$Assignment_2$

November 17, 2022

Assignment Date	24 September 2022
Student Name	Shwetha L S
Student Roll Number	2019504586
Maximum Marks	2 Marks

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

%matplotlib inline
```

0.0.1 Download and Load dataset

[3]:		RowNumb	or	Custome	rTd	Surname	CreditScore	Coography	Gend	or	Λσο	_
		ROWINGIND	ет	Cuscome	IIu	Burname	crearcacare	deography	Gena	ET	Age	\
	0		1	15634	602	Hargrave	619	France	Fema	le	42	
	1		2	15647	311	Hill	608	Spain	Fema	le	41	
	2		3	15619	304	Onio	502	France	Fema	le	42	
	3		4	15701	354	Boni	699	France	Fema	le	39	
	4		5	15737	888	Mitchell	850	Spain	Fema	le	43	
		Tenure		Balance	Num	OfProducts	HasCrCard	IsActiveMe	mber	\		
	0	2		0.00		1	1		1			
	1	1	8	3807.86		1	0		1			
	2	8	15	9660.80		3	1		0			
	3	1		0.00		2	0		0			
	4	2	12	5510.82		1	1		1			

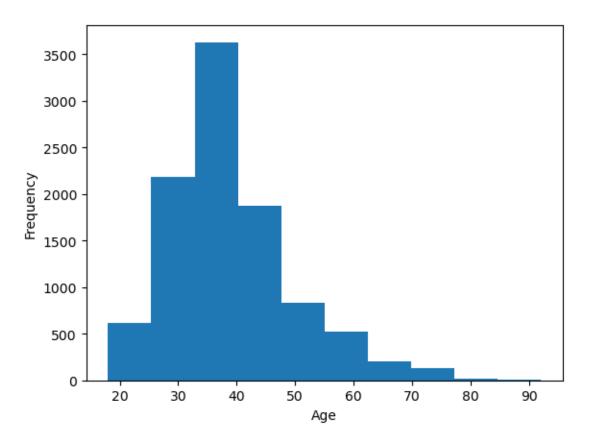
	${ t EstimatedSalary}$	Exited
0	101348.88	1
1	112542.58	0
2	113931.57	1

```
3 93826.63 0
4 79084.10 0
```

0.0.2 Uni-variate Analysis

```
[4]: plt.hist(df["Age"])
  plt.xlabel("Age")
  plt.ylabel("Frequency")
```

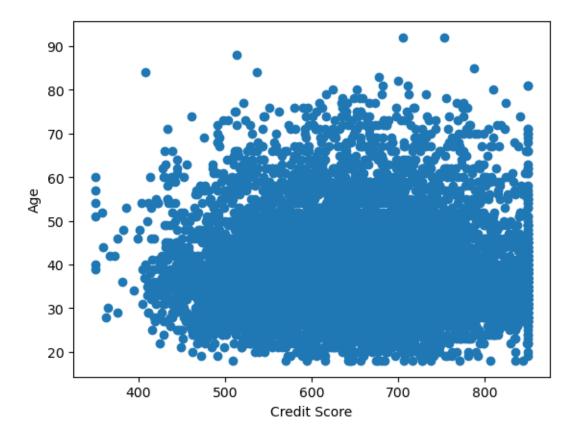
[4]: Text(0, 0.5, 'Frequency')



0.0.3 Bi-variate Analysis

```
[10]: plt.scatter(df["CreditScore"], df["Age"])
    plt.xlabel("Credit Score")
    plt.ylabel("Age")
```

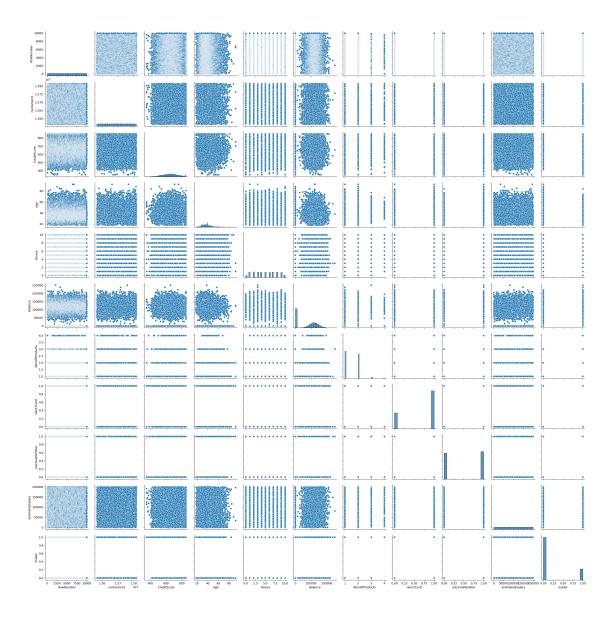
[10]: Text(0, 0.5, 'Age')



0.0.4 Multi-variate Analysis

[11]: sns.pairplot(df)

[11]: <seaborn.axisgrid.PairGrid at 0x7f72c8580250>



0.0.5 Descriptive Statistics

[12]: df.describe()

[12]	:	RowNumber	CustomerId	CreditScore	Age	Tenure	\
	count	10000.00000	1.000000e+04	10000.000000	10000.000000	10000.000000	
	mean	5000.50000	1.569094e+07	650.528800	38.921800	5.012800	
	std	2886.89568	7.193619e+04	96.653299	10.487806	2.892174	
	min	1.00000	1.556570e+07	350.000000	18.000000	0.000000	
	25%	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	
	50%	5000.50000	1.569074e+07	652.000000	37.000000	5.000000	
	75%	7500.25000	1.575323e+07	718.000000	44.000000	7.000000	
	max	10000 00000	1 581569e+07	850 000000	92 000000	10 000000	

	Balance	NumOfProducts	HasCrCard	IsActiveMember	\
count	10000.000000	10000.000000	10000.00000	10000.000000	
mean	76485.889288	1.530200	0.70550	0.515100	
std	62397.405202	0.581654	0.45584	0.499797	
min	0.000000	1.000000	0.00000	0.000000	
25%	0.000000	1.000000	0.00000	0.000000	
50%	97198.540000	1.000000	1.00000	1.000000	
75%	127644.240000	2.000000	1.00000	1.000000	
max	250898.090000	4.000000	1.00000	1.000000	
	EstimatedSalary	z Exited			
count	10000.000000	10000.000000			
mean	100090.239881	0.203700			
std	57510.492818	0.402769			
min	11.580000	0.000000			
25%	51002.110000	0.000000			
50%	100193.915000	0.000000			
75%	149388.247500	0.000000			
max	199992.480000	1.000000			

0.0.6 Handle missing values

```
[13]: print("No of missing values:")
    df.isnull().sum()
```

No of missing values:

[13]: RowNumber CustomerId 0 Surname 0 CreditScore 0 Geography 0 Gender 0 Age 0 0 Tenure Balance 0 0 NumOfProducts HasCrCard 0 IsActiveMember 0 EstimatedSalary 0 Exited dtype: int64

0.0.7 Split Data into independent and dependent variables

```
[14]: x = df.iloc[:,3:13].values
y = df.iloc[:,13:14].values
print("x shape:", x.shape)
print("y shape:", y.shape)

x shape: (10000, 10)
```

0.0.8 Categorical columns encoding

x shape: (10000, 13)

y shape: (10000, 1)

0.0.9 Split the data into training and tesing

x_train shape: (8000, 13)
y_train shape: (8000, 1)
x_test shape: (2000, 13)
y_test shape: (2000, 1)

0.0.10 Scale the independent variables

```
[18]: from sklearn.preprocessing import StandardScaler

sc = StandardScaler()
    x_train = sc.fit_transform(x_train)
    x_test = sc.transform(x_test)
```