USTOMER CHURN PREDICTION 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 Balance Tenure 0 15634602 2 0.00 1 Hargrave 619 France Female 42 1 2 15647311 Hill 608 Spain Female 41 1 83807.86 2 3 15619304 Onio 502 Female 42 159660.80 France 8 3 15701354 699 0.00 Boni France Female 39 125510.82 5 850 15737888 Mitchell Spain Female 43 In [4]: data.tail() Out[4]: **RowNumber CustomerId** Surname CreditScore Geography Gender Age Tenure Balance 9995 9996 15606229 771 39 5 0.00 Obijiaku France Male 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 3 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])

> Number of Rows 10000 Number of Columns 14

print("Number of Columns", data.shape[1])

```
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

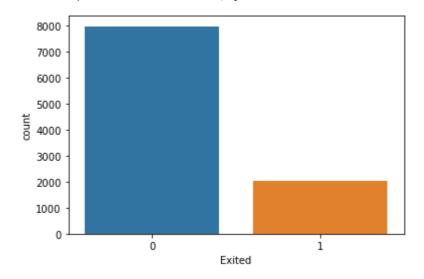
```
Out[22]: 0 7963
1 2037
```

Name: Exited, dtype: int64

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

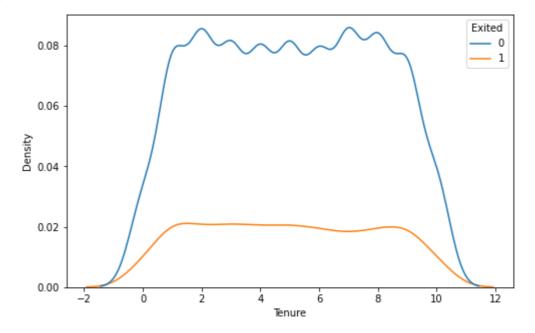
C:\Users\SYS\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: P
ass the following variable as a keyword arg: x. From version 0.12, the only valid po
sitional argument will be `data`, and passing other arguments without an explicit ke
yword will result in an error or misinterpretation.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

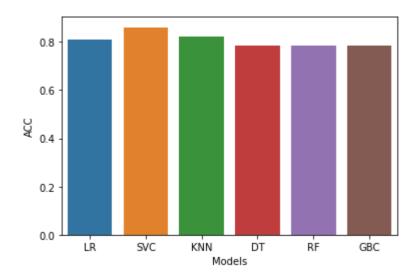
0.47727272727273

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'>

