# CMS COLLEGE OF ENGINEERING AND TECNOLOGY

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DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING.
DETECTING PARKINSONS DISEASE (ASSIGNMENT 2)

<u>DATE</u> : 30-09-2022

**PROBLEM:** PERFORM TASKS ACCORDINGLY

NAME : MUHAMMED MUSAMMIL KP

**OUTPUT:** 

# **SCREENSHOTS:**

## 1.Download the Dataset

#### 2.Load the Dataset

In [1]: import numpy as np
 import pandas as pd
 import seaborn as sns
 import matplotlib.pyplot as plt
 import sklearn

In [2]: data = pd.read\_csv(r"C:\Users\chand\Downloads\IBM-Project\Assignments\Ass-2\Churn\_Modelling.csv")

### 3. Perform below Visualizations

**Univariate Analysis** 

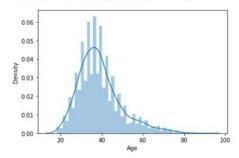
# **Univariate Analysis**

In [3]: sns.distplot(data['Age'])

C:\Users\chand\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar fle xibility) or `histplot` (an axes-level function for histograms).

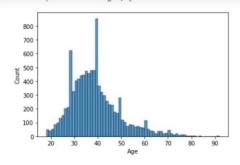
warnings.warn(msg, FutureWarning)

Out[3]: <AxesSubplot:xlabel='Age', ylabel='Density'>



In [4]: sns.histplot(data['Age'])

Out[4]: <AxesSubplot:xlabel='Age', ylabel='Count'>

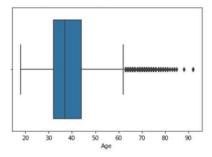


In [5]: sns.boxplot(data['Age'])

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

warnings.warn(

Out[5]: <AxesSubplot:xlabel='Age'>

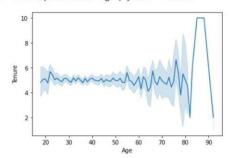


## Bi - Variate Analysis

In [6]: sns.lineplot(data['Age'],data['Tenure'])

C:\Users\chand\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass 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[6]: <AxesSubplot:xlabel='Age', ylabel='Tenure'>

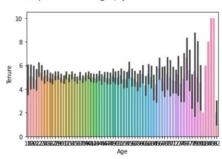


In [7]: sns.barplot(data['Age'],data['Tenure'])

C:\Users\chand\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variables as keyword ar gs: 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[7]: <AxesSubplot:xlabel='Age', ylabel='Tenure'>

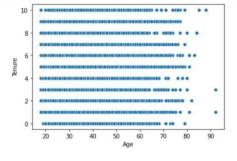


In [8]: sns.scatterplot(data['Age'],data['Tenure'])

C:\Users\chand\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass 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[8]: <AxesSubplot:xlabel='Age', ylabel='Tenure'>



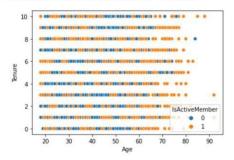
# Multi - Variate Analysis

In [9]: sns.scatterplot(data['Age'],data['Tenure'], hue=data['IsActiveMember'])

C:\Users\chand\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variables as keyword ar gs: 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[9]: <AxesSubplot:xlabel='Age', ylabel='Tenure'>





## 4. Descriptive Statistics

#### In [11]: data.mean()

C:\Users\chand\AppData\Local\Temp\ipykernel\_7968\531903386.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric\_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.

data.mean()

dataillean

Out[11]: RowNumber 5.000500e+03 1.569094e+07 CustomerId CreditScore 6.505288e+02 Age Tenure 3.892180e+01 5.012800e+00 Balance 7.648589e+04 NumOfProducts 1.530200e+00 HasCrCard 7.055000e-01 IsActiveMember EstimatedSalary 5.151000e-01 1.000902e+05 2.037000e-01 Exited dtype: float64

#### In [12]: data.median()

C:\Users\chand\AppData\Local\Temp\ipykernel\_7968\4184645713.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric\_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.

data.median()

Out[12]: RowNumber

5.000500e+03 CustomerId CreditScore 1.569074e+07 6.520000e+02 3.700000e+01 Age Tenure 5.000000e+00 Balance 9.719854e+04 NumOfProducts 1.000000e+00 HasCrCard 1.000000e+00 IsActiveMember 1.000000e+00 EstimatedSalary 1.001939e+05 Exited 0.000000e+00 dtype: float64

In [13]: data.mode()

Out[13]:

	RowNumber	Customerld	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	Estimated Sala
0	1	15565701	Smith	850.0	France	Male	37.0	2.0	0.0	1.0	1.0	1.0	24924.
1	2	15565706	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	N
2	3	15565714	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	N
3	4	15565779	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	N
4	5	15565796	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	N
	1999		***	440.	***	666	200		***			***	
9995	9996	15815628	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	N
9996	9997	15815645	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	N
9997	9998	15815656	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	N
9998	9999	15815660	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	N
9999	10000	15815690	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	N

10000 rows × 14 columns

5. Missing Values

#### omisoning valu

In [14]:	data.isnull().any	()
Out[14]:	RowNumber	False
	CustomerId	False
	Surname	False
	CreditScore	False
	Geography	False
	Gender	False
	Age	False
	Tenure	False
	Balance	False
	NumOfProducts	False
	HasCrCard	False
	IsActiveMember	False
	EstimatedSalary	False
	Exited	False
	dtype: bool	

# There are no missing values

# **6.Handling Outliners**

	RowNumber	Customerld	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
	1000.9	15591167.1	521.0	200000	1.0	0.0	1.0	0.0	0.0	20273.58	0.0
	.quantile([	[0.1,0.5])									
	RowNumber	Customerld	CreditScore	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	<b>Estimated Salary</b>	Exited
	1000.9	15591167.1	521.0	27.0	1.0	0.00	1.0	0.0	0.0	20273.580	0.0
	5000.5	15690738.0	652.0	37.0	5.0	97198.54	1.0	1.0	1.0	100193.915	0.0
a	.quantile([	0.1,0.9])									
	RowNumber	Customerld	CreditScore	Age	Tenure	Balance	e NumOfProduc	ts HasCrCa	rd IsActiveMemb	er EstimatedSala	ry Exited
1	1000.9	15591167.1	521.0	27.0	1.0	0.00	0 1	.0 0	0.0	.0 20273.58	30 0.0
9	9000.1	15790830.7	778.0	53.0	9.0	149244.79	2 2	.0 1	.0 1	.0 179674.70	1.0

# 7.Perform Encoding

	om sklearn	ziiipor e prej	DI OCC3311	15										
: le	<pre>le = preprocessing.LabelEncoder()</pre>													
<pre>oneh = preprocessing.OneHotEncoder()</pre>														
da	data['Age'] = le.fit_transform(data['Age'])													
da	ta.head()													
	RowNumber	Customerld	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	Estimated Salary	
0	1	15634602	Hargrave	619	France	Female	24	2	0.00	1	1	1	101348.88	
1	2	15647311	Hill	608	Spain	Female	23	1	83807.86	1	0	1	112542.58	
_	3	15619304	Onio	502	France	Female	24	8	159660.80	3	1	0	113931.57	
2	4	15701354	Boni	699	France	Female	21	1	0.00	2	0	0	93826.63	
3				2223	0	Female	25	2	125510.82	1	1	1	79084.10	
	5	15737888	Mitchell	850	Spain	remale	25	-	120010.02				73004.10	

## 8. Split into Dependent and Independent variables (X and Y)

```
In [24]: x = data.iloc[:,0:12]
In [25]: x
Out[25]:
            RowNumber Customerld Surname CreditScore Geography Gender Age Tenure Balance NumOfProducts HasCrCard IsActiveMember
        0 1 15634602 Hargrave 619 France Female 24 2 0.00
                                        608
                                                              1 83807.86
          1
                  2 15647311
                                              Spain Female 23
            3 15619304 Onio 502 France Female 24 8 159660.80
                              Boni
                 4 15701354
                                       699
                                                                                  2
                                                                                          0
                                                                                                     0
          3
                                             France Female 21
                                                              1
                                                                     0.00
                  5 15737888 Mitchell 850 Spain Female 25 2 125510.82
                                             France Male 21 5
                9996 15606229 Obijiaku 771
                     15569892 Johnstone
                                       516
                                              France
                                                     Male 17
                                                               10 57369.61
                9998 15584532 Liu
        9997
                                      709 France Female 18 7
                                                                   0.00
                                                                                          0
                9999 15682355 Sabbatini 772 Germany Male 24
                                                                                  2
                                                                                          1
        9998
                                                                3 75075.31
                                                                                                     0
               10000 15628319 Walker 792 France Female 10 4 130142.79
        9999
       10000 rows × 12 columns
In [26]: y = data['Balance']
```

```
In [27]: y
Out[27]: 0
                 83807.86
        2
               159660.80
                    0.00
               125510.82
                 0.00
        9995
               57369.61
        9996
        9997
                    0.00
                 75075.31
         9999
                130142.79
        Name: Balance, Length: 10000, dtype: float64
```

## 9. Scale Independent values

```
In [7]: x = data.iloc[:,0:1]
In [8]: from sklearn.preprocessing import StandardScaler, MinMaxScaler
          sc = StandardScaler()
          x_scaled = sc.fit_transform(x)
In [9]: x_scaled
Out[9]: array([[-1.73187761],
                  [-1.73157701],
[-1.7315312],
[-1.73118479],
                   [ 1.73118479],
                   [ 1.7315479],
[ 1.7315312 ],
[ 1.73187761]])
```

# 10. Split the data into train and test

```
In [12]: from sklearn.model_selection import train_test_split x_train, x_test, y_train, y_test = train_test_split(x_scaled, y, test_size = 0.3, random_state = 0)
In [13]: x_train
Out[13]: array([[ 0.92889885],
                    [ 1.39655257],
[-0.4532777 ],
                      0.60119484],
                    [ 1.67853045],
[-0.78548505]])
In [15]: x_train.shape
Out[15]: (7000, 1)
In [16]: y_train
Out[16]: 7681
                     146193.60
                    0.00
160979.68
           3691
           202
                          0.00
           5625
                    143262.04
                    120074.97
           9225
           4859
                     114440.24
                    161274.05
0.00
108076.33
           3264
           9845
           2732
           Name: Balance, Length: 7000, dtype: float64
```