sort_values(ascending = False)
print(feature)
```

```
Principle_Component_Two
                            0.486513
Principle_Component_One
                            0.166600
Total_Relationship_Count
                            0.089794
                            0.047248
Education_Level
{\tt Dependent\_count}
                            0.045809
                            0.041943
Income Category
Customer Age cat
                            0.032230
Marital_Status
                            0.029365
Months_on_book_cat
                            0.026667
                            0.025580
Gender
                            0.008251
Card\_Category
dtype: float64
```

```
import seaborn as sns
plt.figure(figsize = (28,14))
sns.set(font_scale=2)
sns.barplot(x = feature, y = feature.index)
plt.title("Feature Importance")

plt.xlabel('Score')
plt.ylabel('Features')
plt.show()
```

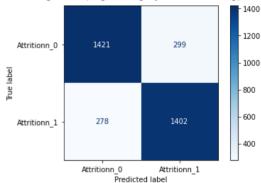
adaboost

```
from sklearn.ensemble import AdaBoostClassifier

ada = AdaBoostClassifier(random_state = 0)
ada.fit(X_train, y_train)
print("Accuracy: %.2f%%" % ((ada.score(X_test, y_test))*100.0))

Accuracy: 83.03%
```

/usr/local/lib/python3.8/dist-packages/sklearn/utils/deprecation.py:87: FutureWarning: Function plot_confusion_matrix is de warnings.warn(msg, category=FutureWarning)

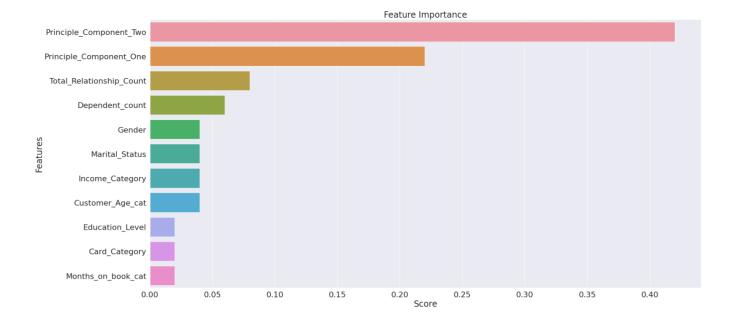


```
feature = pd. Series (ada. feature_importances_, index = X_train.columns).sort_values (ascending = False) print (feature)
```

```
Principle_Component_Two
                           0.42
Principle_Component_One
                           0.22
                           0.08
Total_Relationship_Count
Dependent_count
                           0.06
Gender
                           0.04
Marital_Status
                           0.04
                           0.04
Income_Category
Customer_Age_cat
                           0.04
Education Level
                           0.02
Card_Category
                           0.02
Months on book cat
                           0.02
dtype: float64
```

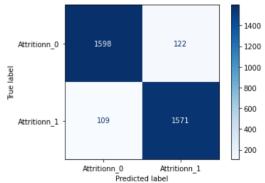
```
import seaborn as sns
plt.figure(figsize = (28,14))
sns.set(font_scale=2)
sns.barplot(x = feature, y = feature.index)
plt.title("Feature Importance")

plt.xlabel('Score')
plt.ylabel('Features')
plt.show()
```



ExtraTree Classifier

/usr/local/lib/python3.8/dist-packages/sklearn/utils/deprecation.py:87: FutureWarning: Function plot_confusion_matrix is de warnings.warn(msg, category=FutureWarning)



Feature Importance

```
feature = pd. Series(extraT.feature_importances_, index = X_train.columns).sort_values(ascending = False) print(feature)
```

Principle_Component_Two 0.408123 Principle_Component_One 0.129841

0.107655 Total_Relationship_Count Education_Level 0.067629 0.065919 Dependent_count 0.057044 Income_Category Marital_Status 0.044225 Customer Age cat 0.043823 Months_on_book_cat 0.036693 Gender 0.025425 0.013623 Card_Category dtype: float64

```
plt.figure(figsize = (28,14))
sns.set(font_scale=2)
sns.barplot(x = feature, y = feature.index)
plt.title("Feature Importance")
plt.xlabel('Score')
plt.ylabel('Features')
plt.show()
```

