```
In [106...
            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
            from sklearn.preprocessing import StandardScaler
            from sklearn.linear_model import LogisticRegression
            from sklearn.metrics import accuracy_score
            from sklearn.metrics import precision_score, recall_score, f1_score
            from sklearn import svm
            from sklearn.neighbors import KNeighborsClassifier
            from sklearn.tree import DecisionTreeClassifier
            from sklearn.ensemble import RandomForestClassifier
            from sklearn.ensemble import GradientBoostingClassifier
  In [2]:
            data = pd.read_csv('C:\Churn_Modelling.csv')
  In [3]:
            data.head()
 Out[3]:
              RowNumber CustomerId Surname CreditScore Geography Gender Age Tenure
                                                                                             Balance N
           0
                             15634602
                                                                                                0.00
                       1
                                      Hargrave
                                                       619
                                                                France
                                                                       Female
                                                                                42
                                                                                         2
           1
                        2
                             15647311
                                           Hill
                                                       608
                                                                Spain
                                                                       Female
                                                                                41
                                                                                            83807.86
           2
                        3
                             15619304
                                          Onio
                                                       502
                                                                       Female
                                                                                42
                                                                                           159660.80
                                                                France
           3
                             15701354
                                                       699
                                                                                                0.00
                                          Boni
                                                                France
                                                                       Female
                                                                                39
                                                                                         2 125510.82
                        5
                            15737888
                                       Mitchell
                                                       850
                                                                Spain
                                                                       Female
                                                                                43
 In [4]:
            data.tail()
 Out[4]:
                 RowNumber CustomerId
                                          Surname CreditScore Geography Gender Age Tenure
                                                                                                 Balance
           9995
                        9996
                                15606229
                                                          771
                                                                                            5
                                                                                                    0.00
                                           Obijiaku
                                                                   France
                                                                             Male
                                                                                    39
           9996
                        9997
                                15569892 Johnstone
                                                          516
                                                                   France
                                                                             Male
                                                                                    35
                                                                                            10
                                                                                                57369.61
           9997
                        9998
                                15584532
                                               Liu
                                                          709
                                                                   France Female
                                                                                    36
                                                                                            7
                                                                                                    0.00
           9998
                        9999
                                15682355
                                          Sabbatini
                                                          772
                                                                 Germany
                                                                             Male
                                                                                    42
                                                                                                75075.31
           9999
                       10000
                                15628319
                                            Walker
                                                          792
                                                                   France
                                                                           Female
                                                                                    28
                                                                                               130142.79
  In [5]:
            data.shape
           (10000, 14)
 Out[5]:
  In [6]:
            print("Number of Rows", data.shape[0])
            print("Number of Columns", data.shape[1])
           Number of Rows 10000
```

Number of Columns 14

```
In [7]:
          data.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 10000 entries, 0 to 9999
         Data columns (total 14 columns):
              Column
                              Non-Null Count Dtype
         ---
              -----
                               -----
              RowNumber
          0
                             10000 non-null int64
          1
              CustomerId
                              10000 non-null int64
          2
              Surname
                              10000 non-null object
          3
              CreditScore
                              10000 non-null int64
                             10000 non-null object
10000 non-null object
              Geography
          5
              Gender
          6
              Age
                              10000 non-null int64
          7
              Tenure
                              10000 non-null int64
              Balance 10000 non-null int64
Balance 10000 non-null float64
          8
          9 NumOfProducts 10000 non-null int64
10 HasCrCard 10000 non-null int64
          11 IsActiveMember 10000 non-null int64
          12 EstimatedSalary 10000 non-null float64
          13 Exited
                               10000 non-null int64
         dtypes: float64(2), int64(9), object(3)
         memory usage: 1.1+ MB
In [10]:
          data.isnull().sum()
Out[10]: RowNumber
                             0
         CustomerId
                            0
         Surname
                            0
         CreditScore
                            0
         Geography
                            0
         Gender
                            0
                            0
         Age
         Tenure
                            0
         Balance
                            0
         NumOfProducts
                            0
         HasCrCard
                             0
         IsActiveMember
                             0
         EstimatedSalary
                             0
         Exited
                             0
         dtype: int64
In [12]:
          data.describe(include = 'all')
```

Out[12]:		RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	
	count	10000.00000	1.000000e+04	10000	10000.000000	10000	10000	10000.000000	1000
	unique	NaN	NaN	2932	NaN	3	2	NaN	
	top	NaN	NaN	Smith	NaN	France	Male	NaN	
	freq	NaN	NaN	32	NaN	5014	5457	NaN	
	mean	5000.50000	1.569094e+07	NaN	650.528800	NaN	NaN	38.921800	
	std	2886.89568	7.193619e+04	NaN	96.653299	NaN	NaN	10.487806	
	min	1.00000	1.556570e+07	NaN	350.000000	NaN	NaN	18.000000	
	25%	2500.75000	1.562853e+07	NaN	584.000000	NaN	NaN	32.000000	
	50%	5000.50000	1.569074e+07	NaN	652.000000	NaN	NaN	37.000000	
	75%	7500.25000	1.575323e+07	NaN	718.000000	NaN	NaN	44.000000	

```
10000.00000 1.581569e+07
                                              NaN
                                                     850.000000
                                                                             NaN
                                                                                     92.000000
            max
                                                                     NaN
In [13]:
          data.columns
         Out[13]:
                dtype='object')
In [15]:
          data = data.drop(['RowNumber', 'CustomerId', 'Surname'],axis = 1)
In [16]:
          data.head()
                                                                NumOfProducts HasCrCard IsActiveMo
            CreditScore
Out[16]:
                       Geography Gender Age
                                              Tenure
                                                        Balance
          0
                   619
                                                    2
                                                           0.00
                            France
                                   Female
                                            42
                                                                             1
                                                                                       1
          1
                   608
                            Spain
                                   Female
                                            41
                                                    1
                                                       83807.86
                                                                             1
                                                                                       0
                   502
                                                      159660.80
                            France
                                   Female
                                            42
                                                    8
                                                                                       1
                                                                             2
                                                                                       0
          3
                   699
                                            39
                                                    1
                                                           0.00
                            France
                                   Female
                   850
                                            43
                                                    2 125510.82
                            Spain
                                   Female
In [17]:
          data['Geography'].unique()
         array(['France', 'Spain', 'Germany'], dtype=object)
Out[17]:
In [19]:
          data = pd.get_dummies(data,drop_first = True)
In [20]:
          data.head()
Out[20]:
            CreditScore
                                     Balance
                                             NumOfProducts HasCrCard
                                                                       IsActiveMember EstimatedSalar
                       Age
                            Tenure
          0
                   619
                                         0.00
                                                                                            101348.8
                                     83807.86
                                                                    0
                                                                                            112542.5
          1
                   608
                         41
                                 1
                                                          1
                                                                                   1
          2
                   502
                         42
                                    159660.80
                                                          3
                                                                    1
                                                                                   0
                                                                                            113931.5
          3
                   699
                         39
                                 1
                                        0.00
                                                          2
                                                                    0
                                                                                   0
                                                                                            93826.€
                   850
                         43
                                 2 125510.82
                                                                                            79084.1
In [80]:
           (data[data['Exited'] == 1].shape[0] / data.shape[0])*100
          20.36999999999997
Out[80]:
In [22]:
          data['Exited'].value_counts()
```

RowNumber

CustomerId Surname

CreditScore Geography Gender

Age

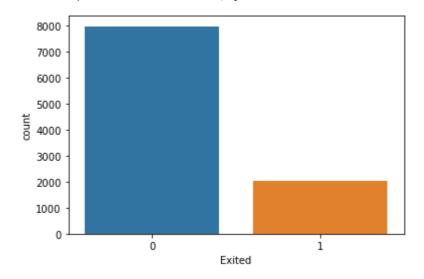
```
Out[22]: 0 7963
1 2037
```

Name: Exited, dtype: int64

```
In [31]: sns.countplot(data['Exited'])
```

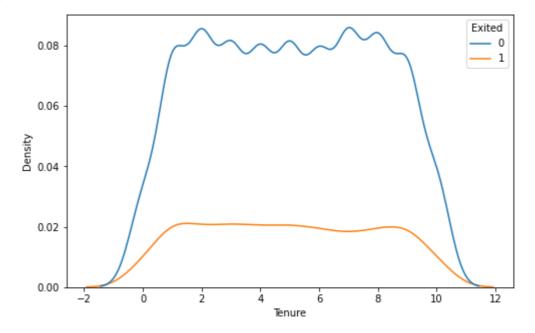
C:\Users\SYS\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: P
ass the following variable as a keyword arg: x. From version 0.12, the only valid po
sitional argument will be `data`, and passing other arguments without an explicit ke
yword will result in an error or misinterpretation.
 warnings.warn(

Out[31]: <AxesSubplot:xlabel='Exited', ylabel='count'>



```
In [82]: plt.figure(figsize=(8,5))
    sns.kdeplot(x = data['Tenure'], hue = data['Exited'])
```

Out[82]: <AxesSubplot:xlabel='Tenure', ylabel='Density'>



```
In [33]: y
```

```
Out[33]: 0
                1
                0
         2
                1
         3
         4
                0
         9995
                0
         9996
         9997
                1
         9998
                1
         9999
         Name: Exited, Length: 10000, dtype: int64
In [44]:
          x_train, x_test, y_train, y_test = train_test_split(x,y,test_size = 0.20,random_stat
In [47]:
           sc = StandardScaler()
In [48]:
          x_train = sc.fit_transform(x_train)
          x_test = sc.transform(x_test)
In [49]:
          x_train
-0.57773517, 0.90750738], [ 1.07927399, -0.18493108, -1.73189531, ..., 1.72916886,
                 -0.57773517, -1.10191942],
                [0.16821031, -0.18493108, 1.3751852, ..., -0.57831252,
                 -0.57773517, -1.10191942],
                [0.37527024, -0.37493284, 1.02995403, ..., -0.57831252,
                 1.73089688, 0.90750738],
                [ 1.56586482,
                             1.14508121, 0.68472287, ..., -0.57831252,
                  1.73089688, 0.90750738]])
        LogisticRegression
In [51]:
          log = LogisticRegression()
In [52]:
          log.fit(x_train, y_train)
Out[52]: LogisticRegression()
In [89]:
          y_pred1 = log.predict(x_test)
In [90]:
          accuracy_score(y_test,y_pred1)
```

0.808

precision_score(y_test, y_pred1)

Out[90]:

In [91]:

```
Out[91]: 0.5891472868217055
  In [92]:
            recall_score(y_test, y_pred1)
  Out[92]: 0.18673218673218672
  In [61]:
            f1_score(y_test, y_pred1)
  Out[61]: 0.2835820895522388
   In [ ]:
            pc = TP / (FP + TP)
rc = TP / (TP + FN)
           SVM
  In [66]:
            svm = svm.SVC()
  In [67]:
            svm.fit(x_train,y_train)
  Out[67]: SVC()
  In [68]:
            y_pred2 = svm.predict(x_test)
  In [93]:
            accuracy_score(y_test,y_pred2)
  Out[93]: 0.861
  In [94]:
            precision_score(y_test, y_pred2)
  Out[94]: 0.8341968911917098
           KNeighbors Classifier
  In [73]:
            knn = KNeighborsClassifier()
```

```
In [73]: knn = KNeighborsClassifier()

In [95]: knn.fit(x_train,y_train)

Out[95]: KNeighborsClassifier()

In [75]: y_pred3 = knn.predict(x_test)

In [96]: accuracy_score(y_test,y_pred3)
```

```
Out[96]: 0.824
In [97]:
           precision_score(y_test, y_pred3)
Out[97]: 0.62222222222222
         Decision Tree Classifier
In [84]:
          dt = DecisionTreeClassifier()
In [85]:
           dt.fit(x_train,y_train)
Out[85]: DecisionTreeClassifier()
In [86]:
           y_pred4 = dt.predict(x_test)
In [98]:
           accuracy_score(y_test,y_pred4)
Out[98]: 0.7865
In [99]:
           precision_score(y_test, y_pred4)
Out[99]: 0.47727272727273
         RandomForestClassifier
In [101...
           rf = RandomForestClassifier()
In [102...
           rf.fit(x_train,y_train)
          RandomForestClassifier()
Out[102...
In [103...
           y_pred5 = dt.predict(x_test)
In [104...
           accuracy_score(y_test,y_pred5)
          0.7865
Out[104...
In [105...
           precision_score(y_test, y_pred5)
```

GradientBoostingClassifier

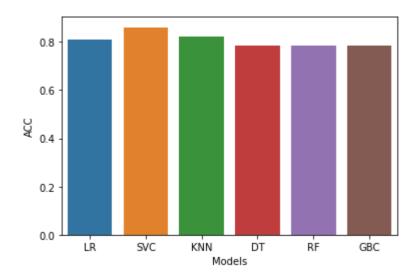
0.47727272727273

Out[105...

```
In [107...
          gbc = GradientBoostingClassifier()
In [108...
          gbc.fit(x_train,y_train)
Out[108...
          GradientBoostingClassifier()
In [109...
          y_pred6 = dt.predict(x_test)
In [110...
          accuracy_score(y_test,y_pred6)
          0.7865
Out[110...
In [111...
          precision_score(y_test, y_pred6)
          0.47727272727273
Out[111...
In [112...
          final_data = pd.DataFrame({'Models': ['LR','SVC','KNN','DT','RF','GBC'],'ACC':[accur
In [113...
          final_data
            Models
                     ACC
Out[113...
          0
                LR 0.8080
          1
               SVC 0.8610
          2
              KNN 0.8240
          3
                DT 0.7865
          4
                RF 0.7865
          5
               GBC 0.7865
In [114...
          sns.barplot(final_data['Models'],final_data['ACC'])
          ass the following variables as keyword args: x, y. From version 0.12, the only valid
          positional argument will be `data`, and passing other arguments without an explicit
          keyword will result in an error or misinterpretation.
           warnings.warn(
```

<AxesSubplot:xlabel='Models', ylabel='ACC'>

Out[114...



In [116...

final_data = pd.DataFrame({'Models': ['LR','SVC','KNN','DT','RF','GBC'],'PRE':[preci

In [118...

final_data

Out[118...

	Models	PRE		
0	LR	0.589147		
1	SVC	0.834197		
2	KNN	0.622222		
3	DT	0.477273		
4	RF	0.477273		
5	GBC	0.477273		

In [119...

sns.barplot(final_data['Models'],final_data['PRE'])

C:\Users\SYS\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: P
ass the following variables as keyword args: x, y. From version 0.12, the only valid
positional argument will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
 warnings.warn(

Out[119... <AxesSubplot:xlabel='Models', ylabel='PRE'>

