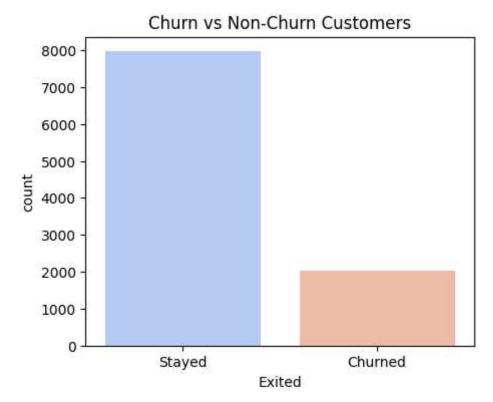
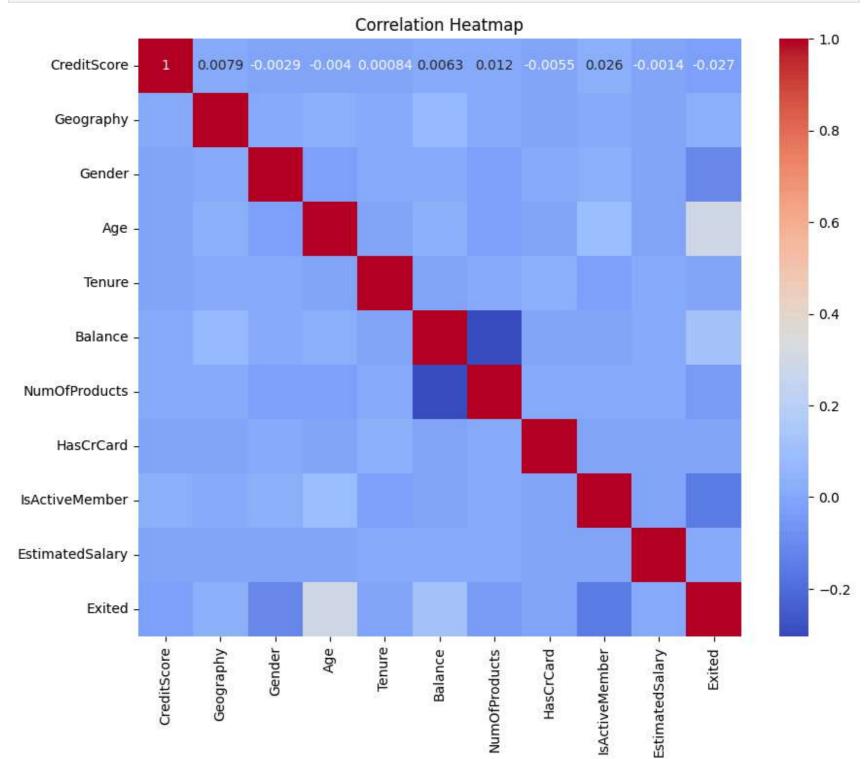
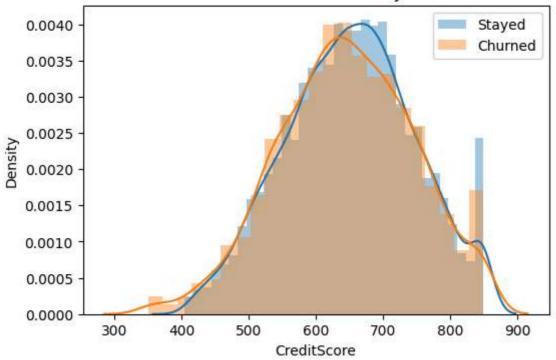
```
In [1]: import numpy as np
        import pandas as pd
        import seaborn as sns
         import matplotlib.pyplot as plt
        from sklearn.model_selection import train_test_split
         from sklearn.preprocessing import LabelEncoder
        from sklearn.linear_model import LogisticRegression
        from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier
        from sklearn.metrics import classification_report, roc_auc_score
In [2]: df = pd.read_csv("C:\\Users\\damma\\Downloads\\Customer Churn Prediction\\Churn_Modelling.csv")
In [3]:
        df.head()
                                                                                      Balance NumOfProducts HasCrCard IsActiveMem
Out[3]:
           RowNumber CustomerId Surname CreditScore Geography Gender Age Tenure
        0
                         15634602 Hargrave
                                                  619
                                                          France
                                                                 Female
                                                                          42
                                                                                  2
                                                                                          0.00
                                                                                                           1
                                                                                                                     1
        1
                     2
                         15647311
                                       Hill
                                                  608
                                                                 Female
                                                                          41
                                                                                      83807.86
                                                                                                                     0
                                                           Spain
        2
                     3
                         15619304
                                      Onio
                                                  502
                                                          France
                                                                 Female
                                                                          42
                                                                                  8
                                                                                    159660.80
                                                                                                           3
                                                                                                                     1
        3
                         15701354
                                                                                          0.00
                                                                                                                     0
                     4
                                      Boni
                                                  699
                                                          France
                                                                 Female
                                                                          39
        4
                     5
                         15737888
                                   Mitchell
                                                  850
                                                           Spain Female
                                                                          43
                                                                                  2 125510.82
                                                                                                                     1
In [4]:
        print("Shape:", df.shape)
        print("\nMissing values:\n", df.isnull().sum())
        print("\nChurn distribution:\n", df['Exited'].value_counts())
        Shape: (10000, 14)
        Missing values:
         RowNumber
                             0
        CustomerId
                            0
        Surname
                            0
                            0
        CreditScore
                            0
        Geography
        Gender
                            0
        Age
        Tenure
                            0
        Balance
                            0
        NumOfProducts
                            0
        HasCrCard
                            0
        IsActiveMember
                            0
        EstimatedSalary
                            0
        Exited
        dtype: int64
        Churn distribution:
         Exited
             7963
             2037
        Name: count, dtype: int64
        plt.figure(figsize=(5,4))
        sns.countplot(x='Exited', data=df, palette='coolwarm')
        plt.title("Churn vs Non-Churn Customers")
        plt.xticks([0,1], ['Stayed', 'Churned'])
        plt.show()
        c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1498: FutureWarning: is_categorical_dtype is de
        precated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
          if pd.api.types.is_categorical_dtype(vector):
        c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1498: FutureWarning: is_categorical_dtype is de
        precated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
          if pd.api.types.is_categorical_dtype(vector):
        c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1498: FutureWarning: is_categorical_dtype is de
        precated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
        if pd.api.types.is_categorical_dtype(vector):
```





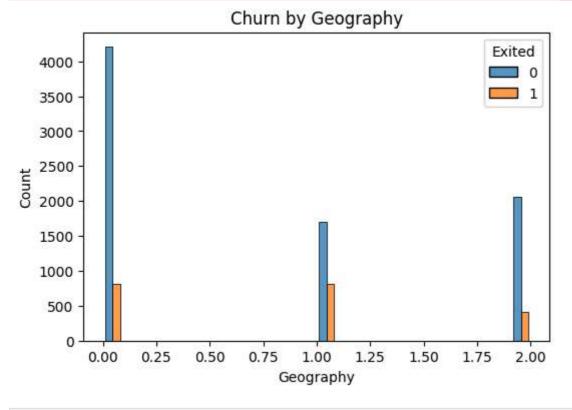
```
In [24]:
         plt.figure(figsize=(6,4))
         sns.distplot(df[df['Exited']==0]['CreditScore'], label='Stayed', hist=True, kde=True)
         sns.distplot(df[df['Exited']==1]['CreditScore'], label='Churned', hist=True, kde=True)
         plt.title("Credit Score Distribution by Churn")
         plt.legend()
         plt.show()
         C:\Users\damma\AppData\Local\Temp\ipykernel_102580\2469896775.py:2: UserWarning:
         `distplot` is a deprecated function and will be removed in seaborn v0.14.0.
         Please adapt your code to use either `displot` (a figure-level function with
         similar flexibility) or `histplot` (an axes-level function for histograms).
         For a guide to updating your code to use the new functions, please see
         https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751
           sns.distplot(df[df['Exited']==0]['CreditScore'], label='Stayed', hist=True, kde=True)
         c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1498: FutureWarning: is_categorical_dtype is de
         precated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
           if pd.api.types.is_categorical_dtype(vector):
         c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is de
         precated and will be removed in a future version. Convert inf values to NaN before operating instead.
           with pd.option_context('mode.use_inf_as_na', True):
         C:\Users\damma\AppData\Local\Temp\ipykernel_102580\2469896775.py:3: UserWarning:
         `distplot` is a deprecated function and will be removed in seaborn v0.14.0.
         Please adapt your code to use either `displot` (a figure-level function with
         similar flexibility) or `histplot` (an axes-level function for histograms).
         For a guide to updating your code to use the new functions, please see
         https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751
           sns.distplot(df[df['Exited']==1]['CreditScore'], label='Churned', hist=True, kde=True)
         c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1498: FutureWarning: is_categorical_dtype is de
         precated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
           if pd.api.types.is_categorical_dtype(vector):
         c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\ oldcore.py:1119: FutureWarning: use inf as na option is de
         precated and will be removed in a future version. Convert inf values to NaN before operating instead.
           with pd.option_context('mode.use_inf_as_na', True):
```

## Credit Score Distribution by Churn



```
In [23]: plt.figure(figsize=(6,4))
    sns.histplot(data=df, x='Geography', hue='Exited', multiple='dodge', shrink=0.8)
    plt.title("Churn by Geography")
    plt.show()
```

```
c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1498: FutureWarning: is_categorical_dtype is de
precated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1498: FutureWarning: is_categorical_dtype is de
precated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1498: FutureWarning: is_categorical_dtype is de
precated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1498: FutureWarning: is_categorical_dtype is de
precated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead
    if pd.api.types.is_categorical_dtype(vector):
c:\Users\damma\anaconda4\envs\tf\lib\site-packages\seaborn\_oldcore.py:1119: FutureWarning: use_inf_as_na option is de
precated and will be removed in a future version. Convert inf values to NaN before operating instead.
    with pd.option_context('mode.use_inf_as_na', True):
```



```
drop_cols = ['RowNumber', 'CustomerId', 'Surname']
In [6]:
         df = df.drop(columns=drop_cols)
In [7]: label_enc = LabelEncoder()
          for col in df.select_dtypes(include=['object']).columns:
             df[col] = label_enc.fit_transform(df[col])
In [8]: X = df.drop(columns=['Exited'])
         y = df['Exited']
In [9]: X_train, X_test, y_train, y_test = train_test_split( X, y, test_size=0.2, random_state=42, stratify=y)
In [10]: log_reg = LogisticRegression(max_iter=1000)
         log_reg.fit(X_train, y_train)
         y_pred_lr = log_reg.predict(X_test)
         print("\nLogistic Regression:")
         print(classification_report(y_test, y_pred_lr))
         print("ROC-AUC:", roc_auc_score(y_test, log_reg.predict_proba(X_test)[:, 1]))
         Logistic Regression:
                       precision
                                    recall f1-score
                                                        support
                    0
                            0.80
                                      1.00
                                                 0.89
                                                           1593
                                      0.00
                                                 0.00
                                                            407
                    1
                            0.00
```

2000

2000

2000

0.80

0.44

0.71

ROC-AUC: 0.554111893094944

0.40

0.63

0.50

0.80

accuracy

macro avg
weighted avg

```
c:\Users\damma\anaconda4\envs\tf\lib\site-packages\sklearn\metrics\ classification.py:1469: UndefinedMetricWarning: Pr
         ecision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` para
         meter to control this behavior.
            _warn_prf(average, modifier, msg_start, len(result))
         c:\Users\damma\anaconda4\envs\tf\lib\site-packages\sklearn\metrics\_classification.py:1469: UndefinedMetricWarning: Pr
         ecision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` para
         meter to control this behavior.
            _warn_prf(average, modifier, msg_start, len(result))
         c:\Users\damma\anaconda4\envs\tf\lib\site-packages\sklearn\metrics\_classification.py:1469: UndefinedMetricWarning: Pr
         ecision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` para
         meter to control this behavior.
         _warn_prf(average, modifier, msg_start, len(result))
In [11]: rf = RandomForestClassifier(n_estimators=100, random_state=42)
         rf.fit(X_train, y_train)
         y_pred_rf = rf.predict(X_test)
         print("\nRandom Forest:")
         print(classification_report(y_test, y_pred_rf))
         print("ROC-AUC:", roc_auc_score(y_test, rf.predict_proba(X_test)[:, 1]))
         Random Forest:
                                   recall f1-score support
                       precision
                    0
                                      0.97
                                                           1593
                            0.88
                                                0.92
                            0.78
                                      0.46
                                                0.58
                                                           407
                                                           2000
             accuracy
                                                0.86
            macro avg
                            0.83
                                      0.71
                                                0.75
                                                           2000
                                                           2000
         weighted avg
                            0.86
                                      0.86
                                                0.85
         ROC-AUC: 0.846416524382626
         gb = GradientBoostingClassifier(random_state=42)
In [12]:
         gb.fit(X_train, y_train)
         y_pred_gb = gb.predict(X_test)
         print("\nGradient Boosting:")
         print(classification_report(y_test, y_pred_gb))
         print("ROC-AUC:", roc_auc_score(y_test, gb.predict_proba(X_test)[:, 1]))
         Gradient Boosting:
                       precision
                                   recall f1-score
                                                       support
                    0
                            0.88
                                      0.97
                                                0.92
                                                           1593
                            0.79
                                      0.48
                                                0.59
                                                           407
                    1
                                                           2000
                                                0.87
             accuracy
            macro avg
                            0.83
                                      0.72
                                                0.76
                                                           2000
                                                           2000
         weighted avg
                            0.86
                                      0.87
                                                0.85
         ROC-AUC: 0.867301045267147
In [26]: # Simple churn check
         age = 40
         balance = 60000
         is_active = True
         if balance < 5000 or not is_active:</pre>
             print("Likely to Churn")
             print("Likely to Stay")
```

Likely to Stay