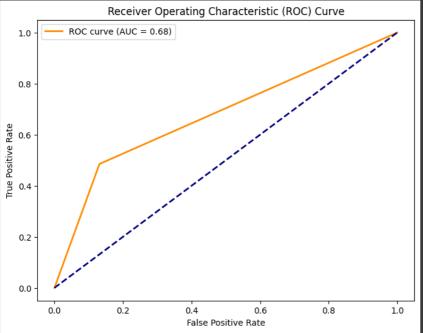
```
# Import necessary libraries
   import pandas as pd
    from \ sklearn.model\_selection \ import \ train\_test\_split
   from sklearn.preprocessing import LabelEncoder, OneHotEncoder
   from sklearn.impute import SimpleImputer
    from sklearn.neighbors import KNeighborsClassifier
    from sklearn.metrics import accuracy_score, classification_report
   # Load dataset
   df = pd.read_csv('/content/drive/MyDrive/Colab Notebooks/Dataset/KNN_Customer Churn_Dataset.csv')
   # Drop unnecessary columns
   df = df.drop(['customerID', 'testIndicator'], axis=1)
   # Encode the target variable 'Churn' into numerical values
   label_encoder = LabelEncoder()
   df['Churn'] = label_encoder.fit_transform(df['Churn'])
   # Identify and one-hot encode categorical columns
   categorical_cols = df.select_dtypes(include=['object']).columns
   df = pd.get_dummies(df, columns=categorical_cols, drop_first=True)
1 # Separate features (X) and target variable (y)
2 X = df.drop('Churn', axis=1)
3 y = df['Churn']
1 # Handle missing values in features
2 imputer = SimpleImputer(strategy='mean') # You can change the strategy as needed
3 X = pd.DataFrame(imputer.fit_transform(X), columns=X.columns)
1 # Split the dataset into training and testing sets
2 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
1 # Initialize KNN classifier
2 knn_classifier = KNeighborsClassifier(n_neighbors=3)
1 # Train the model
2 knn_classifier.fit(X_train, y_train)
             KNeighborsClassifier
    KNeighborsClassifier(n_neighbors=3)
1\ \mbox{\#} Make predictions on the test set
2 y_pred = knn_classifier.predict(X_test)
1 # Evaluate the model
2 accuracy = accuracy_score(y_test, y_pred)
3 print(f'Accuracy: {accuracy:.2f}')
    Accuracy: 0.77
1 # Display classification report
2 print(classification_report(y_test, y_pred))
                  precision recall f1-score support
                       0.82
                                 0.87
                                           0.85
                                                     1036
                       0.57
                                 0.49
                                           0.52
                                                      373
                                           0.77
                                                     1409
        accuracy
       macro avg
                       0.70
                                 0.68
                                           0.69
                                                     1409
    weighted avg
                                           0.76
1 from sklearn.metrics import confusion_matrix
2 import seaborn as sns
{\tt 3} import matplotlib.pyplot as plt
5 # Generate confusion matrix
6 cm = confusion_matrix(y_test, y_pred)
8 # Plot confusion matrix
```

```
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                                                                   01 Customer Churn Prediction.ipynb - Colaboratory
      9 sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', cbar=False
     10 plt.xlabel('Predicted')
11 plt.ylabel('Actual')
     12 plt.title('Confusion Matrix')
                                            Confusion Matrix
                                   900
                                                                       136
              0
           Actual
                                  192
                                                                       181
                                    0
                                                                         1
                                                  Predicted
```

```
1 from sklearn.metrics import roc_curve, auc
 3 # Calculate ROC curve
 4 fpr, tpr, thresholds = roc_curve(y_test, y_pred)
 5 roc_auc = auc(fpr, tpr)
 7 # Plot ROC curve
 8 plt.figure(figsize=(8, 6))
9 plt.plot(fpr, tpr, color='darkorange', lw=2, label=f'ROC curve (AUC = {roc_auc:.2f})')
10 plt.plot([0, 1], [0, 1], color='navy', lw=2, linestyle='--')
11 plt.xlabel('False Positive Rate')
12 plt.ylabel('True Positive Rate')
13 plt.title('Receiver Operating Characteristic (ROC) Curve')
14 plt.legend()
15 plt.show()
```



```
from sklearn.metrics import precision_recall_curve, average_precision_score
# Calculate Precision-Recall curve
average_precision = average_precision_score(y_test, y_pred)
```

```
6
7 # Plot Precision-Recall curve
8 plt.figure(figsize=(8, 6))
9 plt.plot(recall, precision, color='darkorange', lw=2, label=f'Precision-Recall curve (Avg. Precision = {average_prec
10 plt.xlabel('Recall')
11 plt.ylabel('Precision')
12 plt.title('Precision-Recall Curve')
13 plt.legend()
14 plt.show()
```

