```
In [2]: import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   import seaborn as sns
   import warnings
   warnings.filterwarnings("ignore")
```

In [3]: data = pd.read_csv("C:\\Users\\vaibhav vishal\\OneDrive\\Documents\\Churn_M

In [4]: data.head()

Out[4]:

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Bala
0	1	15634602	Hargrave	619	France	Female	42	2	
1	2	15647311	Hill	608	Spain	Female	41	1	8380
2	3	15619304	Onio	502	France	Female	42	8	15966
3	4	15701354	Boni	699	France	Female	39	1	
4	5	15737888	Mitchell	850	Spain	Female	43	2	12551
4									

In [5]: data.info()

<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			
1	CustomerId	10000 non-null	int64			
2	Surname	10000 non-null	object			
3	CreditScore	10000 non-null	int64			
4	Geography	10000 non-null	object			
5	Gender	10000 non-null	object			
6	Age	10000 non-null	int64			
7	Tenure	10000 non-null	int64			
8	Balance	10000 non-null	float64			
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			
dtypos: $float64(2)$ int64(0) object(3)						

dtypes: float64(2), int64(9), object(3)

memory usage: 1.1+ MB

```
In [6]:
        data.isnull().sum()
Out[6]: RowNumber
                            0
        CustomerId
                            0
        Surname
                            0
        CreditScore
                            0
        Geography
                            0
        Gender
                            0
        Age
                            0
        Tenure
                            0
        Balance
                            0
        NumOfProducts
                            0
        HasCrCard
                            0
        IsActiveMember
                            0
        EstimatedSalary
                            0
        Exited
                            0
        dtype: int64
In [7]: data.columns
dtype='object')
        data = data.drop(['RowNumber', 'CustomerId', 'Surname'],axis=1)
In [8]:
        data
Out[8]:
              CreditScore Geography Gender Age Tenure
                                                       Balance NumOfProducts HasCrCard
            0
                     619
                             France
                                   Female
                                            42
                                                   2
                                                          0.00
                                                                           1
                                                                                     1
            1
                     608
                             Spain
                                   Female
                                            41
                                                    1
                                                       83807.86
                                                                           1
                                                                                     0
            2
                     502
                             France
                                   Female
                                            42
                                                   8
                                                      159660.80
                                                                           3
                                                                                     1
            3
                     699
                             France
                                   Female
                                            39
                                                          0.00
                                                                           2
                                                                                     0
                                                   1
            4
                     850
                             Spain
                                   Female
                                            43
                                                      125510.82
                                                                                     1
                      ...
                                       ...
                                            ...
                                                   ...
                                                            ...
         9995
                     771
                             France
                                     Male
                                            39
                                                   5
                                                          0.00
                                                                           2
                                                                                     1
         9996
                     516
                             France
                                     Male
                                            35
                                                   10
                                                       57369.61
                                                                           1
                                                                                     1
                     709
                                                   7
         9997
                             France
                                   Female
                                            36
                                                          0.00
                                                                           1
                                                                                     0
```

9998

9999

772

792

10000 rows × 11 columns

Germany

France Female

Male

42

28

3

75075.31

130142.79

2

1

1

1

```
In [9]: data = pd.get_dummies(data,drop_first = True)
    data.head()
    data = data.astype(int)
    data
```

Out[9]:		CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	Estin
	0	619	42	2	0	1	1	1	
	1	608	41	1	83807	1	0	1	
	2	502	42	8	159660	3	1	0	
	3	699	39	1	0	2	0	0	
	4	850	43	2	125510	1	1	1	
	9995	771	39	5	0	2	1	0	
	9996	516	35	10	57369	1	1	1	
	9997	709	36	7	0	1	0	1	
	9998	772	42	3	75075	2	1	0	
	9999	792	28	4	130142	1	1	0	
	10000	rows × 12 co	olumn	s					

In [10]: data['Exited'].value_counts()

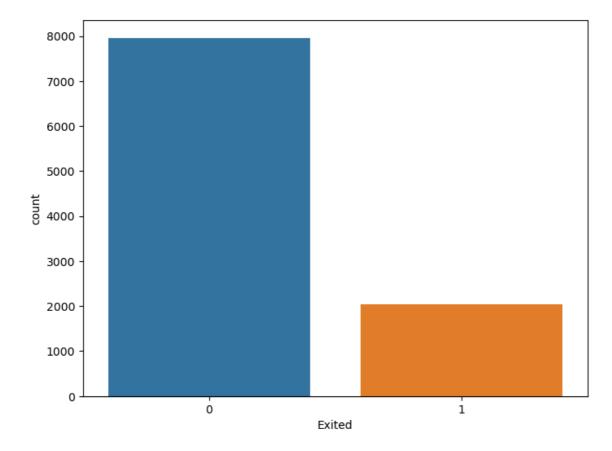
Out[10]: Exited

0 7963 1 2037

Name: count, dtype: int64

```
In [11]: plt.figure(figsize =(8,6))
sns.countplot(x='Exited',data = data)
```

Out[11]: <Axes: xlabel='Exited', ylabel='count'>



```
In [12]: X = data.drop('Exited',axis=1)
y = data['Exited']
```

```
In [13]: pip install imblearn
         Defaulting to user installation because normal site-packages is not writea
         Collecting imblearn
           Obtaining dependency information for imblearn from https://files.pythonh
         osted.org/packages/81/a7/4179e6ebfd654bd0eac0b9c06125b8b4c96a9d0a8ff9e9507
         eb2a26d2d7e/imblearn-0.0-py2.py3-none-any.whl.metadata (https://files.pyth
         onhosted.org/packages/81/a7/4179e6ebfd654bd0eac0b9c06125b8b4c96a9d0a8ff9e9
         507eb2a26d2d7e/imblearn-0.0-py2.py3-none-any.whl.metadata)
           Downloading imblearn-0.0-py2.py3-none-any.whl.metadata (355 bytes)
         Requirement already satisfied: imbalanced-learn in c:\programdata\anaconda
         3\lib\site-packages (from imblearn) (0.10.1)
         Requirement already satisfied: numpy>=1.17.3 in c:\programdata\anaconda3\l
         ib\site-packages (from imbalanced-learn->imblearn) (1.24.3)
         Requirement already satisfied: scipy>=1.3.2 in c:\programdata\anaconda3\li
         b\site-packages (from imbalanced-learn->imblearn) (1.11.1)
         Requirement already satisfied: scikit-learn>=1.0.2 in c:\programdata\anaco
         nda3\lib\site-packages (from imbalanced-learn->imblearn) (1.3.0)
         Requirement already satisfied: joblib>=1.1.1 in c:\programdata\anaconda3\l
         ib\site-packages (from imbalanced-learn->imblearn) (1.2.0)
         Requirement already satisfied: threadpoolctl>=2.0.0 in c:\programdata\anac
         onda3\lib\site-packages (from imbalanced-learn->imblearn) (2.2.0)
         Downloading imblearn-0.0-py2.py3-none-any.whl (1.9 kB)
         Installing collected packages: imblearn
         Successfully installed imblearn-0.0
         Note: you may need to restart the kernel to use updated packages.
In [15]: from sklearn.model_selection import train_test_split
         from sklearn.preprocessing import StandardScaler
         from sklearn.model_selection import train_test_split
         from sklearn.linear_model import LinearRegression , LogisticRegression
         from sklearn.metrics import r2 score
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.ensemble import RandomForestClassifier, GradientBoostingRegres
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.metrics import mean_squared_error
         from sklearn.preprocessing import StandardScaler
         from sklearn.preprocessing import LabelEncoder
         from sklearn.metrics import accuracy score
         from sklearn.metrics import confusion_matrix
         from sklearn.metrics import precision_score, recall_score, f1_score
In [16]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.1, ra
         print('Training Shape: ', X_train.shape)
         print('Testing Shape: ', X_test.shape)
         Training Shape: (9000, 11)
         Testing Shape: (1000, 11)
```

In [17]: | scaler = StandardScaler()

X_train_scaled = scaler.fit_transform(X_train)

X_test_scaled = scaler.transform(X_test)

```
In [18]: |X_train_scaled
Out[18]: array([[-0.47944328, 0.19687202, -0.00234647, ..., -0.5761528,
                  -0.57700814, 0.91105005],
                 [1.04580863, 1.33803657, 1.03625698, ..., -0.5761528]
                  -0.57700814, -1.09763453],
                 [-0.85297437, -0.08841912, 1.03625698, ..., -0.5761528]
                   1.73307782, 0.91105005],
                 [\ 0.86941896,\ -0.08841912,\ -1.38715108,\ \ldots,\ -0.5761528\ ,
                  -0.57700814, -1.09763453],
                 [0.16386025, 0.38706611, 1.03625698, ..., -0.5761528]
                  -0.57700814, 0.91105005],
                 [0.47513615, 1.14784248, -1.38715108, ..., 1.73565068,
                  -0.57700814, 0.91105005]])
In [33]: | from sklearn import svm
         threshold = 0.5
         y_train_classified = [1 if value > threshold else 0 for value in y train]
         svm = svm.SVC()
         svm.fit(X_train_scaled, y_train_classified)
Out[33]: SVC()
         In a Jupyter environment, please rerun this cell to show the HTML representation or
         trust the notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page
         with nbviewer.org.
In [34]: y test classified = [1 if value > threshold else 0 for value in y test]
         accuracy2 = svm.score(X_test_scaled, y_test_classified)
         print("Model Accuracy:", accuracy1)
         Model Accuracy: 0.809
 In [ ]:
In [35]: threshold = 0.5
         y_train_classified = [1 if value > threshold else 0 for value in y_train]
         LR = LogisticRegression()
         LR.fit(X_train_scaled, y_train_classified)
Out[35]: LogisticRegression()
         In a Jupyter environment, please rerun this cell to show the HTML representation or
         trust the notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page
         with nbviewer.org.
In [36]: y_test_classified = [1 if value > threshold else 0 for value in y_test]
         accuracy1 = LR.score(X_test_scaled, y_test_classified)
         print("Model Accuracy:", accuracy2)
         Model Accuracy: 0.865
```

```
In [37]: threshold = 0.5
    y_train_classified = [1 if value > threshold else 0 for value in y_train]
    dt = DecisionTreeClassifier()
    dt.fit(X_train_scaled, y_train_classified)
```

Out[37]: DecisionTreeClassifier()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

Model Accuracy: 0.863

```
In [39]: threshold = 0.5
    y_train_classified = [1 if value > threshold else 0 for value in y_train]
    rf = RandomForestClassifier()
    rf.fit(X_train_scaled, y_train_classified)
```

Out[39]: RandomForestClassifier()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

Model Accuracy: 0.797

```
In [41]: threshold = 0.5
    y_train_classified = [1 if value > threshold else 0 for value in y_train]
    KNN = KNeighborsClassifier()
    KNN.fit(X_train_scaled, y_train_classified)
```

Out[41]: KNeighborsClassifier()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [42]: y_test_classified = [1 if value > threshold else 0 for value in y_test]
    accuracy5 = KNN.score(X_test_scaled, y_test_classified)
    print("Model Accuracy:", accuracy5)
```

Model Accuracy: 0.84

```
In [43]: from sklearn.ensemble import GradientBoostingClassifier
    threshold = 0.5
    y_train_classified = [1 if value > threshold else 0 for value in y_train]
    GBC = GradientBoostingClassifier()
    GBC.fit(X_train_scaled, y_train_classified)
```

Out[43]: GradientBoostingClassifier()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

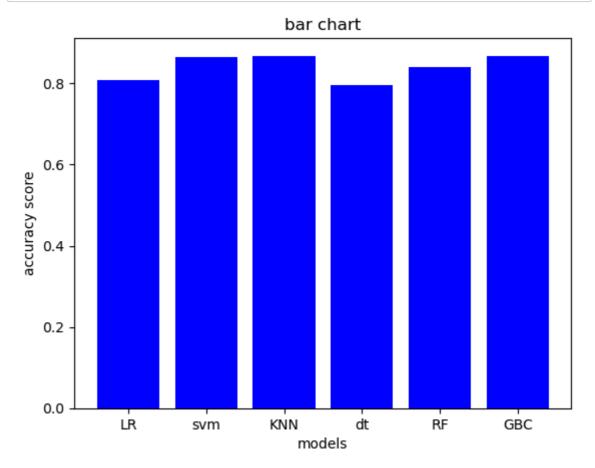
```
In [44]: y_test_classified = [1 if value > threshold else 0 for value in y_test]
accuracy6 = GBC.score(X_test_scaled, y_test_classified)
print("Model Accuracy:", accuracy6)
```

Model Accuracy: 0.867

Out[45]: Model ACC 0 LR 0.809 1 svm 0.865 2 KNN 0.868 3 dt 0.797 4 rf 0.840

GBC 0.867

```
In [46]: x=['LR','svm','KNN','dt','RF','GBC']
    y=[0.809,0.865,0.868,0.797,0.840,0.867]
    plt.bar(x,y,color='blue')
    plt.xlabel('models')
    plt.ylabel('accuracy score')
    plt.title('bar chart')
    plt.show()
```



```
In [ ]:
```