```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model selection import train test split, GridSearchCV, RandomizedSearchCV
from sklearn.preprocessing import StandardScaler
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification report, confusion matrix, accuracy score
!pip install shap

→ Collecting shap

       Downloading shap-0.46.0-cp310-cp310-manylinux_2_12_x86_64.manylinux2010_x86_64.manylinux_2_17_x86_64.manylinux2014_x86_64.whl.metadata
     Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (from shap) (1.26.4)
     Requirement already satisfied: scipy in /usr/local/lib/python3.10/dist-packages (from shap) (1.13.1)
     Requirement already satisfied: scikit-learn in /usr/local/lib/python3.10/dist-packages (from shap) (1.3.2)
     Requirement already satisfied: pandas in /usr/local/lib/python3.10/dist-packages (from shap) (2.1.4)
     Requirement already satisfied: tqdm>=4.27.0 in /usr/local/lib/python3.10/dist-packages (from shap) (4.66.5)
     Requirement already satisfied: packaging>20.9 in /usr/local/lib/python3.10/dist-packages (from shap) (24.1)
     Collecting slicer==0.0.8 (from shap)
       Downloading slicer-0.0.8-py3-none-any.whl.metadata (4.0 kB)
     Requirement already satisfied: numba in /usr/local/lib/python3.10/dist-packages (from shap) (0.60.0)
     Requirement already satisfied: cloudpickle in /usr/local/lib/python3.10/dist-packages (from shap) (2.2.1)
     Requirement already satisfied: llvmlite<0.44,>=0.43.0dev0 in /usr/local/lib/python3.10/dist-packages (from numba->shap) (0.43.0)
     Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.10/dist-packages (from pandas->shap) (2.8.2)
     Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas->shap) (2024.1)
     Requirement already satisfied: tzdata>=2022.1 in /usr/local/lib/python3.10/dist-packages (from pandas->shap) (2024.1)
     Requirement already satisfied: joblib>=1.1.1 in /usr/local/lib/python3.10/dist-packages (from scikit-learn->shap) (1.4.2)
     Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn->shap) (3.5.0)
     Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.8.2->pandas->shap) (1.16.0)
     Downloading shap-0.46.0-cp310-cp310-manylinux_2_12_x86_64.manylinux2010_x86_64.manylinux_2_17_x86_64.manylinux2014_x86_64.whl (540 kB)
                                                - 540.1/540.1 kB 3.7 MB/s eta 0:00:00
     Downloading slicer-0.0.8-py3-none-any.whl (15 kB)
     Installing collected packages: slicer, shap
     Successfully installed shap-0.46.0 slicer-0.0.8
    |\cdot|
import shap
df = pd.read_csv("/content/Churn_Modelling.csv")
print(df.head())
print(df.info())
print(df.describe())
        EstimatedSalary Exited
             101348.88
                              1
     1
              112542.58
                              a
     2
              113931.57
                              1
     3
               93826.63
                              0
               79084.10
     4
                              0
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 10000 entries, 0 to 9999
     Data columns (total 14 columns):
     # Column
                           Non-Null Count Dtype
     ---
     0
         RowNumber
                           10000 non-null int64
          CustomerId
                           10000 non-null int64
     1
                           10000 non-null object
          Surname
          CreditScore
                           10000 non-null int64
          Geography
                           10000 non-null object
         Gender
                           10000 non-null object
                           10000 non-null int64
          Age
          Tenure
                           10000 non-null int64
```

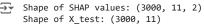
```
584.000000
                                                         32.000000
                                                                         3.000000
     25%
             2500.75000 1.562853e+07
     50%
             5000.50000
                         1.569074e+07
                                          652.000000
                                                         37.000000
                                                                         5.000000
                                                                         7.000000
     75%
             7500.25000 1.575323e+07
                                          718,000000
                                                         44,000000
            10000.00000
                        1.581569e+07
                                          850.000000
                                                         92.000000
                                                                        10.000000
     max
                  Balance NumOfProducts
                                             HasCrCard
                                                        IsActiveMember
     count
             10000.000000
                            10000.000000
                                           10000.00000
                                                          10000.000000
     mean
             76485.889288
                                 1.530200
                                               0.70550
                                                              0.515100
             62397,405202
                                 0.581654
                                               0.45584
                                                              0.499797
     std
                                                              0.000000
     min
                 0.000000
                                 1.000000
                                               0.00000
                 0.000000
     25%
                                 1.000000
                                               0.00000
                                                              0.000000
     50%
             97198.540000
                                 1.000000
                                               1.00000
                                                              1.000000
                                               1.00000
                                                              1.000000
     75%
            127644.240000
                                 2.000000
            250898.090000
                                 4.000000
                                               1.00000
                                                              1.000000
            EstimatedSalary
                                    Exited
               10000.000000 10000.000000
     count
     mean
              100090.239881
                                  0.203700
                                  0.402769
               57510.492818
     std
                  11,580000
                                  0.000000
     min
     25%
               51002.110000
                                  0.000000
                                  0.000000
     50%
              100193.915000
     75%
                                  0.000000
              149388.247500
     max
              199992.480000
                                  1.000000
df = df.drop(['RowNumber', 'CustomerId', 'Surname'], axis=1)
df = pd.get_dummies(df, drop_first=True)
X = df.drop('Exited', axis=1)
y = df['Exited']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
# Train a Random Forest Classifier
rf_model = RandomForestClassifier(random_state=42)
rf_model.fit(X_train, y_train)
y_pred = rf_model.predict(X_test)
print("Confusion Matrix:")
print(confusion_matrix(y_test, y_pred))
print("\nClassification Report:")
print(classification_report(y_test, y_pred))
print("\nAccuracy Score:")
print(accuracy_score(y_test, y_pred))
 Confusion Matrix:
     [[2328
             88]
      [ 310 274]]
     Classification Report:
                   precision
                                 recall f1-score
                                                    support
                0
                         0.88
                                   0.96
                                             0.92
                                                       2416
                         0.76
                                   0.47
                                             0.58
                                                        584
                1
         accuracy
                                             0.87
                                                       3000
        macro avg
                         0.82
                                   0.72
                                             0.75
                                                       3000
                         0.86
                                             0.85
                                                       3000
     weighted avg
                                   0.87
     Accuracy Score:
     0.8673333333333333
# Define the parameter grid for RandomSearch
param_grid = {
     __
'n_estimators': [50, 100, 200],
    'max_features': ['auto', 'sqrt'],
    'max_depth': [10, 20, 30, None],
    'min_samples_split': [2, 5, 10],
    'min_samples_leaf': [1, 2, 4],
    'bootstrap': [True, False]
}
# RandomizedSearchCV
rf_random = RandomizedSearchCV(estimator=rf_model, param_distributions=param_grid, n_iter=100, cv=3, verbose=2, random_state=42, n_jobs=-1)
```

```
rf_random.fit(X_train, y_train)
print(f"Best parameters found by RandomizedSearch: {rf_random.best_params_}")
best_rf_model = rf_random.best_estimator_
y_pred_best = best_rf_model.predict(X_test)
print("Confusion Matrix (Best Model):")
print(confusion_matrix(y_test, y_pred_best))
print("\nClassification Report (Best Model):")
print(classification_report(y_test, y_pred_best))
print("\nAccuracy Score (Best Model):")
print(accuracy_score(y_test, y_pred_best))
         estimator._validate_params()
₹
       File "/usr/local/lib/python3.10/dist-packages/sklearn/base.py", line 638, in _validate_params
         validate_parameter_constraints(
       File "/usr/local/lib/python3.10/dist-packages/sklearn/utils/_param_validation.py", line 96, in validate_parameter_constraints
         raise InvalidParameterError(
     sklearn.utils._param_validation.InvalidParameterError: The 'max_features' parameter of RandomForestClassifier must be an int in the r
     80 fits failed with the following error:
     Traceback (most recent call last):
       File "/usr/local/lib/python3.10/dist-packages/sklearn/model_selection/_validation.py", line 729, in _fit_and_score
         estimator.fit(X_train, y_train, **fit_params)
       File "/usr/local/lib/python3.10/dist-packages/sklearn/base.py", line 1145, in wrapper
         estimator. validate params()
       File "/usr/local/lib/python3.10/dist-packages/sklearn/base.py", line 638, in _validate_params
         validate parameter constraints(
       File "/usr/local/lib/python3.10/dist-packages/sklearn/utils/_param_validation.py", line 96, in validate_parameter_constraints
         raise InvalidParameterError(
     sklearn.utils._param_validation.InvalidParameterError: The 'max_features' parameter of RandomForestClassifier must be an int in the r
       warnings.warn(some_fits_failed_message, FitFailedWarning)
     /usr/local/lib/python3.10/dist-packages/sklearn/model_selection/_search.py:979: UserWarning: One or more of the test scores are non-f
      0.85842867 0.86314308 0.86157155 0.86342859 0.86485724
                                                                   nan
      0.86357134
                        nan 0.85900018
                                             nan
                                                        nan
                                                                   nan
                                             nan 0.86100016
             nan
                        nan
                                  nan
                                                                   nan
             nan
                        nan 0.86314308
                                             nan 0.86400004 0.86057183
             nan
                        nan 0.86028559 0.86199993 0.86199975
                                                                   nan
      0.86342853 0.86157142 0.86285708 0.86342853 0.85614287 0.85585712
             nan
                        nan 0.86214299
                                             nan 0.85700014
                                                                   nan
      0.86314308
                        nan
                                             nan 0.8574284
                                                        nan 0.85985671
                        nan
                                  nan
                                             nan
             nan
      0.8565715
                                  nan 0.8612861 0.86228581
                        nan
                                                                   nan
                        nan 0.86300008 0.86085716
                                                        nan 0.86242899
             nan
      0.86271463
                        nan 0.86100016
                                                        nan 0.86157197
                        nan 0.86399997
             nan
                                             nan
                                                        nan
                                                                   nan
             nan
                        nan
                                  nan
                                             nan
                                                        nan 0.86071477
      0.86242857
                        nan
                                  nan
                                             nan 0.85771434
      0.86485724 0.86328565
                                  nan 0.86171442]
      warnings.warn(
     Best parameters found by RandomizedSearch: {'n_estimators': 50, 'min_samples_split': 2, 'min_samples_leaf': 4, 'max_features': 'sqrt'
     Confusion Matrix (Best Model):
     [[2333 83]
      [ 316 268]]
     Classification Report (Best Model):
                               recall f1-score
                   precision
                                                   support
                0
                        0.88
                                  0.97
                                            0.92
                                                      2416
                1
                        0.76
                                  0.46
                                            0.57
                                                      584
                                            0.87
                                                      3000
         accuracy
                        0.82
                                  0.71
                                            0.75
                                                      3000
        macro avg
                                                     3000
     weighted avg
                        0.86
                                  0.87
                                            0.85
     Accuracy Score (Best Model):
     0.867
    4
# Initialize SHAP explainer
explainer = shap.TreeExplainer(best_rf_model)
# Get SHAP values
shap_values = explainer.shap_values(X_test)
```

```
# IT It s a Dinary Classification, SHAP may return a list of arrays
# Select shap values for class 1 (index 1) for binary classification
if isinstance(shap_values, list):
    shap_values = shap_values[1]

# Check the shape of SHAP values and X_test
print(f"Shape of SHAP values: {shap_values.shape}")
print(f"Shape of X_test: {X_test.shape}")

# Ensure the shapes match
assert shap_values.shape[0] == X_test.shape[0], "Mismatch in the number of samples"
assert shap_values.shape[1] == X_test.shape[1], "Mismatch in the number of features"
# Summary plot for the selected class
shap.summary_plot(shap_values, X_test, feature_names=X.columns)
```





```
# Save the model as a pickle file
import pickle
model_filename = 'best_rf_model.pkl'

# Open a file in write-binary mode and save the model
with open(model_filename, 'wb') as file:
    pickle.dump(best_rf_model, file)
```