# $Assignment\_2$

## November 17, 2022

Assignment Date	24 September 2022
Student Name Student Roll Number Maximum Marks	Prajeen R G 2019504560 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]: df = pd.read_csv("https://drive.google.com/uc?

oid=1_HcMOK8wt4b7FMLkc1V1dv0y6I_9ULzy")

df.head()
```

[3]:	RowNumb	er Custom	erId	Surname	CreditScore	Geography	Gender	Age	\
0		1 1563	4602	Hargrave	619	France	Female	42	
1		2 1564	7311	Hill	608	Spain	Female	41	
2		3 1561	9304	Onio	502	France	Female	42	
3		4 1570	1354	Boni	699	France	Female	39	
4		5 1573	7888	Mitchell	850	Spain	Female	43	
	Tenure	Balance	Num	OfProducts	HasCrCard	IsActiveMe	mber \		
0	2	0.00		1	1		1		
1	1	83807.86		1	0		1		
2	8	159660.80		3	1		0		
3	1	0.00		2	0		0		
4	2	125510.82		1	1		1		

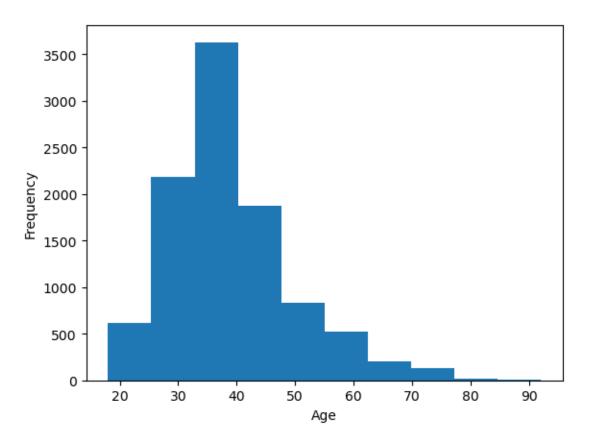
	EstimatedSalary	Exited
0	101348.88	1
1	112542.58	0
2	113931 57	1

```
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")
```

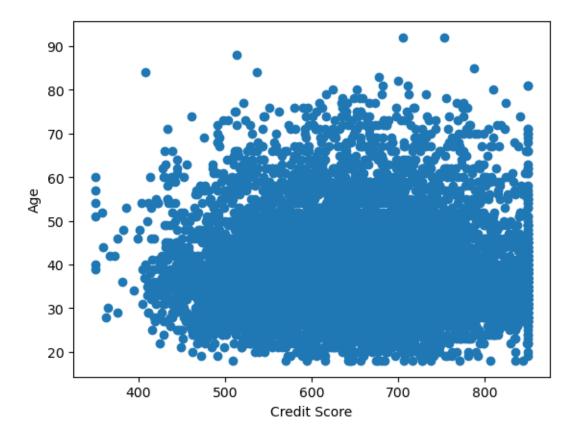
## [4]: Text(0, 0.5, 'Frequency')



## 0.0.3 Bi-variate Analysis

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

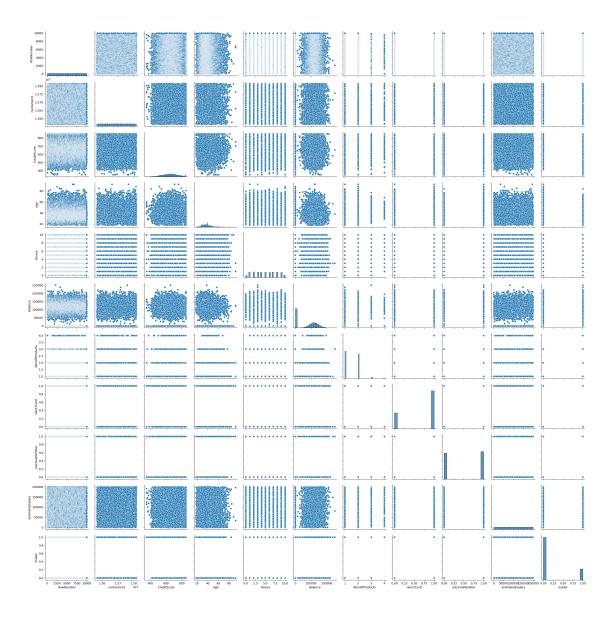
[10]: Text(0, 0.5, 'Age')



## 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)
```